

**BEFORE WE START.... A LITTLE
EXPERIMENT**

37

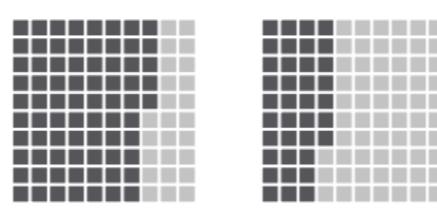
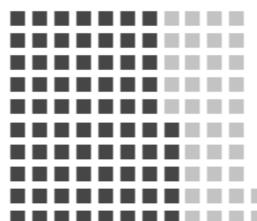
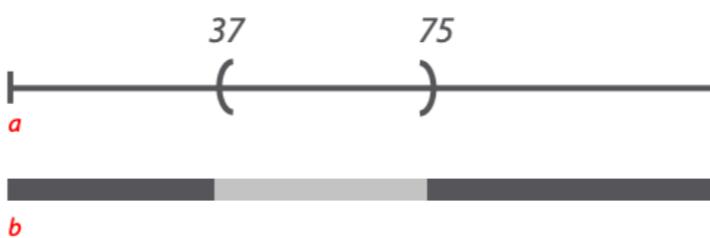
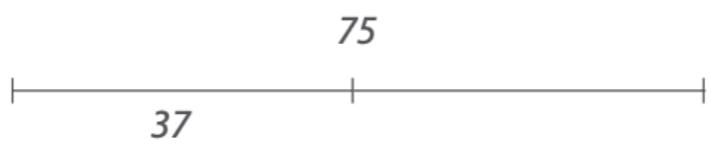
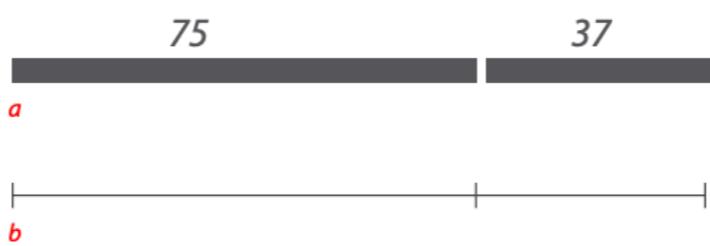
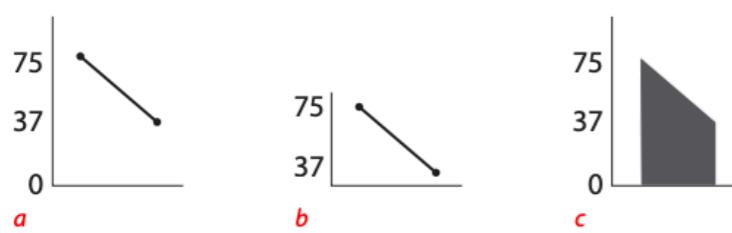
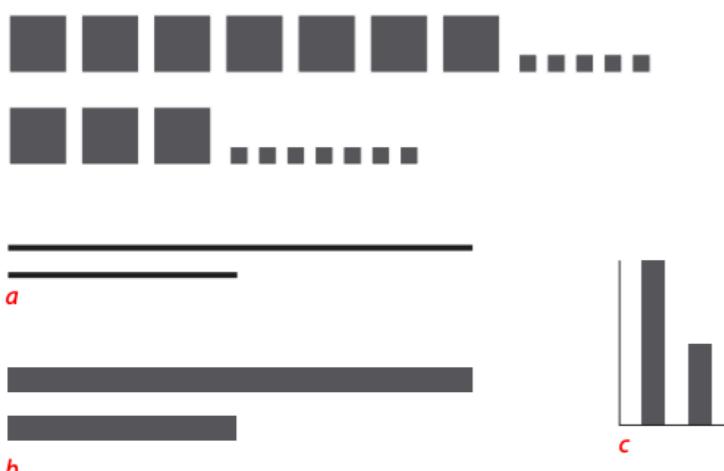
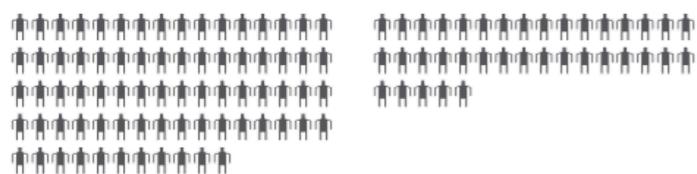
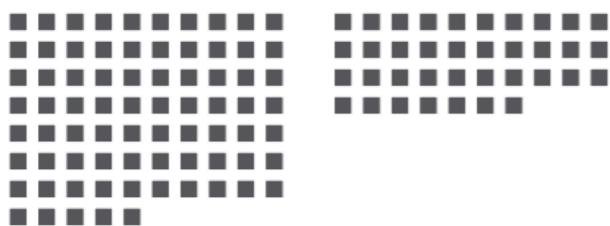
75

1. Grab a set of post-it and a pen; gather up in pairs



2. Try to come up with as many possible representations/encodings for the “data” above as you can, in the paper segments.

Feel free to be creative!





a

b

c



a



b



a



b

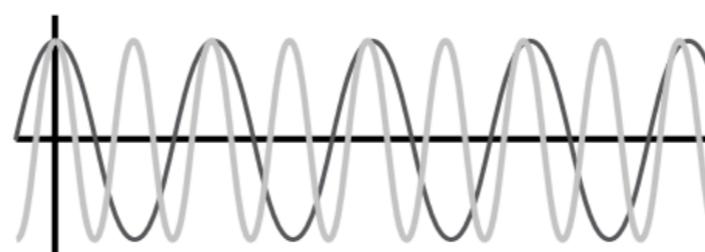
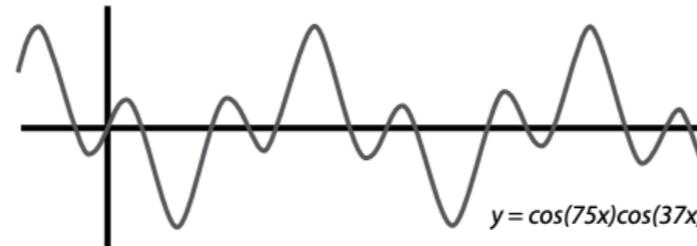
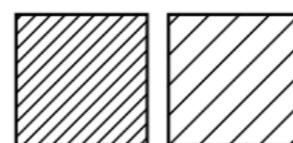
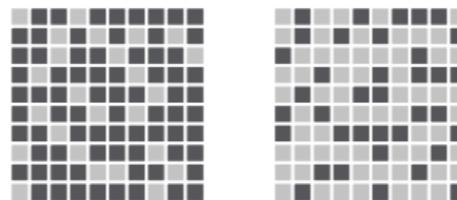
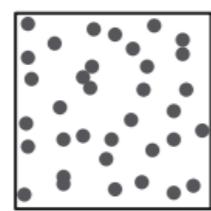
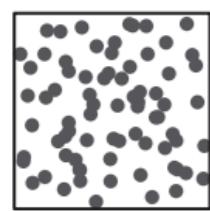
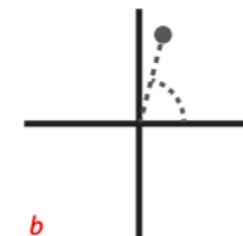
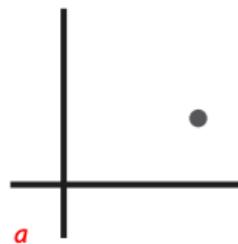


a



b

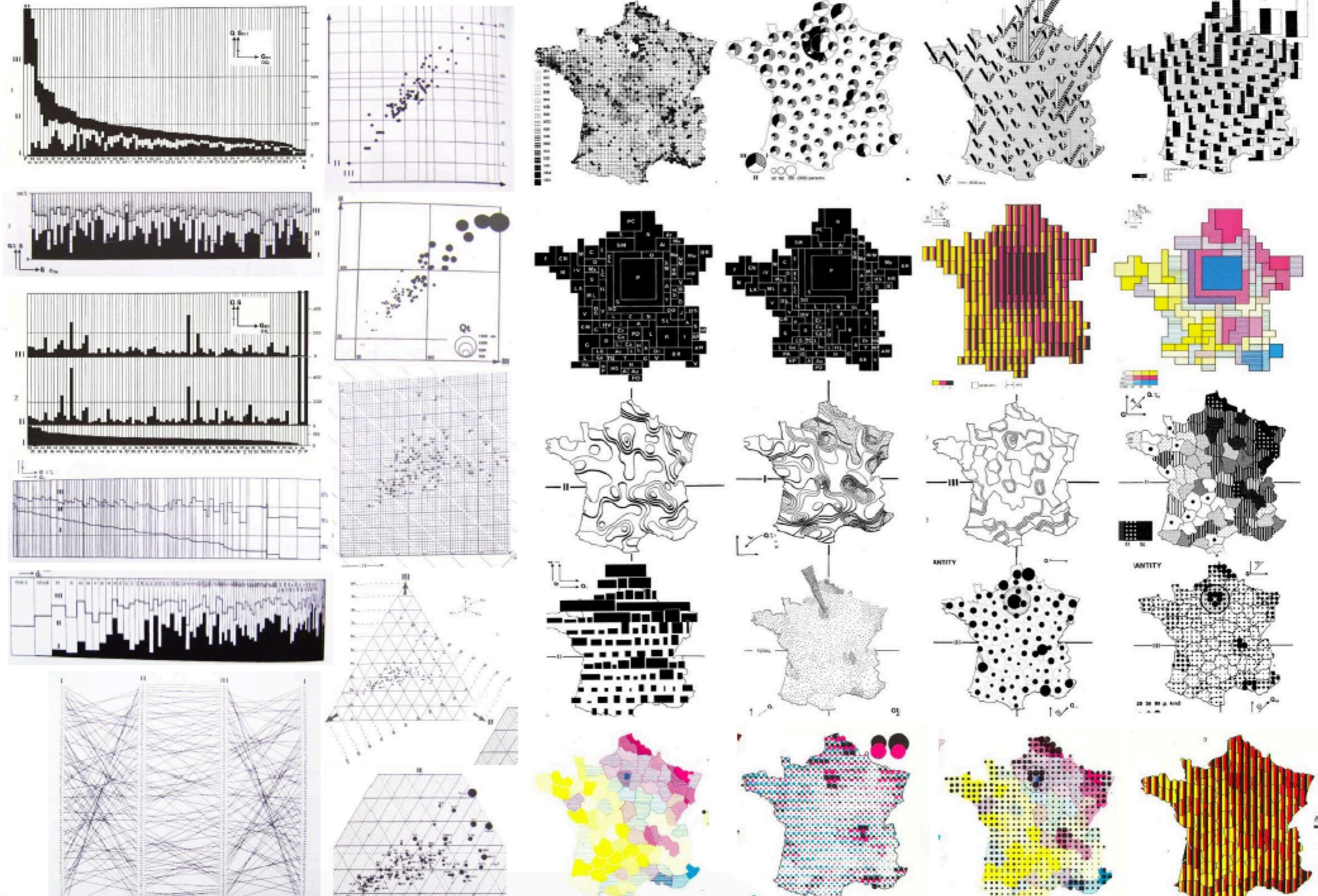




7_s 3_s

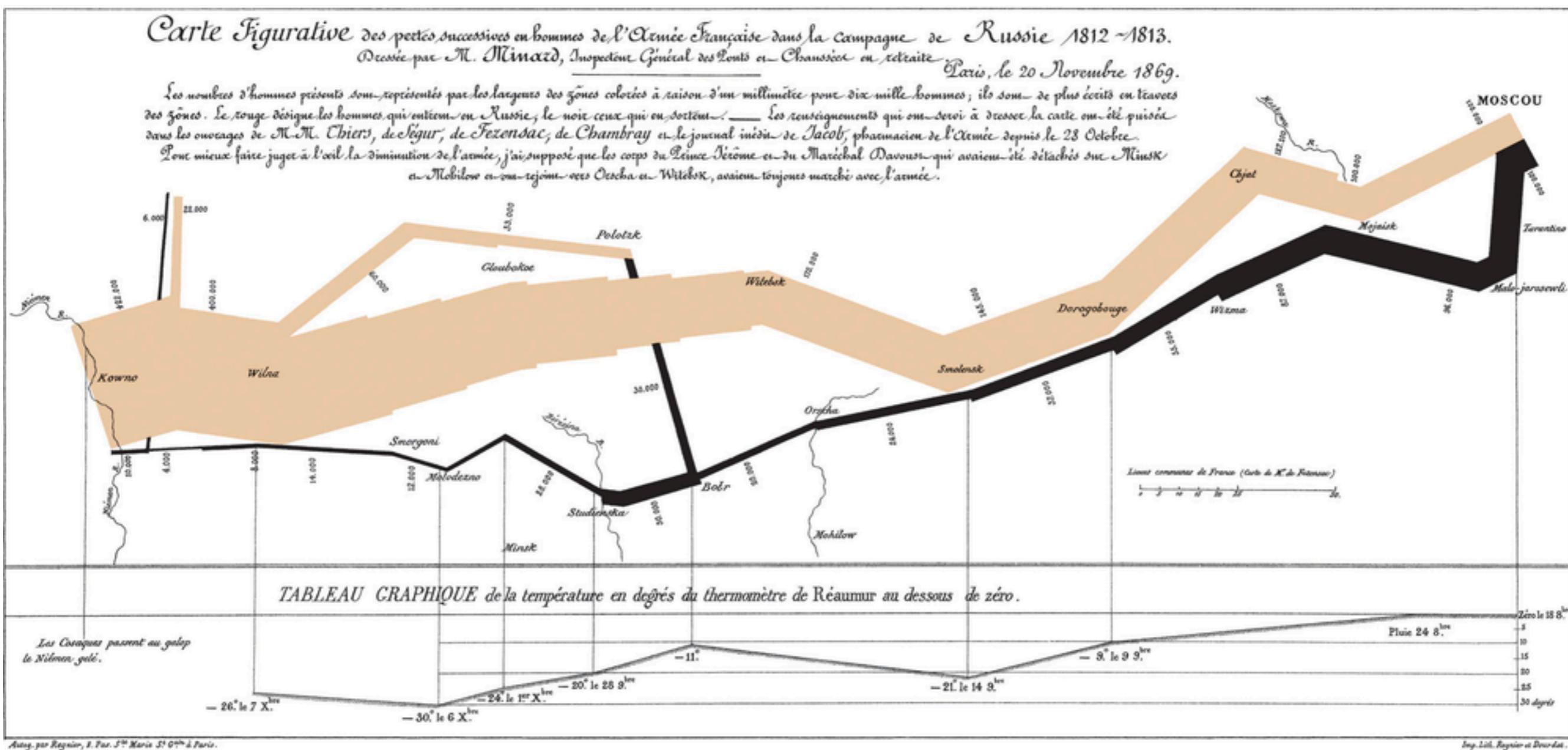
Départements

	Quantités (000)			Proportion %					
	I	II	III	Total	I	II	III		
1 AIN	87	13	40	140	45	38	27		
2 AISNE	94	24	64	182	28	37	34		
3 ALLIER	65	45	57	167	38	37	34		
4 Hautes ALPES	15	8	12	35	43	34	33		
5 Basses ALPES	16	8	13	37	44	21	35		
6 ALPES MARITIMES	31	61	122	214	14	39	57		
7 ARDÈCHE	48	32	25	105	45	21	24		
8 ARDENNES	25	53	21	119	22	47	31		
9 ARIège	35	17	14	66	52	29	24		
10 AUDE	28	48	26	112	25	43	32		
11 AVEYRON	50	20	32	102	49	19	33		
12 BOUCHES-DU-RHÔNE	79	92	29	199	54	34	21		
13 CALVADOS	79	55	69	193	36	38	36		
14 CANTAL	45	13	29	83	56	16	21		
15 CHARENTE	65	26	39	120	47	28	25		
17 CHARENTE-MARitime	79	39	63	182	43	21	36		
18 CHIERS	43	41	36	120	26	34	39		
19 COISE	64	23	29	116	35	28	23		
21 CÔTE D'OR	43	44	29	112	39	29	31		
22 CÔTES-DU-NORD	131	33	62	226	58	15	27		
23 CRÈUSE	59	13	17	89	60	15	19		
24 DORDOGNE	104	34	42	180	58	19	23		
25 DOLUS	25	67	39	132	25	47	39		
26 DRÔME	46	38	35	119	39	38	39		
27 EURE	48	52	45	145	28	36	31		
28 EURE & LOIR	64	27	38	129	41	25	34		
29 FINISTERE	154	26	89	269	50	22	27		
30 GARD	40	31	32	104	29	26	36		
31 HAUTE-GARONNE	64	67	64	215	39	31	39		
32 GERS	43	19	16	78	71	11	19		
33 GIROUD	115	197	179	392	30	27	43		
34 HERAULT	92	69	71	172	26	23	41		
35 ILLE & V.	137	68	82	279	49	23	39		
36 ISÈRE	94	39	32	126	45	26	29		
37 INDRE & LO.	61	61	65	187	39	26	35		
38 ISERE	68	126	78	222	24	48	28		
39 JURA	39	24	27	100	39	34	27		
40 LANDS	79	25	29	122	53	20	29		
41 LOIRE & CHER	51	27	29	108	47	25	26		
42 LOIRE	56	169	82	219	19	54	27		
43 Ille LOIRE	81	23	22	97	54	24	22		
44 LOIRE INF.	101	109	105	324	29	34	34		
45 LOIRET	51	51	34	136	32	28	35		
46 LOT	41	10	16	67	61	15	24		
47 LOT & GAR.	79	24	20	124	37	19	24		
48 LOZÈRE	22	5	7	34	64	13	21		
49 MAINES & LO.	104	65	65	234	44	28	28		
50 MANCE	116	43	56	214	34	20	36		
51 MARNE	64	57	67	188	39	35	34		
52 Ille MARNE	79	26	28	91	31	35	32		
53 MAYENNE	74	23	29	122	39	29	39		
54 MONTBELLÉGNY	23	127	91	241	8	53	39		
55 MOSSE	24	31	27	82	55	27	38		
56 MORBIAN	132	47	59	238	55	20	25		
57 MUSSELLE	36	179	94	309	94	39	39		
58 NEUVIC	34	21	23	70	34	38	29		
59 NOUO	81	493	295	809	9	56	25		
60 ORNE	49	63	55	161	63	36	31		
61 OISE	65	39	36	139	36	26	39		
62 P. I. C.	94	242	127	473	20	51	29		
63 PYRÉNÉES-ORIENT.	60	79	69	228	69	35	36		
64 Pyrénées-Orient.	60	69	68	197	68	39	33		
65 Pyrénées-Orient.	37	27	29	90	40	29	39		
66 PYRÉNÉES-ORIENT.	35	29	29	90	40	29	39		
67 HAUTE-PYRÉNÉE	76	122	114	312	24	39	39		
68 HAUTE-PYRÉNÉE	49	121	74	295	17	38	39		
69 HAUTE-SAVOIE	44	213	198	455	10	47	39		
70 Haute-Savoie	34	32	29	89	39	26	39		
71 SAONE & LO.	94	77	62	233	41	33	39		
72 SAÔNE	87	45	58	199	40	24	34		
73 SAÔNE	84	39	35	117	39	32	39		
74 Haute-Saône	52	42	45	138	27	36	39		
75 PAIX	7	275	949	1327	0	38	41		
76 SAÔNE INF.	8	274	550	1129	1	51	39		
77 SAÔNE & LO.	87	72	36	185	20	39	39		
78 SAÔNE & LO.	46	229	356	830	0	49	39		
79 SAÔNE-SAÔNE	71	29	39	133	53	29	39		
80 SEMUR	57	60	61	186	21	38	39		
81 TARN	55	47	33	125	41	35	37		
82 TARN & G.	48	18	16	59	35	29	37		
83 VAR	82	59	61	184	20	51	39		
84 VAUCLUSE	40	30	41	111	36	37	39		
85 VENDÉE	110	38	40	182	59	26	39		
86 VIENNE	60	29	39	128	47	39	39		
87 Haute-VIENNE	64	47	43	156	41	39	39		
88 VORGES	36	95	63	174	21	34	39		
89 YONNE	41	28	37	106	39	26	39		
90 BULLENT	3	25	13	41	0	60	39		
91	444	444	444	1332	6795	6926	16225	28	39



