

Wizualizacja danych

Wykład 1: Wizualizacja danych

Mariusz Tarnopolski

Instytut Astronomii UMK

DataViz, 2024/2025



- 15 godz. konwersatorium = wykład + dyskusje + problemy (dwugodzinne bloki, co dwa tygodnie wg harmonogramu)
- [Website](#)
- Konsultacje — wg potrzeb, proszę się umawiać przez maila:
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Zaliczenie (warunki inkrementalne)

- uczestnictwo w zajęciach (maks. 1 nieob. nieuspr. $\Rightarrow$ 3.0)
- aktywność ($\Rightarrow$ 4.0)
- omówienie* jednego wykresu/diagramu/ilustracji/itp. z artykułów jakie pojawią się w czasie semestru na <https://arxiv.org/> ($\Rightarrow$ 5.0):

*(1) Coś więcej niż „za mała czcionka” lub ”to jest wykres kołowy”. (2) W formie pisemnej elektronicznej, przesłanej mailem: w treści maila lub w załączonym pliku. Równoważność 5 min. prezentacji, czyli ok. 0.5 strony A4. Podać link do artykułu (identyfikator arXiv) oraz numer omawianego wykresu.

arXiv is a free distribution service and an open-access archive for 2,257,602 scholarly articles in the fields of physics, mathematics, computer science, quantitative biology, quantitative finance, statistics, electrical engineering and systems science, and economics. Materials on this site are not peer-reviewed by arXiv.

Subject search and browse:

Physics Search Form Interface Catchup

News

Read about recent news and updates on [arXiv's blog](#). (View the former "what's new" pages here).

Physics

- Astrophysics ([astro-ph](#) new, recent, search)
includes: Astrophysics of Galaxies; Cosmology and Nongalactic Astrophysics; Earth and Planetary Astrophysics; High Energy Astrophysical Phenomena; Instrumentation and Methods for Astrophysics; Solar and Stellar Astrophysics
- Condensed Matter ([cond-mat](#) new, recent, search)
includes: Disordered Systems and Neural Networks; Materials Science; Mesoscale and Nanoscale Physics; Other Condensed Matter; Quantum Gases; Soft Condensed Matter; Statistical Mechanics; Strongly Correlated Electrons; Superconductivity
- General Relativity and Quantum Cosmology ([gr-qc](#) new, recent, search)
- High Energy Physics - Experiment ([hep-ex](#) new, recent, search)
- High Energy Physics - Lattice ([hep-lat](#) new, recent, search)
- High Energy Physics - Phenomenology ([hep-ph](#) new, recent, search)
- High Energy Physics - Theory ([hep-th](#) new, recent, search)
- Mathematical Physics ([math-ph](#) new, recent, search)
- Nonlinear Sciences ([nlin](#) new, recent, search)
includes: Adaptation and Self-Organizing Systems; Cellular Automata and Lattice Gases; Chaotic Dynamics; Exactly Solvable and Integrable Systems; Pattern Formation and Solitons
- Nuclear Experiment ([nucl-ex](#) new, recent, search)
- Nuclear Theory ([nucl-th](#) new, recent, search)
- Physics ([physics](#) new, recent, search)
includes: Accelerator Physics; Applied Physics; Atmospheric and Oceanic Physics; Atomic and Molecular Clusters; Atomic Physics; Biological Physics; Chemical Physics; Classical Physics; Computational Physics; Data Analysis; Statistics and Probability; Fluid Dynamics; General Physics; Geophysics; History and Philosophy of Physics; Instrumentation and Detectors; Medical Physics; Optics; Physics and Society; Physics Education; Plasma Physics; Popular Physics; Space Physics
- Quantum Physics ([quant-ph](#) new, recent, search)

Mathematics

- Mathematics ([math](#) new, recent, search)
includes (see detailed description): Algebraic Geometry; Algebraic Topology; Analysis of PDEs; Category Theory; Classical Analysis and ODEs; Combinatorics; Commutative Algebra; Complex Variables; Differential Geometry; Dynamical Systems; Functional Analysis; General Mathematics; General Topology; Geometric Topology; Group Theory; History and Overview; Information Theory; K-Theory and Homology; Logic; Mathematical Physics; Metric Geometry; Number Theory; Numerical Analysis; Operator Algebras; Optimization and Control; Probability; Quantum Algebra; Representation Theory; Rings and Algebras; Spectral Theory; Statistics Theory; Symplectic Geometry

