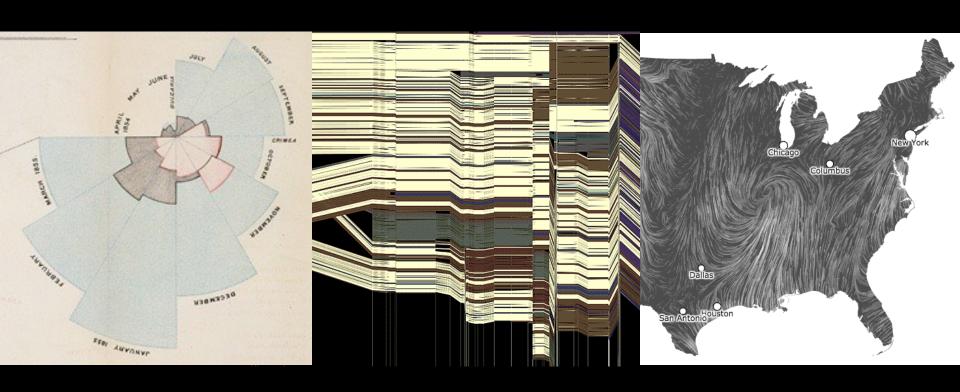
#### CSE 442 - Data Visualization

# Data and Image Models



Jeffrey Heer University of Washington

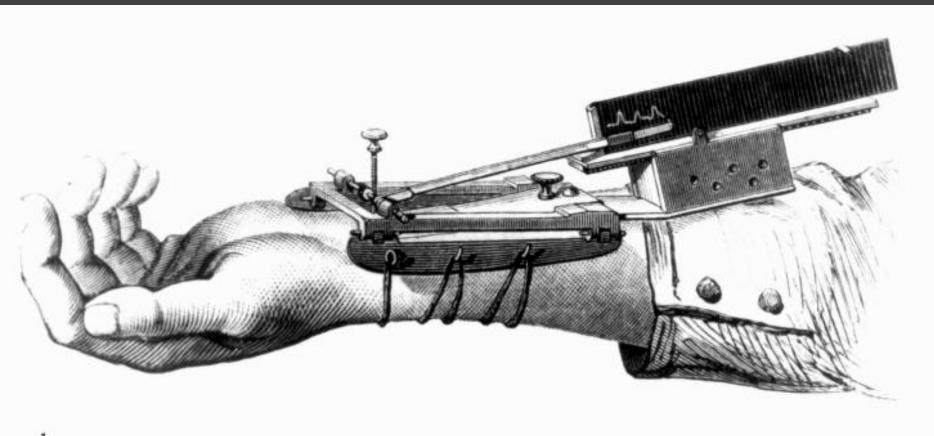
# Last Week: Value of Visualization

#### The Value of Visualization

- Record information
  Blueprints, photographs, seismographs, ...
  Analyze data to support reasoning
  Develop and assess hypotheses
  Find patterns / Discover errors in data
  Expand memory
- **Communicate** information to others