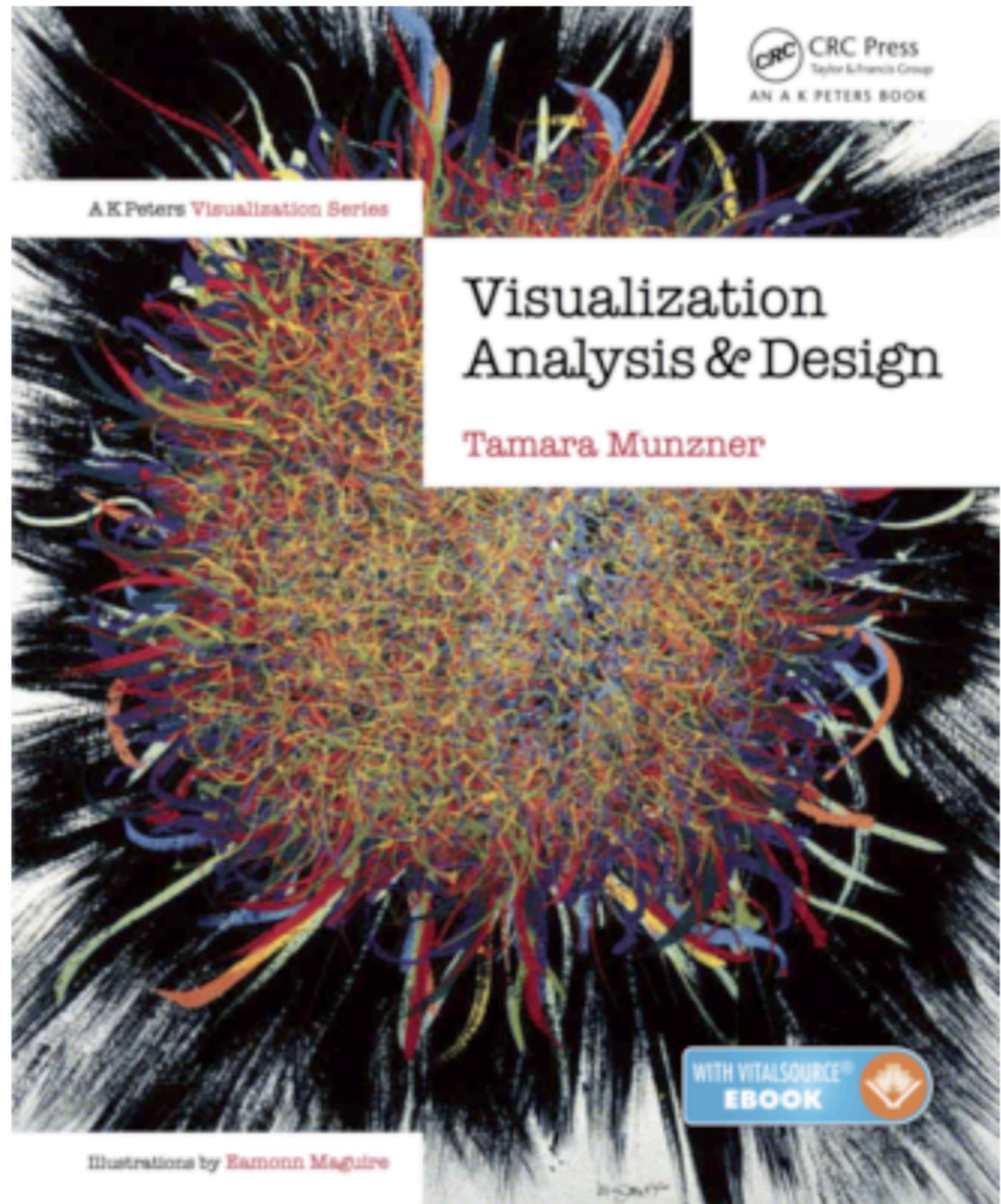
Bertin defined the design space then showed how it could be used to generate many different types of visualizations; using a simple dataset as an example. (Semiology of Graphics, pg 100 ~ 138). Side note French edition appendix on text, not translated into English.

Visualization theory and principles



Text-book on data visualization
(1st of its kind)

If you can only get one,
THIS is the one!



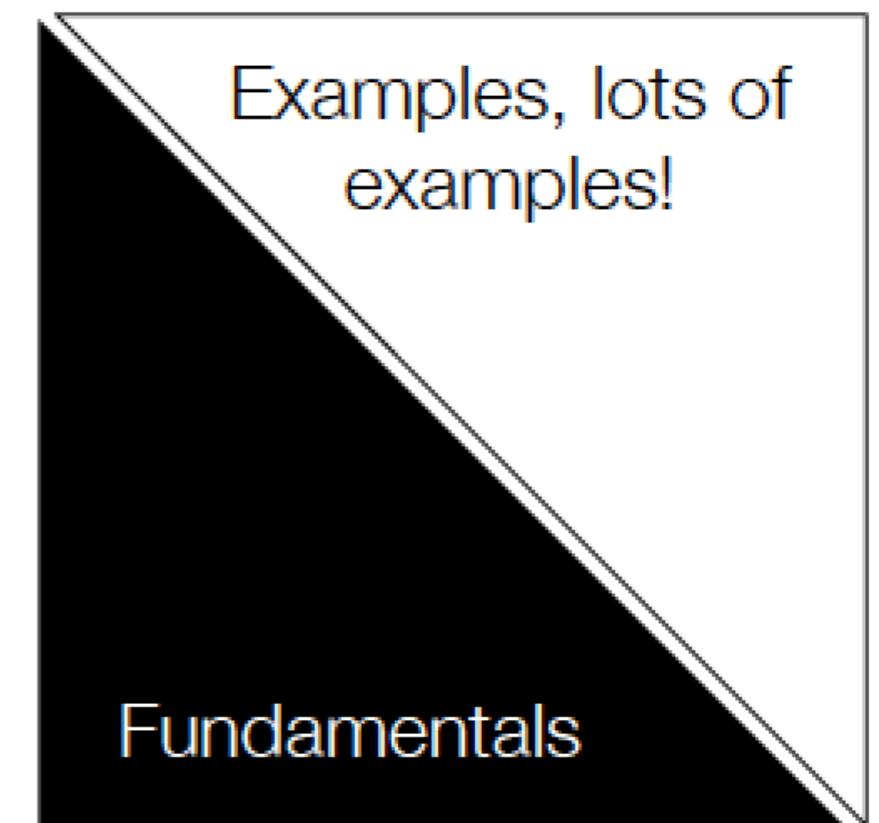
Munzner, Tamara. Visualization Analysis and Design. CRC Press, 2014.

TAKE AWAY MESSAGE....

It takes **experience!**

But despite of that, we all have the “necessary prerequisites” to be a good visualization designer....

.... we just need to learn how to **see** things a bit differently!

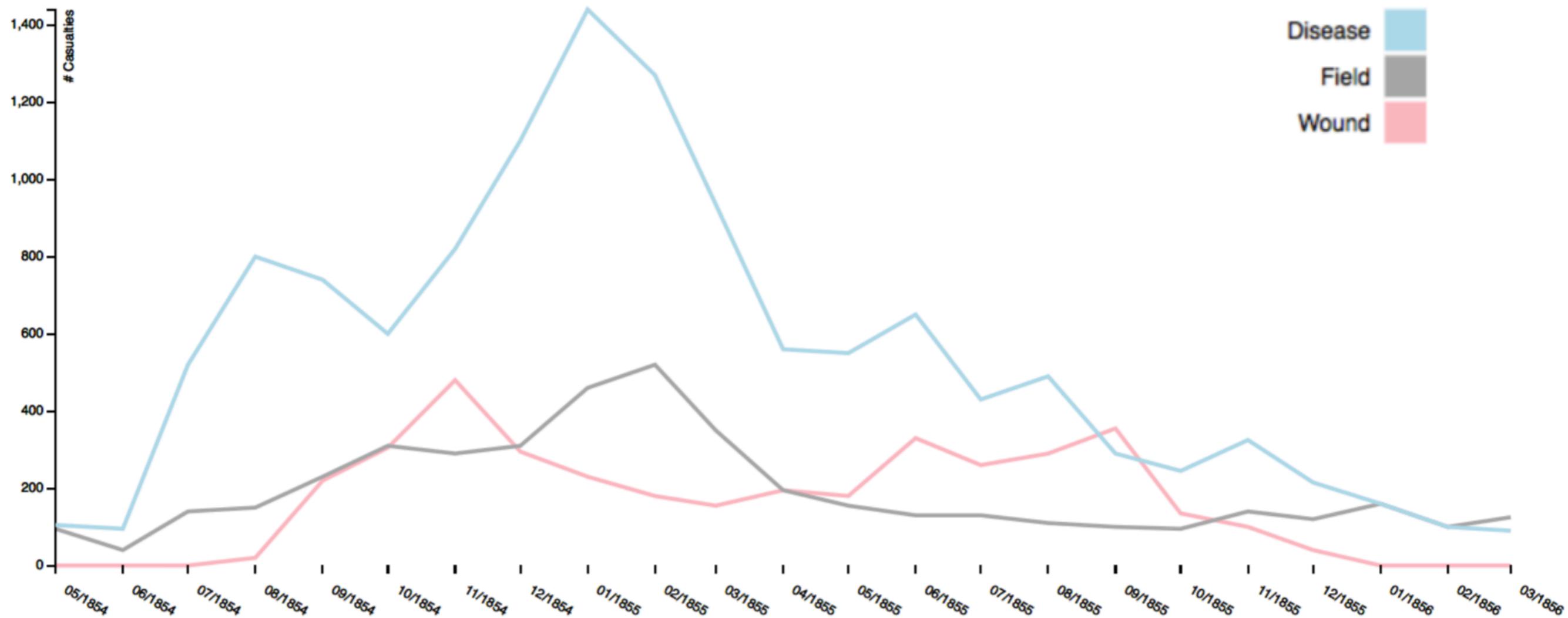


British Casualties in the Crimean War

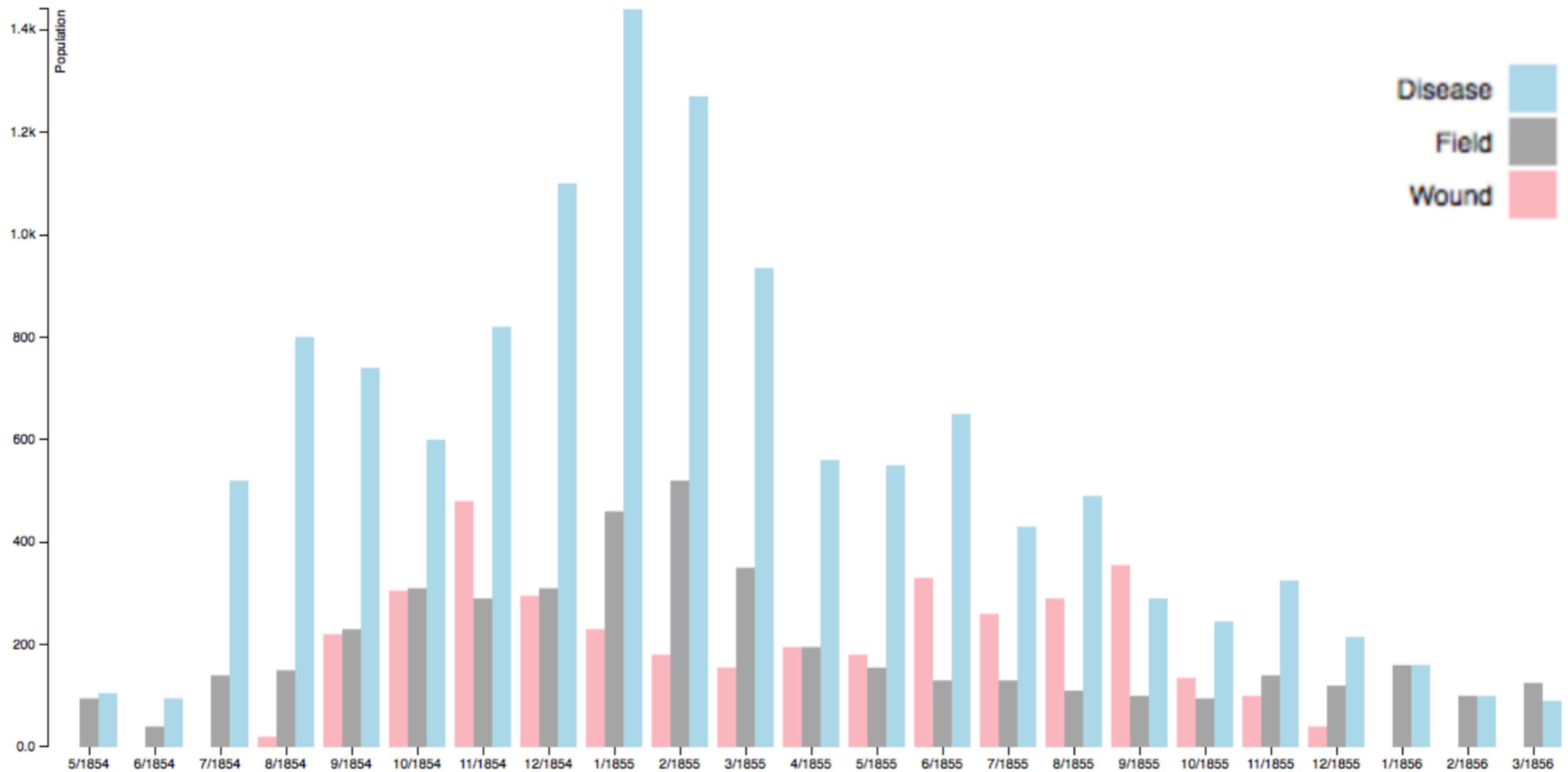
Data	Wound	Field	Disease
05/1854	0	95	105
06/1854	0	40	95
07/1854	0	140	520
08/1854	20	150	800
09/1854	220	230	740
10/1854	305	310	600
11/1854	480	290	820
12/1854	295	310	1100
01/1855	230	460	1440
02/1855	180	520	1270
03/1855	155	350	935
04/1855	195	195	560
05/1855	180	155	550
06/1855	330	130	650
07/1855	260	130	430
08/1855	290	110	490
09/1855	355	100	290
10/1855	135	95	245
11/1855	100	140	325
12/1855	40	120	215
01/1856	0	160	160
02/1856	0	100	100
03/1856	0	125	90

- ▶ Month with highest casualty rates in the field?
- ▶ Months in which deaths by wound exceeds deaths in the field?
- ▶ Month with highest total casualty rate?
- ▶ Months in which % of deaths by disease was below 50%?

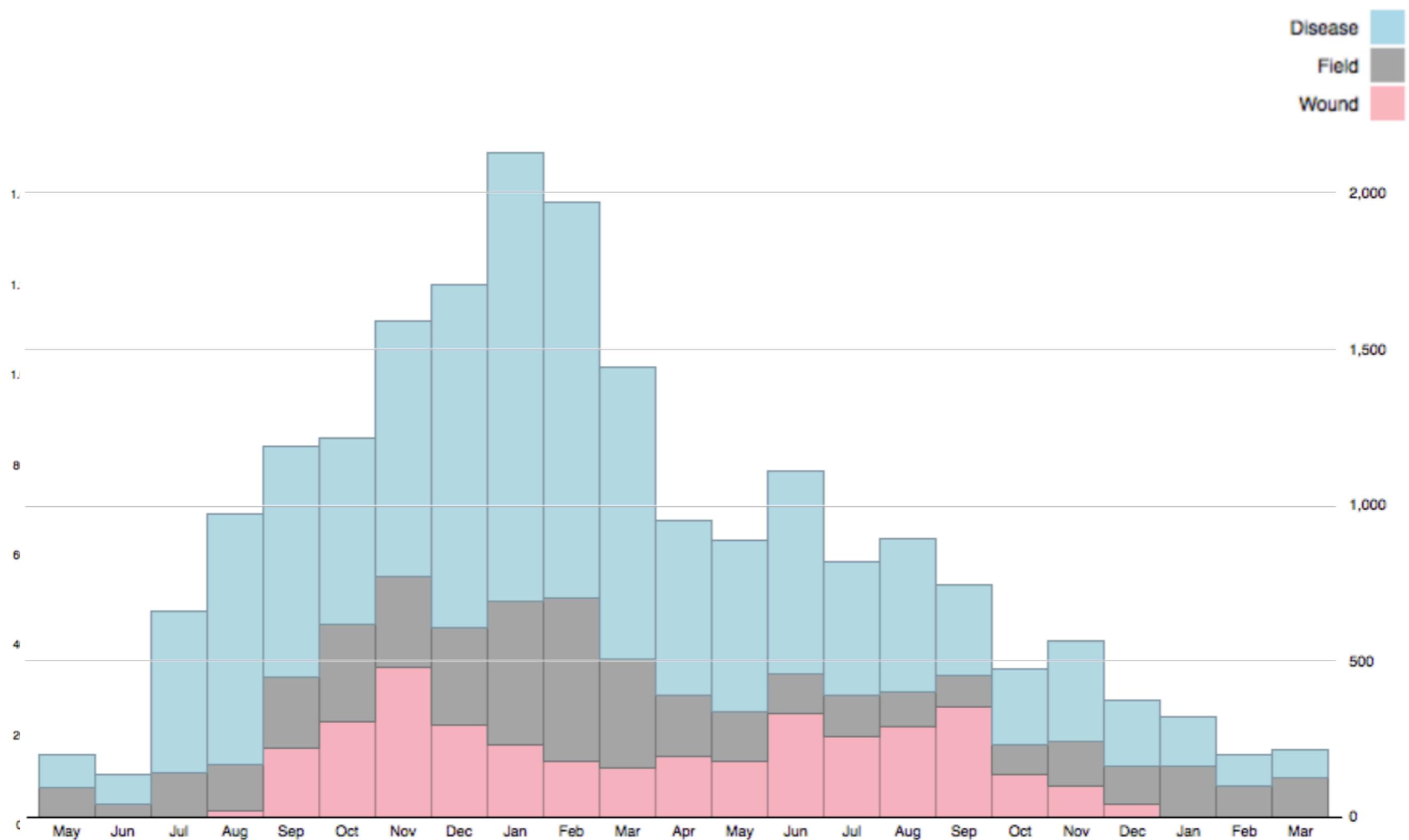
Month with highest casualty rates in the field?



