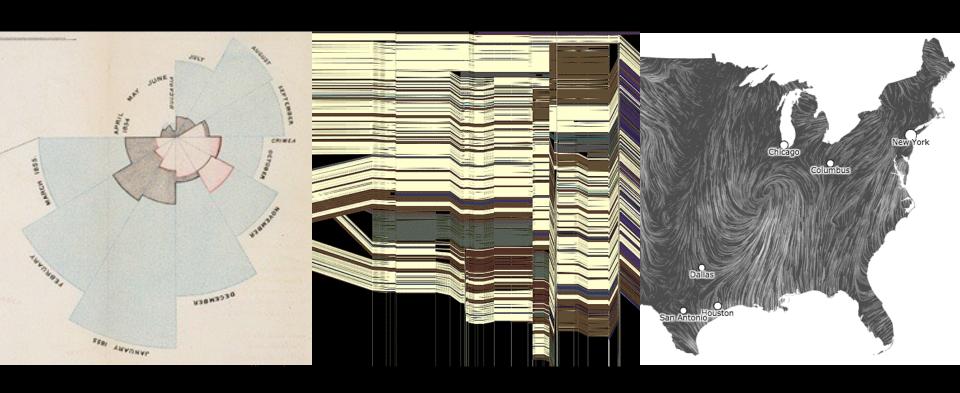
#### CSE 442 - Data Visualization

# Data and Image Models



Jeffrey Heer University of Washington

## The Big Picture

#### task

questions, goals assumptions

#### data

physical data type conceptual data type

#### domain

metadata semantics conventions processing algorithms

mapping visual encoding

image visual channel graphical marks

## **Topics**

Properties of Data

Properties of Images

Mapping Data to Images

# Data Models

# Data Models / Conceptual Models

Data models are formal descriptions

Math: sets with operations on them

Example: integers with + and x operators

**Conceptual models** are mental constructions Include semantics and support reasoning

Examples (data vs. conceptual)

1D floats vs. temperatures

3D vector of floats vs. spatial location

## Taxonomy of Data Types (?)

1D (sets and sequences)

**Temporal** 

2D (maps)

3D (shapes)

nD (relational)

Trees (hierarchies)

Networks (graphs)

Are there others?

The eyes have it: A task by data type taxonomy for information visualization [Shneiderman 96]

- N Nominal (labels or categories)
  - · Fruits: apples, oranges, ...

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- O Ordered
  - · Quality of meat: Grade A, AA, AAA

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  - · Only differences (i.e., intervals) may be compared

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  - Only differences (i.e., intervals) may be compared
- Q Ratio (zero fixed)
  - Physical measurement: Length, Mass, Time duration, ...
  - Counts and amounts

- N Nominal (labels or categories)
  - Operations: =, ≠
- O Ordered
  - Operations: =,  $\neq$ , <, >
- Q Interval (location of zero arbitrary)
  - Operations: =,  $\neq$ , <, >, =
  - · Can measure distances or spans
- Q Ratio (zero fixed)
  - Operations: =, ≠, <, >, -, %
  - Can measure ratios or proportions

#### From Data Model to N, O, Q

#### Data Model

32.5, 54.0, -17.3, ...

Floating point numbers

#### **Conceptual Model**

Temperature (°C)

#### Data Type

Burned vs. Not-Burned (N)

Hot, Warm, Cold (O)

Temperature Value (Q-interval)

#### **Dimensions & Measures**

**Dimensions** (~ independent variables)
Often discrete variables describing data (N, O)
Categories, dates, binned quantities

Measures (~ dependent variables)

Data values that can be aggregated (Q)

Numbers to be analyzed

Aggregate as sum, count, avg, std. dev...

Not a strict distinction. The same variable may be treated either way depending on the task.

# Example: U.S. Census Data

#### Example: U.S. Census Data

People Count: # of people in group

**Year**: 1850 - 2000 (every decade)

**Age**: 0 - 90+

Sex: Male, Female

Marital Status: Single, Married, Divorced, ...

#### Example: U.S. Census

People Count

Year

Age

Sex

**Marital Status** 

2,348 data points

	Α	В	С	D	Е
1	year	age	marst	sex	people
2	1850	0	0	1	1483789
3	1850	0	0	2	1450376
4	1850	5	0	1	1411067
5	1850	5	0	2	1359668
6	1850	10	0	1	1260099
7	1850	10	0	2	1216114
8	1850	15	0	1	1077133
9	1850	15	0	2	1110619
10	1850	20	0	1	1017281
11	1850	20	0	2	1003841
12	1850	25	0	1	862547
13	1850	25	0	2	799482
14	1850	30	0	1	730638
15	1850	30	0	2	639636
16	1850	35	0	1	588487
17	1850	35	0	2	505012
18	1850	40	0	1	475911
19	1850	40	0	2	428185
20	1850	45	0	1	384211
21	1850	45	0	2	341254
22	1850	50	0	1	321343
23	1850	50	0	2	286580
24	1850	55	0	1	194080
25	1850	55	0	2	187208
26	1850	60	0	1	174976
27	1850	60	0	2	162236
28	1850	65	0	1	106827
29	1850	65	0	2	105534
30	1850	70	0	1	73677
31	1850	70	0	2	71762
32	1850	75	0	1	40834
33	1850	75	0	2	40229
34	1850	80	0	1	23449
35	1850	80	0	2	22949
36	1850	85	0	1	8186
37	1850	85	0	2	10511
38	1850	90	0	1	5259
39	1850	90	0	2	6569
40	1860	0	0	1	2120846
41	1860	0	0	2	2092162

#### Census: N, O, Q-Interval, Q-Ratio?

**People Count** Q-Ratio

**Year** Q-Interval (O)

**Age** Q-Ratio (O)

Sex

**Marital Status** \( \)

#### Census: Dimension or Measure?

**People Count** Measure

**Year** Dimension

Age Depends!

