

**The influence of the Gestalt principles similarity and proximity on the processing of  
information in graphs: An eye tracking study.**

Marlies de Brouwer  
ANR 810146

Master thesis  
Communication- and Information sciences  
Data Journalism

Tilburg School of Humanities  
Tilburg University, Tilburg

First supervisor: Dr. R. Cozijn  
Second supervisor: Drs. H.A.F.J. van der Kaa MBA

July 2014

## **Preface**

After five years of being a student at Tilburg University, I have finished my master Data Journalism. In my bachelor I studied Text and Communication and one of the courses was Research Text and Communication. One of the most interesting things I did during that course was eye tracking. I thought it was fascinating that by interpreting eye movements a small part of the complex brain was explained. This course has inspired me to do research with eye tracking. In my bachelor I had no chance to use the eye tracker by myself. In my master I chose the track Data Journalism. My interest for journalism and visualizations had grown. After one year of doing courses, I decided to conduct a research which investigates the usefulness of infographics. I wanted to investigate this with eye movements.

By conducting the research I learned the most in five years. I could chose my own topic and I had a lot of freedom to conduct the research. However, I could not do this research on my own. My supervisor, Rein Cozijn, helped me a lot and was really flexible. He helped me with great insights and detailed feedback. In addition, I want to thank my second supervisor Hille van der Kaa for inspiring me to do Data Journalism. Of course, my thesis was not finished without the support of my family and friends.

Marlies de Brouwer

June, 2014

## **Abstract**

Recently, a new type of graph has emerged: the infographic. The infographic tells the whole story, instead of just supporting the text, like standard graphs do. The influence of the Gestalt principles similarity and proximity on the processing of information in graphs is studied in an eye tracking experiment. In addition, the entertainment and usability value of graph types is investigated.

The results showed that the total viewing times and the viewing times on the target region were the same regardless of the use of similarity and proximity. The total viewing times were shorter for easy infographics than for difficult infographics. The axes were observed first and after that the data area for the bar charts. For line graphs and scatterplots, the axes and the data area were observed first equally often. Infographics were the least learnable and the least efficient compared to the other graphs, but equally entertaining in comparison with bar charts and pie charts.

It is concluded that the use of the Gestalt principles does not help the viewers to extract information faster from infographics. Furthermore, viewers are used to interpret bar charts in a similar way. Infographics should not be used for information extraction.

The conclusion that the use of Gestalt principles does not help the viewers to extract information faster from infographics can be explained by the lines around the subtopics and the variance in infographics. The differences in viewing times between easy and difficult infographics can be explained by the number of elements displayed per subtopic and their visual organization. When information needs to be extracted fast, the number of elements in subtopics needs to be minimized. For future research the infographics need to be checked for difficulty, and the lines around the subtopics need to be deleted. The viewing behavior on bar charts can be explained by the nature of the graphs and the information on the axes. The axes of bar charts need to be clear as possible. Future research should focus on the specifics of the viewing behaviors.

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## **1. Theoretical outline**

In this section the theoretical outline for this research is described. The study investigates the processing of information in graphs, and their usability and entertainment value.

### **1.1 Introduction**

Journalists have been using graphs to support their work for many years. For example, USA Today was a groundbreaking newspaper in 1982. The newspaper was the first to use graphs and images for the visualization of information (Siricharoen, 2013). Newspapers use graphs to visualize complex processes and data making the information understandable for the readers (see, e.g. Utt & Pasternack, 1993; Ware, 1999). The choice of the graph type that is used for supporting an article depends on the data, the audience, and the medium, according to Tufte (1983). An example that the data demands the choice of graph type, is described by Zack and Tversky (1990). They claim that bar charts facilitate discrete comparisons and line graphs trends.

Recently a new type of graph has emerged. These graphs are called infographics and are entertaining for the viewers. They use images, numbers, graphics, and text to visualize the journalistic story. Standard graphs such as bar charts, have long been used to support storytelling. The text conveys the story, and the image provides supporting evidence or related details. However, infographics attempt to combine narratives with graphics. In a sense, they tell the story instead of the text (see, e.g., Segel & Heer, 2010; Siricharoen, 2013). News organizations including the New York Times, Washington Post and the Guardian use infographics in their media.

Information that is displayed in graphs needs to be processed by the viewers, and the difficulty of processing information is influenced by the way information is displayed. There are many guidelines for creating graphs. The practical guidelines of Tufte (1983) are followed often in newsrooms. For example, designers should not use distracting patterns, too much color, no shadow effect, or 3D. This is called chart junk. Also, animations need to be used with care, because it is really hard to understand animations. However, little empirical evidence for this approach has been found, and the way in which readers perceive the graphs that are created with this approach has been questioned (see, e.g., David, 1992). In addition, other guidelines are taken into account by designers. Kosslyn (1994) claims that it is important that the colors in graphs are selected with care, and that colors have to be used to indicate categories. This guideline is based on the Gestalt principles of Kohler (1929) which clarify how a design is perceived by the viewers. For example, the Gestalt principle similarity

states that objects with the same color are perceived as objects belonging to the same group. The principle of proximity states that objects that are placed close together, are perceived as objects belonging to the same group. If several bars in a bar charts have the same color or are placed close together, these bars are perceived as belonging to the same group. The use of the Gestalt principles similarity and proximity of Kohler (1929) are the topic of this study. The central research question of this study is: *What is the influence of the Gestalt Principles proximity and similarity on the processing of information in graphs?* In addition, the usability and the entertainment value of the graph types are investigated.

This chapter will go into graphics, their design and processing. In the first paragraph, the different types of graphs that are used for visualizing data are described. The second paragraph discusses the Gestalt principles of Kohler (1929) and the proximity compatibility principle of Wickens and Carswell (1995). This literature provides presumptive evidence that the processing of information is influenced by taking these principles into account. The third paragraph discusses the processing of information in graphs. This includes literature on the three steps viewers have to take in order to understand graphs (Pinker, 1990), the perceptual organization hypothesis by Shah, Meyer and Hegarty (1999), and the cognitive load theory of Sweller (1994). The fourth paragraph discusses the usability and the entertainment value of different types of graphs. Finally, the research question and the hypotheses are formulated.

## **1.2 Types of graphs**

As mentioned in the introduction, graphs support the work of journalists (Siricharoen, 2013). Data and information visualizations show quantitative and qualitative information, so the viewers may notice patterns, trends, and anomalies (Friendly, 2008). Newspapers use for example bar charts, line graphs, scatterplots, and pie charts to show data. Apparently, every type of graph has his own purpose. Friendly (2008) describes the graph types which William Playfair has developed. Line graphs were developed in 1786 to show the changes in economic subjects over time. An example of a line graph is the development of the national debt versus time. He also developed bar charts in 1786 that showed relations of discrete series, for example imports from and exports to England. Pie charts and circle diagrams were developed in 1801 to show part-whole relations.

Zack and Tversky (1999) investigated the purposes of bar charts and line graphs. They found that bar charts are best suited to display discrete comparisons, and line graphs to display trends between data points. The information that is displayed in bar charts needs to be

described in terms of comparisons between individual discrete data points. The terms to describe these comparisons are: higher, lower, larger than, and smaller than. Information that is displayed in line graphs needs to be described as trends between data points. The best terms to be used here are: increase and decrease.

Scatterplots were constructed for the purpose of showing the association between two variables. Readers deduce from a scatter of points a linear or non-linear relationship. Scatterplots also display information about the strength of the relationship between the two variables (see, e.g., Doherty & Anderson, 2009).

Pie charts are used to show numerical proportions. The chart is divided into sections with areas that are proportional to the quantities they represent. Consequently, pie charts show part-whole relations (see, e.g., Friendly, 2008).

As mentioned in the introduction, standard graphs such as bar charts, have long been used to support the text. The text conveys the story, and the image provides supporting evidence or relate details. However, new types of visualization attempt to combine narratives with graphics. These graphs are called infographics. An infographic is a graphic visual representation of information, data or knowledge with a combination of words, numbers, and images. These graphs intend to clarify and integrate difficult information quickly and clearly. In a sense, they tell the story instead of the text (see, e.g., Segel & Heer, 2010; Siricharoen, 2013).

### **1.3 Designing graphs**

Information in graphs needs to be categorized according to guidelines. Kosslyn (1994) claims that it is important that colors in graphs are selected with care, and that colors have to be used to indicate categories. In addition, Tversky (1997) describes that categories between objects can be created by using for example proximity, similarity, and boundaries between objects. The grouping of objects is based on the Gestalt principles of Kohler (1929).

The Gestalt principles of Kohler (1929) offer guidelines for designing graphs. The Gestalt principles were defined a century ago, and they still provide guidelines for designers and data journalists. When designing graphs, it is important to take into account these principles.

These principles clarify how a design is perceived by viewers. If these guidelines are not considered by designers, the effect may be that the intentions of the design are not understood. For example, if two objects have the same color, and the objects have no interaction, the readers still try to integrate these objects because of their similarity (Kohler, 1929). Kohler (1929) distinguished five Gestalt principles: *continuity*, *enclosure*, *connectedness*, *proximity* and *similarity*.

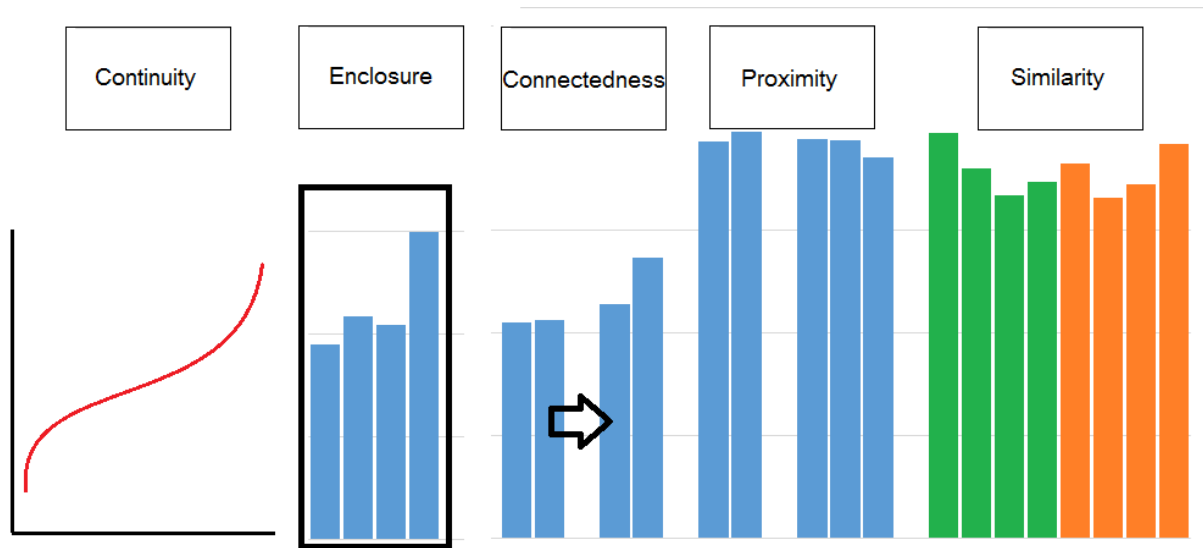


Figure 1.1: Example of the use of Gestalt principles of Kohler: continuity, enclosure, connectedness, proximity, and similarity.

The Gestalt principle *continuity* states that objects with fluent lines are noticed with ease, however, objects with straight lines that suddenly change direction are harder to perceive. For example, the line graph in Figure 1.1 is easy to perceive because of the use of fluent lines to connect data points instead of using straight lines to connect the points. An explanation for this principle is that fluent lines are more natural. These fluent forms are less organized, but they are easier to perceive. Another principle, *enclosure*, states that objects which are placed in a certain area with sharp boundaries are perceived as belonging to the same group. For example, when several bars in a bar chart are framed, these bars are perceived as belonging to the same group (see Figure 1.1). The principle *connectedness* states that groups can be made by lines or arrows. For example, when an arrow is placed between two graphs, these two graphs are perceived as belonging to the same group (see Figure 1.1).

The Gestalt principle *proximity* states that objects that are placed close together, are perceived as belonging to the same group. Casasanto (2008) claims that people describe objects that are lookalikes as “close” and objects that are perceived as dissimilar as “far away”. For example, when several bars are placed apart from other bars, these several bars are

perceived as belonging to the same group (see Figure 1.1). The Gestalt principle *similarity* states that objects that are the same in color, shape, orientation, size, and value are perceived as objects belonging to the same group. For example, when some bars in a bar chart have the same color and other bars have another color, the bars with the same color are perceived as objects belonging to the same group (see Figure 1.1). Proximity and similarity are the subject of investigation in this study.

Wickens and Carswell (1995) state in their *proximity compatibility principle* that the design of the graph depends on the task. This principle is based upon two kinds of proximity: *perceptual* and *processing proximity*. The level of perceptual proximity depends on the level of processing proximity: the extent in which graphical elements are perceptually similar depends on whether two or more graphical elements need to be integrated. When this principle is not taken into account, it is harder for the viewers to process the information that is shown in the graph.

*Processing proximity* refers to whether two or more graphical elements need to be integrated when performing a task. The processing proximity is high when several elements need to be combined, compared, or integrated. When the processing proximity is high, the attention is divided among multiple graphical elements because multiple pieces of information need to be considered simultaneously. When the processing proximity is high, it is advised to use high perceptual proximity. *Perceptual proximity* refers to the extent in which graphical elements are similar to each other. Examples of perceptual proximity are color, proximity, and shape. Graphical elements that are perceptually similar have high perceptual proximity. An example of a task is to compare the semester results for the year 2011 in the bar chart that is displayed in Figure 1.2 [1]. The processing proximity is high, because the bars that display the results in 2011 need to be compared. The level of processing proximity demands the perceptual proximity. Therefore, the two bars that display the results for 2011 need to be perceptually similar to each other giving these bars the same color, as seen in Figure 1.2 [1]. Because the two bars are colored red and the other bars are colored blue, the two red bars are perceptually similar based on their color.

The processing proximity is low when several elements of information need to be processed independently or when information needs to be filtered to extract a single value. When the processing proximity is low, the attention is focused on single elements because individual pieces of information need to be considered. When elements need to be processed separately, it is advised to use low perceptual proximity. For example, only specific

information elements are highlighted. An example of a task is to verify whether the turnover is less than 1200 euro in 2011-1 in the bar chart displayed in Figure 1.2 [2]. The processing proximity is low, because only one graphical element needs to be filtered out. Therefore, the level of perceptual proximity needs to be low as well, so the bar that displays the value of 2011-1 is highlighted as seen in Figure 1.2 [2].



Figure 1.2: An example of the proximity compatibility principle. [1] An example of high processing proximity: The turnover in 2011-1 is lower than in 2011-2. The level of perceptual proximity needs to be high as well, so the two red bars that display the results are perceptual similar based on their color. [2] An example of low processing proximity: The turnover in 2011-1 is less than 1200 euro. The level of perceptual proximity needs to be low as well, so one graphical element, the bar of 2011-1, is highlighted.



In short, Kohler (1929) describes the observation and interpretation of groups. The Gestalt principles of Kohler (1929) provide insight into how a design is perceived by viewers. The Gestalt principles similarity and proximity are used in this study. These principles state that when objects have the same color or are placed together, these objects are perceived as belonging to the same group. The proximity compatibility principle of Wickens and Carswell (1995) states that the task demands the design of the graph. When the task demands integrating graphical elements, the elements that need to be integrated in order to perform the task, need to be perceptually similar.

#### 1.4 The processing of information in graphs

According to Pinker (1990) three processes need to be completed before a graph can be understood: Viewers code the visual features, identify their quantitative properties, and relate them to the referents of the graph. These processes are incremental and interactive, so they occur for each part of the graph, separately (Carpenter & Shah, 1998).

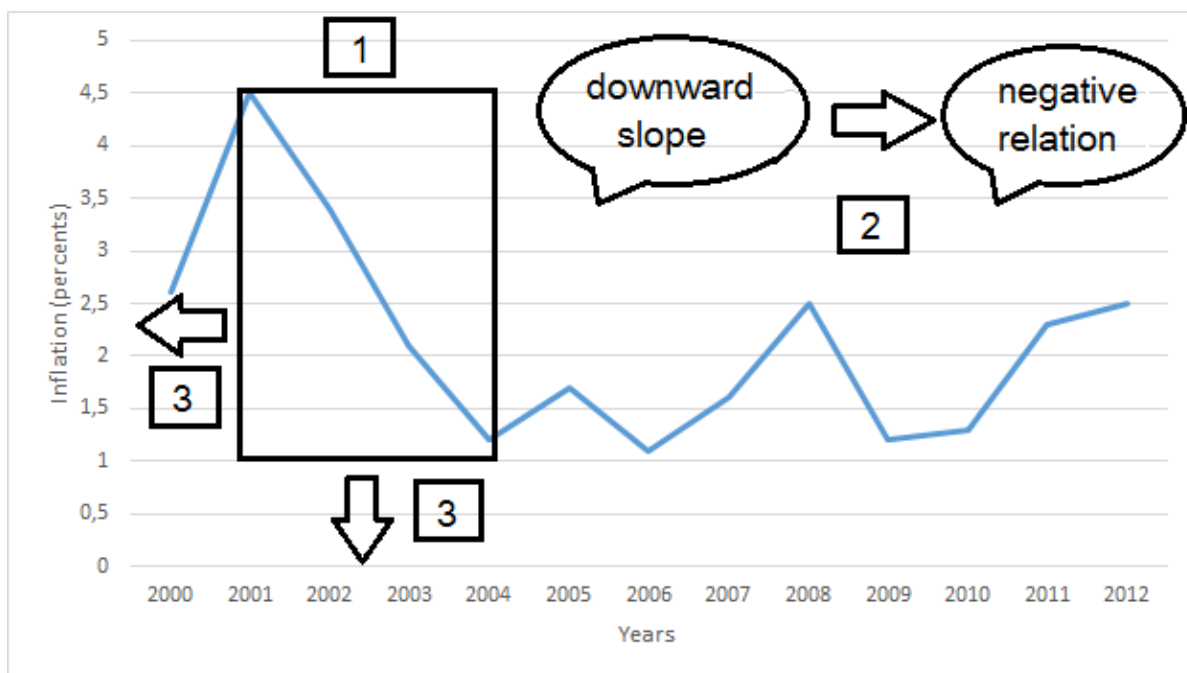


Figure 1.3: The three steps to comprehend graphs. First, code a part the visual feature. Second, identify their quantitative properties. Third, the quantitative properties are related to the referents of the graph (Pinker, 1990).

First, the viewers need to code the visual series and identify the most important visual features of the graph. For example, when line graphs are observed, the viewers need to notice the slope of the line. As seen in Figure 1.3, the viewers identify the most important visual features of the graph [1].

Second, the quantitative properties which are displayed by these features, need to be identified. According to Pinker (1990) and Carpenter and Shah (1998), two forms are used to identify the quantitative properties: visual chunking and complex inferential processes. In one form, the viewers code a *visual chunk*, visual chunks are parts of the perceptual organization of graphs, and associate this visual chunk with a quantitative fact or relationship automatically. For example, if the line of the line graph has a downward slope, the viewers identify a negative relation between the points on the line. In Figure 1.3, the viewers notice the downward slope and identify a negative relation [2]. In the other form, viewers must rely on complex inferential processes. The viewers need to infer the quantitative interpretation. It happens that viewers do not have the knowledge to associate the visual chunk to the quantitative property, for example the viewers do not know that an upward slope represents a linear relation. In addition, it happens that separate individual visual chunks need to be processed, for example the viewers need to compare or relate information in two different visual chunks.

When relevant quantitative information is accessible in visual chunks directly, for example a line in a line graph, pattern perception and association processes are sufficient to interpret quantitative information. The viewers can use their knowledge about the relations that are displayed in line graphs. For example, a downward slope means a negative relation between  $x$  and  $y$ . It is possible to decrease the information processing load of a graph. That can be achieved by supporting the pattern perception and the association processes and by reducing the need for inferential processing. In order to reduce the inferential processing load, the quantitative information needs to be presented in visual chunks directly. Shah, Mayer and Hegarty (1999) describe the *perceptual organization hypothesis* which clarifies the possibilities to reduce the inferential processing load. The definition they use for visual chunks is based on the Gestalt principles of perceptual organization. An example of this perceptual organization is that bars that are placed together or points that are connected by a line, are perceived as a visual chunk based on the Gestalt principle proximity. The perceptual organization hypothesis claims that the interpretation of the viewers is not influenced by the format but by the presentation of quantitative properties in the visual chunks in graphs. When information is displayed in visual chunks, for example grouped bars in bar charts, the viewers use their pattern perception to interpret the relationships.

Third, the quantitative properties need to be related to the variables which are displayed in the graph. The viewers need to infer the referents from the labels, title, and the axes. In Figure 1.3, the variables on the axes represent the inflation in percentages per year. The negative slope is related to the referents: a negative slope represents a decrease in inflation [3].

The processing of information in graphs increases the working memory load. The cognitive load theory of Sweller (1994) describes the factors that determine the difficulty of learning materials. The capacity of working memory is the most important element of the cognitive load theory of Sweller. The cognitive load is the total activity that takes place in memory. The load depends mainly on the number of elements that need to be paid attention to. The theory claims that when the cognitive load is higher than the capacity of working memory, the construction of meaning is delayed. Sweller described three forms of cognitive load: *intrinsic*, *extraneous*, and *relevant cognitive load*.

The *intrinsic cognitive load* is determined by the complexity of the information. The number of elements, and the interaction between these elements determines the complexity of the information. The element interactivity is determined by estimating the number of interacting elements. The interaction is low if the number of elements that need to be integrated is not larger than two. A high number of elements that need to be integrated leads to a high load of working memory.

The *extraneous cognitive load* is not determined by the information itself, but by the manner in which information is displayed. The extraneous cognitive load is a disturbance for the processing of information. An inappropriate design can lead to high extraneous cognitive load. For example, as seen in Figure 1.4, if the quarterly results of a company per year are distributed over several bar charts per year [1] to [4], and the results between years need to be compared, the extraneous cognitive load is high because the viewers need to integrate the different bar charts. When the information of the years is displayed in one bar chart [5], the extraneous cognitive load is low, because the information that needs to be compared is presented in the same place.



Figure 1.4: An example of lowering the extraneous cognitive load by displaying the information of bar charts [1] to [4] into bar chart [5].

The *relevant cognitive load* is determined by the mental activities that are related to the learning process. This form of cognitive load is related to the construction of schemes. Gaining knowledge implies the adaptation of an existent schema, or the construction of a new schema. In conclusion, the relevant cognitive load is related by the knowledge about topics. For example, expansion of an existing schema about the economic crisis is possible by reading an article about the latest development on this topic.

The cognitive load theory of Sweller (1994) clarifies how difficult it is for people to extract information from graphs. The intrinsic and extraneous cognitive load need to be as low as possible, so that more capacity remains for the expansion or construction of knowledge in long-term memory.

A way to reduce the extraneous cognitive load is to use the Gestalt principles similarity and proximity of Kohler (1929). By dividing the information into groups, chunking, the extraneous cognitive load is lowered. For example, the viewers need to compare the quarterly results in a bar chart of which the bars have the same color and are not separated, as seen in Figure 1.5 [1]. The viewers need to integrate the bars per year and they need to compare the development of the results per year with each other. This task leads to high element interactivity. The extraneous cognitive load can be lowered by chunking the information. This can be done visually by using the Gestalt principles of Kohler (1929), and color the bars or separate them as shown in Figure 1.5 [2] and [3]. According to Wickens and Carswell (1995) the information needs to be chunked when the task demands integrating or comparing graphical elements. When the bars per year are chunked, for example by using different colors or inserting spaces, the extraneous cognitive load is lowered, because the categories are visually indicated. The bars in a chunk need to be integrated, and according to Shah, Mayer and Hegarty (1999) the processing of information is faster when information is chunked because chunking supports the pattern perception and association processes.



Figure 1.5: An example of lowering the extraneous cognitive load by using the Gestalt principle similarity. In bar chart [1] the bars have the same color and are not separated. The extraneous cognitive load is lowered by chunking the information in bar chart [2] and [3]. The categories are indicated by color [2] or by inserted spaces [3].

In short, based on the literature of Pinker (1990), and Carpenter and Shah (1998) three steps are distinguished to comprehend graphs: Viewers code the visual chunks, identify their quantitative properties, and relate them to the referents of the graph. These steps describe the viewing behavior when observing graphs. The data area is observed first, and after that the

axes to infer the referents. In addition, three forms of cognitive load can be distinguished by Sweller (1994): intrinsic, extraneous, and relevant. The extraneous cognitive load can be lowered by taken into account the Gestalt principles. When information in graphs is chunked, the processing of information is faster, because the chunking facilitates the identification of categories.

### **1.5 The usability and entertainment values of graph types**

Nielsen (1994) describes five components that define the usability of interfaces: *learnability*, *efficiency*, *memorability*, *errors*, and *satisfaction*. Usability is a quality attribute that assesses how easy a user interface is to use. Although, the usability components of Nielsen (1994) have been defined for measuring the usability of interfaces, they are applicable to the usability of graphs, too. The component *learnability* defines the difficulty for users to accomplish basic tasks the first time they encounter the design. The component *efficiency* defines the extent to which the interface is efficient to use when the users have learned the design. The component *memorability* defines the extent to which the users remember the interface which means that when users return to the design after a period of not using it, they reestablish the information easily. The component *errors* defines the extent to which users make errors in using the interface, and whether the errors can be easily recovered. Finally, the component *satisfaction* defines the extent to which the interface is pleasant to use.

The present study will focus on three of the components: learnability, efficiency, and memorability. Bar charts and line graphs for example, always have the same elements: bars or lines, two axes, and sometimes a legend. The interface of these graphs is learned, therefore the viewers can extract information from these graphs by following the same viewing behavior. However, infographics consist of several images, text, and graphs. Infographics are always different, no infographic is the same. For every infographic, readers need to construct a new strategy to extract information visually.

The main purpose of graphs is to visualize data effectively. However, graphs are used to entertain readers, as well. Examples of entertainment values are fun to watch and attractiveness. The question is how entertaining the several types of graphs are.

### **1.6 Research question and hypotheses**

The main research question addressed in this study is: *What is the influence of the Gestalt Principles proximity and similarity on the processing of information in graphs?*

In addition, the usability and entertainment value of graphs are investigated.

The processing of information is investigated in an eye tracking study. Eye tracking measures the eye movements of people. This method is used to show the cognitive aspects of the processing of information. Eye tracking is used in several disciplines, for instance in language processing studies and usability studies. When viewers observe graphs, the eyes jump from one position in the graph to another. The mean length of that jump is called a saccade. During these jumps the eyes do not process visual information. The visual information is processed between the two jumps. This is called a fixation. The measures that are often used in eye-tracking studies are gaze duration and total viewing time. The gaze duration is defined as the sum of fixation durations and the saccades durations on an area of interest, for example a chunk, until the viewers switch to another area of interest. The total viewing time is defined as the sum of the fixation durations and the saccades durations on the graph (Cozijn, 2006).

Based on the literature that is described in the previous paragraphs it is expected that the Gestalt principles proximity and similarity influence the processing of graphs. The Gestalt principles (Kohler, 1929) and the proximity compatibility principle (Wickens and Carswell, 1995) implicate that clustering information leads to faster processing of information by lowering the extraneous cognitive load, because the graphical elements are clustered by chunking. The chunks contain information needed to perform the task and the information in these chunks is processed by the viewers. The processing of information in these chunks takes time.

**Hypothesis 1a:** The viewing times of a graph are shorter with the presence of the Gestalt principle similarity than with the absence of this principle.

**Hypothesis 1b:** The viewing times of a graph are shorter with the presence of the Gestalt principle proximity than with the absence of this principle.

**Hypothesis 1c:** The viewing times of the chunk that contains the information needed to perform the task are longer with the presence of the Gestalt principle similarity than with the absence of this principle.

**Hypothesis 1d:** The viewing times of the chunk that contains the information needed to perform the task are longer with the presence of the Gestalt principle proximity than with the absence of this principle.



Based on the theories of Pinker (1990), and Carpenter and Shah (1998) that are described in the previous paragraphs, it is expected that viewers apply the same viewing behavior when viewing standard graphs. These processes are: Code visual chunks, identify their quantitative properties, and relate them to the referents of the graph. These processes occur for each part of the graph, separately. For bar charts, line graphs, and scatterplots this would mean: Code the data area, identify its quantitative properties, and relate it to the axes.

**Hypothesis 2:** The viewers observe the data area first and after that the axes when viewing bar charts, line graphs, and scatterplots.

Based on the literature that is described in the previous paragraphs it is expected that the usability and entertainment value differ among the types of graphs. The interface of the standard graphs, bar charts, line graphs, pie charts and scatterplots, is learned. The viewers can extract information from them by using the same viewing behavior. However, no infographic is the same. For every infographic the viewers need to construct a new strategy to extract information visually. Because of the variation of infographics, they will be more fun to watch.

**Hypothesis 3a:** It is more difficult to extract information from infographics than from bar charts, line graphs, pie charts, and scatterplots.

**Hypothesis 3b:** Infographics are more entertaining than bar charts, line graphs, pie charts, and scatterplots.

## **2. Material research**

The present study investigated the influence of similarity (color) and proximity on the processing of information, whether there were different viewing behaviors for bar charts, line graphs and scatterplots, and the usability and entertainment value for graph types was researched. An online material research has been conducted to investigate whether the use of color and proximity in bar charts<sup>1</sup> had an influence on the difficulty of verifying statements. In addition, the difficulty of the statements was investigated for additional bar charts, line graphs, scatterplots, pie charts, and infographics to select the stimuli for the experiment.

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<sup>1</sup> The material research was conducted on bar charts, however, infographics were selected for the main research. Bar charts were not used in the main research because of their simplicity. This was based on the results of the material research (see section 2.2 and 2.3 for more information).

## 2.1 Method

### 2.1.1 Participants

Twenty participants, Communication- and Information sciences students of which 11 bachelors, 1 premaster and 8 masters, filled out the survey (14 women and 6 men). The mean age of the participants was 21.95 (SD = 2.23).

### 2.1.2 Materials

50 graphs were created of which 8 bar charts were manipulated with color and proximity, 8 additional bar charts, 10 line graphs, 8 scatterplots, 8 pie charts, and 8 infographics (see Appendix A).

The manipulated bar charts showed data about a certain topic that could be clustered. For example, in Figure 2.1, the data consisted of quarterly results clustered per year. In some bar charts, the data could be clustered by subtopic (see Figure 2.2).

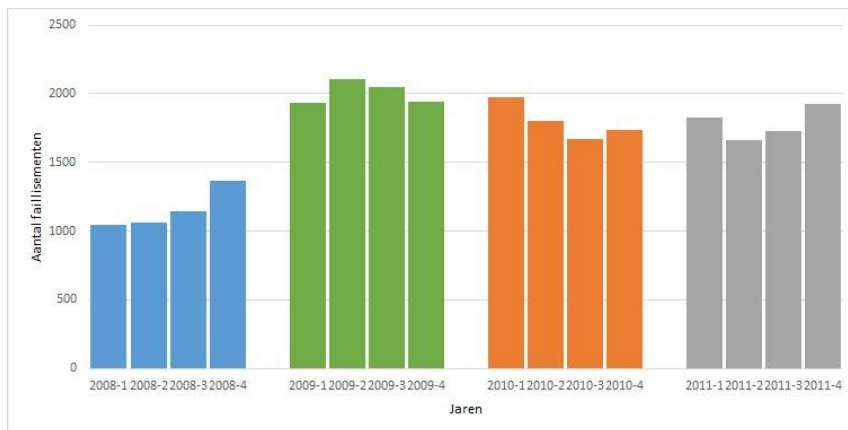


Figure 2.1: Example of a bar chart in which the data are clustered by year.

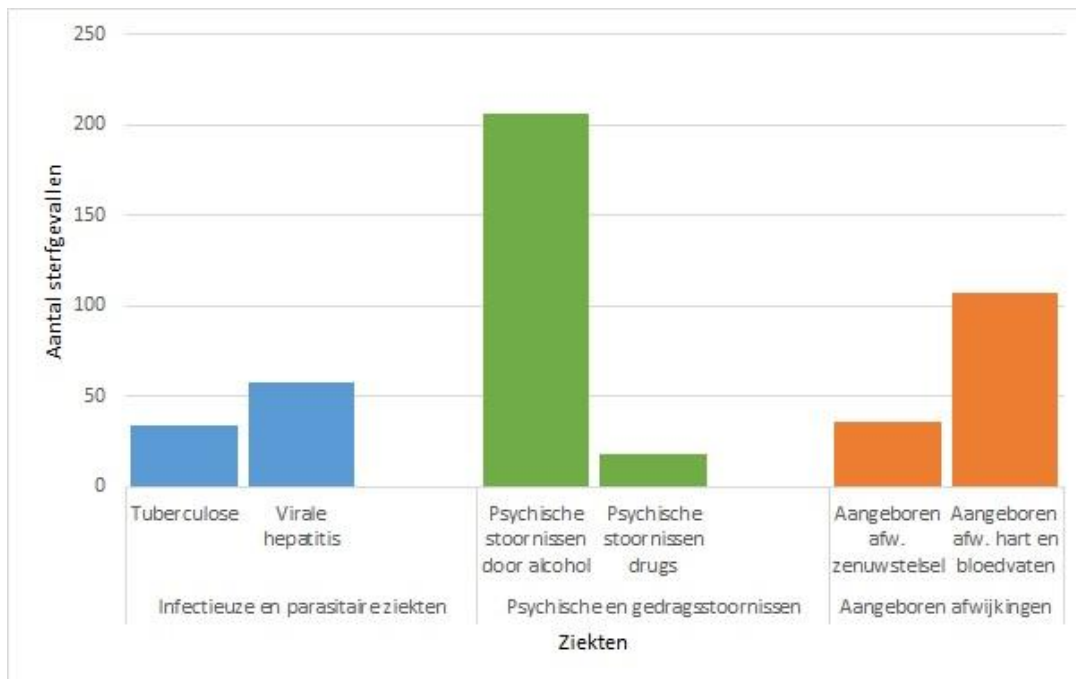


Figure 2.2: Example of a bar chart in which the data are clustered by subtopic. It shows different diseases in the subtopics, and these subtopics consist of different elements, for example tuberculosis.

Eight topics were selected for creating the bar charts. Six out of eight themes were chosen from the data of a statistics website ([www.cbs.nl](http://www.cbs.nl)): immigrants, bankruptcy, health, revenue, causes of death and unemployment. The data for the other two topics were used from the annual report of Tilburg University 2012: amount of diplomas per year and courses. In addition, 42 other graphs were created: 8 scatterplots, 8 additional bar charts, 10 line graphs (four line graphs with one line and six with several lines), 8 pie charts, and 8 infographics. Some of the topics for the line graphs were obtained from the statistic website ([www.cbs.nl](http://www.cbs.nl)). The topics for the other graphs were taken from the internet. The data were altered, so that they no longer reflected the information from which they were taken. The data was fake. The material was distributed over four versions.

The 8 bar charts were manipulated (see Figure 2.3). Bar chart [1] was not manipulated with color and proximity. To manipulate similarity the clusters were colored differently [2], the data was clustered per year. To manipulate proximity space was inserted between the clusters [3]. The inserted space was as large as the bar width. In addition, bar charts were manipulated with color and proximity, so space was inserted and the bars were colored differently [4]. The text on the axes was placed horizontally or in an angle of 45 degrees, if possible.

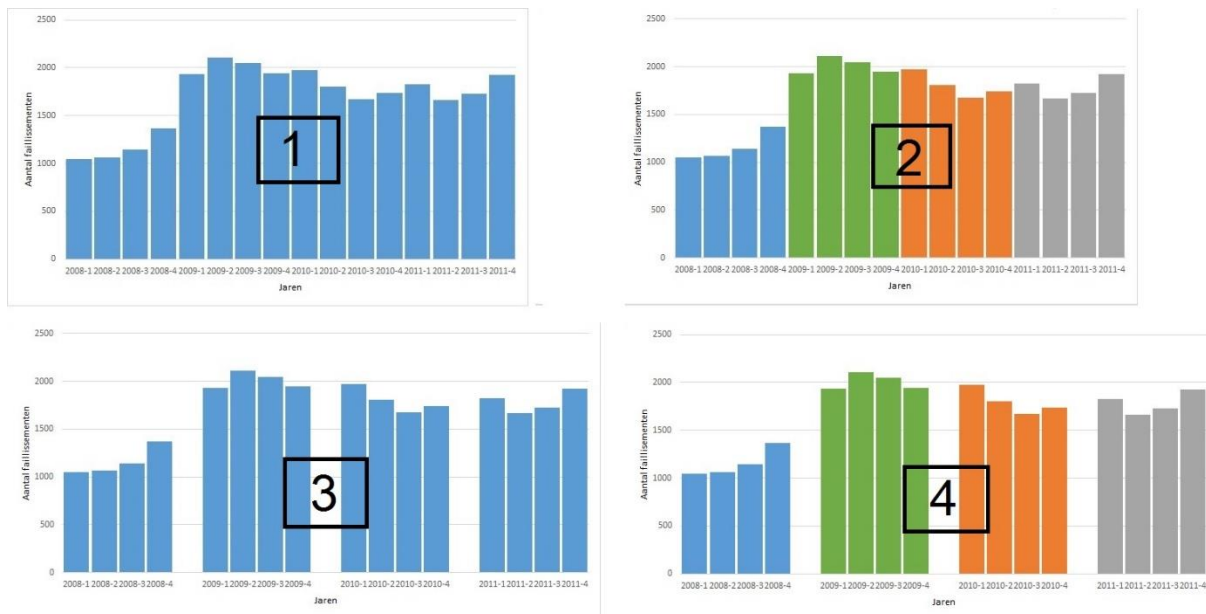


Figure 2.3: An example of manipulated bar charts (bankruptcy). [1] was not manipulated with color or proximity. [2] is manipulated with color, [3] with proximity, and [4] with color and proximity.

The 42 other graphs were not manipulated. The additional bar charts consisted of two axes, and vertical bars. The line graphs consisted of two axes, one line or multiple lines and a legend. The scatterplots consisted of two axes and dots. The pie chart consisted of three or more parts and a legend. The infographics consisted of several graphs, images and numbers.

For each graph, a statement was created that had to be verified (see Appendix A). The participants needed to compare clusters in the experimental graphs. For example, the statement for the bar charts in Figure 2.3 was “The number of bankruptcies is higher in 2009-1 than in 2008-4” (true).

For additional bar charts, the participants needed to compare the bars. For example, the statement for the bar chart in Figure 2.4 was “During the fourth quarter every year the least number of trucks are sold” (false).

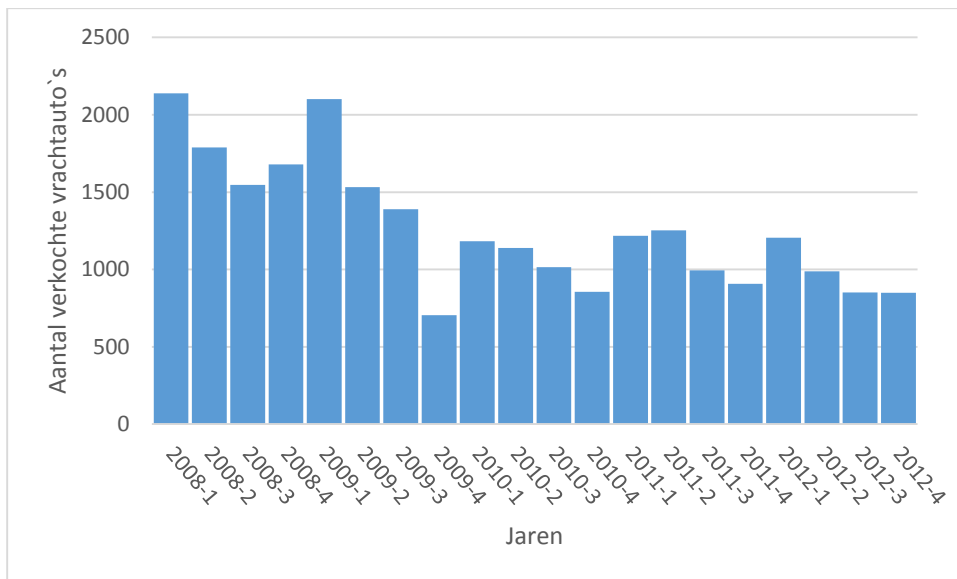


Figure 2.4: Bar chart Cars (x axis: years, y axis: number of sold cars). The bars needed to be compared for verifying the statement “During the fourth quarter every year the least number of trucks are sold” (false).

Also, for bar charts values needed to be verified. For example, the statement for the bar chart in Figure 2.5 was “In 2003 175000 kg CO2 is emitted” (false).

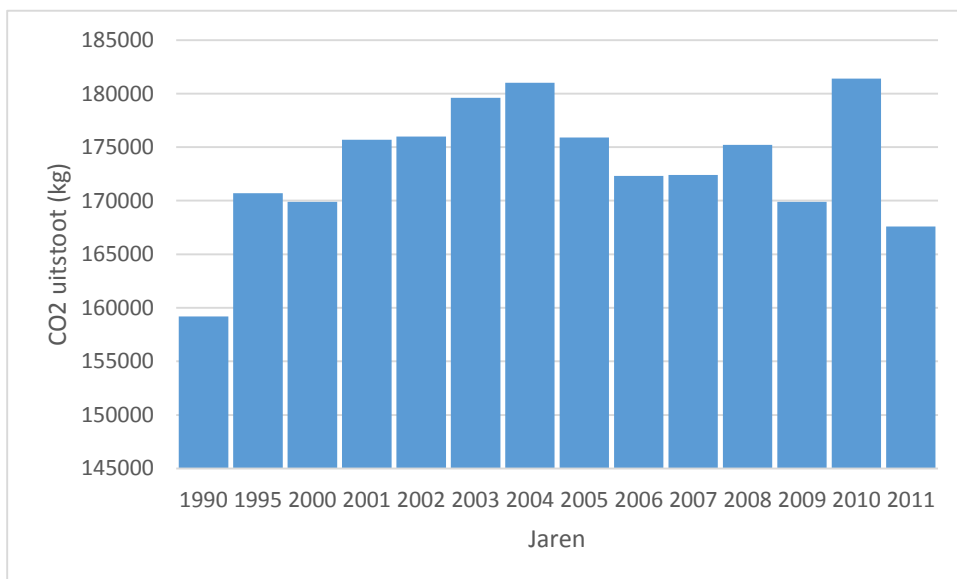


Figure 2.5: Bar chart CO2. A value needed to be verified for verifying the statement “In 2003 175000 kg CO2 is emitted” (false).

For line graphs with multiple lines, the participants needed to compare the multiple lines. For example, the statement for the line graph in Figure 2.6 was “When a car drives 120 kilometers per hour, the car uses almost twice as much fuel in the third gear as in the sixth gear” (true).

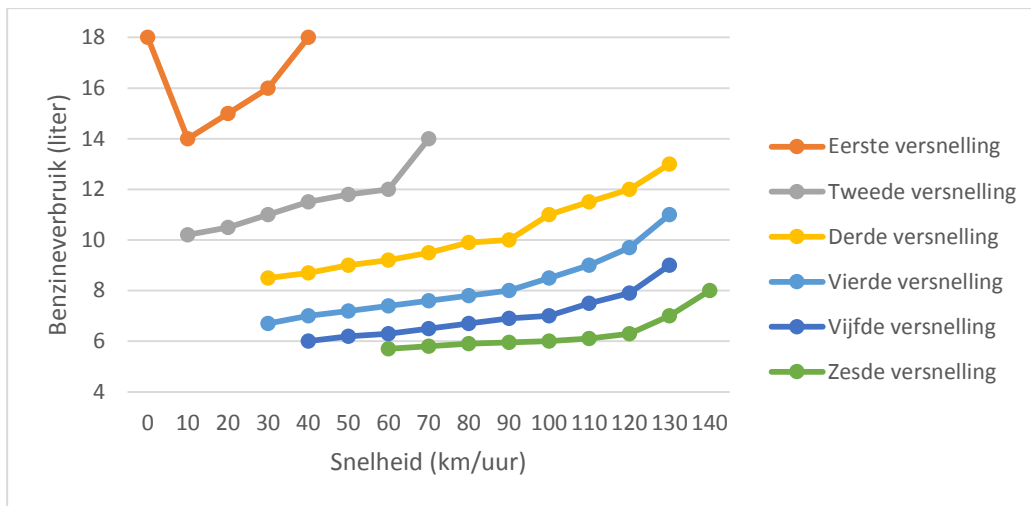


Figure 2.6: Line graph Fuel (x axis: speed (km/hour), y axis: fuel consumption (liter), legend: first gear to sixth gear). The line needed to be compared for verifying the statement “When a car drives 120 kilometers per hour, the car uses almost twice as much fuel in the third gear as in the sixth gear” (true).

For the line graphs with one line, the participants needed to describe trends. For example, the statement for the line graphs in Figure 2.7 was “The inflation is increased to 2.5 percent in 2008” (true).

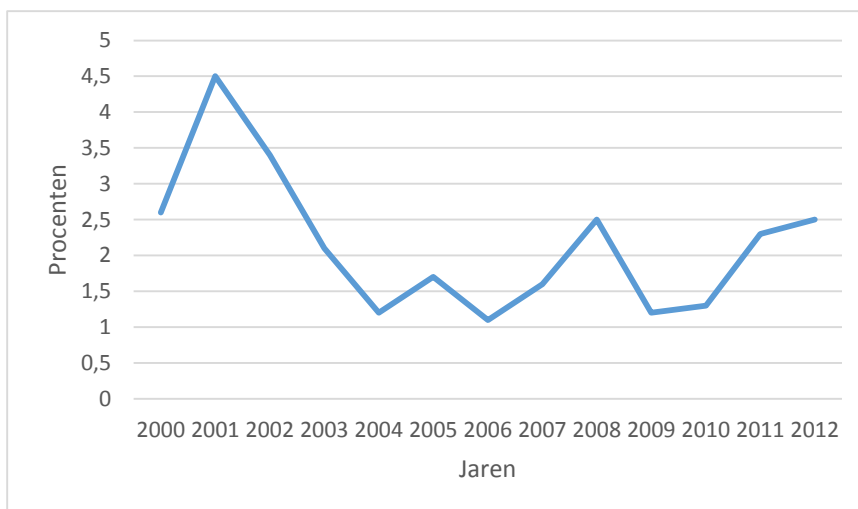


Figure 2.7: Line graph Inflation (x axis: years, y axis: percentages). Trends needed to be observed for verifying the statement “The inflation is increased to 2.5 percent in 2008” (true).

For scatterplots, the relation between the x and the y axis was verified. For example, the statement for the scatterplot in Figure 2.8 was “The relation between gaining weight and the total taken calories is positive” (true).

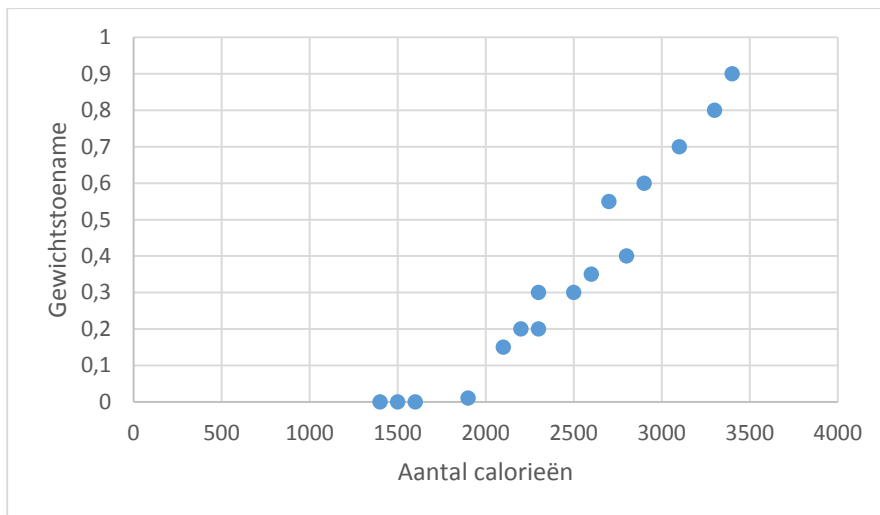


Figure 2.8: Scatterplot Gaining weight (x axis: number of calories, y axis: gaining weight). The relation between the x axis and the y axis needed to be observed for verifying the statement “The relation between gaining weight and the total taken calories is positive” (true).

Also, for scatterplots the data points needed to be counted. For example, the statement for the scatterplot in Figure 2.9 was “Most of the students take 45-55 minutes to do an exam” (true). In addition, if..then relations needed to be verified for scatterplots. The statement for the scatterplot in Figure 2.10 was “If the car is 5 years old, the price of the car would be minimal 10.000 euro” (false).

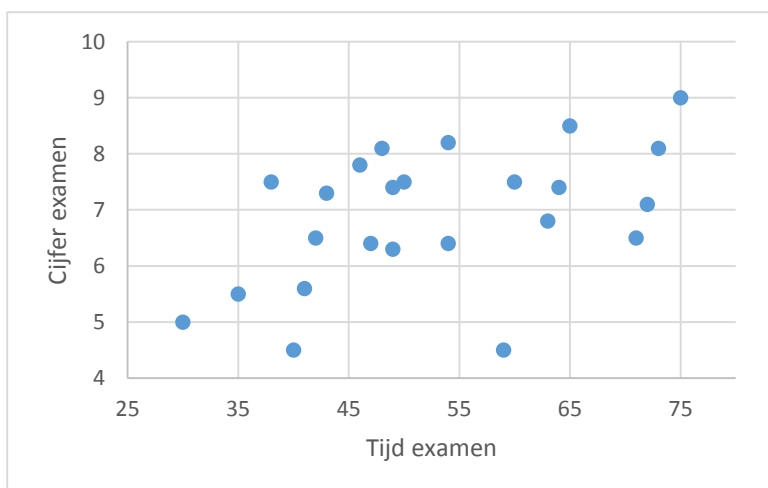


Figure 2.9: Scatterplot Exam time (x axis: time exam, y axis: grade exam). The dots needed to be counted for verifying the statement “Most of the students take 45-55 minutes to do an exam” (true).

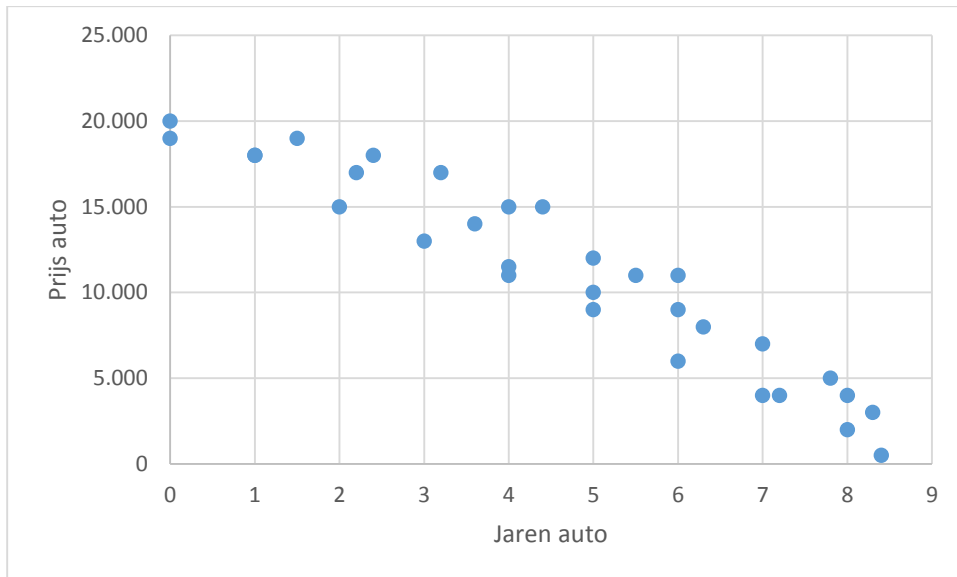


Figure 2.10: Scatterplot Pricecar (x axis: years car, y axis: price car). The if .. then relation needed to be verified for verifying the statement “If the car is 5 years old, the price of the car would be at least 10.000 euro” (false).

For pie charts the different parts needed to be compared. For example, the statement for the pie chart in Figure 2.11 was “The second favorite transport for students is the bicycle” (true).

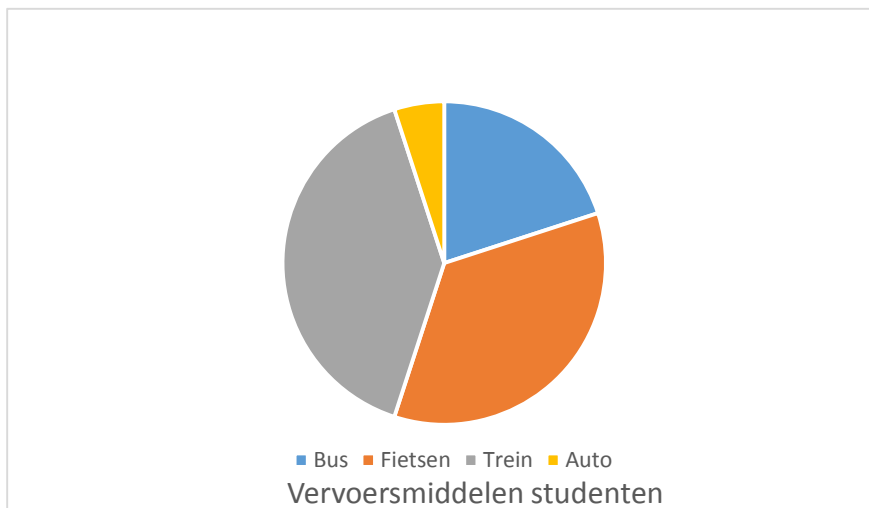


Figure 2.11: Pie chart Transport (blue: bus, orange: bicycle, grey: train, yellow: car). The parts needed to be compared for verifying the statement “The second favorite transport for students is the bicycle” (true).

For infographics one information element was chosen, for example a bar chart in the infographic. A statement was created for this element. For example, the statement for the infographic in Figure 2.12 was “Students receive more money from their parents per month than they earn by themselves” (false).