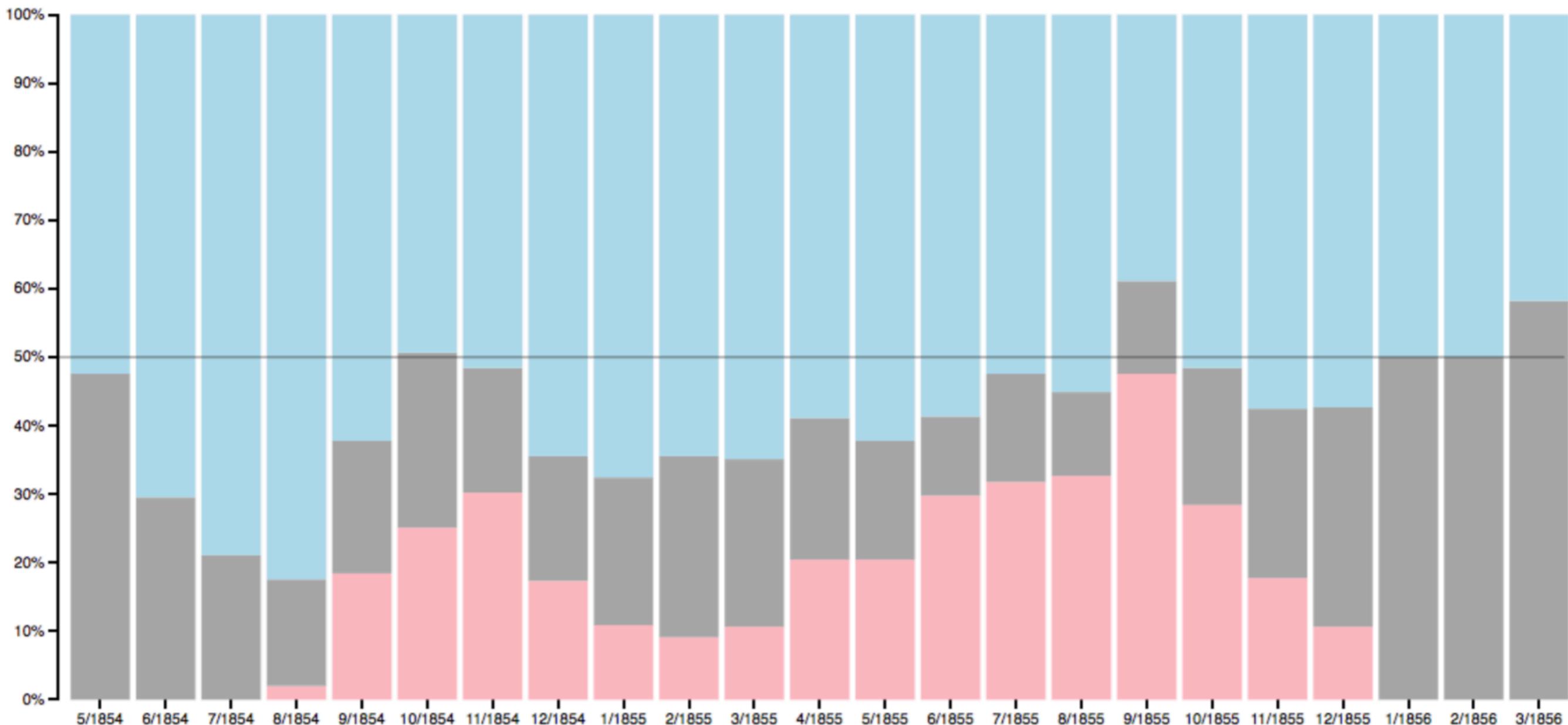
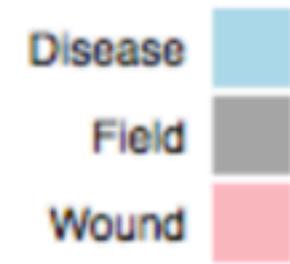
Months in which deaths by wound exceeds deaths in the field?



Month with highest total casualty rate?

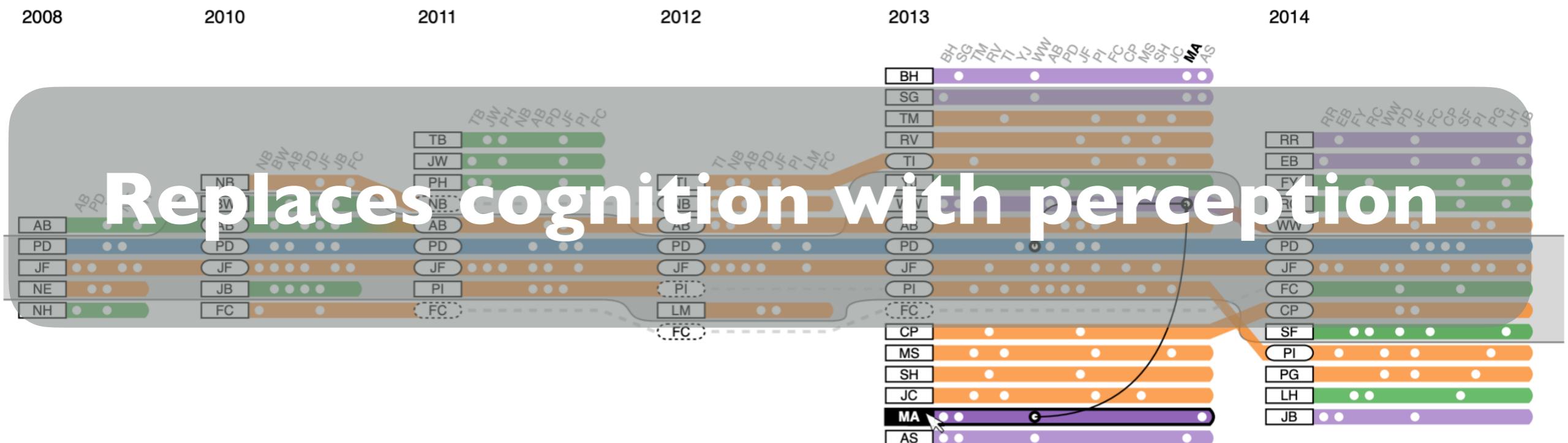


Months in which % of deaths by disease was below 50%?

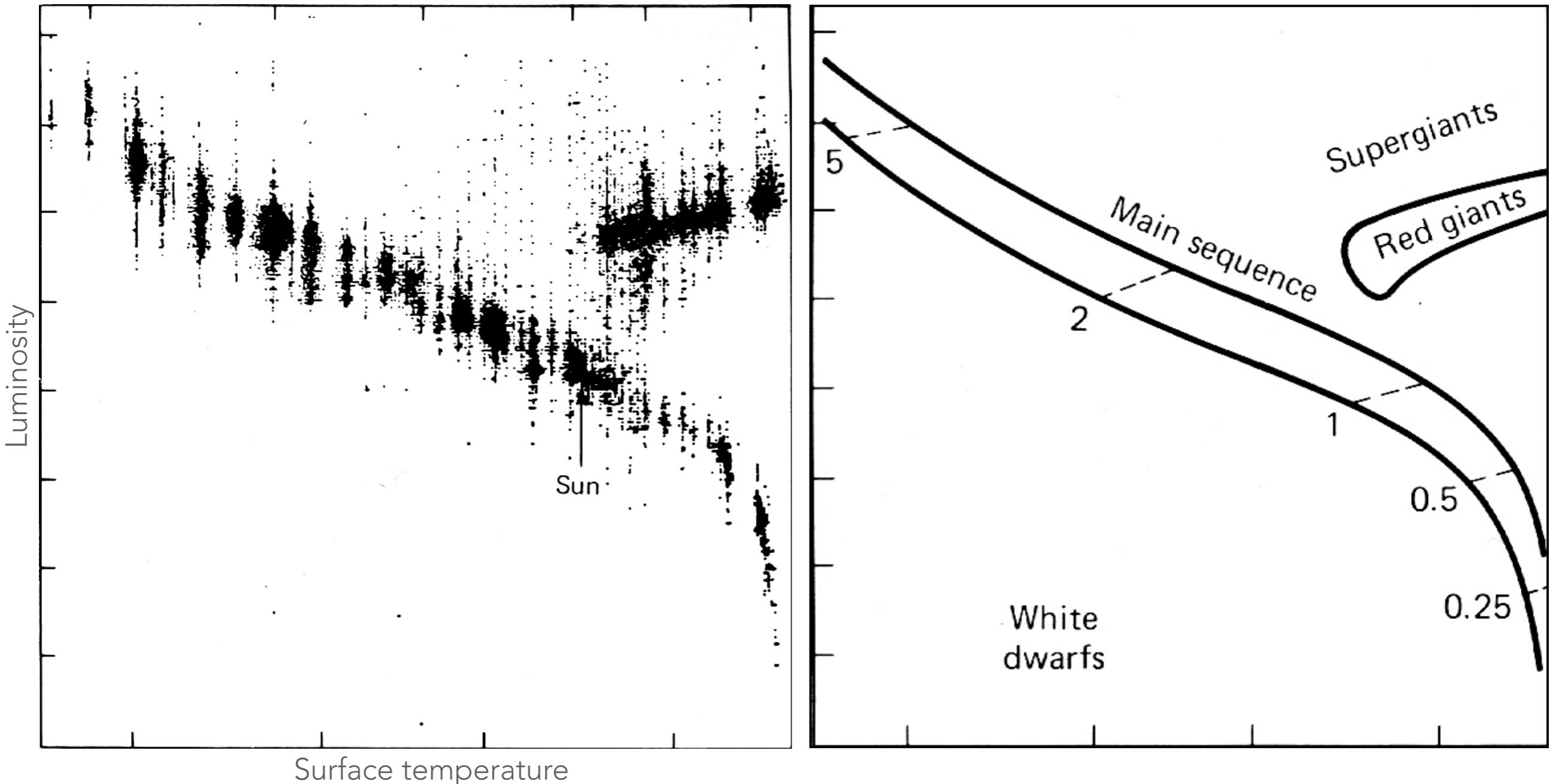


WHY VISUALIZATION?

Computer-based visualization systems provide visual representations of datasets designed to help people carry out tasks more effectively.

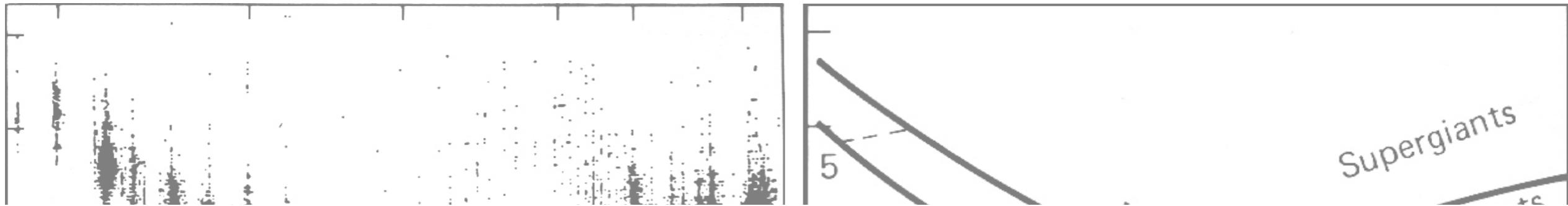


PERCEPTION AND COGNITION



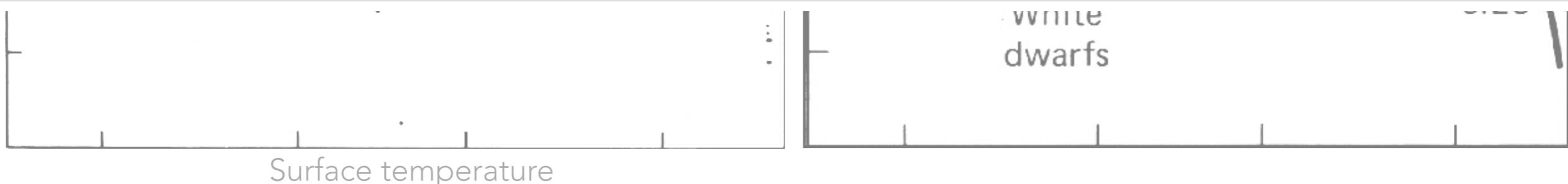
Hertzsprung Russell Diagram and its interpretation

PERCEPTION AND COGNITION



*“Why should we be interested in visualization? Because the human visual system is a pattern seeker of enormous power and subtlety. The eye and the visual cortex of the brain form a **massively parallel processor** that provides **the highest-bandwidth channel** into human cognitive centers.”*

Ware, Colin. Information visualization: perception for design. Elsevier, 2013.



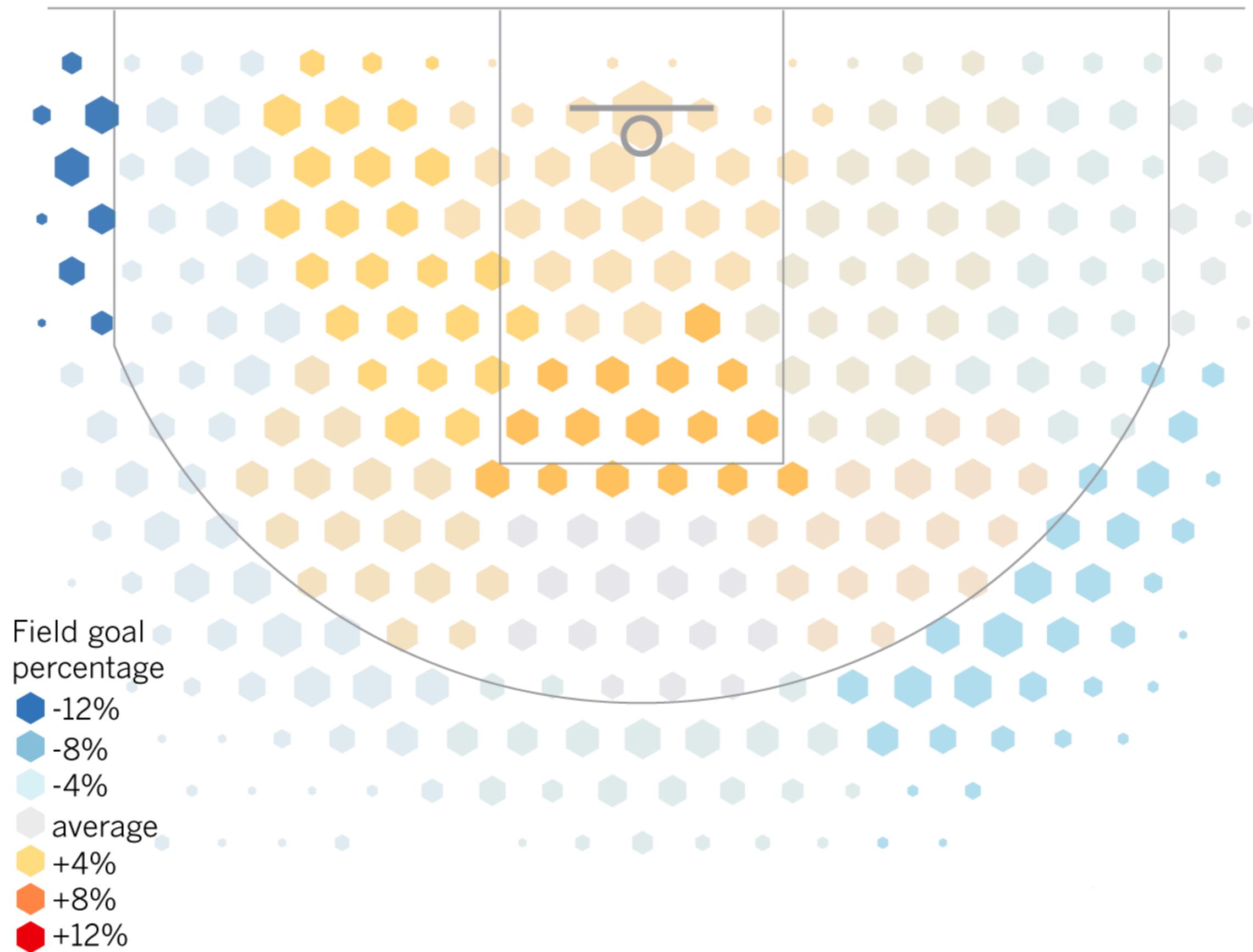
Hertzsprung Russell Diagram and its interpretation

[Source: J.D. Fekete et al. “The Value of Information Visualization”, 2008]

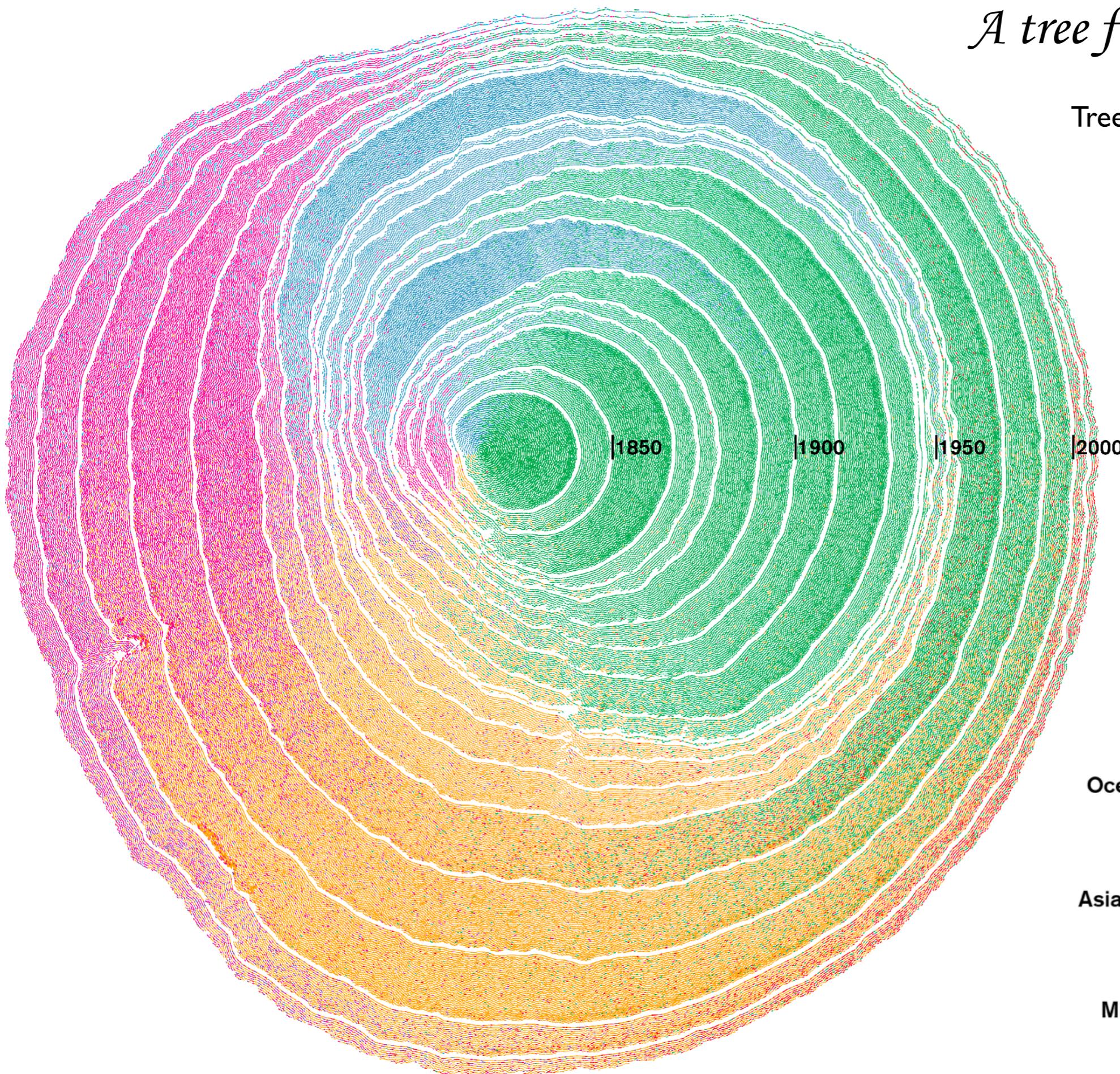
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285960799918712845268101495969124567781
874241649645757659608149596912456701285
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142416496357598475921765968474891728482
285958819829450968504850695847612124044
074674898985171495969124567659608020860
608**3**6541649645759064**3**980479248576960781
285960799918712845268101495969124567781
874241649645757659608149596912456701285
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22359164964575958819825096**3**576596080596