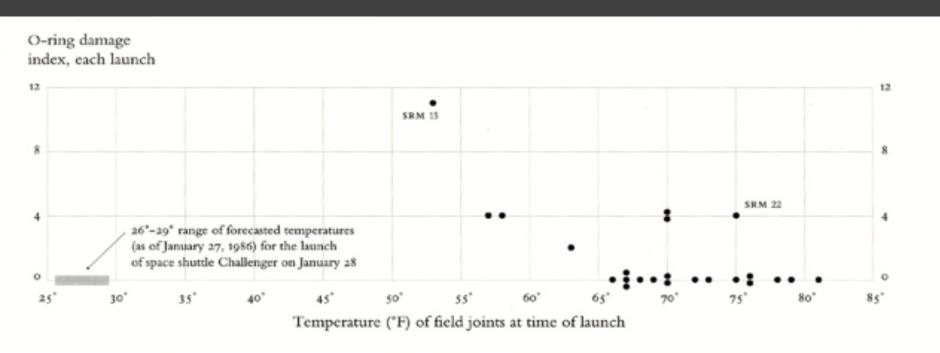
  Share and persuade

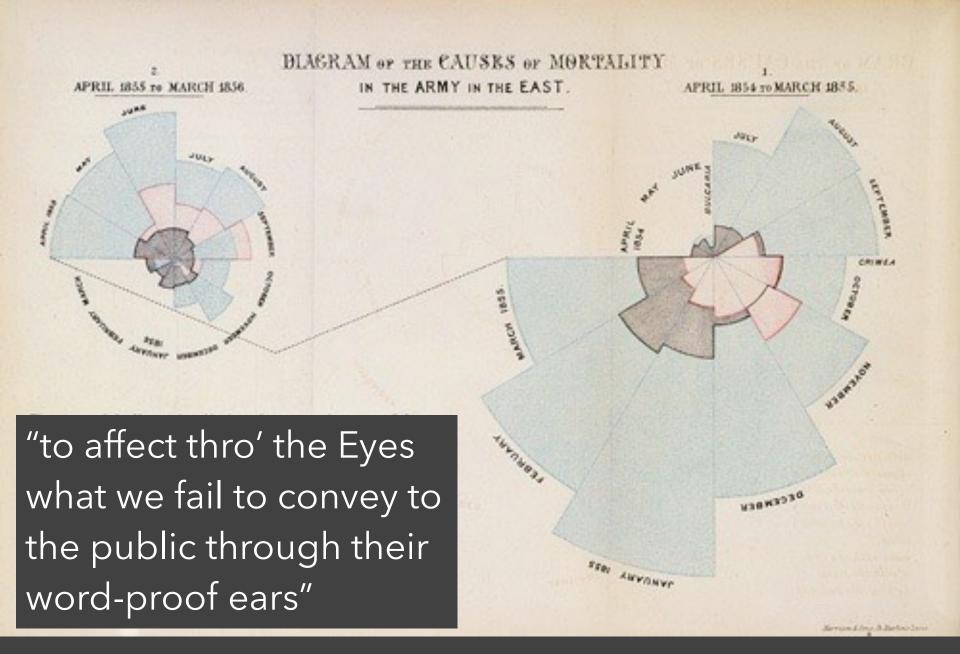
  Collaborate and revise



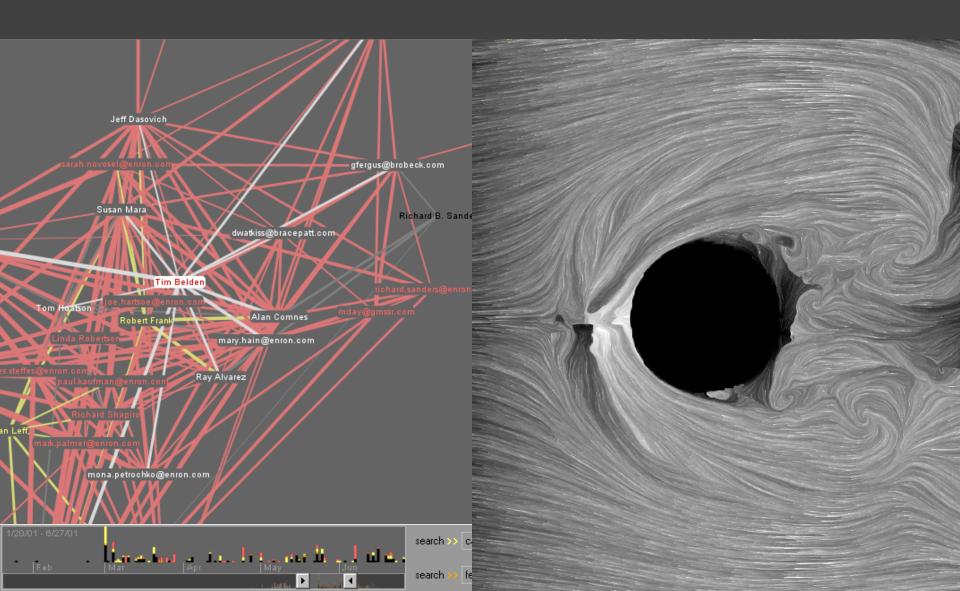
Marey's sphygmograph in use, 1860. La méthode graphique dans les sciences expérimentales et principalement en physiologie et en médecine.

#### Make a Decision: Challenger





#### InfoVis vs. SciVis?



#### Informative vs. Aesthetic?

#### wind map

April 1, 2015

11:35 pm EST (time of forecast download)

top speed: 30.5 mph average: 10.2 mph





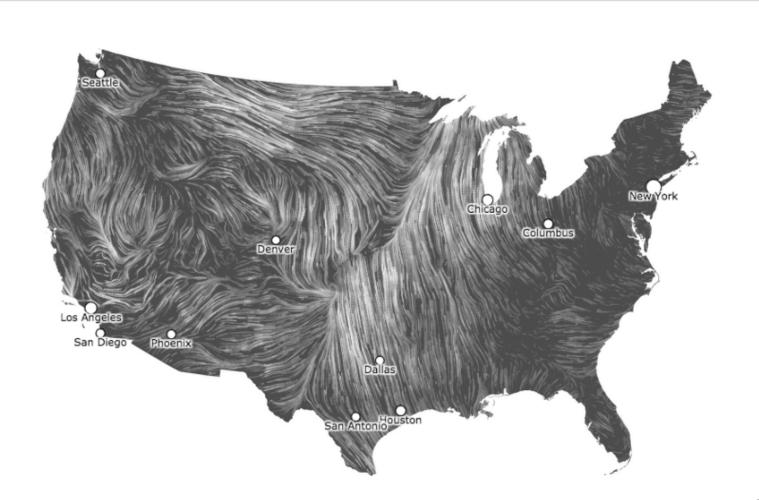












## Data & Image Models

### 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, ...

- N Nominal (labels or categories)
  - · Fruits: apples, oranges, ...
- O Ordered
  - · Quality of meat: Grade A, AA, AAA

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  - Dates: Jan, 19, 2006; Location: (LAT 33.98, LONG -118.45)
  - · Only differences (i.e. intervals) may be compared

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  - · Fruits: apples, oranges, ...
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- Q Interval (location of zero arbitrary)
  - Dates: Jan, 19, 2006; Location: (LAT 33.98, LONG -118.45)
  - Only differences (i.e. intervals) may be compared
- Q Ratio (zero fixed)
  - · Physical measurement: Length, Mass, Temp, ...
  - 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: =,  $\neq$ , <, >, -, %
  - 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)

#### **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

1 2 3 4	year 1850 1850	age O	marst	sex	people
3	1850	0			
4			0	1	1483789
	4050	0	0	2	1450376
	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?