Computer Science

- Computing Research Repository ([CoRR](#) new, recent, search)
includes (see detailed description): Artificial Intelligence; Computation and Language; Computational Complexity; Computational Engineering, Finance, and Science; Computational Geometry; Computer Science and Game Theory; Computer Vision and Pattern Recognition; Computers and Society; Cryptography and Security; Data Structures and Algorithms; Databases; Digital Libraries; Discrete Mathematics; Distributed, Parallel, and Cluster Computing; Emerging Technologies; Formal Languages and Automata Theory; General Literature; Graphics; Hardware Architecture; Human-Computer Interaction; Information Retrieval; Information Theory; Logic in Computer Science; Machine Learning; Mathematical Software; Multiagent Systems; Multimedia; Networking and Internet Architecture; Neural and Evolutionary Computing; Numerical Analysis; Operating Systems; Other Computer Science; Performance; Programming Languages; Robotics; Social and Information Networks; Software Engineering; Sound; Symbolic Computation; Systems and Control

Quantitative Biology

- Quantitative Biology ([q-bio](#) new, recent, search)
includes (see detailed description): Biomolecules; Cell Behavior; Genomics; Molecular Networks; Neurons and Cognition; Other Quantitative Biology; Populations and Evolution; Quantitative Methods; Subcellular Processes; Tissues and Organs

Quantitative Finance

- Quantitative Finance ([q-fin](#) new, recent, search)
includes (see detailed description): Computational Finance; Economics; General Finance; Mathematical Finance; Portfolio Management; Pricing of Securities; Risk Management; Statistical Finance; Trading and Market Microstructure

Plan ramowy kursu

- 1 Historia, cele i przykłady DataViz; cele kursu
- 2 Podstawowe rodzaje wykresów
- 3 Fundamentalne zasady projektowania graficznego
- 4 Typografia
- 5 Teoria koloru
- 6 Aspekty kognitywne (w tym złudzenia optyczne)
- 7 Przegląd dobrych & złych wykresów naukowych
- 8 Przegląd (głównie złych) wykresów z mediów
- 9 Wykresy specjalne
- 10 Wizualizacje 3D
- 11 Schematy blokowe (*flowcharts*)
- 12 Dane wielowymiarowe*: redukcja wymiaru, współrzędne równoległe
- 13 Tabele*

*Tylko jeśli czas pozwoli.

Literatura podstawowa

Graphical Perception: Theory, Experimentation, and Application to the Development of Graphical Methods

WILLIAM S. CLEVELAND and ROBERT MUGSEL*

The subject of graphical methods for data analysis and for data presentation needs no scientific foundation. In this article we take a first step in the direction of establishing such a foundation. Our approach is based on graphical perception—the visual decoding of information encoded on graphs—and it includes both theory and experimental data to test the theory. The theory deals with a small but important piece of the whole process of graphical perception. The first part is an identification of a set of elementary perceptual units that are common to most people extract quantitative information from graphs. The second part is an ordering of the units on the basis of how accurately people perform them. Elements of the theory are tested by experimentation in which subjects record their judgments of the quantitative information on graphs. The experiments validate these elements but also suggest that the set of elementary units should be expanded. The theory provides a guideline for graph construction. Graphical design elements are ranked in the ordering as possible. This principle is applied to a variety of graphs, including bar charts, divided bar charts, pie charts, and statistical maps with shading. The conclusion is that initial design on these present graphs is needed, and so replacement we offer alternative graphical formats—charts, dot charts with grouping, and pie charts with shading.

KEY WORDS: Computer graphics; Psychophysics.

1. INTRODUCTION

Nearly 200 years ago William Playfair (1786) began the modern use of graphs for coding of data. More than 50 years later, in 1845, Charles Minors published *The American Statistical Association* about the relative merits of a small number of graphical methods for data analysis. In 1907, Crockett and Vignati (1921) used Minors (1921). Tansie graphs are a vital part of communication in science and technology, business, education, and the mass media.