Kobe Bryant vs. league average (1996-2016)

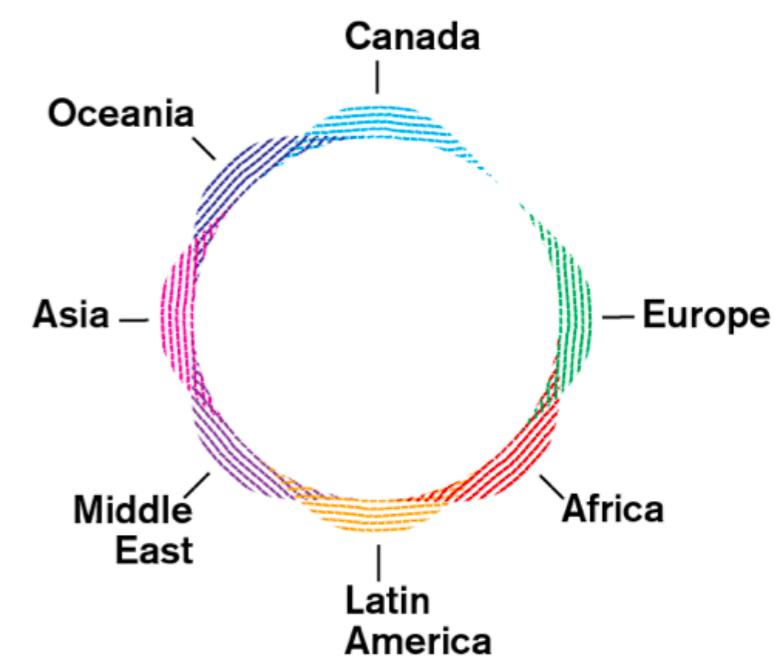


A tree for U.S. immigration



Tree rings showing immigration for
1830-2016

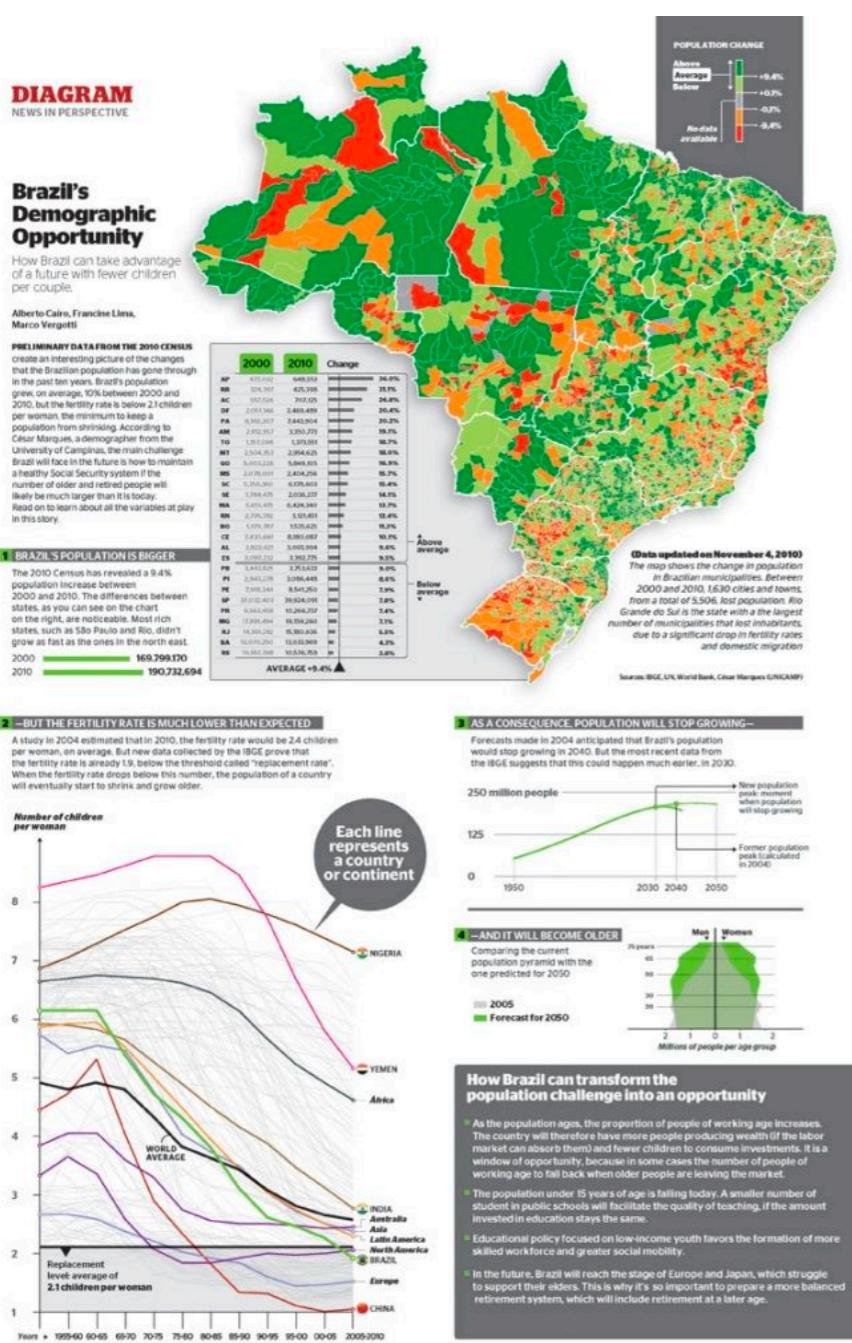
Each dot corresponds to
100 immigrants



MAIN USES OF DATA VISUALIZATION

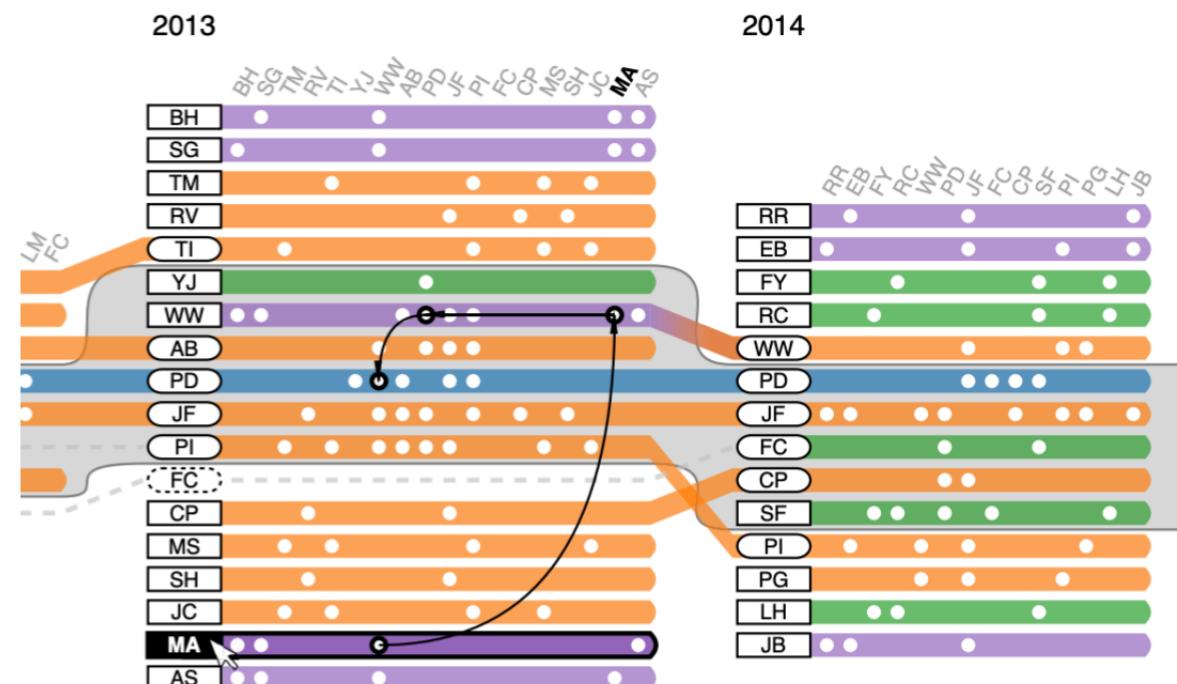
EXPLANATORY

Storytelling



EXPLORATORY

Acquiring insights



This is your brain on visualization....

PERCEPTION FUNDAMENTALS

Visualization design is based on the mapping between data and visual representation

Data



Graphics

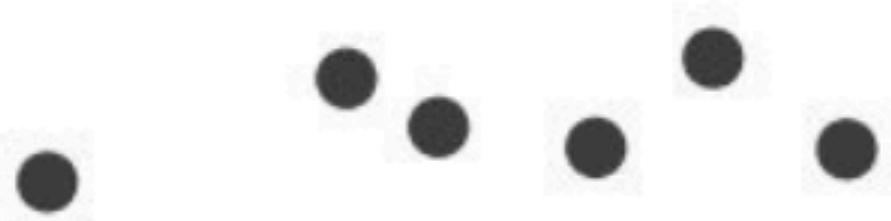
Which representations can I use to encode data?

Which representations are more suitable to ensure I'm conveying the right message?

MARKS

Geometric primitives

→ Points



→ Lines

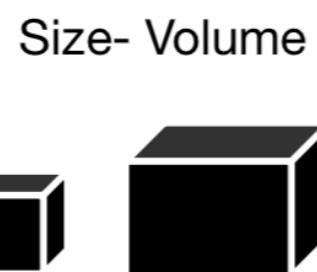
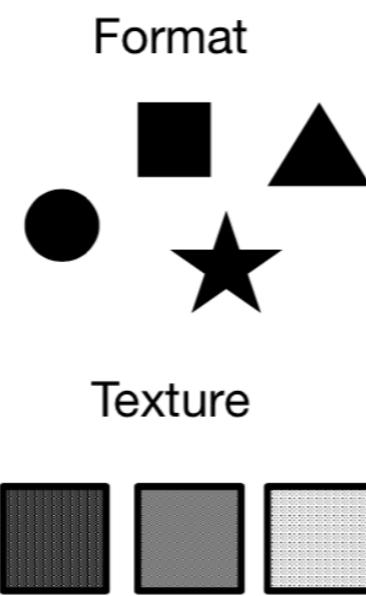
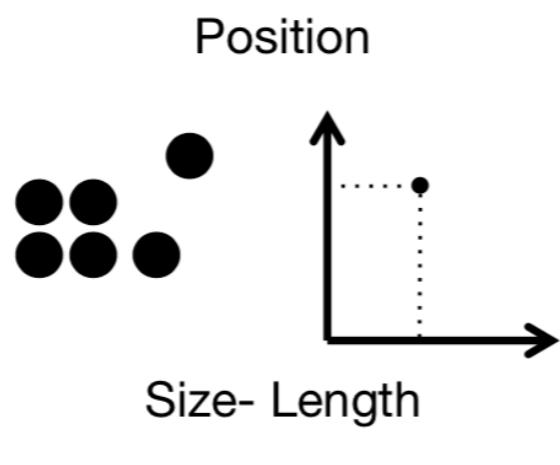


→ Areas



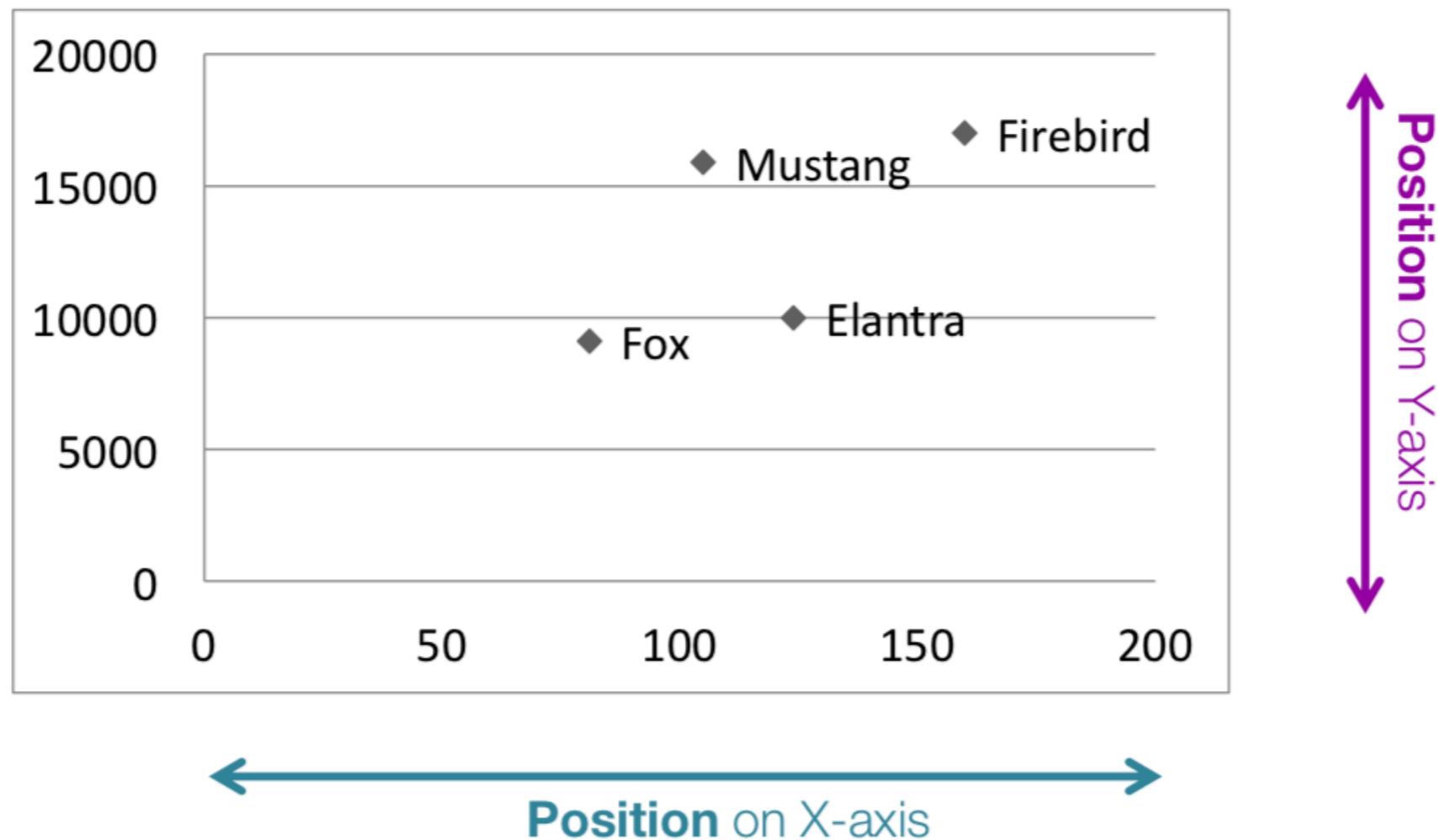
CHANNELS (VISUAL VARIABLES)