**Sex** Dimension

Marital Status Dimension

#### Census Data Demo

https://observablehq.com/@uwdata/us-population-1850-2000-json

# Data Tables & Transformations

#### Relational Data Model

Represent data as a **table** (or *relation*)

Each **row** (or *tuple*) represents a record Each record is a fixed-length tuple

Each **column** (or *field*) represents a variable Each field has a *name* and a *data type* 

A table's **schema** is the set of names and types

A database is a collection of tables (relations)

Operations on Data Tables: table(s) in, table out

Operations on Data Tables: table(s) in, table out Project (select): select a set of columns Filter (where): remove unwanted rows Sort (order by): order records Aggregate (group by, sum, min, max, ...): partition rows into groups + summarize Combine (join, union, ...): integrate data from multiple tables

Project (select): select a set of columns
select day, stock

day	stock	price
10/3	AMZN	957.10
10/3	MSFT	74.26
10/4	AMZN	965.45
10/4	MSFT	74.69



day	stock
10/3	AMZN
10/3	MSFT
10/4	AMZN
10/4	MSFT

Filter (where): remove unwanted rows

select \* where price > 100

day	stock	price
10/3	AMZN	957.10
10/3	MSFT	74.26
10/4	AMZN	965.45
10/4	MSFT	74.69



day	stock	price
10/3	AMZN	957.10
10/4	AMZN	965.45

**Sort** (order by): order records select \* order by stock

day	stock	price
10/3	AMZN	957.10
10/3	MSFT	74.26
10/4	AMZN	965.45
10/4	MSFT	74.69



day	stock	price
10/3	AMZN	957.10
10/4	AMZN	965.45
10/3	MSFT	74.26
10/4	MSFT	74.69

**Aggregate** (group by, sum, min, max, ...): select stock, min(price) group by stock

day	stock	price
10/3	AMZN	957.10
10/3	MSFT	74.26
10/4	AMZN	965.45
10/4	MSFT	74.69



stock	min(price)
AMZN	957.10
MSFT	74.26

Join (join) multiple tables together

day	stock	price
10/3	AMZN	957.10
10/3	MSFT	74.26
10/4	AMZN	965.45
10/4	MSFT	74.69

	day	stock	price	min
	10/3	AMZN	957.10	957.10
•	10/3	MSFT	74.26	74.26
	10/4	AMZN	965.45	957.10
	10/4	MSFT	74.69	74.26

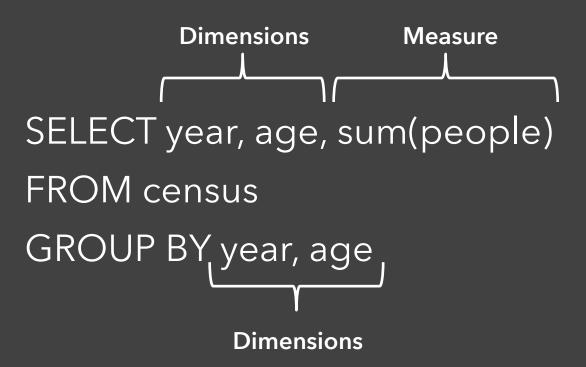
stock	min
AMZN	957.10
MSFT	74.26

select t.day, t.stock, t.price, a.min
from table as t, aggregate as a
where t.stock = a.stock

#### Roll-Up and Drill-Down

Want to examine population by year and age?

Roll-up the data along the desired dimensions



#### Roll-Up and Drill-Down

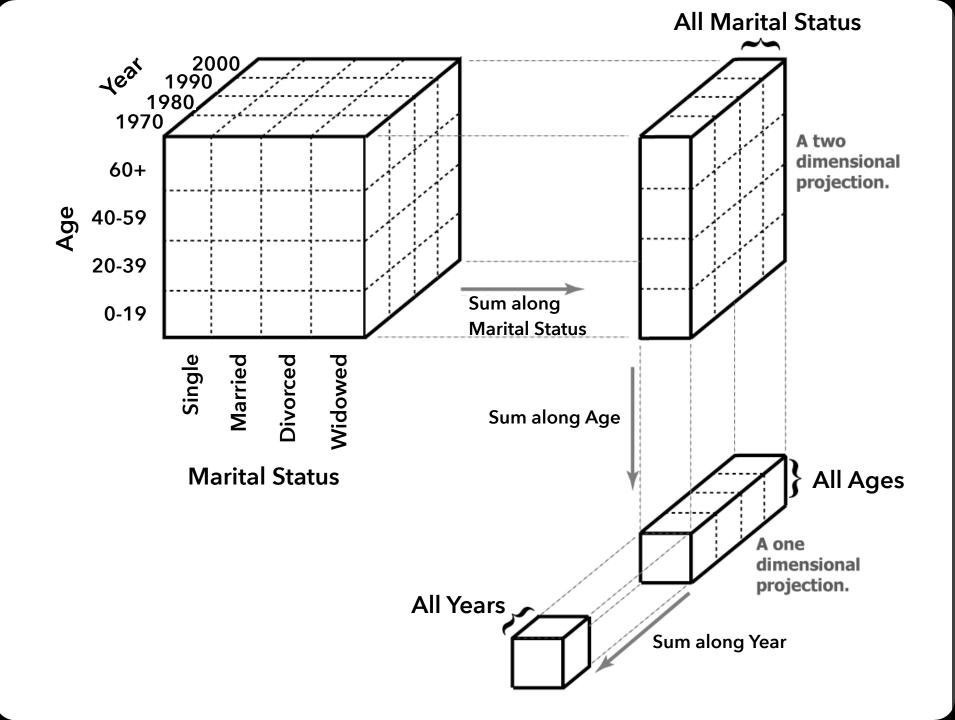
Want to see the breakdown by marital status?