Still, graph design for data analysis and presentation is largely unsystematic. This is why Cox (1977) argued, "There is a major need for a theory of graphical methods." In "a very limited (1977) regard" to showing, constructing, and comparing graphical methods we have little to go on by tradition, use of theory, and a kind of intergenerational passing of information. . . . there is neither theory nor consistent body of experiment as a guide" (p. 28-29).

There is, of course, much good common sense about how to make a graph. There are many theories on graph construction (e.g., Schmitt and Schmitt 1970). But practice has been successful (e.g., Tufte 1983), graph designers have thought intensely about graphical methods for data analysis (e.g., Tansie 1977, Cleveland et al. 1983), and computer graphics have shown great ability in the generation of traditional single-theme charts (Barrett, Smith, and Shneider 1978, Cleveland 1983, Cleveland and McGee 1984, Cleveland National Research Institute 1979) provide guidelines for making graphs. But the manual advice, "This standard . . . sets forth the best current usage, and offers standards by general agreement rather than by scientific test" (p. 16).

In this article we approach the science of graphs through human graphical perception. The approach includes both theory and experimentation to test it.

The first part of the theory is a list of elementary perceptual units that people perform in extracting quantitative information from graphs. In the second part we hypothesize an ordering of the elementary units based on how accurately people perform them. We do not expect that the accuracy of quantitative extraction in the units of a graph to which one might want to transfer a

The theory is testable; we use it to predict the relative performance of existing graphs, and there are no experiments to check the actual performance. The experiments are of two types: In one, since the graphs are drawn, the evidence appears so strong that it is taken prima facie to have established the case. When a strong effect is perceived by the authors' eyes and noses, it is likely that it will appear to most other people as well. In

the second, the evidence is less strong, and the experiments are designed to test the theory.

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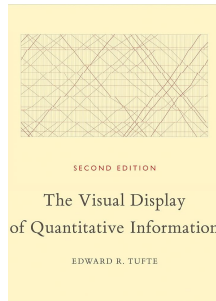
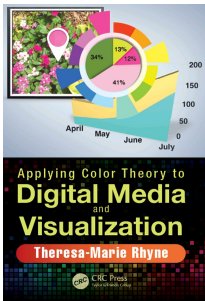
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Literatura dodatkowa

Graphs, Tables, and Figures in Scientific Publications: The Good, the Bad, and How Not to Be the Latter

Lorenz E. Fraundhaar, Kerin C. Chang, MEd, MS

Graphs, figures, and tables can save readers time and energy, aid their understanding of an article, and reduce the word count of the same text. However, many graphics submitted to and published in scientific journals fail to meet their potential and include mistakes that jeopardize their clarity. Many formats are available for presenting data, as well as a variety of techniques for enhancing interpretability. When the appropriate format is used to display data, it conveys the greatest amount of information in the clearest fashion, complements the text, and improves readers' understanding. The aims of this article are to draw attention to the necessity of well-constructed graphs, tables, and figures in scientific publications, and to show how to create them. (*J Biol Chem* 2012;287:11611-11616. Copyright © 2012 by the American Society for Biochemistry and Molecular Biology. All rights reserved.)

Key words: Graphs, tables, charts, figures, manuscripts.

In the current or recent research article to be done, the information that gives due to conclusions. However, unless proper presentation, the data can be misinterpreted or worse ignored. To avoid this, the authors must use well-constructed graphs, figures, and tables to display data and trends, and to communicate information. Illustrations or graphics have a number of proven advantages over text. They present complex data and relationships in a way that is easier to interpret and understand.¹⁻³ Second, they reduce reading time by summarizing and highlighting key findings or trends so that they are readily visible,^{4,5} and third, they reduce the overall word count.

Well-designed graphics make data easier to understand by explicitly displaying trends and key points, and they are valuable assets to any paper or presentation. These visual aids assist readers' understanding and improve the accuracy of their interpretation of the data.⁶ On the other hand, poorly constructed graphics also have a number of drawbacks. They can mislead, distort, or oversimplify data, and they can be time-consuming to create and publish in scientific journals. The aims of this article are to bring to light common mistakes in graph, table, and figure construction, and to show how to create high-quality graphics that will enhance reader acceptance and understanding of data.