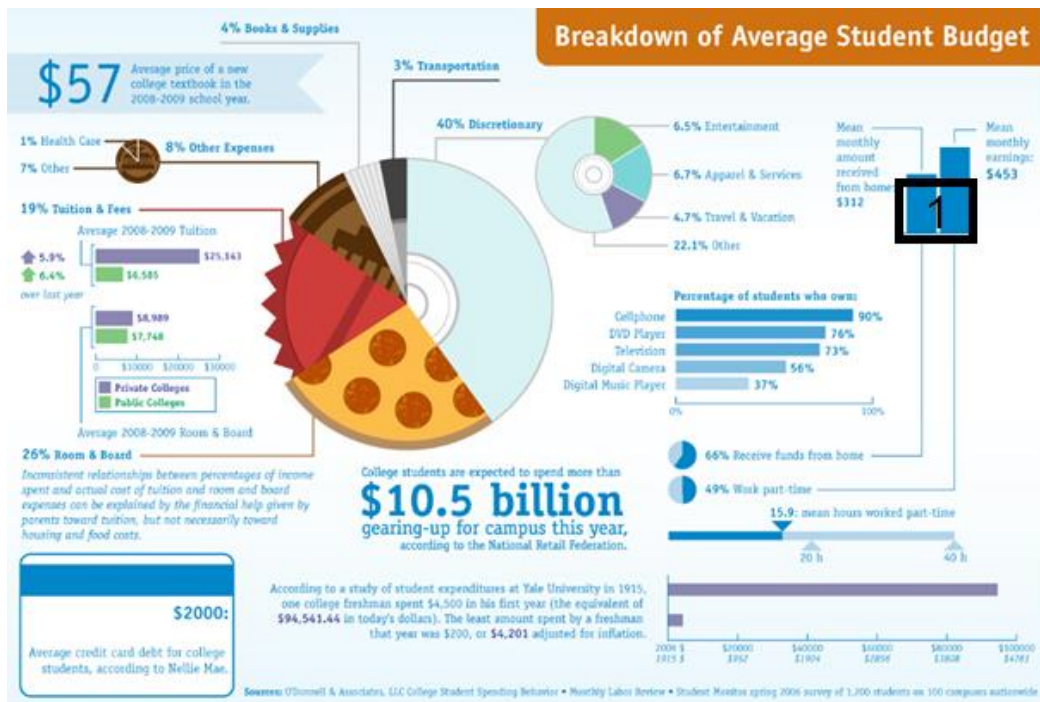


Figure 2.12: Infographic Students. One information element was chosen [1]. The information needed to be observed for verifying the statement “Students receive more money from their parents per month than they earn by themselves” (false).

Reprinted from *Breakdown of Average Student Budget* (n.d.), by Westwood College. Copyright by Westwood College. Retrieved from <http://www.westwood.edu/resources/student-budget>.

The statement for the experimental bar charts were true, and the statements for the additional bar charts were false, so the participants could not develop a strategy for the verification. For line graphs, scatterplots, pie charts, and infographics one half of the statement that were created were false and the other half was true.

### 2.1.3 Design

Two independent variables were manipulated: similarity and proximity. For the bar charts four conditions were created: no similarity and no proximity, similarity and no proximity, no similarity and proximity, and similarity and proximity. The experiment used a within subjects design.

The graphs were distributed over four versions. The participants evaluated eight different experimental bar charts in which each of the four conditions was used twice. In addition, the other types of graphs were distributed over four versions otherwise the survey was too extensively. Version 1 and version 2 consisted of each 18 graphs: eight manipulated bar charts, so each condition twice, two additional bar charts, two line graphs, two scatterplots, two pie charts, and two infographics. Version 3 and version 4 consisted of each 19 graphs: eight manipulated bar charts, so each condition twice, two additional bar charts, three line graphs, two scatterplots, two pie charts, and two infographics. For each version there was an equally distribution of true and false statements. The participants were assigned to a version randomly.

#### 2.1.4 Instruments

The difficulty of the statement was measured on a 5-point Likert scale (1 = really difficulty, 5 = really easy). Also, the verification of the statement, true or false, was filled out by the participants. In addition, gender, age and education (bachelor, premaster or master) were filled out. Also, the browser meta info was saved, so participants could be eliminated that did not view the graph in a proper way. For example, if participants filled out the survey on their mobile phone or tablet, it was possible that the graphs did not fit the page and influenced the evaluation of difficulty.

#### 2.1.5 Procedure

The participants read the instruction of the survey first. The instruction explained that the participant should read and verify the statement based on the information in the graph. When they finished reading, they clicked on the start button to start the survey. First, the participants filled in demographic information (gender, age, and education). Then they started the experiment. They studied a graph and verified the statement. Finally, the participants filled out to which extent it was difficulty to verify the statement. When the questions were answered for that particular graph, the participants clicked on the “go further” button. The participants could go back in the survey to change their answer. The survey took 10 to 15 minutes to complete.

## 2.2 Results

The effects on Difficulty, and on Correct Answers, were evaluated with univariate analyses of variance with within factors Item (immigrants, bankruptcy, health, revenue, causes of death and unemployment, diploma, and courses), Color (presence or absence), Proximity (presence or absence), and Type of Graph (bar charts, line graphs, and scatterplots). These factors were analyzed between participants, because of loss of data. In addition, pairwise comparisons were performed, when necessary with Bonferroni correction. Effect sizes were determined with eta squares.

### 2.2.1 The experimental bar charts

Table 2.1 shows the mean Difficulty for the experimental bar charts for Item, Color and Proximity. There was an effect of Item  $F(7, 124) = 4.089$ ;  $MSE = 3$ ,  $p < .001$ ,  $\eta^2 = .188$ . A pairwise comparison showed that, the statement for Immigrants was more difficult to verify than the statement for Bankruptcy:  $p < .025$ , Revenue:  $p < .005$ , and Diploma:  $p < .050$ . In addition, the statement for Unemployment was more difficult to verify than the statement for Revenue:  $p < .050$ . There was no effect of Color  $F(1, 124) < 1$ , no effect of Proximity  $F(1, 124) < 1$ , no interaction between Color and Proximity  $F(1, 124) < 1$ , no interaction between Item and Color  $F(1, 124) < 1$ , no interaction between Item and Proximity  $F(1, 124) < 1$ , and no interaction between Item, Color and Proximity  $F(1, 124) < 1$ .

Table 2.1: Mean Difficulty per Item as a function of Color (present/absent) and Proximity (present/absent) (1= really difficult, 5= really easy).

Item	Color	Proximity	
		Present	Absent
Bankruptcy	Present	4.2 (.38)	4.4 (.38)
	Absent	4.4 (.38)	4.4 (.38)
Unemployment	Present	3.4 (.38)	3.6 (.38)
	Absent	3.8 (.38)	3.6 (.38)
Courses	Present	4.3 (.43)	3.6 (.38)
	Absent	3.2 (.38)	3.8 (.38)
Revenue	Present	4.6 (.38)	4.2 (.38)
	Absent	4.8 (.43)	4.4 (.38)
Diploma	Present	4.2 (.38)	4.2 (.38)
	Absent	4.0 (.38)	4.8 (.38)
Deaths	Present	4.0 (.38)	4.4 (.38)
	Absent	4.0 (.43)	4.0 (.38)
Health	Present	3.6 (.38)	3.5 (.43)
	Absent	4.2 (.38)	4.2 (.38)
Immigrants	Present	3.6 (.38)	3.4 (.38)
	Absent	3.0 (.38)	3.6 (.38)

In total, 93.6 percent of the statements were verified correctly. Table 2.2 shows the total Correct Answers for Color and Proximity. There was no effect of Color  $F(1,76) < 1$ , no effect of Proximity  $F(1, 76) < 1$ , and no interaction between Color and Proximity  $F(1, 76) < 1$ .

Table 2.2: Correct Answers (percentage) as a function of Color (present/absent) and Proximity (present/absent).

Proximity	Color	
	Absent	Present
Absent	92.5	92.5
Present	95.0	95.0

### 2.2.2 The other types of graphs

The other graphs were analyzed for difficulty and number of correct answers, as well.

Table 2.3 shows the mean Difficulty for Type of Graph. There was an effect of Type of Graph  $F(4, 199) = 15.108$ ;  $MSE = 16$ ,  $p < .001$ ,  $\eta^2 = .233$ . A pairwise comparison showed that infographics were more difficult than bar charts:  $p < .001$ , line graphs:  $p < .005$ , and pie charts:  $p = .001$ . In addition, scatterplots were more difficult than bar charts:  $p < .001$ , line graphs:  $p < .025$ , and pie charts:  $p < .005$ . Also, line graphs were more difficult than bar charts:  $p < .010$ . Consequently, infographics and scatterplots were the most difficult graphs in comparison with the other types of graph.

Table 2.3: Mean Difficulty (1 = very difficult, 5 = very easy) as a function of Type of Graph (bar chart/line graph/infographic/pie chart/scatterplot).

Type of Graph	Difficulty (sd)
Bar chart	4.39 (0.17)
Line graph	3.63 (0.15)
Infographic	2.82 (0.17)
Pie chart	3.80 (0.17)
Scatterplot	2.91 (0.17)

For the other types of graphs, there was no effect of Difficulty for Items: Line graphs:  $F(9, 39) = 1.821$ ;  $MSE = 1.821$ ,  $p = .095$ , no differences in Difficulty between the one line graphs and the multiple line graphs  $F(1, 47) < 1$ , Bar charts:  $F(7, 31) < 1$ , Infographics:  $F(7, 31) < 1$ , Pie charts:  $F(7, 31) = 1.747$ ;  $MSE = 2.116$ ,  $p = .135$ , and Scatterplots:  $F(7, 30) = 2.295$ ;  $MSE = 2.230$ ,  $p = .053$ .

In total, 82.4% of the statements of the other graphs were verified correctly. Table 2.4 shows the total Correct Answers for Type of Graph. There was an effect of Type of Graph  $F(4, 95) = 2.964$ ;  $MSE = .213$ ,  $p < .025$ ,  $\eta^2 = .111$ . A pairwise comparison showed that bar charts more often resulted in an correct answer than line graphs:  $p = .071$ , infographics:  $p < .050$ , and scatterplots:  $p = .093$ .

Table 2.4: Correct Answers (percentages) as a function of Type of Graph (bar chart/line graph/infographic/pie chart/scatterplot).

Type of Graph	Correct answers
Bar chart	100.0
Line graph	76.7
Infographic	75.0
Pie chart	80.0
Scatterplot	77.5

### 2.3 Conclusion

Quite unexpectedly, for the experimental bar charts the difficulty was the same regardless of the presence or absence of color and proximity. In addition, most of the statements were verified correctly. It seems that the clustering of information did not help in extracting information from the graph. A possible explanation for this conclusion is the simplicity of the bar charts. Bar charts are common graphs, and it seems that viewers know how to interpret them and do not need color or proximity to extract information easier. The main research investigated the influence of similarity and proximity on the processing of information in graphs. Bar charts would be of no use in the main research, because the difficulty of extracting information was the same regardless of the clustering of information.

However, for the other types of graphs, the results showed that the statements of scatterplots and infographics were the most difficult to verify. Also, infographics resulted in more incorrect answers. Because of difficulties in verifying the statements for infographics, infographics were used for the experiment. Chunking information by using color and proximity could help the viewers in extracting information from infographics. Consequently, infographics were used for the experiment because of the faced difficulties in the material research, increasing usage for information transfer, and the capability for clustering information. The infographics in the material research were not used as experimental infographics in the experiment, because of the difficulty of clustering information in these infographics.

The infographics for the experiment were selected from newspapers, and were not tested before, because that was beyond the limited timeframe of the thesis. The specific explanation of the manipulation of the infographics is described in Chapter 3. The graphs with the lowest difficulty score (1 = really difficult) were selected for the experiment.

### **3. Method**

The present study investigated the influence of similarity and proximity on the processing of information in infographics, whether there were different viewing behaviors for bar charts, line graphs, and scatterplots, and the usability and entertainment value for graph types is researched. Pie charts and additional infographics were used as fillers in the experiment. As explained in Chapter 2, infographics, and not bar charts, were used as experimental graphs because of difficulties in verifying statements for infographics. Eight new infographics were selected for the experiment. The manipulation of the eight infographics is described below. 30 graphs were selected from the online material research that were used in the experiment: 6 bar charts, 6 line graphs, 6 scatterplots, 6 pie charts and 6 infographics. Two additional infographics were selected from the internet. In addition, 3 practice items were used.

#### **3.1 Materials**

The material consisted of 43 graphs: 8 infographics that were manipulated with color and proximity, 6 bar charts, 6 line graphs, 6 scatterplots, 6 pie charts, 8 additional infographics, and 3 practice items (see Appendices A and B). The infographics were created for this experiment as described below. The other graphs were selected from the material research as described above.

The eight experimental infographics were selected from qualitative newspapers. Four graphs were used in the Guardian: Afghanistan, Iraq, Murder and Crime, three in the Wall Street Journal: Facebook, Retirement and Car, and one in the USA Today: War. The topics of the infographics were different: information about improvised explosive devices in Afghanistan, war in Iraq, the users of Facebook, the money that is left for retirement, information about electronic cars, opinions about the war in Afghanistan, information about murders, and data about crime. To select infographics for the experiment, they needed to meet two requirements: The infographics consisted of two or more chunks, and the use of color in the infographics had no function. Chunks in infographics are mentioned as subtopics.

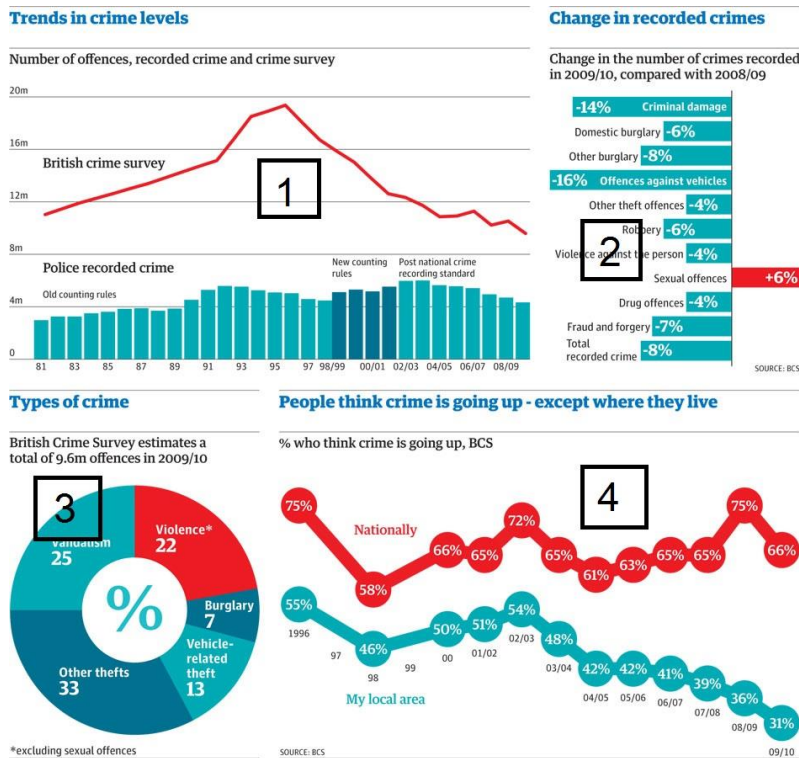


Figure 3.1: Example of infographic: Infographic Crime consisted of four subtopics [1] – [4], and the colors red and blue had no clear function.

Reprinted from *Crime statistics: get the rates were you life* (2010), by The Guardian.

Copyright 2010 by The Guardian. Retrieved from

<http://www.theguardian.com/news/datablog/2010/jul/15/crime-statistics-police-force>.

For example, Figure 3.1 consisted of four subtopics: [1] trend in crime level, [2] change in recorded crimes, [3] types of crime, and [4] opinions about crime level in local area. In addition, the colors red and blue had no function in this infographic, because the red line in subtopic [1] had no relation with the red bar in subtopic [2]. The subtopics were spaced and colored as described below.

In addition, 32 other graphs were used in the experiment: six bar charts, six line graphs and six scatterplots for investigating the viewing behavior, six pie charts and eight additional infographics that were used as fillers. As seen in the material research (Chapter 2), the graphs with the lowest difficulty score were selected for the experiment.

The eight experimental infographics were manipulated, as seen in Figure 3.2. The infographic Afghanistan in Figure 3.2 consisted of three subtopic [1] to [3]: [1] IEDs exploded and cleared, [2] civilian victims of IEDs measured by the data base, and [3] IEDs attacks per year. Infographic [A] was not manipulated with color and proximity. Infographic [B] was manipulated with color: the subtopic [1] to [3] were distinguished by different colors.



For example, the elements, as well the graph title, in subtopic [1] were colored black. Infographic [C] was manipulated with proximity: the subtopics [1] to [3] were distinguished by inserting space. Between the subtopics [1] and [2], and between the subtopics [2] and [3] a space was inserted of 100 pixels. This space had the same color as the background of the infographic, so the participants would not notice that an empty space had been inserted.

The size of the infographics were adjusted to the resolution of the monitor (1680 x 1050). The maximum length and width depended on the inserted spaces between the subtopics. For example in Figure 3.2, the infographic consisted of three subtopics, so two empty spaces needed to be inserted to distinct the subtopics which meant that the base infographic had a height of 850 pixels. The ratio of the infographics was the same. The infographics were placed on a grey screen of 1680 x 1050. The infographics were placed in the center of the 1680 x 1050 grey picture.

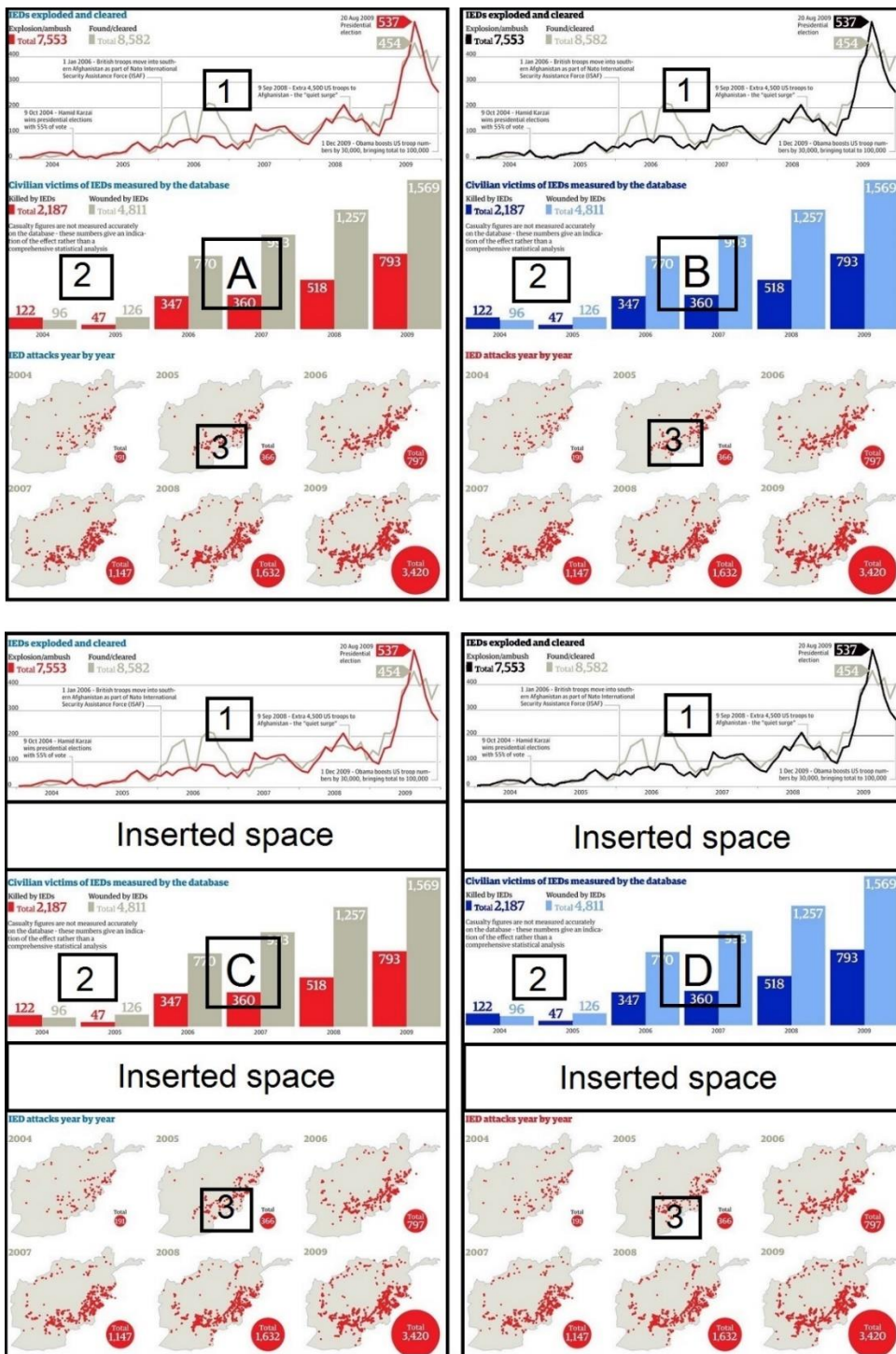


Figure 3.2: An example of the manipulation of the infographic Afghanistan. The subtopics [1] to [3] had a different color [B], between the subtopics an empty space of 100 pixels was inserted [C] or the subtopics were colored differently and empty spaces were inserted [D]. The inserted space had the same color as the background of the infographic. Adapted from *Wikileaks` Afghanistan war logs* (2011), by The Guardian. Copyright 2011 by The Guardian. Retrieved from: <http://www.theguardian.com/news/datablog/2010/jul/27/wikileaks-afghanistan-data-datajournalism>.

The most difficult statement for bar charts, line graphs, pie charts and scatterplots, and the statements for the additional infographics were used from the material research (see Chapter 2). The statements for the experimental infographics were created in the same way as for the additional infographics as seen in Chapter 2: One information element was chosen and a statement was created about that information element. It was made sure that the statements for the experimental infographics were verified in one of the subtopics. The task was to verify the statement by searching for the right subtopic (see Appendices A and B).

The statement was placed 60 pixels to the left and 60 pixels below the top left corner. One statement was placed at the position 20x20, because the graph was too large (infographic Iraq). The statement was maximum 200 pixels width, so all the statements had the same width. There were 20 false and 20 true statements. For the experimental infographics, the statements were true, and the statements for the additional infographics that served as fillers were false in which the infographics with true statements that were selected from the material research were changed into false statements. For the line graphs, pie charts, bar charts and scatterplots half of the statements were false and half were true.

### **3.2 Instruments**

The participants answered per type of graph four questions about the entertainment value and six questions about the usability per type of graph. The participants judged on a seven-point Likert scale (1= strongly disagree, 7 = strongly agree) the entertainment value and the usability value (see Appendix C). The entertainment value consisted of two positive questions: “I think a line graph is attractive” and “I think a line graph is fun to watch”, and two negative questions: “I think a line graph is boring” and “I think a line graph is confusing”. The six usability questions consisted of two questions that measured the learnability of the graph: “I have a hard time to understand a line graph” and “I know very fast how to interpret a line graph”, two questions that measured the efficiency of the graph: “I can extract information fast in a line graph” and “I think that I need to make unnecessary eye movements when I am reading a line graph”, and two questions that measured the memorability of the graph: “I can memorize information in a line graph easily” and “I have a hard time to explain how you should interpret a line graph”. Each type of usability included a positive and a negative question. In addition, the survey consisted of two ranking questions for attractiveness and difficulty (1 = not attractive/ difficulty, 5 = attractive/easy). Also, participants filled in their gender, age, and education.

### **3.3 Participants**

In total 64 participants, 43 women (67.2%), participated in the experiment of which four participants participated in the pilot. The data of these 64 participants were used for analyzing the survey, but not for the eye movement data (see below). The participants studied Communication- and Information Sciences: Four (6.3%) masters, 33 (51.6%) bachelors, and 27 (42.2%) premasters. The minimum age was 18 and the maximum age was 25. The mean age was 21.20 (sd 2.25).

The data of 45 participants, 31 women (68.9%) and 14 men (31.1%), were analyzed for the eye tracking experiment. The eye movement data of the four participants that participated in the pilot were not analyzed. Also, the eye movement data of four participants were excluded because the calibration failed: the calibration had a technical defect for two participants, and two participants wore thick lenses. In addition, the eye movement data of 11 participants were excluded, because no eye movements were registered at the statement or the data was uninterpretable. 19.6% of the participants were excluded from analyses. The participants received study credits for their participation.

### **3.4 Equipment**

To record the eye movements of the participants, the SMI RED 250 eye tracking device, with a sampling rate of 250Hz, was used. This tracking device was placed below a Dell computer. The monitor was 22'' with a resolution of 1680 x 1050. The software Experiment Centre 3.3 was used to make and run the experiment. Besides that, a webcam was placed on the computer screen.

### **3.5 Design**

The independent variables for the eye tracking experiment were color, proximity, and type of graph. The presence or absence of color and distance in infographics was meant by the independent variables color and proximity. For the infographics, four conditions were made: no similarity and no proximity, similarity and no proximity, no similarity and proximity, and similarity and proximity. Type of graph were the different graphs: bar charts, line graphs and scatterplots. The experiment used a within subjects design.

The graphs were distributed semi-randomly over four lists. In total, each list consisted of 40 graphs of which eight experimental infographics. The participants viewed eight different infographics in which each of the four conditions was used twice. The 32 other types of graphs, the bar charts, line graphs, scatterplots, pie charts, and the additional infographics, were the same among the lists.

The experimental infographics were distributed semi-randomly. In addition, the true and false statements were distributed semi-randomly, as well. In the order of graphs the topics of the graphs were taken into account. It was made sure that the topic of a filler did not influence the viewing behavior of the experimental infographics. Consequently, an experimental infographic was not be placed after a filler with the same topic.

### **3.6 Pilot**

Four master student of Communication- and Information Sciences participated in the pilot of the experiment. The pilot was conducted to test the experiment for problems. In addition, the behavior of the participants could be tested. The information was used to improve the procedure of the experiment and the experiment itself. Based on the pilot, some typos were removed and the procedure of the experiment was improved.

### **3.7 Procedure**

The experiment was conducted in two weeks. The experiment lasted per participant one hour, approximately. The instruction, the calibration and the practice items lasted 15 minutes. The eye tracking experiment lasted 30 minutes. The survey lasted 10 minutes. The experiment was conducted in a sound free booth (ca. minus 40 decibel).

Some demographical information was noted in the logbook (gender, wearing lenses or glasses, dominant eye, and color blindness). After that, the participant sat down on a chair behind the computer which displayed the experiment in a sound free booth. The goal of the experiment was explained. The participant viewed different types of graphs and should examine the graphs as they normally would after they were shown a statement that they had to verify. In addition, it was explained that the eye movements were recorded by the eye tracker. Besides that, the participants were told the webcam recorded the participant. After the instruction, the calibration was started.

The participant was explained the procedure of the experiment. The participant was told that the first screen consisted of a fixation cross in the left top corner. By viewing this fixation cross one second, a new screen appeared with a statement placed at the left top corner

and a graph in the center. It was stressed that the statement should be read first, before the participant should view the graph. It was told that after the verification of the statement, the participants needed to hit the spacebar. A screen with a fixation cross in the center appeared. By fixating at the plus, the answer screen appeared. By fixating the answer (left true and right false) the trial finished and the next trial appeared.

There were three practice items. When the three practice items were done, the participants were asked whether everything was clear. The participants could start the experiment by hitting the spacebar. When the experiment was finished, the participants left the sound free booth to fill in the survey.

### **3.8 Data analysis**

The eye movements were analyzed with the software Fixation (Cozijn, 2006). This software displayed the fixations of the participants over the graphs. Areas of interest were created for the graphs. The analyzes were performed on these areas of interest. For infographics, the areas of interest were the subtopics and the statement, see for example Figure 3.3. The subtopic that contained the information needed for verification was marked as target region. For bar charts, line graphs, and scatterplots, the areas of interest were the axes (and legend), the data area, and the statement (see Figure 3.4).

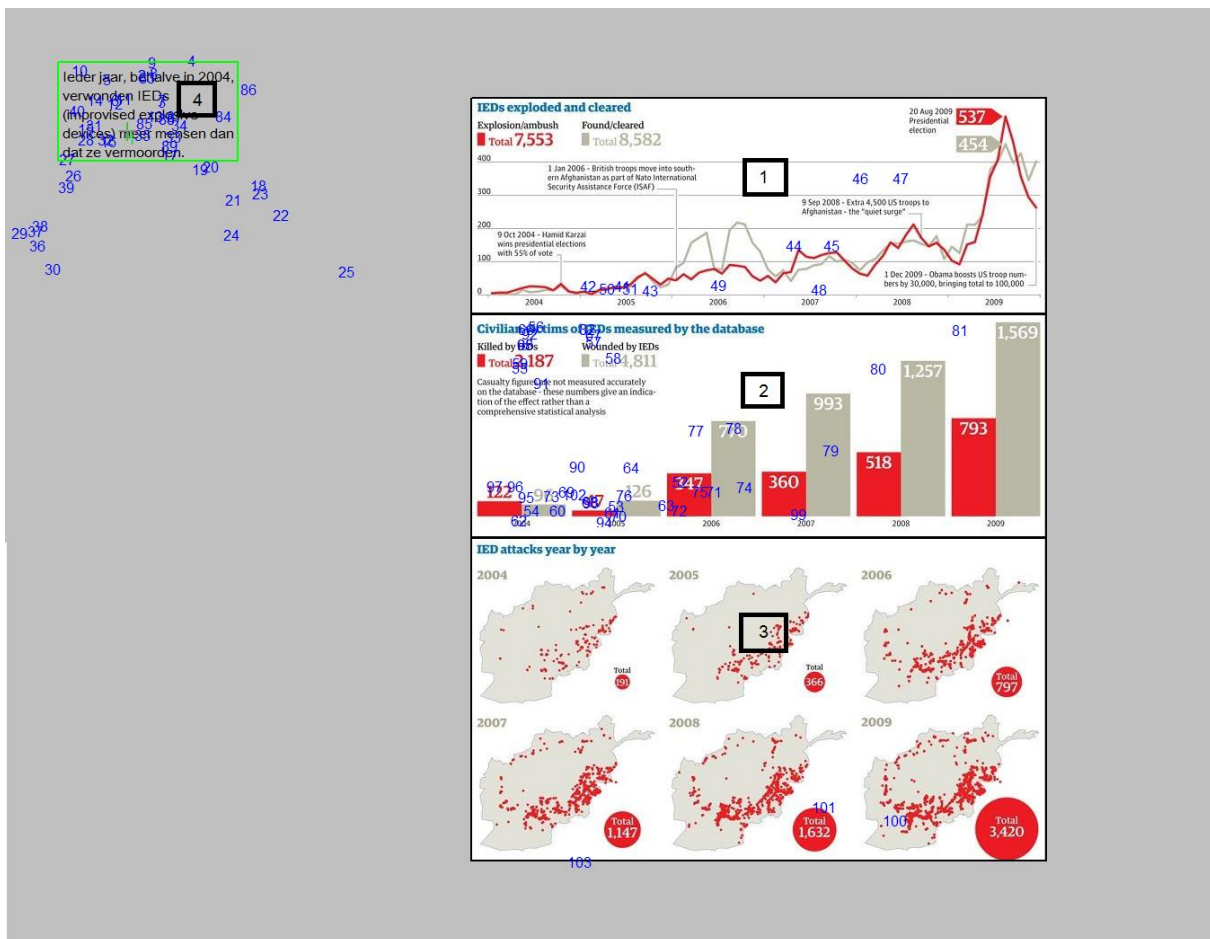


Figure 3.3: Example of areas of interest in infographics [1] to [4]. The areas of interest are indicated by squares. The infographic Afghanistan consisted of three subtopics which meant that three areas of interest were created for this infographic [1] to [3]. In addition, the statement was an area of interest in infographics [4].

Adapted from Wikileaks` *Afghanistan war logs* (2011), by The Guardian. Copyright 2011 by The Guardian. Retrieved from:

<http://www.theguardian.com/news/datablog/2010/jul/27/wikileaks-afghanistan-data-datajournalism>.

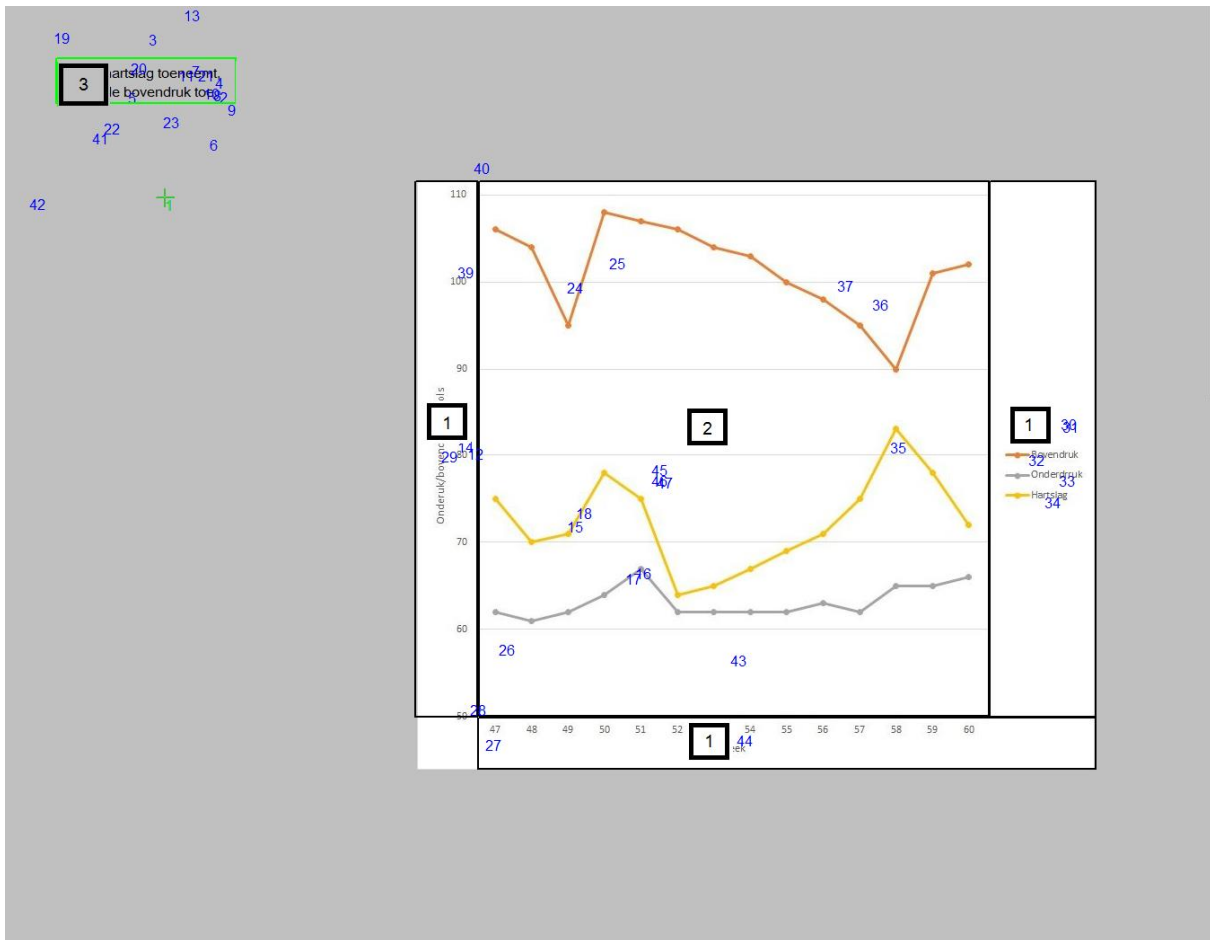


Figure 3.4: Example of areas of interest in line graphs [1] to [3]. The areas of interest are indicated by squares. The line graph heartbeat consisted of three areas of interest: [1] the axes and the legend, [2] the data area, and [3] the statement.

The fixations were assigned to the areas of interest by the software automatically and were corrected when needed. As seen in Figure 3.3, not all fixations were assigned to the area of interest of the statement automatically. The fixations that were not assigned to an area of interest, needed to be assigned manually. Also, fixations that were assigned to a particular area of interest automatically could be unassigned and assigned to another area of interest manually. For example, the participants viewed the different subtopics. If the duration of a fixation was less than 100 milliseconds and the next fixation was on another area of interest, that fixation was assigned to that area of interest. To judge which answer the participants had given, the last fixation on the answer screen was used.

The data of four graphs (two line graphs and two scatterplots) were excluded: priceperkm, fuel, age men women and reading score. The reason was that the lines and the scatters were close to the axes, making it very difficult to distinguish the eye movements at the axes from those of the line or scatter (see Figure 3.5).



Finally, for eight participants the eye tracking recording failed on one or more graphs (in total 13 graphs). There was no data registered at the statement, so the data was incomplete. These trials were also excluded from the experiment. 15.4% of the experimental graphs were excluded, and overall 16.5% of the total data was excluded.

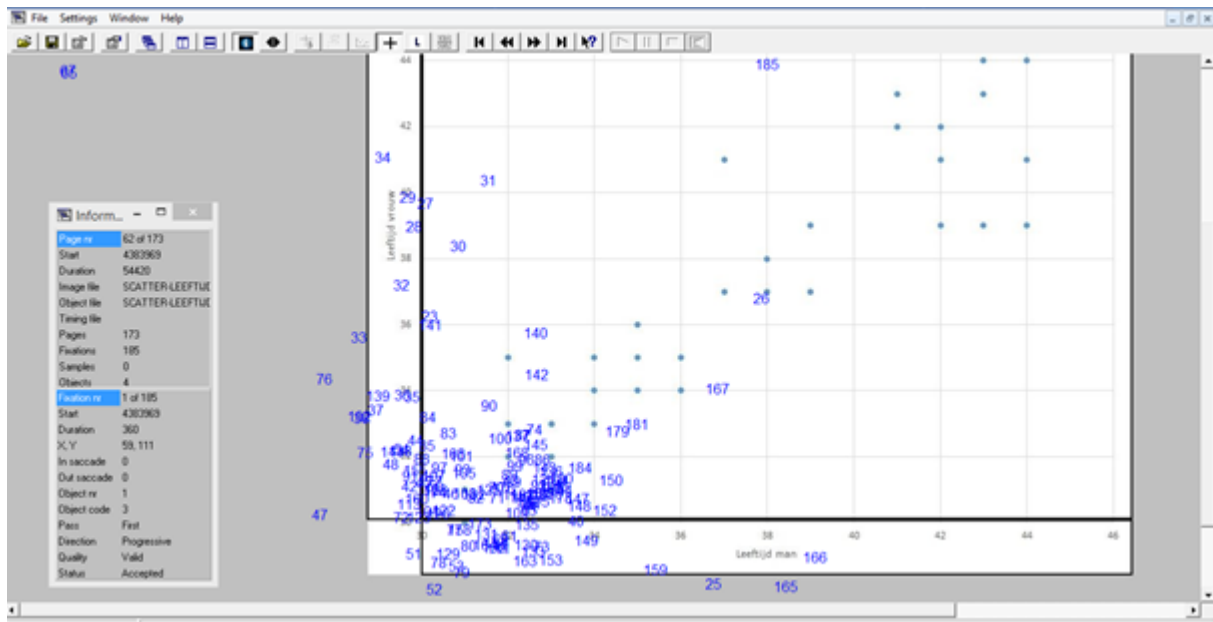


Figure 3.5: Example of uninterpretable data: The scatters were so close to the axes, that no clear distinction could be made between the fixations on the axes and on the data area. This made the data uninterpretable.

The software stored the viewing paths of the participants. The page duration, gaze duration, number of switches, and viewing frequency were stored. In addition, the correct answers were analyzed. Correct answers were the correct answers per participant.

Page duration was the time from onset of the graph on screen until the spacebar was hit to leave the page.

Gaze duration was the viewing time of an area of interest until switched to another area of interest. It included the fixation durations and saccade durations on that area of interest.

Number of switches were the number of times a participant changed between the statement, the target (the subtopic that contained the information for verifying the statement), and the other subtopics for infographics. When a participant fixated at the statement and after that at the target, this is one change. If the participant would observe another subtopic after that, that would count for another change.

Number of switches were the number of times of changing between the statement, the data area, the axes, and the legend for bar charts, line graphs, and scatterplots.

The viewing frequency was the count of the occurrences of first viewing an area of interest in the graph after reading the statement.

When preparing the data for the analyses, large differences in the number of fixations were noticed between the infographics. For instance, a participant fixated the infographic Afghanistan 74 times, Iraq was fixated 208 times, Facebook was fixated 173 times, Murder was fixated 54 times, Auto was fixated 35 times, War was fixated 29 times, Retirement was fixated 127 times, and Crime was fixated 97 times. Consequently, some infographics were fixated more than other infographics. Therefore, the Page Duration per infographic was analyzed. To decrease the variation in the data, two groups of infographics were made. Four of the eight infographics had more fixations and longer viewing times than the other four infographics. When an infographic is observed, the information is processed. So, when infographics are observed longer than other infographics, these infographics are more difficult to process. In this way, two groups could be made: easy infographics and difficult infographics. The mean Page Duration for Difficulty was analyzed (1.7% of the data were outliers) (easy: 28183 ms difficult: 52353 ms). There was an effect of Difficulty  $F(1, 341) = 170.863$ ;  $MSE = 50925829848$ ,  $p < .001$ ,  $\eta^2 = .334$ . The Page Duration was longer for difficult infographics than for easy infographics. The difficult infographics were Iraq, Facebook, Retirement and Crime, and the easy infographics consisted of Afghanistan, Murder, Auto and War. When relevant the data was analyzed for easy and difficult infographics, separately.

In the survey, per type of graph, the entertainment, learnability, efficiency, and memorability value was measured on a seven-point Likert scale. The different items for each component was tested for internal consistency. The attractiveness and the difficulty of the graphs was measured on one item. Attractiveness was the order of the type of graphs for attractiveness. Difficulty was the order of the type of graphs for difficulty.

The entertainment value was measured with four items. The Cronbach's Alpha was .695. One of the items, (" I think this graph is confusing"), did not correlate with the other three items. When this item was deleted, the Cronbach's Alpha was .885. This indicated a high level of internal consistency for the entertainment scale. The learnability of the graphs were measured with two items. The Cronbach's Alpha was .880. The efficiency was measured with two items. The items were internally consistent, Cronbach's Alpha was .864. The memorability was measured with two items. The memorability scale was internally consistent, Cronbach's Alpha .734.

## **4. Results**

In this paragraph the results are reported in relation to the hypotheses. First, the eye tracking results are reported and then the results of the survey. The eye tracking results for the infographics and for the other types of graphs are reported, separately. For infographics the viewing times on the subtopic that contained information for verifying the statement which is called the target region, and the total viewing times on the infographics and the statements were analyzed. For the other types of graphs the viewing times on the data area and the axes, and the total viewing times on the graphs and the statements were analyzed. The eye tracking sections start with a description of the analyses that have been performed. The dependent variables were page duration, gaze duration, number of switches, and correct answers. In addition, the viewing frequency was reported for the other types of graphs. The survey section starts with the description of the dependent variables, and the analyses that have been performed. The dependent variables for the survey were entertainment, learnability, memorability, efficiency, attractiveness, and difficulty. Finally, a summary of the results is given.

### **4.1 Eye tracking: infographics**

The effects on Page Duration and on Gaze Duration were evaluated with univariate analyses of variance with within factors Color (presence or absence), and Proximity (presence or absence). For these analyses between participants analyses were performed, because of loss of data. In addition, the effects on Page Duration and on Gaze Duration were evaluated with univariate analyses of variance with within factors Color (presence or absence), Proximity (presence or absence), and Difficulty (easy or difficult). These analyses were executed with item analyses. The effects on Number of Switches, and on Correct Answers, were evaluated with univariate analysis of variance with within factors Color (presence or absence), Proximity (presence or absence), and Difficulty (easy or difficult). These analyses were executed with item analyses, too. In addition, pairwise comparisons were performed with Bonferroni correction, when necessary. The effect sizes were determined with eta squares. Outliers, the deviation of the participant mean was larger than three times the standard deviation, in viewing times were excluded.

#### 4.1.1 Page duration

Table 4.1 shows the mean Page Durations for Color and Proximity (0.9% of the data were outliers). There was no effect of Color  $F(1, 164) < 1$ , no effect of Proximity  $F(1, 164) = 1.508, p = .221$ , and no interaction between Color and Proximity  $F(1, 164) = 1.037, p = .310$ .

Table 4.1: Mean Page Durations (ms) as a function of Color (present/absent) and Proximity (present/absent).

Color	Proximity	
	Present	Absent
Present	41390	36876
Absent	40431	40010

Table 4.2 shows the mean Page Durations for Difficulty, Color and Proximity (0.7% of the data were outliers). There was an effect of Difficulty  $F(1, 24) = 46.363; MSE = 4855303783, p < .001, \eta^2 = .665$ . The difficult infographics were observed longer than the easy infographics. There was no effect of Color  $F(1, 24) < 1$ , no effect of effect of Proximity  $F(1, 24) < 1$ , no interaction between Difficulty and Color  $F(1, 24) < 1$ , no interaction between Difficulty and Proximity  $F(1, 24) < 1$ , no interaction between Color and Proximity  $F(1, 24) < 1$ , and no interaction between Difficulty, Color and Proximity  $F(1, 24) < 1$ .

Table 4.2: Mean Page Durations (ms) as a function of Difficulty (easy/difficult), Color (present/absent) and Proximity (present/absent).

Difficulty	Color	Proximity	
		Present	Absent
Easy	Present	34573	31769
	Absent	31496	29882
Difficult	Present	57171	55765
	Absent	55516	57811

#### 4.1.2 Gaze Duration on the target region

Table 4.3 shows the mean Gaze Durations on the target region for Color and Proximity (1.9% of the data were outliers). The target region, a subtopic, contained information that need to be used for verifying the statement. There was no effect of Color  $F(1, 164) < 1$ , no effect of Proximity  $F(1, 164) = 2.277, p = .133$ , and no interaction between Color and Proximity  $F(1, 164) = 2.524, p = .114$ .

Table 4.3: Mean Gaze Durations (ms) on the target region as a function of Color (present/absent) and Proximity (present/absent).

Color	Proximity	
	Present	Absent
Present	3789	3079
Absent	3210	3229

Table 4.4 shows the mean Gaze Durations on the target region for Difficulty, Color and Proximity (1.9% of the data were outliers). There was no effect of Difficulty  $F(1, 24) < 1$ , no effect of Color  $F(1, 24) < 1$ , no effect of Proximity  $F(1, 24) = 1.666; MSE = 1143676, p = .209$ , no interaction between Difficulty and Color  $F(1, 24) = 1.414; MSE = 970556, p = .246$ , no interaction between Difficulty and Proximity  $F(1, 24) < 1$ , no interaction between Color and Proximity  $F(1, 24) = 1.622; MSE = 1113295, p = .215$ , and no interaction between Difficulty, Color and Proximity  $F(1, 24) < 1$ .

Table 4.4: Mean Gaze Durations (ms) on the target region as a function of Difficulty (easy/difficult), Color (present/absent) and Proximity (present/absent).

Difficulty	Color	Proximity	
		Present	Absent
Easy	Present	3587	2572
	Absent	3189	3404
Difficult	Present	3587	3099
	Absent	2976	2750

In the easy group, there was no effect of Color  $F(1, 12) = 1.101$ ;  $MSE = 188089$ ,  $p = .315$ . There was a trend of Proximity  $F(1, 12) = 3.734$ ;  $MSE = 637887$ ,  $p = .077$ ,  $\eta^2 = .237$ . When the infographics were easy and the Proximity principle was present, the target region was observed longer than Proximity was absent. There was an interaction between Color and Proximity  $F(1, 12) = 8.876$ ;  $MSE = 1516412$ ,  $p < .015$ ,  $\eta^2 = .425$ . A pairwise comparison showed that the target region was observed longer at the Color/Proximity condition than at the Color/No Proximity condition:  $p < .050$ . In addition, the target region was observed longer at the No Color/No Proximity condition than at the Color/No Proximity condition:  $p = .088$ . There was no interaction between Difficulty and Proximity  $F(1, 24) < 1$ .

In the difficult group, there was no effect of Color  $F(1, 12) < 1$ , no effect of Proximity  $F(1, 12) < 1$ , and no interaction between Color and Proximity  $F(1, 12) < 1$ .

When Color was present, there was no effect of Difficulty  $F(1, 14) < 1$ . In addition, when Color was absent, there was no effect of Difficulty  $F(1, 14) < 1$ . When Proximity was present, there was no effect of Difficulty  $F(1, 14) < 1$ . In addition, when Proximity was absent, there was no effect of Difficulty  $F(1, 14) < 1$ .

#### 4.1.3 Number of Switches

Table 4.5 shows the Number of Switches for Difficulty, Color and Proximity. There was an effect of Difficulty  $F(1, 24) = 9.503$ ;  $MSE = 214$ ,  $p = .005$ ,  $\eta^2 = .284$ . There were more switches while observing difficult infographics than easy infographics. There was no effect of Color  $F(1, 24) < 1$ , no effect of Proximity  $F(1, 24) < 1$ , no interaction between Difficulty and Color  $F(1, 24) < 1$ , no interaction between Difficulty and Proximity  $F(1, 24) < 1$ , and no interaction between Difficulty, Color and Proximity  $F(1, 24) < 1$ .

Table 4.5: Mean Number of Switches as a function of Difficulty (easy/difficult), Color (present/absent) and Proximity (present/absent) (minimum = 4.8, maximum = 26.5).

Difficulty	Color	Proximity	
		Present	Absent
Easy	Present	7.4	8.5
	Absent	6.8	7.4
Difficult	Present	11.8	13.6
	Absent	12.6	12.8

#### 4.1.4 Correct Answers

Overall, 83.7% of the answers were correct for infographics. For the easy infographics, 78.9% of the answers were correct. Difficult infographics were answered correctly 88.6% of the time. Table 4.6 shows the Correct Answers for Difficulty, Color and Proximity. There was no effect of Difficulty  $F(1, 24) = 1.441$ ;  $MSE = .076$ ,  $p = .242$ , no effect of Color  $F(1, 24) < 1$ , no effect of Proximity  $F(1, 24) < 1$ , no interaction between Difficulty and Color  $F(1, 24) < 1$ , no interaction between Difficulty and Proximity  $F(1, 24) < 1$ , no interaction between Color and Proximity  $F(1, 24) < 1$ , and no interaction between Color, Proximity and Difficulty  $F(1, 24) < 1$ .

Table 4.6: Correct Answers (percentages) as a function of Difficulty, (easy/difficult), Color (present/absent) and Proximity (present/absent).

Difficulty	Color	Proximity	
		Present	Absent
Easy	Present	84.5	77.0
	Absent	74.3	79.5
Difficult	Present	88.6	82.2
	Absent	95.5	88.2

Table 4.7 shows the Page Durations for Correct Answers for Difficulty, Color and Proximity (1.0% of the data were outliers). There was an effect of Difficulty  $F(1, 24) = 70.089$ ;  $MSE = 4281159453$ ,  $p < .001$ ,  $\eta^2 = .745$ . When the infographics were difficult, the infographics were observed longer than when the infographics were easy. There was no effect of Color  $F(1, 24) < 1$ , no effect of effect of Proximity  $F(1, 24) < 1$ , no interaction between Difficulty and Color  $F(1, 24) < 1$ , no interaction between Difficulty and Proximity  $F(1, 24) < 1$ , no interaction between Color and Proximity  $F(1, 24) < 1$ , and no interaction between Difficulty, Color and Proximity  $F(1, 24) < 1$ .

Table 4.7: Mean Page Durations (ms) as a function of Difficulty (easy/difficulty), Color (present/absent) and Proximity (present/absent).

Difficulty	Color	Proximity	
		Present	Absent
Easy	Present	28543	28028
	Absent	28113	29356
Difficulty	Present	54391	51569
	Absent	47132	53481

Table 4.8 shows the Gaze Durations on the target region for Correct Answers for Difficulty, Color and Proximity (1.9% of the data were outliers). There was no effect of Difficulty  $F(1, 24) < 1$ , no effect of Color  $F(1, 24) < 1$ , no effect of Proximity  $F(1, 24) < 1$ , no interaction between Difficulty and Color  $F(1, 24) = 1.986$ ;  $MSE = 2131160$ ,  $p = .172$ , no interaction between Difficulty and Proximity  $F(1, 24) < 1$ , no interaction between Color and Proximity  $F(1, 24) = 1.763$ ;  $MSE = 1892476$ ,  $p = .197$ , no interaction between Difficulty, Color and Proximity  $F(1, 24) < 1$ .

Table 4.8: Mean Gaze Durations on the target region (ms) as a function of Difficulty (easy/difficult), Color (present/absent) and Proximity (present/absent).

Difficulty	Color	Proximity	
		Present	Absent
Easy	Present	3421	2816
	Absent	3148	4142
Difficulty	Present	3653	3097
	Absent	2974	2764

#### 4.2 Eye tracking: bar charts, line graphs and scatterplots

The effects on Page Duration, on Number of Switches and on Correct Answers were evaluated with univariate analyses of variance with within factor Type of Graph (bar charts, line graphs, and scatterplots). These dependent variables were analyzed between participants, because of loss of data. The effect on Gaze Duration and on Viewing Frequency were evaluated with univariate analyses of variance with within factors Type of Graph (bar charts,



line graphs, and scatterplots), and Object (axes and data area). These dependent variables were analyzed between participants, because of loss of data. In addition, pairwise comparisons were performed with Bonferroni correction, when necessary. The viewing frequency on the axes and the data areas were evaluated with Chi Square tests. The effect sizes were determined with eta squares.

#### 4.2.1 Page Duration

Table 4.9 shows the mean Page Durations for Type of Graph (1.4% of the data were outliers). There was no effect of Type of Graph  $F(2, 129) = 2.265$ ;  $MSE = 60110156$ ,  $p = .108$ .

Table 4.9: Mean Page Durations (ms) as a function of Type of Graph (bar charts/line graphs/scatterplots).

Type of Graph	Mean Page Durations
Bar chart	19441
Line graph	20615
Scatterplot	18303

#### 4.2.2 Gaze Duration

Table 4.10 shows the mean Gaze Durations for Object and Type of Graph (2.2% of the data were outliers). Overall, there was an effect of Object  $F(1, 264) = 17.782$ ;  $MSE = 893410$ ,  $p = .001$ ,  $\eta^2 = .063$ . The data areas were observed longer than the axes. There was no effect of Type of Graph  $F(2, 264) = 2.832$ ;  $MSE = 142303$ ,  $p = .061$ . There was an interaction between Object and Type of Graph  $F(2, 264) = 73.199$ ;  $MSE = 3677777$ ,  $p < .001$ ,  $\eta^2 = .357$ . The data area was observed longer for scatterplots than for bar charts. For bar charts, there was an effect of Object  $F(1, 82) = 70.408$ ;  $MSE = 2601032$ ,  $p < .001$ ,  $\eta^2 = .462$ . The axes were observed longer than the data area. For line graphs and scatterplots, there were also effects of Object  $F(1, 82) = 43.075$ ;  $MSE = 1527911$ ,  $p < .001$ ,  $\eta^2 = .344$ , and  $F(1, 82) = 53.965$ ;  $MSE = 4150675$ ,  $p < .001$ ,  $\eta^2 = .397$ , but in a different direction. The data area was observed longer than the axes.

Table 4.10: Mean Gaze Durations (ms) as a function of Object (axes/data area) and Type of Graph (bar charts/line graphs/scatterplots).