Control appearance of marks

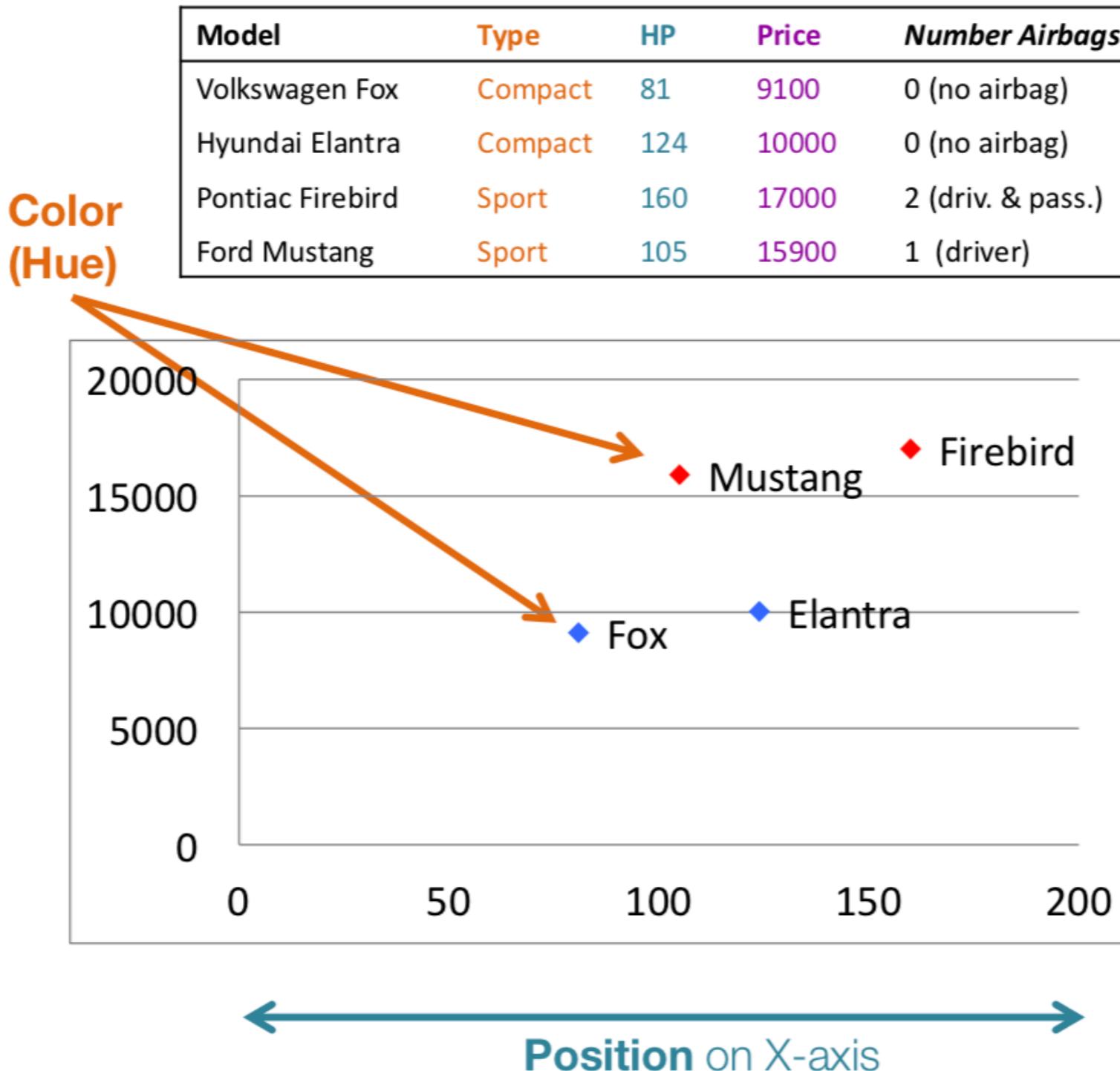


MAPPING DATA TO VISUAL VARIABLES

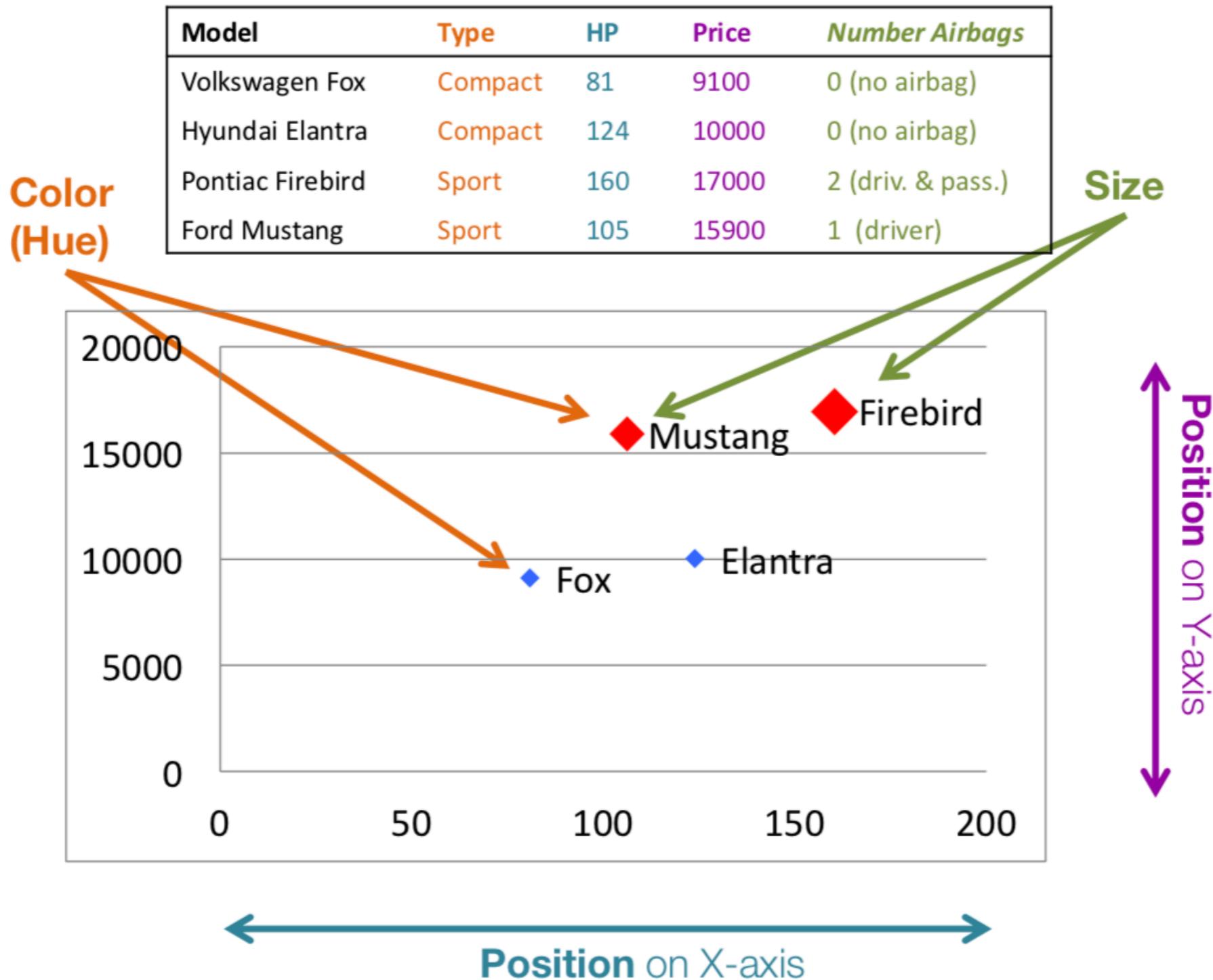
Model	Type	HP	Price	Number Airbags
Volkswagen Fox	Compact	81	9100	0 (no airbag)
Hyundai Elantra	Compact	124	10000	0 (no airbag)
Pontiac Firebird	Sport	160	17000	2 (driv. & pass.)
Ford Mustang	Sport	105	15900	1 (driver)



MAPPING DATA TO VISUAL VARIABLES



MAPPING DATA TO VISUAL VARIABLES



Which representations can I use to encode data?

Expressiveness principle: use adequate/suitable data representations

Encodings should convey all, and only, the information of associated attributes.

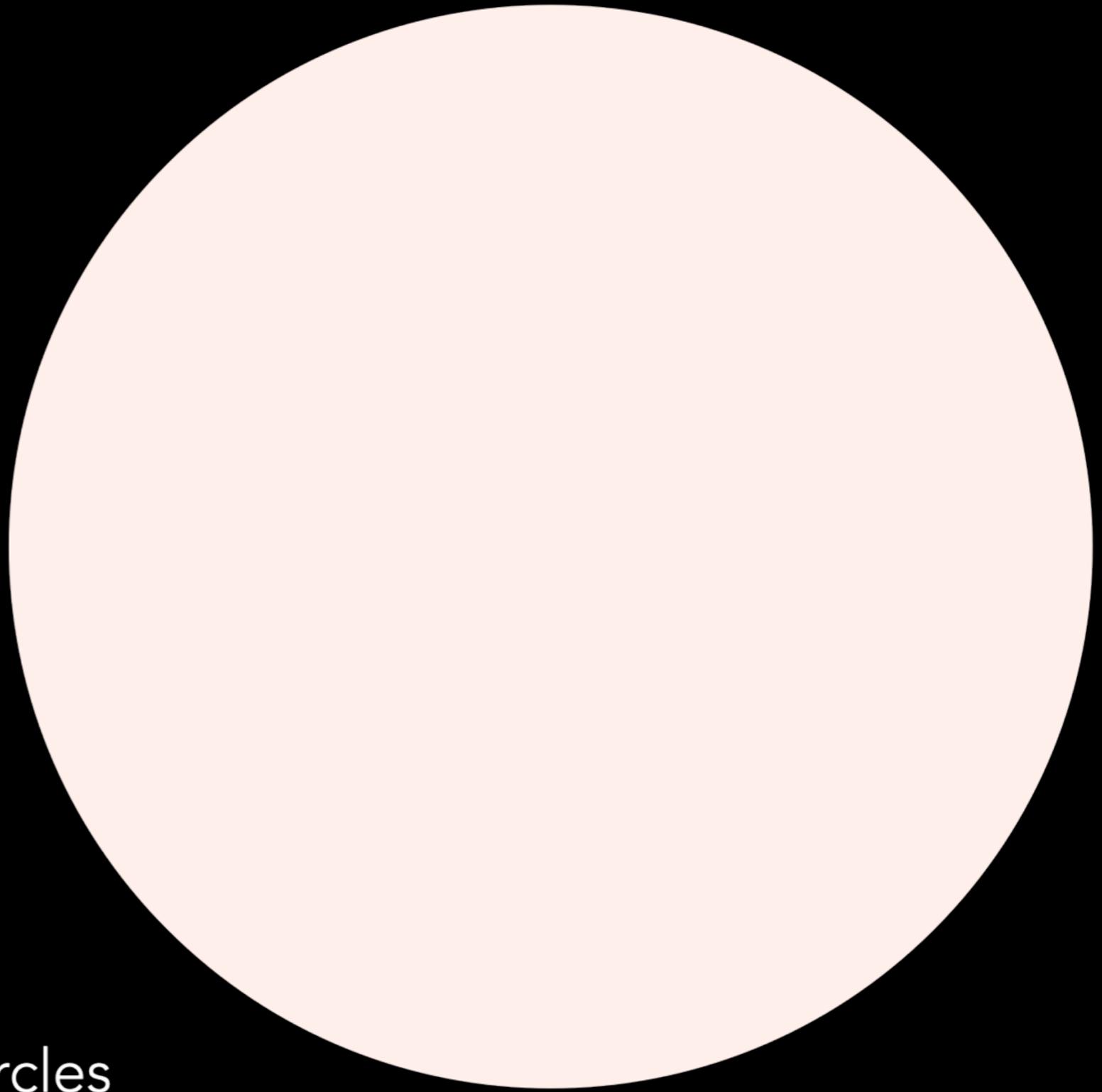
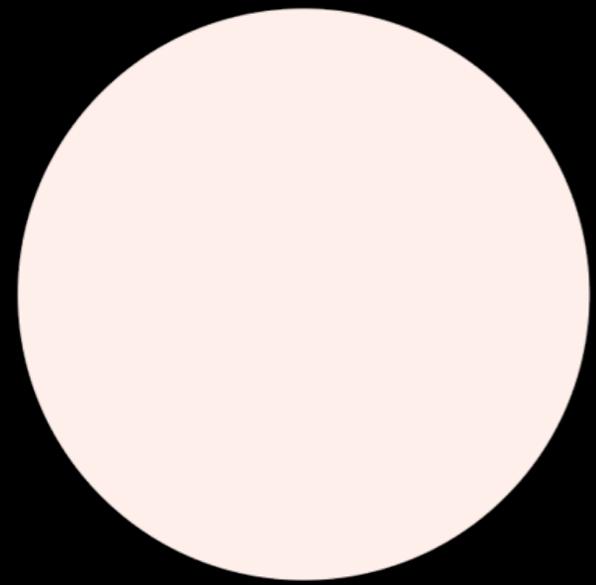
e.g. Ordinal data representation should convey “order”; similarly, “categorical data” should not be shown in a way that implies order.

**Which representations are more suitable to ensure
I'm conveying the right message?**

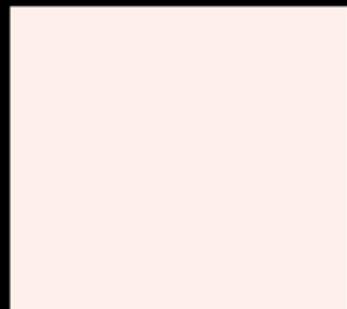
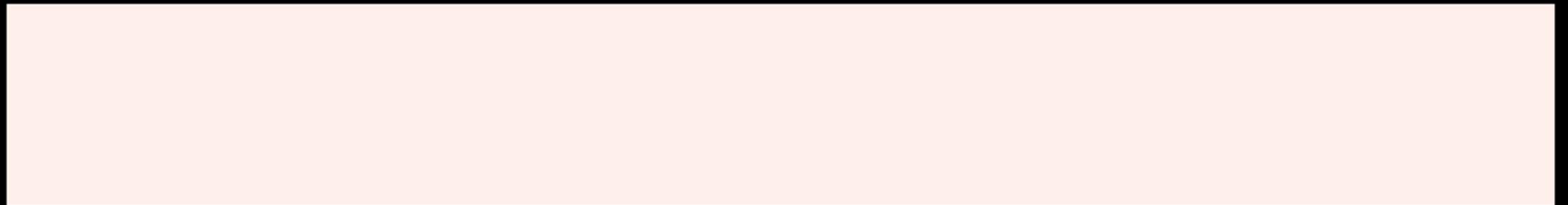
**Effectiveness principle:
choosing the best representation to your data**

Importance of attributes should match the “saliency” of the channel;

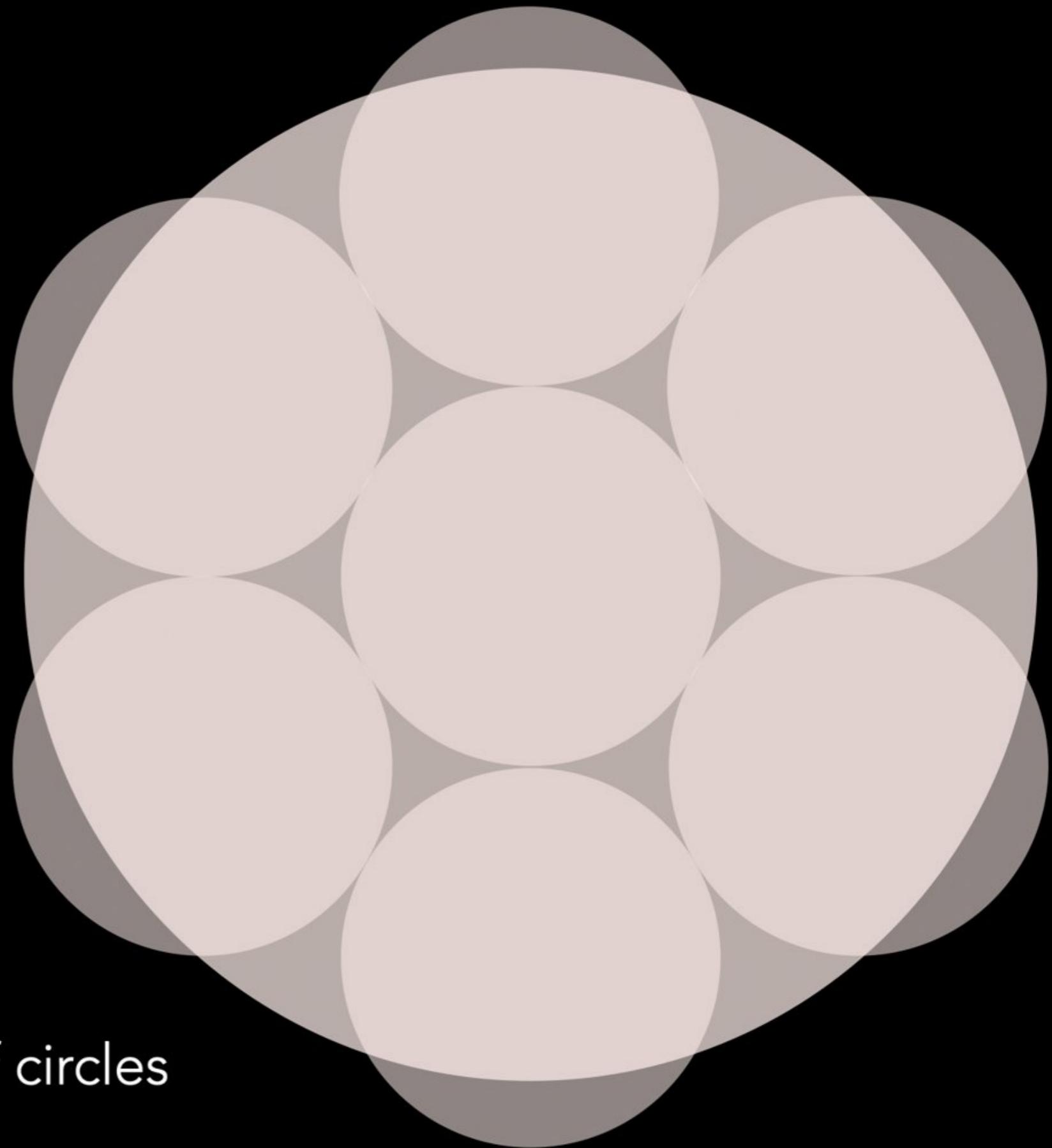
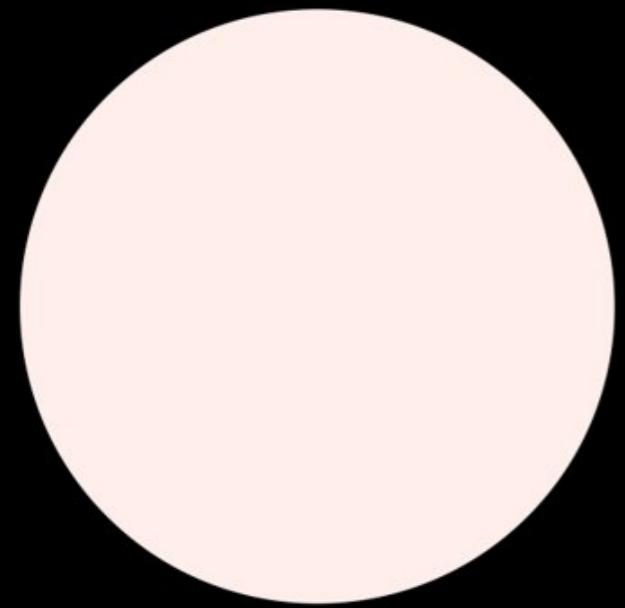
Most important attributes should be encoded using the most effective and noticeable channels.