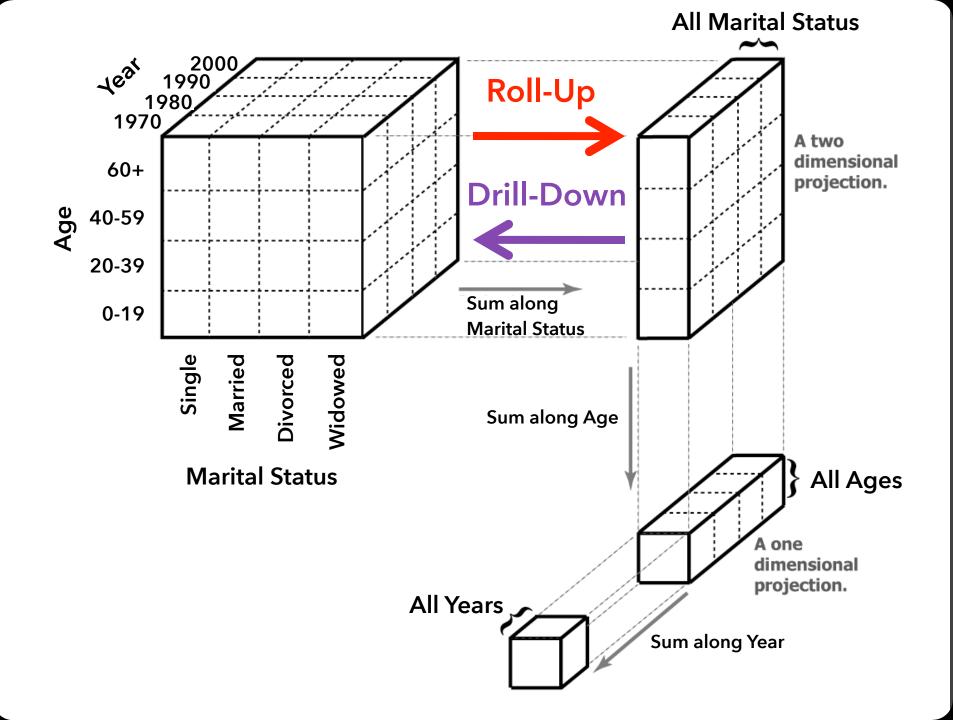
**Drill-down** into additional dimensions

SELECT year, age, marst, sum(people)

FROM census

GROUP BY year, age, marst





#### ORIGINAL

YEAR	AGE	MARST	SEX	PEOPLE
1850	0	0	1	1,483,789
1850	5	0	1	1,411,067
1860	0	0	1	2,120,846
1860	5	0	1	1,804,467
• • •				

#### PIVOTED (or CROSS-TABULATION)

AGE	MARST	SEX	1850	1860
0	0	1	1,483,789	2,120,846
5	0	1	1,411,067	1,804,467

Which format might we prefer? Why?

#### Tidy Data [Wickham 2014]

How do rows, columns, and tables match up with observations, variables, and types? In "tidy" data:

- 1. Each variable forms a column.
- 2. Each observation forms a row.
- 3. Each type of observational unit forms a table.

The advantage is that this provides a flexible starting point for analysis, transformation, and visualization.

Our pivoted table variant was not "tidy"!

(This is a variant of <u>normalized forms</u> in DB theory)

#### Common Data Formats

#### CSV: Comma-Separated Values (d3.csv)

```
year,age,marst,sex,people
1850,0,0,1,1483789
1850,5,0,1,1411067
```

• • •

### Common Data Formats

### CSV: Comma-Separated Values (d3.csv)

```
year,age,marst,sex,people 1850,0,0,1,1483789 1850,5,0,1,1411067
```

### JSON: JavaScript Object Notation (d3.json)

```
[
    {"year":1850,"age":0,"marst":0,"sex":1,"people":1483789},
    {"year":1850,"age":5,"marst":0,"sex":1,"people":1411067},
    ...
]
```

# Administrivia

#### Winston Elkin

#### Academic Background

Doggy School Level 1\*

PhD, UW CSE, 2019 - ???

Dissertation: Evaluating the Impact of Tree Species on a Dog's Desire

to Pee on it: an Autoethnography

#### **Industry Internships**

Meta 2022, Meta 2023. Even has the badge to prove it.



<sup>\*</sup> Indicates candidate did not complete all requirements but passed due to being very cute.

### A1: Expository Visualization

#### Design a static visualization for a data set.

The climate of a place can have a tremendous impact on people's lived experience. You will examine average monthly climate measurements for six major U.S. cities, roughly covering the edges of the continental United States.

You must choose the message you want to convey. What question(s) do you want to answer? What insight do you want to communicate?

### A1: Expository Visualization

Pick a **guiding question**, use it to title your vis. Design a **static visualization** for that question. You are free to **use any tools** (inc. pen & paper).