WHAT NOT TO DO

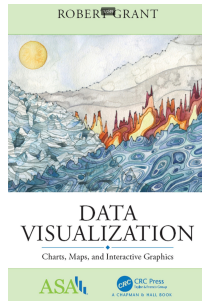
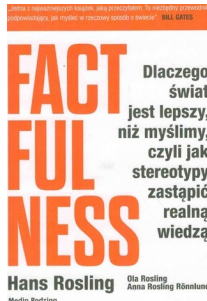
Many submitted and even published manuscripts contain illustrations of "unpublished quality."⁷ Table 1 details common errors that lead to the less-than-ideal graphical and table presentation. The most glaring oversight is neglecting the most important information about the data that should be displayed.

Many common graphical and graphical formats that are easy to use and have a problem of design elements. However, they tend to produce nonsensical



P. Biecek

ESEJE



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oraz dodatkowe materiały podane w trakcie zajęć.

Oprogramowanie

(o tym ten kurs nie jest)

python



matplotlib, seaborn,
plotnine, bokeh, ...

Matlab



Excel



Impress



R



ggplot2, ...

Mathematica



Photoshop



gimp



Java Script



d3.js, ...

gnuplot



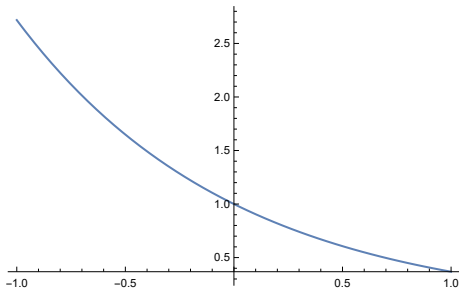
Illustrator



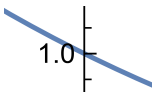
inkscape



Formaty plików

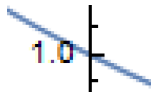


Wektorowe (.eps, .pdf, .svg):



- nieskończona rozdzielczość

Rastrowe (.jpeg, .png, .tiff):



- Zdjęcia
- Sposób na zmniejszenie wielkości plików

- 1 Funkcja ilustracyjna:
wykres jest wart więcej niż tysiąc słów, tj. opisanie przebiegu wielkości w czasie, złożona zależność jednej wielkości od drugiej, rozkład całości na składowe, itd.
- 2 Funkcja eksploratywna:
walidacja; poszukiwanie wzorców, np. okresowości w czasie/przestrzeni, nadreprezentacja pewnej subpopulacji, grupowanie, itd.

1a) Kwartet Anscombe'a

Some of these points are illustrated by four fictitious data sets, each consisting of eleven (x, y) pairs, shown in the table. For the first three data sets the x -values are the same, and they are listed only once.

Data set	1-3	1	2	3	4	4	
Variable	x	y	y	y	x	y	
Obs. no.	1 :	10.0	8.04	9.14	7.46 :	8.0	6.58
	2 :	8.0	6.95	8.14	6.77 :	8.0	5.76
	3 :	13.0	7.58	8.74	12.74 :	8.0	7.71
	4 :	9.0	8.81	8.77	7.11 :	8.0	8.84
	5 :	11.0	8.33	9.26	7.81 :	8.0	8.47
	6 :	14.0	9.96	8.10	8.84 :	8.0	7.04
	7 :	6.0	7.24	6.13	6.08 :	8.0	5.25
	8 :	4.0	4.26	3.10	5.39 :	19.0	12.50
	9 :	12.0	10.84	9.13	8.15 :	8.0	5.56
	10 :	7.0	4.82	7.26	6.42 :	8.0	7.91
	11 :	5.0	5.68	4.74	5.73 :	8.0	6.89

TABLE. Four data sets, each comprising 11 (x, y) pairs.