**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

# 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

Projection (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

Sorting (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

Selection (where): filter 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

Aggregation (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	965.45
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	965.45
10/3	MSFT	74.26	74.26
10/4	AMZN	965.45	965.45
10/4	MSFT	74.69	74.26

stock	min
AMZN	965.45
MSFT	74.26

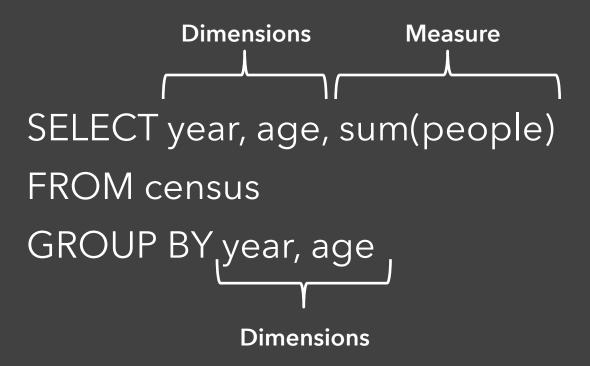
select t.day, t.stock, t.price, a.min
from table as t, aggregate as a
where t.stock = a.stock

Operations on Data Tables: table(s) in, table out Projection (select): select a set of columns Selection (where): filter rows Sorting (order by): order records Aggregation (group by, sum, min, max, ...): partition rows into groups + summarize Combination (join, union, ...): integrate data from multiple tables

### 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

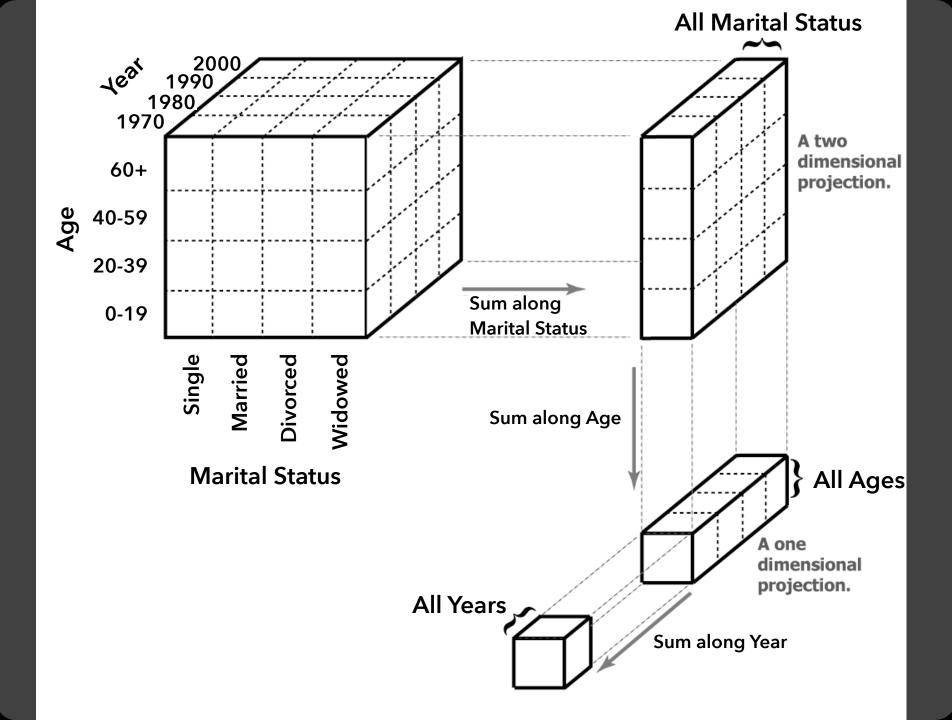
Want to see the breakdown by marital status?