**Deliverables** (via Gradescope; see A1 page) Image of your visualization (PNG or JPG format) Short description + design rationale (≤ 4 paragraphs)

Due by **11:59 pm, Fri Oct 4**.

### **Tableau Tutorial**

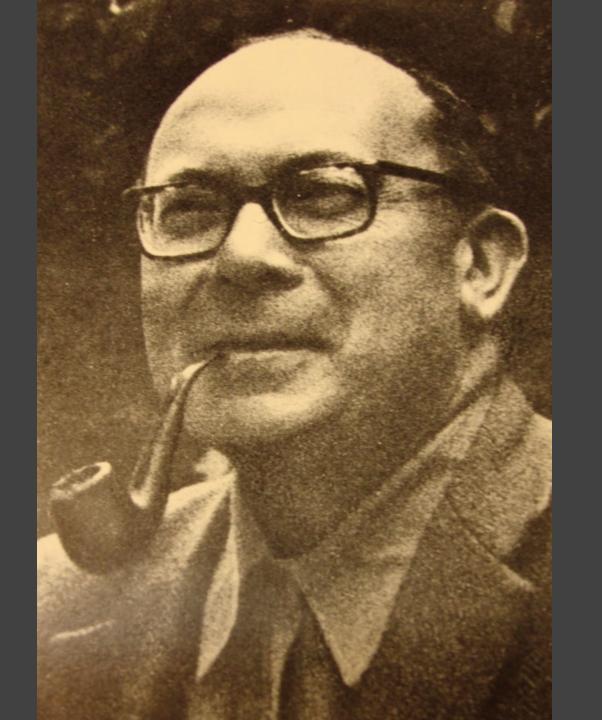
Friday Oct. 4, 4:00-5:30pm on Zoom

Introduction to Tableau, a graphical (no-code) visualization tool. You are free to use Tableau for A1 and/or A2.

Led by Shaan. Zoom link is available on Canvas. The tutorial will be recorded.

Download Tableau and sign up for a student license prior to tutorial!

# Image Models



### Visual Language is a Sign System



**Jacques Bertin** 

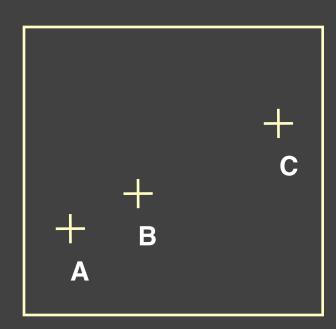
Images perceived as a set of signs

Sender encodes information in signs

Receiver decodes information from signs

Sémiologie Graphique, 1967

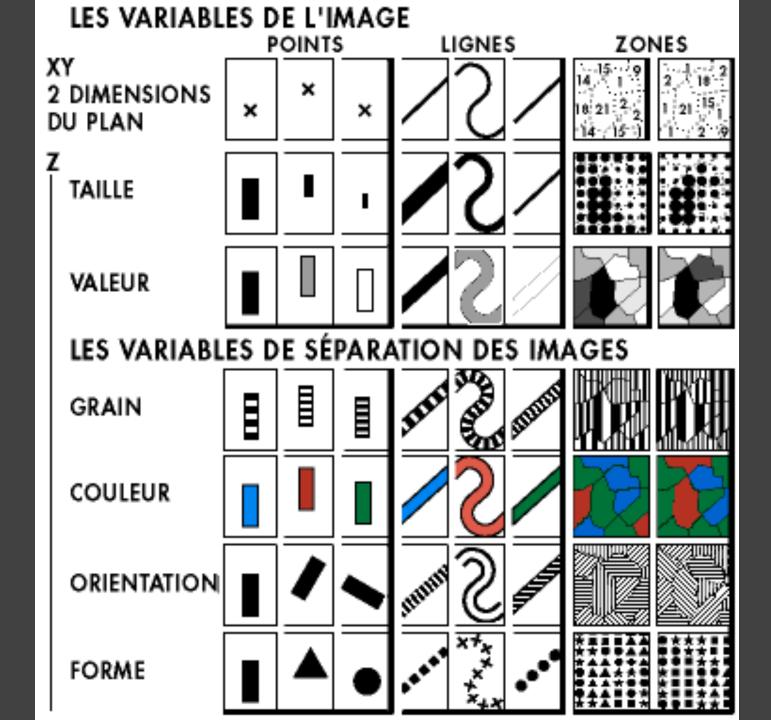
# Bertin's Semiology of Graphics



- 1. A, B, C are distinguishable
- 2. B is between A and C.
- 3. BC is twice as long as AB.

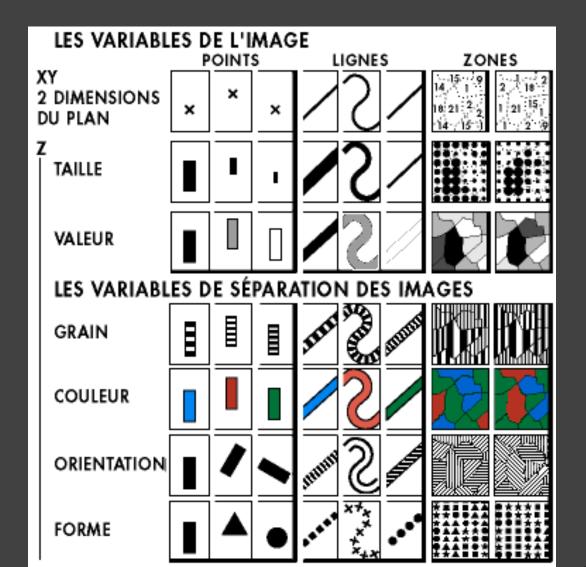
:. Encode quantitative variables

"Resemblance, order and proportional are the three signfields in graphics." - Bertin