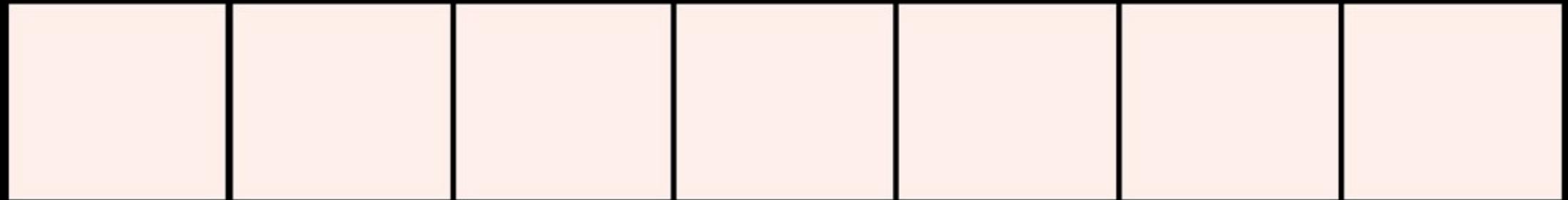
Compare **area** of circles



Compare **length** of bars



Compare the **area** of circles

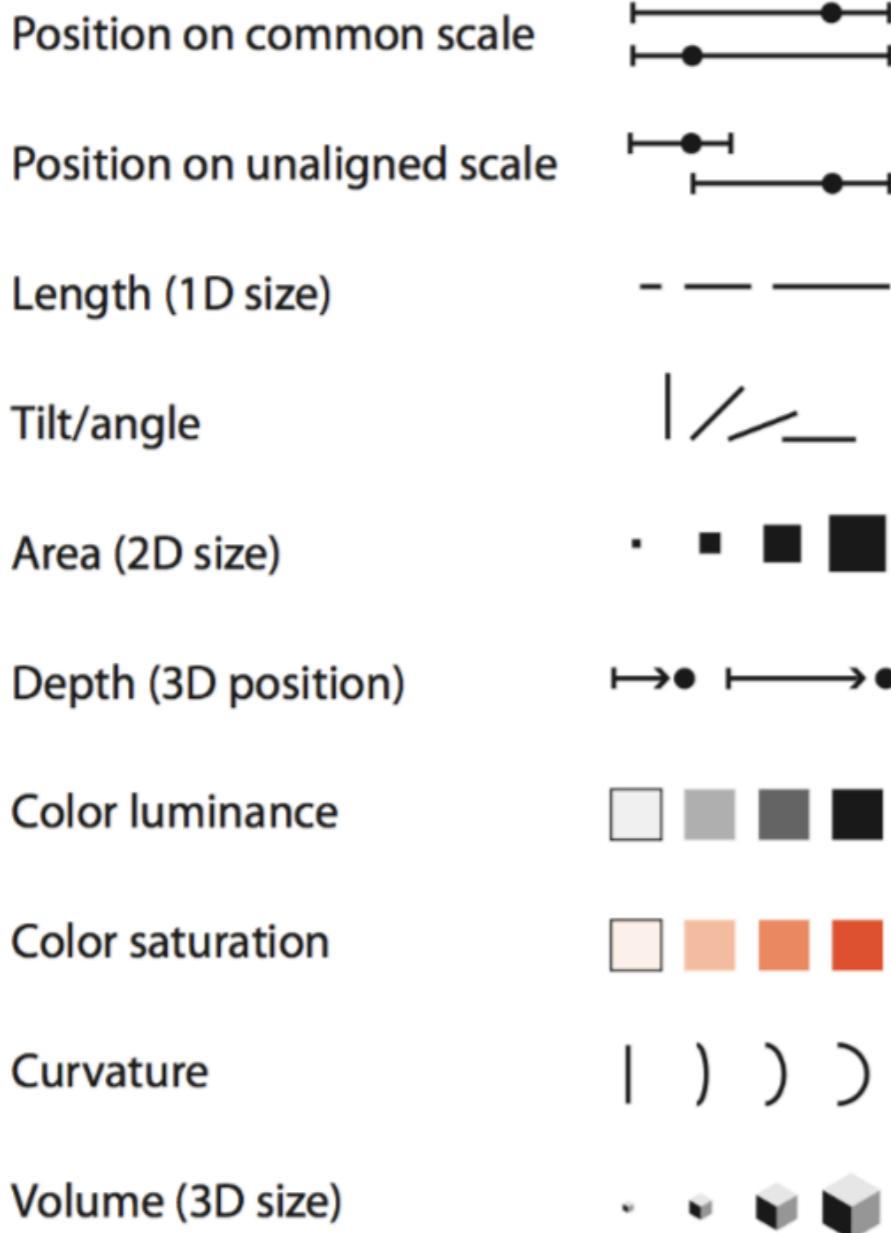


Compare **length** of bars

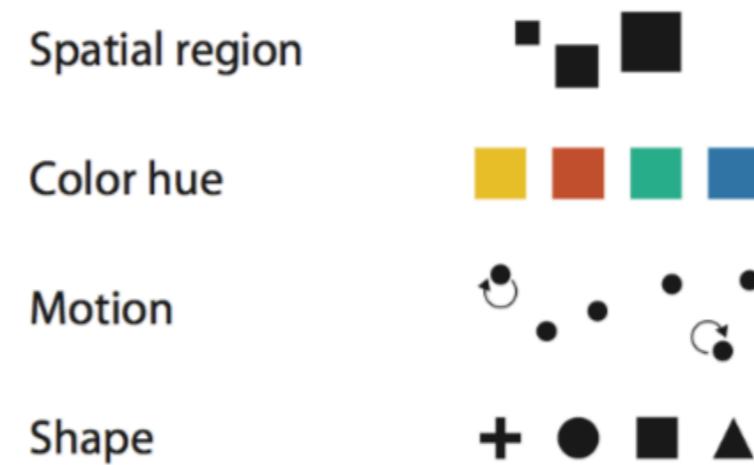
EFFECTIVENESS PRINCIPLE

Some variables are perceptually better than others

④ Magnitude Channels: Ordered Attributes



④ Identity Channels: Categorical Attributes



ATTRIBUTE TYPES

Not Ordered

Ordered

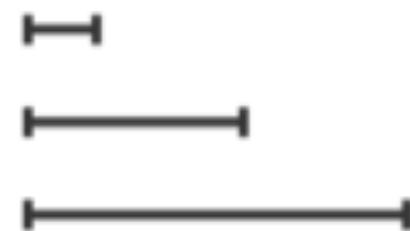
Categorical



Ordinal



Quantitative



Ordering Direction

Sequential



Diverging



Circular

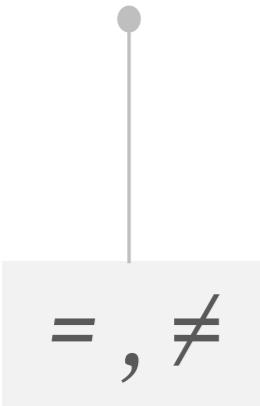


ATTRIBUTE TYPES

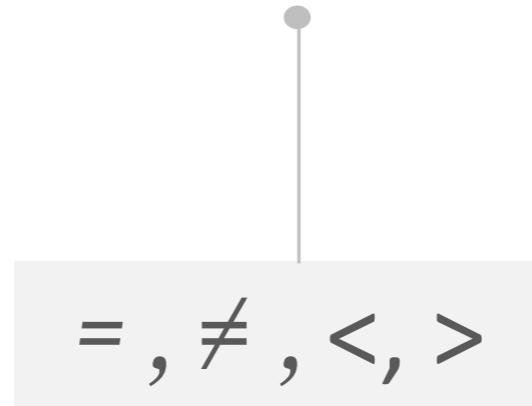
Not Ordered

Ordered

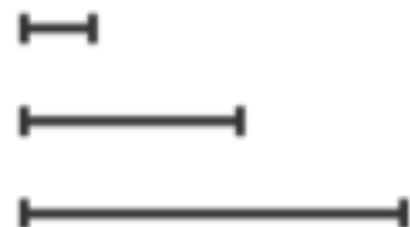
Categorical



Ordinal



Quantitative



ATTRIBUTE TYPES

Not Ordered

Ordered

Categorical



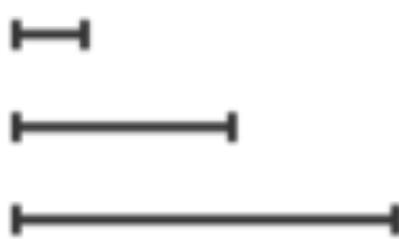
=, ≠

Ordinal

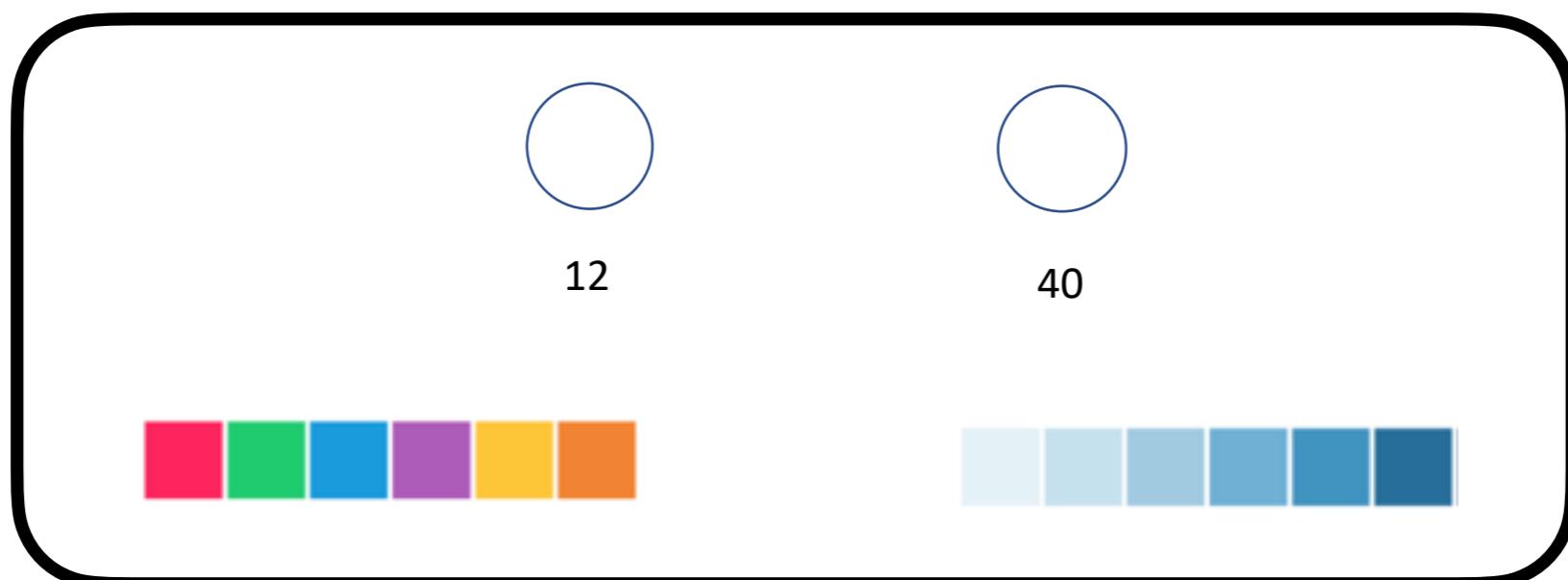


=, ≠, <, >

Quantitative



=, ≠, <, >, +, -, ÷