Type of Graph	Object	
	Axes	Data Area
Bar chart	965	625
Line graph	703	964
Scatterplot	660	1090

#### 4.2.3 Gaze Duration of the first view on the graphs

Table 4.11 shows the mean Gaze Durations of the first view on the graphs for Object and Type of Graph (1.9% of the data were outliers). Overall, there was an effect of Object  $F(1, 227) = 136.594$ ;  $MSE = 34301756$ ,  $p < .001$ ,  $\eta^2 = .376$ . Axes were observed longer than the data area during the first view on the graph. There was an effect of Type of Graph  $F(2, 227) = 7.973$ ;  $MSE = 2002283$ ,  $p < .001$ ,  $\eta^2 = .066$ . A pairwise comparison showed that during the first view, an object was observed longer in bar charts than in scatterplots:  $p < .005$ . These effects were due to the interaction between Object and Type of Graph  $F(2, 227) = 7.859$ ;  $MSE = 1973440$ ,  $p < .001$ ,  $\eta^2 = .065$ . The axes were observed longer for bar charts than for scatterplots. For bar charts, there was an effect of Object  $F(1, 75) = 73.939$ ;  $MSE = 23987577$ ,  $p < .001$ ,  $\eta^2 = .496$ . During the first view on the graph, the axes were observed longer than the data area. For line graphs, there was an effect of Object  $F(1, 70) = 30.257$ ;  $MSE = 9046383$ ,  $p < .001$ ,  $\eta^2 = .302$ . During the first view on the graph, the axes were observed longer than the data area. For scatterplots, there was an effect of Object  $F(1, 64) = 30.725$ ;  $MSE = 4374069$ ,  $p < .001$ ,  $\eta^2 = .324$ . The axes were observed longer than the data area during the first view on the graph.

Table 4.11: Mean Gaze Durations of the first view on the graphs (ms) as a function of Object (axes/data area) and Type of Graph (bar charts/line graphs/scatterplots).

Type of Graph	Object	
	Axes	Data Area
Bar chart	1458	360
Line graph	1074	387
Scatterplot	831	334

#### 4.2.4 Viewing Frequency

Table 4.12 shows the Viewing Frequency for Type of Graph. Overall, there was an effect of Object  $\chi^2(1) = 13632, p = .001$ . For bar charts, there was an effect of Object  $\chi^2(1) = 27.930, p < .001$ . After reading the statement, the first view was more often on the axes than on the data area. For line graphs and scatterplots, there were no effects of Object  $\chi(1) = .022, p = .881$ , and  $\chi(1) = .360, p = .549$ , respectively.

Table 4.12: Viewing Frequency as a function of Object (axes/data area) and Type of Graph (bar charts/line graphs/scatterplots) (Overall, 362 times axes and 265 times data area).

Type of Graph	Object	
	Axes	Data Area
Bar chart	179	92
Line graph	90	88
Scatterplot	93	85

#### 4.2.5 Number of Switches

Table 4.13 shows the mean Number of Switches for Type of Graph. There was an effect of Type of Graph  $F(2, 123) = 12.138; MSE = 107, p < .001, \eta^2 = .165$ . A pairwise comparison showed that there were more switches when reading line graphs than when reading bar charts:  $p < .025$  and scatterplots  $p < .001$ .

Table 4.13: Mean Number of Switches as a function of Type of Graph (bar charts/line graphs/scatterplots) (minimum = 3.7, maximum = 19.9).

Type of Graph	Number of Switches
Bar chart	8.2
Line graph	10.0
Scatterplot	7.0

#### 4.2.6 Correct Answers

Table 4.14 shows the total Correct Answers for Type of Graph. There was an effect of Type of Graph  $F(2, 123) = 10.210$ ;  $MSE = .291$ ,  $p < .001$ ,  $\eta^2 = .142$ . A pairwise comparison showed that scatterplots more often resulted in an incorrect answer compared to the other graph types. The other graph types were bar charts:  $p < .001$  and line graphs:  $p < .005$ .

Table 4.14: Correct Answers (percentages) as a function of Type of Graph (bar charts/line graphs/scatterplots).

Type of Graph	Correct Answers
Bar chart	94.8
Line graph	91.6
Scatterplot	79.2

Table 4.15 shows the mean Page Durations for Correct Answers for Type of Graph (1.2% of the data were outliers). There is an effect of Type of Graph  $F(2, 123) = 3.994$ ;  $MSE = 103550055$ ,  $p < .025$ ,  $\eta^2 = .061$ . A pairwise comparison showed that line graphs were observed longer than scatterplots:  $p < .025$ .

Table 4.15: Mean Page Durations (ms) for Correct Answers as a function of Type of Graph (bar charts/line graphs/scatterplots).

Type of Graph	Page Duration
Bar chart	19433
Line graph	20272
Scatterplot	17305

Table 4.16 shows the mean Gaze Durations for Correct Answers for Object and Type of Graph (2.0% of the data were outliers). There was an effect of Object  $F(2, 264) = 12.822$ ;  $MSE = 903298$ ,  $p < .001$ ,  $\eta^2 = .046$ . The data area was observed longer than the axes. There was an effect of Type of Graph  $F(2, 264) = 3.597$ ;  $MSE = 253413$ ,  $p < .050$ ,  $\eta^2 = .027$ . A pairwise comparison showed that the objects were observed longer for scatterplots than for bar charts:  $p = .087$ . These effects were due to the interaction between Object and Type of Graph  $F(2, 264) = 52.055$ ;  $MSE = 3667185$ ,  $p < .001$ ,  $\eta^2 = .283$ . The data area was observed

longer for scatterplots than for bar charts. For bar charts, there was an effect of Object  $F(1, 82) = 66.806$ ;  $MSE = 2560264$ ,  $p < .001$ ,  $\eta^2 = .449$ . The axes were observed longer than the data area. For line graphs and scatterplots, there were effects of Object  $F(1, 82) = 43.634$ ;  $MSE = 1457548$ ,  $p < .001$ ,  $\eta^2 = .347$ , and  $F(1, 82) = 29.197$ ;  $MSE = 4185748$ ,  $p < .001$ ,  $\eta^2 = .263$ , but in a different direction. The data area was observed longer than the axes.

Table 4.16: Mean Gaze Durations (ms) for Correct Answers as a function of Object (axes/data area) and Type of Graph (bar charts/line graphs/scatterplots).

Type of Graph	Object	
	Axes	Data Area
Bar chart	948	611
Line graph	707	962
Scatterplot	669	1101

### 4.3 Survey: bar charts, line graphs, infographics, pie charts and scatterplots

#### 4.3.1 Dependent variables

Entertainment is the extent to which a graph is attractive, fun to watch and amusing.

Learnability is the extent to which a graph is easy to understand and to interpret.

Efficiency is the extent to which information is extracted visually with ease and information is verified in a short time.

Memorability is the extent to which information of a graph is memorized easily and the interpretation of a graph is easy to explain.

Attractiveness is the order of the type of graphs for attractiveness.

Difficulty is the order of the type of graphs for difficulty.

#### 4.3.2 Analyses

The homogeneity of the dependent variables was determined with Cronbach's Alpha. The effects on Entertainment, Learnability, Efficiency, Memorability, Attractiveness, and Difficulty were evaluated with General Linear Models with within factors Type of Graph (bar charts, line graphs, and scatterplots). For these analyses within participants analyses were performed. The effect sizes were determined with eta squares.

### 4.3.3 Overview results survey

An overview of the results for the survey is given in Table 4.17 and in Figure 4.1.

Table 4.17: The mean scores and standard deviation of Entertainment, Learnability, Efficiency, Memorability, Attractiveness, and Difficulty.

(Entertainment/Learnability/Memorability/Efficiency: 1 = strongly disagree, 7 = strongly agree, Attractiveness/Difficulty: 1 = unattractive/difficult, 5 = attractive/easy) as a function of Type of Graph (bar chart/line graph/infographic/pie chart/scatterplot).

Factor	Bar charts	Line graphs	Infographics	Pie Charts	Scatterplots
Entertainment	4.46 (1.15)	3.84 (1.09)	5.07 (1.45)	4.65 (1.00)	3.48 (1.09)
Learnability	5.66 (0.76)	5.70 (0.89)	3.46 (1.22)	6.09 (0.97)	4.45 (1.24)
Memorability	5.34 (0.84)	5.21 (0.84)	3.84 (1.09)	5.70 (0.92)	3.80 (1.32)
Efficiency	5.13 (1.06)	5.09 (0.99)	2.51 (1.13)	5.61 (0.99)	3.77 (1.14)
Attractiveness	3.52 (0.18)	3.09 (0.19)	3.41 (0.25)	3.58 (0.19)	2.34 (0.24)
Difficulty	4.00 (0.17)	3.78 (0.15)	1.89 (0.22)	4.00 (0.19)	2.27 (0.19)

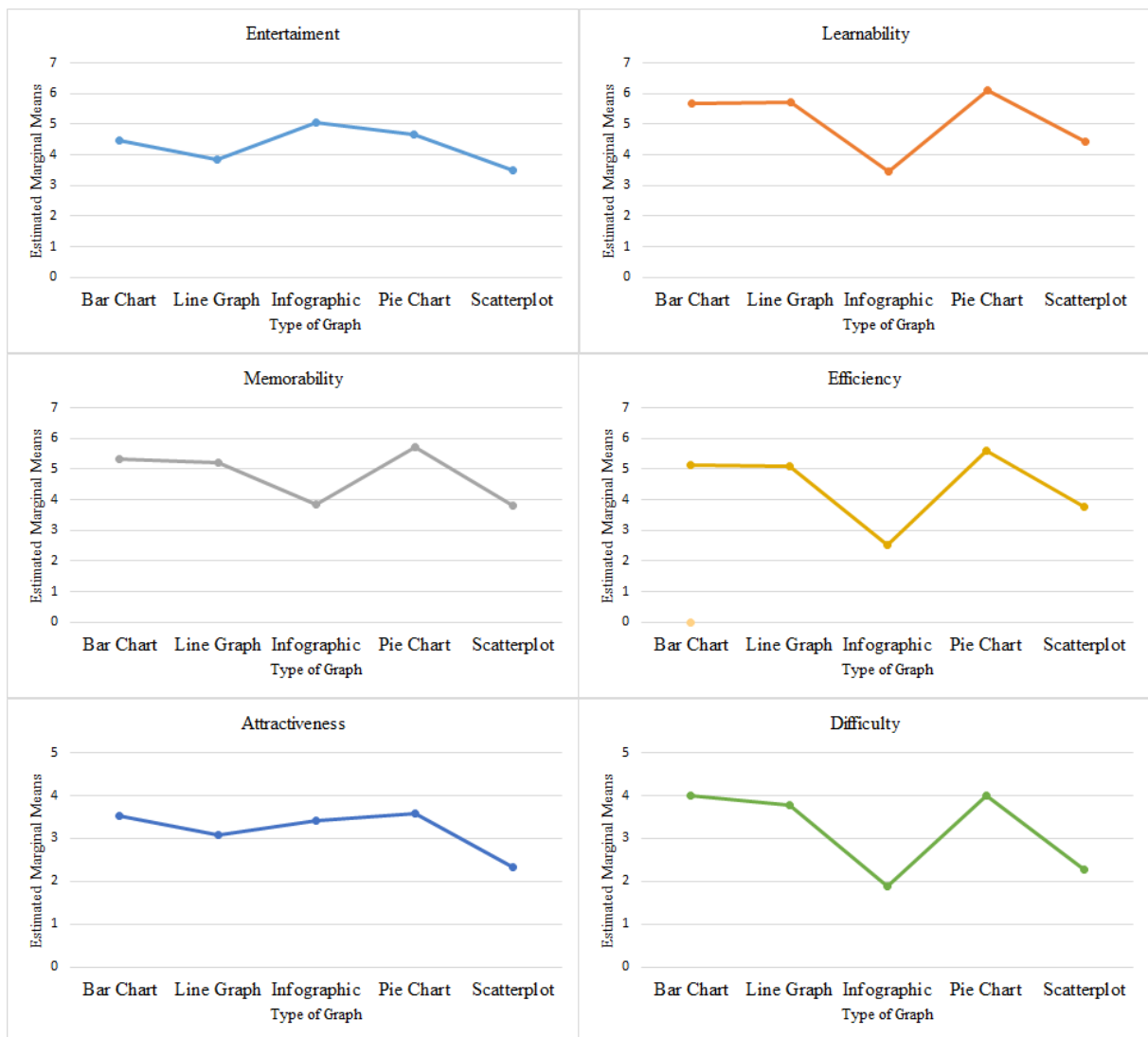


Figure 4.1: Mean scores of Entertainment, Learnability, Memorability, Efficiency, Attractiveness and Difficulty value (1 = strongly disagree, 7 = strongly agree, except Attractiveness and Difficulty: 1 = unattractive/difficulty, 5 = attractive/easy) in relation with the Type of Graph (bar chart/line graph/infographic/pie chart/scatterplot).

#### 4.3.4 Entertainment

Entertainment value was measured on three items for Type of Graph. The mean Entertainment value for Type of Graph is displayed in Table 4.17 and in Figure 4.1, as well the Learnability value, Efficiency value, Attractiveness order and Difficulty Order. There was an effect of Type of Graph  $F(4, 252) = 21.139$ ;  $MSE = 26$ ,  $p < .001$ ,  $\eta^2 = .251$ .

A pairwise comparison showed that line graphs were found to be less entertaining than bar charts:  $p = .001$ , infographics:  $p < .001$ , and pie charts:  $p < .001$ . Also, scatterplots were found to be less entertaining than bar charts:  $p < .001$ , infographics:  $p < .001$ , and pie charts:  $p < .001$ .

Consequently, line graphs and scatterplots were the least entertaining. There were no significant differences between bar charts, infographics and pie charts. These graphs were found to be equally entertaining.

#### 4.3.5 Learnability

The Learnability for Type of Graph was measured on two items. The means of Learnability are shown in Table 4.17 and Figure 4.1. There was an effect of Type of Graph  $F(4, 252) = 82.257$ ;  $MSE = 76$ ,  $p < .001$ ,  $\eta^2 = .566$ . A pairwise comparison showed that infographics were found to be less learnable than bar charts:  $p < .001$ , line graphs:  $p < .001$ , scatterplots:  $p < .001$ , and pie charts:  $p < .001$ . Also, scatterplots were found to be less learnable than bar charts:  $p < .001$ , line graphs:  $p < .001$ , and pie charts:  $p < .001$ . In addition, pie chart were more learnable than line graphs:  $p < .050$ , and bar charts:  $p = .015$ .

Consequently, infographics and scatterplots were the least learnable, of which infographics were less learnable than scatterplots. Pie charts were the most learnable in comparison with the other graphs. Line graphs and bar charts were found to be equally learnable.

#### 4.3.6 Memorability

The Memorability for Type of Graph was measured on two items. The means of Memorability are shown in Table 4.17 and Figure 4.1. There was an effect of Type of Graph  $F(4, 252) = 53.898$ ;  $MSE = 51$ ,  $p < .001$ ,  $\eta^2 = .461$ . A pairwise comparison showed that scatterplots were found to be less memorable than bar charts:  $p < .001$ , line graphs:  $p < .001$ , and pie charts:  $p < .001$ . Also, infographics were found to be less memorable than bar charts:  $p < .001$ , line graphs:  $p < .001$ , and pie charts:  $p < .001$ . Pie charts were found to be more memorable than bar charts:  $p < .050$ , and line graphs:  $p < .015$ .

Consequently, infographics and scatterplots were the least memorable. Pie charts were the most memorable in comparison with the other graphs. Line graphs and bar chart were found to be equally memorable.



#### 4.3.7 Efficiency

The Efficiency for Type of Graph was measured on two items. The means of Efficiency per type of graph are displayed in Table 4.17 and Figure 4.1. There was an effect of Type of Graph  $F(4,252) = 105.167$ ;  $MSE = 103$ ,  $p < .001$ ,  $\eta^2 = .625$ . A pairwise comparison showed that infographics were less efficient than bar charts:  $p < .001$ , line graphs:  $p < .001$ , pie charts:  $p < .001$ , and scatterplots:  $p < .001$ . Also, scatterplots were found to be less efficient than bar charts:  $p < .001$ , line graphs:  $p < .001$ , and pie charts:  $p < .001$ . Pie charts were found to be more efficient than bar charts:  $p < .050$ , and line graphs:  $p < .015$ .

Consequently, infographics and scatterplots were the least efficient, of which infographics were less efficient than scatterplots. Pie charts were found to be the most efficient. Bar charts and line graphs were found to be equally efficient.

#### 4.3.8 Attractiveness

Attractiveness was measured on one item. The means of Attractiveness for Type of Graph are displayed in Table 4.17 and Figure 4.1. There was an effect of Type of Graph  $F(4, 252) = 7.486$ ;  $MSE = 16$ ,  $p < .001$ ,  $\eta^2 = .106$ . A pairwise comparison showed that scatterplots were found to be less attractive than bar charts:  $p < .001$ , line graphs:  $p = .005$ , infographics:  $p < .050$ , and pie charts:  $p < .001$ .

Consequently, scatterplots were found to be the least attractive compared with the other graphs. Line graphs, bar charts, pie charts and infographics were found to be equally attractive.

#### 4.3.9 Difficulty

Difficulty was measured on one item. The means of Difficulty for Type of Graph are displayed in Table 4.17 and Figure 4.1. There was an effect of Type of Graph  $F(4, 252) = 48.312$ ;  $MSE = 67$ ,  $p < .001$ ,  $\eta^2 = .434$ . A pairwise comparison showed that infographics were more difficult than bar charts:  $p < .001$ , line graphs:  $p < .001$ , and pie charts:  $p < .001$ . Also, scatterplots were more difficult than bar charts:  $p < .001$ , line graphs:  $p < .001$ , and pie charts:  $p < .001$ .

Consequently, infographics and scatterplots were the most difficult. There were no significant differences in Difficulty between line graphs, bar charts and pie charts.

## **4.4 Summary results**

### **4.4.1 Infographics**

The hypotheses 1a and 1b were rejected.

Hypothesis 1a: The viewing times of a graph are shorter with the presence of the Gestalt principle similarity than with the absence of this principle.

Hypothesis 1b: The viewing times of a graph are shorter with the presence of the Gestalt principle proximity than with the absence of this principle.

The viewing times were the same regardless of the presence or absence of Color and Proximity. However, the viewing times were longer when the infographics were difficult than when the infographics were easy.

The hypotheses 1c and 1d were rejected.

Hypothesis 1c: The viewing times of the chunk that contains the information needed to perform the task are longer with the presence of the Gestalt principle similarity than with the absence of this principle.

Hypothesis 1d: The viewing times of the chunk that contains the information needed to perform the task are longer with the presence of the Gestalt principle proximity than with the absence of this principle.

The viewing times of the target were the same regardless of the presence or absence of Color and Proximity. However, in the easy group, the viewing times on the target were longer with the presence of Proximity than with the absence of Proximity. In addition, interactions between Color and Proximity were found in the easy group. The viewing times on the target were longer with the presence of Color and Proximity than with the presence of Color and absence of Proximity. The viewing times on the target region were longer with the absence of Color and the absence of Proximity than with the presence of Color and the absence of Proximity.

Additional analyses gave the following results:

Number of Switches: There were more switches when the infographics were difficult than when the infographics were easy. The number of switches were the same regardless of the presence or absence of Color and Proximity.

Correct Answers: Overall, 83.7% of the answers were correct. There were the same total correct answers regardless of the difficulty of the infographics and regardless of the presence or absence of Color and Proximity. The viewing time and gaze duration were the same regardless of the presence or absence of Color and Proximity. However, the page duration was longer when the infographics were difficult than when the infographics were easy.

#### 4.4.2 Bar charts, line graphs and scatterplots

The hypothesis 2 was partially confirmed.

Hypothesis 2: The viewers observe the data area first and after that the axes when viewing bar charts, line graphs, and scatterplots.

For bar charts, the axis was observed first and after that the data area. For line graphs and scatterplots, the axis and the data area were observed first equally often.

Additional analysis gave the following results:

Page duration: The viewing time was the same regardless of the type of graphs.

Gaze Duration: Overall, the data area was observed longer than the axes. In addition, the data area was observed longer for scatterplots than for bar charts. For bar charts, the axes were observed longer than the data area. For line graphs and scatterplots the reverse was true: the data area was observed longer than the axes.

Gaze Duration of the first view on the graphs: Overall, during the first view on the graph, the axes were observed longer than the data area. The gaze durations during the first view on the axes was longer for bar charts than for scatterplots. In addition, regardless of the type of graph, the axes were observed longer than the data area.

Number of Switches: For line graphs, there were more switches than for bar charts and scatterplots.

Correct Answer: More incorrect answers were given after reading scatterplots than after reading line graphs and bar charts. For the page durations, the viewing time for line graphs were longer than for scatterplots when a correct answer had been given. For the gaze durations, the data area was observed longer than the axes. In addition, the data area was observed longer for scatterplots than for bar charts. Also, the axes were observed longer than the data area for bar charts, and for line graphs and scatterplots the reverse was true: the data area was observed longer than the axes if correct answers had been given.

#### 4.4.3 Survey

Hypothesis 3a was partially confirmed.

Hypothesis 3a: It is more difficult to extract information from infographics than from bar charts, line graphs, pie charts, and scatterplots.

Scatterplots and infographics were judged to be less learnable than bar charts, line graphs and pie charts, and infographics were less learnable than scatterplots. Infographics and scatterplots were judged to be less memorable than bar charts, line graphs and pie charts. Also, pie charts were found to be the most memorable. In addition, infographics and scatterplots were judged to be less efficient than bar charts, line graphs and pie charts, and infographics were less efficient than scatterplots. Pie charts were the most efficient. Infographics and scatterplots were found to be equally difficult. These type of graphs were more difficult than bar charts, line graphs and pie charts. Consequently, infographics were the least learnable and efficient, but scatterplots and infographics were found to be equally memorable and difficult.

Hypothesis 3b was rejected.

Hypothesis 3b: Infographics are more entertaining than bar charts, line graphs, pie charts, and scatterplots.

Bar charts, infographics and pie charts were found to be equally entertaining. Line graphs, bar charts, pie charts and infographics were found to be equally attractive.

## 5 Conclusion and Discussion

In this paragraph, the conclusions of the results of the eye tracking experiment and the survey in relation to the hypotheses are reported. Thereafter, the results are discussed in relation to the theoretical outline. Finally, recommendations for designing graphs and for future research are given.

### 5.1 Conclusion

#### 5.1.1 Infographics

The present research was set up to examine whether the Gestalt principles similarity and proximity influence the processing of information in graphs. It was predicted that the total viewing times of the infographics would be shorter, and the total viewing times on the subtopic that contained information for performing the task would be longer with the presence

of similarity and proximity than without the presence of similarity and proximity. This prediction was not confirmed by the experiment. The total viewing times of the graph and the viewing times of the target were the same regardless of the use of similarity and proximity. Consequently, the use of Gestalt principles does not help the viewers to extract information faster from infographics. Possible explanations for this conclusion are discussed below.

#### 5.1.2 Bar charts, line graphs and scatterplots

The present study also examined the viewing behavior on bar charts, line graphs, and scatterplots. The prediction was that the visual chunk, the data area, would be observed before the areas of the axes. The results partially confirmed this expectation. For bar charts, the axes were observed before the data area. For line graphs and scatterplots, however, the axes and data area were observed first equally often. It seems that, viewers are used to interpret bar charts in a similar way. They apply a certain viewing behavior while extracting information from bar charts in which the axes are interpreted first. This may implicate that the axes of bar charts are more clear and easier to interpret than those of line graphs and scatterplots. Possible explanations for this conclusion are discussed below.

#### 5.1.3 Survey

The survey examined which type of graph type is more entertaining or more suitable for information transfer. The prediction was that infographics would be entertaining, but not suitable to extract information. This was partially confirmed by the results. Infographics were the least learnable and the least efficient in comparison to the other types of graphs, but equally entertaining in comparison to bar charts and pie charts. Consequently, infographics should not be used for information extraction. This conclusion is discussed below.

## **5.2 Discussion**

### 5.2.1 The processing of information in infographics

As explained in Chapter 1, the Gestalt principles of Kohler (1929) and the proximity compability principle of Wickens and Carswell (1995) implicate that clustering information leads to faster processing of information by lowering the extraneous cognitive load. The experiment does not support this claim. The total viewing times and the viewing times on the subtopic that contained information for verifying the statement were the same regardless of the presence of similarity and proximity. It was concluded that the use of Gestalt principles does not help the viewers to extract information faster from infographics.

The conclusion that the Gestalt principles does not help the viewers to extract information faster from infographics can be explained by the lines around the subtopics. It seems that the viewers already experienced the subtopics as different subtopics and did not need color and distance between them.

Another explanation for the lack of differences between the viewing times is the variance in the infographics. The eye movement data showed large differences in viewing behaviors on the infographics. Some infographics had a large amount of fixations and some infographics had much fewer fixations. As a result, the infographics were split up into two groups. As seen in the data analysis section (Chapter 3), two groups of infographics were made: easy infographics and difficult infographics. The reason for this division was not the total correct answers per group but the number of fixations and the total viewing times. Four out of eight infographics had substantially more fixations and longer viewing times than the other four infographics. When a subtopic is observed, information is processed. When infographics are observed longer, the information processes take longer, so it was be concluded that these infographics were more difficult to process than the other infographics.

But why did the information process take longer for four out of the eight infographics? There are two possible explanations for the differences in total viewing times between easy and difficult infographics: The number of elements that are displayed per subtopic (see, for example Figure 5.1), and the visual organization of chunks.

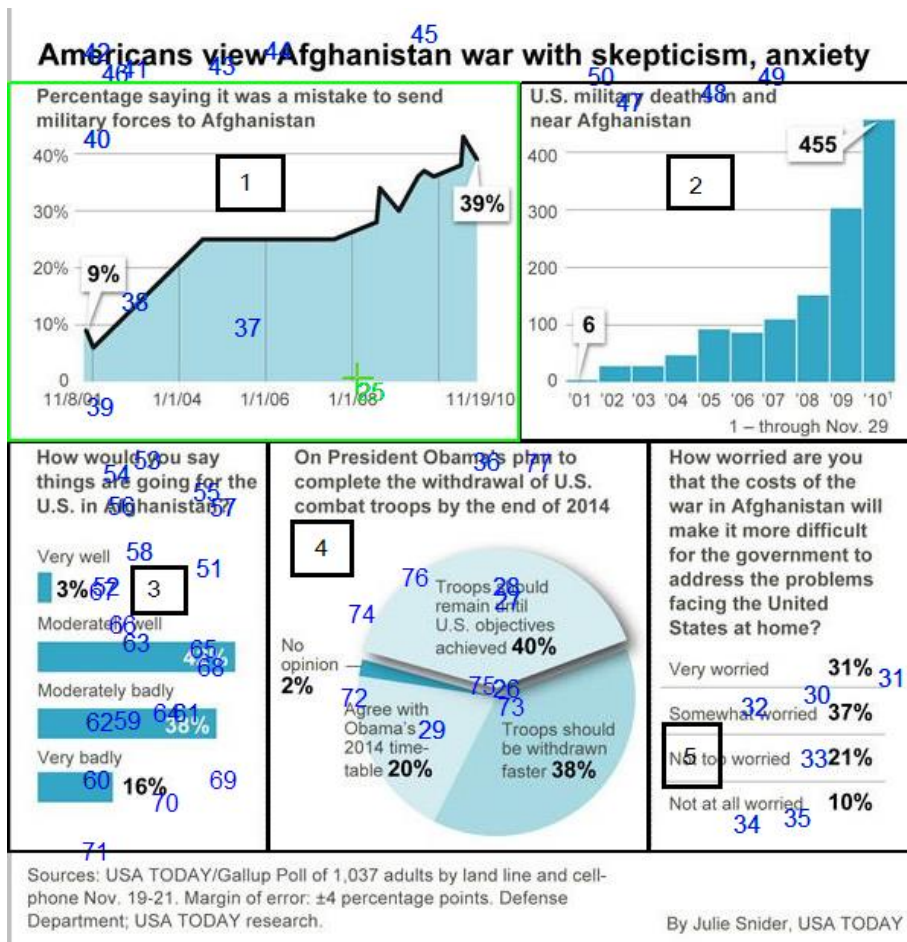


Figure 5.1: Infographic War: The numbers [1] to [5] indicate the subtopics in the infographic. Each subtopic consists of a graph and a title.

Adapted from *Americans view Afghanistan war with skepticism, anxiety* (2010), by the USA Today. Copyright 2010 by the USA Today. Retrieved from [http://usatoday30.usatoday.com/news/washington/2010-12-01-1Awar01\\_CV\\_N.htm?csp=34](http://usatoday30.usatoday.com/news/washington/2010-12-01-1Awar01_CV_N.htm?csp=34).

In Figure 5.1, the infographic War consists of five subtopics: [1] opinions about the mission in Afghanistan, [2] statistics about the deaths in and near Afghanistan, [3] opinions about how things are going for the US in Afghanistan, [4] opinions about withdrawal combat troops, and [5] opinions about money issues. Every subtopic that is shown in the infographics consists of several elements: graphs, texts and images. In this case, the subtopics consist of a graph and a graph title.

As a comparison, Figure 5.2 contains only three subtopics. However, each subtopic consists of many more elements.

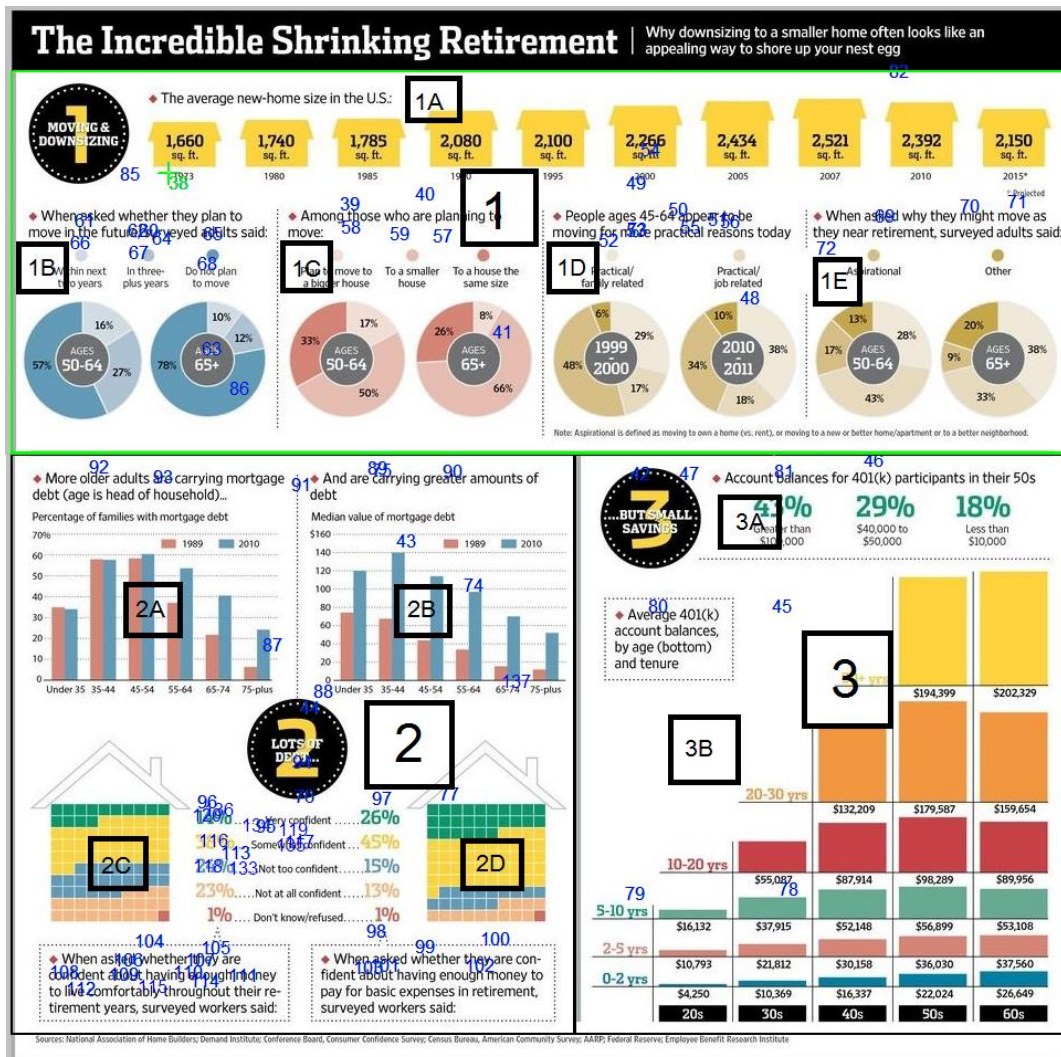


Figure 5.2: Infographic Retirement: The numbers [1] to [3] indicate the subtopics in the infographic. The subtopics consist of the topics [1A] to [3B]. Each subtopic [1-3] consists of more than one graph and title.

Adapted from *The Incredible Shrinking Retirement* (2012), by The Wall Street Journal.

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<http://online.wsj.com/news/articles/SB10001424052970204005004578080561639348492>.

The infographic Retirement consists of three subtopics: [1] moving and downsizing, [2] debt, and [3] savings. Subtopic [1] includes five elements: [1A-1E]. [1A] consists of images with a title, and [1B] to [1E] consist each of two pie charts with titles. All these elements contain information about the subtopic moving and downsizing. Subtopic [2] consists of the two bar charts [2A] and [2B] with titles, and two images that show data about the confidence in retirement [2C] and [2D] and corresponding texts about the content of the two images. Subtopic [3] has one line of percentages and corresponding title [3A], and one graph [3B] with title.



In summary, in Figure 5.1 each subtopic consisted of one graph with a corresponding title, and in Figure 5.2 each subtopic consisted of several graphs and images with corresponding titles. Consequently, the total elements that were shown per subtopic for difficult infographics, such as Retirement, were higher than for the easier infographics, such as War. So, the total amount of information that needed to be processed may have caused the longer total viewing times of the difficult infographics.

The possible reasons for the differences in the viewing times between easy and difficult infographics, i.e., the number of elements per subtopic and visual organization of the graph, are substantiated by the viewing behavior of participants. Figure 5.3 shows the viewing behavior of a participant.

First of all, the participant viewed the statement (fixations 1-25). Thereafter, the participant switched to subtopic [1], and viewed [1B], [1C], and [1A] consecutively (fixations 26-36). Subtopic [1] did not contain information that enabled the verification of the statement, so the participant switched to subtopic [2]. First, the title of graph [2A] was viewed and then the graph title of [2B] (fixations 37-58). Instead of viewing the whole subtopic, the participant switched to subtopic [3]. The participant viewed [3A] and [3B], not consecutively but by switching back and forth from [3A] and [3B] (fixations 59-82). Thereafter, the participant switched back to subtopic [2] and viewed [2D], and switched several times to the statement and back to [2D] (fixations 83-138). While processing this last element, the statement could be verified.

In summary, the participant did not view a whole subtopic before switching to another subtopic. It seems that the participant viewed some elements in the subtopic and then decided that it did not contain information to verify the statement, so the participant switched to another subtopic, instead of inspecting the remaining elements of the subtopics. It seems that the participant assumed that the elements in the subtopic contained the same or similar information.

Consequently, it seems that the viewers decided whether the statement could be verified in a subtopic by reading only a few elements in that subtopic, suggesting that they assumed that the other elements in the subtopic contain the same or similar information. From this it may be concluded that clustering information may have a negative effect on verifying statements if infographics have many elements per subtopic. If an element in a subtopic does not contain information that can be used to verify the statement, a switch is made to another subtopic. Because easy infographics have fewer elements per subtopic, the viewers see



the participant, first [2A] was observed and after that [2B]. Because viewers are used to read from left to right, [3A] was observed after [2B]. So instead of inspecting subtopic [2] completely, the viewers switched to subtopic [3]. Consequently, the viewers needed to switch back to a subtopic in difficult infographics to observe the other elements which contained different information related to the subtopic, so the viewing times were longer for difficult infographics than for easy infographics.

### 5.2.2 The processing of information in the target region in infographics

Because of the higher information level per subtopic in difficult infographics, one would expect the duration on the subtopic that contained information for verifying the statement to be longer for the difficult infographics than for the easy infographics. Unfortunately, this was not the case. When the viewers observed the subtopic that contained information for verification, the viewing times were the same regardless of the difficulty of the infographic. However, there was a small effect for the easy infographics. The viewing times on these subtopics were longer when distance was used between the subtopics than when no distance was used between them. In addition, the viewing times were longer when colors were used to distinct subtopics and distance was used between them than when colors were used to distinct subtopics and no distance was used between them. Also, viewing times on the subtopic that contained information for verifying the statement were longer when no color was used to distinct subtopics and no distance was used between them than when color was used and no distance was used.

As explained in Chapter 1, the cognitive load theory of Sweller (1994) may give an explanation for the absence of differences in viewing times on the subtopic that contained information for the verification between easy and difficult infographics. For the difficult infographics, there were more elements in the target, but the graphs did not interact, so the viewers did not have to integrate the different graphs with each other to verify the statement. So, the intrinsic cognitive load for the difficult infographics was not higher than for the easy infographics. It is possible that the complexity of the information was the same regardless of the difficulty and, therefore, the viewing times of the target did not differ.

### 5.2.3 The number of switches in infographics

There were more switches between the different subtopics and the statement while viewing difficult infographics than easy infographics. This could not be attributed to a difference in the number of subtopics in the infographics. The total number of subtopics for easy infographics was 18 and for difficulty infographics 14. As seen in Figures 5.1 and 5.2, the

number of elements per subtopic were higher for difficult infographics than for easy infographics. This led to viewing behavior as seen in Figure 5.3. It may seem that the elements were perceived as one topic, even though every graph had its own contribution. When one element, for example a graph title, was viewed and the element contained information that was not useful to verifying the statement, the subtopic was not viewed further even though the other elements of the subtopic may have helped, and a switch was made to another subtopic (see Figure 5.3). Consequently, the viewers switched back more often to the subtopic of which one element was viewed. For the easy infographics, most of the time the subtopic consisted of one graph and a title (see Figure 5.1). If the element was not useful to verify the statement, a switch was made to another subtopic. The viewers did not have to switch back to a subtopic that they already viewed, because the elements were observed at once.

#### 5.2.4 No distinctive task for easy and difficult infographics

There were as many correct answers for easy infographics as for difficult infographics, and for infographics with the presence of color or proximity as for infographics without the presence of color or proximity. This implicates that the task was not distinctive enough. A suggestion for another, elaborate, task is asking for an explanation of the infographics. A prediction is that the visual organization is described clearly and is named for infographics with the presence of color or proximity than for infographics without the presence of color or proximity, so the participants name the different chunks for infographics with the presence of color or proximity. In addition, a prediction is that longer and more explanations are given for difficult infographics than for easy infographics, because, as explained above, the difficult infographics contain more information elements.

#### 5.2.5 The axes in bar charts

As explained in Chapter 1, Pinker (1990) and Carpenter and Shah (1998) distinguish three steps to comprehend graphs: Code the visual chunks, identify their quantitative properties, and relate them to the referents of the graph. It was predicted that the data area, the visual chunk, would be observed before the areas of the axes. This prediction was partially confirmed by the experiment.

For bar charts, the axes were observed before the data area. For line graphs and scatterplots, however, the axes and the data area were observed first equally often. It seems that viewers are used to interpret bar charts in a similar way. Also, the axes were observed longer during the first view on the graph for bar charts than for scatterplots. There are two possible explanations for the viewing behavior on the axes for bar charts: The nature of the graphs, and the information which is displayed on the axes.

The research of Zack and Tversky (1999) may explain why for bar charts the axes were observed first. As described in Chapter 1, they claim that information in bar charts needs to be described in terms of comparisons between discrete individual data points. Terms that need to be used are: higher, lower, bigger than, smaller than. Information in line graphs need to be described as trends between continuous data points. The terms that need to be used are: increase and decrease. The results of the gaze duration confirmed the different purposes of the graphs: the axes were observed longer than the data area for bar charts (discrete comparison) and the data area was observed longer than the axes for line graphs and scatterplots (describing trends).

In addition, because of the purpose of the graph types, the axes of the graphs are interpreted in different ways. There is a difference in the ease of the interpretation of the axes for the several graph types. For bar charts it is clear how many data points are displayed. One bar displays one data point. However, for line graphs it is not clear how many data points are displayed. For line graphs the number of data points is infinite between the beginning and the end of the lines. Therefore, the axes are not so easy to interpret. Scatterplots describe a relation between the axes. The data points need to be interpreted as cluttered around a line, so viewers may have the same viewing strategy as with line graphs. Consequently, because of the nature of the bar charts, it is more useful for viewers to view and interpret the axes first. This leads to a viewing behavior in which the axes are observed first.

The second explanation for the viewing behavior on the axes for bar charts may be the amount of information that is displayed on their axes. The viewing times during the first view were longer on the axes for bar charts than for scatterplots. The information on the axes of the bar charts may be more complex than for scatterplots. The axes of the scatterplots contained single numbers, for example a reading score. The information on the axes of bar charts contained more complex information, for example the number 2010-4, which means the fourth quarter of 2010. The viewers needed to decode this number in order to interpret the graph.

### 5.2.6 The amount of information in line graphs

For line graphs, there were more switches between the statements, the data area, the axes and the legend than for bar charts and scatterplots. A possible explanation for this result is that the line graphs contained more information than the other two graphs, so more switches had to be made. In addition, in order to verify the statement for line graphs the viewers needed to compare multiple lines, observe the two axes, and read the legend to know what the lines represented. Consequently, the viewers needed to observe more objects in order to verify the statement, and more switches needed to be made.

The viewing times for line graphs were longer than for scatterplots when a correct answer had been given. The cognitive load theory of Sweller (1994), as explained in Chapter 1, could give a possible explanation. The elements that were displayed in the line graphs were multiple lines, two axes and a legend. All these elements needed to be interpreted in order to verify the statement, so the intrinsic load was high. The lines needed to be interpreted and after that the values needed to be compared. Also, for viewing line graphs, the viewers needed to observe the different areas more often. For scatterplots, only one axis needed to be interpreted and the number of dots needed to be compared, and therefore the intrinsic load may have been lower. Perhaps the absence of a difference with bar charts was caused by the fact that for bar charts both the axes needed to be interpreted in order to verify the statement.

### 5.2.7 The difficulty of the scatterplot

The most incorrect answers were given after reading scatterplots. It seemed that viewers found scatterplots more difficult to process than line graphs and bar charts (this is confirmed by the survey). This may be due to the fact that the participants had little experience by interpreting scatterplots. This was noticed by the participants themselves, too. They reported that they had difficulties to interpret scatterplots.

The finding that the data area was observed longer for scatterplots than for bar charts may be explained by the task. For the bar charts two individual data points needed to be compared. For scatterplots the viewers needed to count the dots of the scatterplot and compare them. For example, in Figure 5.4, the statement “most of the students take 45-55 minutes to do an exam” (in graph “Exam time”) (see Appendix A), forced viewers to count the dots. Counting the dots per timeslot takes time. Because the viewers needed to remember more numbers, the viewing time of the objects was longer for scatterplots than for bar charts.

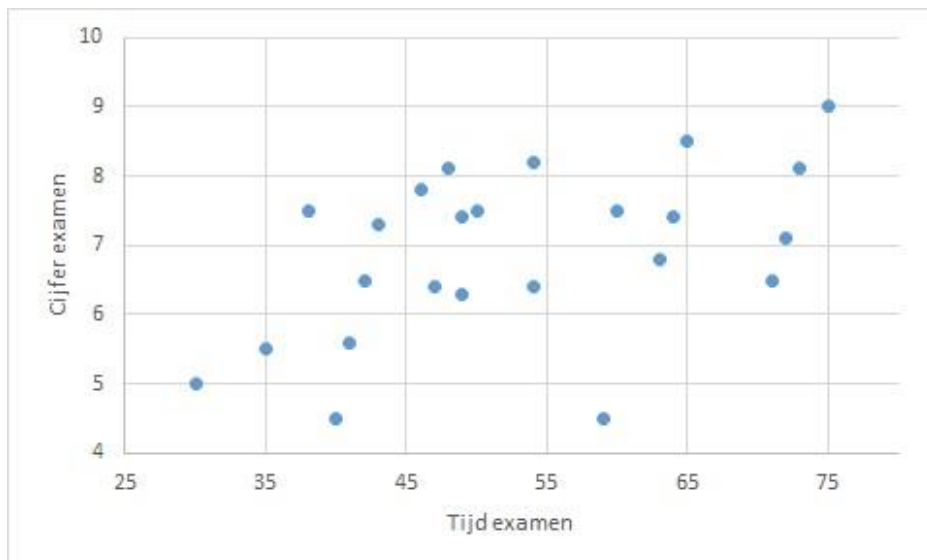


Figure 5.4: Scatterplot Exam time: The participants needed to count the dots per timeslot in order to verify the statement.

### 5.2.8 The absence of the high entertainment value for infographics

As explained in Chapter 1, infographics use images, numbers, graphics, and text to visualize the journalistic story (see, e.g., Segel & Heer, 2010; Siricharoen, 2013). No infographic is the same. Because of the variation in infographics, it was expected that they were more entertaining than the other standard graphs such as bar charts and line graphs. The experiment did not support this expectation. Infographics were equally entertaining in comparison to bar charts and pie charts. A possible explanation for this result is the difficulty of the task.

According to the results, bar charts, infographics and, pie charts were found to be equally entertaining. Also, line graphs, bar charts, pie charts, and infographics were found to be equally attractive. It is striking that there was almost no distinction between the types of graphs for entertainment and attractiveness. During the introduction of the experiment, the viewers faced difficulties in interpreting scatterplots, and they verified the statement more often incorrectly than the other type of graphs. So, this may be an explanation that scatterplots were judged as not entertaining. A possible explanation for the result that infographics, bar charts and pie charts were judged equally entertaining, may be explained by the experimental task, verifying the statements.

### 5.2.9 The usability value of the graph types

As explained in Chapter 1, the interface of the standard graphs, such as bar charts, is learned. The viewers can extract information from them by using the same viewing behavior. However, no infographic is the same. For every infographic the viewers need to construct a

new strategy to extract information visually. It was predicted that it is more difficult to extract information from infographic than from bar charts, line graphs, pie charts, and scatterplots. This prediction was partially confirmed by the experiment. Infographics were judged to be the least learnable and the least efficient of the graphs. It was concluded that infographics should not be used for extracting information from.

A possible explanation why infographics were the least learnable is the standardization of the other graphs, as explained in Chapter 1. Bar charts, line graphs, scatterplots, and pie charts are very common graphs. This means that people know what to expect when they see bar charts, line graphs, scatterplots, and pie charts. Bar charts always consist of two axes and bars, line graphs consist of two axes, one or more lines and sometimes a legend, scatterplots consist of two axes and dots, and pie charts consist of several parts and a legend. Infographics are different: no infographic is the same. For every infographic, the viewers need to develop a strategy to extract information visually. Analyzing the graph takes time, as can be seen from the eye movements. The eye movements show the inefficiency of infographics.

Viewers judged infographics and scatterplots to be less memorable than the other types of graphs. The amount of information was high in the infographics used in this experiment, so it was difficult to memorize this information. Because of the differences between the infographics, there was no clear viewing behavior. Also, infographics were judged as the most difficult graphs. The results of the total viewing times confirmed that. The amount of information of infographics led to difficulty in processing.

In addition, the results of the scatterplots are explained. Scatterplots and infographics were found to be the least memorable and difficult. An explanation for the low memorability may be the difficulty. Self-report in the experiment and the results of the survey showed that viewers faced difficulties while viewing scatterplots. So, the viewers cannot describe their viewing strategy, and memorize the content, easily.

Pie charts were judged as the most memorable and the most efficient graphs. A possible explanation for this result may be that pie charts always have the same appearance. So, the reader can apply the same strategy to each pie chart. Consequently, the viewers do not have to make unnecessary eye movements and the viewing times are shorter.



## 5.3 Recommendations and future research

### 5.3.1 Infographics

Difficult infographics led to longer viewing times and more switches than easy infographics (see explanation above). Consequently, when information needs to be extracted fast, the number of elements that are displayed in the subtopics needs to be minimized. For future research the infographics need to be checked for difficulty, and the lines around the subtopics need to be deleted. When this is the case, future research can focus on whether clustering information may have an effect on viewing times. The prediction is that the total viewing times are shorter with the presence of similarity and proximity than without the presence of similarity and proximity. Also, future research can focus on the influence of the visual organization of infographics on the processing of information. As described above, a possible explanation for the different viewing times between infographics could be the visual organization. Viewers are used to read from left to right. To investigate whether viewers view from left to right, the elements in the subtopics can be placed on a horizontal line in one condition, as seen in Figure 5.5, and on a vertical line in the other condition, as seen in Figure 5.6. The prediction is that the total viewing times are shorter when the elements in the subtopic are placed horizontally than when these elements are placed vertically.

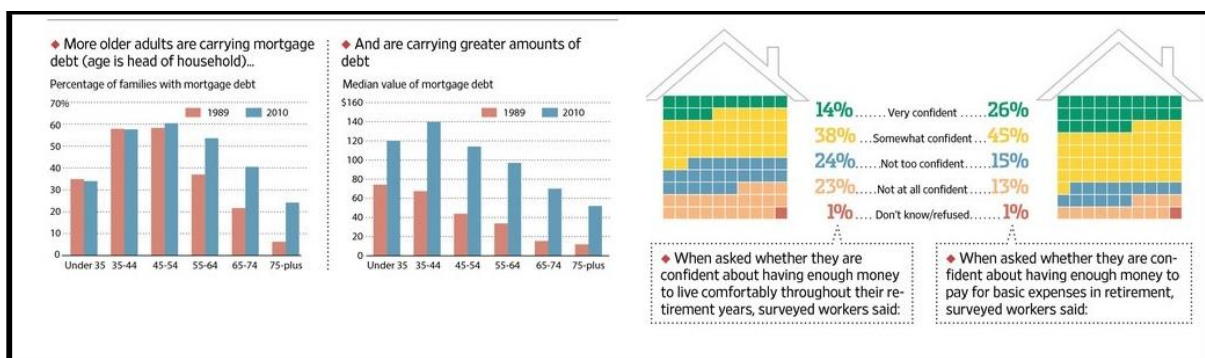


Figure 5.5: The elements in one subtopic are placed on a horizontal line.

Adapted from *The Incredible Shrinking Retirement* (2012), by The Wall Street Journal.

Copyright 2012 by The Wall Street Journal. Retrieved from

<http://online.wsj.com/news/articles/SB10001424052970204005004578080561639348492>.

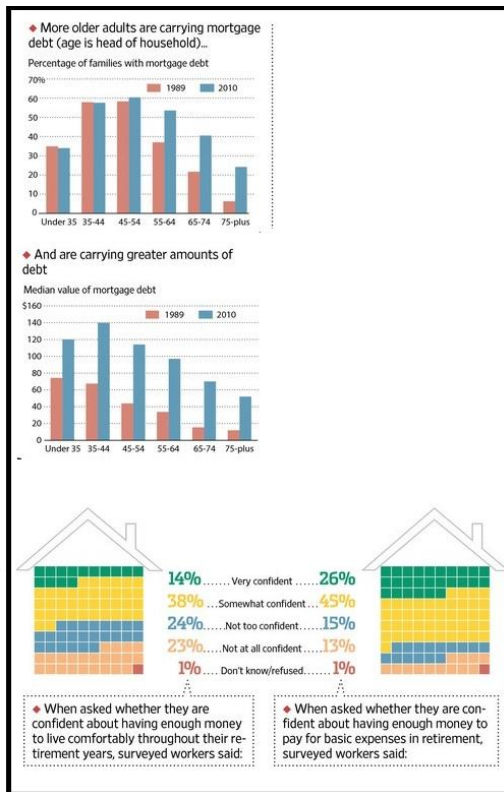


Figure 5.6: The elements in one subtopic are placed on a vertical line.

Adapted from *The Incredible Shrinking Retirement* (2012), by The Wall Street Journal.

Copyright 2012 by The Wall Street Journal. Retrieved from

<http://online.wsj.com/news/articles/SB10001424052970204005004578080561639348492>.

In addition, infographics were not the most efficient, memorable, and learnable graphs in comparison to the other types of graphs. Infographics should not be used for information extraction. Viewers faced large difficulties when extracting information from infographics. However, it was predicted that infographics would be the most entertaining graph in comparison with the other types of graphs. Nevertheless, infographics were not the most entertaining graphs. As mentioned, above, it seems that the task could be a cause for the lower predicted entertainment value. Future research could ask about the entertainment value before information need to be extracted and could ask the entertainment value after extracting information. A suggestion for research is that the participants first judge an infographic on entertainment, then verify statements, and after that judge the entertainment value again. One would expect that the infographics are judged as more entertaining before verifying the statement than after the verification.

### 5.3.2 Bar charts, line graphs, scatterplots, and pie charts

During the first view, viewers observed the axes longer than the data area, whereby the axes were observed longer for bar charts than scatterplots. In addition, the axes were observed first for bar charts. Consequently, the information on the axes needs to be clear as possible, especially for bar charts. Future research could focus on why the axes were observed first for bar charts and not for line graphs and scatterplots. As explained above, an explanation may be the purpose of the graph which describing and comparing concrete data points. The data points are represented directly by the height of the bars. A suggestion for research is to complicate the detection of the individual data points by deleting the spacing between the bars, so the graph consists of one part as in Figure 5.7, showing that the axes are observed first because of the data points in the graph, and the more clear axes. A prediction is that the participants will spend less time on the axes and more time on the data area without spacing than with spacing between the bars. Also, the first view on the axes will be lower for bar charts without spacing than with spacing.

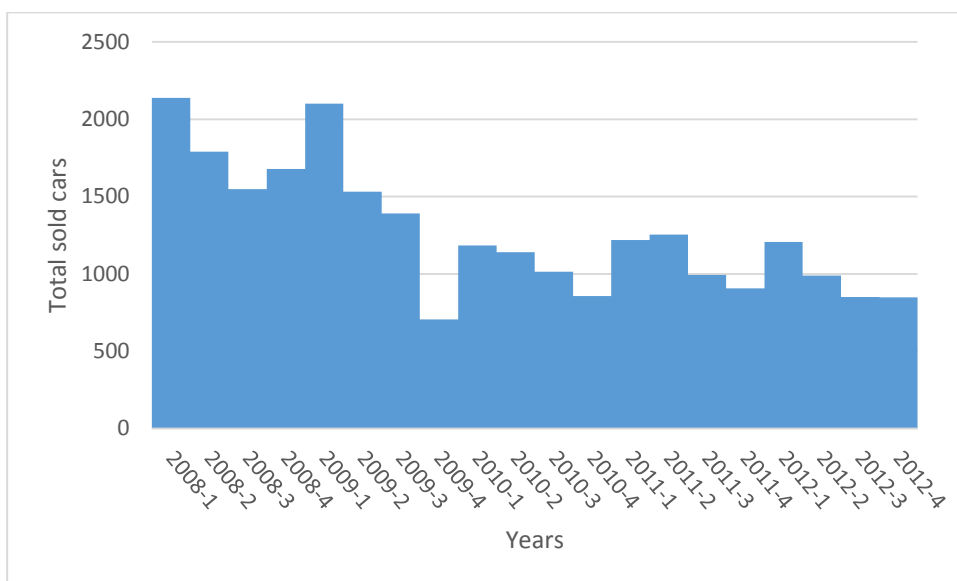


Figure 5.7: By deleting the spacing between the bars, the data points may be more difficult to verify.