Each of the four data sets yields the same standard output from a typical regression program, namely

Number of observations $(n) = 11$

Mean of the x 's $(\bar{x}) = 9.0$

Mean of the y 's $(\bar{y}) = 7.5$

Regression coefficient (b_1) of y on $x = 0.5$

Equation of regression line: $y = 3 + 0.5x$

Sum of squares of $x - \bar{x} = 110.0$

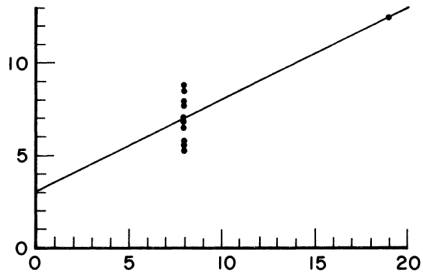
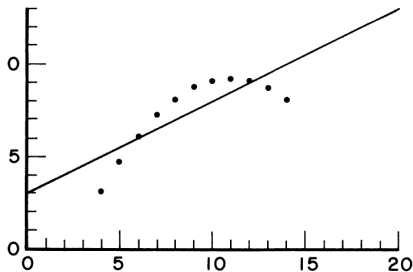
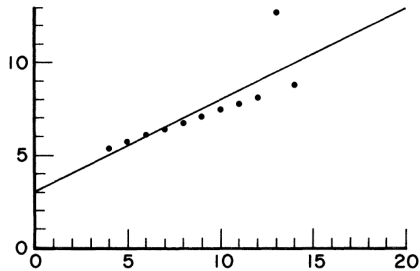
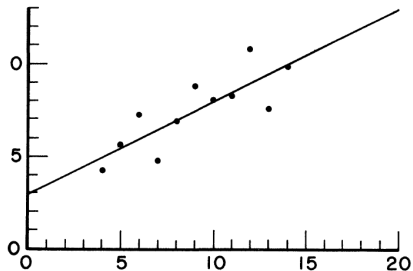
Regression sum of squares = 27.50 (1 d.f.)

Residual sum of squares of $y = 13.75$ (9 d.f.)

Estimated standard error of $b_1 = 0.118$

Multiple $R^2 = 0.667$

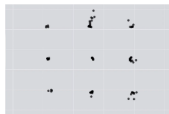
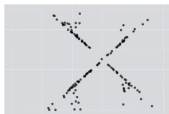
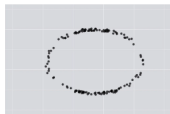
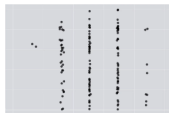
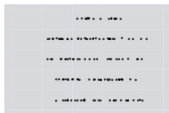
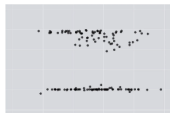
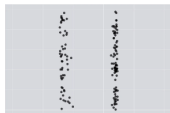
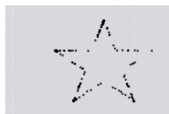
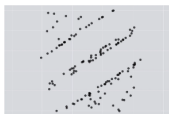
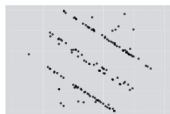
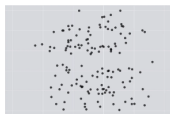
1a) Kwartet Anscombe'a



1b) DataSaurus

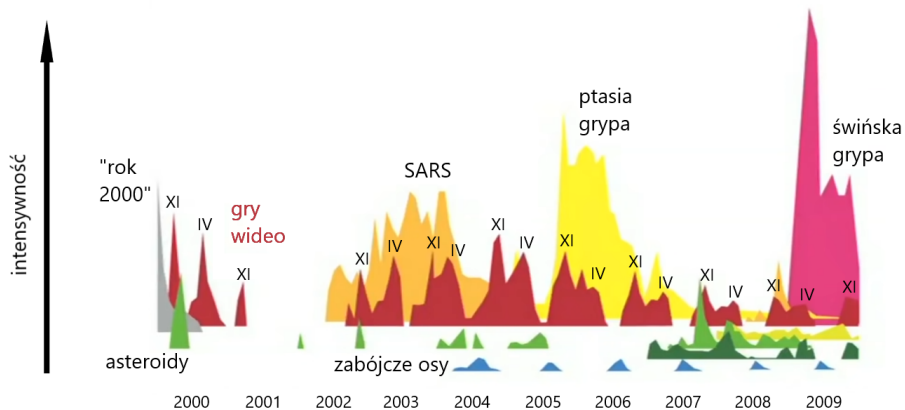


X Mean: 54.26
Y Mean: 47.83
X SD : 16.76
Y SD : 26.93
Corr. : -0.06



<https://www.research.autodesk.com/publications/same-stats-different-graphs/>

2) A timeline of global media panic



<https://www.youtube.com/watch?v=5Zg-C8AAIGg>

[https://media.nature.com/original/magazine-assets/d41586-021-01429-6/
d41586-021-01429-6.pdf](https://media.nature.com/original/magazine-assets/d41586-021-01429-6/d41586-021-01429-6.pdf)

60 000 pne.



