DATA & IMAGE MODELS

CS 448B | Fall 2023

MANEESH AGRAWALA

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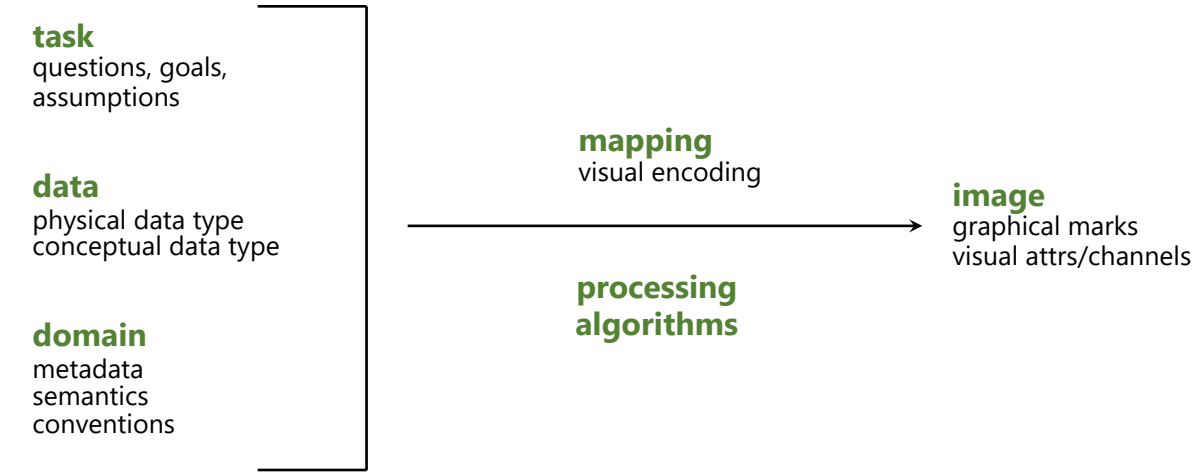
The big picture



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TODAY

Learning Objectives

1. Identify *properties* of data and images
2. Decide how to *encode data using visual attributes/channels*
3. Define concepts of *expressiveness* and *effectiveness*
4. Develop *automated chart design* algorithm

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DATA

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Please select which fun-size Halloween treat you would most want to receive as a trick-or-treater.

Which would you prefer as a trick-or-treater?

Battle: : Candy

Werther's Original Caramel

Pop Rocks

Skip

http://walthickey.com/2017/10/18/whats-the-best-halloween-candy/

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DATA TABLE

Halloween Candy Power Ranking Dataset

	competitorname	chocolate	fruity	caramel	peanutyalmondy	nougat	crispedricewafer	hard	bar	pluribus	sugarpercent	pricepercent	winpercent
2	100 Grand	1	0	1	0	0	1	0	1	0	.73199999	.86000001	66.971725
3	3 Musketeers	1	0	0	0	1	0	0	1	0	.60399997	.51099998	67.602936
4	One dime	0	0	0	0	0	0	0	0	0	.011	.116	32.261086
5	One quarter	0	0	0	0	0	0	0	0	0	.011	.51099998	46.116505
6	Air Heads	0	1	0	0	0	0	0	0	0	.90600002	.51099998	52.341465
7	Almond Joy	1	0	0	1	0	0	0	1	0	.465	.76700002	50.347546
8	Baby Ruth	1	0	1	1	1	0	0	1	0	.60399997	.76700002	56.914547
9	Boston Baked Beans	0	0	0	1	0	0	0	0	1	.31299999	.51099998	23.417824
10	Candy Corn	0	0	0	0	0	0	0	0	1	.90600002	.32499999	38.010963
11	Caramel Apple Pops	0	1	1	0	0	0	0	0	0	.60399997	.32499999	34.517681
12	Charleston Chew	1	0	0	0	1	0	0	1	0	.60399997	.51099998	38.975037
13	Chewey Lemonhead Fruit Mix	0	1	0	0	0	0	0	0	1	.73199999	.51099998	36.017628

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Dataset

Data Field

Halloween Candy Power Ranking Dataset

	competitorname	chocolate	fruity	caramel	peanutyalmondy	nougat	crispedricewafer	hard	bar	pluribus	sugarpercent	pricepercent	winpercent
2	100 Grand	1	0	1	0	0	1	0	1	0	.73199999	.86000001	66.971725
3	3 Musketeers	1	0	0	0	1	0	0	1	0	.60399997	.51099998	67.602936
4	One dime	0	0	0	0	0	0	0	0	0	.011	.116	32.261086
5	One quarter	0	0	0	0	0	0	0	0	0	.011	.51099998	46.116505
6	Air Heads	0	1	0	0	0	0	0	0	0	.90600002	.51099998	52.341465
7	Almond Joy	1	0	0	1	0	0	0	1	0	.465	.76700002	50.347546
8	Baby Ruth	1	0	1	1	1	0	0	1	0	.60399997	.76700002	56.914547
9	Boston Baked Beans	0	0	0	1	0	0	0	0	1	.31299999	.51099998	23.417824
10	Candy Corn	0	0	0	0	0	0	0	0	1	.90600002	.32499999	38.010963
11	Caramel Apple Pops	0	1	1	0	0	0	0	0	0	.60399997	.32499999	34.517681
12	Charleston Chew	1	0	0	0	1	0	0	1	0	.60399997	.51099998	38.975037
13	Chewey Lemonhead Fruit Mix	0	1	0	0	0	0	0	0	1	.73199999	.51099998	36.017628

Data Item/Observation

Cell Value

https://fivethirtyeight.com/videos/the-ultimate-halloween-candy-power-ranking/

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TIDY DATA [Wickham 2014]

How do rows and columns, match up with data fields, and observations?

In *tidy data*

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

Flexible starting point for analysis, transformation, and visualization

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Data models are formal descriptions

Math: Sets with operations on them

Example: integers with + and × operators

Conceptual models are mental constructions

Include semantics and support reasoning

Examples (data vs. conceptual)

1D floats vs. temperature

3D vector of floats vs. spatial location

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DATA MODEL

	competitorname	chocolate	fruity	caramel	peanutyalmondy	nougat	crispedricewafer	hard	bar	pluribus	sugarpercent	pricepercent	winpercent
2	100 Grand	1	0	1	0	0	1	0	1	0	.73199999	.86000001	66.971725
3	3 Musketeers	1	0	0	0	1	0	0	1	0	.60399997	.51099998	67.602936
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6	Air Heads	0	1	0	0	0	0	0	0	0	.90600002	.51099998	52.341465
7	Almond Joy	1	0	0	1	0	0	0	1	0	.465	.76700002	50.347546
8	Baby Ruth	1	0	1	1	1	0	0	1	0	.60399997	.76700002	56.914547
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13	Chewey Lemonhead Fruit Mix	0	1	0	0	0	0	0	0	1	.73199999	.51099998	36.017628
	string	bool	bool	bool	bool	bool	bool	bool	bool	bool	float	float	float