In addition, future research could focus on the specifics of the viewing behavior. It seems that viewers may have the same viewing strategy for viewing bar charts. They start viewing the axes. However, for line graphs and scatterplots it seems that viewers do not apply the same behavior for every graph. By conducting a research with a think aloud method, the viewing behavior may become clear.

The number of switches was higher for line graphs than for bar charts and scatterplots. This may be due to the legend of the line graph which needs to be checked as well. The number of switches may be lower by labelling the lines in the graph. The total information is the same, but the extraneous cognitive load is lower. A prediction is that the total viewing times, and the number of switches will be lower for line graphs with labelling, than for line graphs with a legend.

Every graph has its own purpose. Scatterplots show a relation between the x and y axis by which the scatter of this relation is displayed. Scientists use scatterplots, because they have the knowledge to interpret them. However, it seems that viewers faced difficulties while interpreting them and judged scatterplots as less memorable than the other types of graphs. A scatterplot needs to be used for scientific audiences. Future research could focus on the question why scatterplots are difficult to interpret. A suggestion for research is to ask several types of questions to verify which type of question causes difficulties. In addition, the think aloud method could be used to investigate the processing of graphs. It is predicted that the participants face the least difficulties by verifying statements about trends between the two axes (for example the relation between x and y is positive).

Pie charts were the most memorable and efficient type of graphs. If the data can be visualized into pie charts, it is recommend to do so, because the viewers extract information easily and the charts are found to be entertaining as well by the viewers.

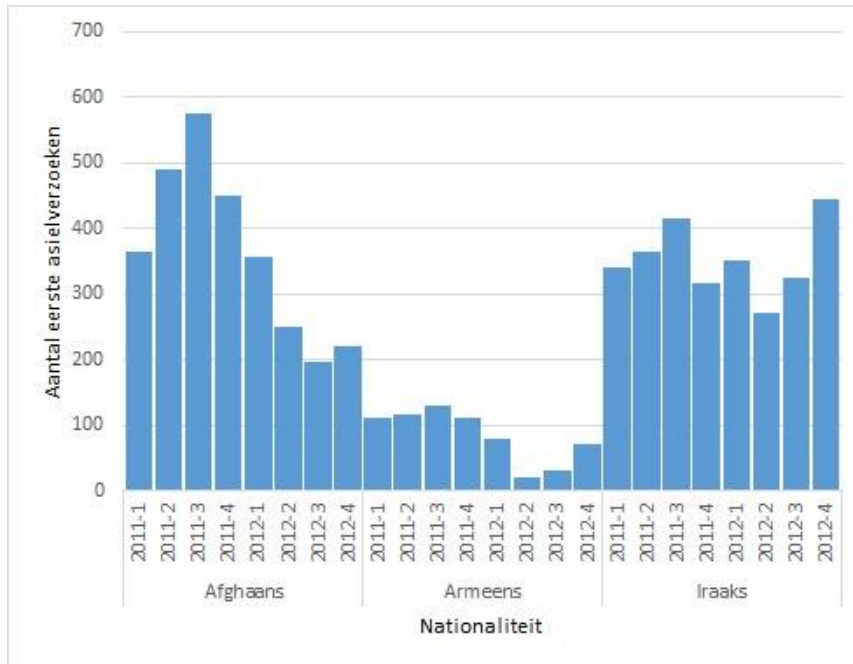
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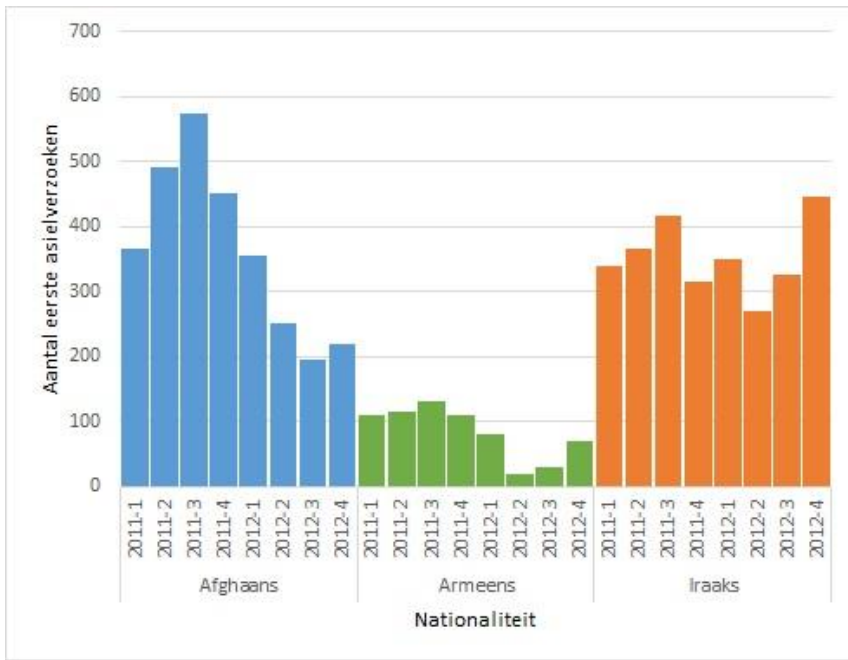
## Appendix A material research

Appendix A consists of the material used in the material research. In addition, the following graphs were used in the experiment: fuel, heartbeat, internet, priceperkm, smartphones, holiday for line graphs, movie genre, music, area, SP, cake, transport for pie charts, exam time, age men women, reading score, pricecar, salary, time study for scatterplots, immigrants, CO2, SP, courses, cars, unemployment for bar charts, boat, students, high school, Pinterest, MacDonalds-Starbucks, cats-dogs for infographics. Chapter 2,3, and 5 refer to this appendix



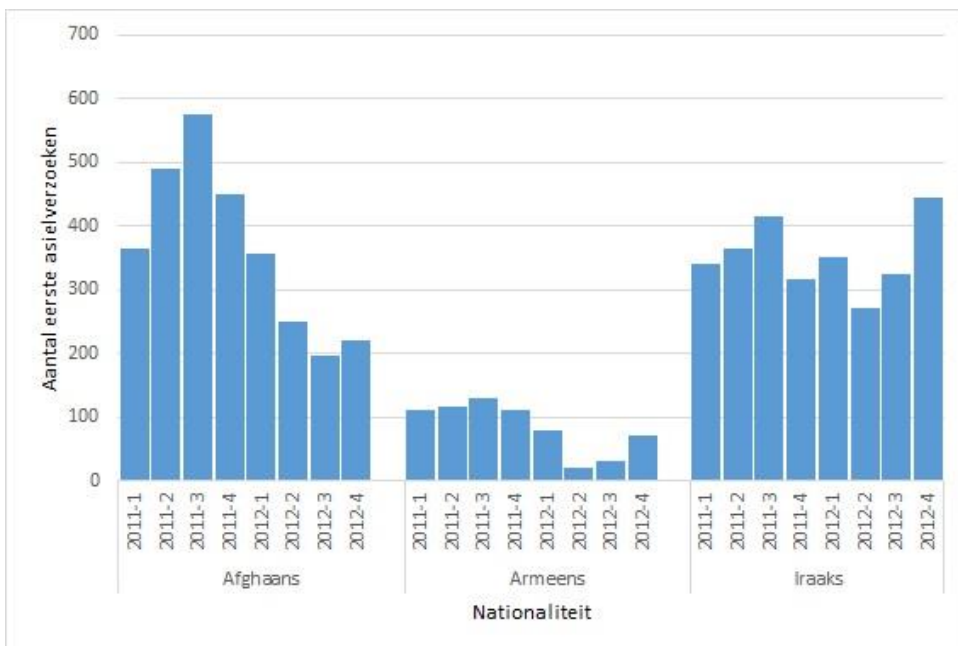
Immigrants (x axis: nationality: Afghans, Armenian, Iraqi, y axis: number of first asylum applications): no color and no proximity

Statement: The number of first asylum applications is higher for Afghans than for Iraqis in the second quarter of 2011 (true).



Immigrants (x axis: nationality: Afghans, Armenian, Iraqi, y axis: number of first asylum applications): color and no proximity

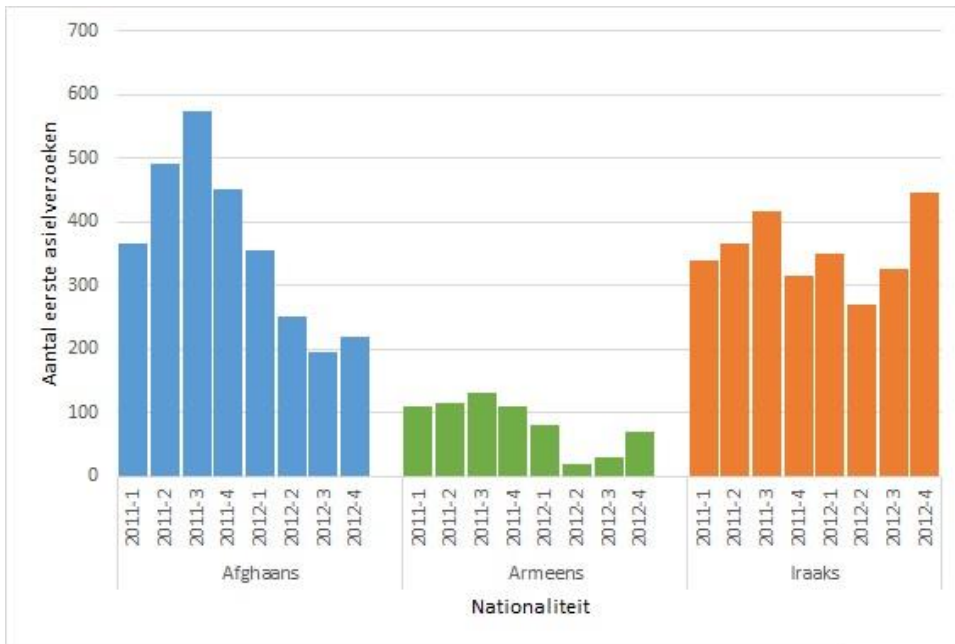
Statement: The number of first asylum applications is higher for Afghans than for Iraqis in the second quarter of 2011 (true).



Immigrants (x axis: nationality: Afghans, Armenian, Iraqi, y axis: number of first asylum applications): proximity and no color

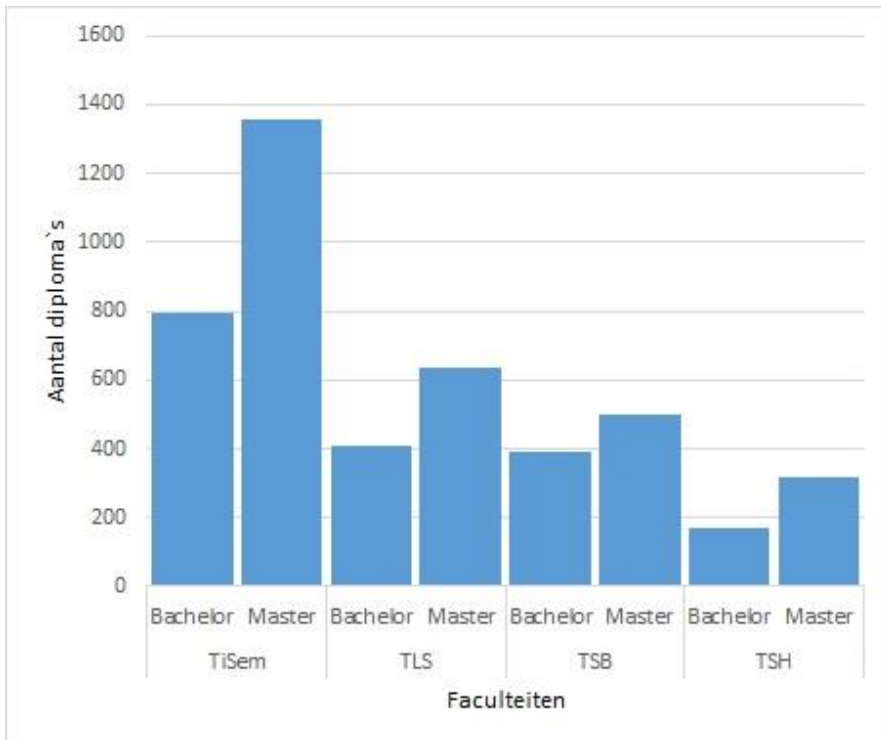
Statement: The number of first asylum applications is higher for Afghans than for Iraqis in the second quarter of 2011 (true).





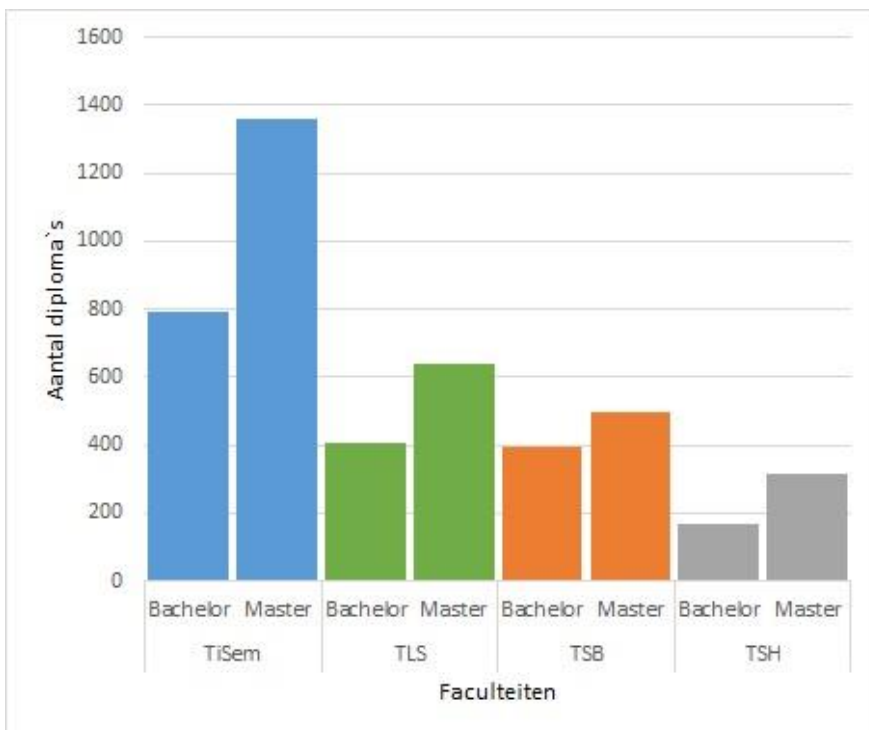
Immigrants (x axis: nationality: Afghans, Armenian, Iraqi, y axis: number of first asylum applications): color and proximity

Statement: The number of first asylum applications is higher for Afghans than for Iraqis in the second quarter of 2011 (true).



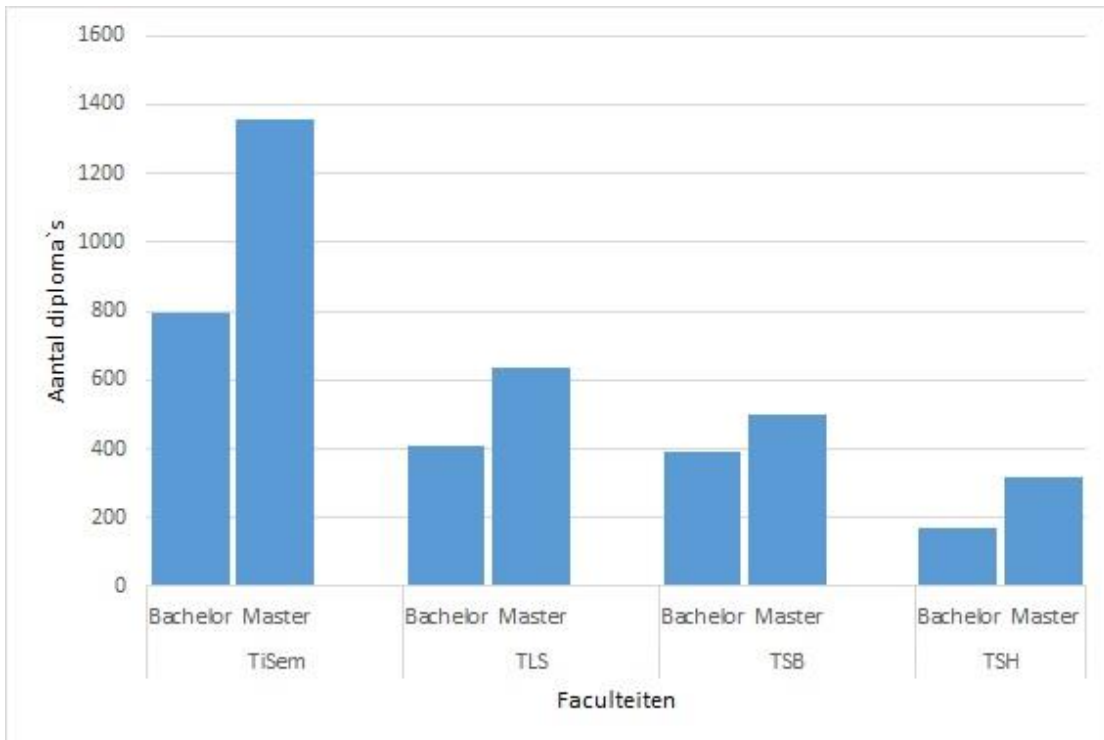
Diploma (x axis: Schools, y axis: number of diplomas): no color and no proximity

Statement: The difference in the number of diplomas for bachelor and master is the highest for TiSem (true).



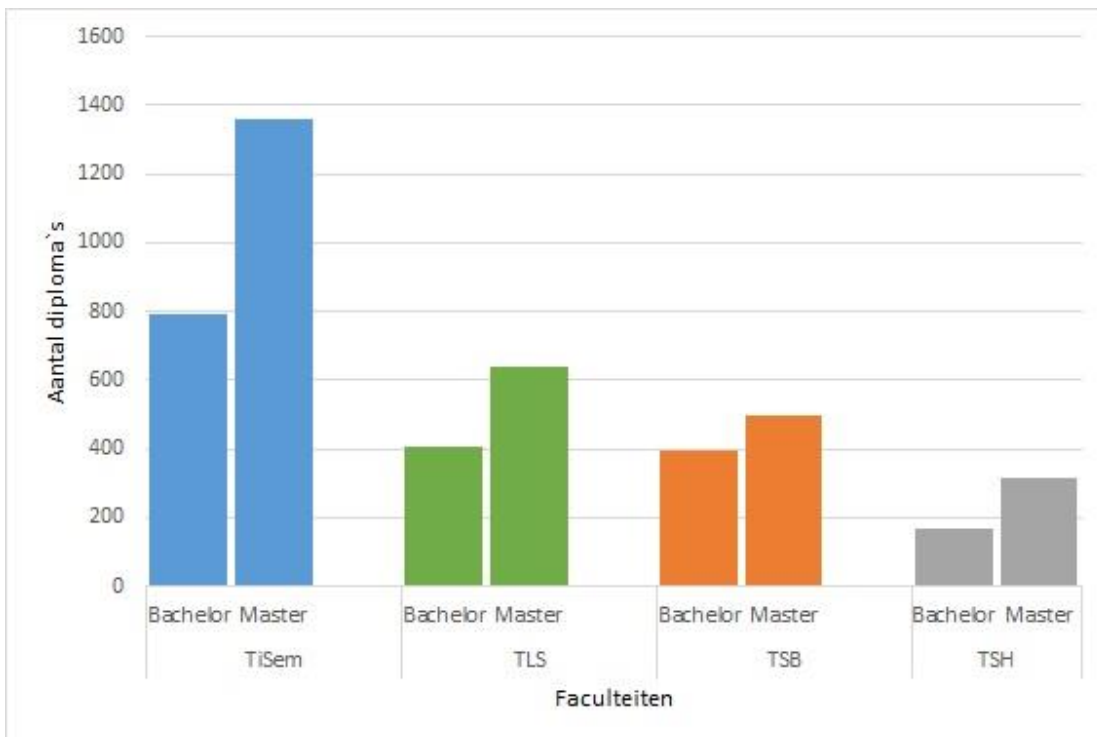
Diploma (x axis: Schools, y axis: number of diplomas): color and no proximity

Statement: The difference in the number of diplomas for bachelor and master is the highest for TiSem (true).



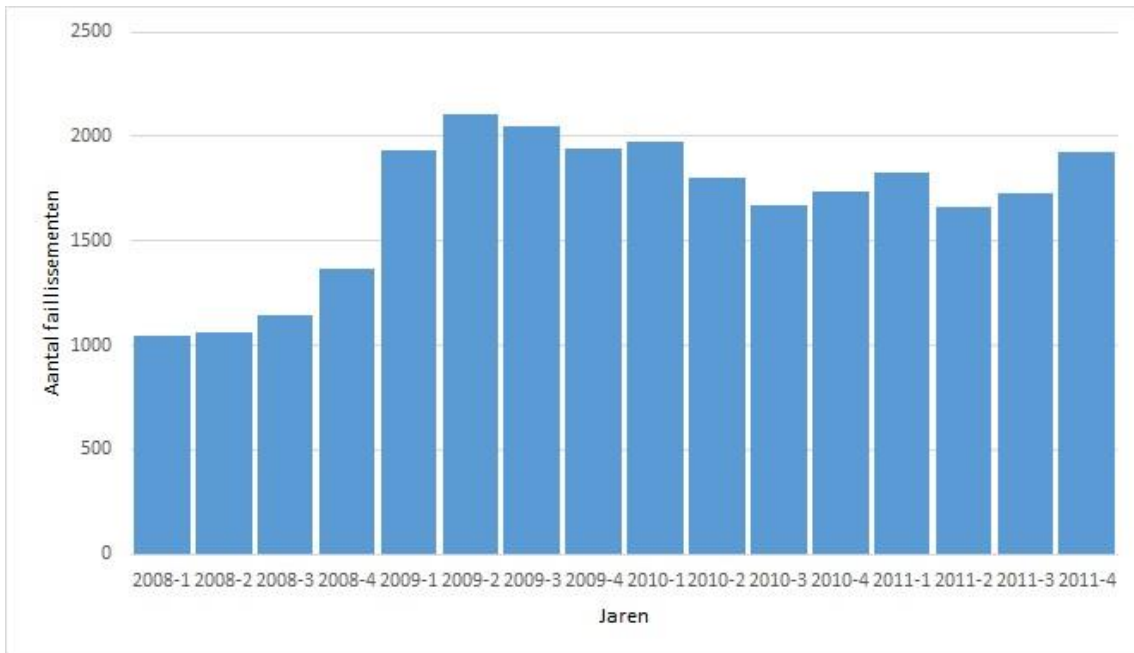
Diploma (x axis: Schools, y axis: number of diplomas): no color and proximity

Statement: The difference in the number of diplomas for bachelor and master is the highest for TiSem (true).

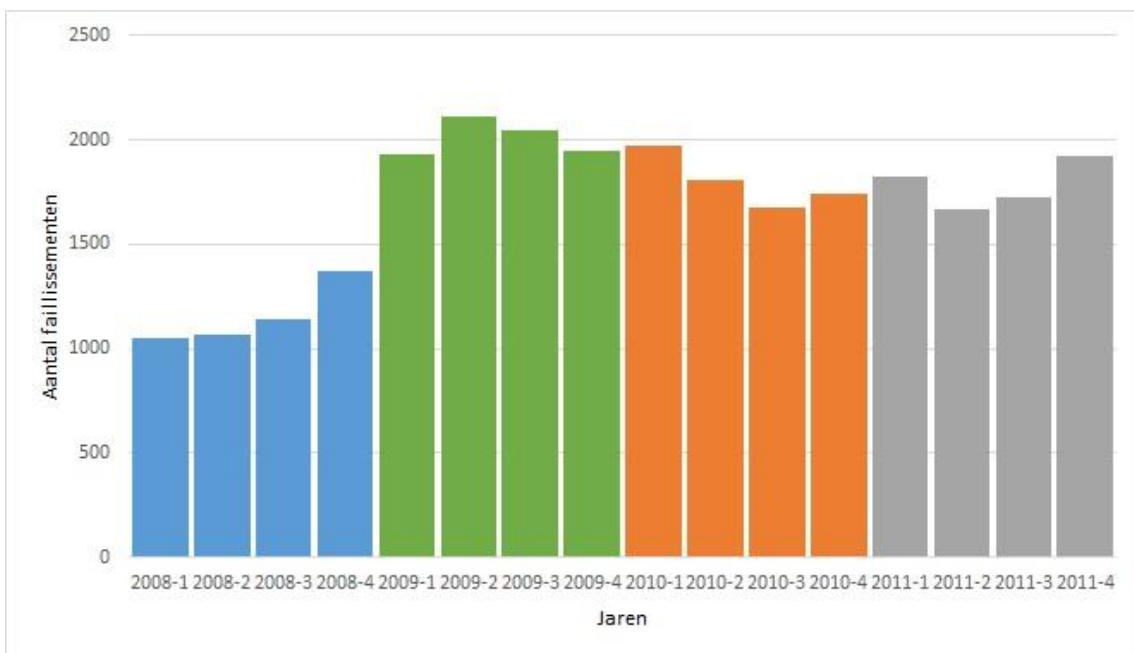


Diploma (x axis: Schools, y axis: number of diplomas): color and proximity

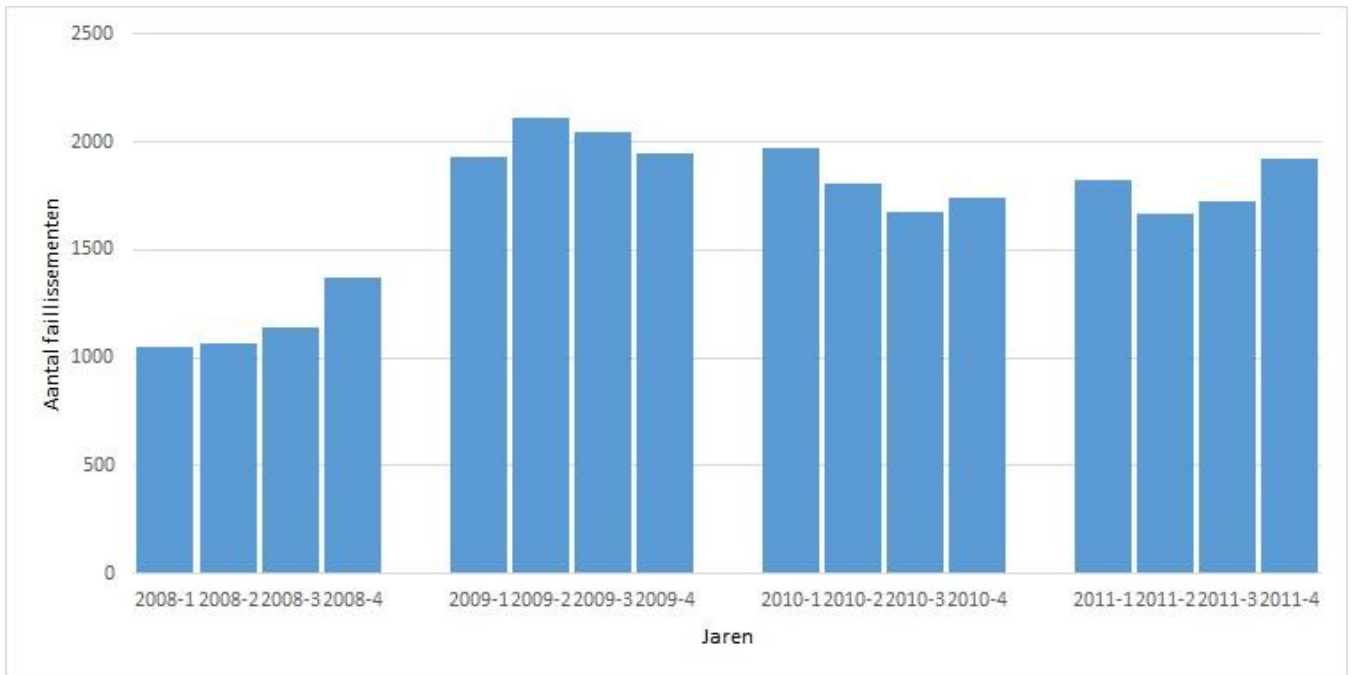
Statement: The difference in the number of diplomas for bachelor and master is the highest for TiSem (true).



Bankruptcy (x axis: years, y axis: number of bankruptcies): no color and no proximity  
 Statement: The number of bankruptcies is higher in 2009-1 than in 2008-4 (true).

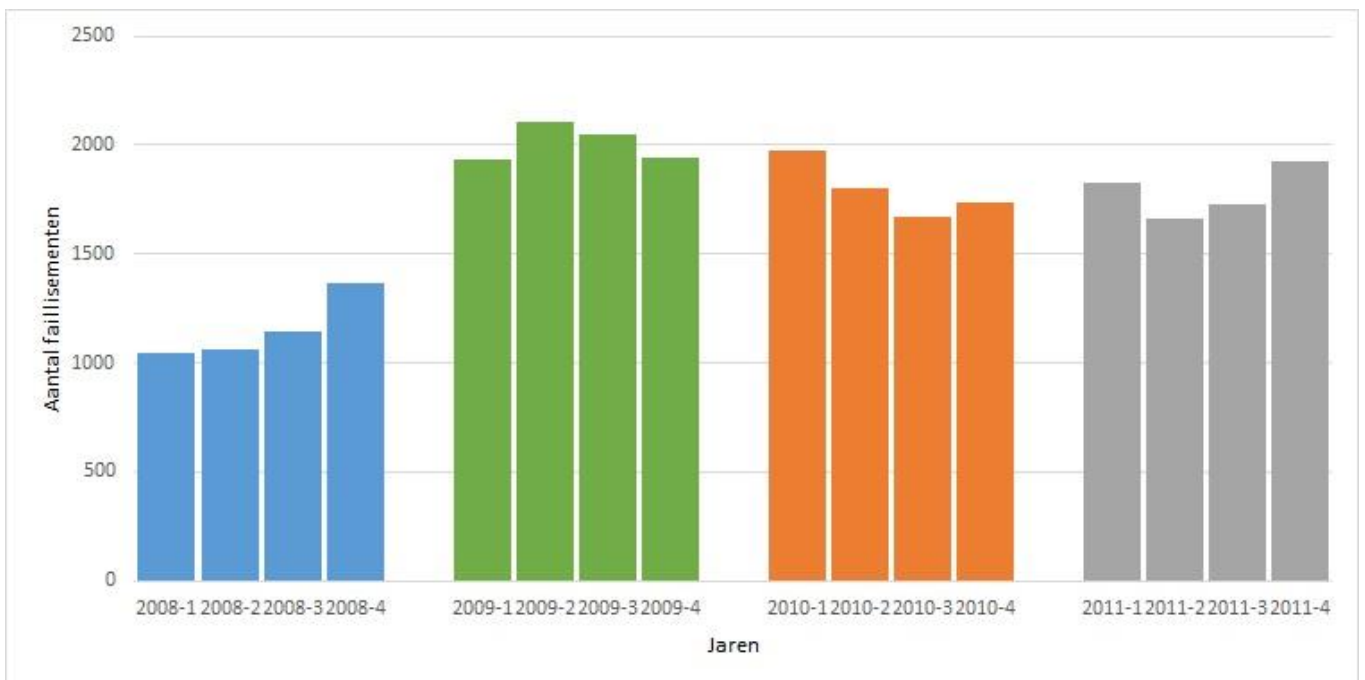


Bankruptcy (x axis: years, y axis: number of bankruptcies): color and no proximity  
 Statement: The number of bankruptcies is higher in 2009-1 than in 2008-4 (true).



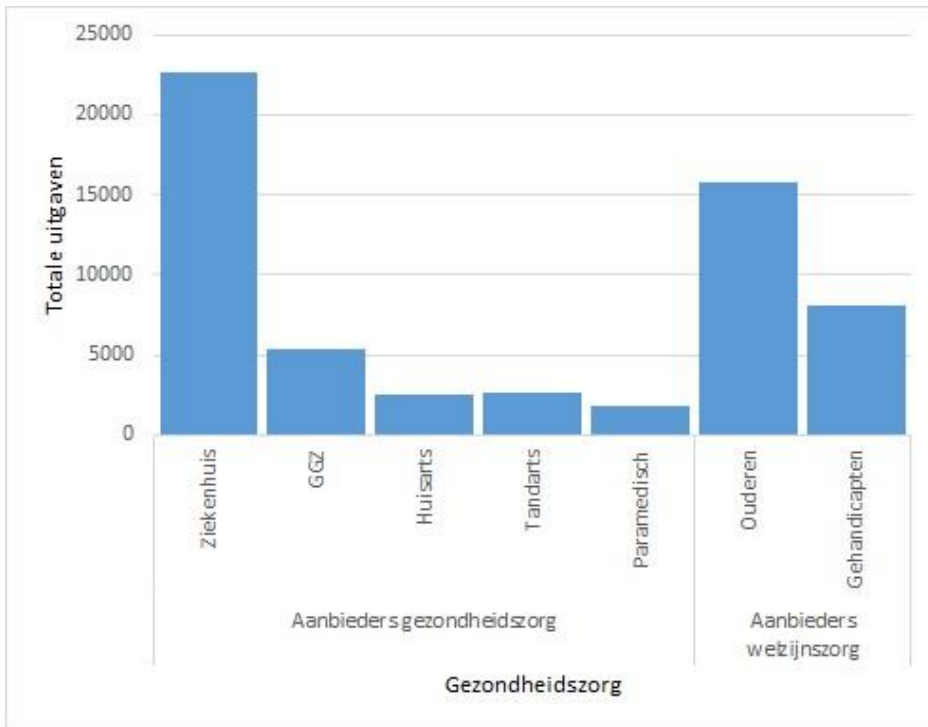
Bankruptcy (x axis: years, y axis: number of bankruptcies): no color and proximity

Statement: The number of bankruptcies is higher in 2009-1 than in 2008-4 (true).



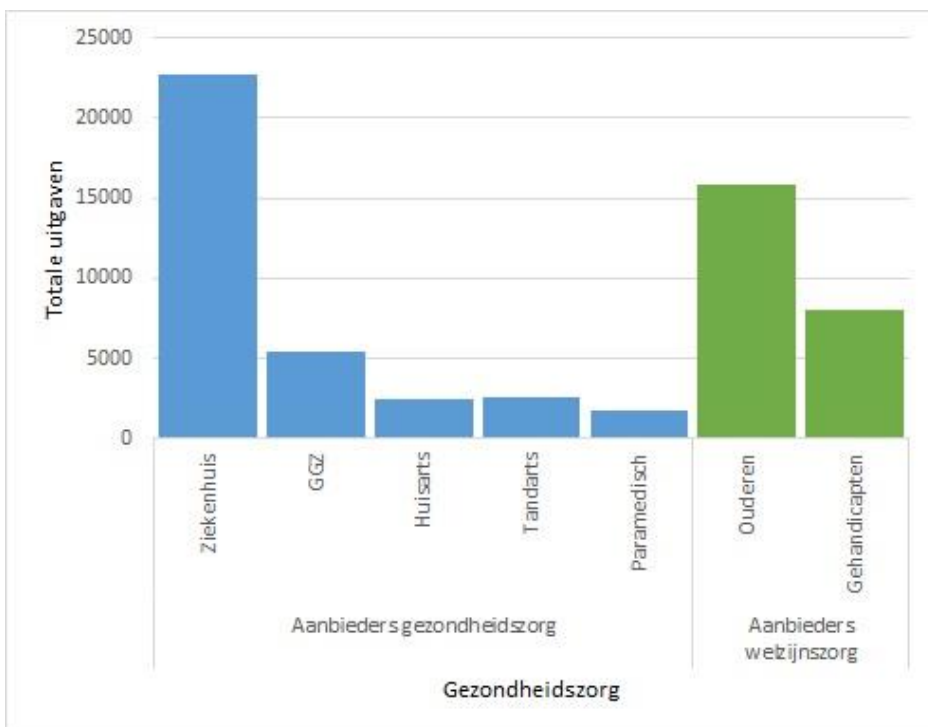
Bankruptcy (x axis: years, y axis: number of bankruptcies): color and proximity

Statement: The number of bankruptcies is higher in 2009-1 than in 2008-4 (true).



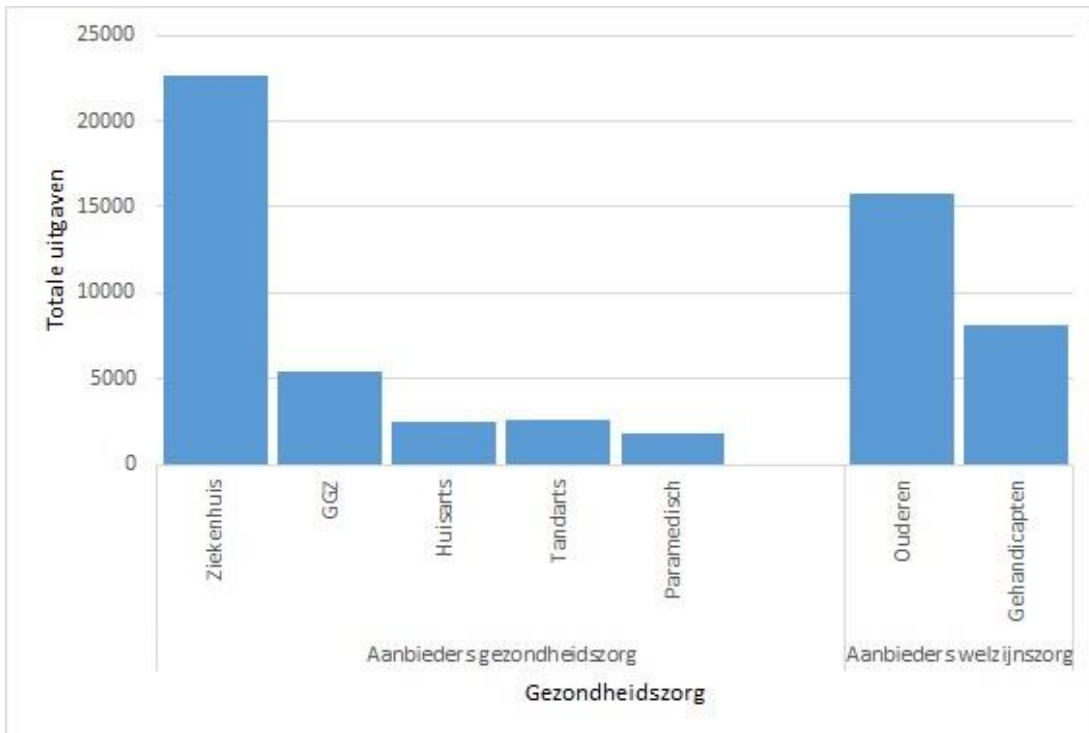
Health (x axis: healthcare: hospital, GGZ, general practitioner, dentist, paramedic and elderly, disabled, y axis: total spending): no color and no proximity

Statement: The total spending is higher in 2010 for hospitals than for the elderly (true).



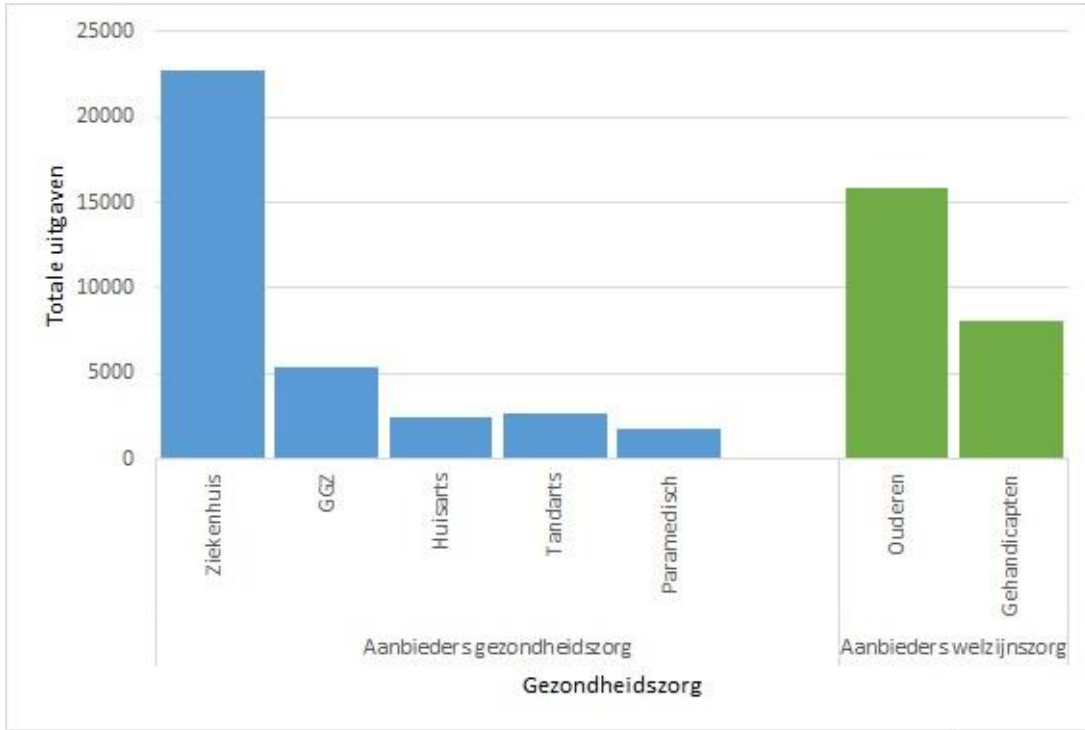
Health (x axis: healthcare: hospital, GGZ, general practitioner, dentist, paramedic and elderly, disabled, y axis: total spending): color and no proximity

Statement: The total spending is higher in 2010 for hospitals than for the elderly (true).



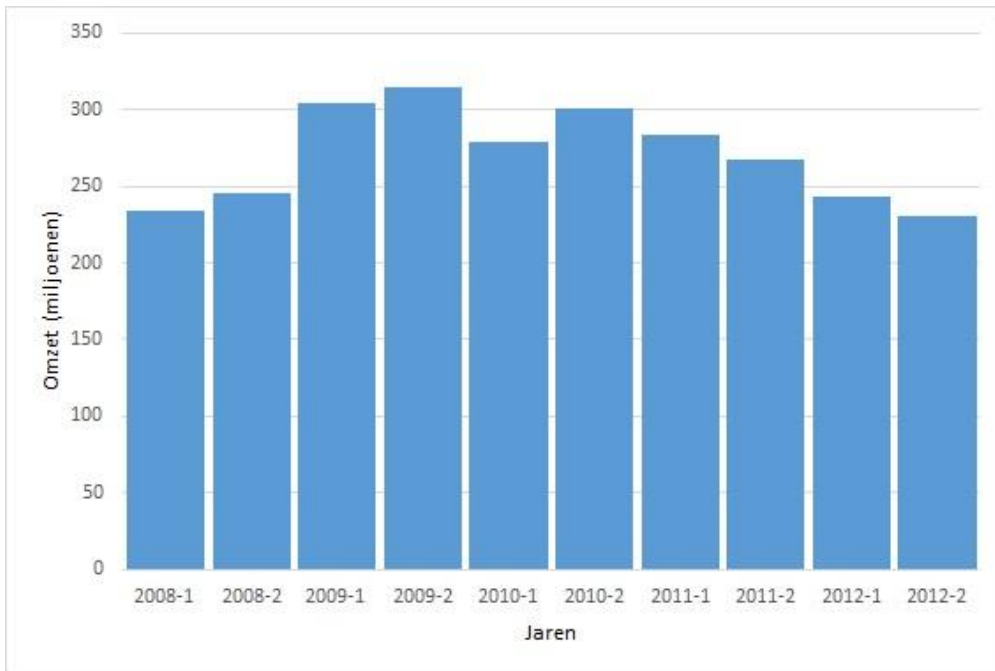
Health (x axis: healthcare: hospital, GGZ, general practitioner, dentist, paramedic and elderly, disabled, y axis: total spending): no color and proximity

Statement: The total spending is higher in 2010 for hospitals than for the elderly (true).



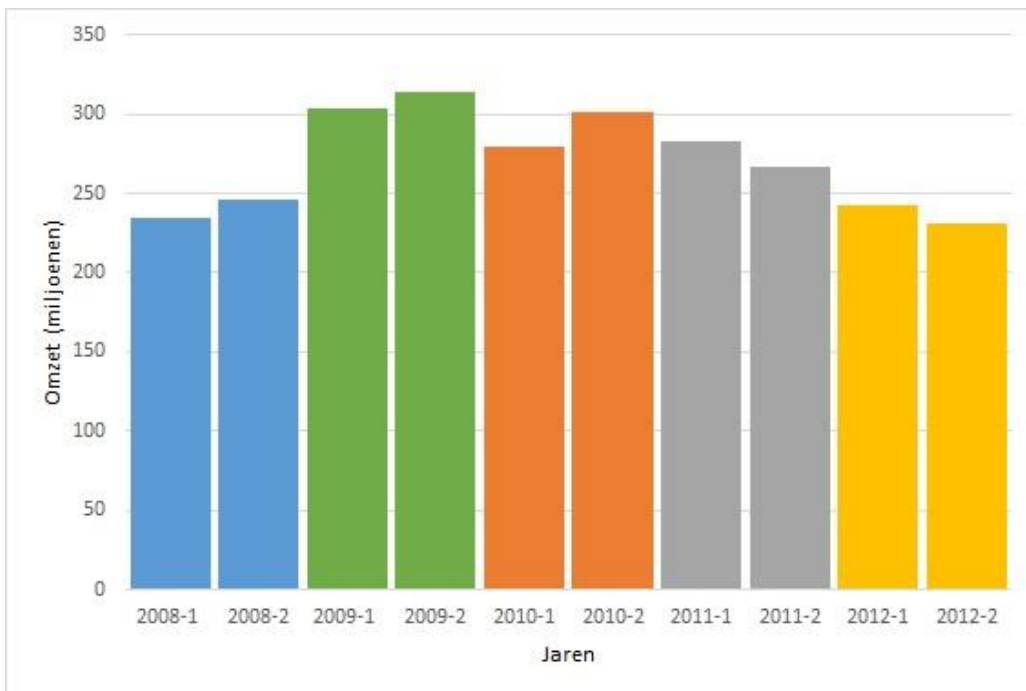
Health (x axis: healthcare: hospital, GGZ, general practitioner, dentist, paramedic and elderly, disabled, y axis: total spending): color and proximity

Statement: The total spending is higher in 2010 for hospitals than for the elderly (true).



Revenue (x axis: years, y axis: revenue (millions)): no color and no proximity

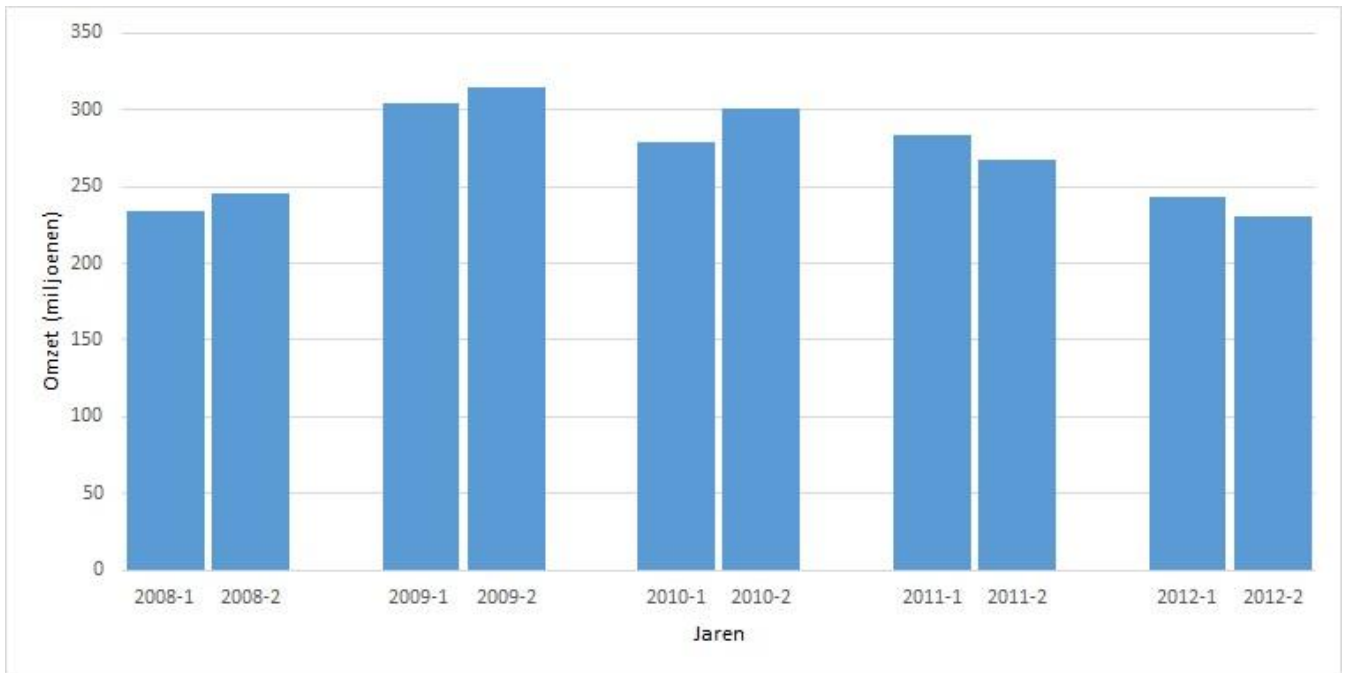
Statement: The revenue in 2010 in the second semester is higher than in the first semester (true).



Revenue (x axis: years, y axis: revenue (millions)): color and no proximity

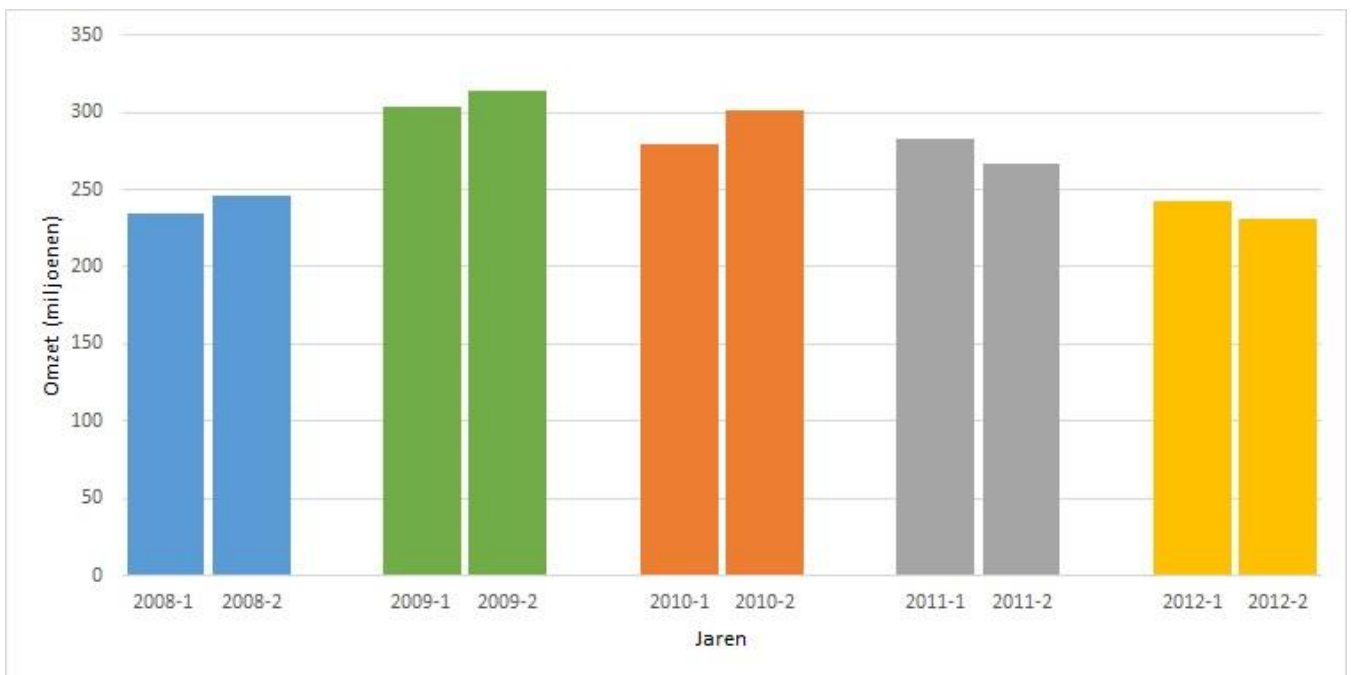
Statement: The revenue in 2010 in the second semester is higher than in the first semester (true).





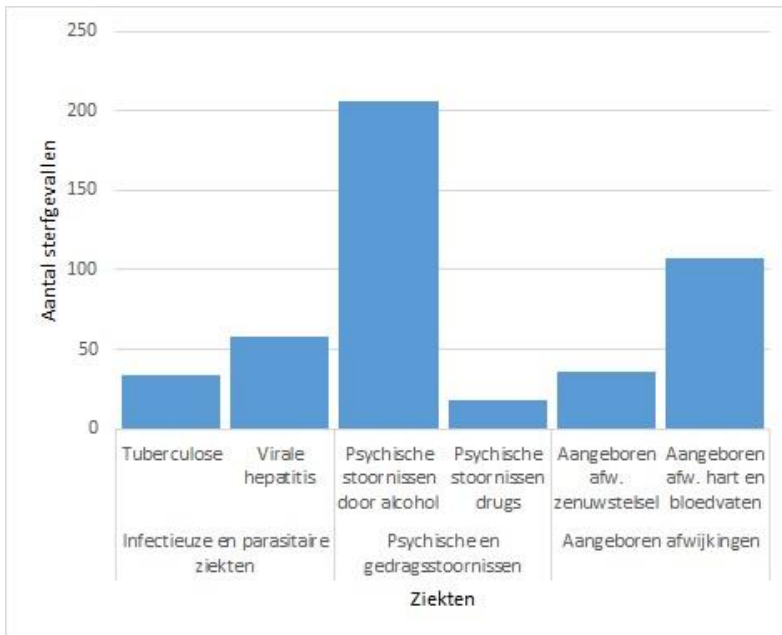
Revenue (x axis: years, y axis: revenue (millions): no color and proximity

Statement: The revenue in 2010 in the second semester is higher than in the first semester (true).