<https://media.nature.com/original/magazine-assets/d41586-021-01429-6/d41586-021-01429-6.pdf>

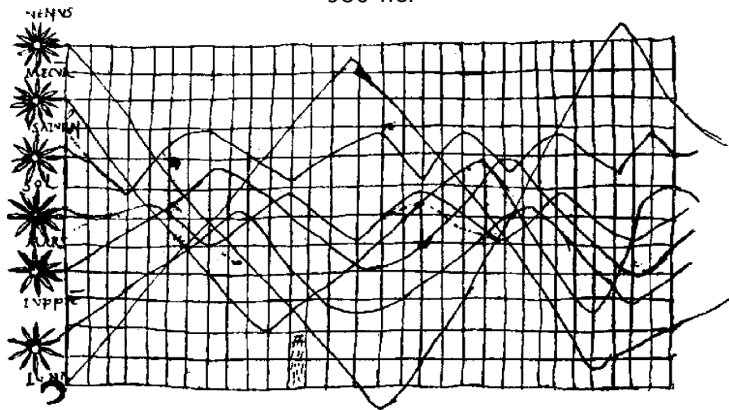
40 000 pne.



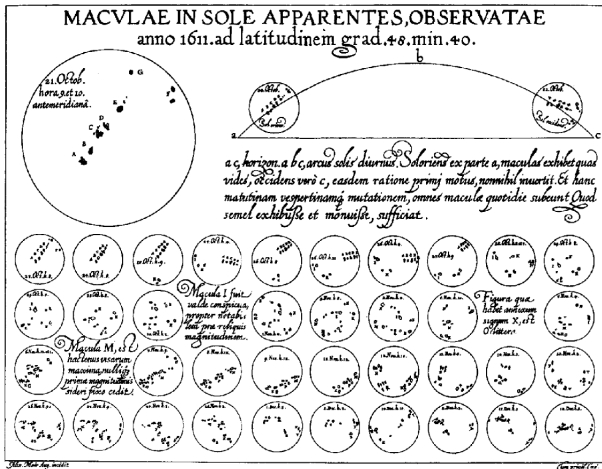
<https://blog.panoply.io/history-of-data-visualization>

Inklinacje ciał niebieskich

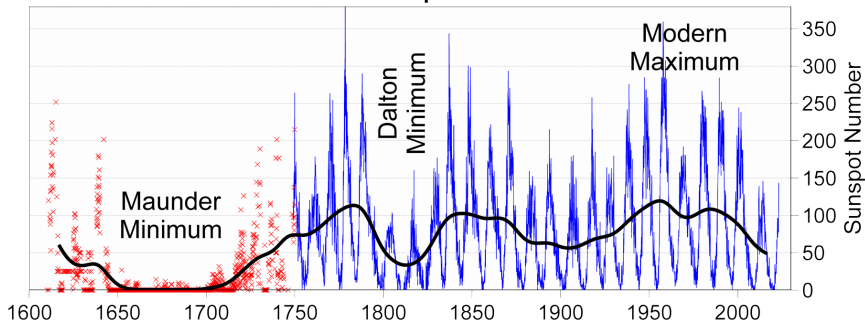
950 ne.