ATTRIBUTE TYPES

Not Ordered

Ordered

Categorical



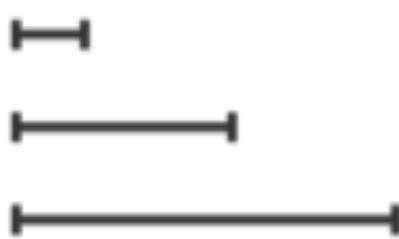
=, ≠

Ordinal

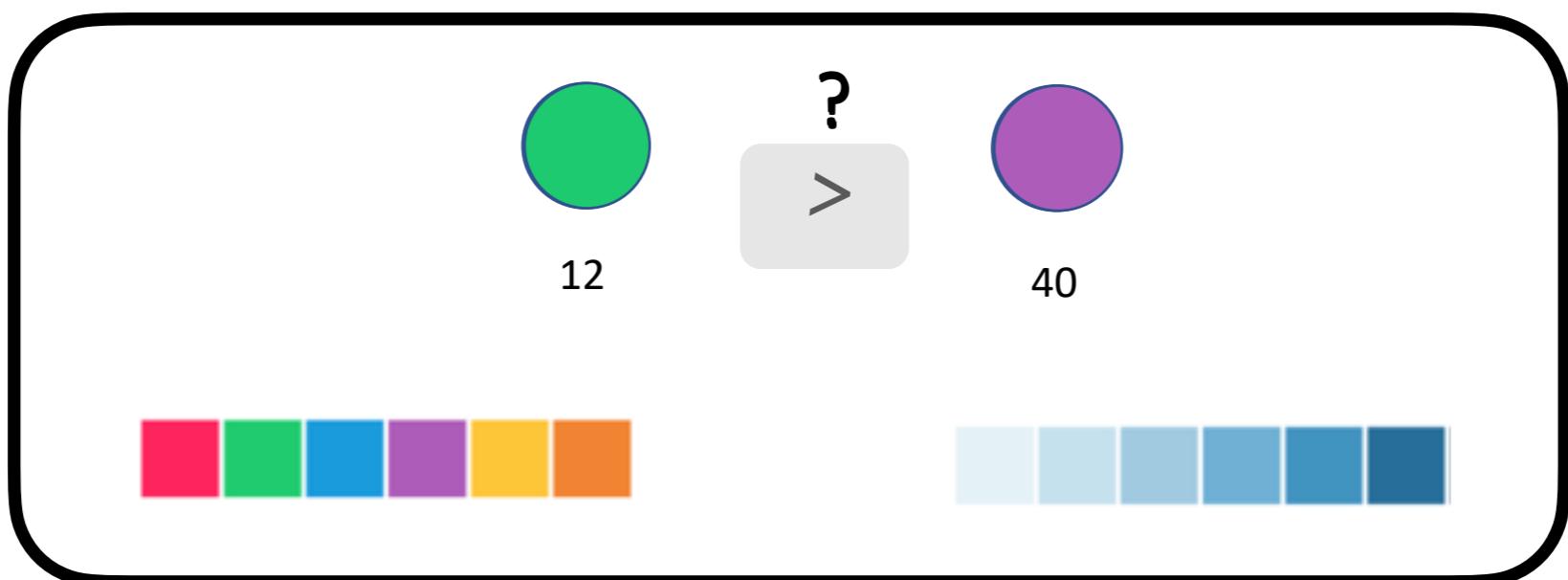


=, ≠, <, >

Quantitative



=, ≠, <, >, +, -, ÷



ATTRIBUTE TYPES

Not Ordered

Ordered

Categorical



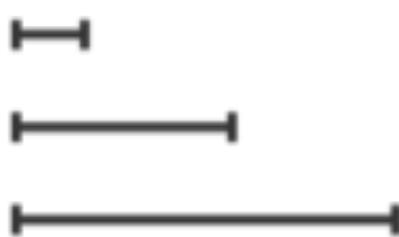
=, ≠

Ordinal

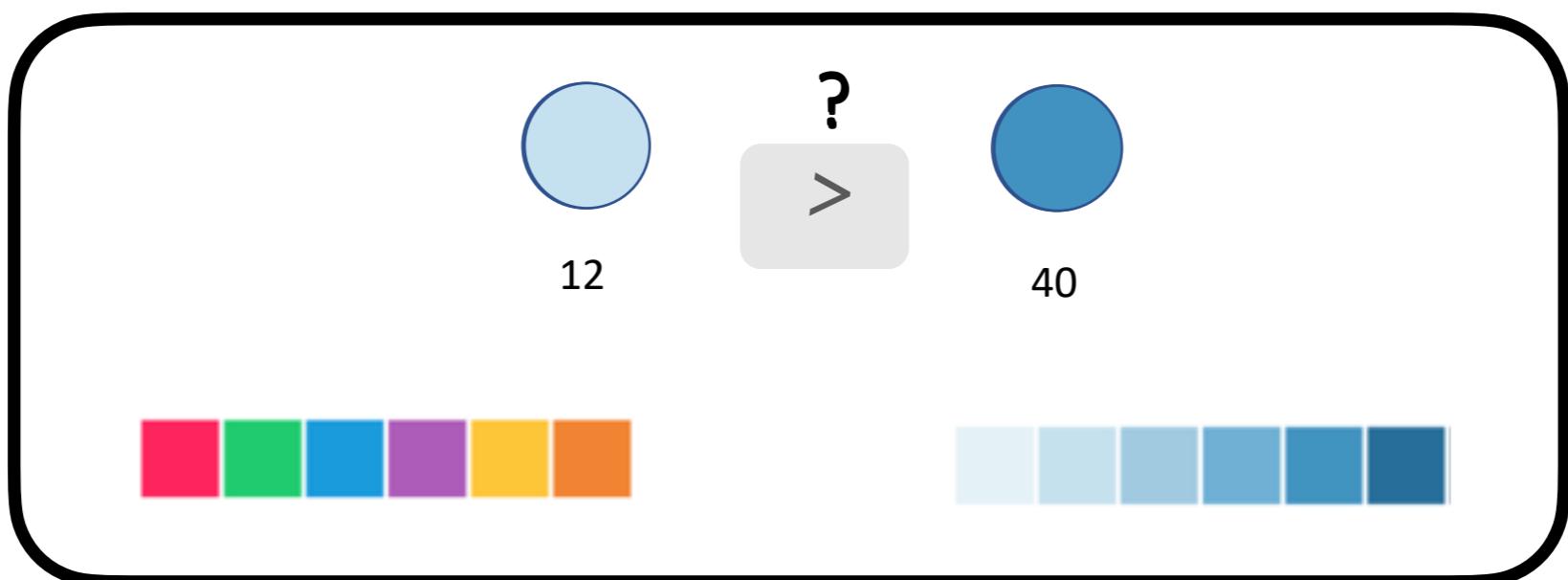


=, ≠, <, >

Quantitative



=, ≠, <, >, +, -, ÷



ATTRIBUTE TYPES

Not Ordered

Ordered

Categorical



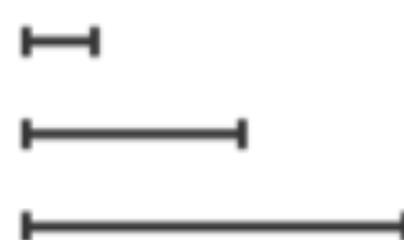
=, ≠

Ordinal

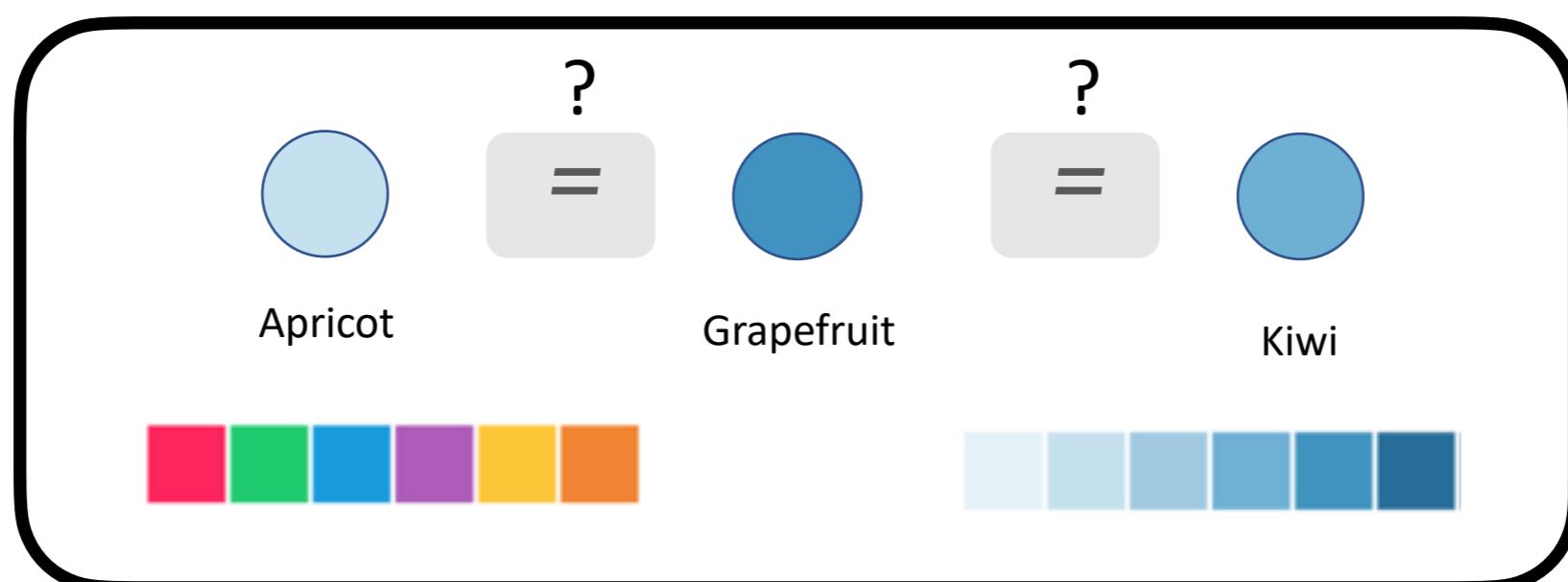


=, ≠, <, >

Quantitative



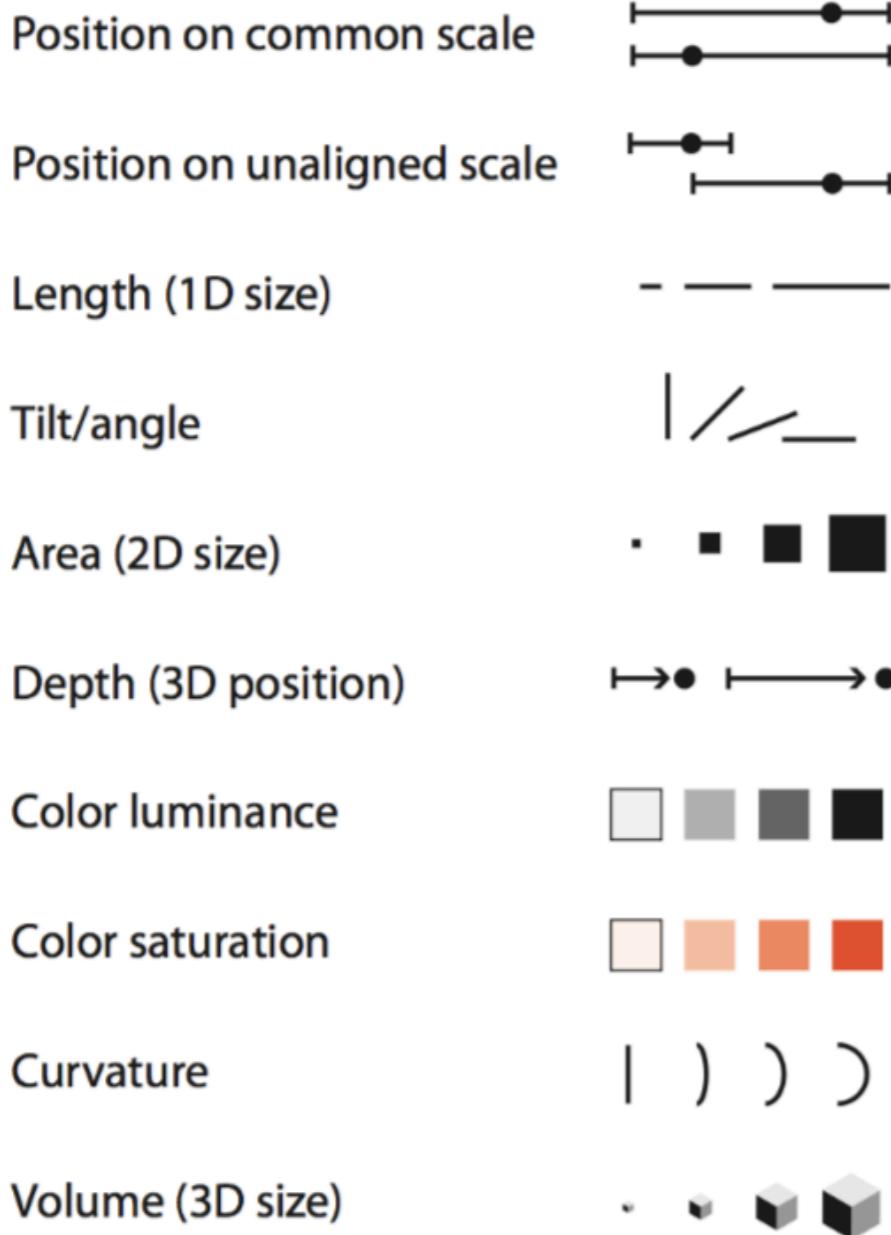
=, ≠, <, >, +, -, ÷



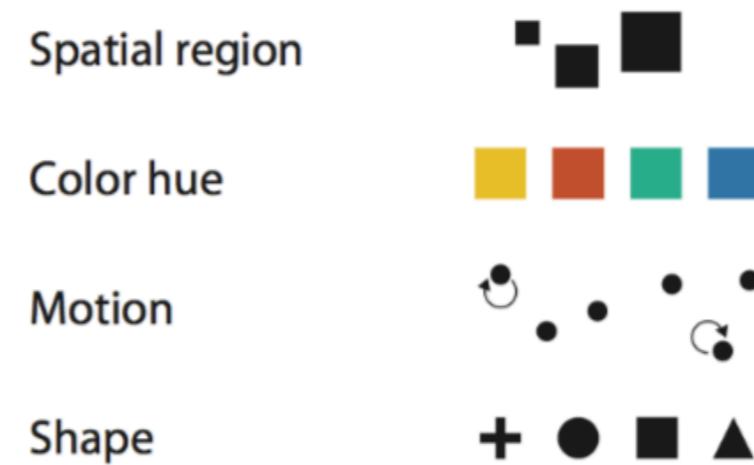
EFFECTIVENESS PRINCIPLE

Some variables are perceptually better than others

④ Magnitude Channels: Ordered Attributes

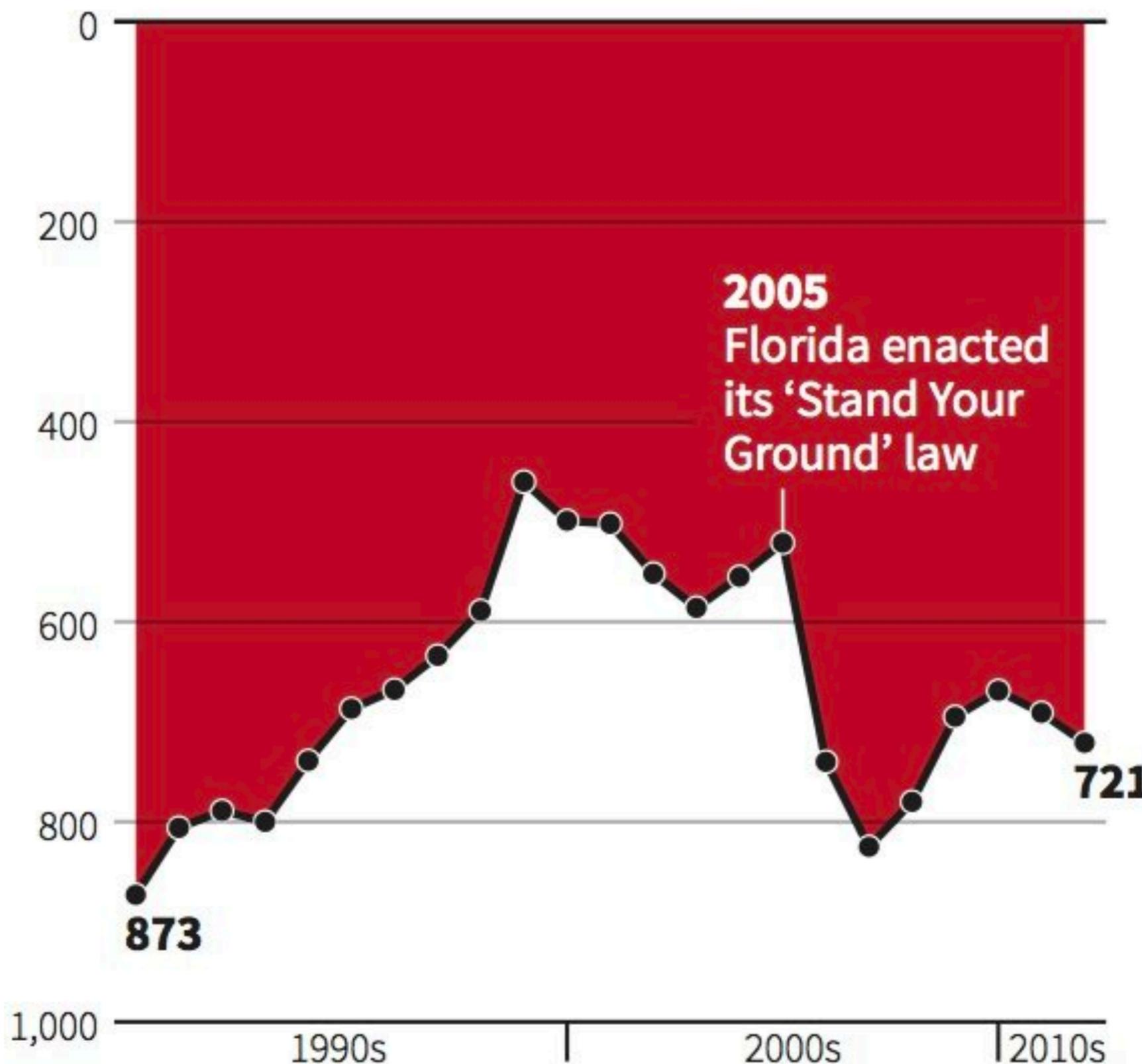


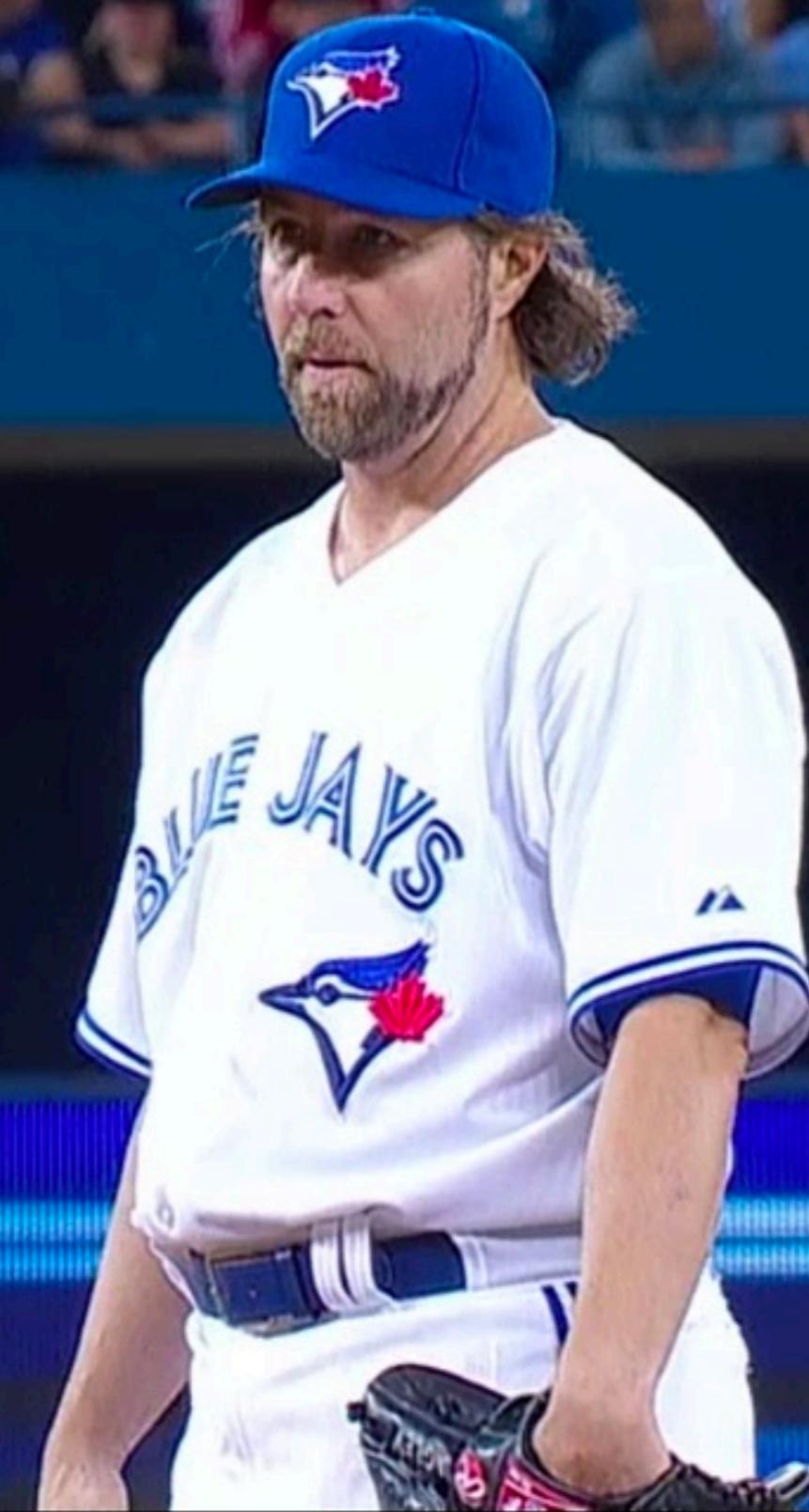
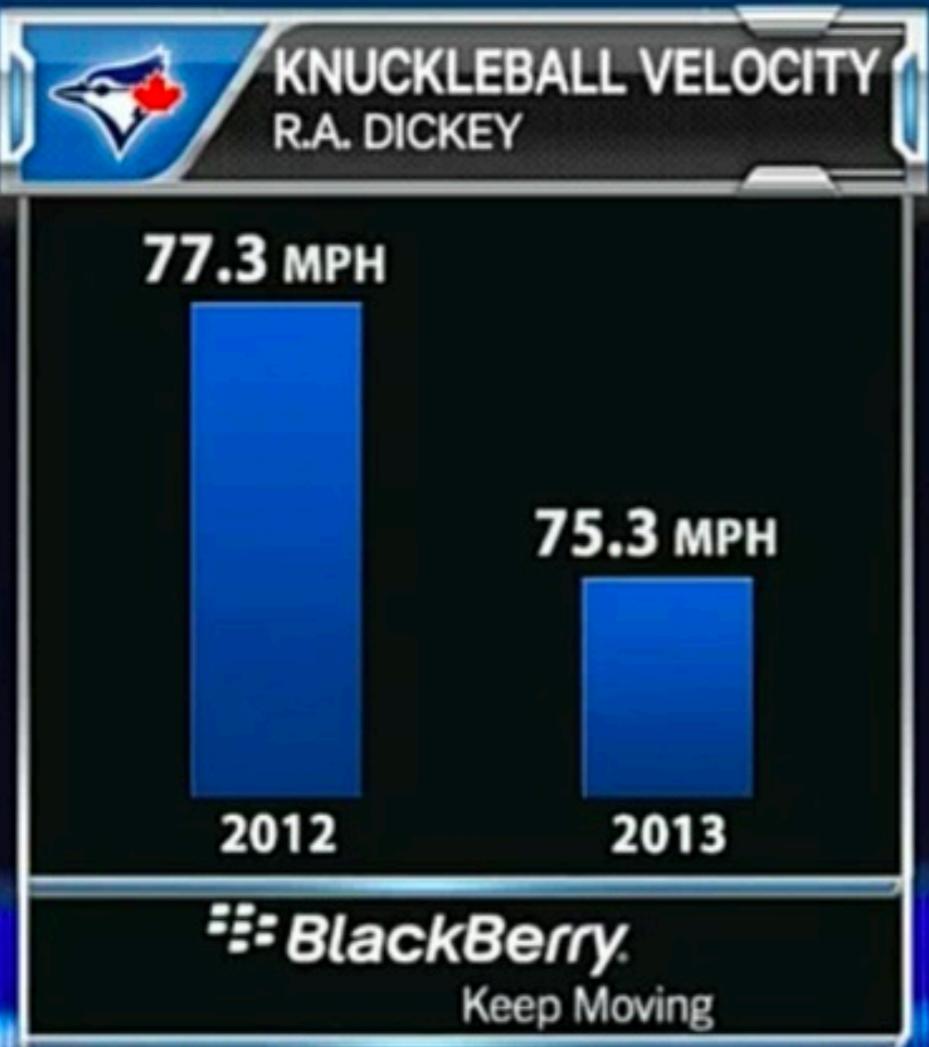
④ Identity Channels: Categorical Attributes