How is data stored in the database?

https://github.com/fivethirtyeight/data/tree/master/candy-power-ranking

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CONCEPTUAL MODEL

Header	Description
chocolate	Does it contain chocolate?
fruity	Is it fruit flavored?
caramel	Is there caramel in the candy?
peanutalmondy	Does it contain peanuts, peanut butter or almonds?
nougat	Does it contain nougat?
crispedricewafer	Does it contain crisped rice, wafers, or a cookie component?
hard	Is it a hard candy?
bar	Is it a candy bar?
pluribus	Is it one of many candies in a bag or box?
sugarpercent	The percentile of sugar it falls under within the data set.
pricepercent	The unit price percentile compared to the rest of the set.
winpercent	The overall win percentage according to 269,000 matchups.

https://github.com/fivethirtyeight/data/tree/master/candy-power-ranking

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crispedricewafer	Does it contain crisped rice or cookies?
hard	Is it a hard candy?
bar	Is it a candy bar?
pluribus	Is it one of many candies in a bad?
sugarpercent	The percentile of sugar (across dataset)
pricepercent	The unit price percentile (across dataset)
winpercent	The overall win percentage in 269K contests

Domain specific understanding about the data

Supports analysis and reasoning

https://github.com/fivethirtyeight/data/tree/master/candy-power-ranking

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DATA TYPES

N - Nominal (labels)
Fruits: Apples, oranges, …
Operations: =, ≠

O - Ordered
Quality of meat: Grade A, AA, AAA
Operations: =, ≠, <, >

Q - Interval (location of zero arbitrary)
Dates: Jan, 19, 2016; Loc.: (LAT 33.98, LON -118.45)
Like a geometric point. Cannot compare directly
Only differences (i.e. intervals) may be compared
Operations =, ≠, <, >, -

Q - Ratio (location of zero fixed)
Physical measurement: Length, Mass, …
Counts and amounts
Like a geometric vector, origin is meaningful
Operations: =, ≠, <, >, -, ÷

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NOMINAL, ORDINAL, QUANTITATIVE

Header	Description	
competitorname	Name of candy	N
chocolate	Does it contain chocolate?	N (maybe O)
fruity	Is it fruit flavored?	N (maybe O)
caramel	Is there caramel in the candy?	N (maybe O)
peanutalmondy	Does it contain peanuts or almonds?	N (maybe O)
nougat	Does it contain nougat?	N (maybe O)
crispedricewafer	Does it contain crisped rice or cookies?	N (maybe O)
hard	Is it a hard candy?	N (maybe O)
bar	Is it a candy bar?	N (maybe O)
pluribus	Is it one of many candies in a bad?	N (maybe O)
sugarpercent	The percentile of sugar (across dataset)	Q-Ratio
pricepercent	The unit price percentile (across dataset)	Q-Ratio
winpercent	The overall win percentage in 269K contests	Q-Ratio

https://github.com/fivethirtyeight/data/tree/master/candy-power-ranking

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DATA TYPES

DIMENSIONS

Dimensions are often the **independent** variables

Dimensions contain **qualitative values that describe the data item** (such as names, dates, or geographical data)

MEASURES

Measures are often the **dependent** variables

Measures contain numeric, **quantitative values that you can measure**. Measures can be aggregated (sum, count, average, std. deviation).

	competitorname	chocolate	fruity	caramel	peanutyalmondy	nougat	crispedricewafer	hard	bar	pluribus	sugarpercent	pricepercent	winpercent
2	100 Grand	1	0	1	0	0	1	0	1	0	.73199999	.86000001	66.971725
3	3 Musketeers	1	0	0	0	1	0	0	1	0	.60399997	.51099998	67.602936
4	One dime	0	0	0	0	0	0	0	0	0	.011	.116	32.261086

NOTE: Distinction is not strict. The same variable may be treated either way depending on the task

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DIMENSION OR MEASURE

Header	Description
competitorname	Name of candy
chocolate	Does it contain chocolate?
fruity	Is it fruit flavored?
caramel	Is there caramel in the candy?
peanutalmondy	Does it contain peanuts or almonds?
nougat	Does it contain nougat?
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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, …

2348 data points

	A	B	C	D	E
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

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CENSUS N, O, Q

People Count: **Q-Ratio**
Year: **Q-Interval** (maybe O)
Age: **Q-Ratio** (maybe O)
Sex: **N**
Marital Status: **N**

	A	B	C	D	E
1	year	age	marst	sex	people
2	1850	0	0	1	1483789
3	1850	0	0	2	1450376
4	1850	5	0	1	1411067
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24	1850	55	0	1	194080
25	1850	55	0	2	187208

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CENSUS DIM., MEAS.

People Count: Measure
Year: Dimension
Age: Depends!
Sex: Dimension
Marital Status: Dimension

	A	B	C	D	E
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
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12	1850	25	0	1	862547
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14	1850	30	0	1	730638
15	1850	30	0	2	639636
16	1850	35	0	1	588487
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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

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DATA TABLES & TRANSFORMATIONS

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RELATIONAL ALGEBRA [Codd 1970] / SQL

Operations on data tables: table(s) in, table out

Projection (SELECT) – select a set of columns

Selection (WHERE) – filter rows

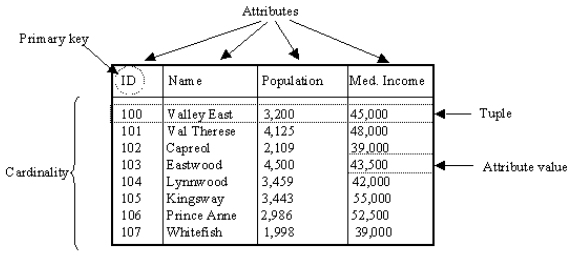
Sorting (ORDER BY) – order rows

Aggregation (GROUP BY, SUM, MIN, …)

partition rows into groups and summarize

Combination (JOIN, UNION, …)

integrate data from multiple tables



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RELATIONAL ALGEBRA [Codd 1970] / SQL

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

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RELATIONAL ALGEBRA [Codd 1970] / SQL

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

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RELATIONAL ALGEBRA [Codd 1970] / SQL

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

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RELATIONAL ALGEBRA [Codd 1970] / SQL