Scheiner, 1611



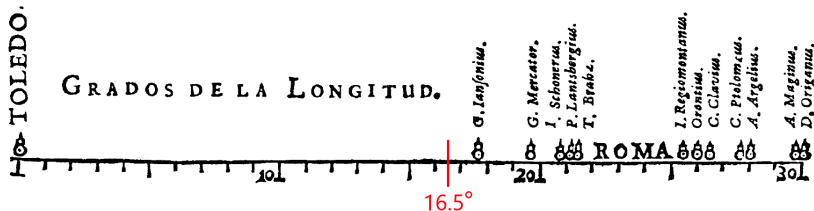
400 Years of Sunspot Observations



Odległość Toledo-Rzym

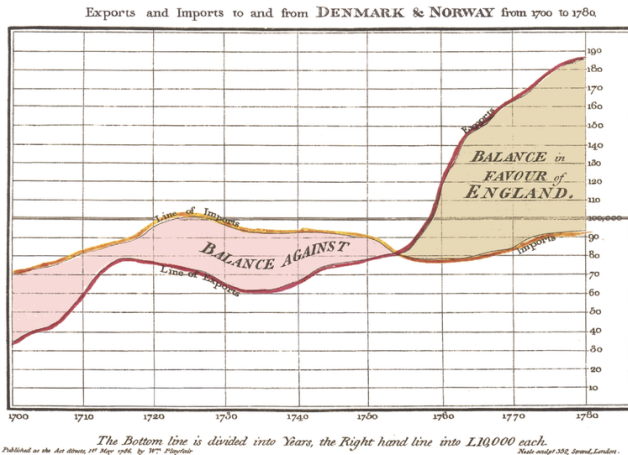
jako różnica długości geograficznych

Van Langren, 1644



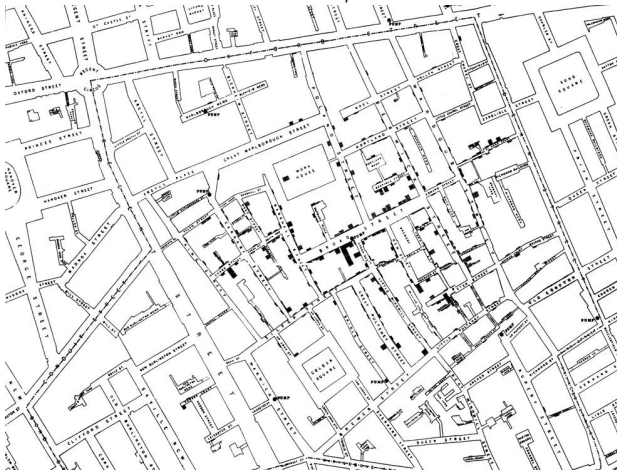
Równie dobrze można było przedstawić tabelę wartości, ale wykres jest bardziej obrazowy.

Playfair, 1821



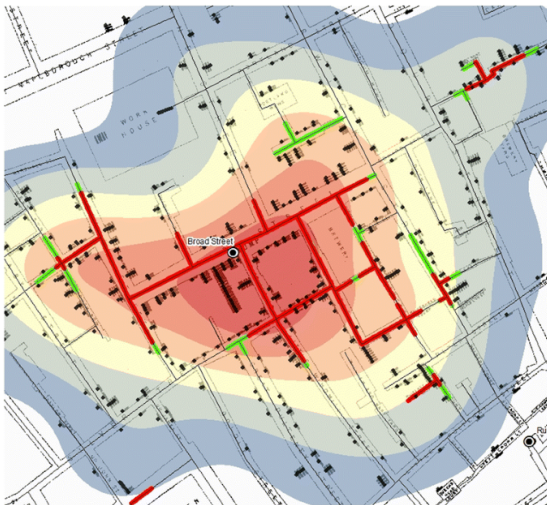
XIX-wieczne epidemie cholery

John Snow, 1854



<https://ij-healthgeographics.biomedcentral.com/articles/10.1186/s12942-015-0011-y>

John Snow, 1854



<https://ij-healthgeographics.biomedcentral.com/articles/10.1186/s12942-015-0011-y>

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www.jstor.org/stable/2246327