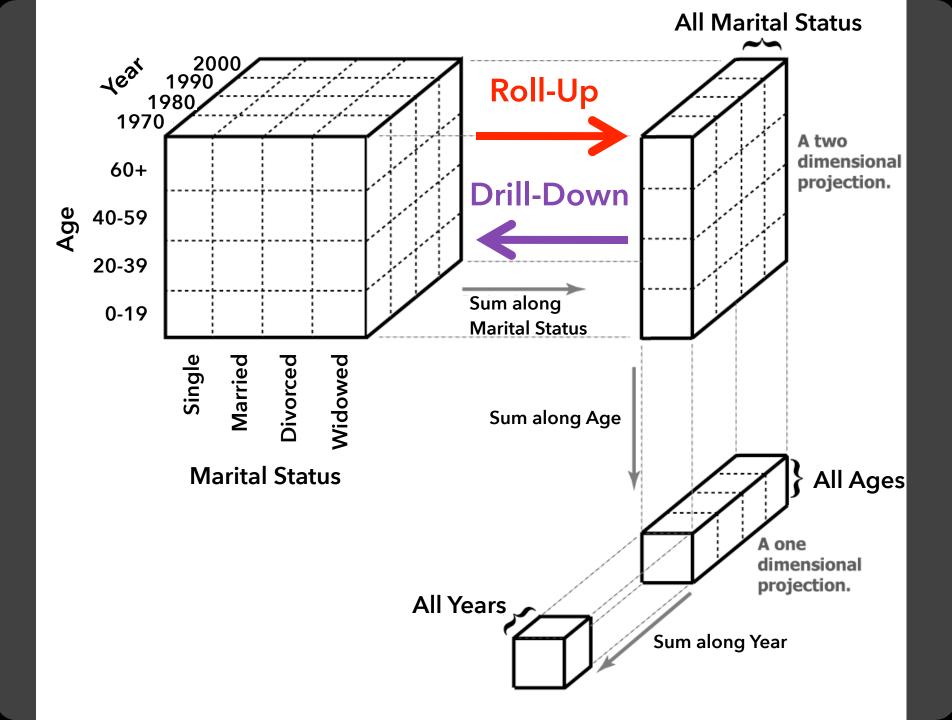
**Drill-down** into additional dimensions

SELECT year, age, marst, sum(people)

FROM census

GROUP BY year, age, marst





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
• • •				

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?

### 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

### A2: Exploratory Data Analysis

Use visualization software to form & answer questions

#### First steps:

Step 1: Pick domain & data

Step 2: Pose questions

Step 3: Profile the data

Iterate as needed

#### **Create visualizations**

Interact with data Refine your questions

#### **Author a report**

Screenshots of most insightful views (10+) Include titles and captions for each view

File Edit View Forms Data Analysis Table Bookmark Collaborate Window Help

Data

Sample - Coffee Ch. 
Files

Sample - Coffee Ch. 
Files

Cohemen 
Froduct Type

Froduct Ty

Due by 5:00pm Monday, Oct 16

### Online Discussion

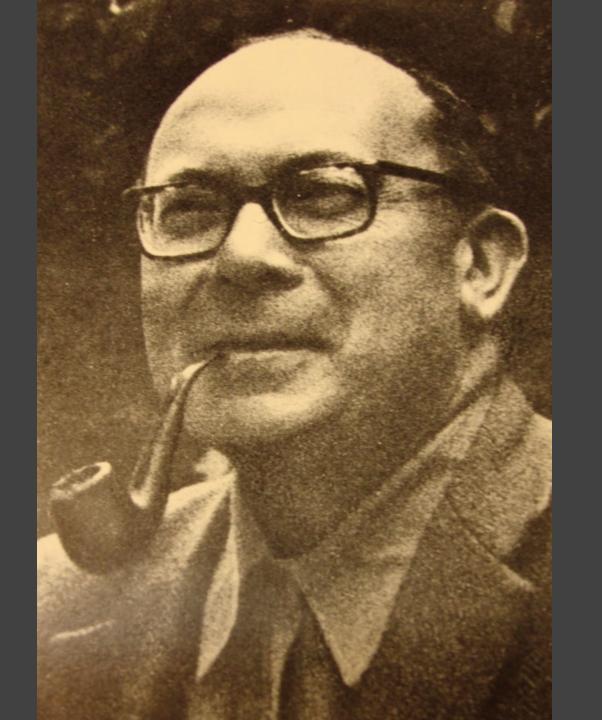
Post discussion comments on class Canvas forum.

One comment per week (ending week 8).

Comments must be posted by Friday 11:59pm.

You have 1 "pass" for the quarter.