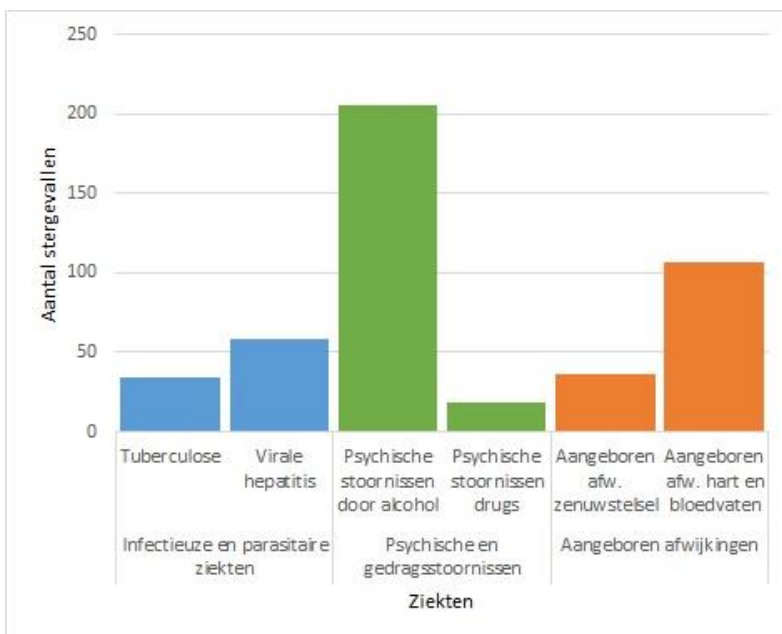
Revenue (x axis: years, y axis: revenue (millions): color and proximity

Statement: The revenue in 2010 in the second semester is higher than in the first semester (true).



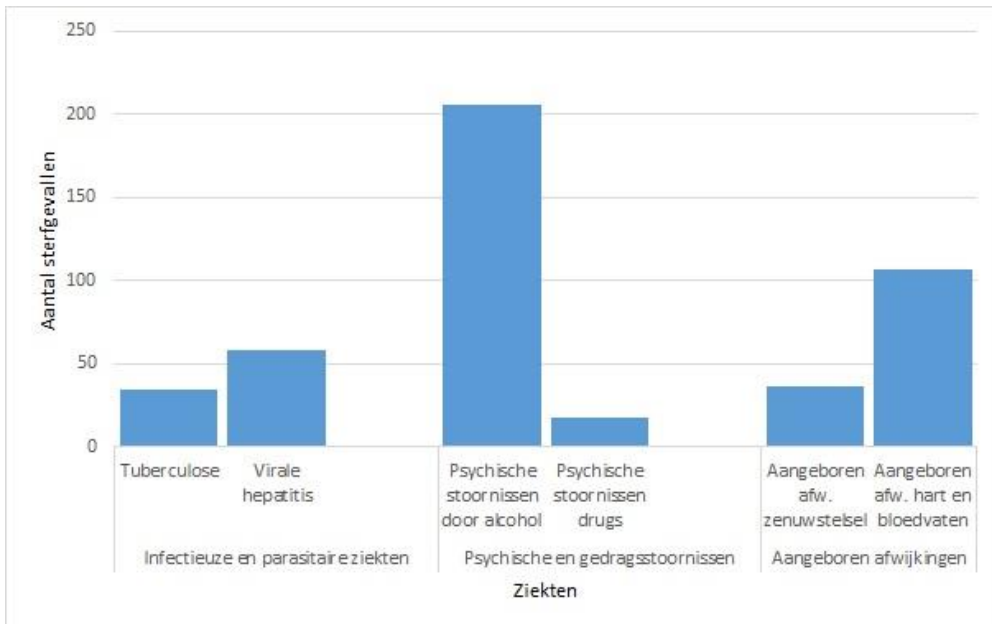
Deaths (x axis: diseases: tuberculosis, viral hepatitis, psychical disorders by alcohol, psychical disorder by drugs, congenital abnormality nervous system, and congenital abnormality heart and blood vessels, y axis: number of deaths): no color and no proximity

Statement: The number of deaths is higher for congenital abnormality heart and blood vessels than for viral hepatitis (true).



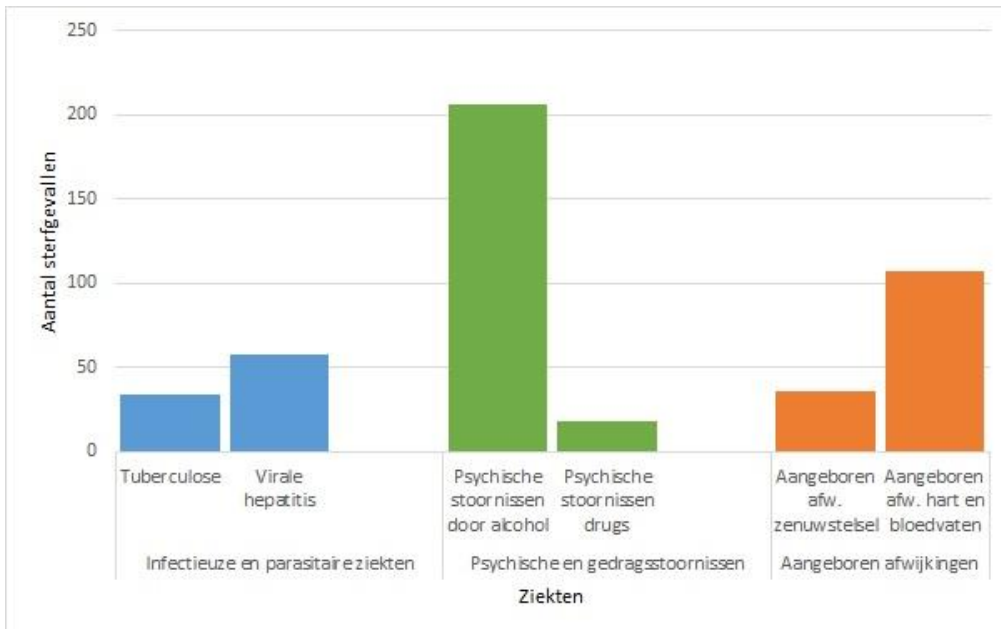
Deaths (x axis: diseases: tuberculosis, viral hepatitis, psychical disorders by alcohol, psychical disorder by drugs, congenital abnormality nervous system, and congenital abnormality heart and blood vessels, y axis: number of deaths): color and no proximity

Statement: The number of deaths is higher for congenital abnormality heart and blood vessels than for viral hepatitis (true).



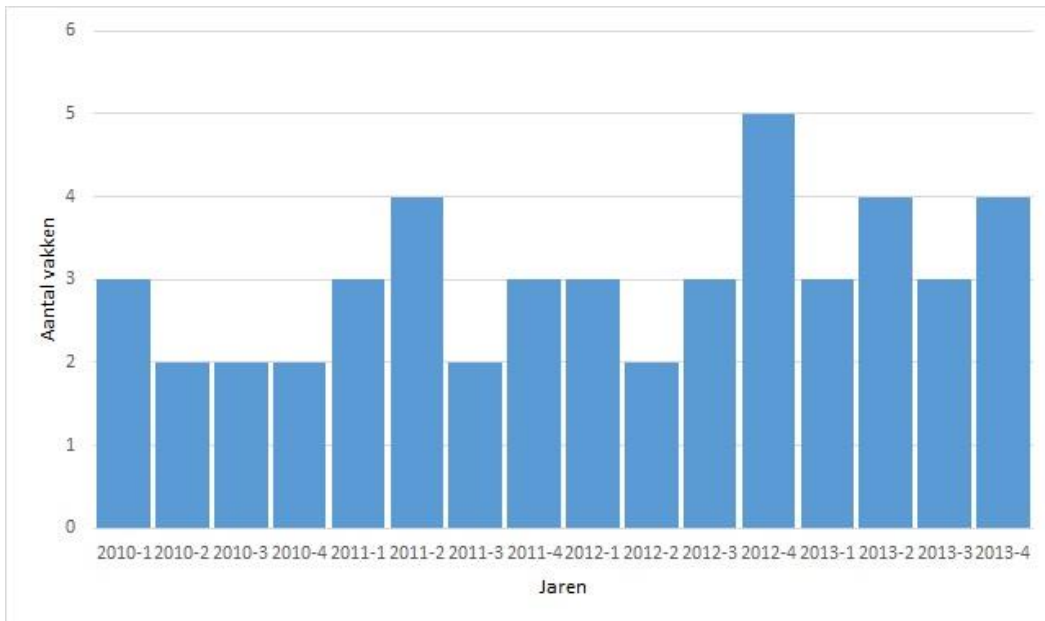
Deaths (x axis: diseases: tuberculosis, viral hepatitis, physical disorders by alcohol, physical disorder by drugs, congenital abnormality nervous system, and congenital abnormality heart and blood vessels, y axis: number of deaths): no color and proximity

Statement: The number of deaths is higher for congenital abnormality heart and blood vessels than for viral hepatitis (true).



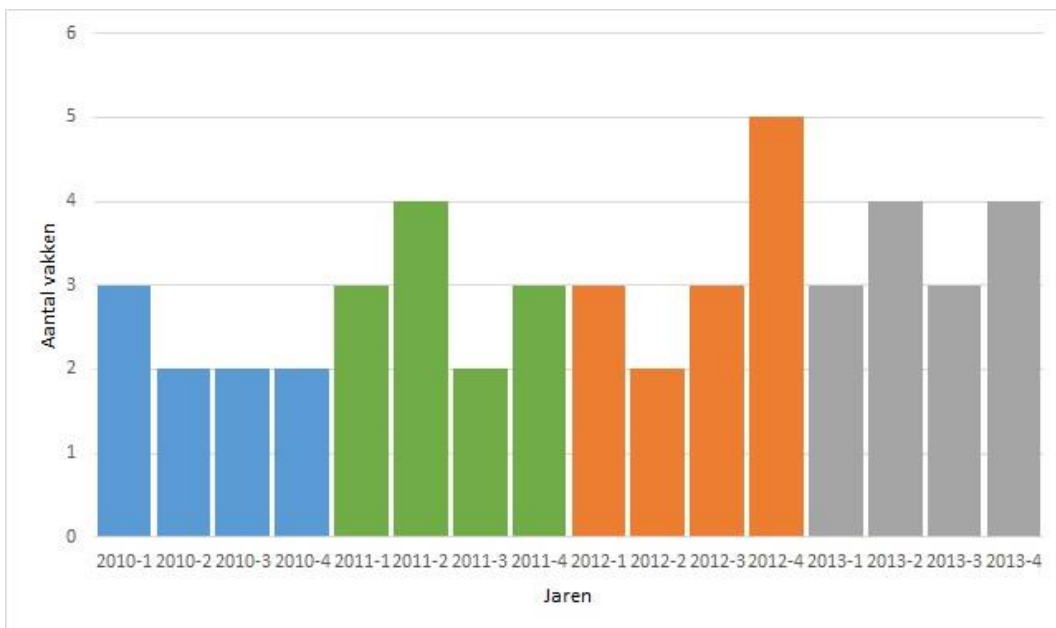
Deaths (x axis: diseases: tuberculosis, viral hepatitis, physical disorders by alcohol, physical disorder by drugs, congenital abnormality nervous system, and congenital abnormality heart and blood vessels, y axis: number of deaths): color and proximity

Statement: The number of deaths is higher for congenital abnormality heart and blood vessels than for viral hepatitis (true).



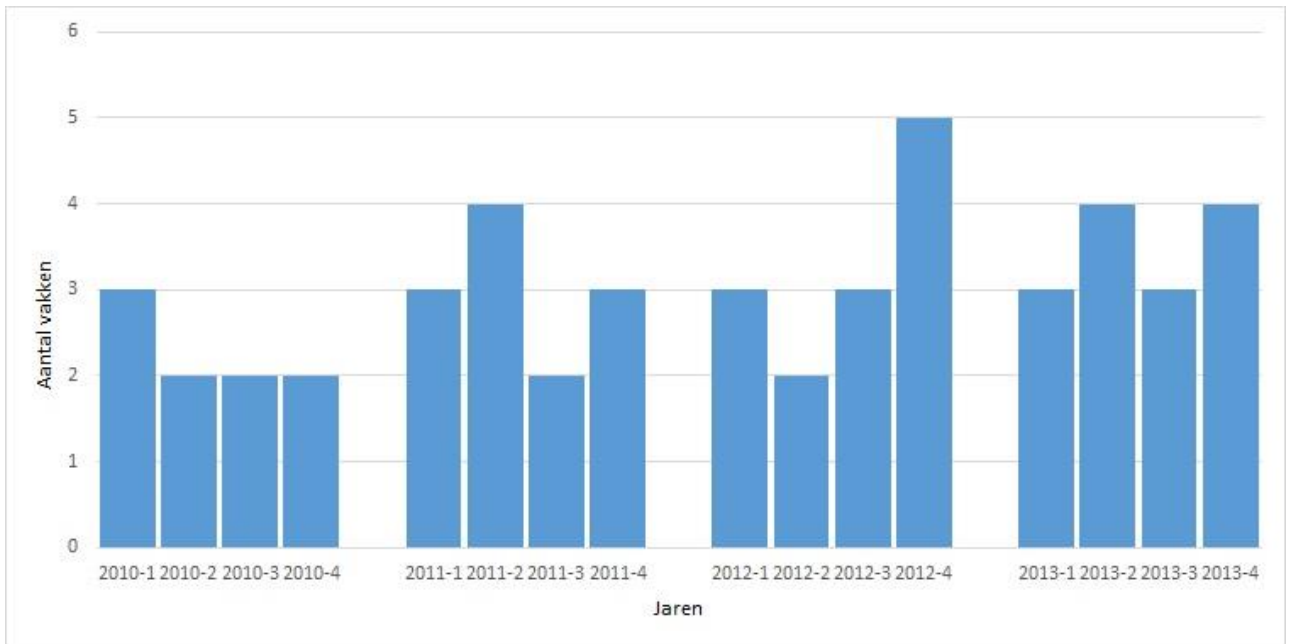
Courses (x axis: years, y axis: number of courses): no color and no proximity

Statement: Twice as much courses are provided during the second quarter of 2013 than during the second quarter of 2010 (true).



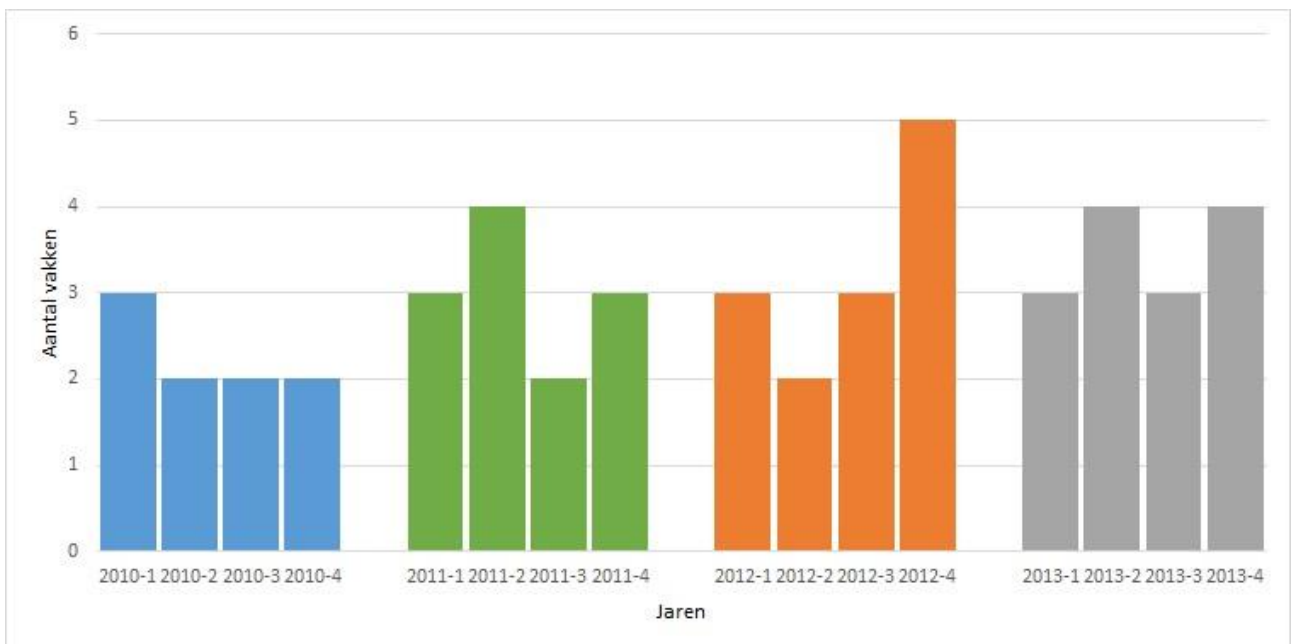
Courses (x axis: years, y axis: number of courses): color and no proximity

Statement: Twice as much courses are provided during the second quarter of 2013 than during the second quarter of 2010 (true).



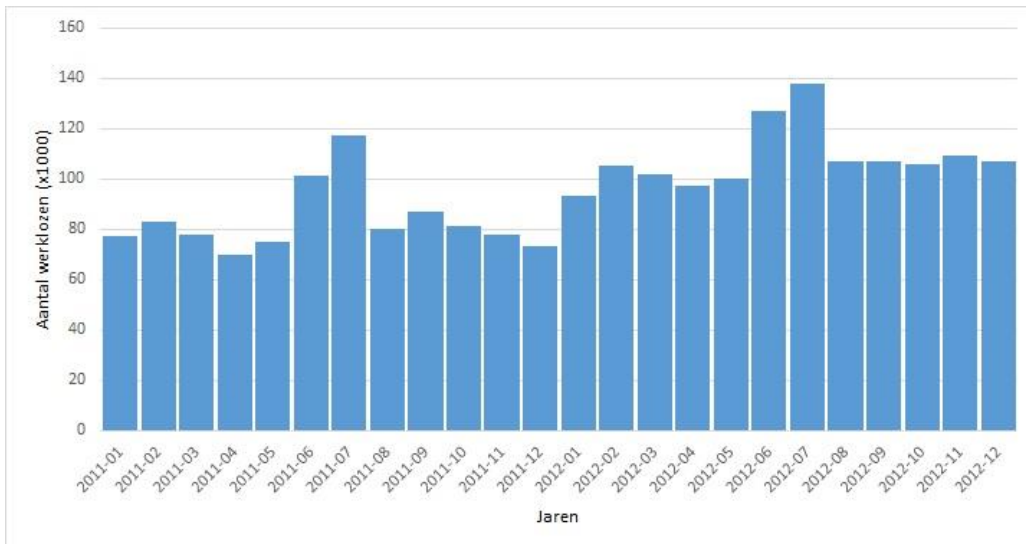
Courses (x axis: years, y axis: number of courses): no color and proximity

Statement: Twice as much courses are provided during the second quarter of 2013 than during the second quarter of 2010 (true).



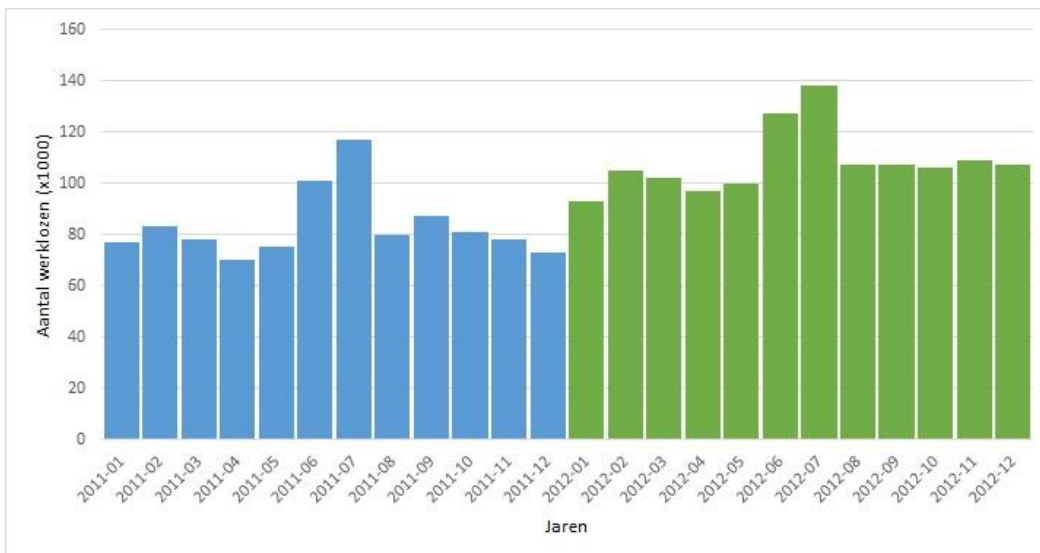
Courses (x axis: years, y axis: number of courses): color and proximity

Statement: Twice as much courses are provided during the second quarter of 2013 than during the second quarter of 2010 (true).



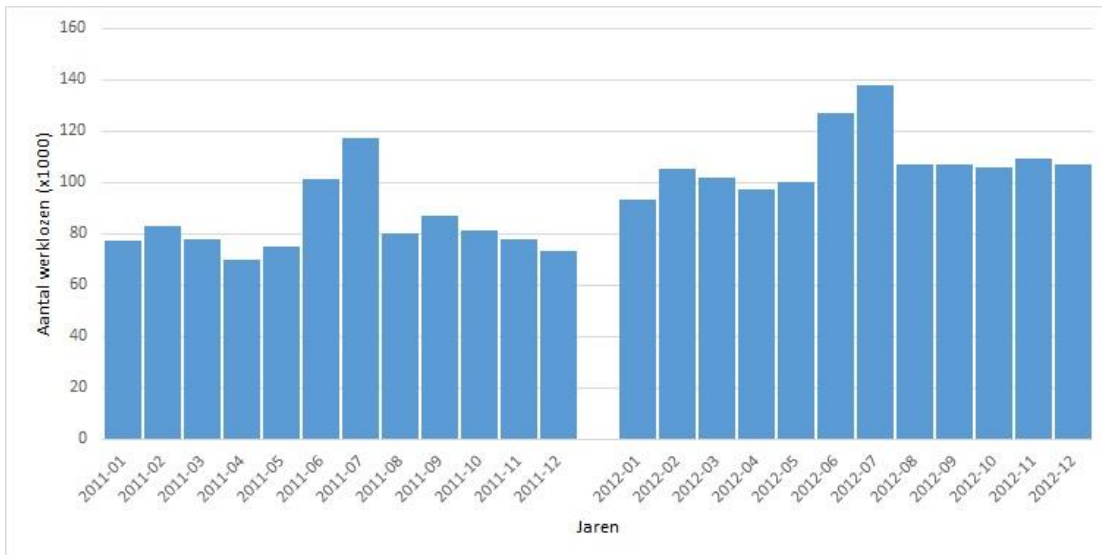
Unemployment (x axis: years, y axis: number of unemployed): no color and no proximity

Statement: The number of employed is higher in July 2012 than in July 2011 (true).

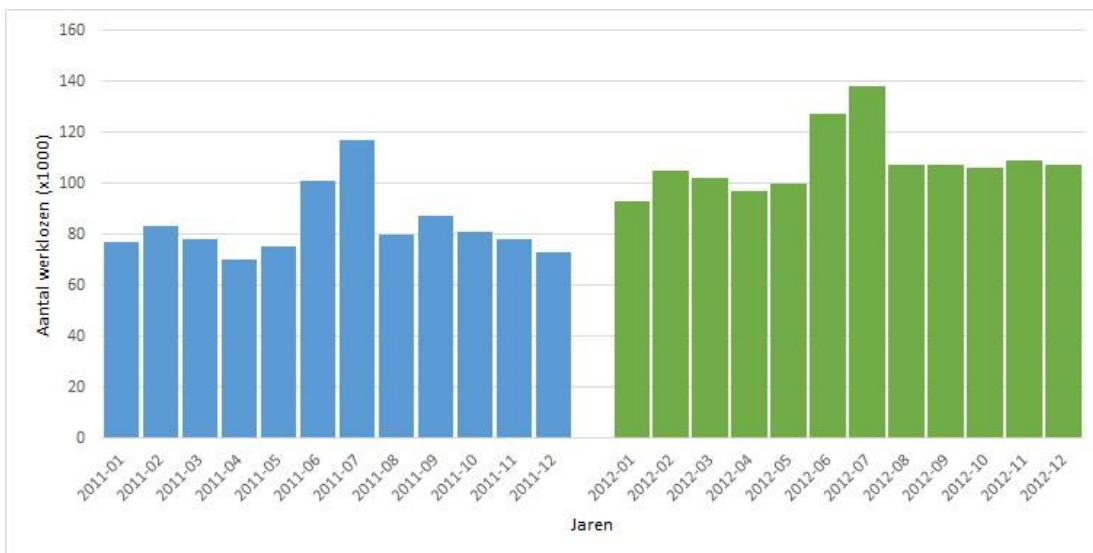


Unemployment (x axis: years, y axis: number of unemployed): color and no proximity

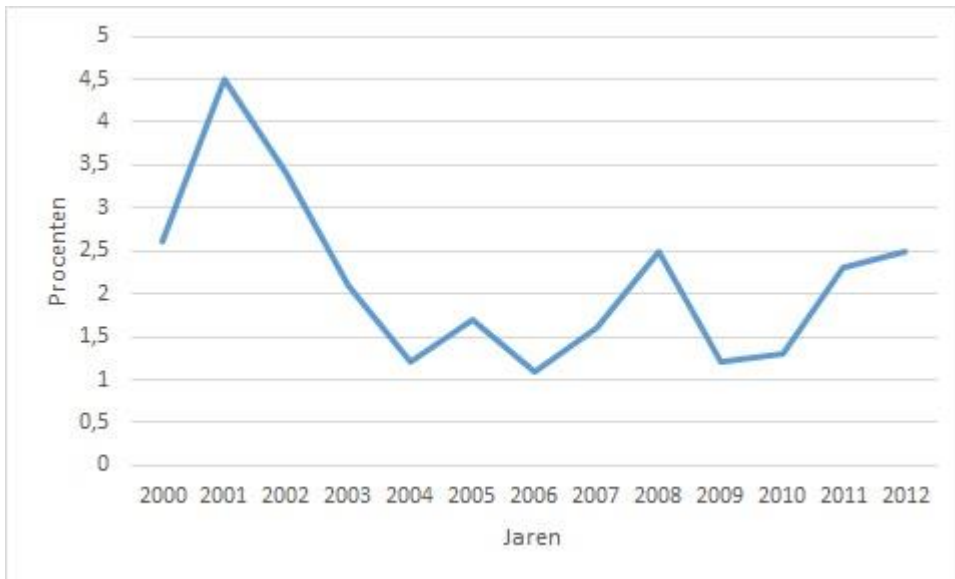
Statement: The number of employed is higher in July 2012 than in July 2011 (true).



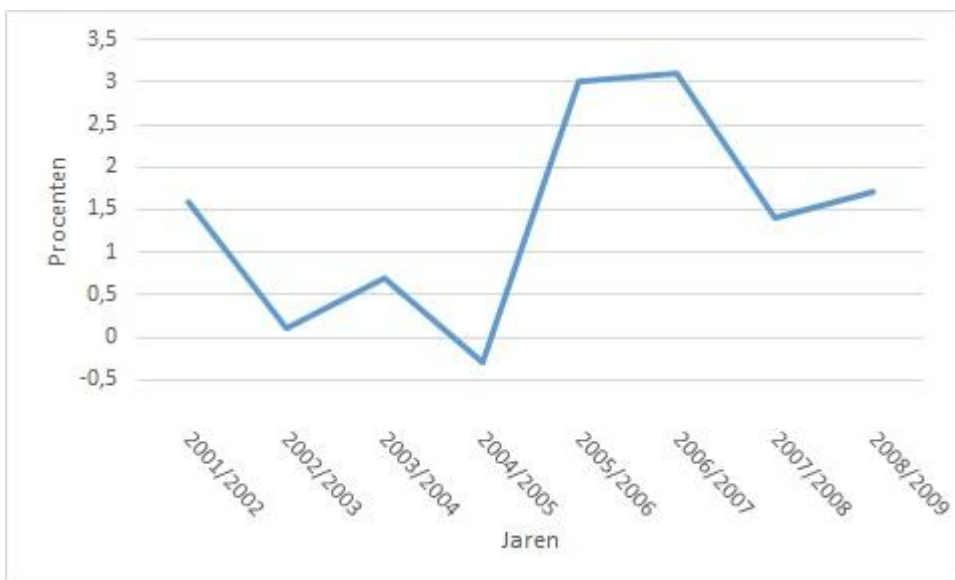
Unemployment (x axis: years, y axis: number of unemployed): no color and proximity  
 Statement: The number of employed is higher in July 2012 than in July 2011 (true).



Unemployment (x axis: years, y axis: number of unemployed): color and proximity  
 Statement: The number of employed is higher in July 2012 than in July 2011 (true).

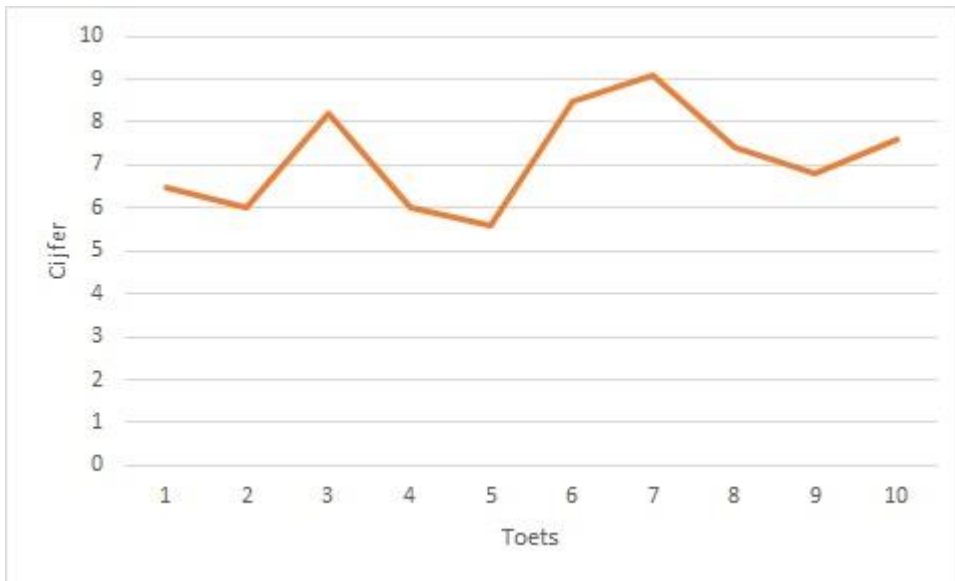


Inflation (x axis: years, y axis: percentages): The inflation is increased to 2.5 percent in 2008 (true).

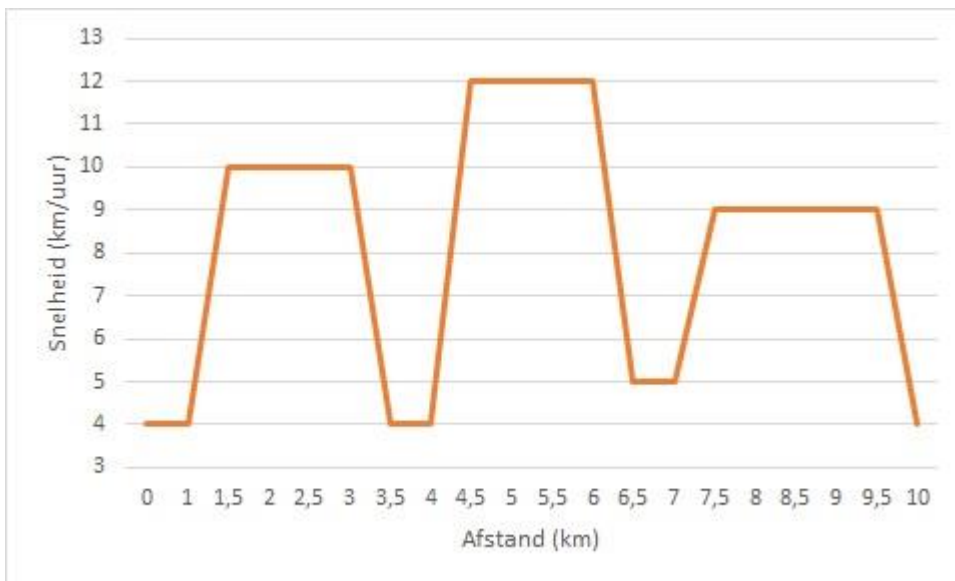


Buying power (x axis: years, y axis: percentages): In 2005/2006 the buying power is increased to 3 percent (true).

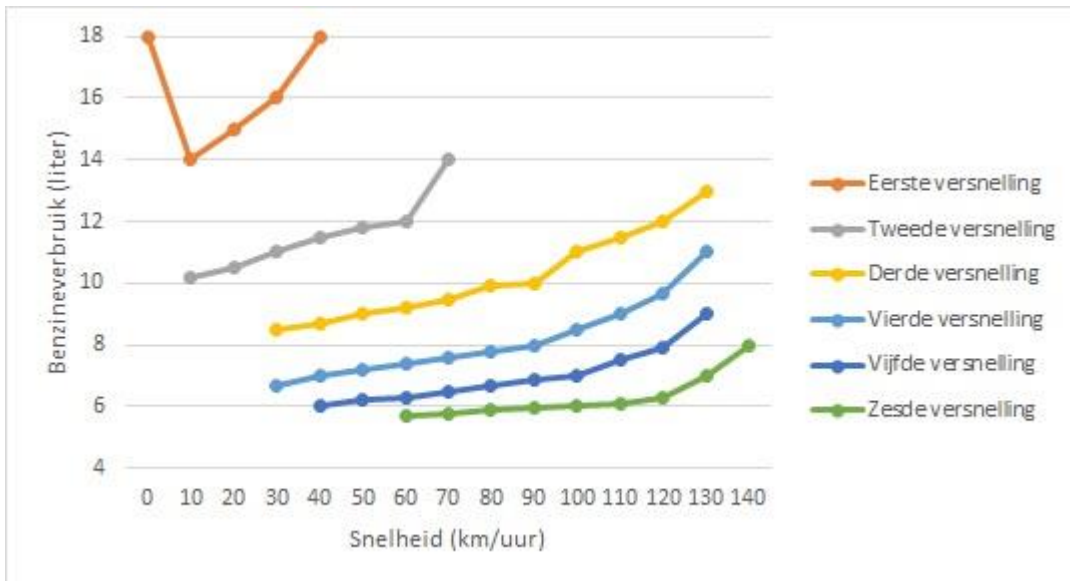




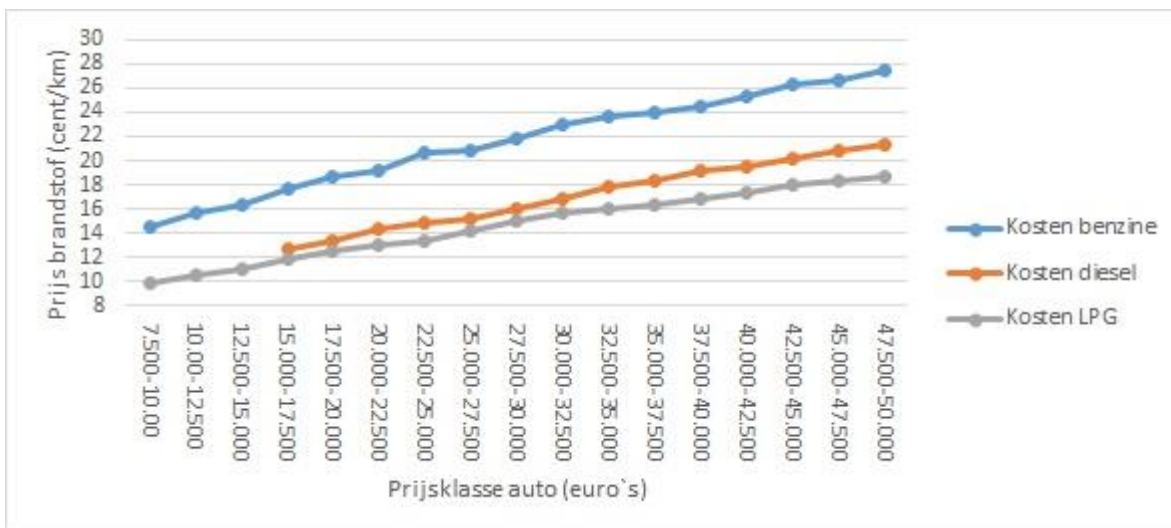
Grade (x axis: exam, y axis: grade): When this student got an eight or higher, he got at the next exam a lower grade (false).



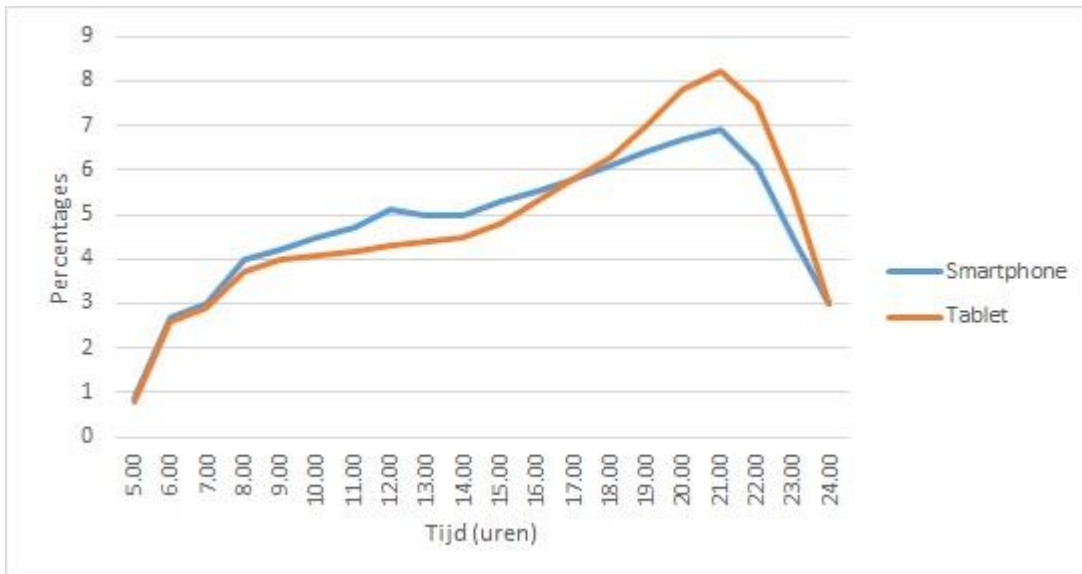
Running (x axis distance (k.), y axis: speed (km/hour)): This runner runs intervals of 1.5 kilometers (false).



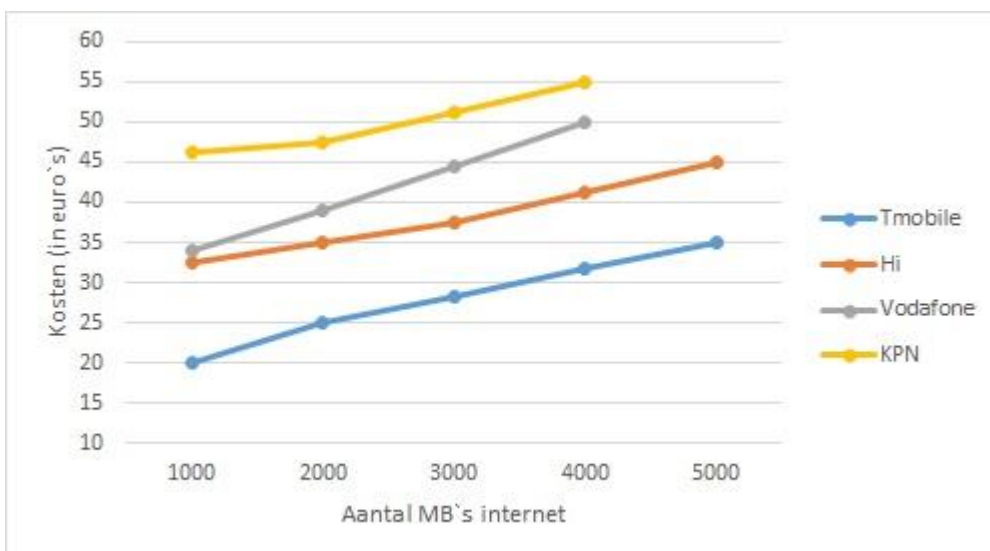
Fuel (x axis: speed (km/hour), y axis: fuel consumption (liter), legend: first gear to sixth gear): When a car drives 120 kilometers per hour, the car uses almost twice as much fuel in the third gear as in the sixth gear (true).



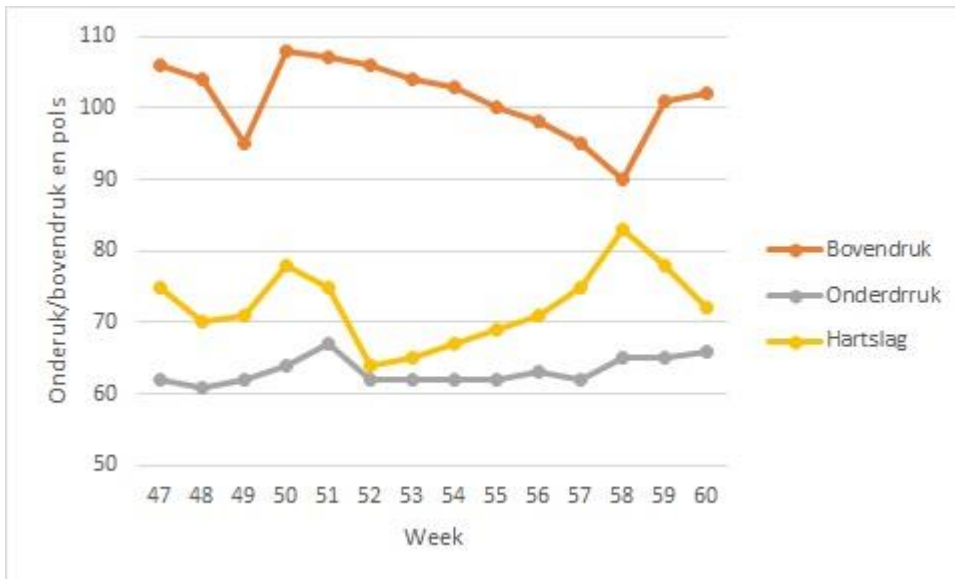
Priceperkm (x axis: price range (euro), y axis: price fuel (cent/km), legend: costs fuel, costs diesel, costs LPG): The price per kilometer increases faster for more expensive fuel cars than for more expensive LPG cars (true).



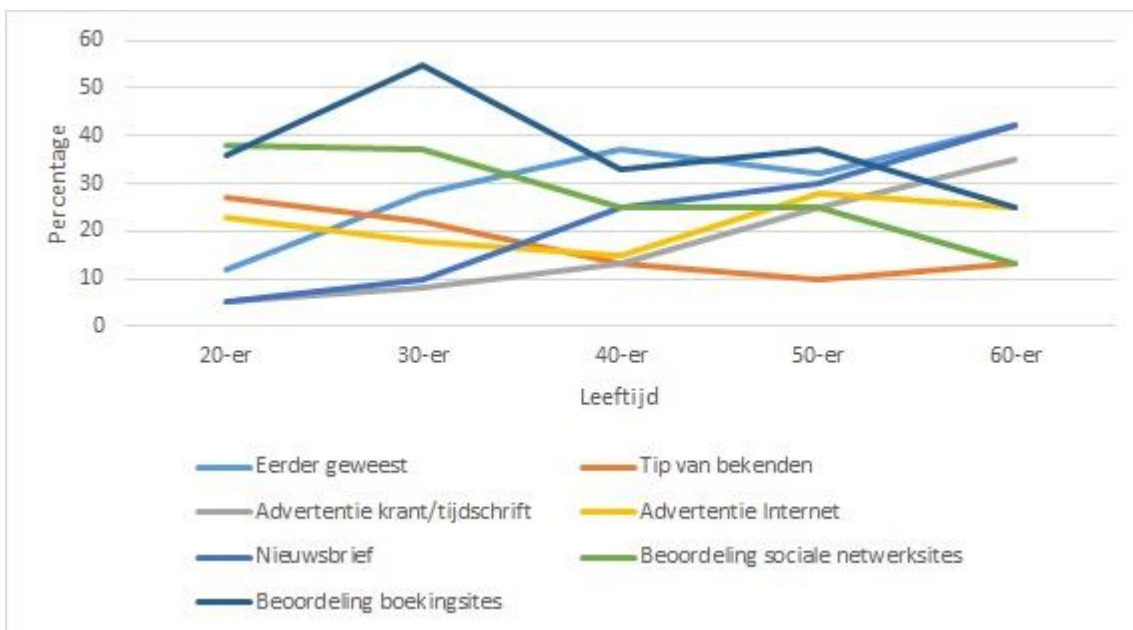
Smartphones (x axis: time (hours), y axis: percentages, legend: smartphone, tablet): The use of smartphones is higher till five o'clock than the use of tablets (true).



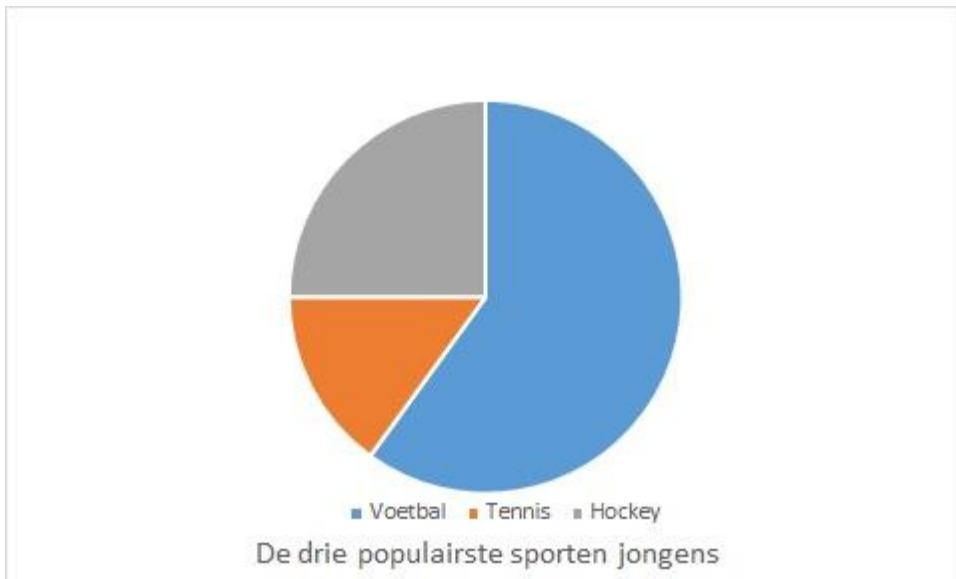
Internet (x axis: number of MBs internet, y axis: costs (in euros), legend: Tmobile, HI, Vodafone, KPN): At KPN: The more MBs you get in your internet bundle, the cheaper the bill (false).



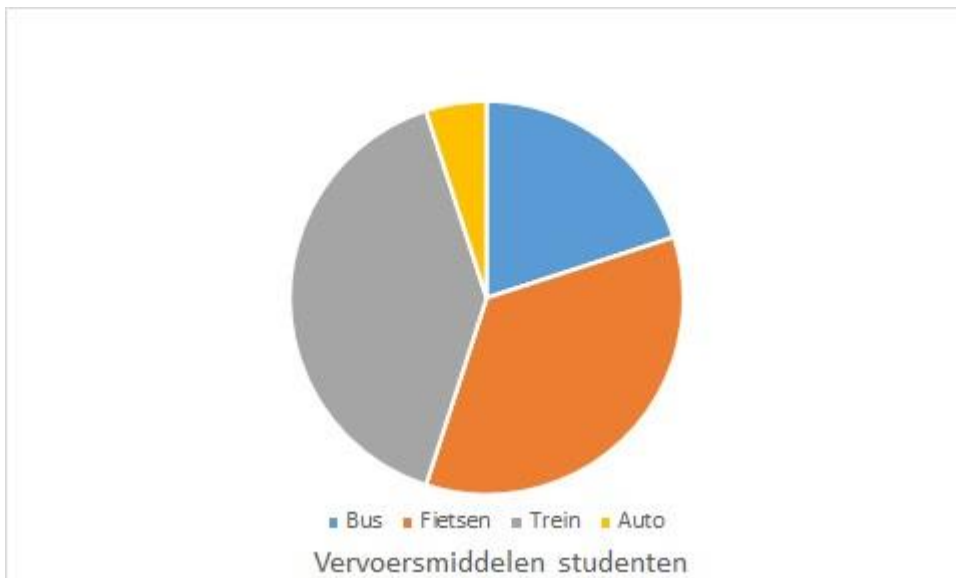
Heartbeat (x axis: week, y axis: pressure top pressure and wrist, legend: top pressure, pressure, heartbeat): If the heart beat increases, the top pressure increases (false).



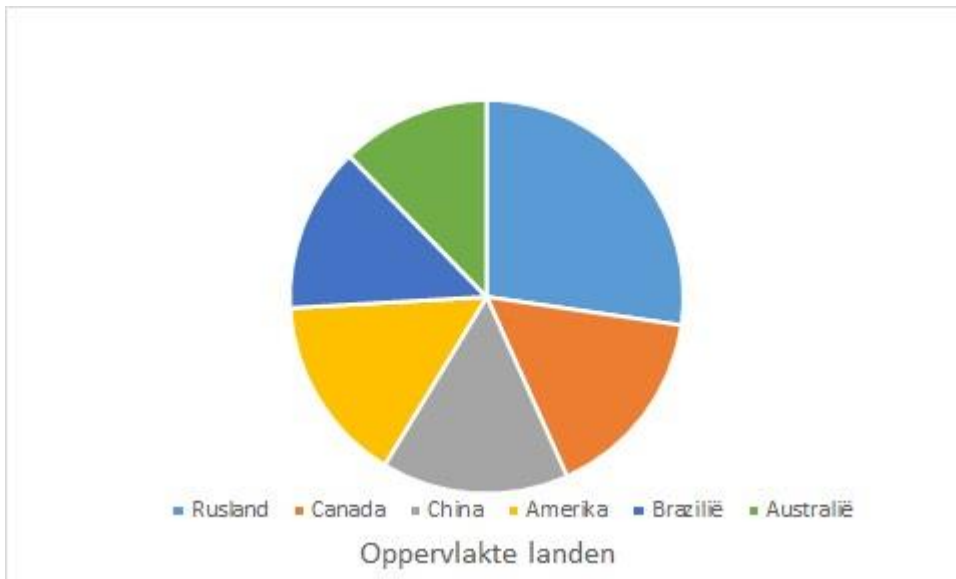
Holiday (x axis: age, y axis: percentages, legend: been previously, advertorial newspaper/magazine, newsletter, review booking sites, tip from friends, advertorial internet, review social network sites): Elderly find quicker a holiday via advertorials on the internet than via advertorials in the newspaper (false).



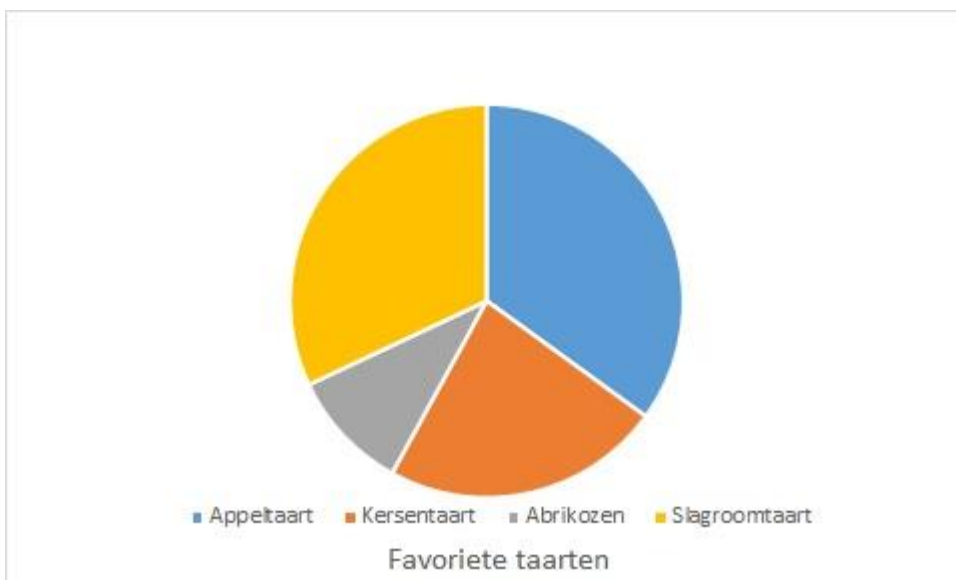
Sport (blue: soccer, orange: tennis, grey: hockey): More boys play hockey than tennis (true).



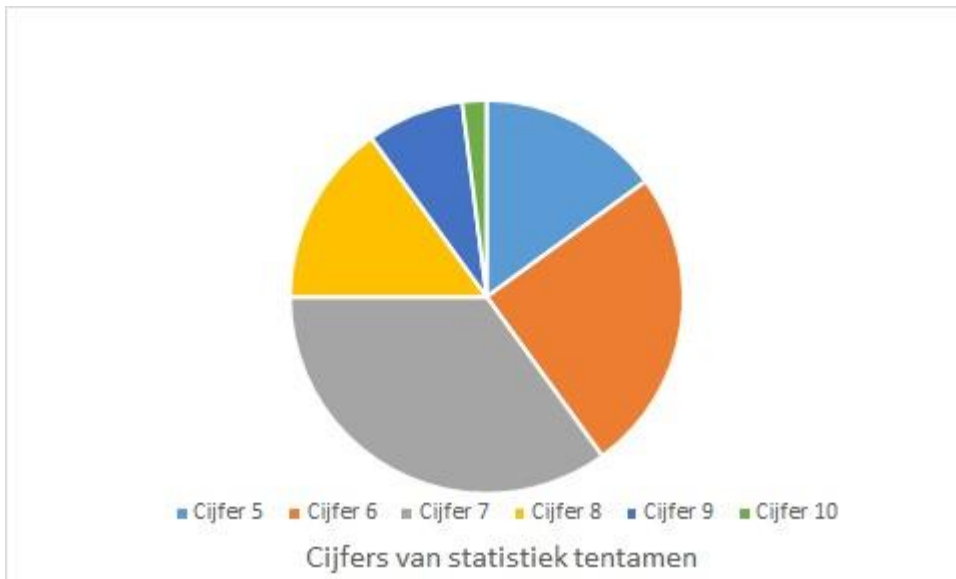
Transport (blue: bus, orange: bicycles, grey: train, yellow: car): The second favourite transport for students is the bicycle (true).