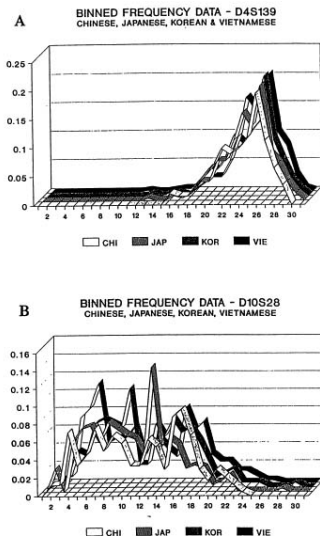
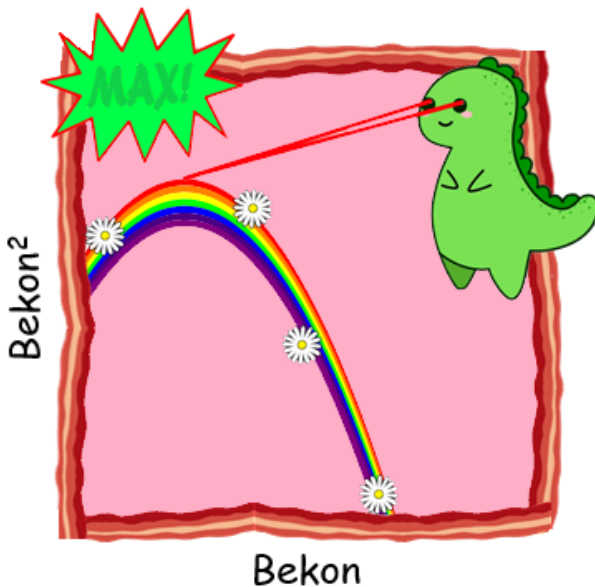
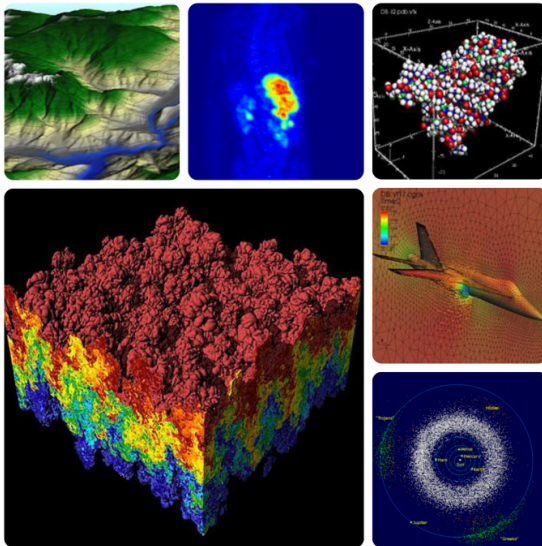


FIG. 4. Fixed bin distribution (histogram) for two loci and four Asian subpopulations (used with permission from John Hartmann): the boundaries of the 30 bins (vertical axis) are determined by the FBI; these bins are not of equal length. Sample sizes (numbers of individuals) for Chinese, Japanese, Korean and Vietnamese are 103, 125, 93 and 215 for D4S139 and 120, 137, 100 and 193 for D10S28. The horizontal axis is the bin number; bins are not of equal length.

Walory estetyczne/artystyczne są subiektywne...

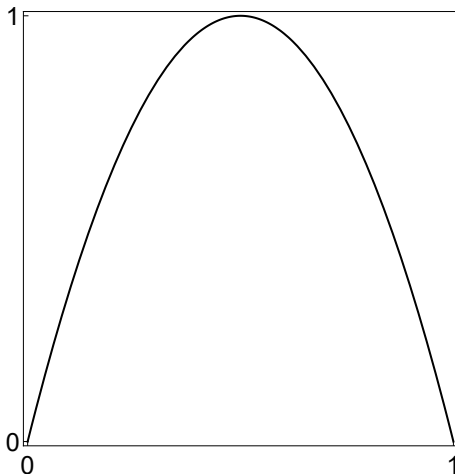


...dlatego należy dążyć do tego:



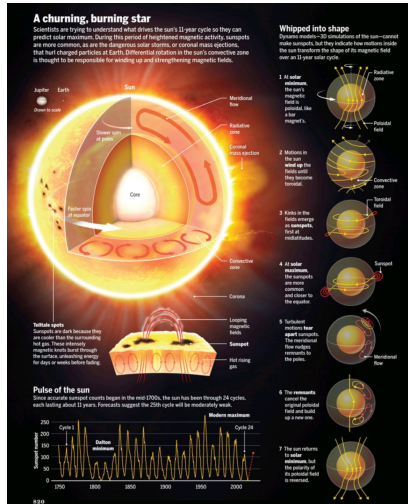
https://en.wikipedia.org/wiki/Scientific_visualization

...ale to też jest w porządku*:



* jeśli spełnia swój cel

...zaś to* wykracza poza ramy niniejszego kursu



*Tj. infografiki.

Building science graphics, Jen Christiansen

Mariusz Tarnopolski (IA UMK)

DataViz

Wykład 1 (2024/25)

27 / 27