Aggregation (GROUP BY, SUM, MIN, …)

```
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

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RELATIONAL ALGEBRA [Codd 1970] / SQL

Combination (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

stock	min
AMZN	957.10
MSFT	74.26

→

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

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

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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?

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ANNOUNCEMENTS

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CLASS PARTICIPATION REQUIREMENTS

Complete required **readings** and **notebooks** before class

Attend class and be a part of the in-class discussion

Post at least 1 discussion substantive comment/question per week

Due by 8pm the following Sunday
1 free pass for the quarter

Class home page
https://magrawala.github.io/cs448b-fa23/

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READING/NOTEBOOK/LECTURE RESPONSES

Good responses typically exhibit one or more

Critiques of arguments made in the papers/lectures
Analysis of implications or future directions for ideas in readings/lectures
Insightful questions about the readings/lectures

Responses should not be summaries

Should be substantive (1-2 paragraphs is typical)

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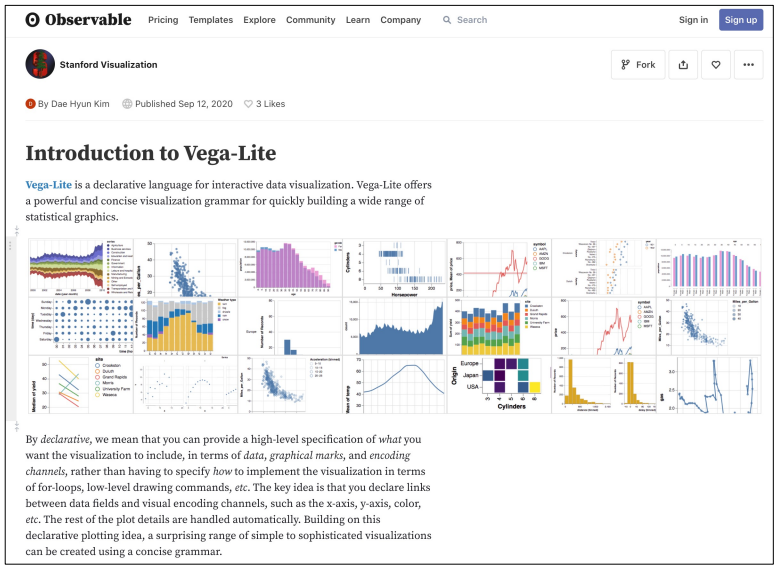
OBSERVABLE NOTEBOOKS / VEGA-LITE



Vega-Lite is a *declarative* API for programming visualizations

Do the exercises (fork notebook)

Lec. on Wed 10/4 will assume you have done the 1st three notebooks

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ASSIGNMENT 1: VISUALIZATION DESIGN

Due TODAY

Design a static visualization for a data set

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

Data: Stanford Undergraduate Majors

Stanford University publishes a variety of datasets through the Stanford Institutional Rsearch & Decision Support website. They have published a data table containing information about the number of Stanford undergraduates obtaining a Bachelor's degree in 75 different fields of study from 2003 to 2022. We have filtered and wrangled this data to the top 10 fields of study by cummulative degrees conferred over the time period to produce a dataset with the following attributes:

- **Year:** Academic year between 2003 and 2022. (Academic years run July-June so Year=2003 covers July 2002 to June 2003.)