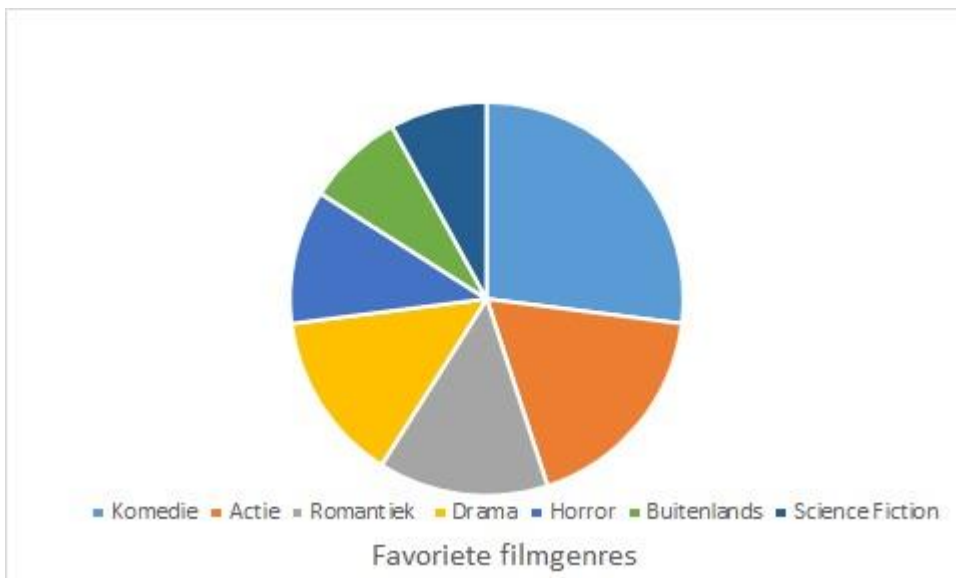
Area (blue: Russia, orange: Canada, grey: China, yellow America: dark blue: Brazil, green: Australia): The area of Canada is smaller than the area of Russia (true).



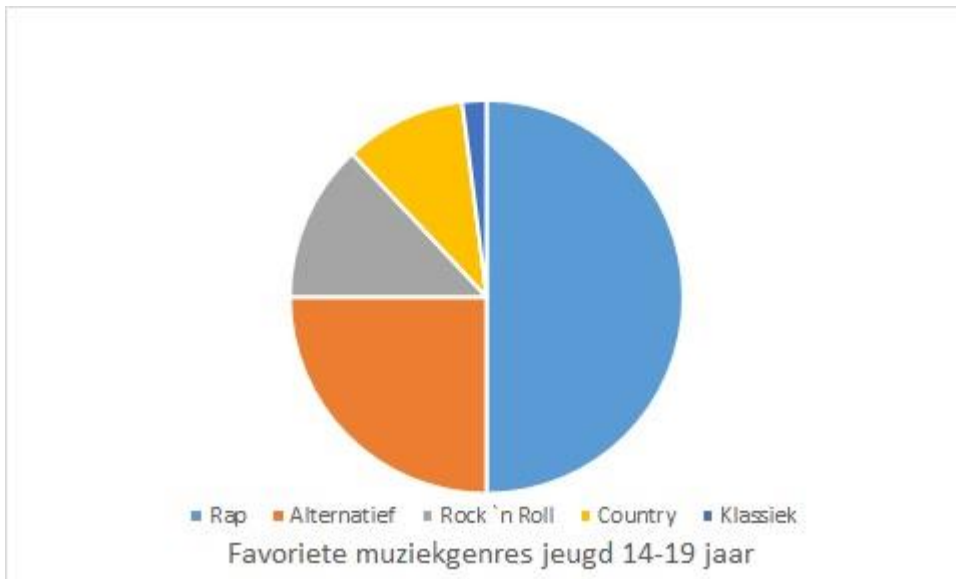
Cake (blue: apple pie, orange: cherry pie, grey: apricot pie, yellow: cream pie): The last choice for most of the people on a party is apricot pie (true).



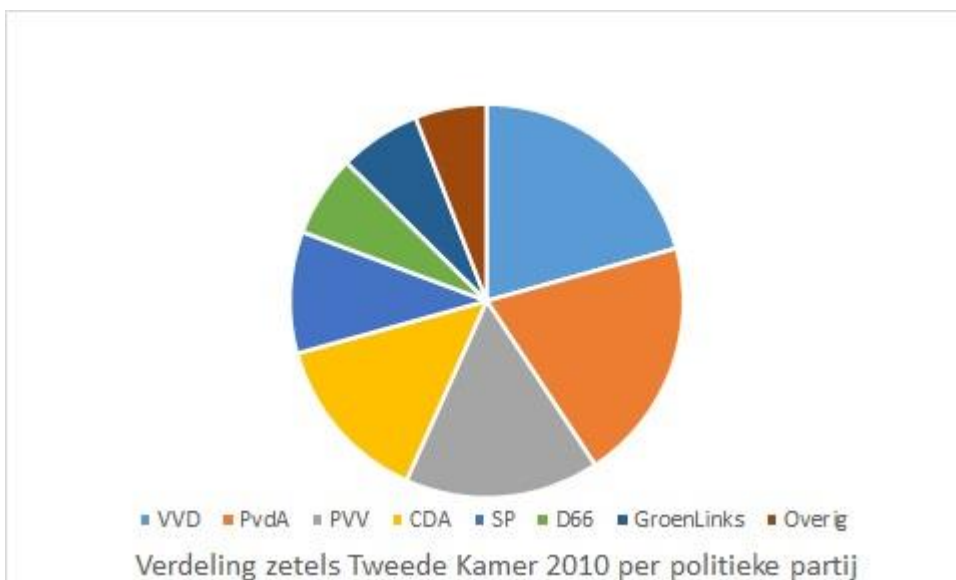
Statistics (blue grade 5, orange: grade 6, grey: grade 7, yellow: grade 8, dark blue: grade 9, green: grade 10): More students got a 7 than a 6 (false).



Movie genre (blue: comedy, orange: action, grey: romantics, yellow: drama, dark blue: horror, green: foreign, darkest blue: science fiction): The second favourite movie genre is horror (false).

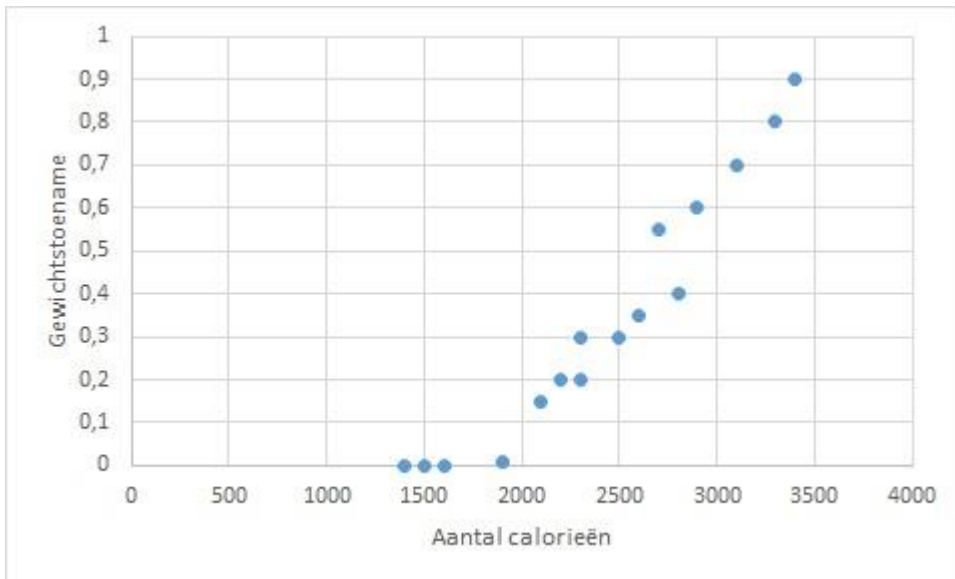


Music (blue: rap, orange: alternative, grey: rock `n roll, yellow: country, dark blue: classic):  
 The youth like less alternative music than country music (false).

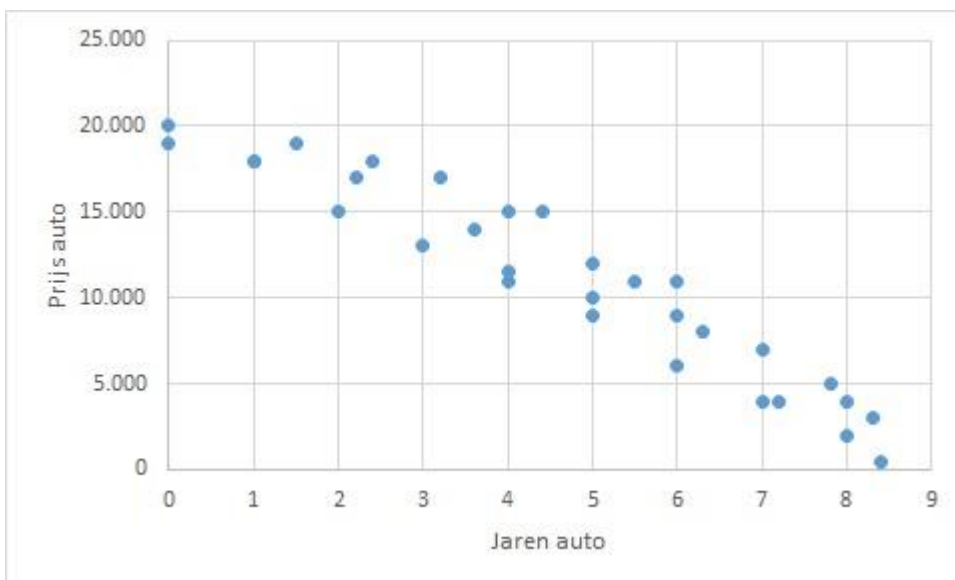


SP (brown: remaining): The least number of people voted on SP (false).

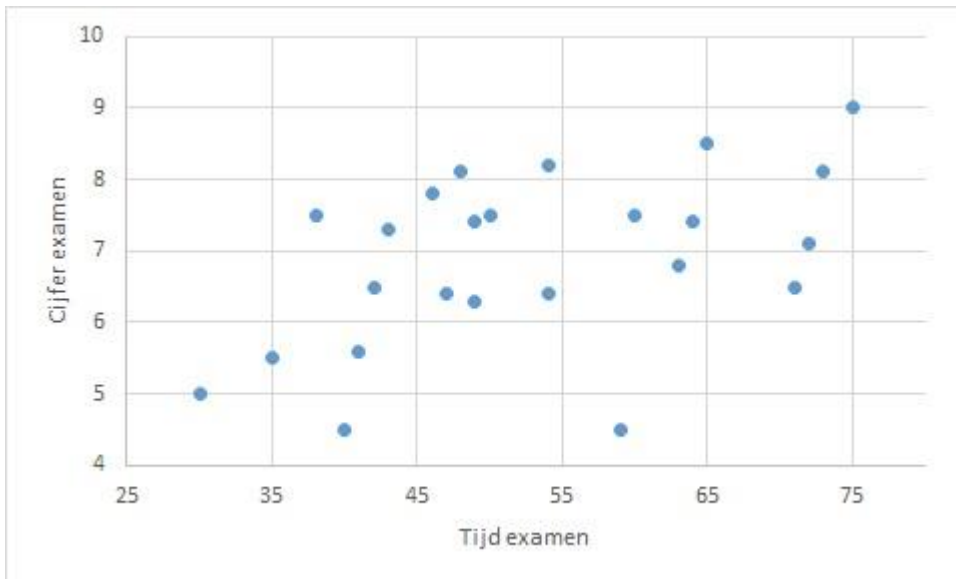




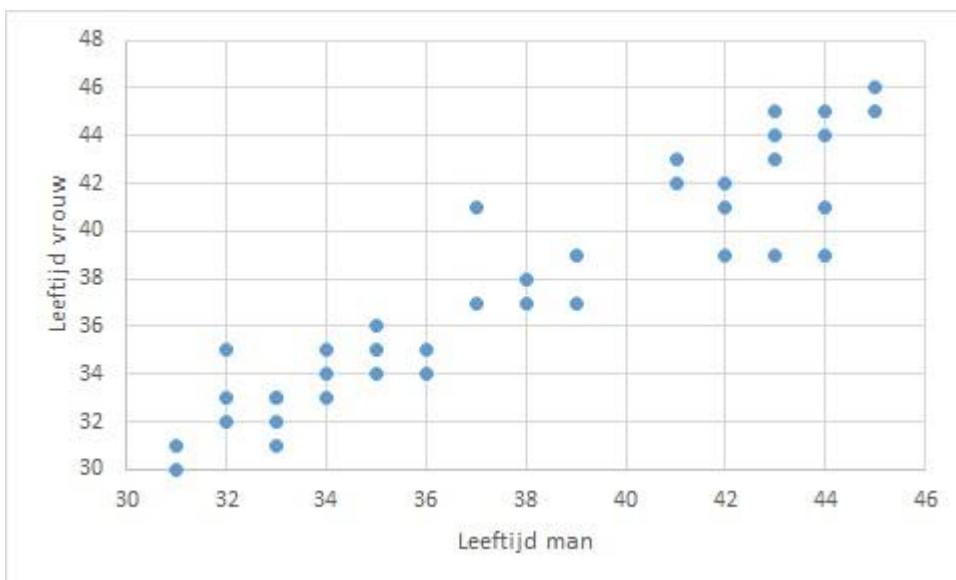
Gaining weight (x axis: number of calories, y axis: gaining weight): The relation between gaining weight and the total taken calories is positive (true).



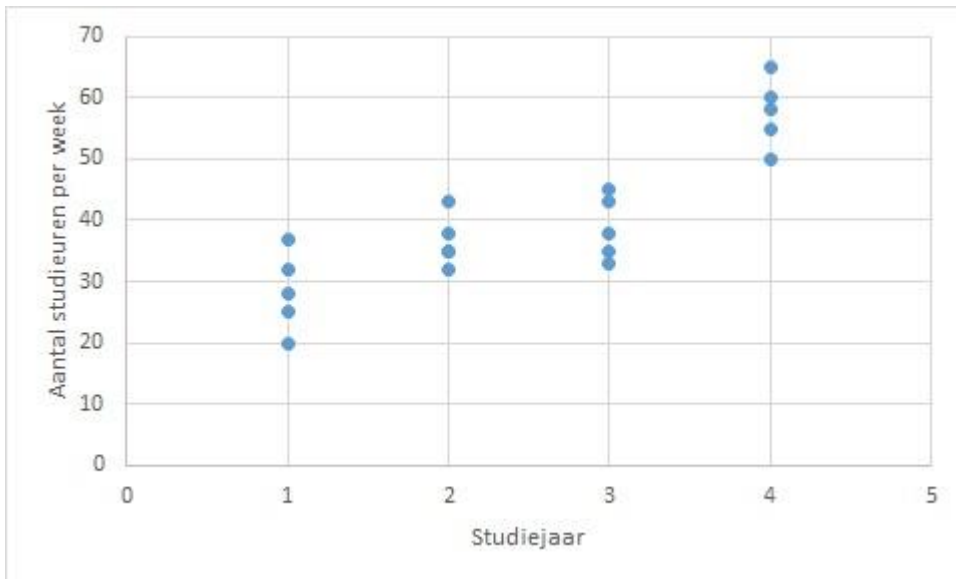
Pricecar (x axis: years car, y axis: price car): If the car is 5 years old, the price of the car would be at least 10.000 euro (false).



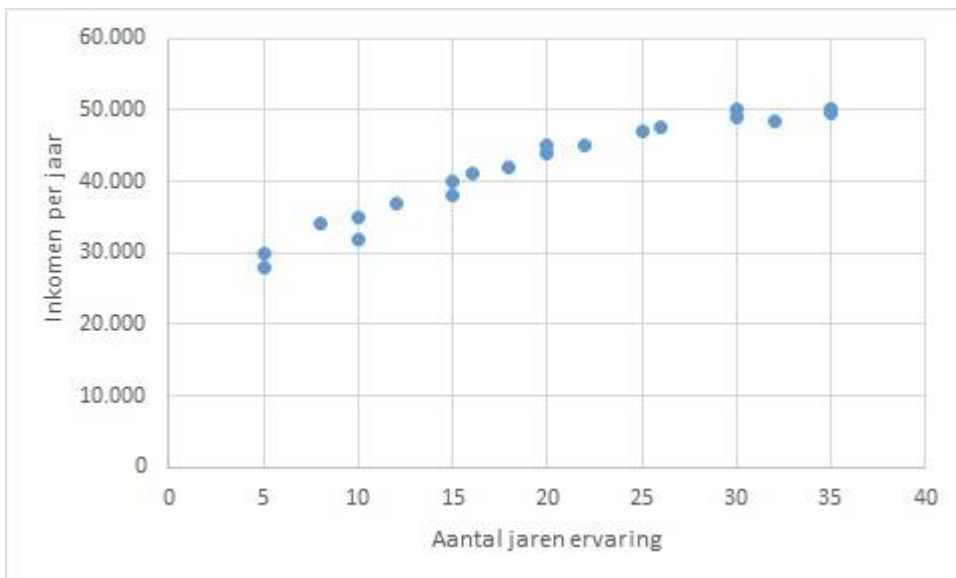
Exam time (x axis time exam, y axis: grade exam): Most of the students take 45-55 minutes to do an exam (true).



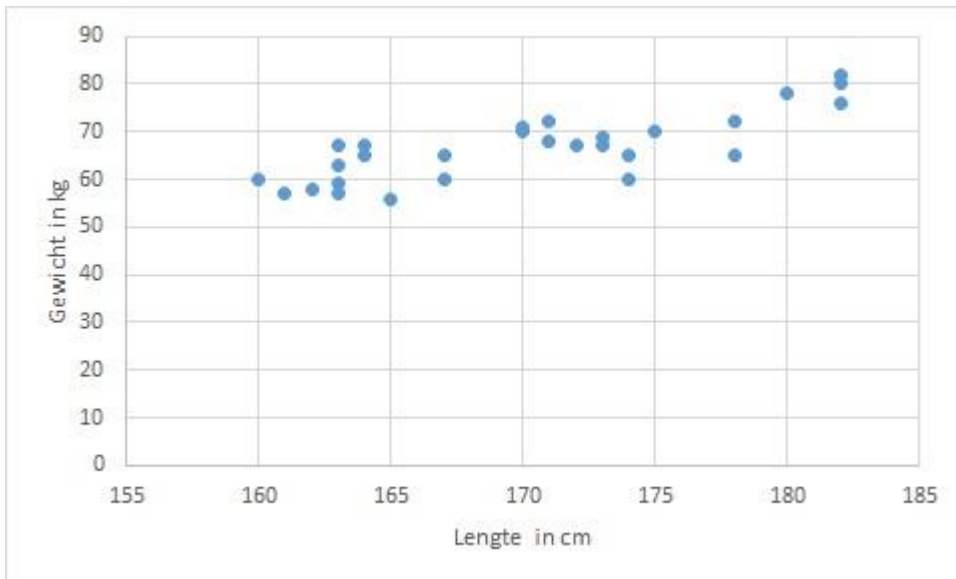
Age men women (x axis age men, y axis: age women): If the man is 32, the woman is 32 or older (true).



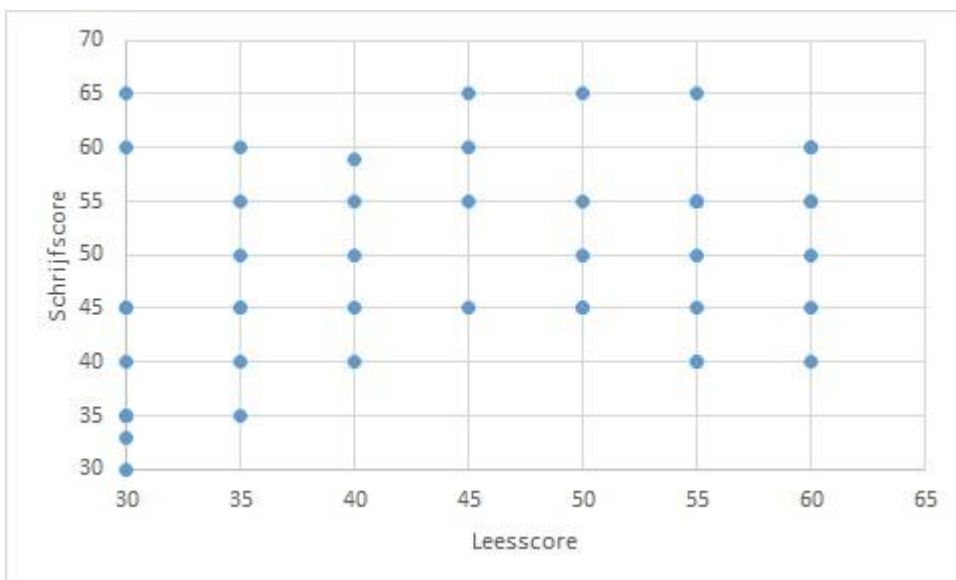
Time study (x axis: study year, y axis: number of study hours per week): Most of the students spend 30-40 hours a week to study (true).



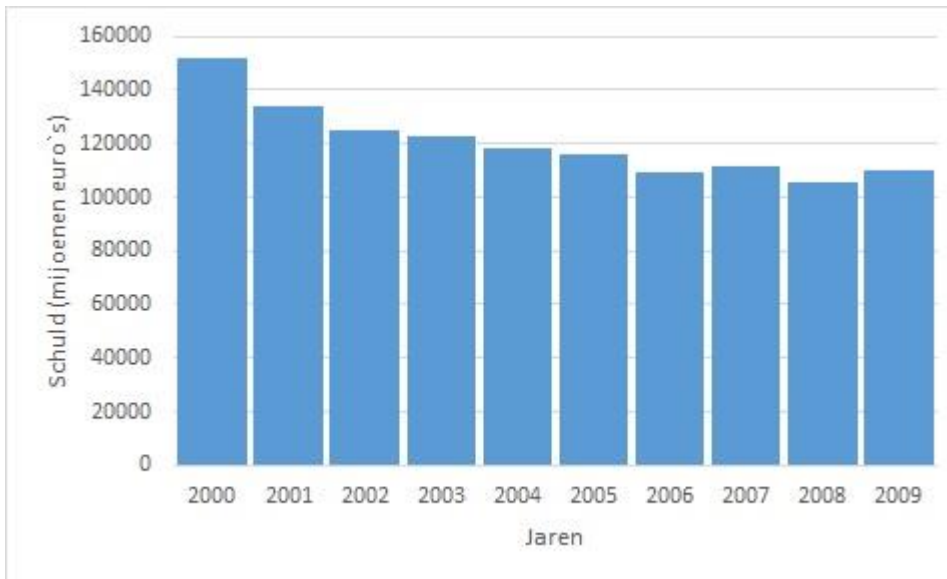
Experience ( x axis: number of years of experience, y axis: income per year): Most of the employees earn 30.000-40.000 euros per year (false).



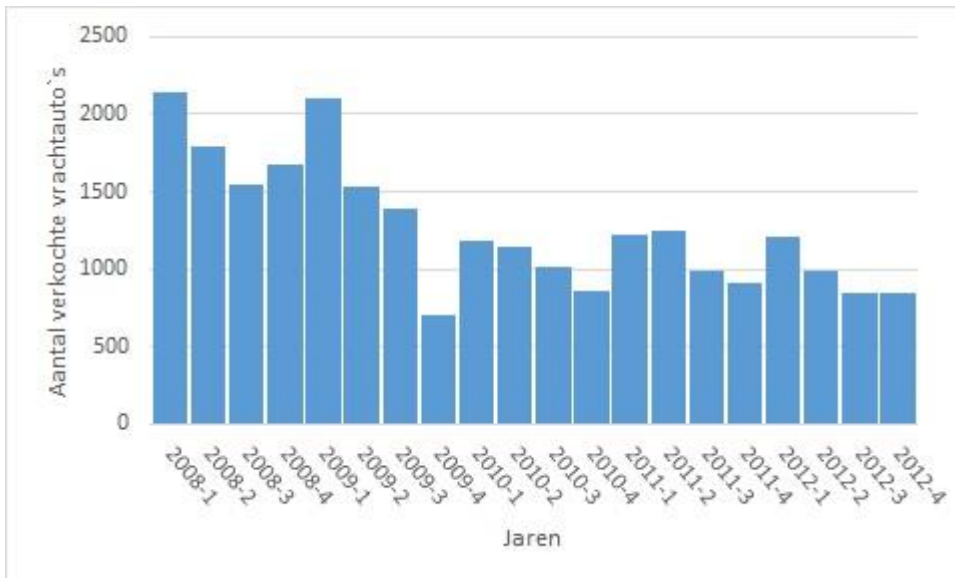
Length weight (x axis: length in cm, y axis: weight in kg): The relation between length and weight is strongly positive (false).



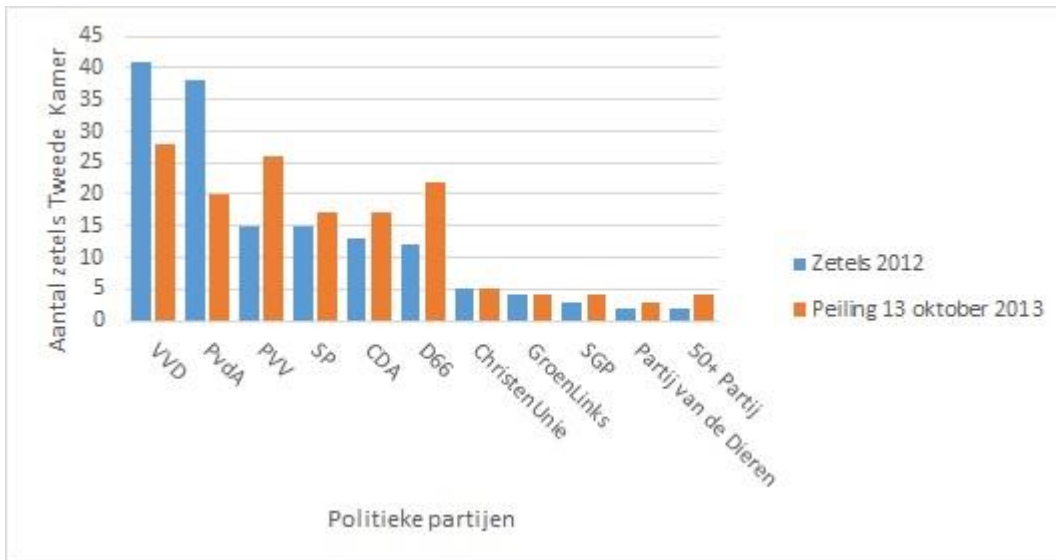
Reading score (x axis: reading score, y axis: writing score): A reading score of 55 is the most common (false).



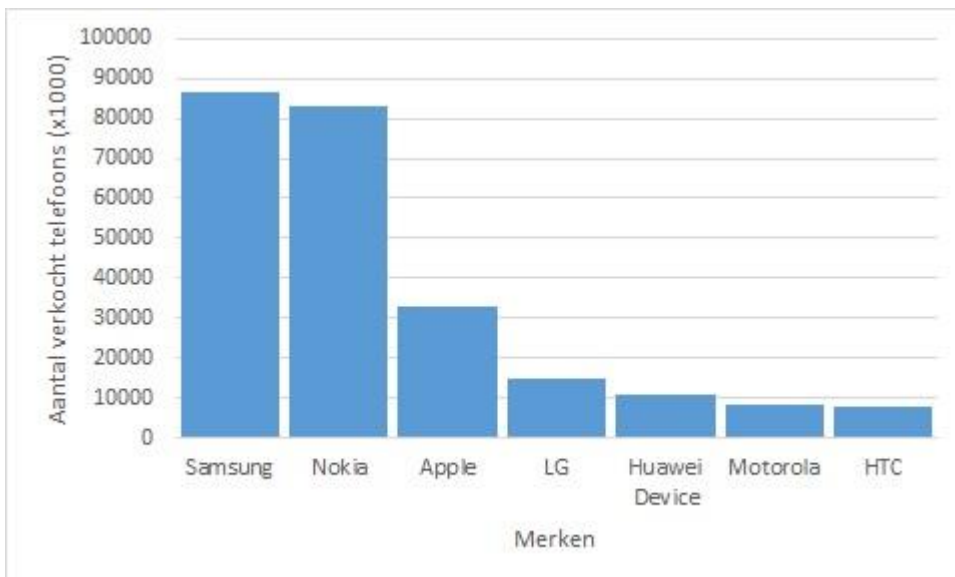
Debt (x axis: year, y axis: debt (in millions euros): The debt was higher in 2004 than in 2003 (false).



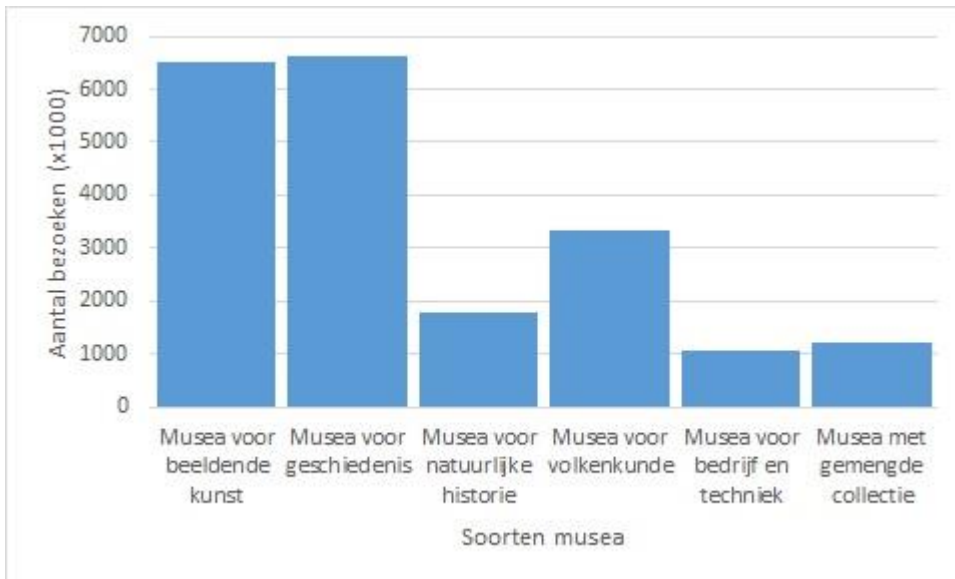
Cars (x axis: years, y axis: number of sold trucks): During the fourth quarter every year the least number of trucks are sold (false).



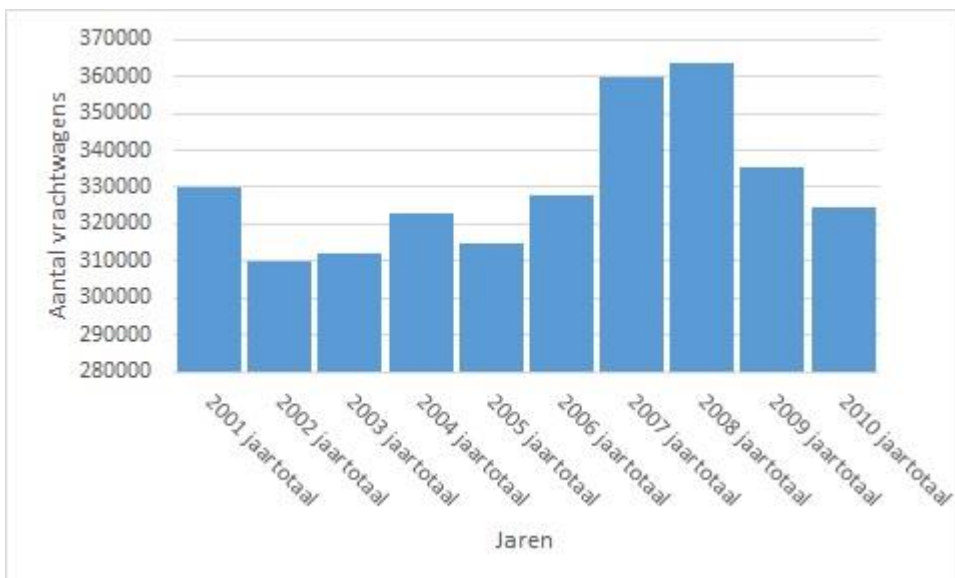
SP (x axis: political parties, y axis: number of seats Tweede Kamer, legend: seats 2012, poll 13 October 2013): The SP loses seats in “de Tweede Kamer” according to the poll (false).



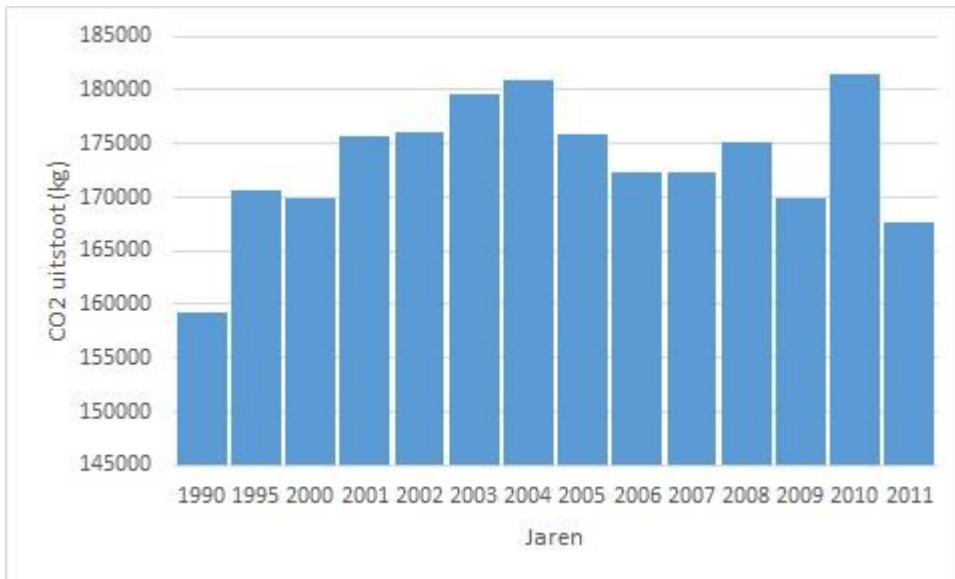
Brands (x axis: brands, y axis: number of sold phones (x1000)): Apple sold 40 million phones (false).



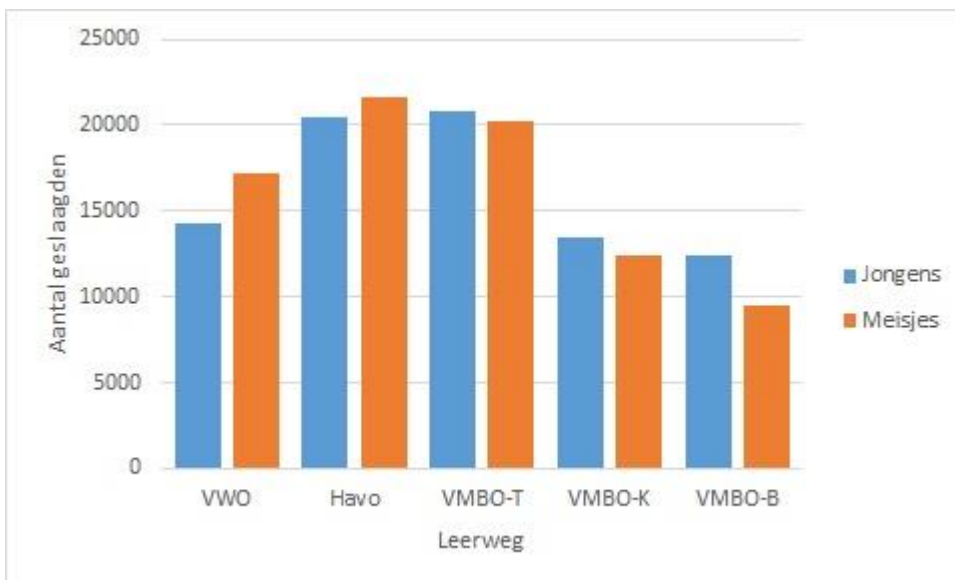
Museum ( axis: types of museums: fine arts, history, natural history, ethnology, business and technology, mixed collection, y axis: number of visitors (x 1000)): Museums with mixed collections got the least number of visitors (false).



Driving cars (x axis: years, y axis: number of trucks): In 2007 drove the least trucks (false).

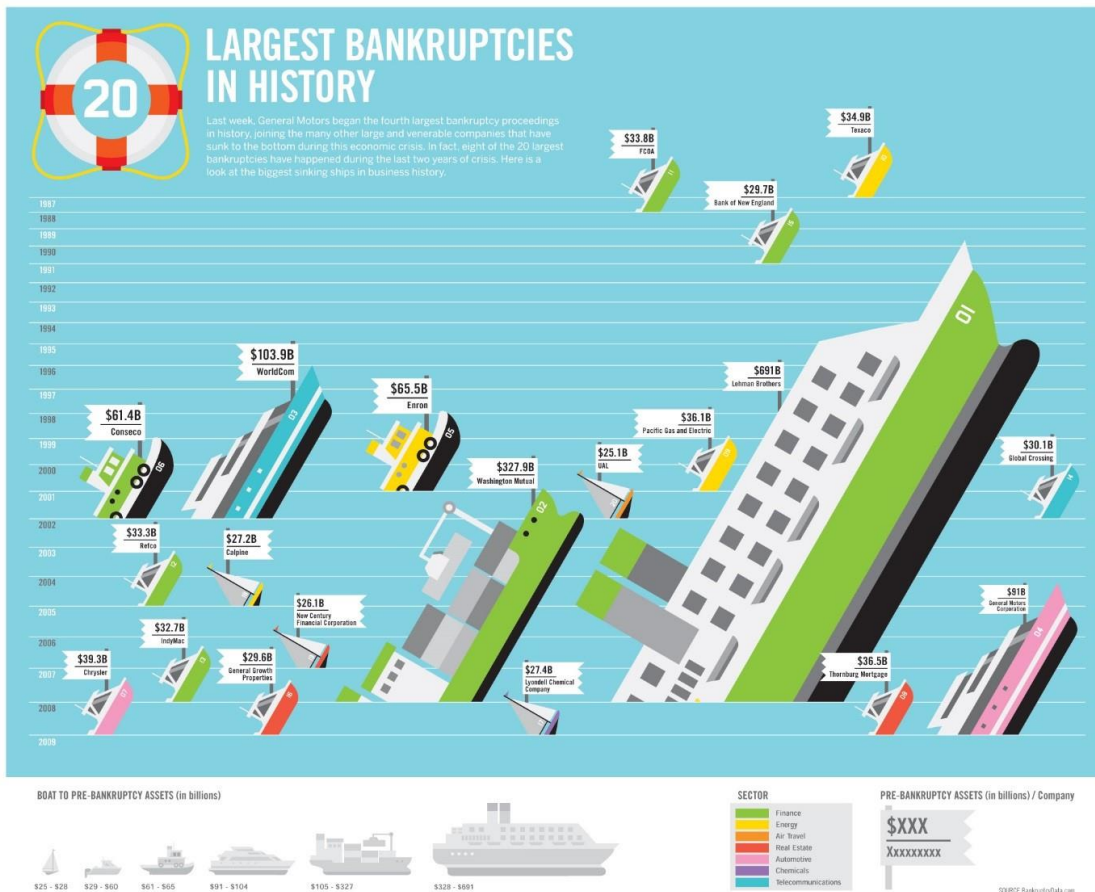


CO2 (x axis: years, y axis: CO2 emissions (kg)): In 2003 175000 kg CO2 is emitted (false).



School (x axis school, y axis: number of graduated, legend: boys, girls): Of all schools most of the girls graduate from VMBO-T (false).





Boat: Reprinted from *Largest Bankruptcies in History* (n.d.), by News by Design Copyright by News by Design. Retrieved from <http://news-by-design.com/the-biggest-bankruptcies-of-all-time/>

Statement: Most of the banks went bankrupt in 2008 (false).

**Built by the US firm Lockheed Martin, the F-16 Falcon comes at a price of \$14.6 million per plane. The US has 732 active.**  
**\$18.6 BILLION** LOCKHEED MARTIN'S 2008 AIRCRAFT AND WEAPON SALES, THE HIGHEST IN THE WORLD

**118 HELICOPTERS SOLD TO IRAQ BY U.S. COMPANIES, 1990-1991**  
**19** IRACH HELICOPTERS DESTROYED BY U.S. IN 1991 GULF WAR

There are roughly 106,000 tanks in existence worldwide.  
**22%** PERCENTAGE IN THE MIDDLE EAST  
**8%** PERCENTAGE IN THE UNITED STATES

US aircraft carriers cost about \$4.6 billion apiece, and are generally named after past presidents.

The USA leads the modern space race with 110 military satellites in orbit. Russia has 40 and the rest of the world 20.

A one megaton bomb has an annihilation radius of 1.7 miles, inside of which most everything gets vaporized.  
**\$3.5 TRILLION** AMOUNT SPENT BY THE USA FROM 1945-1995 IN NUCLEAR PREPARATION

There are over 100 million unexploded landmines in 64 countries, taking an average daily toll of 70 lives.  
**\$3** COST OF A CHEAP LANDMINE  
**\$300** AVERAGE COST TO DEACTIVATE ONE LANDMINE

**\$25** COST OF A RUSSIAN KALASHNIKOV (AK-47) IN UZBEKISTAN

**83** MINIMUM NUMBER OF GUNS PER 100 PEOPLE IN THE USA  
**2** NUMBER IN THE NETHERLANDS

**4.4** PEOPLE KILLED IN THE U.S. WITH FIREARMS, PER 100,000  
**0.5** NUMBER IN FRANCE

**\$8,053** MAXIMUM FINE FOR CARRYING A KNIFE LONGER THAN 3 INCHES IN ENGLAND

**252** NUMBER OF 1997 IRISH ROBBERIES USING A GUN  
**590** NUMBER USING A SYRINGE

**1.8 MILLION** SCISSORS CONFISCATED AT U.S. AIRPORTS, 2002  
 MANY WERE DONATED TO THE U.S. BOY SCOUTS

### THE NUCLEAR FAMILY

A couple of one megaton nuclear bombs would obliterate Manhattan instantly. Russia currently has over 18,000 such bombs, and the USA has nearly 11,000. In 2001 George W. Bush and Russian President Vladimir Putin pledged by 2012 to reduce each of their countries' nuclear arsenals to "operationally deployed" warheads (those positioned for rapid use on delivery vehicles). As for the rest, Russia (and most of the world) wants to have them destroyed, but Bush has decided to keep them in storage, where they could easily be rearmend and deployed. Welcome to the family.

**USA** 10,729  
**RUSSIA** 1700 - 2200 "operationally deployed" warheads (those positioned for rapid use on delivery vehicles). As for the rest, Russia (and most of the world) wants to have them destroyed, but Bush has decided to keep them in storage, where they could easily be rearmend and deployed. Welcome to the family.

— 100 NUCLEAR BOMBS

**CHINA** 400  
**FRANCE** 300  
**UK** 200  
**ISRAEL** 200P  
**INDIA** 40  
**PAKISTAN** 24-48

### CHOOSE YOUR WEAPON

THE GLOBAL ARMS TRADE

**BANNED!**  
 COUNTRIES SUBJECT TO INTERNATIONAL ARMS EMBARGOS FOR ANY LENGTH OF TIME 1996-2008

- AFGHANISTAN
- ANGOLA
- AZERBAIJAN
- BOSNIA
- BURUNDI
- CHINA
- CONGO
- CROATIA
- ETHIOPIA
- INDONESIA
- IRAQ
- LIBERIA
- LEBYA
- MACEDONIA
- NIGERIA
- RWANDA
- SIERRA LEONE
- SLOVENIA
- SOMALIA
- SUDAN
- YUGOSLAVIA

### SPEND IT ON ME!

PER CAPITA DEFENSE SPENDING, 2002

World military spending in 2001 totalled \$839 billion, accounting for 2.9% of world GDP and a global average of \$137 per capita. Despite its massive defense budget (over 40% of the world total), the USA's per capita military spending is dwarfed by Israel's, a soaring \$1,625 per person.

**ISRAEL** \$1,625  
**USA** \$1,196  
**S. ARABIA** \$918  
**UK** \$602  
**FRANCE** \$562  
**JAPAN** \$367  
**ITALY** \$365  
**GERMANY** \$332  
**S. KOREA** \$279  
**IRAN** \$262  
**TURKEY** \$150  
**RUSSIA** \$78  
**BRAZIL** \$56  
**INDIA** \$12  
**CHINA** \$24

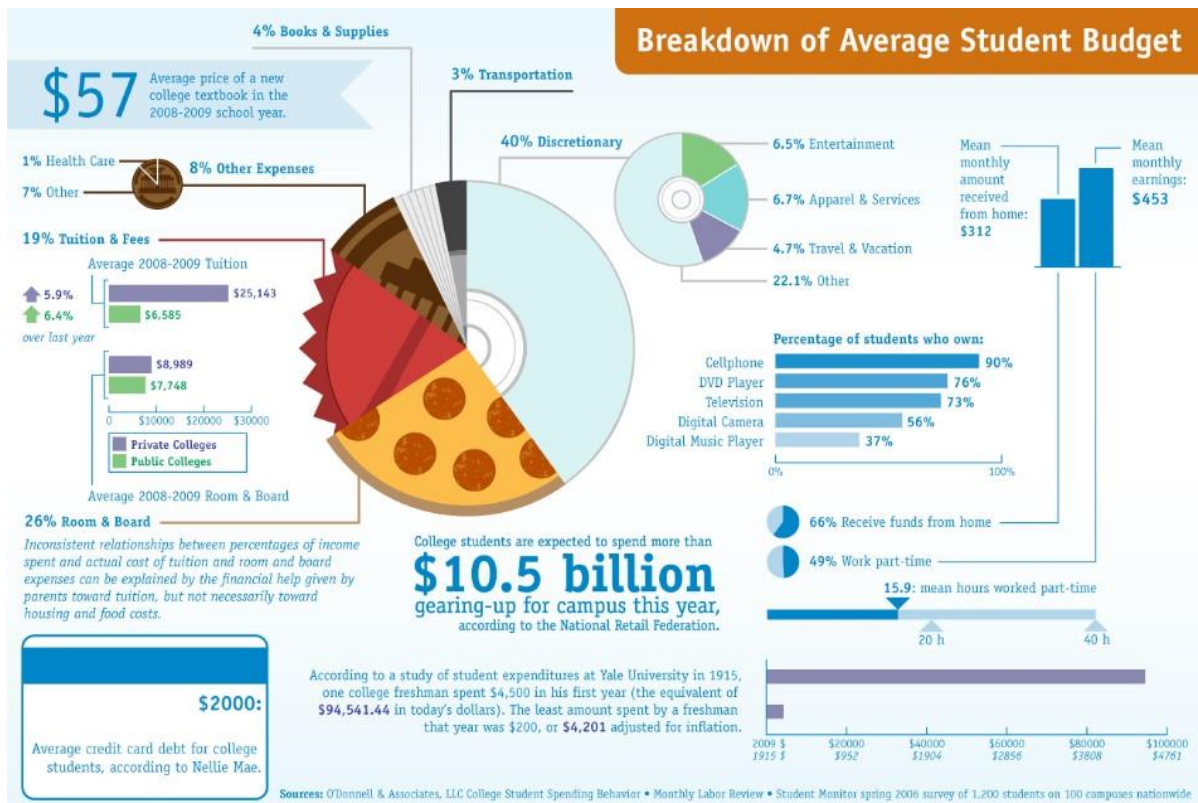
**THE BIG SPENDER(S)**  
 TOP 15 DEFENSE BUDGETS, 2002  
 MARKET VALUE IN BILLIONS

**USA** 467  
**FRANCE** 33.6  
**JAPAN** 27.7  
**UK** 21.1  
**GERMANY** 21.1  
**ITALY** 17.9  
**S. KOREA** 13.5  
**INDIA** 11.4  
**RUSSIA** 10.0  
**BRAZIL** 10.0

SOURCES: WASHINGTON POST - CIA WORLD FACTBOOK - U.S. DEPARTMENT OF DEFENSE - STOCKHOLM INTERNATIONAL PEACE INSTITUTE - WORLD WATCH INSTITUTE - THE MILITARY BALANCE 2002 YEARBOOK - IRISH NATIONAL POLICE - 'YOUR RIGHTS ONLINE'

Army: Reprinted from *Choose your Weapon* (n.d.). Copyright holder unknown. Retrieved from <http://www.princeton.edu/~ina/infographics/weapons.html>.

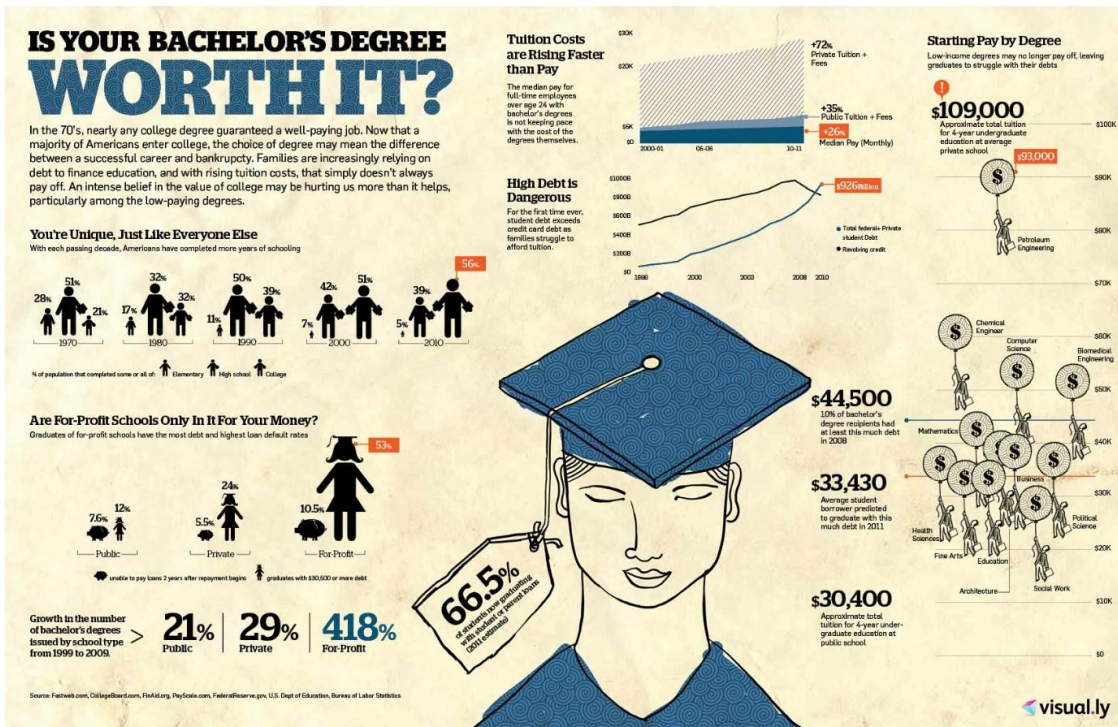
Statement: It costs 100 times more to deactivate a landmine than to buy it (true).



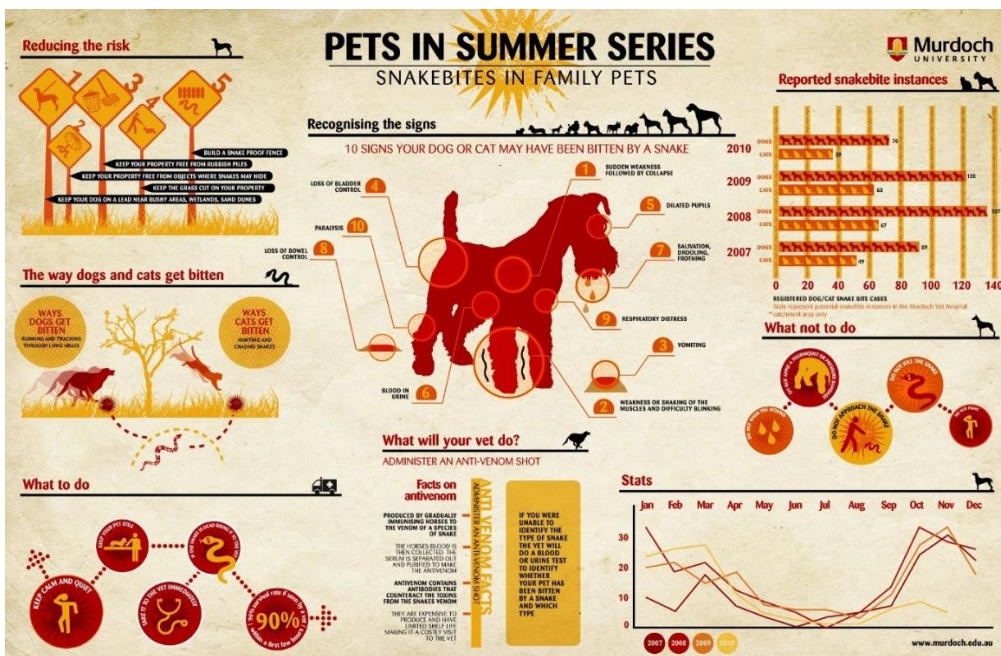
Students: Reprinted from *Breakdown of Average Student Budget* (n.d.), by Westwood College. Copyright by Westwood College. Retrieved from <http://www.westwood.edu/resources/student-budget>.

Statement: Students receive more money from their parents per month than they earn by themselves (false).





High school: Reprinted from *Is your Bachelor's Degree Worth it?*(n.d.), by Visual.ly.  
 Copyright by Visual.ly. Retrieved from <http://visual.ly/your-bachelors-degree-worth-it>.  
 Statement: More and more Americans have completed more years of schooling after high school (true).



Cat-dogs: Reprinted from *Pets in Summers Series* (n.d.), by The Daily Infographic. Copyright The Daily Infographic. Retrieved from <http://dailyinfographic.com/wp-content/uploads/2013/05/10-signs-your-dog-or-cat-has-been-bitten-by-a-snake-infographic.jpg>  
 Statement: Most of the snake bites occur with dogs (true).