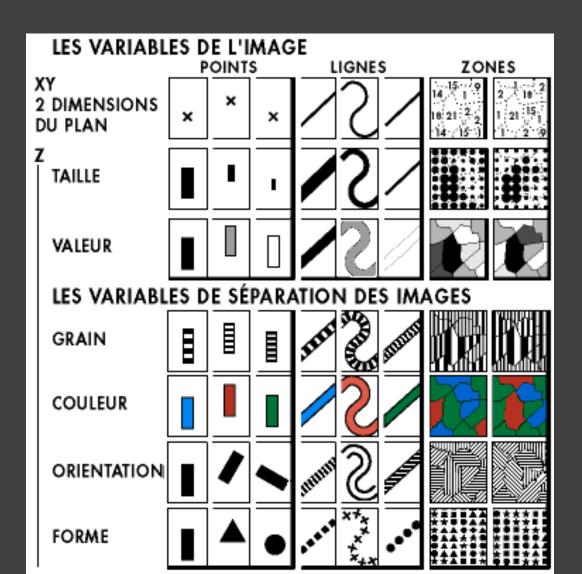
### Visual Encoding Variables

Position (x 2)
Size
Value
Texture
Color
Orientation
Shape



### Visual Encoding Variables

Position Length Area Volume Value Texture Color Orientation Shape **Transparency** Blur / Focus ...



### Information in Hue and Value

Value is perceived as ordered

∴ Encode ordinal variables (O)



: Encode continuous variables (Q) [not as well]



Hue is normally perceived as unordered

:. Encode nominal variables (N) using color



# Bertin's Levels of Organization

Position	N	О	Q
Size	N	0	Q
Value	N	0	Q
Texture	N	О	
Color	N		
Orientation	N		

Shape

Nominal

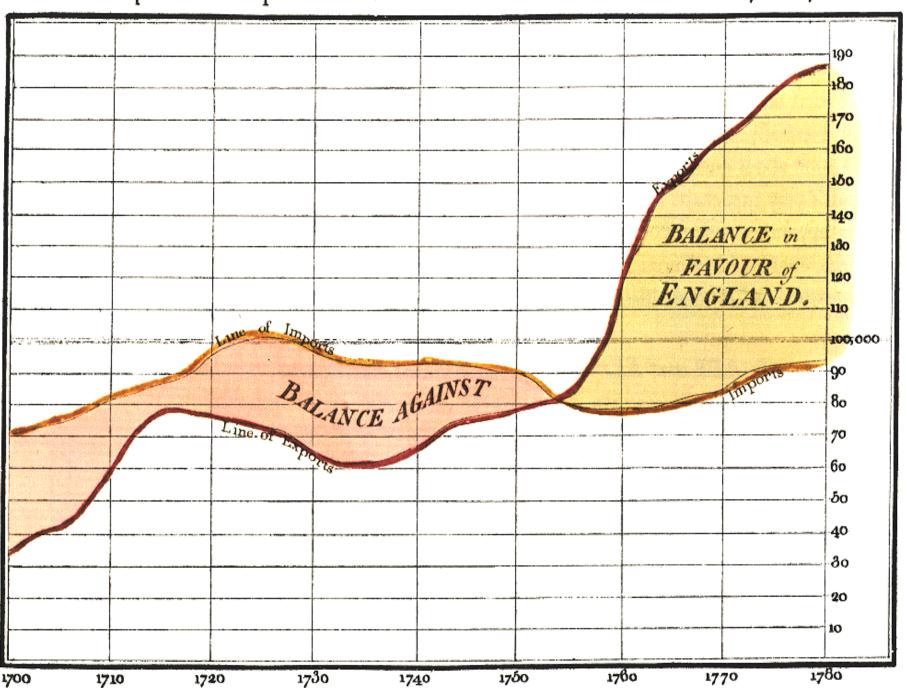
**O**rdinal

**Q**uantitative

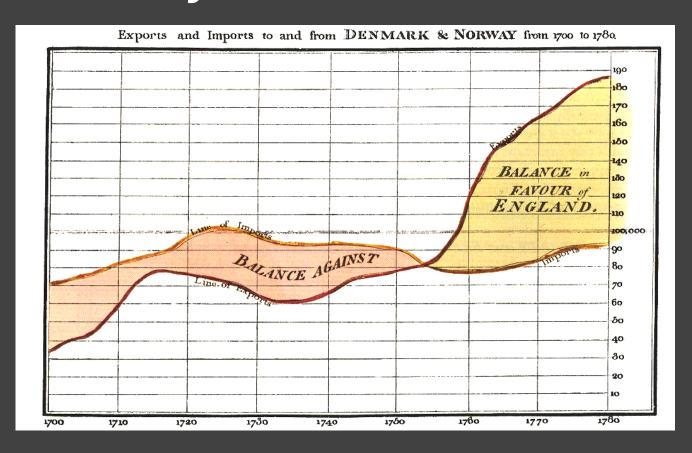
Note:  $\mathbf{Q} \subset \mathbf{O} \subset \mathbf{N}$ 

# Deconstructions

Exports and Imports to and from DENMARK & NORWAY from 1700 to 1780.