- **FieldOfStudy:** Field in which degree was obtained.
- **Count:** Number of students earning a Bachelor's degree.

The extracted dataset is available in csv format: TopFieldsStanfordBachelors.csv

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ASSIGNMENT 2: EXP. DATA ANALYSIS

Due 10/16 11:30am

Use Tableau or Vega-Lite to formulate & answer data questions

First steps
- Step 1: Pick domain & data
- Step 2: Pose questions
- Step 3: Profile data
- Iterate as needed

Create visualizations
- See different views of data
- Refine questions

Author a report
- Screenshots of most insightful views (8+)
- Include titles and captions for each view

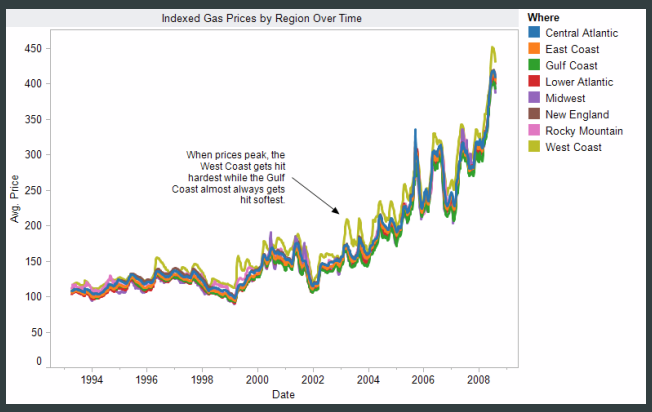

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IMAGE

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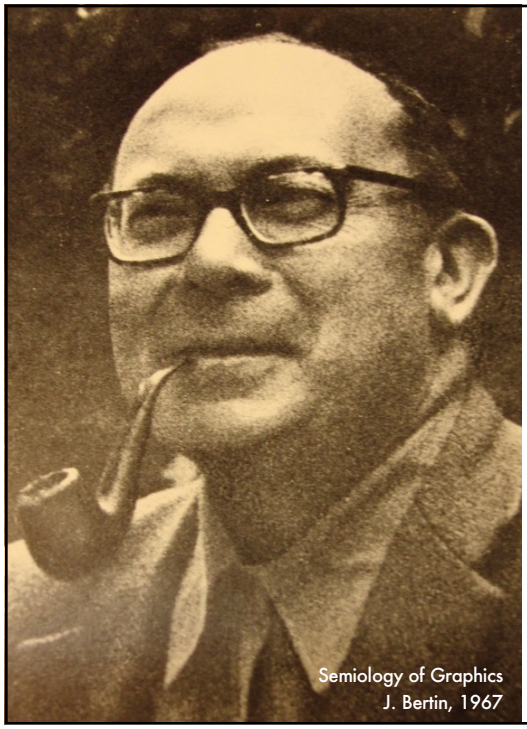


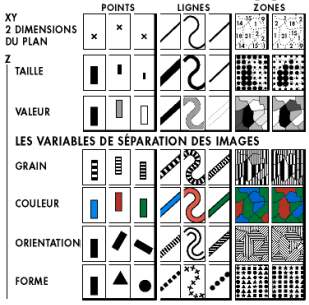
MARKS & VISUAL ATTRs

Marks: geometric primitives

points lines areas

Visual Attributes: control mark appearance

Position (2x)
Size
Value
Texture
Color
Orientation
Shape



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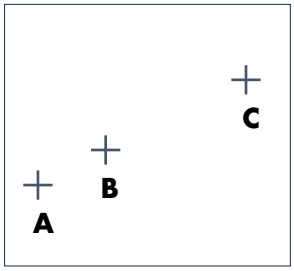
CODING INFORMATION IN POSITION



1. A, B, C are distinguishable
2. Three points are colinear: B 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

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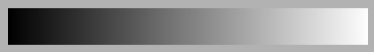
CODING INFORMATION IN COLOR

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

50

BERTIN'S "LEVELS OF ORGANIZATION"

Position	**N**	**O**	**Q**
Size	**N**	**O**	**Q**
Value	**N**	**O**	**Q**
Texture	**N**	**O**	
Color	**N**		
Orientation	**N**		
Shape	**N**		

N Nominal
O Ordered
Q Quantitative

Note: Q ⊂ O ⊂ N

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VISUAL ENCODING

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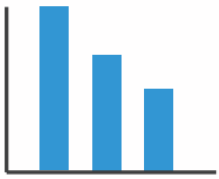
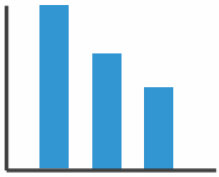
ENCODINGS: MAP DATA to MARK ATTRIBUTES



mark: rect
data → size (height)

53

ENCODINGS: MAP DATA to MARK ATTRIBUTES

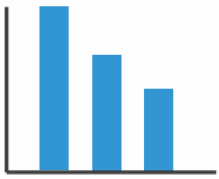


mark: rect
data → size (height)

mark: points
$data_1$ → x-pos
$data_2$ → y-pos

54

ENCODINGS: MAP DATA to MARK ATTRIBUTES



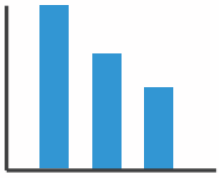
mark: rect
data → size (height)

mark: points
$data_1$ → x-pos
$data_2$ → y-pos

mark: points
$data_1$ → x-pos
$data_2$ → y-pos
$data_3$ → color

55

ENCODINGS: MAP DATA to MARK ATTRIBUTES