# Image Models



### Visual Language is a Sign System



**Jacques Bertin** 

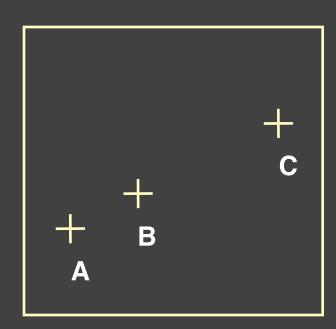
Images perceived as a set of signs

Sender encodes information in signs

Receiver decodes information from signs

Sémiologie Graphique, 1967

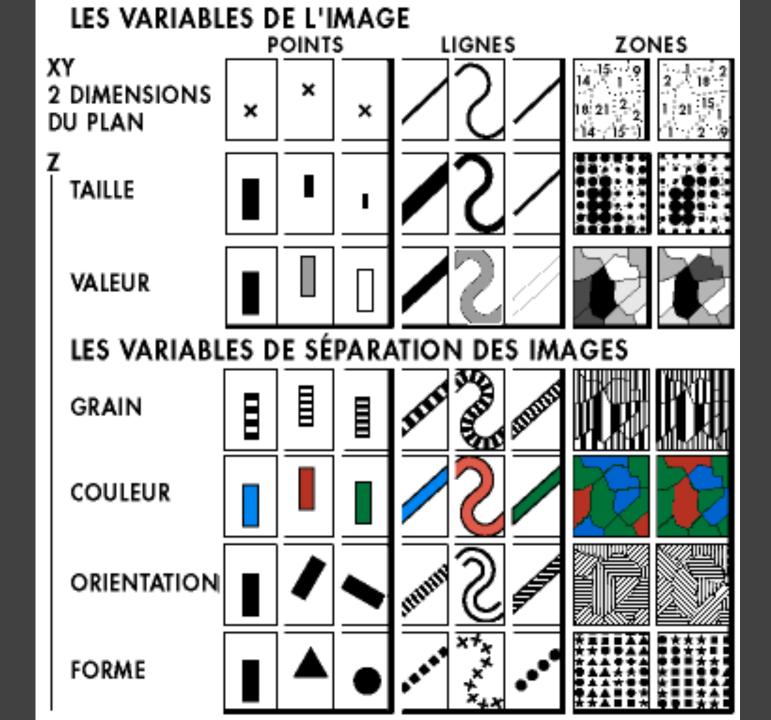
# Bertin's Semiology of Graphics



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

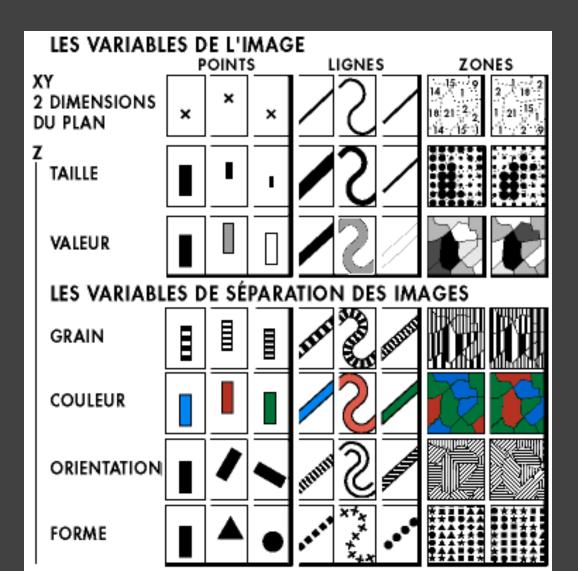
.: Encode quantitative variables

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



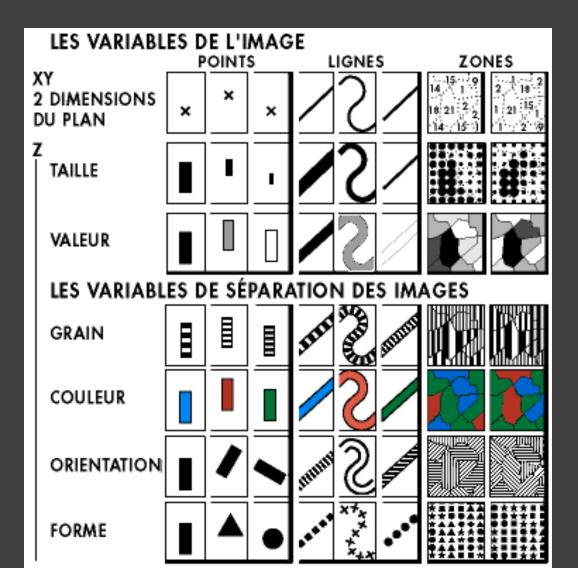