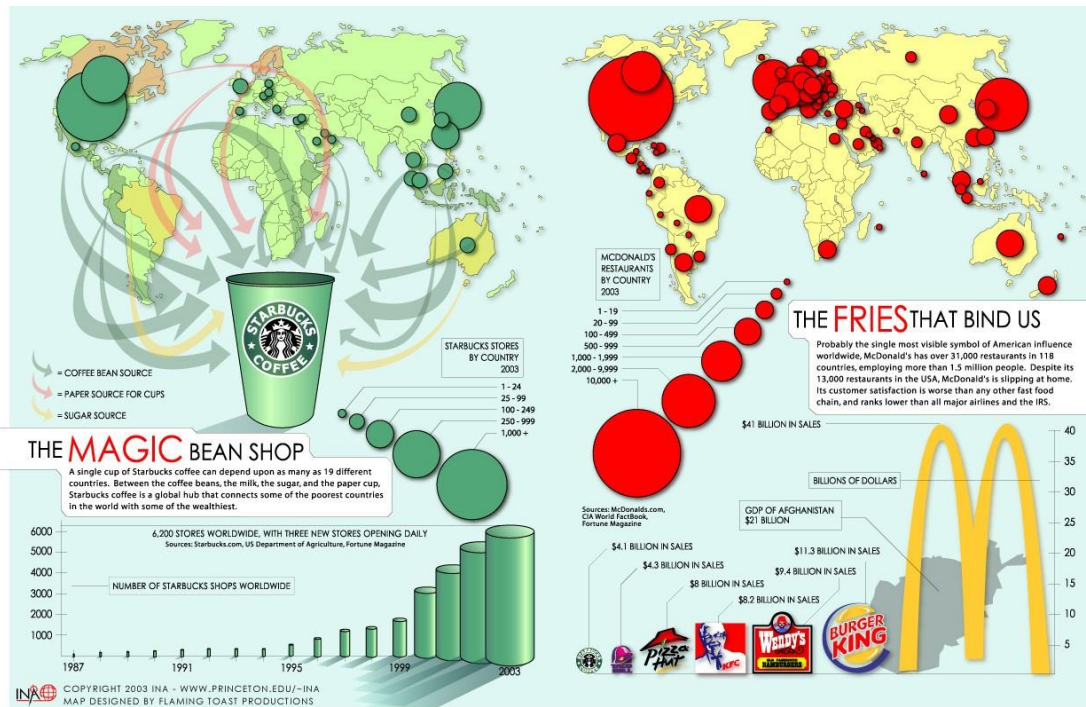
# very Pinteresting



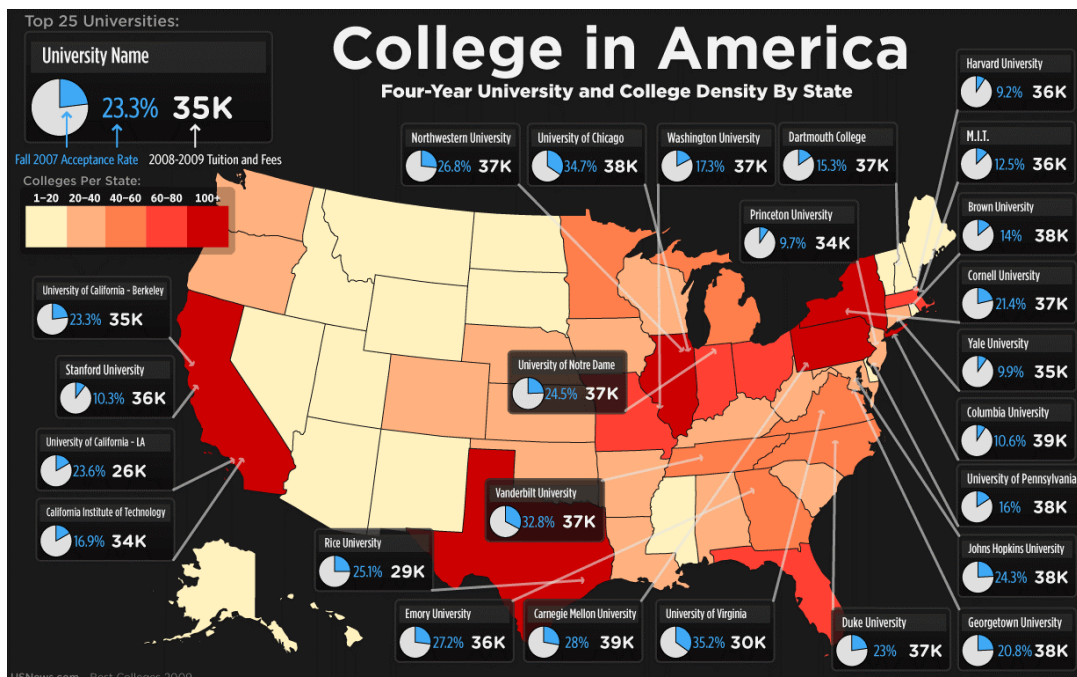
Pinterest: Reprinted from *Very Pinteresting* (n.d.), by Adverblog. Copyright by Adverblog. Retrieved from [http://www.adverblog.com/2012/02/26/interesting-facts-about-pinterest-infographic/pinterest\\_infographic2/](http://www.adverblog.com/2012/02/26/interesting-facts-about-pinterest-infographic/pinterest_infographic2/).

Statement: Pinterest is used the most by adolescents (false).





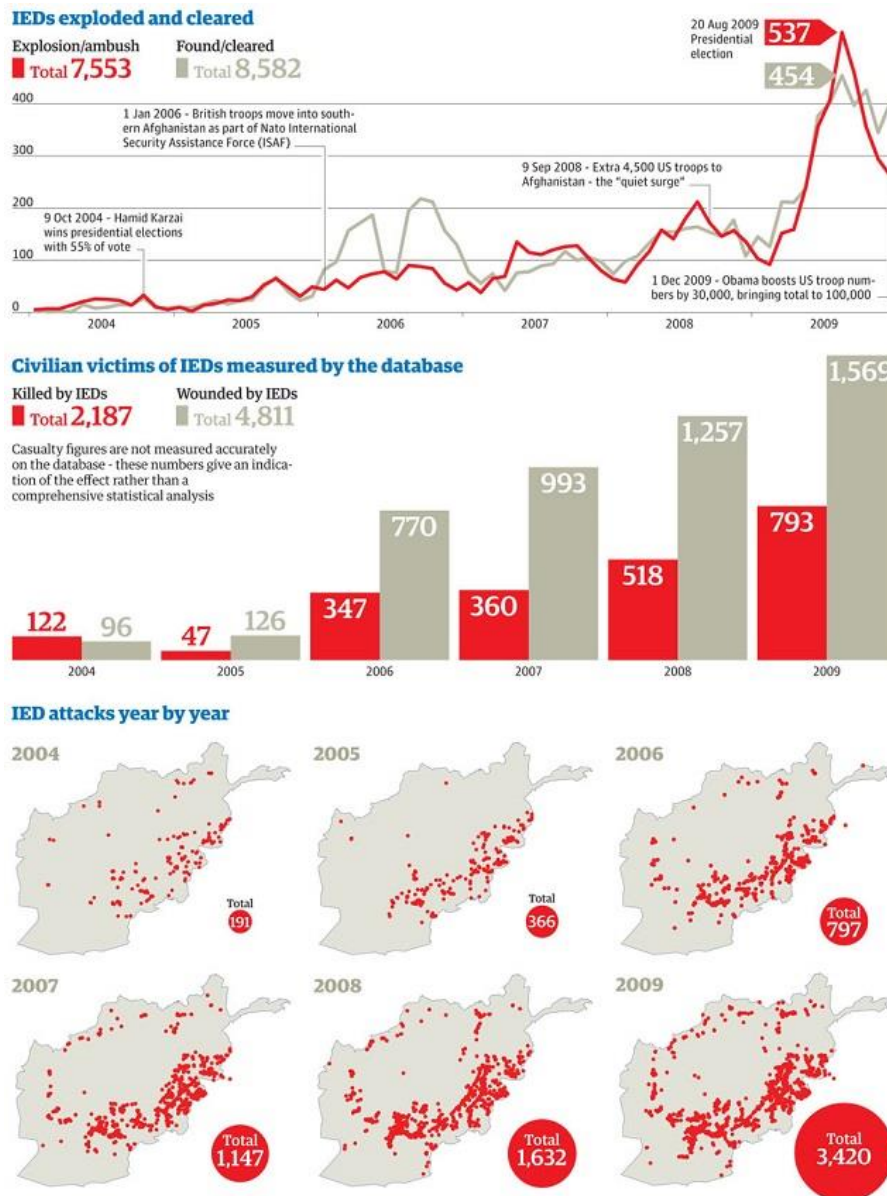
MacDonalds-Starbucks: Reprinted from [Starbucks and McDonalds] (n.d.). Copyright holder unknown. Retrieved from <http://www.princeton.edu/~ina/images/infographics/starbucks.jpg>.  
Statement: Most of the countries that export the paper for the coffee cups of Starbucks do not have a Starbucks in their own country (true).



College: Reprinted from *College in America* (n.d.). Copyright holder unknown. Retrieved from [http://blog.socrato.com/wp-content/uploads/2010/11/us\\_colleges\\_infographic.gif](http://blog.socrato.com/wp-content/uploads/2010/11/us_colleges_infographic.gif).  
Statement: The universities with the lowest tuition are suited in an area where the least universities are located (false).

## Appendix B experimental infographics

Appendix B consists of the experimental infographics and two additional infographics used in the experiment (see Appendix A for the remaining graphs of the experiment). Chapter 3 refers to this appendix.



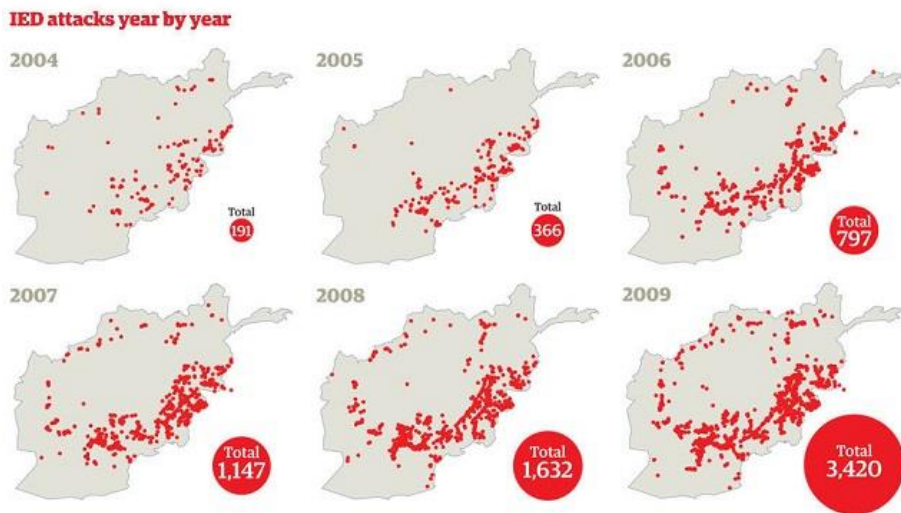
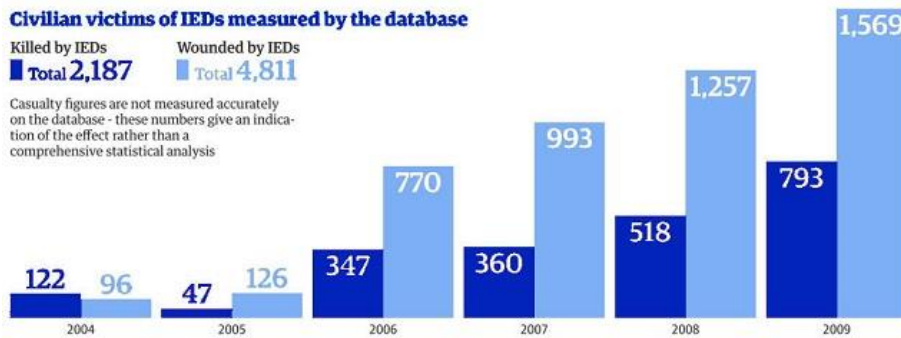
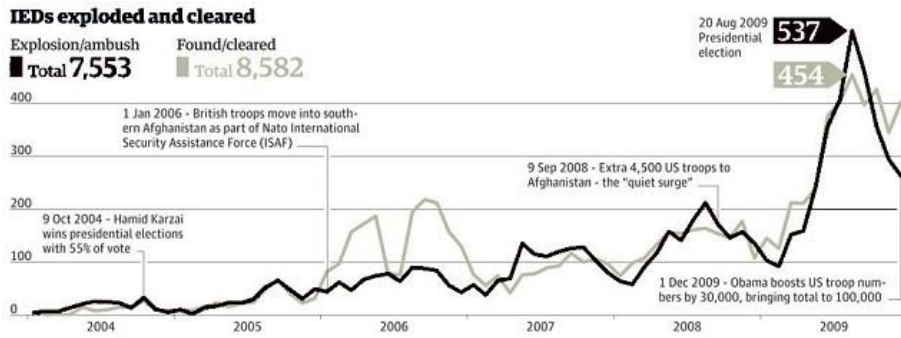
Afghanistan: Reprinted from Wikileaks` *Afghanistan war logs* (2011), by The Guardian.

Copyright 2011 by The Guardian. Retrieved from:

<http://www.theguardian.com/news/datablog/2010/jul/27/wikileaks-afghanistan-data-datajournalism>.

Condition: No color and no proximity.

Statement: Every year, except in the year 2004, IEDs (improvised explosive devices) wound more people than they kill (true).



Afghanistan: Adapted from *Wikileaks` Afghanistan war logs* (2011), by The Guardian.

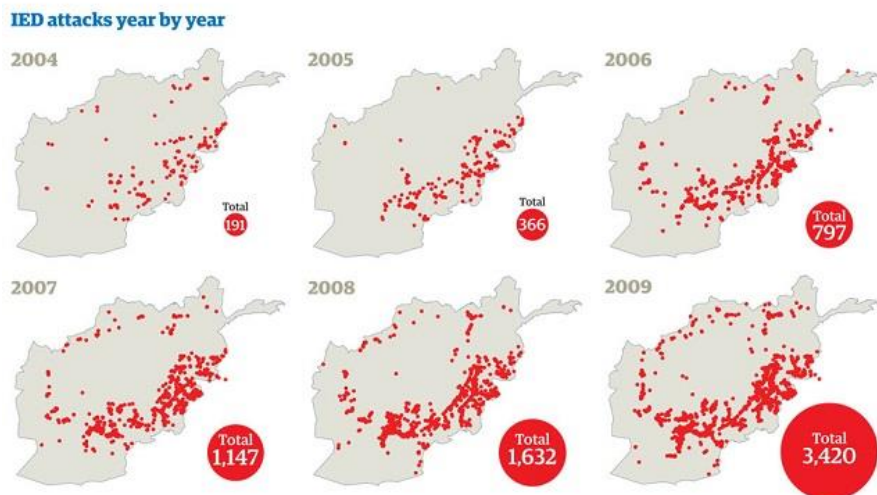
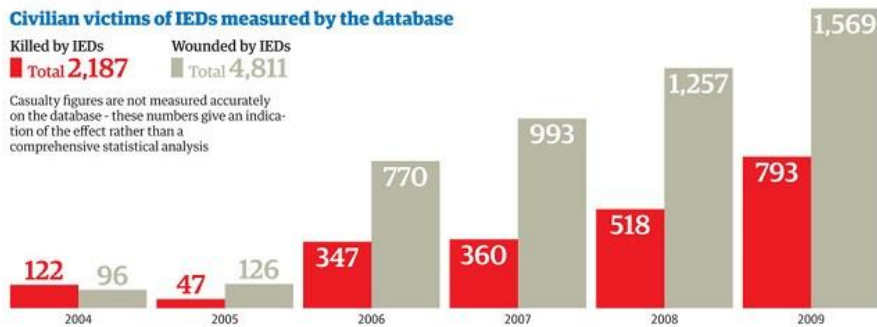
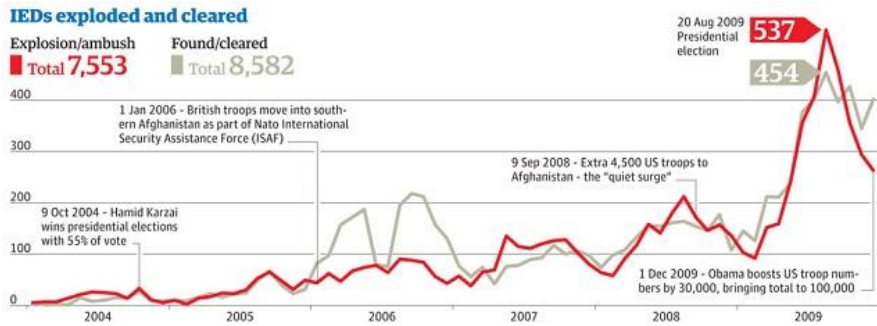
Copyright 2011 by The Guardian. Retrieved from:

<http://www.theguardian.com/news/datablog/2010/jul/27/wikileaks-afghanistan-data-datajournalism>.

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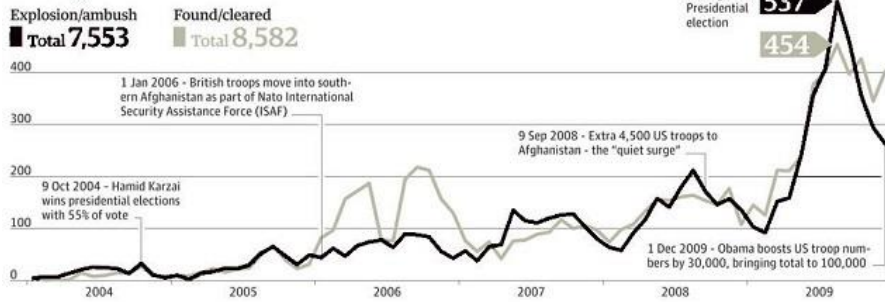
Copyright 2011 by The Guardian. Retrieved from:

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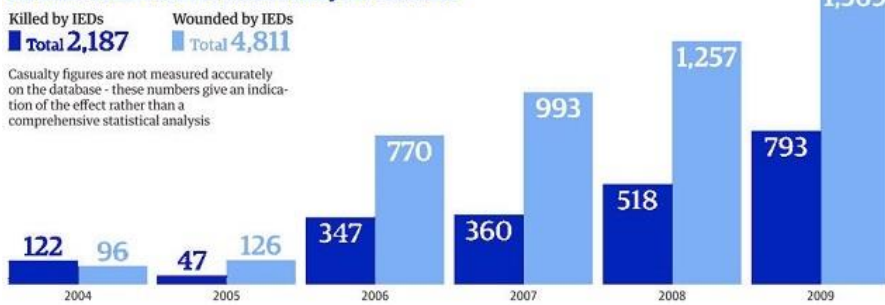
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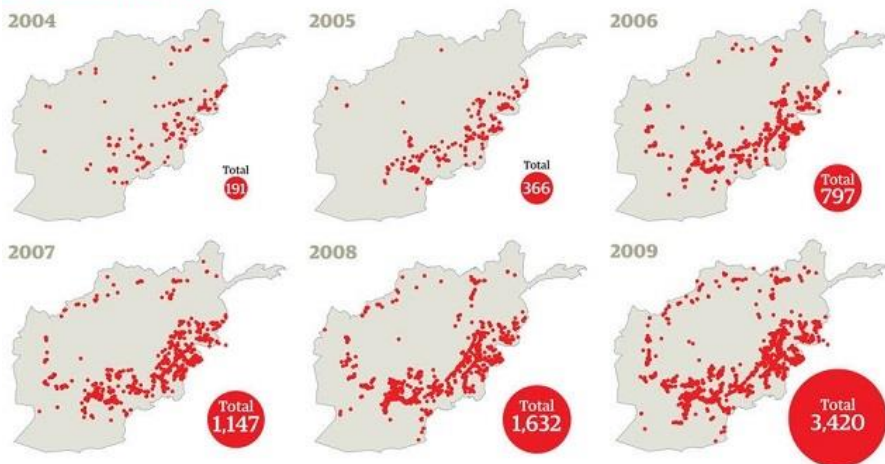
### IEDs exploded and cleared



### Civilian victims of IEDs measured by the database



### IED attacks year by year



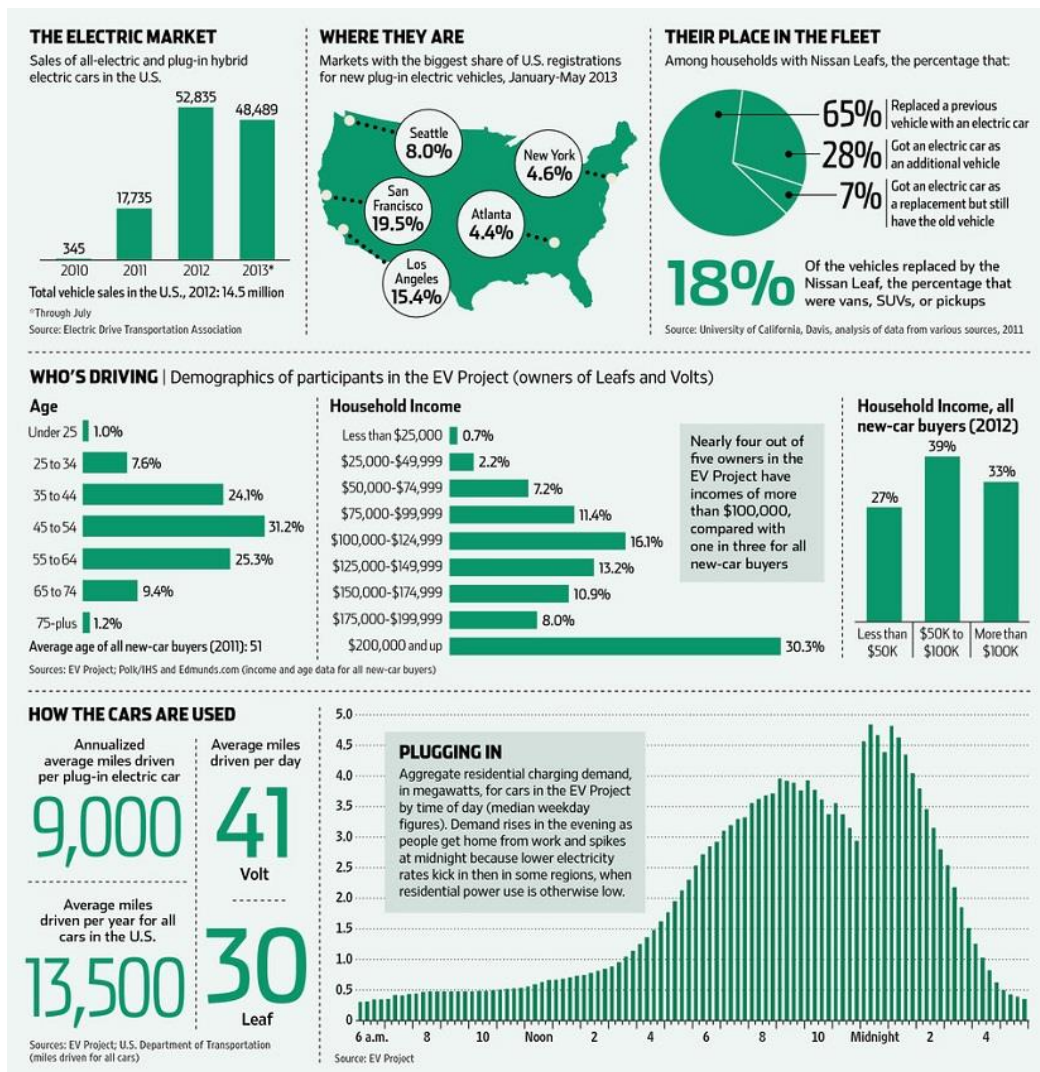
Afghanistan: Adapted from *Wikileaks` Afghanistan war logs* (2011), by The Guardian.

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<http://www.theguardian.com/news/datablog/2010/jul/27/wikileaks-afghanistan-data-datajournalism>.

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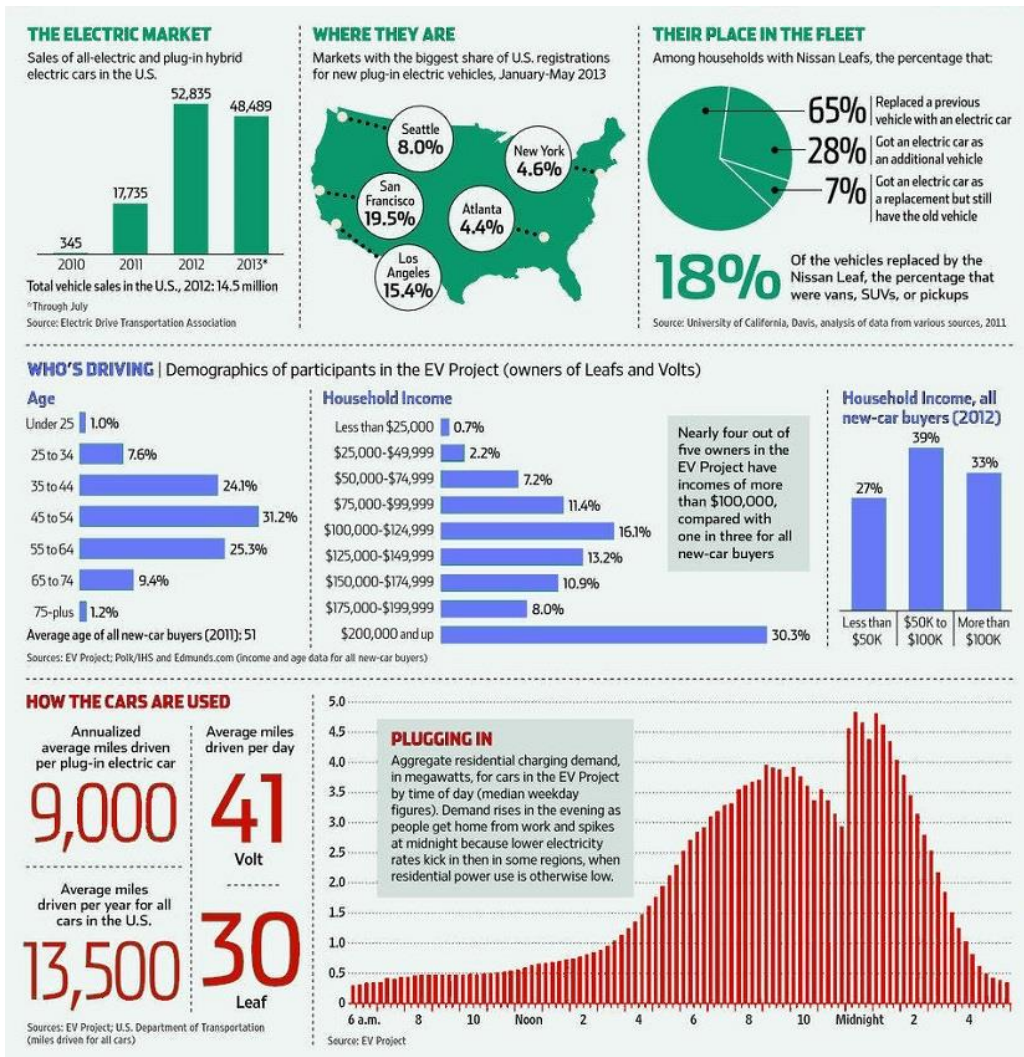
[http://online.wsj.com/news/articles/SB10001424127887324123004579055112450200336?ru=yahoo&mod=yahoo\\_hs&mg=reno64-](http://online.wsj.com/news/articles/SB10001424127887324123004579055112450200336?ru=yahoo&mod=yahoo_hs&mg=reno64-)

[wsj&url=http%3A%2F%2Fonline.wsj.com%2Farticle%2FSB10001424127887324123004579055112450200336.html%3Fru%3Dyahoo%26mod%3Dyahoo\\_hs#project=ECARSCHRTPRNT&articleTabs=interactive.](http://online.wsj.com/news/articles/SB10001424127887324123004579055112450200336.html#ru%3Dyahoo%26mod%3Dyahoo_hs#project=ECARSCHRTPRNT&articleTabs=interactive)

Condition: No color and no proximity.

Statement: Of all people who are driving an electric car, earns the largest group more than 200,000 dollar per year (true).





Car: Adapted from *Who Drives Electric Cars?* (2013), by The Wall Street Journal.

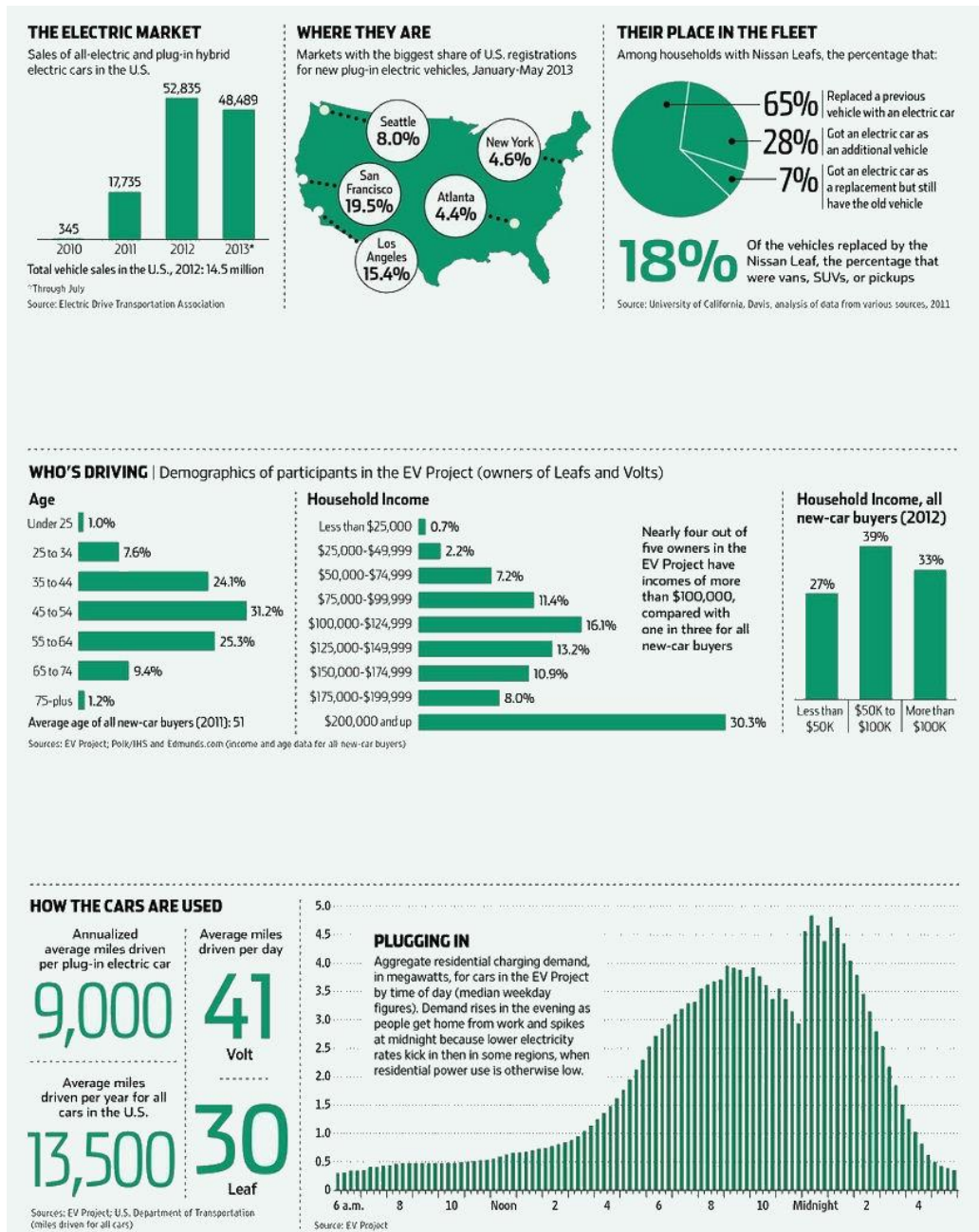
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[http://online.wsj.com/news/articles/SB10001424127887324123004579055112450200336?ru=yahoo&mod=yahoo\\_hs&mg=reno64-](http://online.wsj.com/news/articles/SB10001424127887324123004579055112450200336?ru=yahoo&mod=yahoo_hs&mg=reno64-)

[wsj&url=http%3A%2F%2Fonline.wsj.com%2Farticle%2FSB10001424127887324123004579055112450200336.html%3Fru%3Dyahoo%26mod%3Dyahoo\\_hs#project=ECARSCHRTPTNT&articleTabs=interactive.](http://online.wsj.com/news/articles/SB10001424127887324123004579055112450200336.html%3Fru%3Dyahoo%26mod%3Dyahoo_hs#project=ECARSCHRTPTNT&articleTabs=interactive)

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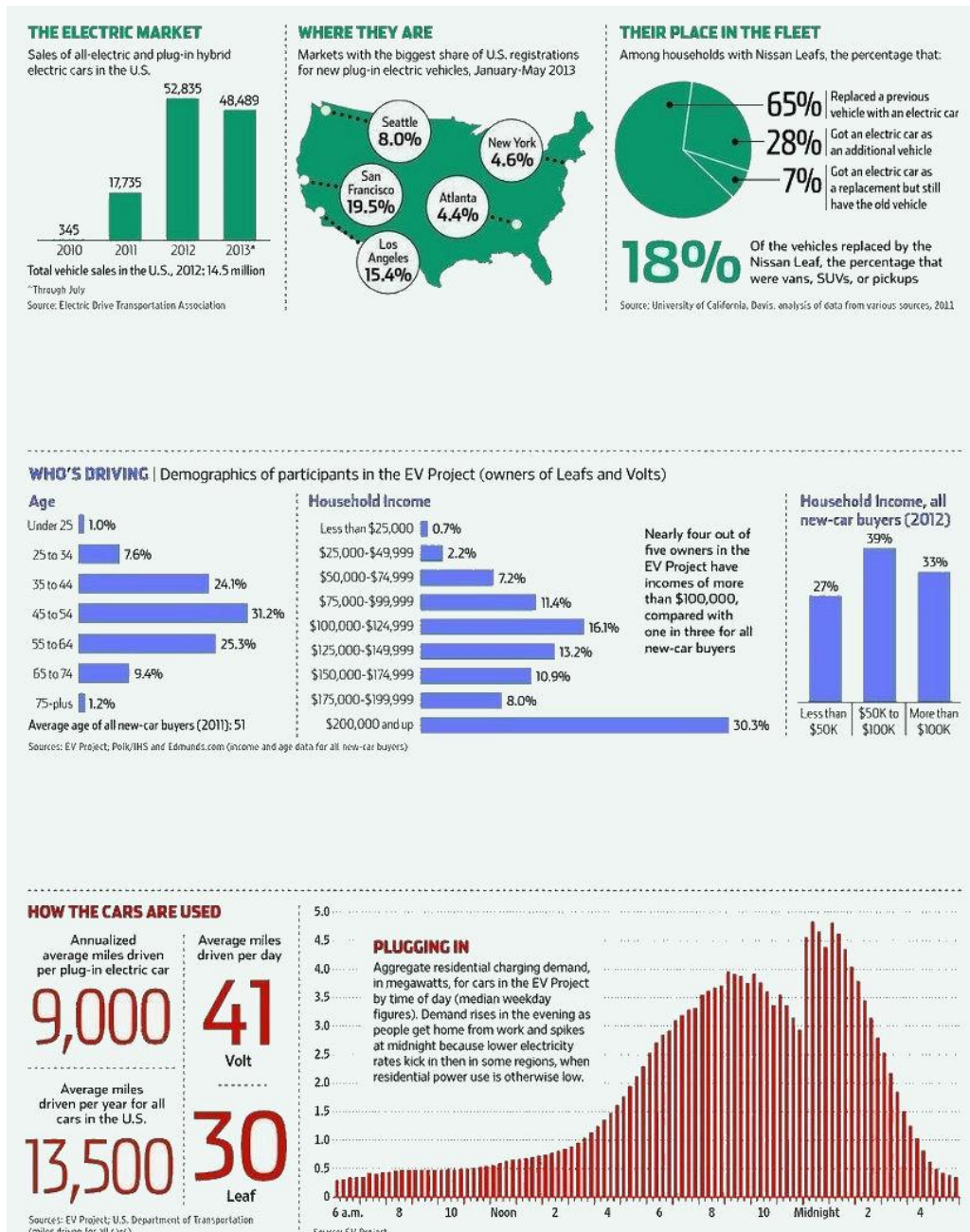


Car: Adapted from *Who Drives Electric Cars?* (2013), by The Wall Street Journal.

Copyright 2013 by The Wall Street Journal. Retrieved from [http://online.wsj.com/news/articles/SB10001424127887324123004579055112450200336?ru=yahoo&mod=yahoo\\_hs&mg=reno64-wsj&url=http%3A%2F%2Fonline.wsj.com%2Farticle%2FSB10001424127887324123004579055112450200336.html%3Fru%3Dyahoo%26mod%3Dyahoo\\_hs#project=ECARSCHRTPRNT&articleTabs=interactive](http://online.wsj.com/news/articles/SB10001424127887324123004579055112450200336?ru=yahoo&mod=yahoo_hs&mg=reno64-wsj&url=http%3A%2F%2Fonline.wsj.com%2Farticle%2FSB10001424127887324123004579055112450200336.html%3Fru%3Dyahoo%26mod%3Dyahoo_hs#project=ECARSCHRTPRNT&articleTabs=interactive).

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[wsj&url=http%3A%2F%2Fonline.wsj.com%2Farticle%2FSB10001424127887324123004579055112450200336.html%3Fru%3Dyahoo%26mod%3Dyahoo\\_hs#project=ECARSCHRTPRNT&articleTabs=interactive.](http://online.wsj.com/news/articles/SB10001424127887324123004579055112450200336.html?ru=yahoo&mod=yahoo_hs#project=ECARSCHRTPRNT&articleTabs=interactive)

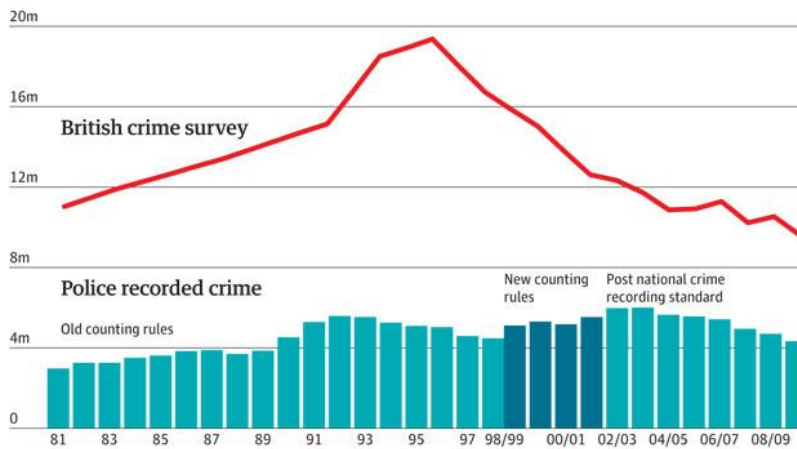
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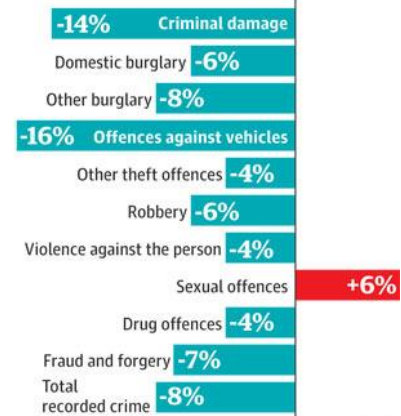
## Trends in crime levels

Number of offences, recorded crime and crime survey



## Change in recorded crimes

Change in the number of crimes recorded in 2009/10, compared with 2008/09



SOURCE: BCS

## Types of crime

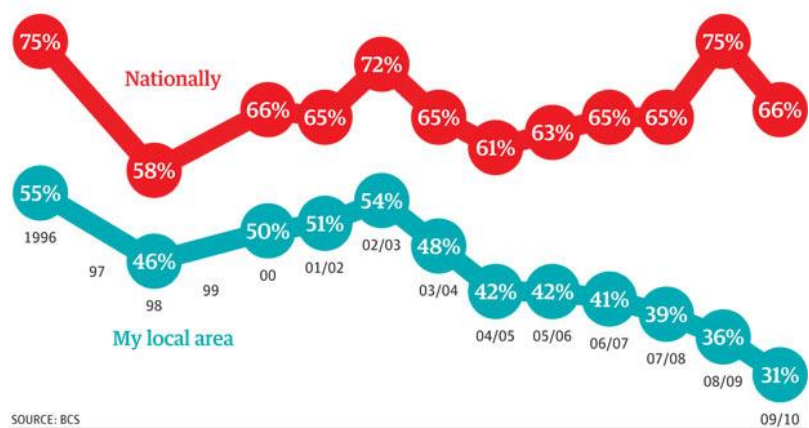
British Crime Survey estimates a total of 9.6m offences in 2009/10



\*excluding sexual offences

## People think crime is going up - except where they live

% who think crime is going up, BCS



SOURCE: BCS

Crime: Reprinted from *Crime statistics: get the rates were you life* (2010), by The Guardian.

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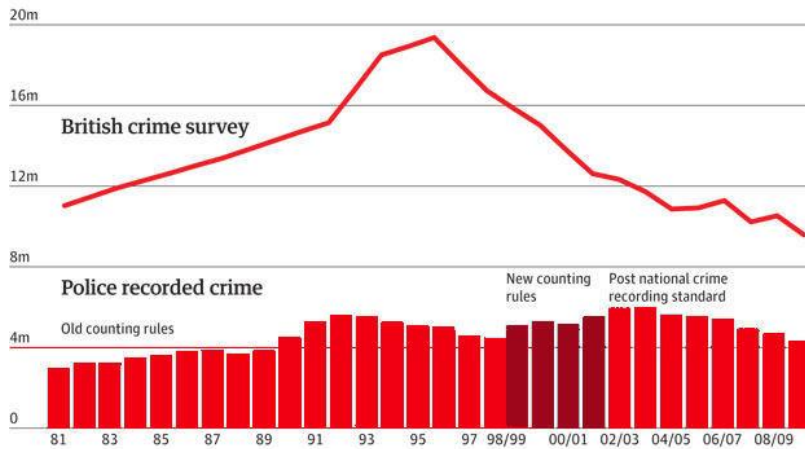
<http://www.theguardian.com/news/datablog/2010/jul/15/crime-statistics-police-force>.

Condition: No color and no proximity.

Statement: Of all types of crimes, the offences against vehicles in 2009/2010 decreased the most compared to 2008/2009 (true).

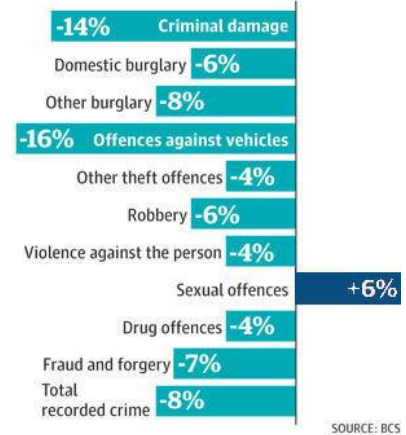
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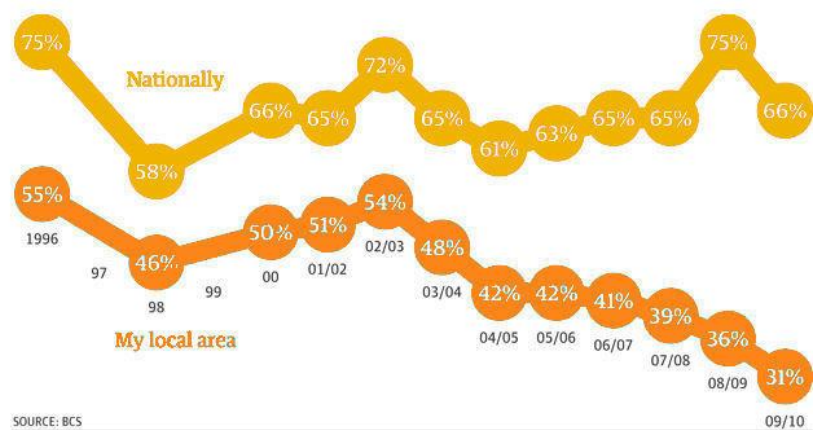
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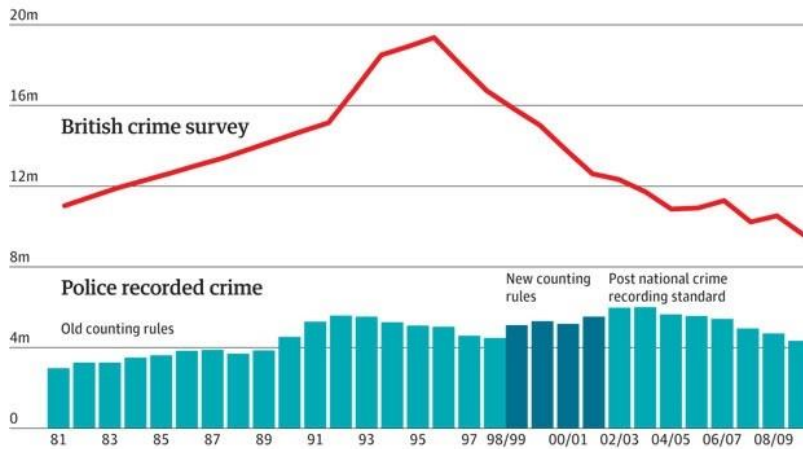
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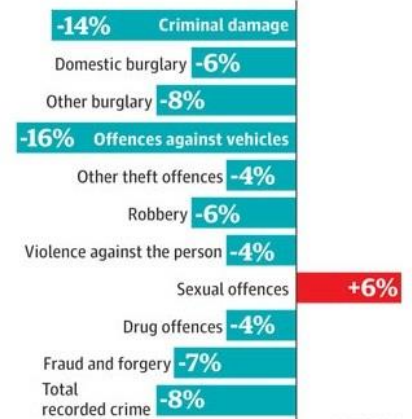
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Number of offences, recorded crime and crime survey



## Change in recorded crimes

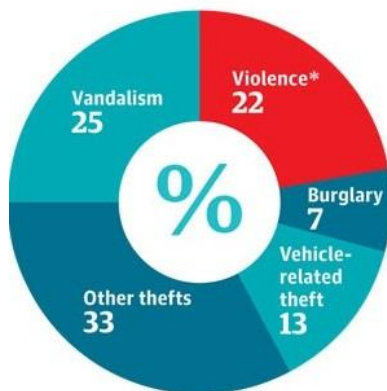
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## Types of crime

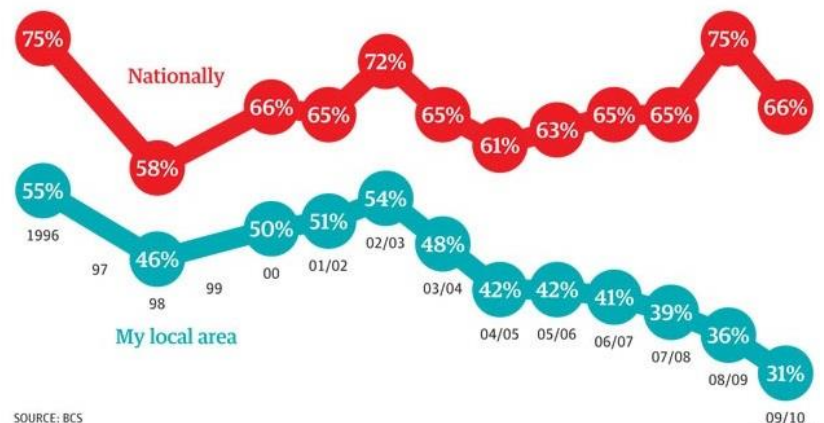
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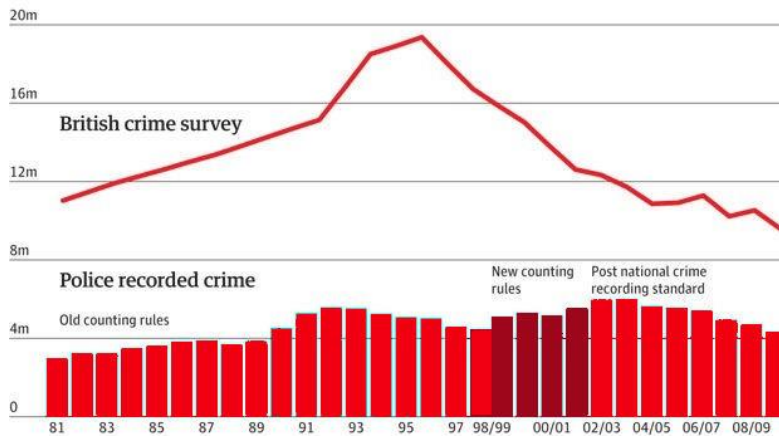
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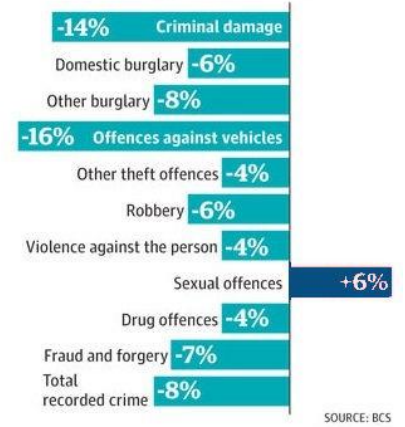
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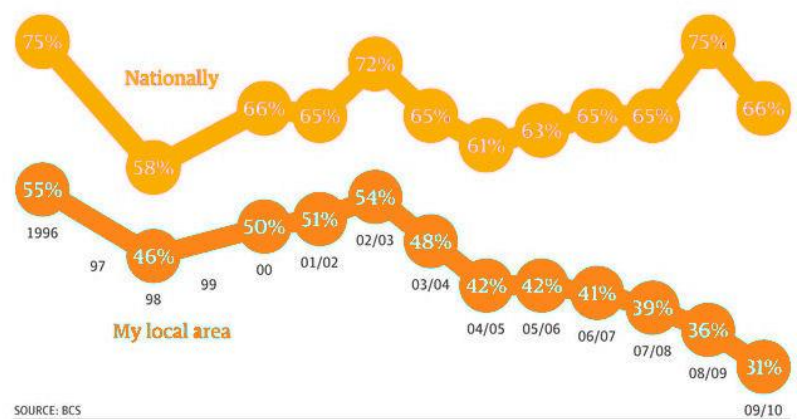
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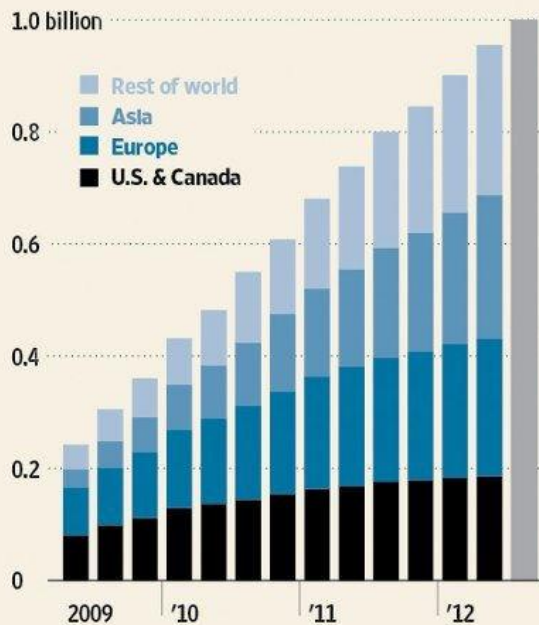
## Facebook Nation | The social network reaches 1 billion active users

If Facebook were a nation, it would have the third-largest population. But if you compared its revenue to national economies, it would rank 156th.

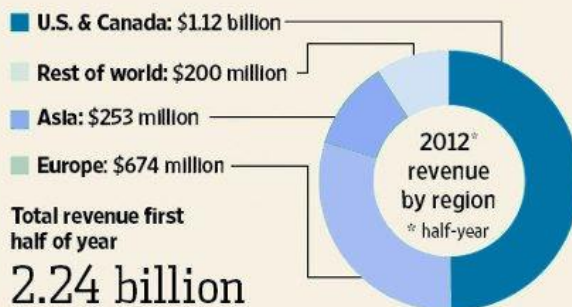
Total population, billions of people		Ranking		2011 gross domestic product, billions of dollars
1.34	China	#1	#154	Swaziland 3.98
1.24	India	#2	#155	Fiji 3.81
1 billion users*	Facebook	#3	#156	Facebook \$3.71 billion
0.31	U.S.	#4	#157	Barbados 3.69
0.24	Indonesia	#5	#158	Togo 3.59

\*monthly active users

Facebook's monthly active users by region, quarterly (current breakdown not available)



Quarterly revenue



Note: For purposes of reporting MAUs and revenue by geographic region, Europe includes Russia and Turkey, Asia includes Australia and New Zealand, and rest of world includes Africa, Latin America, and the Middle East.

Sources: The World Bank (economic and population data); the company (regional data); FactSet (quarterly revenue)

The Wall Street Journal

Facebook: Adapted from *Facebook Nation* (2012), by The Wall Street Journal. Copyright 2012 by The Wall Street Journal. Retrieved from <http://online.wsj.com/news/articles/SB10000872396390443635404578036164027386112?mod=e2fb&mg=reno64-wsj&url=http%3A%2F%2Fonline.wsj.com%2Farticle%2FSB10000872396390443635404578036164027386112.html%3Fmod%3De2fb>.

Condition: No color and no proximity.

Statement: The amount of users in the US & Canada and Europe increase less rapidly than in Asia or the rest of the world (true).

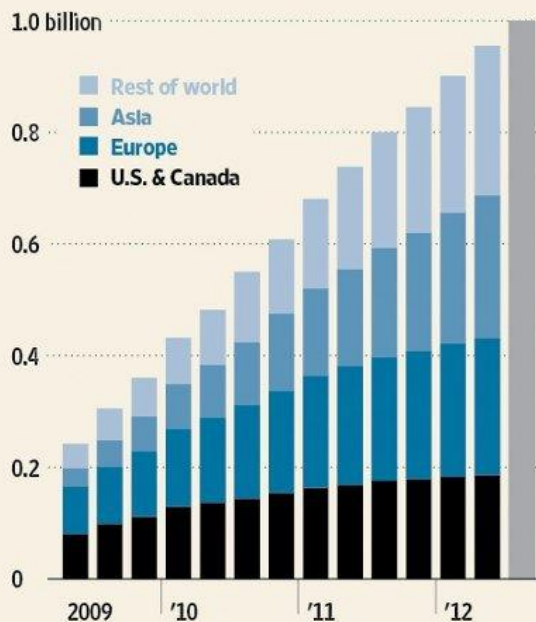


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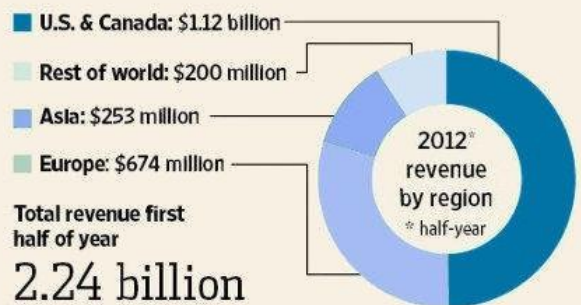
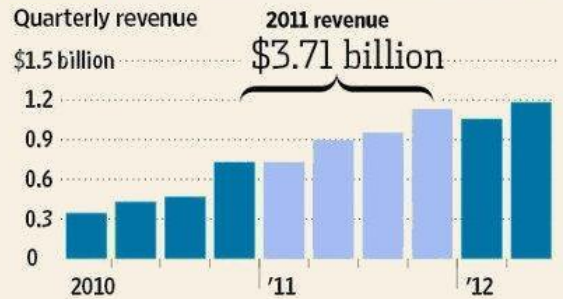
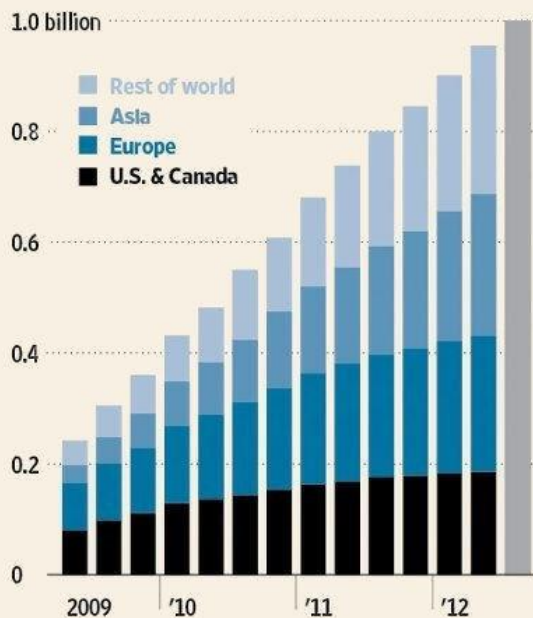
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Facebook: Adapted from *Facebook Nation* (2012), by The Wall Street Journal. Copyright 2012 by The Wall Street Journal. Retrieved from <http://online.wsj.com/news/articles/SB10000872396390443635404578036164027386112?mod=e2fb&mg=reno64-wsj&url=http%3A%2F%2Fonline.wsj.com%2Farticle%2FSB10000872396390443635404578036164027386112.html%3Fmod%3De2fb>.