mark: rect
data → size (height)

mark: points
$data_1$ → x-pos
$data_2$ → y-pos

mark: points
$data_1$ → x-pos
$data_2$ → y-pos
$data_3$ → color

mark: points
$data_1$ → x-pos
$data_2$ → y-pos
$data_3$ → color
$data_4$ → size

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DECONSTRUCTIONS

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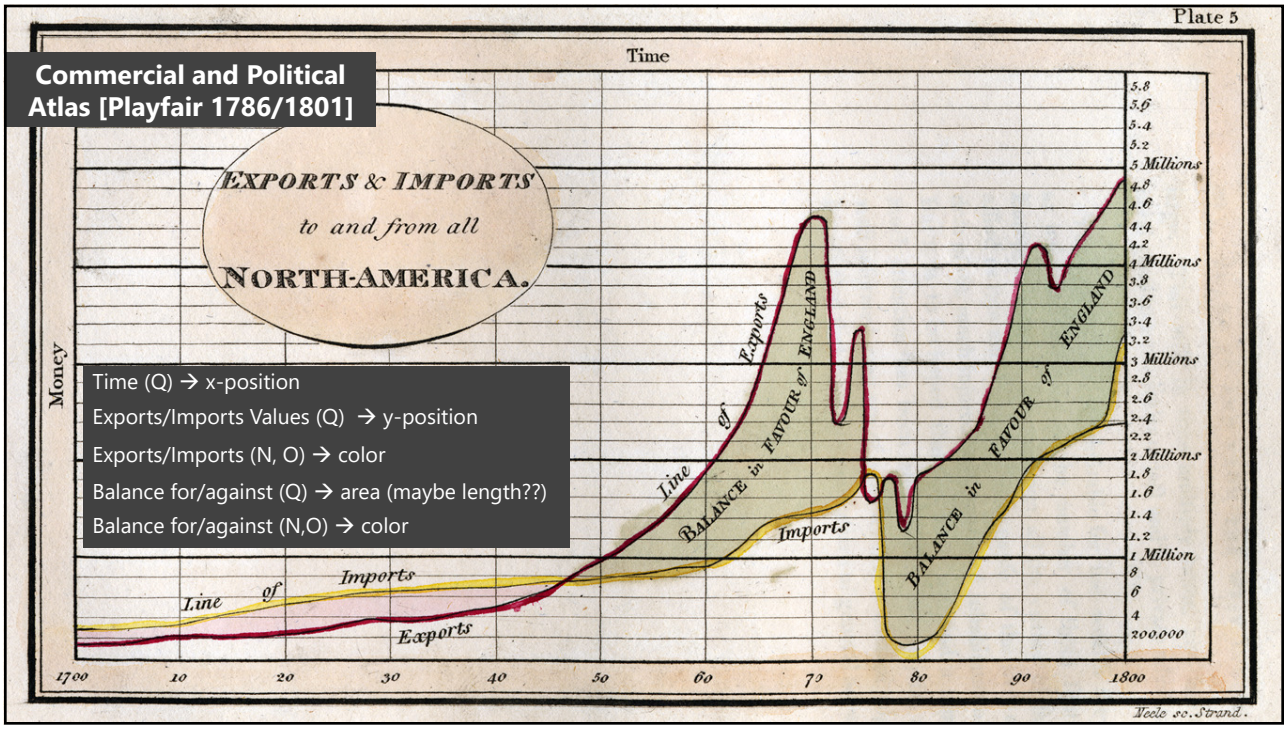
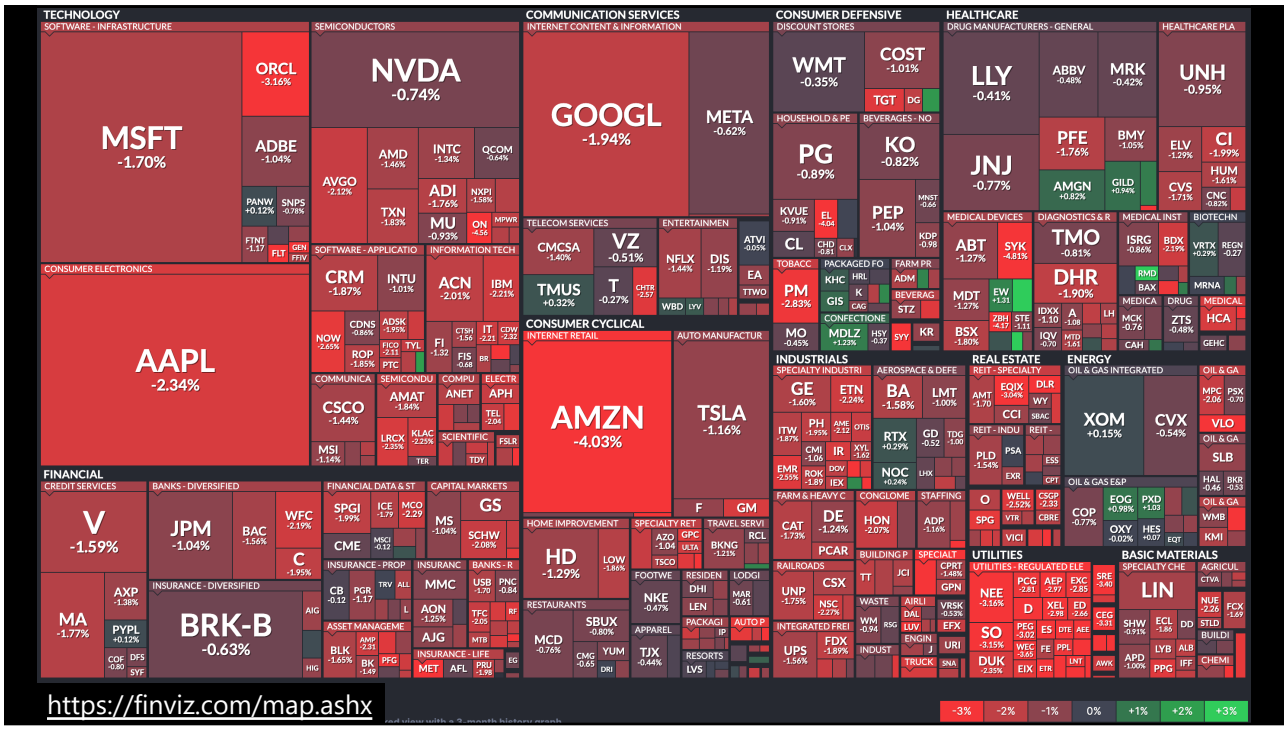

Commercial and Political Atlas [Playfair 1786/1801]
Plate 5
Time
EXPORTS & IMPORTS
to and from all
NORTH-AMERICA.
Money
Time (Q) → x-position
Exports/Imports Values (Q) → y-position
Exports/Imports (N, O) → color
Balance for/against (Q) → area (maybe length??)
Balance for/against (N,O) → color
Line of Imports
Exports
Line of Exports
BALANCE in FAVOUR of ENGLAND
Imports
BALANCE in FAVOUR of ENGLAND
5 Millions
4 Millions
3 Millions
2 Millions
1 Million
200,000
1700
10
20
30
40
50
60
70
80
90
1800
Neele sc. Strand.

60


TECHNOLOGY
SOFTWARE - INFRASTRUCTURE
MSFT -1.70%
ORCL -3.16%
ADBE -1.04%
PANW +0.12%
SNPS -0.78%
FTNT -1.17%
GEN
FLT
FFIV
CONSUMER ELECTRONICS
AAPL -2.34%
SEMICONDUCTORS
NVDA -0.74%
AVGO -2.12%
AMD -1.46%
INTC -1.34%
QCOM -0.64%
TXN -1.83%
ADI -1.76%
NXPI -1.58%
MU -0.93%
ON -4.56%
MPWR
SOFTWARE - APPLICATIO
INFORMATION TECH
CRM -1.87%
INTU -1.01%
ACN -2.01%
IBM -2.21%
NOW -2.65%
CDNS -0.86%
ADSK -1.95%
ROP -1.85%
FICO -2.11%
TYL
FI -1.32
CTSH -1.56
IT -2.21
CDW -2.32
FIS -0.68
BR
COMMUNICA
SEMICONDU
COMPU
ELECTR
CSCO -1.44%
AMAT -1.84%
ANET
APH
TEL -2.04
MSI -1.14%
LRCX -2.35%
KLAC -2.22%
TER
SCIENTIFIC
TDY
FSLR
FINANCIAL
CREDIT SERVICES
BANKS - DIVERSIFIED
V -1.59%
JPM -1.04%
BAC -1.56%
WFC -2.19%
C -1.95%
AXP -1.38%
MA -1.77%
PYPL +0.12%
COF -0.80
DFS
SYF
INSURANCE - DIVERSIFIED
BRK-B -0.63%
AIG
HIG
FINANCIAL DATA & ST
CAPITAL MARKETS
SPGI -1.99%
ICE -1.79
MCO -2.29
GS
MS -1.04%
SCHW -2.08%
CME
MSCI -0.12
INSURANCE - PROP
INSURANC
BANKS - R
CB -0.12
PGR -1.17
TRV
ALL
L
MMC
USB -1.70
PNC -0.84
TFC -2.05
RF
AON -1.25%
MTB
AJG
ASSET MANAGEME
INSURANCE - LIFE
BLK -1.65%
AMP -2.31
BK -1.49
PFG
MET
AFL
PRU -1.98
EG
COMMUNICATION SERVICES
INTERNET CONTENT & INFORMATION
GOOGL -1.94%
META -0.62%
TELECOM SERVICES
ENTERTAINMEN
CMCSA -1.40%
VZ -0.51%
TMUS +0.32%
T -0.27%
CHTR -2.57
NFLX -1.44%
DIS -1.19%
WBD
LYV
ATVI -0.05%
EA
TTWO
CONSUMER CYCLICAL
INTERNET RETAIL
AUTO MANUFACTUR
AMZN -4.03%
TSLA -1.16%
F
GM
HOME IMPROVEMENT
SPECIALTY RET
TRAVEL SERVI
HD -1.29%
LOW -1.86%
AZO -1.04
GPC
ULTA