### Visual Encoding Variables

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

N	О	Q
N	О	Q
N	0	Q
N	О	
N		
N		
	N N N	N O N O N O

Shape

**N**ominal

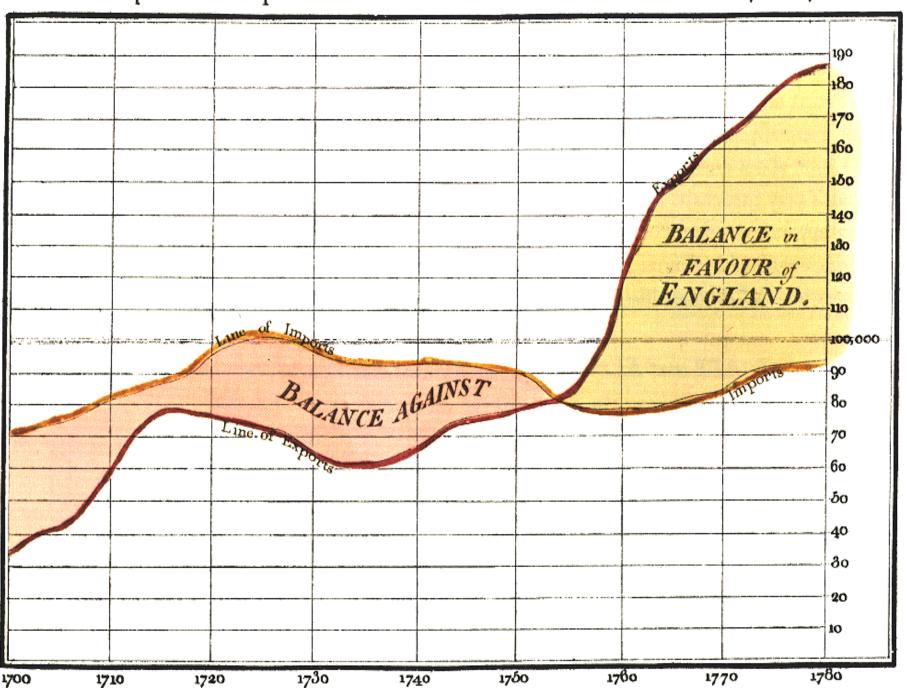
**O**rdinal

**Q**uantitative

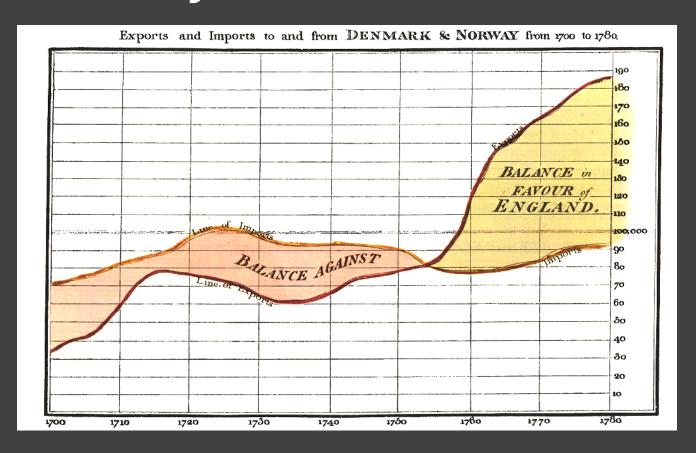
Note: **Q C O C N** 

## Deconstructions

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



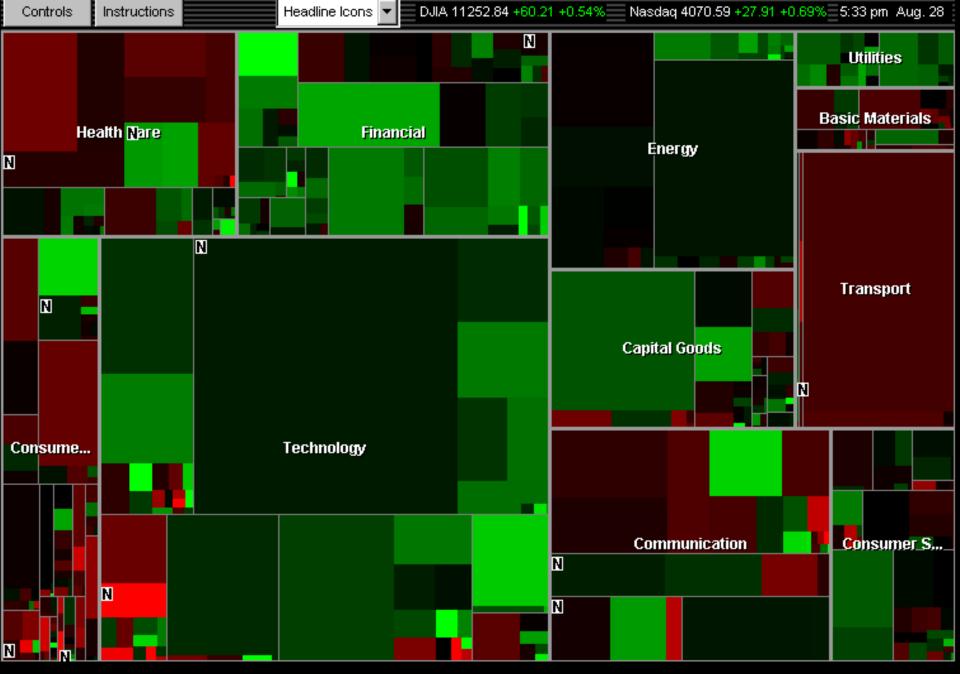
### William Playfair, 1786



X-axis: year (Q)

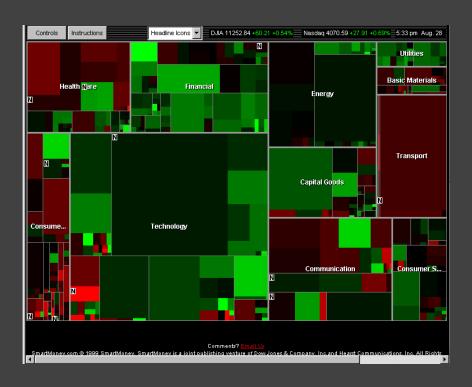
Y-axis: currency (Q)

Color: imports/exports (N, O)