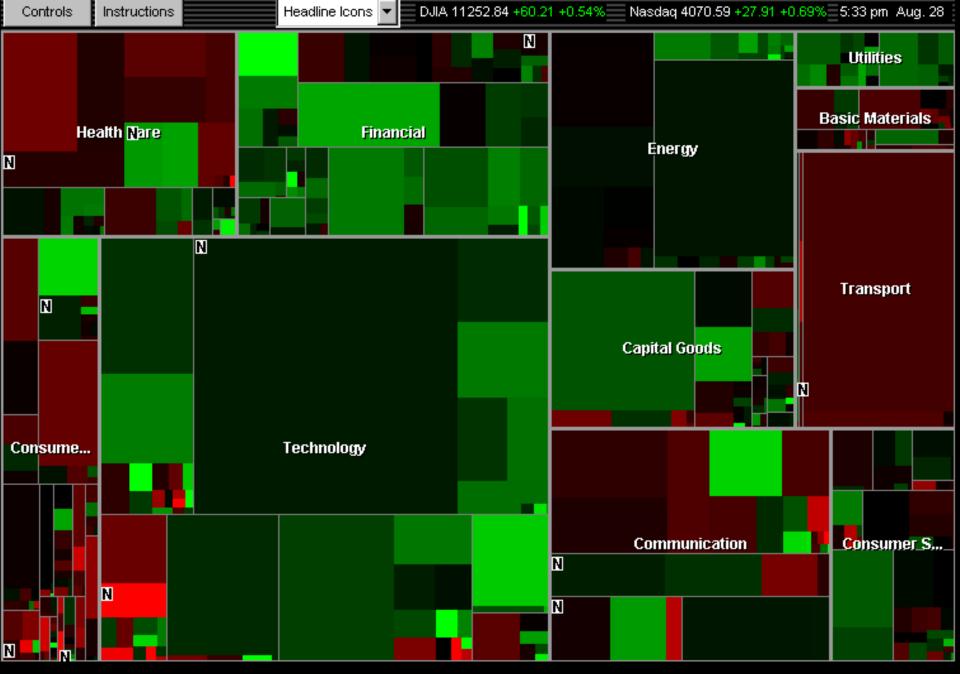
# William Playfair, 1786



X-axis: year (Q)

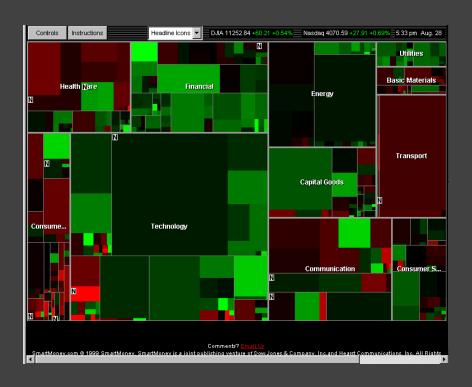
Y-axis: currency (Q)

Color: imports/exports (N, O)



http://www.smartmoney.com/marketmap/

## Wattenberg's Map of the Market



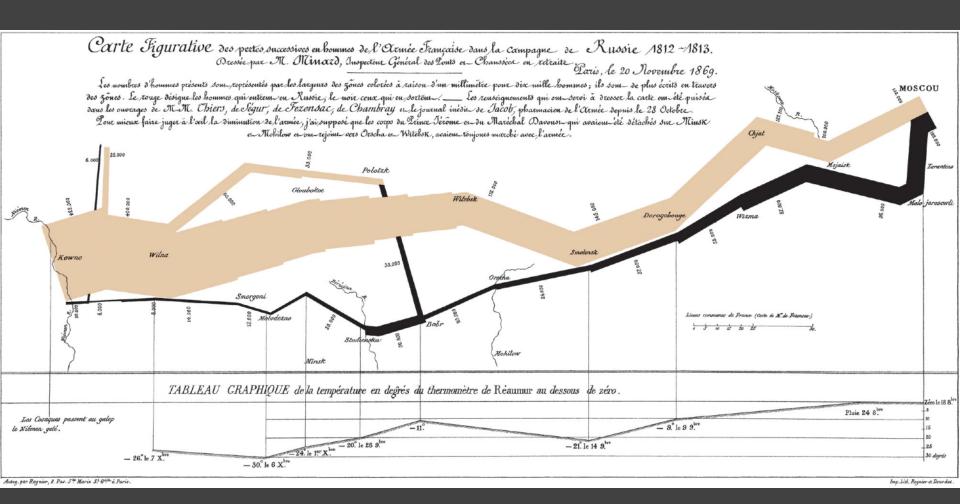
Rectangle Area: market cap (Q)

Rectangle Position: market sector (N), market cap (Q)

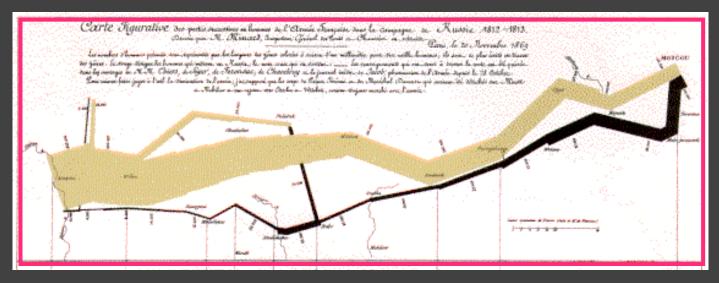
Color Hue: loss vs. gain (N, O)

Color Value: magnitude of loss or gain (Q)