TSCO
BKNG -1.21%
RCL
FOOTWE
RESIDEN
LODGI
NKE -0.47%
DHI
LEN
MAR -0.61
RESTAURANTS
APPAREL
PACKAGI
IP
AUTO P
SBUX -0.80%
MCD -0.76%
CMG -0.65
YUM
DRI
TJX -0.44%
RESORTS
LVS
CONSUMER DEFENSIVE
DISCOUNT STORES
WMT -0.35%
COST -1.01%
TGT
DG
HOUSEHOLD & PE
BEVERAGES - NO
PG -0.89%
KO -0.82%
KVUE -0.91%
EL -4.04
CHD -0.81
CLX
PEP -1.04%
MNST -0.66
KDP -0.98
TOBACC
PACKAGED FO
FARM PR
CL
KHC
HRL
K
ADM
BEVERAG
PM -2.83%
GIS
CAG
STZ
CONFECTIONE
MDLZ +1.23%
HSY -0.37
SYY
KR
MO -0.45%
BSX -1.80%
INDUSTRIALS
SPECIALTY INDUSTRI
AEROSPACE & DEFE
GE -1.60%
ETN -2.24%
BA -1.58%
LMT 1.00%
ITW -1.87%
PH -1.95%
AME -2.12
OTIS
RTX +0.29%
GD -0.32
TDG 1.00
EMR -2.55%
ROK -1.89
DOV
IEX
CMI -1.06
IR
XYL -1.62
NOC +0.24%
LHX
FARM & HEAVY C
CONGLOME
STAFFING
CAT -1.73%
DE 1.24%
HON -2.07%
ADP -1.16%
PCAR
RAILROADS
BUILDING P
SPECIALT
UNP -1.75%
CSX
NSC -2.27%
TT
JCI
CPRT -1.48%
GPN
INTEGRATED FREI
WASTE
AIRLI
VRSK -0.53%
EFX
UPS -1.56%
FDX -1.89%
WM -0.94
RSG
LUV
DAL
ENGIN
J
URI
INDUST
TRUCK
SNA
HEALTHCARE
DRUG MANUFACTURERS - GENERAL
LLY -0.41%
ABBV -0.48%
MRK -0.42%
JNJ -0.77%
PFE -1.76%
BMY -1.05%
AMGN +0.82%
GILD +0.96%
HEALTHCARE PLA
UNH -0.95%
ELV -1.29%
CI -1.99%
CVS -1.71%
HUM -1.61%
CNC -0.82%
MEDICAL DEVICES
DIAGNOSTICS & R
MEDICAL INST
BIOTECHN
ABT -1.27%
SYK -4.81%
TMO -0.81%
ISRG -0.86%
BDX -2.19%
VRTX +0.29%
REGN -0.27
DHR -1.90%
RMD
BAX
MRNA
MDT -1.27%
EW +1.31
IDXX -1.10
A -1.08
LH
MEDICA
DRUG
MEDICAL
MCK -0.76
ZTS -0.48%
HCA
ZBH -4.17
STE -1.11
IQV -0.70
MTD -1.61
BSX -1.80%
CAH
GEHC
REAL ESTATE
REIT - SPECIALTY
EQIX -2.04%
DLR
AMT -1.70
WY
CCI
SBAC
REIT - INDU
REIT -
PLD -1.54%
PSA
ESS
EXR
CPT
O
WELL -2.52%
CSGP -2.33
SPG
VTR
CBRE
VICI
ENERGY
OIL & GAS INTEGRATED
XOM +0.15%
CVX -0.54%
OIL & GA
MPC -2.06
PSX -0.79
VLO
SLB
OIL & GAS E&P
COP -0.77%
EOG +0.98%
PXD +1.03
OXY -0.02%
HES +0.07
EQT
HAL -0.46
BKR -0.53
KMI
WMB
UTILITIES
UTILITIES - REGULATED ELE
NEE -3.16%
PCG -2.81
AEP -2.97
EXC -2.85
SRE -1.40
D
XEL -2.99
ED -2.66
CEG -3.31
SO -3.15%
PEG -3.02
ES
DTE
AEE
DUK -2.35%
WEC -3.01
FE
PPL
EIX
ETR
LNT
AWK
BASIC MATERIALS
SPECIALTY CHE
AGRICUL
LIN
CTVA
NUE -2.26
FCX -1.69
STLD
BUILDI
SHW -0.91%
ECL -1.86
DD
LYB
ALB
APD -1.00%
PPG
IFF
CHEMI
https://finviz.com/map.ashx
-3%
-2%
-1%
0%
+1%
+2%
+3%

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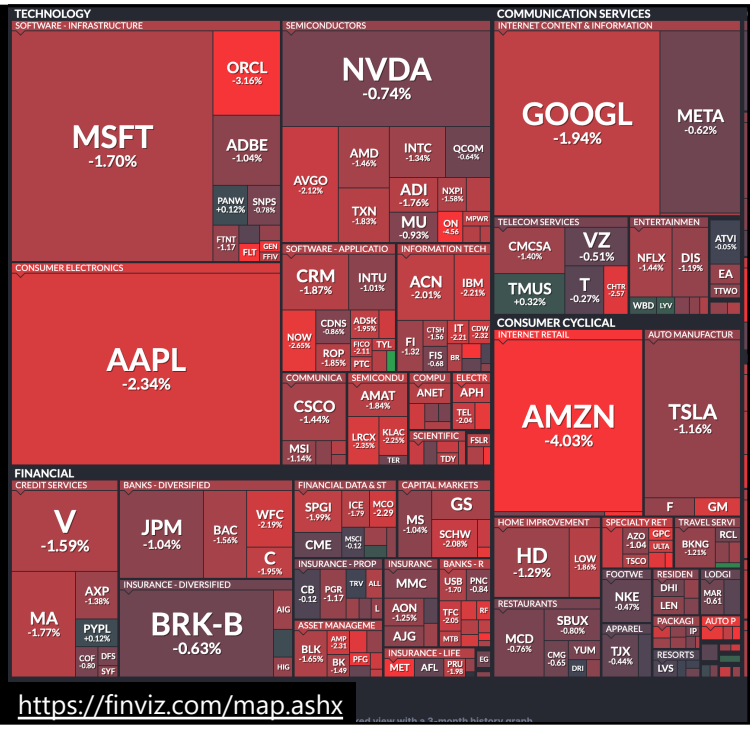
ENCODINGS

market cap (Q) → rectangle size
mkt sector (N), mkt cap (Q) → rect. pos.
loss vs. gain (N, O) → color hue
magnitude of loss or gain (Q) → color value