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

### Wattenberg's Map of the Market



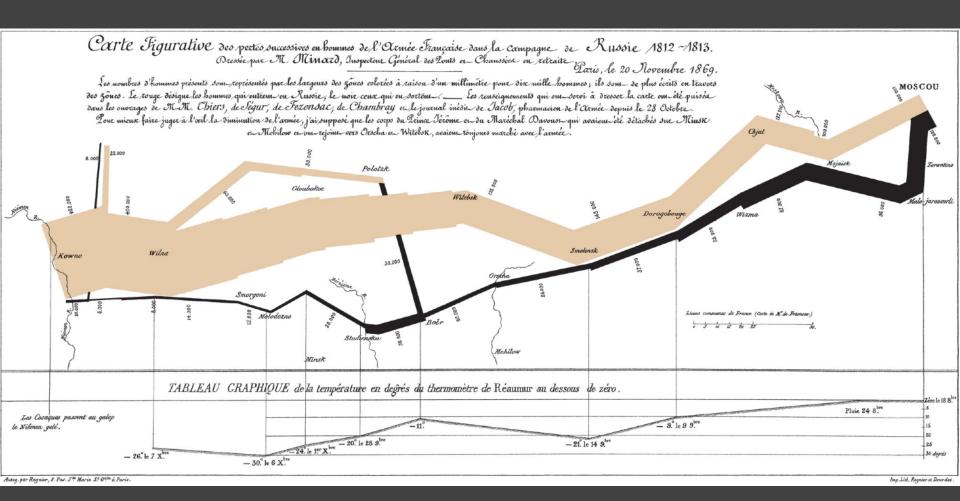
**Rectangle Area**: market cap (Q)

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

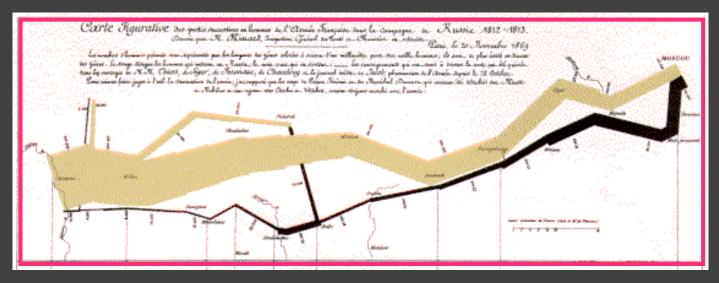
Color Hue: loss vs. gain (N, O)

Color Value: magnitude of loss or gain (Q)

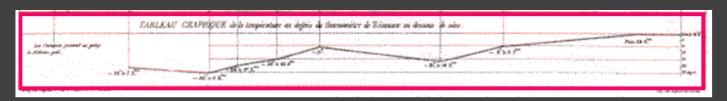
### Minard 1869: Napoleon's March



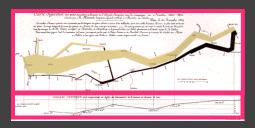
## Single-Axis Composition











# Mark Composition

Y-axis: temperature (Q)

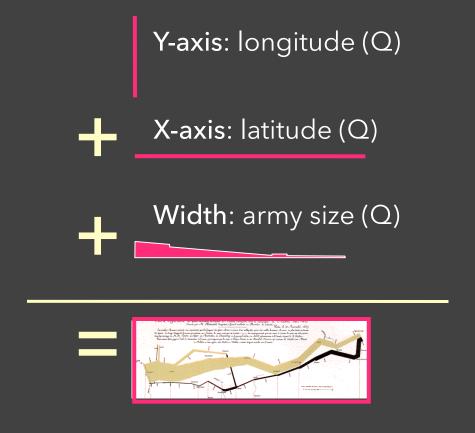


X-axis: longitude (Q) / time (O)



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

# Mark Composition



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

longitude (Q)

latitude (Q)

army size (Q)

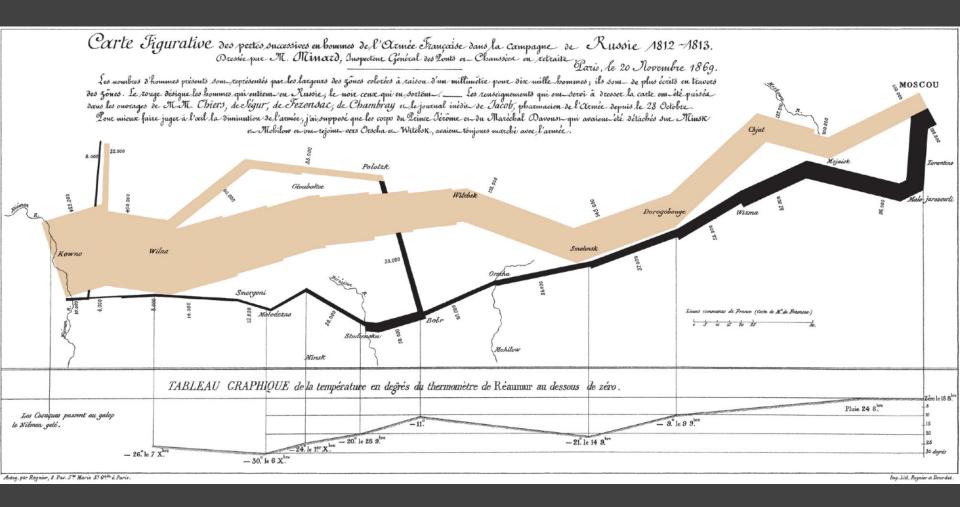




temperature (Q)

latitude (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 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]