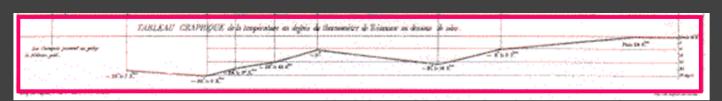
# Minard 1869: Napoleon's March



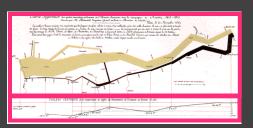
# Single-Axis Composition









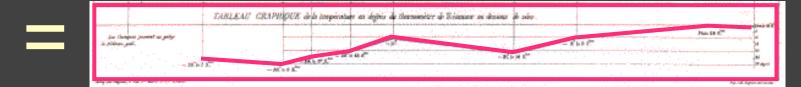


# Mark Composition

Y-axis: temperature (Q)

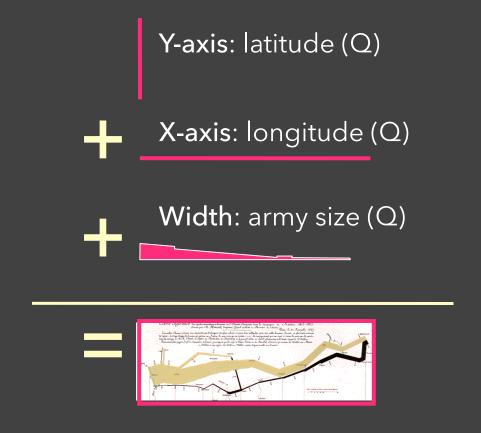


X-axis: longitude (Q) / time (O)



Temp over space/time  $(Q \times Q)$ 

# Mark Composition



Army position  $(Q \times Q)$  and army size (Q)

latitude (Q)

longitude (Q)

army size (Q)

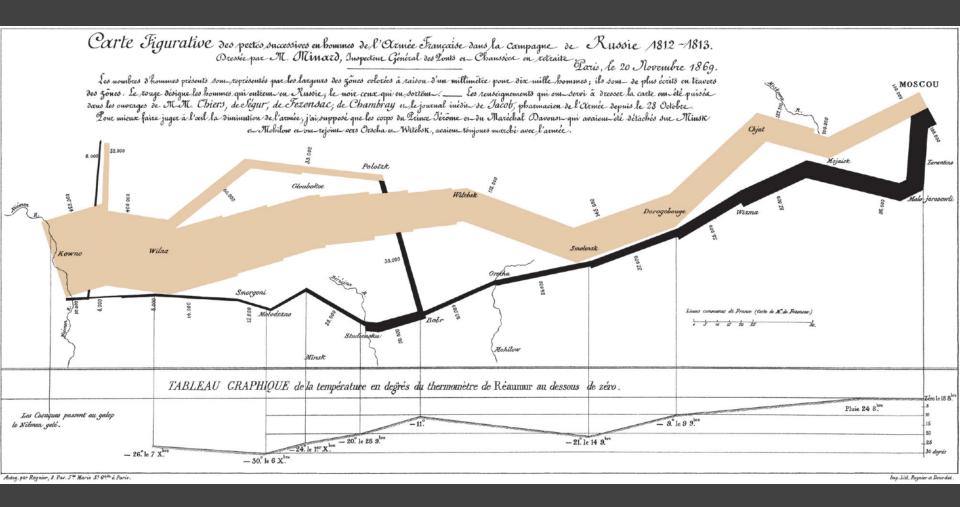




temperature (Q)

longitude (Q) / time (O)

# Minard 1869: Napoleon's March



Depicts at least 5 quantitative variables. Any others?

# Formalizing Design

### **Choosing Visual Encodings**

Assume k visual encodings and n data attributes. We would like to pick the "best" encoding among a combinatorial set of possibilities of size  $(n+1)^k$ 

#### **Principle of Consistency**

The properties of the image (visual variables) should match the properties of the data.

#### Principle of Importance Ordering

Encode the most important information in the most effective way.

### Design Criteria [Mackinlay 86]

#### Expressiveness

A set of facts is *expressible* in a visual language in the sentences (i.e. the visualizations) in the language express all the facts in the set of data, and only the facts in the data.

#### **Effectiveness**

A visualization is more effective than another visualization if the information conveyed by one visualization is more readily perceived than the information in the other visualization.

### Design Criteria [Mackinlay 86]

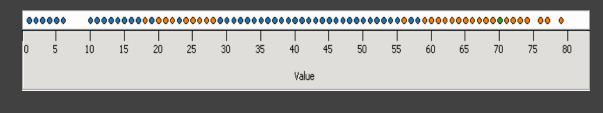
#### **Expressiveness**

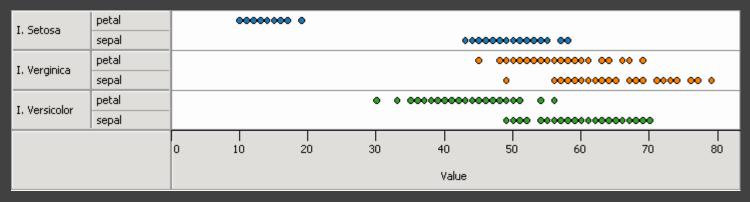
A set of facts is *expressible* in a visual language if the sentences (i.e. the visualizations) in the language express all the facts in the set of data, and only the facts in the data.