https://finviz.com/map.ashx

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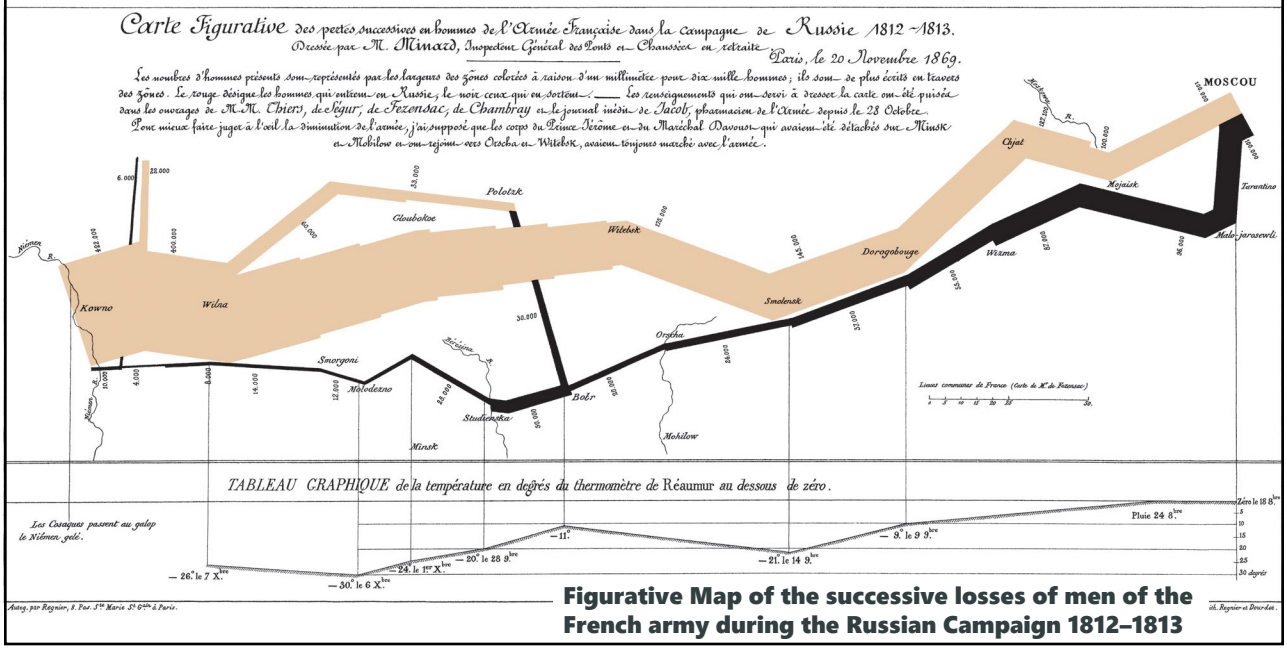
MINARD's MARCH on MOSCOW



Figurative Map of the successive losses of men of the French army during the Russian Campaign 1812–1813

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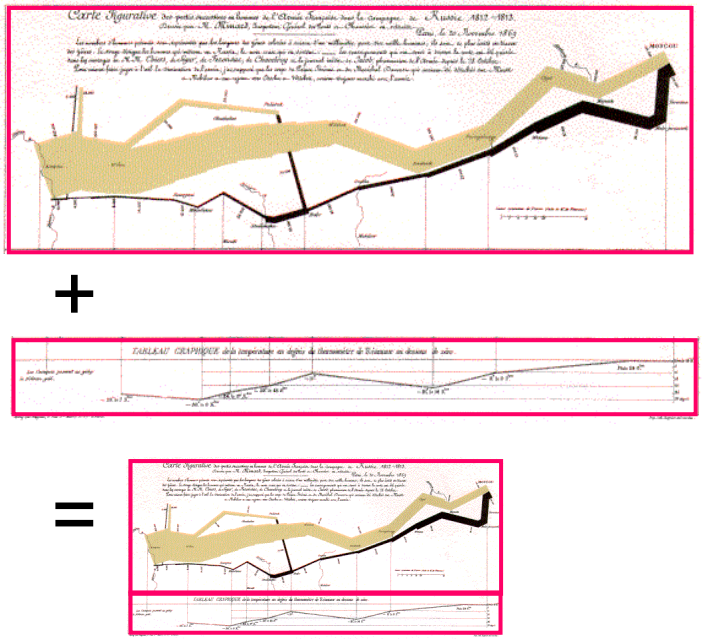

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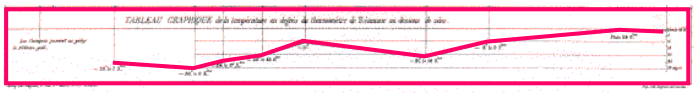
MARK COMPOSITION

temperature (Q) → y-position

\+

longitude (Q), time (O) → x-position

=



temp across space & time (Q x Q,O)

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MARK COMPOSITION

latitude (Q) → y-position

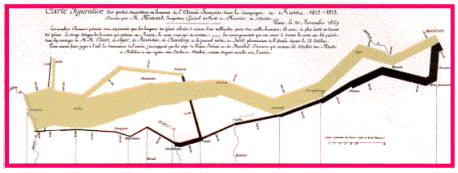
\+

longitude (Q) → x-position

\+

army size (Q) → width

=



army position (Q x Q) and army size (Q)

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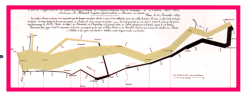
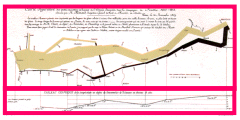
latitude (Q)

longitude (Q)

army size (Q)





temperature (Q)

longitude (Q), time(O)

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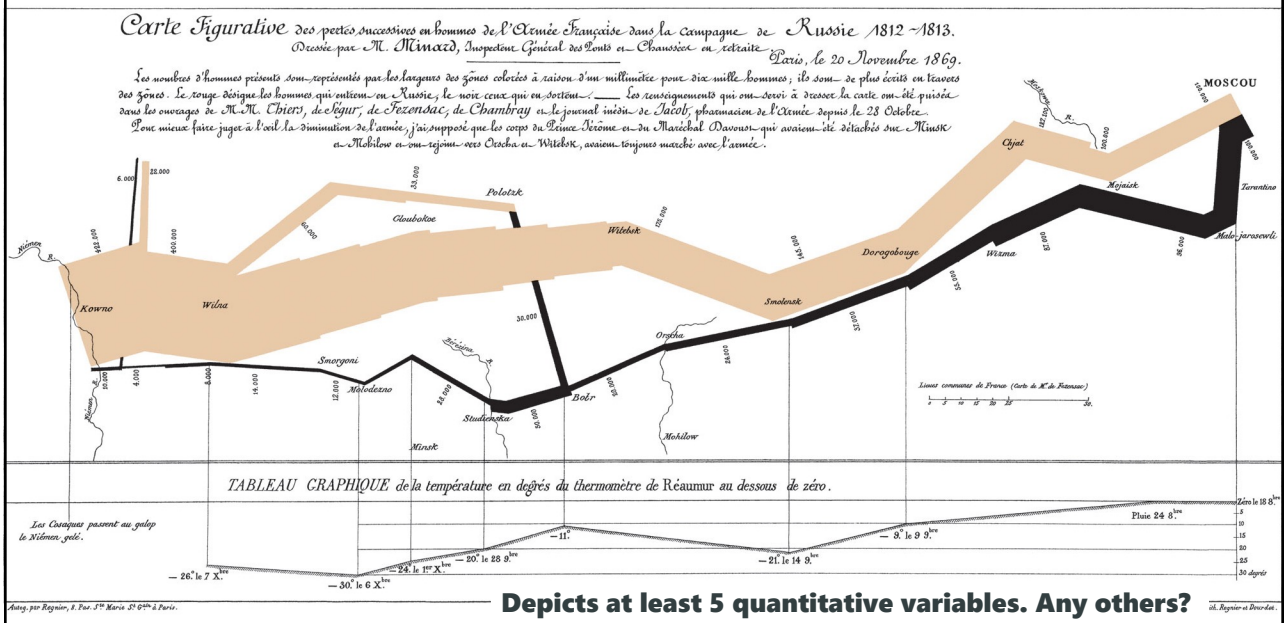
MINARD's MARCH on MOSCOW



Depicts at least 5 quantitative variables. Any others?

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FORMALIZING DESIGN

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COMBINATORICS OF ENCODINGS

Challenge:

Assume k visual attributes/channels and n data fields

Pick the best encoding from the exponential number of possibilities $(n+1)^k$

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PRINCIPLES

Challenge

Assume k visual attributes/channels and n data fields

Pick the best encoding from the exponential number of possibilities $(n+1)^k$

Principle of Consistency

Properties of image (visual variables) should match properties of data

Principle of Importance Ordering

Encode most important information in the most effective way

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EXPRESSIVENESS CRITERIA [Mackinlay 1986]

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.

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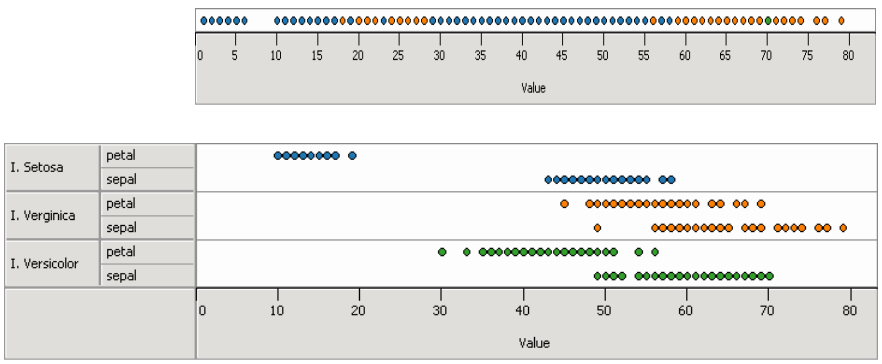
CANNOT EXPRESS ALL THE FACTS

Horizontal dot plot

A one-to-many (1 → N) relation cannot be expressed in a single horizontal dot plot because multiple tuples are mapped to the same position



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EXPRESSES FACTS NOT IN THE DATA