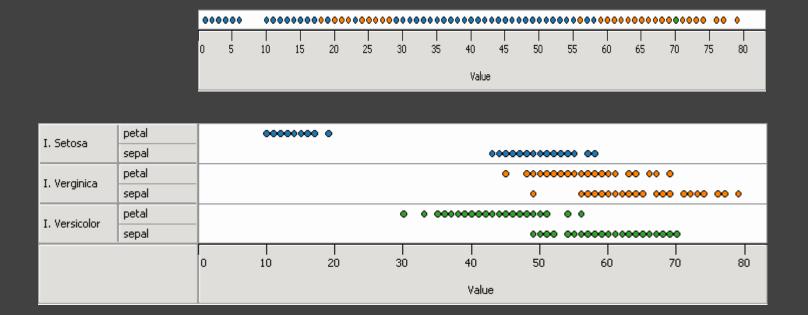
#### Expressiveness

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

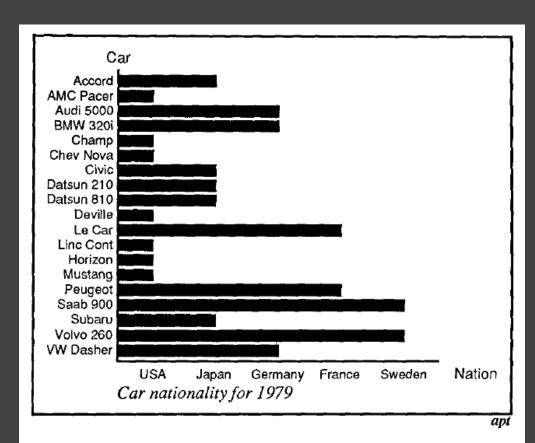


Fig. 11. Incorrect use of a bar chart for the *Nation* relation. The lengths of the bars suggest an ordering on the vertical axis, as if the USA cars were longer or better than the other cars, which is not true for the *Nation* relation.

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**

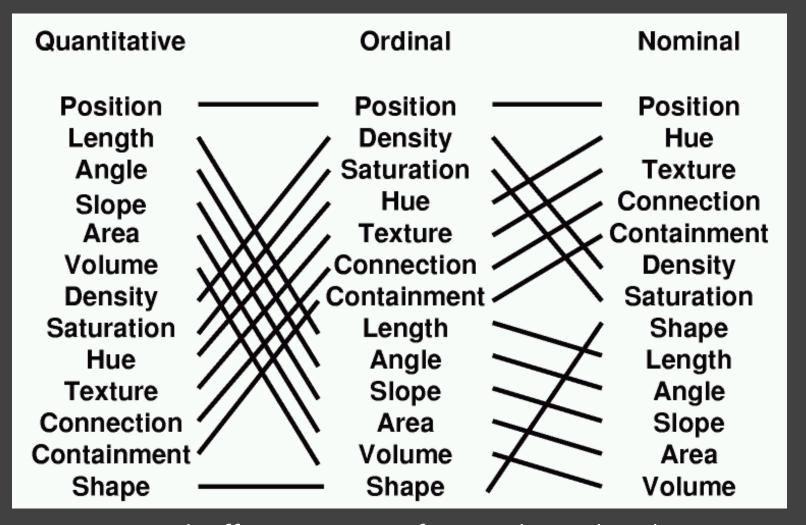
A visualization is more effective than another visualization if the information conveyed by one visualization is more readily perceived than the information in the other visualization.

### 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

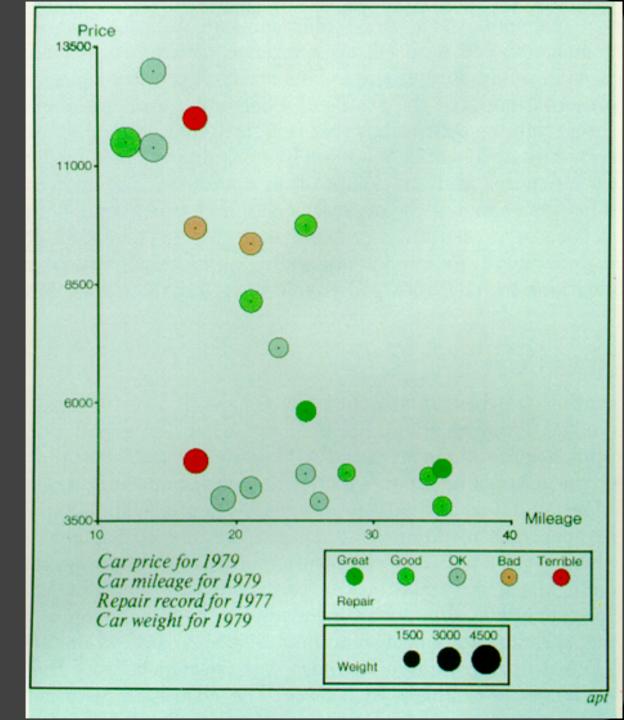
Output the "most effective" visualization

### **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
<u>Also: 3D structure, animation, illustration, ...</u>

Does not consider interaction

Does not consider semantics / conventions

Assumes single visualization as output

### 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...