#### Effectiveness

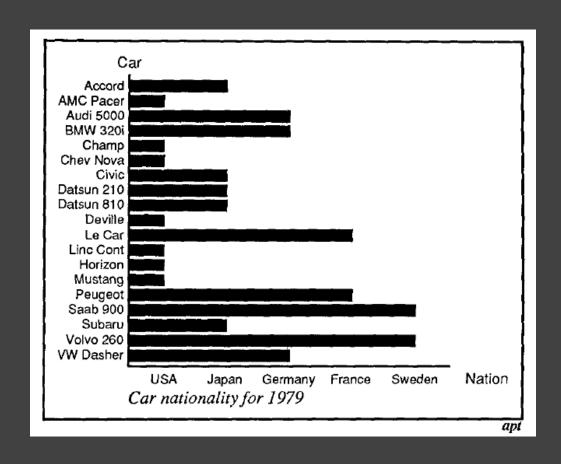
### Can not express the facts

A multivariate relation may be *inexpressive* in a single horizontal dot plot because multiple records are mapped to the same position.





### Expresses facts not in the data



A length is interpreted as a quantitative value.

### Design Criteria [Mackinlay 86]

#### **Expressiveness**

A set of facts is *expressible* in a visual language if the sentences (i.e. the visualizations) in the language express all the facts in the set of data, and only the facts in the data.

#### Effectiveness

### Design Criteria [Mackinlay 86]

#### **Expressiveness**

A set of facts is *expressible* in a visual language if the sentences (i.e. the visualizations) in the language express all the facts in the set of data, and only the facts in the data.

#### **Effectiveness**

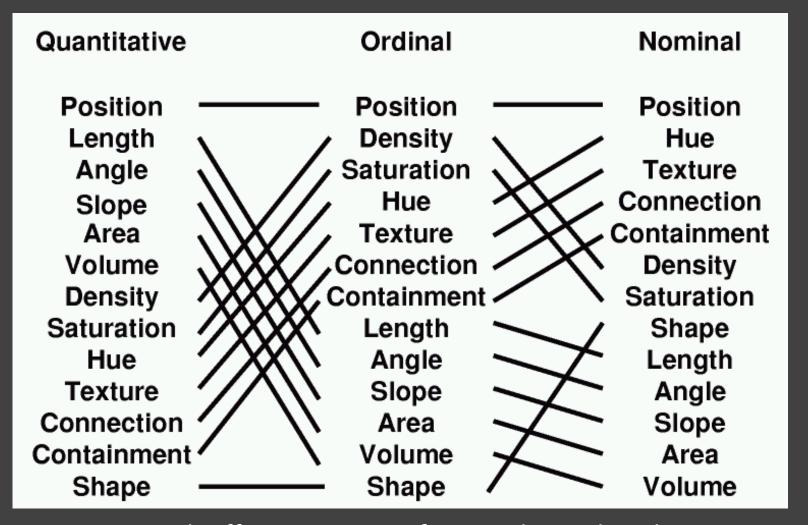
A visualization is more effective than another visualization if the information conveyed by one visualization is more readily perceived than the information in the other visualization.

### Design Criteria Translated

Tell the truth and nothing but the truth (don't lie, and don't lie by omission)

Use encodings that people decode better (where better = faster and/or more accurate)

# Mackinlay's Ranking



Conjectured effectiveness of encodings by data type

### Mackinlay's Design Algorithm

APT - "A Presentation Tool", 1986

User formally specifies data model and type Input: ordered list of data variables to show

#### APT searches over design space

Test expressiveness of each visual encoding Generate encodings that pass test Rank by perceptual effectiveness criteria

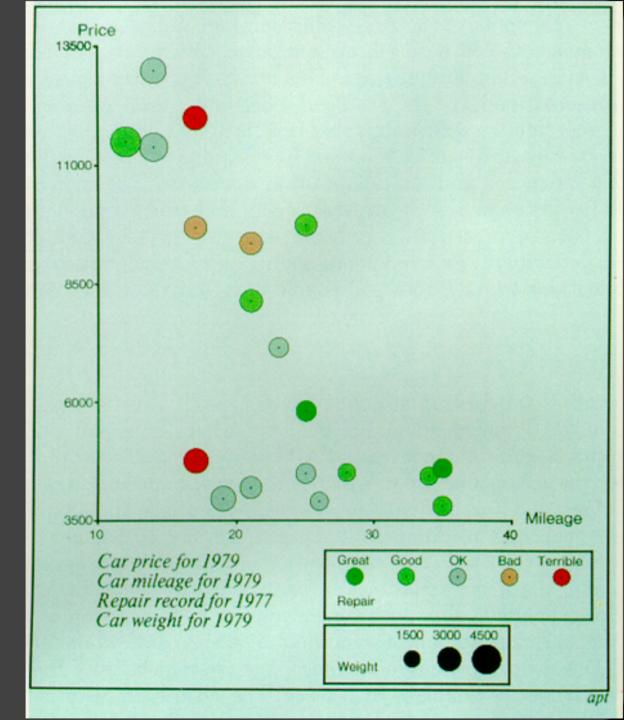
Output the "most effective" visualization

### **APT**

Automatically generate chart for car data

#### Input variables:

- 1. Price
- 2. Mileage
- 3. Repair
- 4. Weight



### Limitations of APT?

### **Limitations of APT**

**Does not cover many visualization techniques**Networks, hierarchies, maps, diagrams
Also: 3D structure, animation, illustration, ...

Does not consider interaction

Does not consider semantics / conventions

Assumes single visualization as output

Still an active area of research, e.g., the **Draco visualization design knowledge base** 

### Summary: Data & Image Models

#### Formal specification

Data model: relational data; N,O,Q types Image model: visual encoding channels Encodings map data to visual variables

### Choose expressive and effective encodings Rule-based tests of expressiveness Perceptual effectiveness rankings

**Question**: how do we establish effectiveness criteria? Subject of perception lectures...