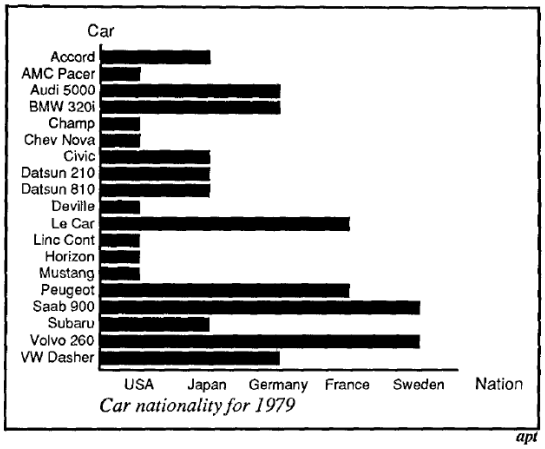

Fig. 11. Incorrect use of a bar chart for the *Nation* relation. The lengths of the bars suggest an ordering on the vertical axis, as if the USA cars were longer or better than the other cars, which is not true for the *Nation* relation.

Length is interpreted as encoding a quantitative value

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EFFECTIVENESS CRITERIA [Mackinlay 1986]

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.

Subject of the Perception Lecture

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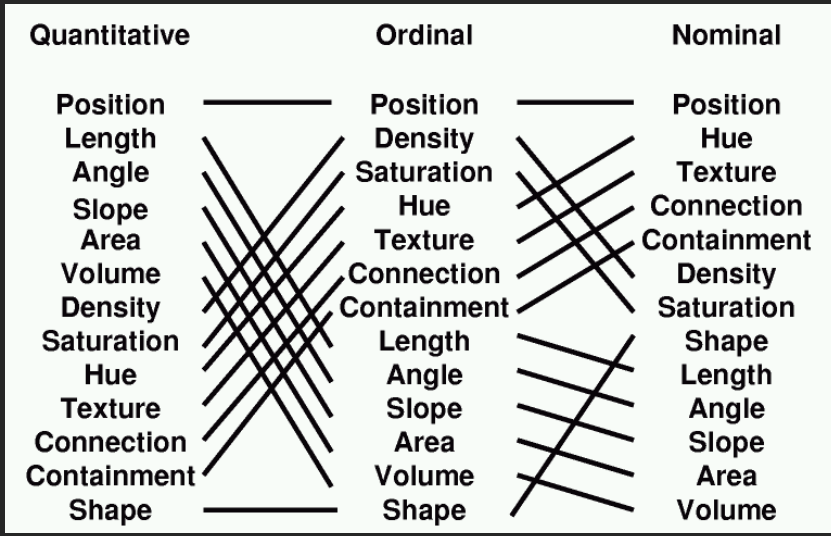
MACKINLAY'S RANKING



Conjectured *effectiveness* of encodings by data type

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AUTOMATIC CHART DESIGN [Mackinlay 1986]

APT – "A Presentation Tool"

User formally specifies data model and type
Input: list of data variables ordered by importance

APT searches over the design space
Tests expressiveness of each visual encoding (rule-based)
Generates encodings that pass test
Rank by perceptual effectiveness criteria

Outputs *most effective* visualization

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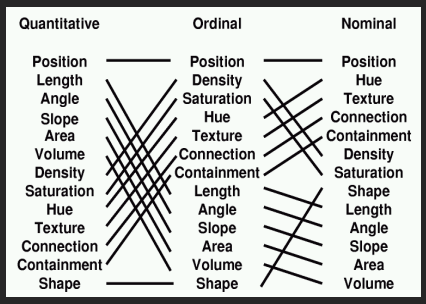
APT [Mackinlay 1986]

Encode most important data using highest ranking visual variable for the data type

Price	Mileage	Weight	Repair
13,500	22	3000	great
7,200	31	1500	ok
11,300	12	4200	terrible
…	…	…	…

→

1. Price (Q)
2. Mileage (Q)
3. Weight (Q)
4. Repair (N)



→

mark: lines

Price (Q) → y-pos
Mileage (Q) → x-pos
Weight (Q) → size
Repair (N) → color

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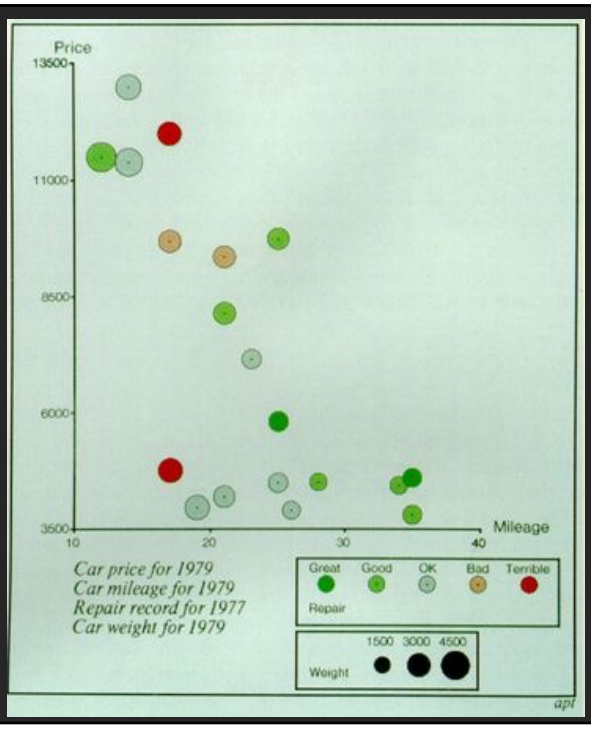
APT [Mackinlay 1986]

Automatically generated chart for cars data

Cars Data

1. Price (Q)
2. Mileage (Q)
3. Weight (Q)
4. Repair (Q)



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LIMITATIONS

Does not cover many visualization techniques

- Networks, maps, diagrams
- Also, 3D, animation, illustration, …

Does not consider interaction

Does not consider semantics or conventions

Assumes single visualization as output

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SUMMARY

Formal specification

- Data model: tidy data, N,O,Q types
- Image model: marks, visual attributes/channels
- Encodings map data to mark attributes/channels

Choose *expressive* and *effective* encodings

- Rule-based test of expressiveness
- Perceptual effectiveness rankings

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