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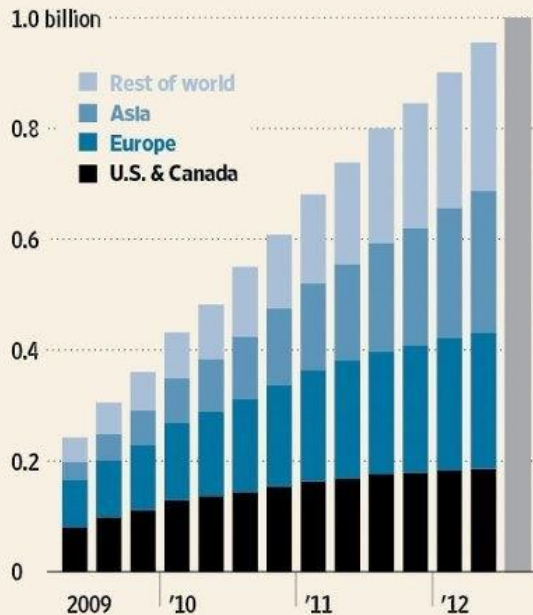
Total population, billions of people

Ranking

2011 gross domestic product, billions of dollars

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2011 revenue



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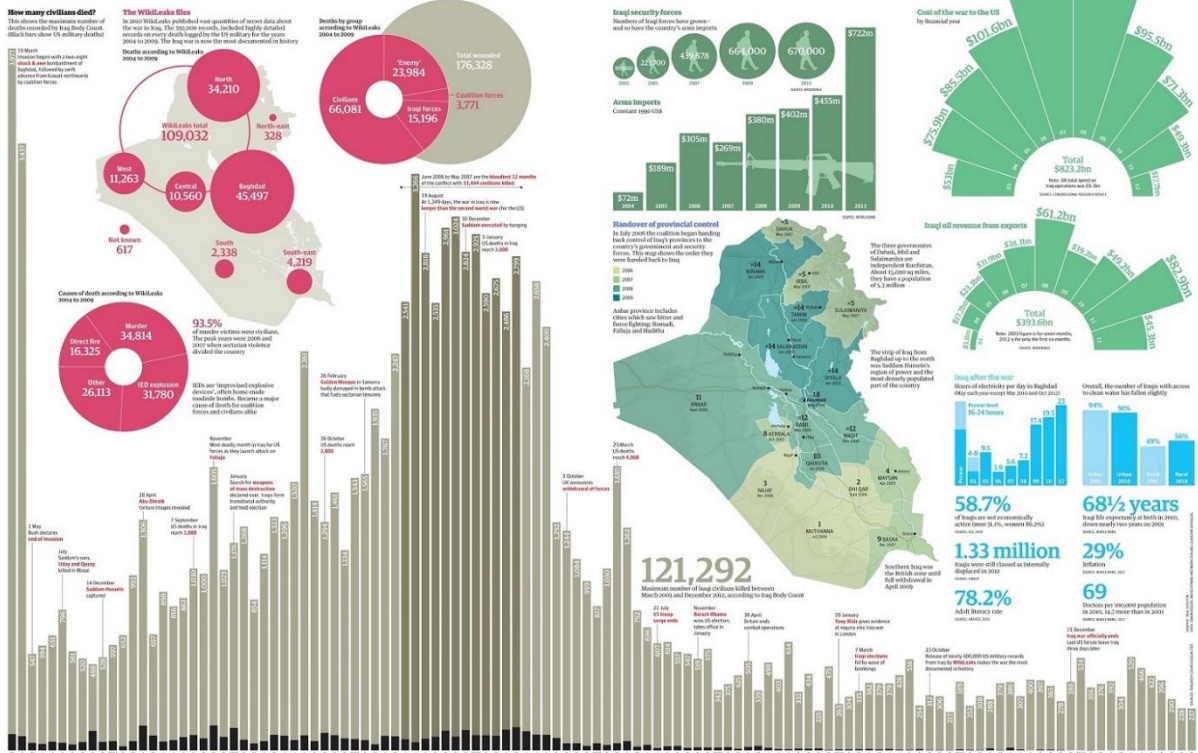
Facebook: Adapted from *Facebook Nation* (2012), by The Wall Street Journal. Copyright 2012 by The Wall Street Journal. Retrieved from <http://online.wsj.com/news/articles/SB10000872396390443635404578036164027386112?mod=e2fb&mg=reno64-wsj&url=http%3A%2F%2Fonline.wsj.com%2Farticle%2FSB10000872396390443635404578036164027386112.html%3Fmod%3De2fb>.

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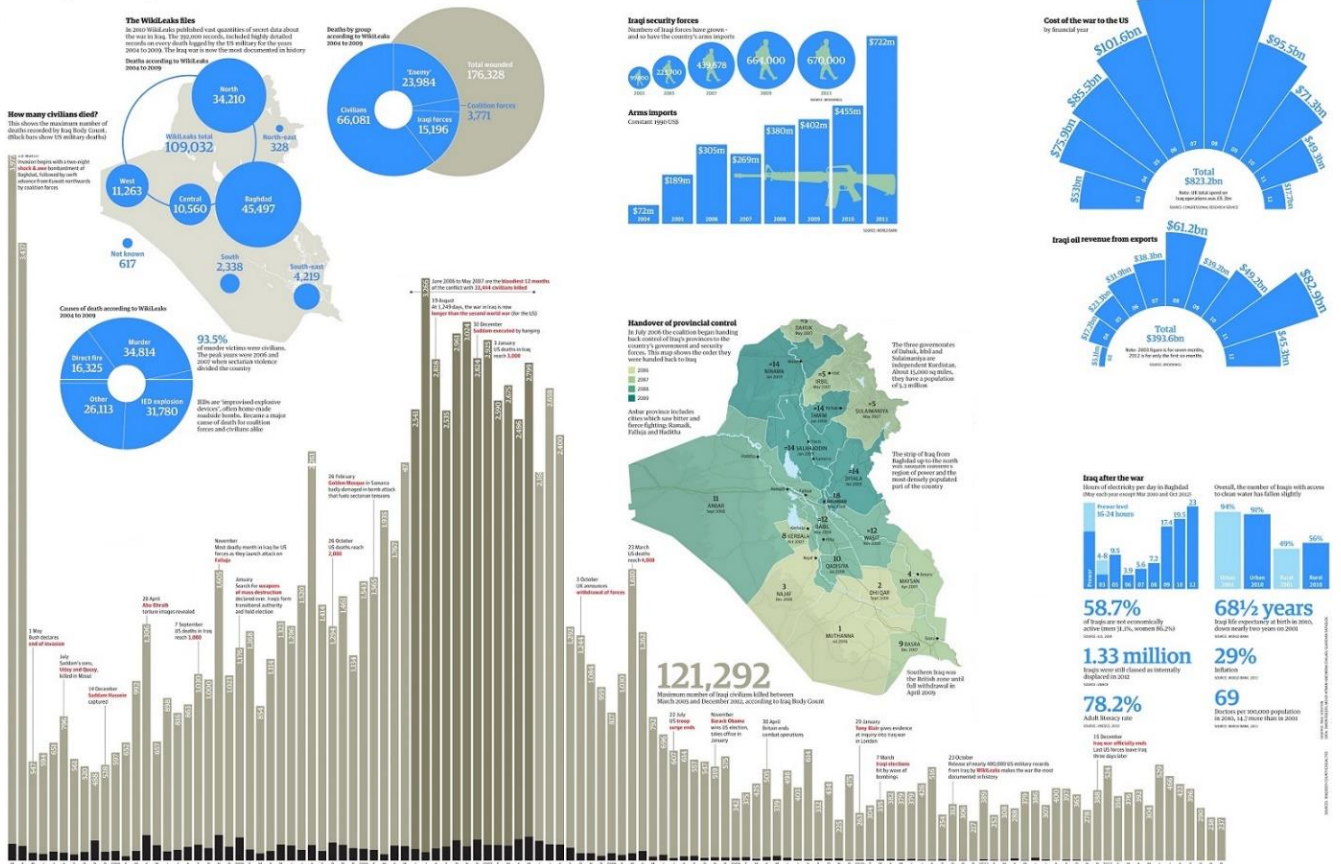
# Iraq: ten years since the invasion



Iraq: Reprinted from *Iraq: Ten Years since the Invasion* (2013), by The Guardian. Copyright 2013 by The Guardian. Retrieved from <http://www.theguardian.com/news/datablog/2013/mar/14/iraq-ten-years-visualised>. Condition: Color and no proximity. Statement: The arm import is higher in 2010 than in 2009 (true).

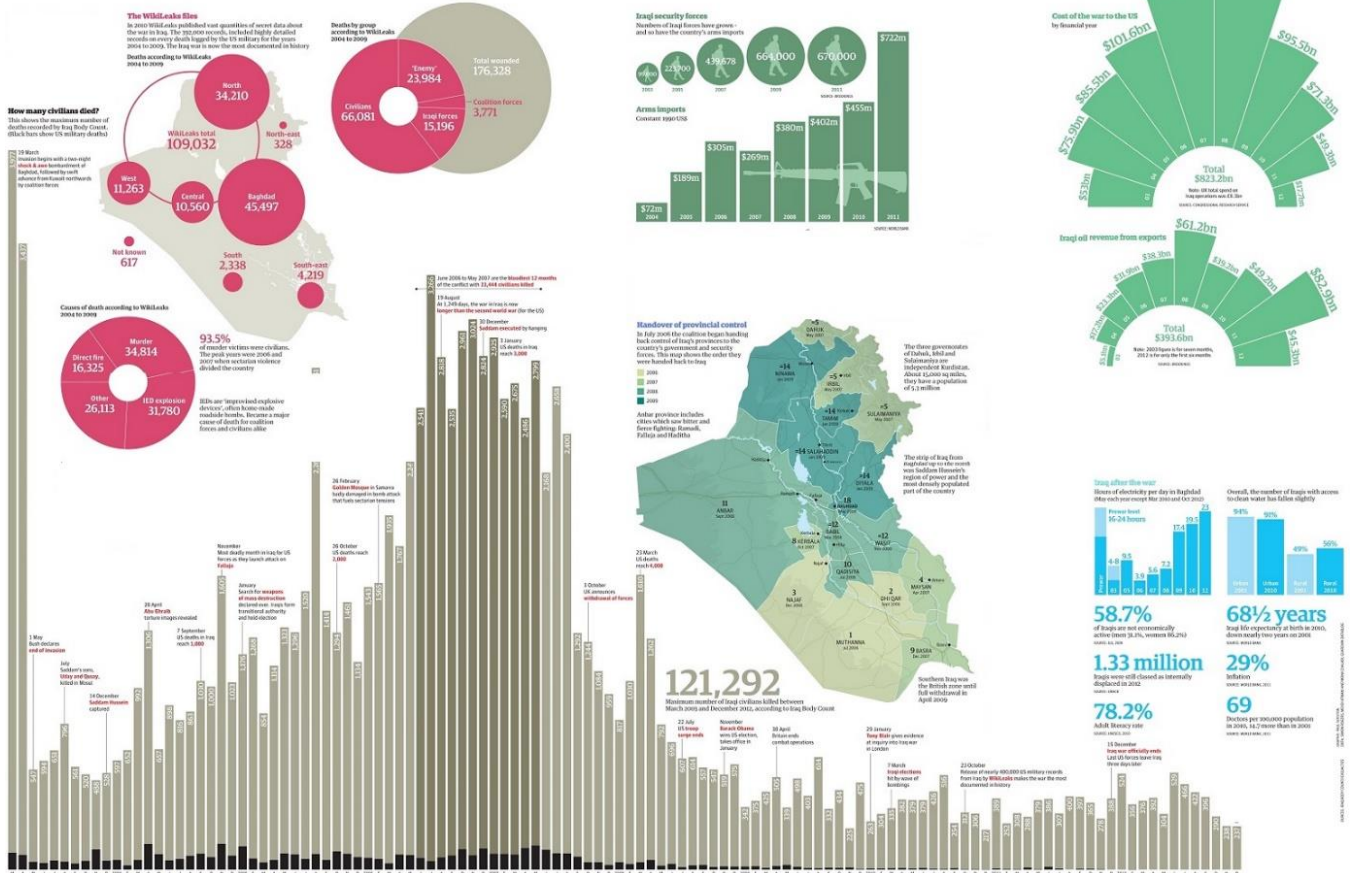


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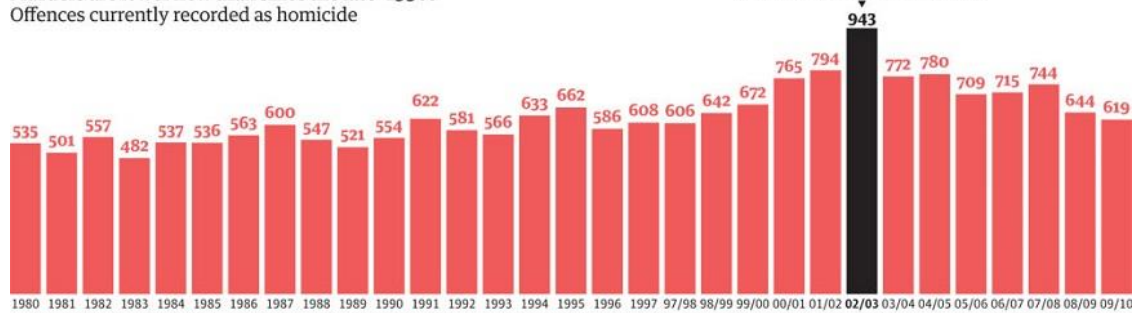


Iraq: Adapted from *Iraq: Ten Years since the Invasion* (2013), by The Guardian. Copyright 2013 by The Guardian. Retrieved from <http://www.theguardian.com/news/datablog/2013/mar/14/iraq-ten-years-visualised>. Condition: Color and proximity. Statement: The arm import is higher in 2010 than in 2009 (true).

## Murder Deaths in numbers

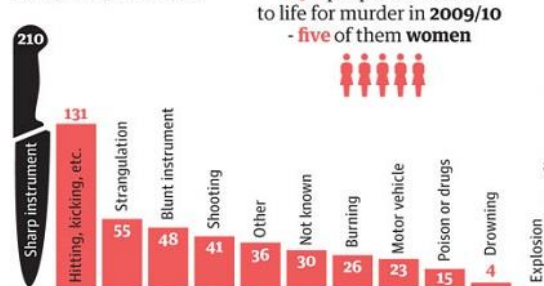
**Murders are lower now than since the late -1990s**  
Offences currently recorded as homicide

Includes 218 Harold Shpman murders



### How do victims die?

2009/2010, both sexes

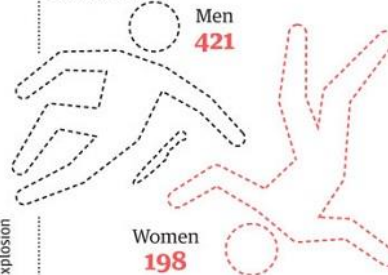


**146** people sentenced to life for murder in **2009/10** - **five** of them **women**



### The gender gap

2009/2010



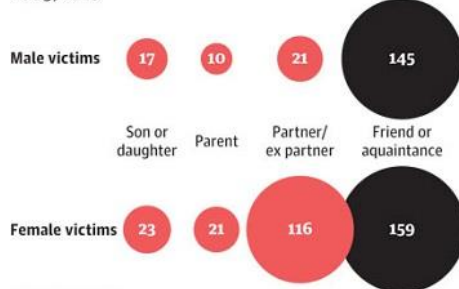
### Who murders?

2009/2010



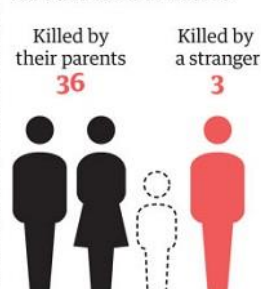
### How do murder victims know their murderer?

2009/2010



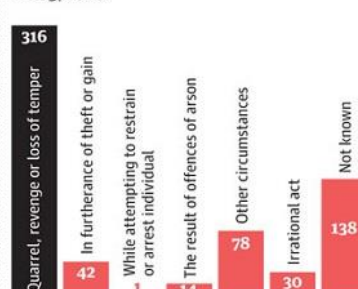
### Child murders

Male and female under 16



### Why do murders happen?

2009/2010



SOURCE: HOME OFFICE

Murder: Reprinted from *Murder Deaths in Numbers* (2011) by The Guardian. Copyright 2011 by The Guardian. Retrieved from <http://www.theguardian.com/news/datablog/2011/jan/20/ukcrime-criminal-justice>.

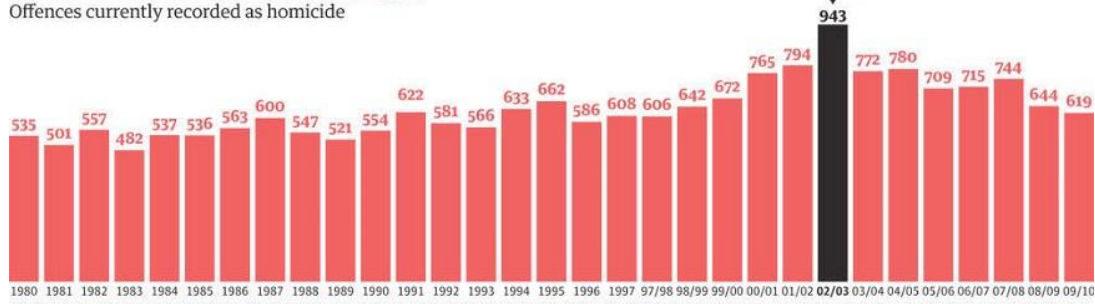
Condition: No color and no proximity.

Statement: Most of the murders are friends with or know the victim (true).

## Murder Deaths in numbers

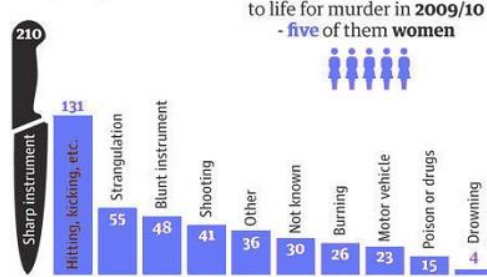
Murders are lower now than since the late -1990s  
Offences currently recorded as homicide

Includes 218 Harold Shpman murders



### How do victims die?

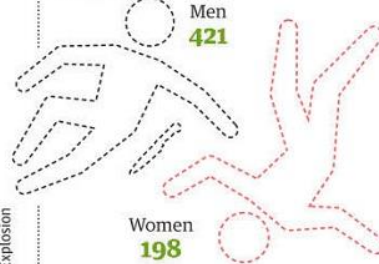
2009/2010, both sexes



146 people sentenced to life for murder in 2009/10 - five of them women

### The gender gap

2009/2010



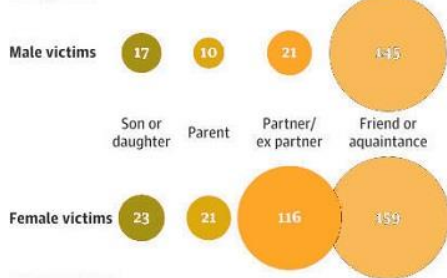
### Who murders?

2009/2010



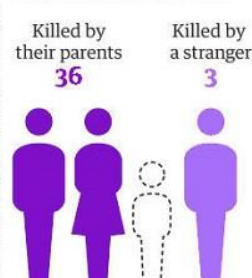
### How do murder victims know their murderer?

2009/2010



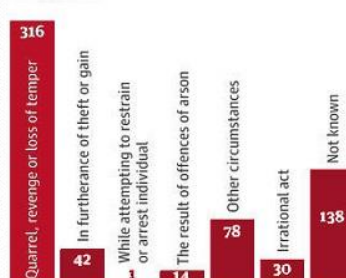
### Child murders

Male and female under 16



### Why do murders happen?

2009/2010



SOURCE: HOME OFFICE

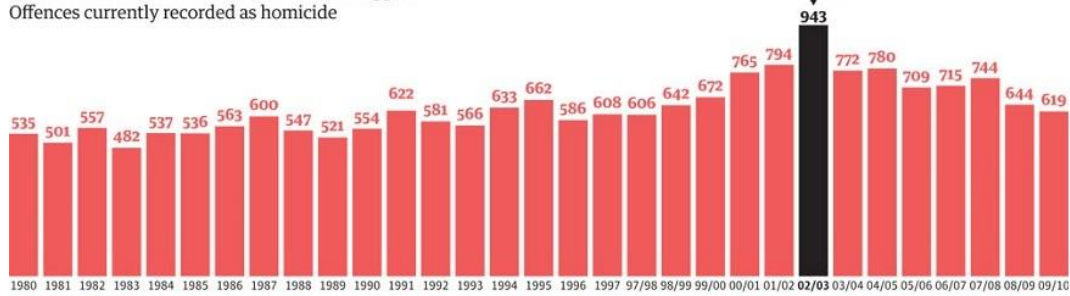
Murder: Adapted from *Murder Deaths in Numbers* (2011) by The Guardian. Copyright 2011 by The Guardian. Retrieved from <http://www.theguardian.com/news/datablog/2011/jan/20/ukcrime-criminal-justice>.  
Condition: Color and no proximity.  
Statement: Most of the murders are friends with or know the victim (true).



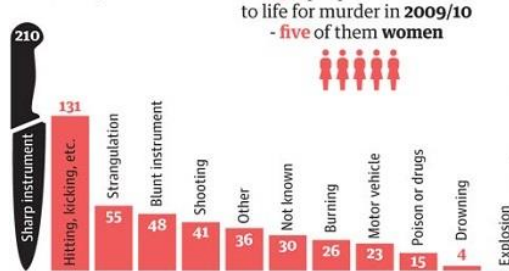
## Murder Deaths in numbers

**Murders are lower now than since the late -1990s**  
Offences currently recorded as homicide

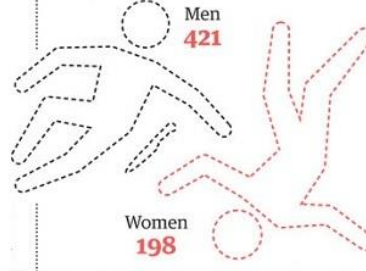
Includes 218 Harold Shipman murders



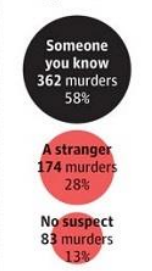
**How do victims die?**  
2009/2010, both sexes



**The gender gap**  
2009/2010



**Who murders?**  
2009/2010

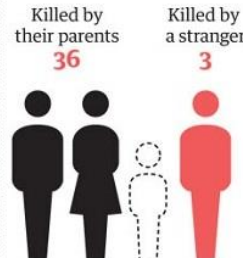


**How do murder victims know their murderer?**  
2009/2010



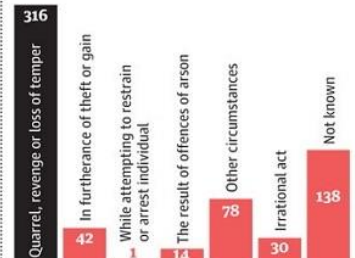
**Child murders**

Male and female under 16



**Why do murders happen?**

2009/2010



Murder: Adapted from *Murder Deaths in Numbers* (2011) by The Guardian. Copyright 2011

by The Guardian. Retrieved from

<http://www.theguardian.com/news/datablog/2011/jan/20/ukcrime-criminal-justice>.

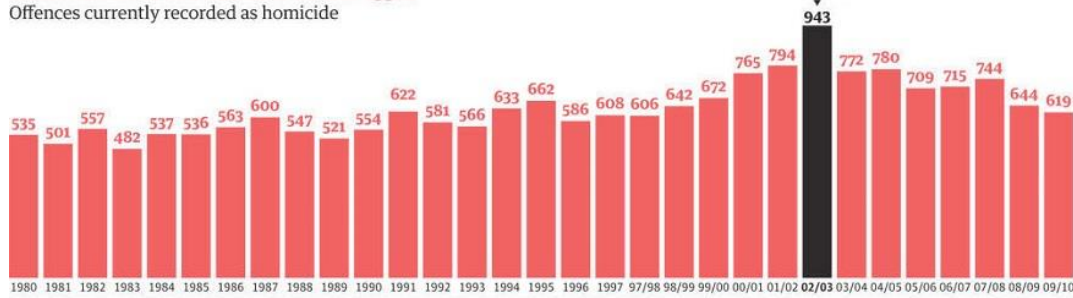
Condition: No color and proximity.

Statement: Most of the murders are friends with or know the victim (true).

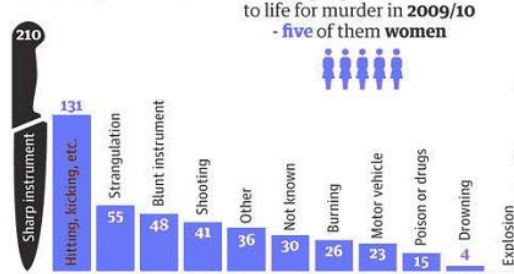
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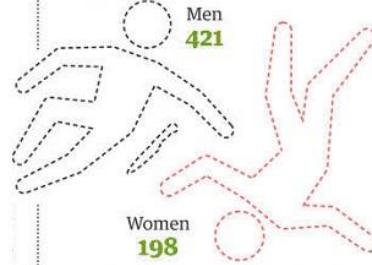
Includes 218 Harold Shpman murders



### How do victims die? 2009/2010, both sexes



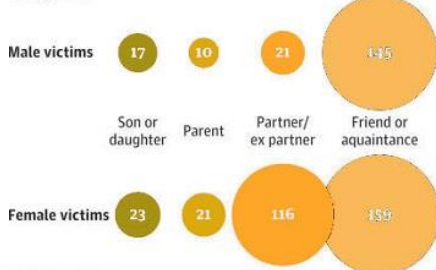
### The gender gap 2009/2010



### Who murders? 2009/2010

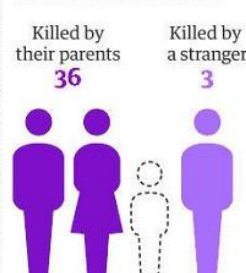


### How do murder victims know their murderer? 2009/2010

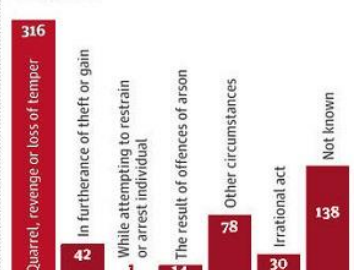


SOURCE: HOME OFFICE

### Child murders Male and female under 16



### Why do murders happen? 2009/2010



Murder: Adapted from *Murder Deaths in Numbers* (2011) by The Guardian. Copyright 2011

by The Guardian. Retrieved from

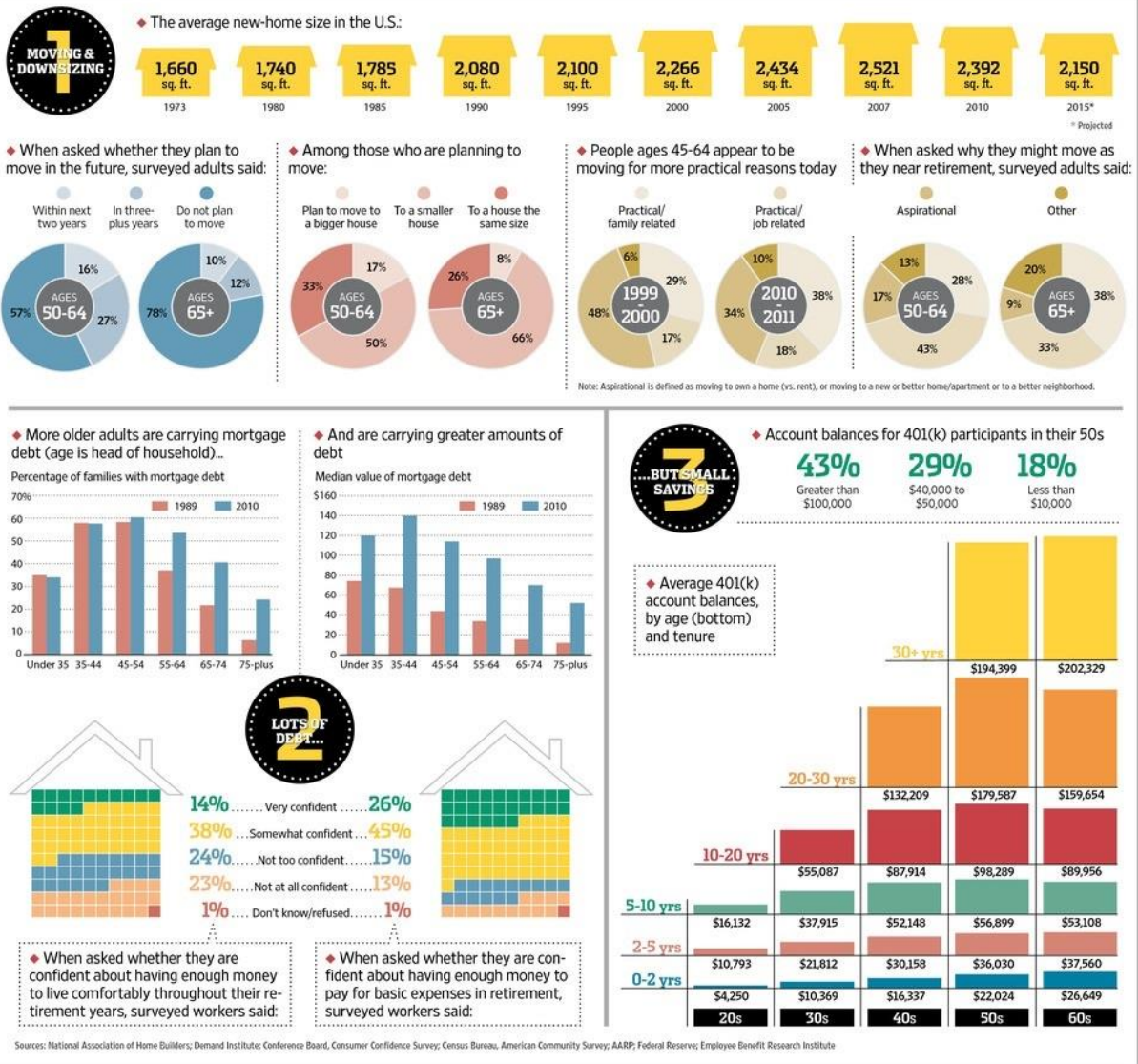
<http://www.theguardian.com/news/datablog/2011/jan/20/ukcrime-criminal-justice>.

Condition: Color and proximity.

Statement: Most of the murders are friends with or know the victim (true).

# The Incredible Shrinking Retirement

Why downsizing to a smaller home often looks like an appealing way to shore up your nest egg

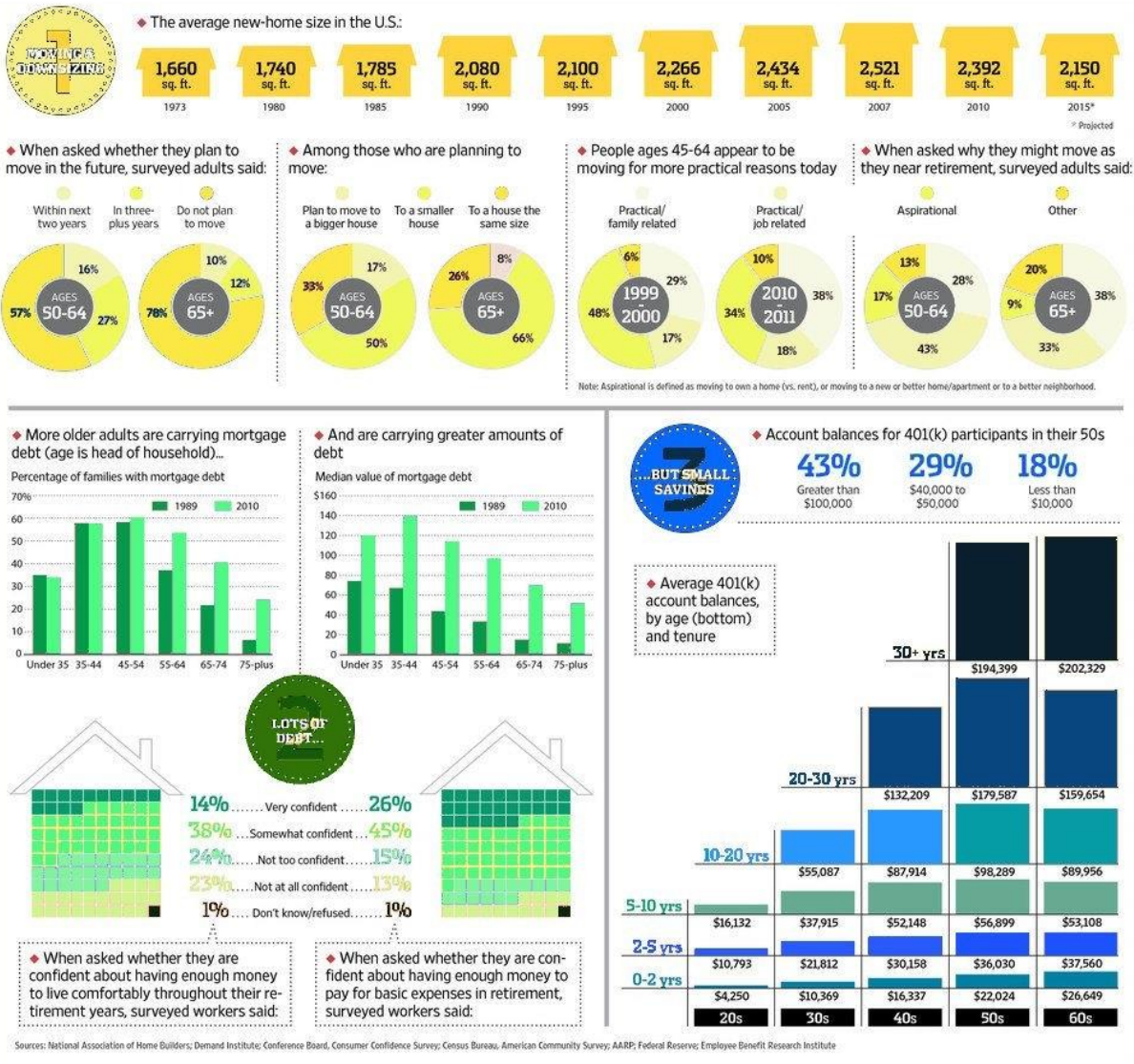


Retirement: Reprinted from *The Incredible Shrinking Retirement* (2012), by The Wall Street Journal. Copyright 2012 by The Wall Street Journal. Retrieved from <http://online.wsj.com/news/articles/SB10001424052970204005004578080561639348492>. Condition: No color and no proximity. Statement: More than half of the respondents has confidence in having enough money during their retirement (true).



# The Incredible Shrinking Retirement

Why downsizing to a smaller home often looks like an appealing way to shore up your nest egg



Retirement: Adapted from *The Incredible Shrinking Retirement* (2012), by The Wall Street Journal. Copyright 2012 by The Wall Street Journal. Retrieved from <http://online.wsj.com/news/articles/SB10001424052970204005004578080561639348492>.

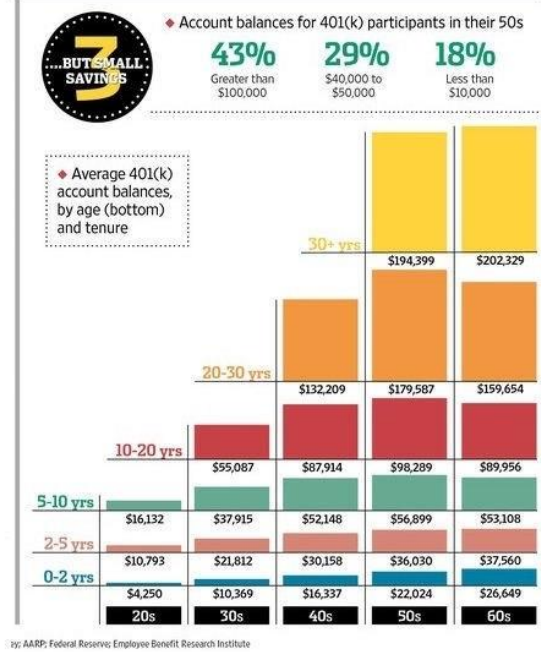
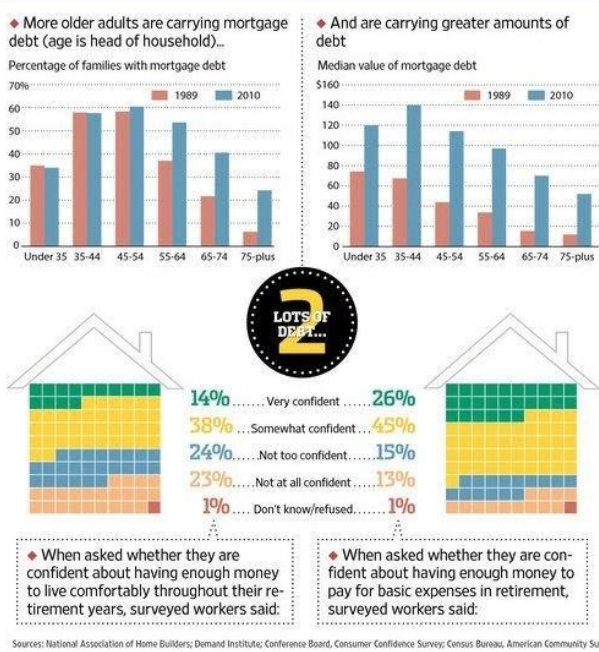
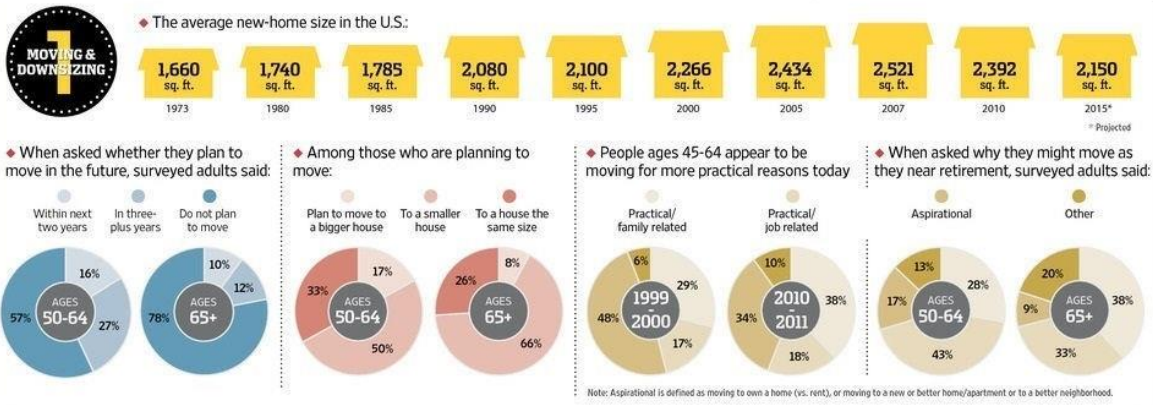
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# The Incredible Shrinking Retirement |

Why downsizing to a smaller home often looks like an appealing way to shore up your nest egg



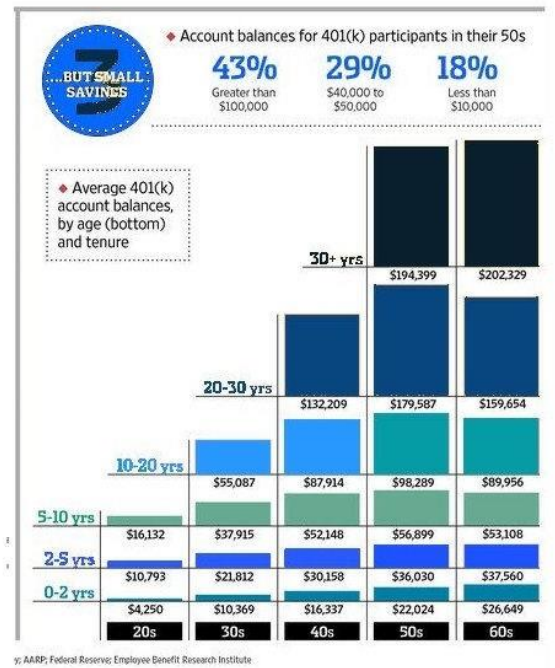
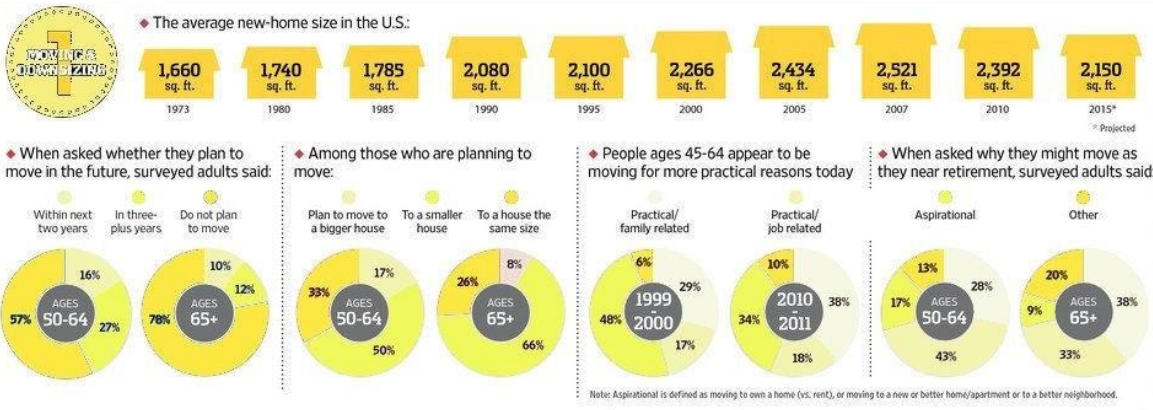
Retirement: Adapted from *The Incredible Shrinking Retirement* (2012), by The Wall Street Journal. Copyright 2012 by The Wall Street Journal. Retrieved from <http://online.wsj.com/news/articles/SB10001424052970204005004578080561639348492>.

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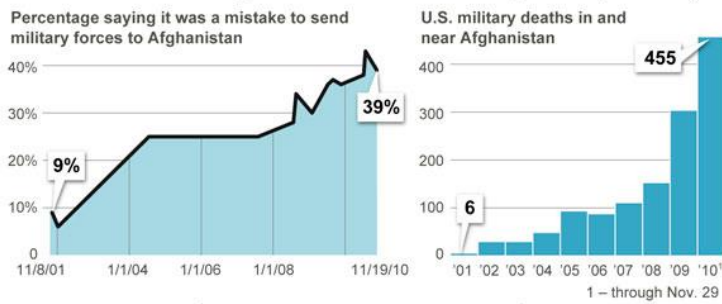


Retirement: Adapted from *The Incredible Shrinking Retirement* (2012), by The Wall Street Journal. Copyright 2012 by The Wall Street Journal. Retrieved from <http://online.wsj.com/news/articles/SB10001424052970204005004578080561639348492>.

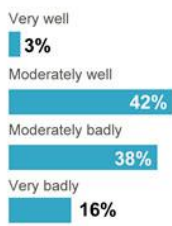
Condition: Color and proximity.

Statement: More than half of the respondents has confidence in having enough money during their retirement (true).

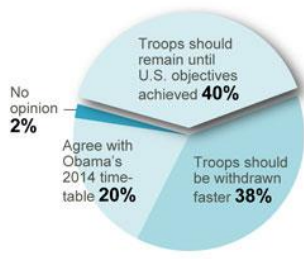
## Americans view Afghanistan war with skepticism, anxiety



How would you say things are going for the U.S. in Afghanistan?



On President Obama's plan to complete the withdrawal of U.S. combat troops by the end of 2014



How worried are you that the costs of the war in Afghanistan will make it more difficult for the government to address the problems facing the United States at home?



Sources: USA TODAY/Gallup Poll of 1,037 adults by land line and cell-phone Nov. 19-21. Margin of error: ±4 percentage points. Defense Department; USA TODAY research.

By Julie Snider, USA TODAY

War: Reprinted from *Americans view Afghanistan war with scepticism, anxiety* (2010), by the USA Today. Copyright 2010 by the USA Today. Retrieved from [http://usatoday30.usatoday.com/news/washington/2010-12-01-1Awar01\\_CV\\_N.htm?csp=34](http://usatoday30.usatoday.com/news/washington/2010-12-01-1Awar01_CV_N.htm?csp=34). Condition: No color and no proximity. Statement: Most of the Americans think that the mission in Afghanistan is going bad (true).

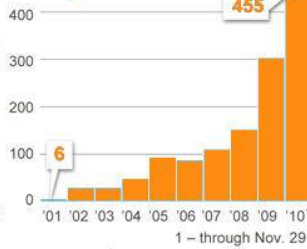


## Americans view Afghanistan war with skepticism, anxiety

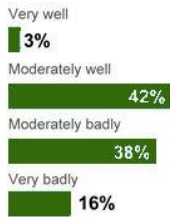
Percentage saying it was a mistake to send military forces to Afghanistan



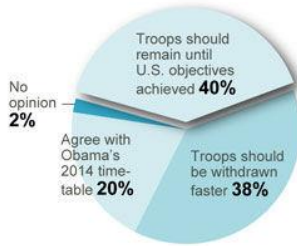
U.S. military deaths in and near Afghanistan



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On President Obama's plan to complete the withdrawal of U.S. combat troops by the end of 2014.



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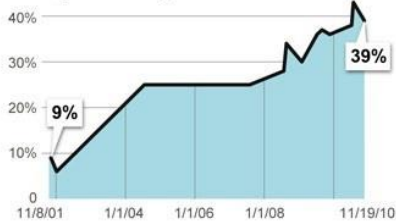
Sources: USA TODAY/Gallup Poll of 1,037 adults by land line and cell-phone Nov. 19-21. Margin of error: ±4 percentage points. Defense Department; USA TODAY research.

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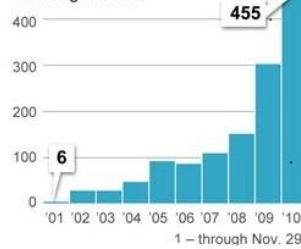
War: Adapted from *Americans view Afghanistan war with scepticism, anxiety* (2010), by the USA Today. Copyright 2010 by the USA Today. Retrieved from [http://usatoday30.usatoday.com/news/washington/2010-12-01-1Awar01\\_CV\\_N.htm?csp=34](http://usatoday30.usatoday.com/news/washington/2010-12-01-1Awar01_CV_N.htm?csp=34). Condition: Color and no proximity. Statement: Most of the Americans think that the mission in Afghanistan is going bad (true).

## Americans view Afghanistan war with skepticism, anxiety

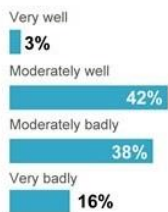
Percentage saying it was a mistake to send military forces to Afghanistan



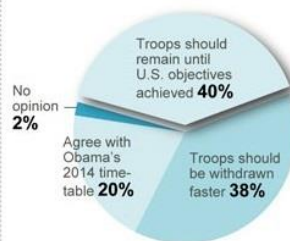
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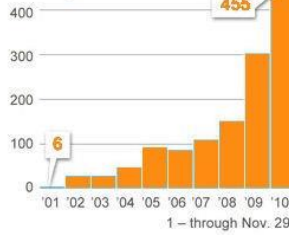
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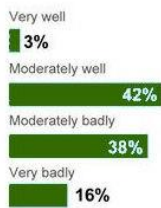
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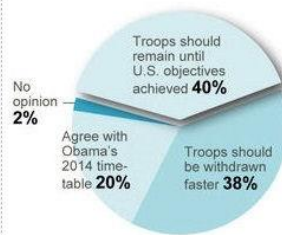
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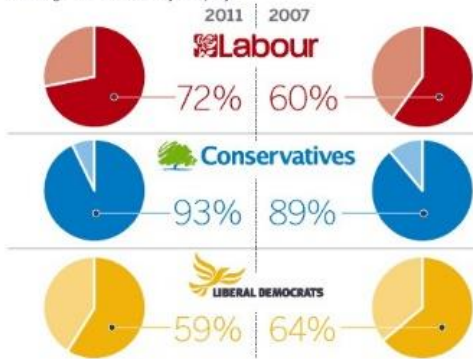
Statement: Most of the Americans think that the mission in Afghanistan is going bad (true).



## Sign of a shifting political landscape

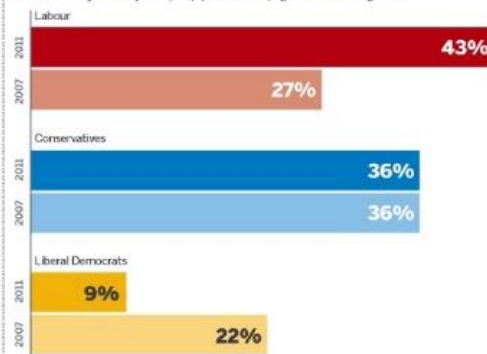
The Tories are fielding more candidates than Labour or the Liberal Democrats...

Percentage of seats contested by each party

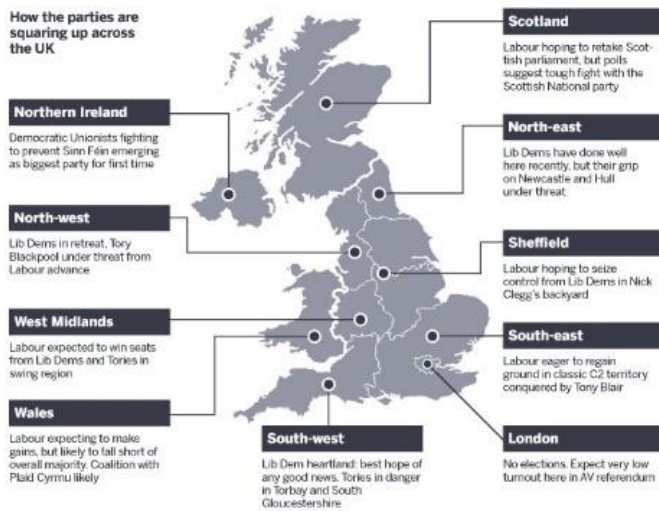


...but Labour is expecting big gains after a year of painful decisions by the coalition

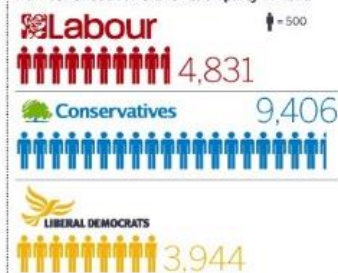
YouGov/Sunday Times opinion poll (April 8-10 2011) against 2007 voting share



How the parties are squaring up across the UK



Number of councillors for each party in 2010



What's on offer

All or a third of the seats are being contested in:

- 36 metropolitan councils
- 49 unitary councils
- 194 district councils
- 5 mayoralties:
  - Bedford
  - Middlesbrough
  - Torbay
  - Mansfield
  - Leicester

Who is voting?

All of the country will vote in the AV referendum. But council elections are not taking place in unitary counties such as:

- Cornwall
- Wiltshire
- Shropshire
- Isle of Wight
- Durham and Northumbria
- London

Sources: Labour and Conservative party calculations; Rankings and Threshold; FT research

Financial Infographics 09 [www.infographicity.com](http://www.infographicity.com)

Politics: Reprinted from *Sign of a Shifting Landscape* (n.d.), by Infographicity. Copyright by Infographicity. Retrieved from <http://infographicity.com/financial-infographics-09-2/>. Statement: The Conservatives will win votes after a year of govern (false).



## **Appendix C survey**

Appendix C consists of the survey used in the experiment. Chapter 3 refers to this appendix. The survey consisted of 10 questions about the entertainment and usability value for every type of graph. An example is given for the line graphs. After that, two ranking questions were filled out.

### **Entertainment**

Ik vind een lijngrafiek aantrekkelijk (I think a line graph is attractive)

Ik vind een lijngrafiek leuk om te bekijken (I think a line graph is fun to watch)

Ik vind een lijngrafiek saai (I think a line graph is boring)

Ik vind een lijngrafiek verwarrend (I think a line graph is confusing)

### **Usability**

#### *Learnability*

Ik vind een lijngrafiek moeilijk om te begrijpen

(I have a hard time to understand a line graph)

Ik heb snel door hoe ik een lijngrafiek moet interpreteren

(I know very fast how to interpret a line graph)

#### *Efficiency*

Ik kan snel informatie opzoeken in een lijngrafiek

(I can extract information fast in a line graph)

Ik vind dat ik veel overbodige oogbewegingen moet maken als ik een lijngrafiek lees

(I think that I need to make unnecessary eye movemens when I am reading a line graph)

#### *Memorability*

Ik kan informatie die in een lijngrafiek wordt gegeven eenvoudig onthouden

(I can memorize information in a line graph easily)

Ik vind het moeilijk om aan iemand anders uit leggen hoe je een lijngrafiek moet interpreteren

(I have a hard time to explain how you should interpret a line graph)

### **Ranking**

Waardeer the verschillende soorten grafieken naar aantrekkelijkheid (1= onaantrekkelijk, 5 = aantrekkelijk).

(Rank the attractiveness per type of graph, 1 = not attractive, 5 = attractive).

Waardeer de verschillende soorten grafieken naar moeilijkheidsgraad (1= moeilijk, 5 = makkelijk).

(Rank the difficulty per type of graph, 1 = difficult, 5 = easy).