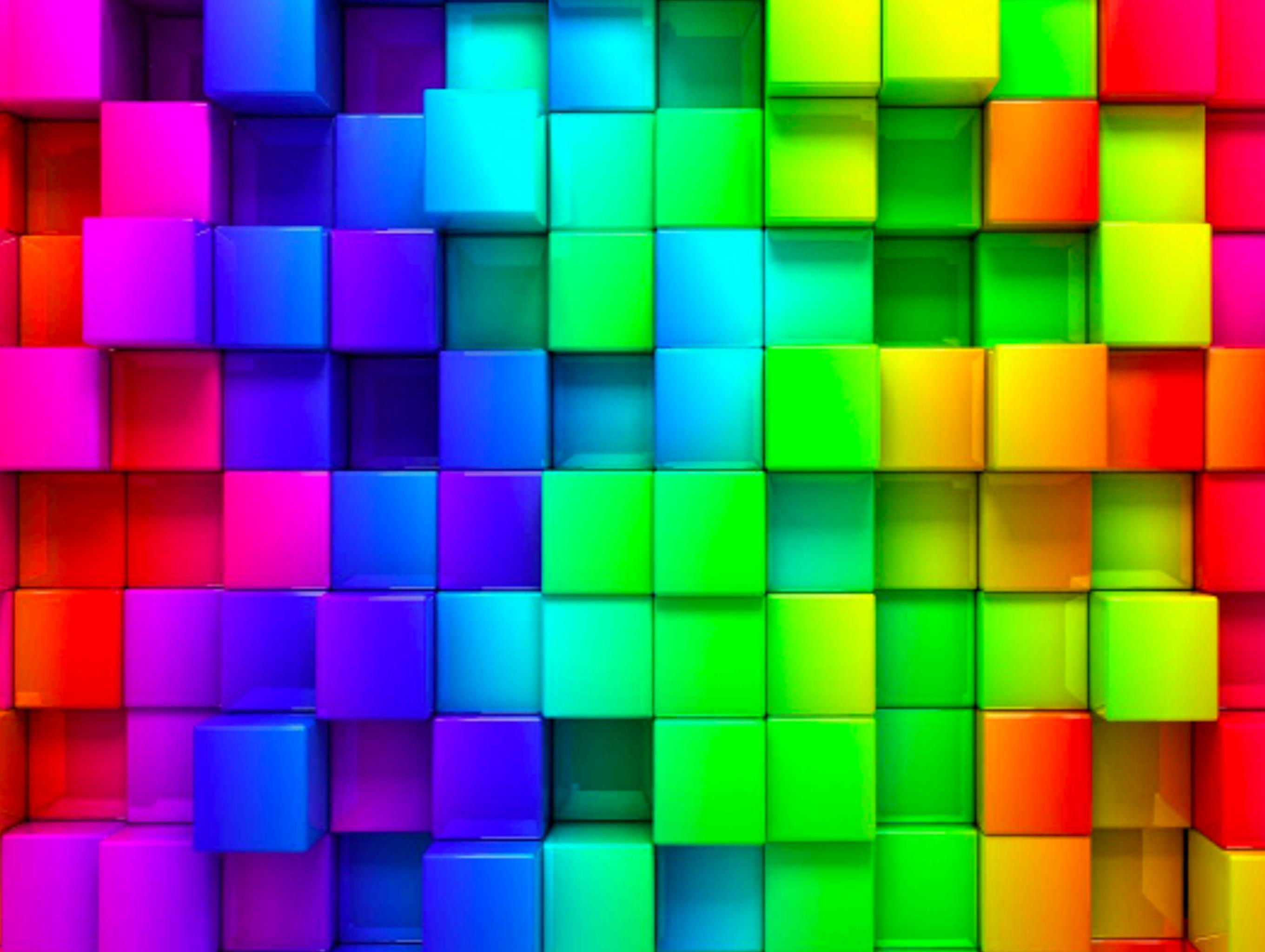


Gun deaths in Florida

Number of murders committed using firearms







COLORMAPS

Not Ordered

Categorical

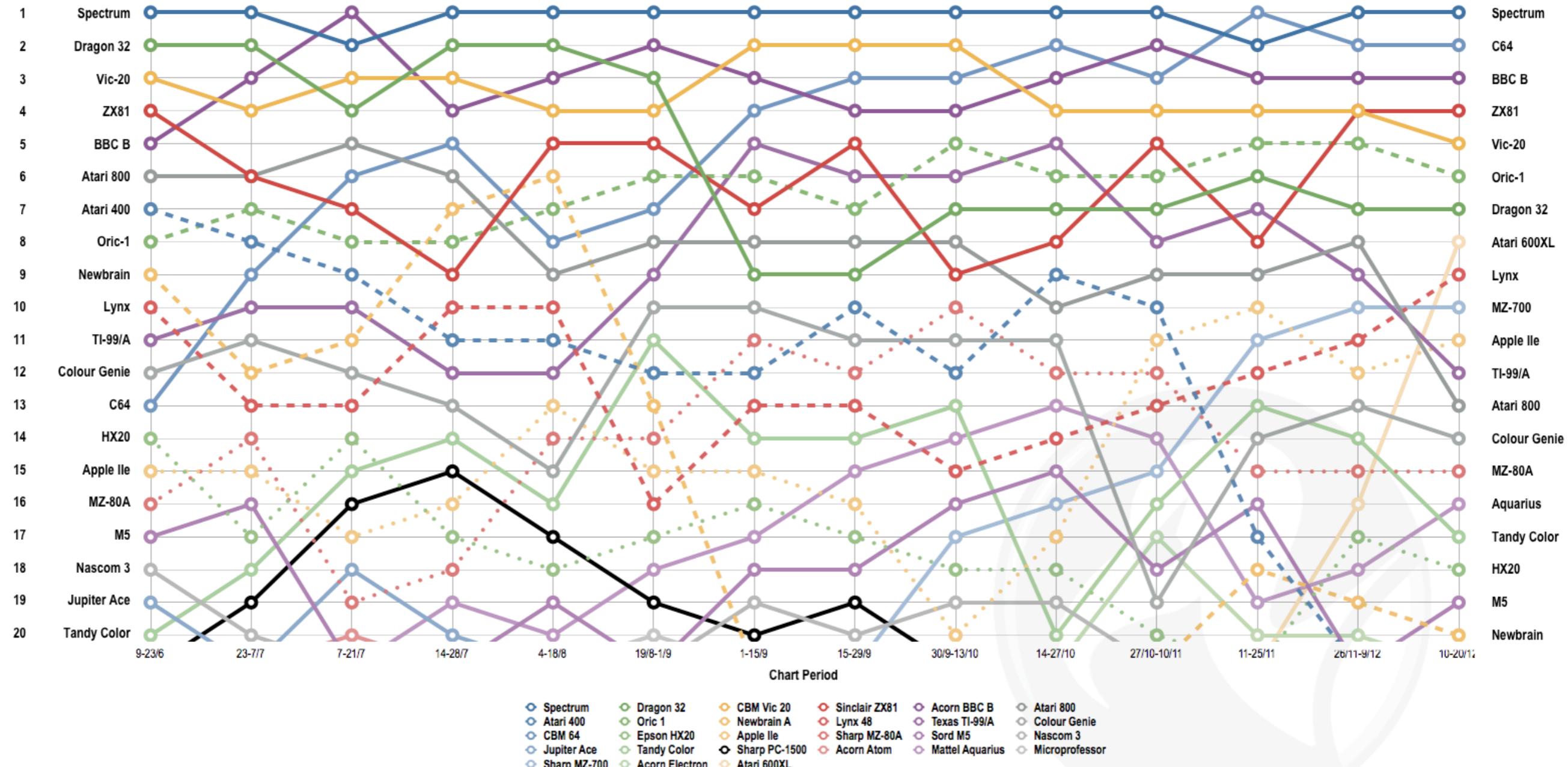


Ordered

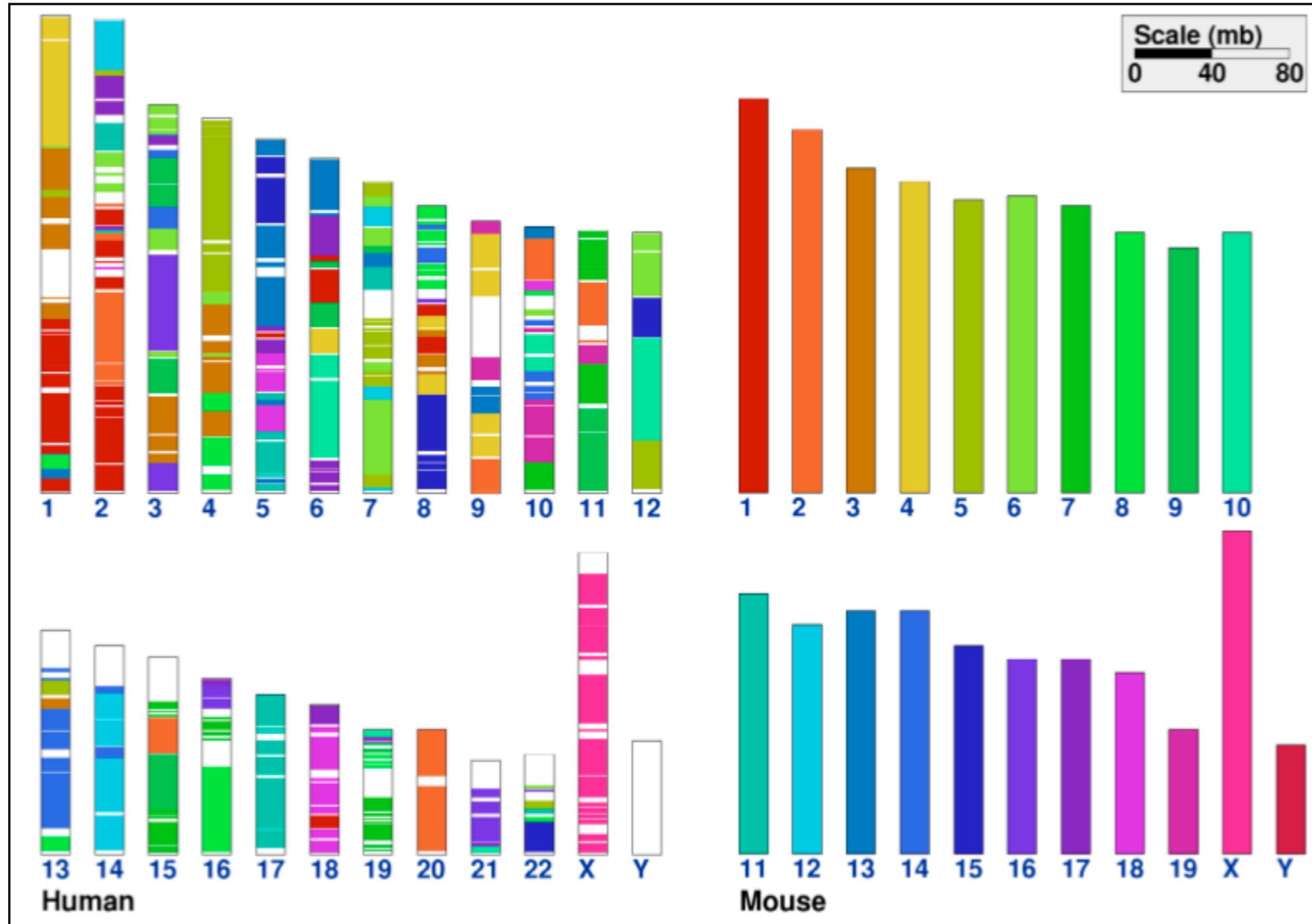
Ordinal, Quantitative



COLOR DISCRIMINABILITY



COLOR DISCRIMINABILITY



Cinteny: flexible analysis and visualization of synteny and genome rearrangements in multiple organisms.
Sinha and Meller. BMC Bioinformatics, 8:82, 2007.

PERCEPTUAL LINEARITY

SANFORD AND SELNICK

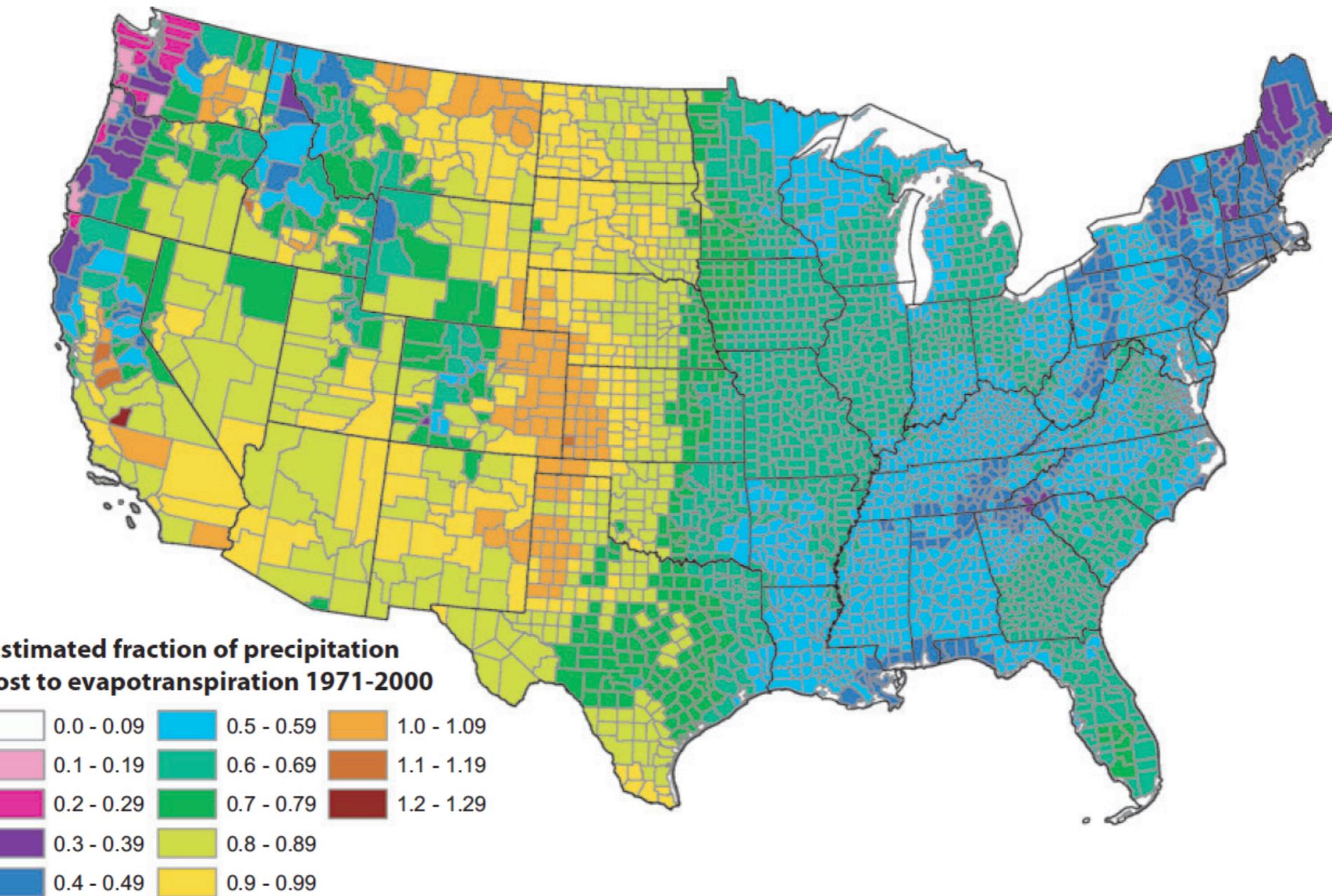


FIGURE 13. Estimated Mean Annual Ratio of Actual Evapotranspiration (ET) to Precipitation (P) for the Conterminous U.S. for the Period 1971-2000. Estimates are based on the regression equation in Table 1 that includes land cover. Calculations of ET/P were made first at the 800-m resolution of the PRISM climate data. The mean values for the counties (shown) were then calculated by averaging the 800-m values within each county. Areas with fractions >1 are agricultural counties that either import surface water or mine deep groundwater.

COLORMAPS

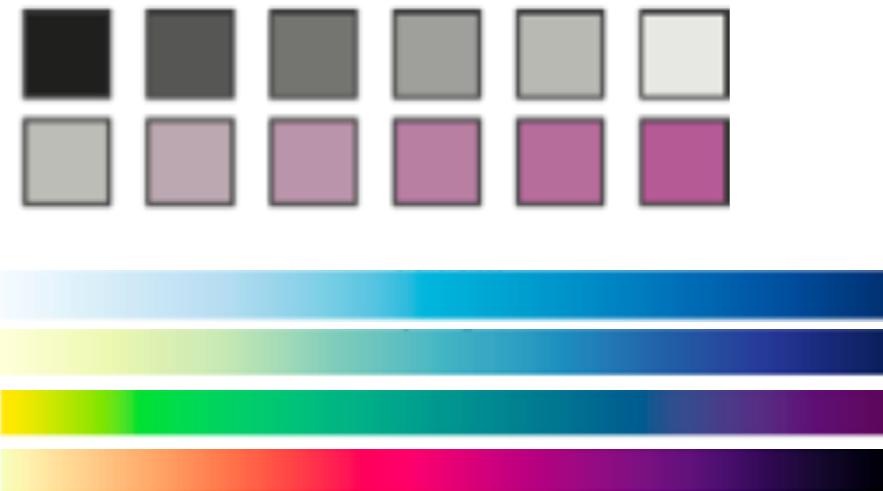
Not Ordered

Categorical

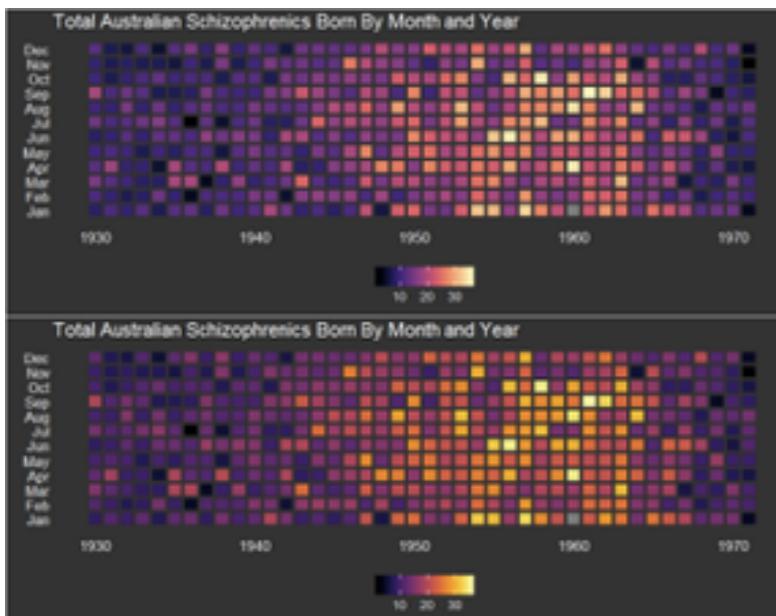


Ordered

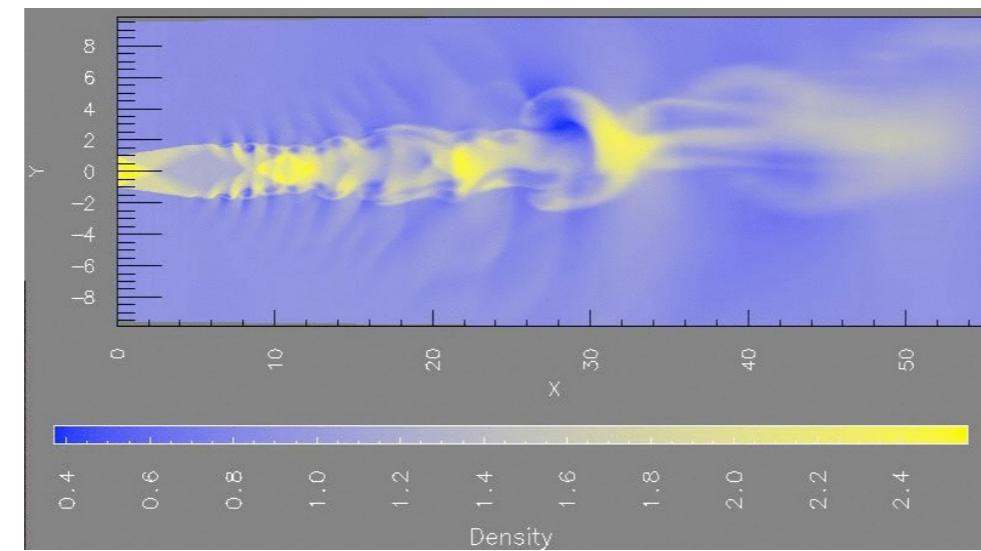
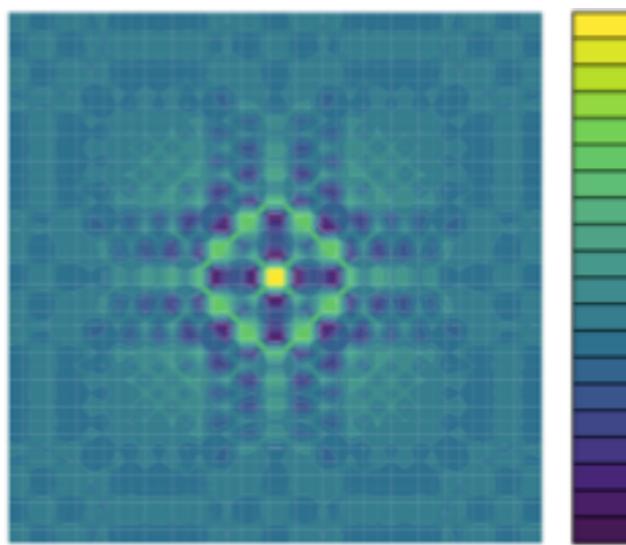
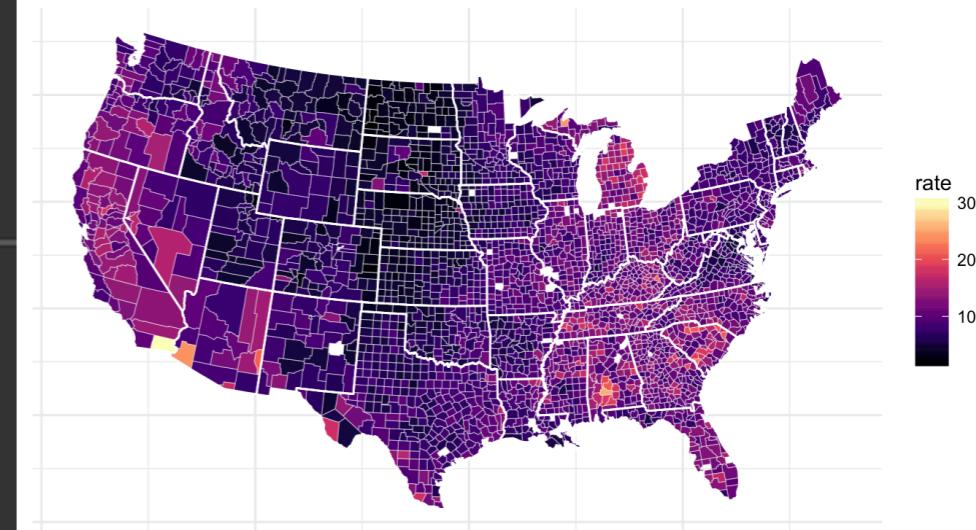
Ordinal, Quantitative



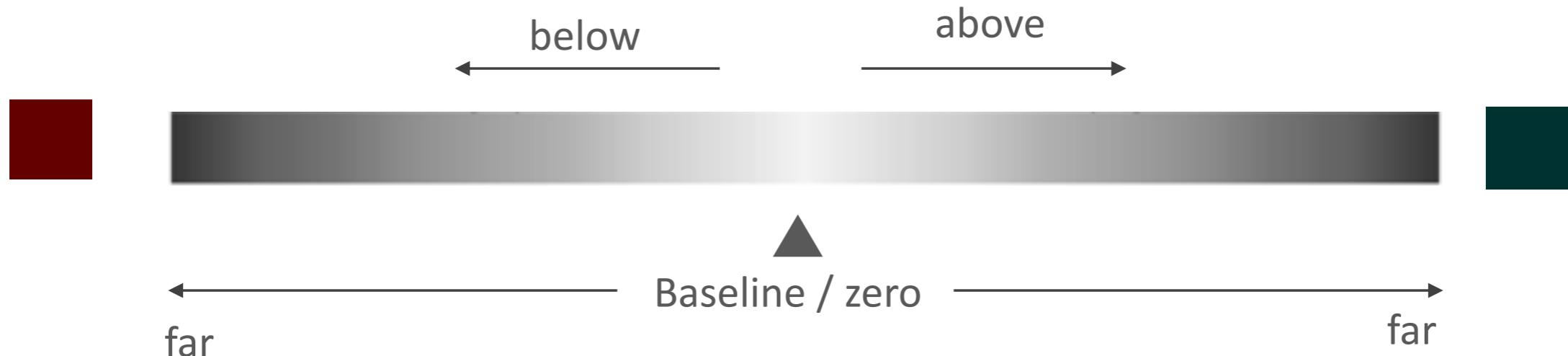
ORDERED COLORMAPS



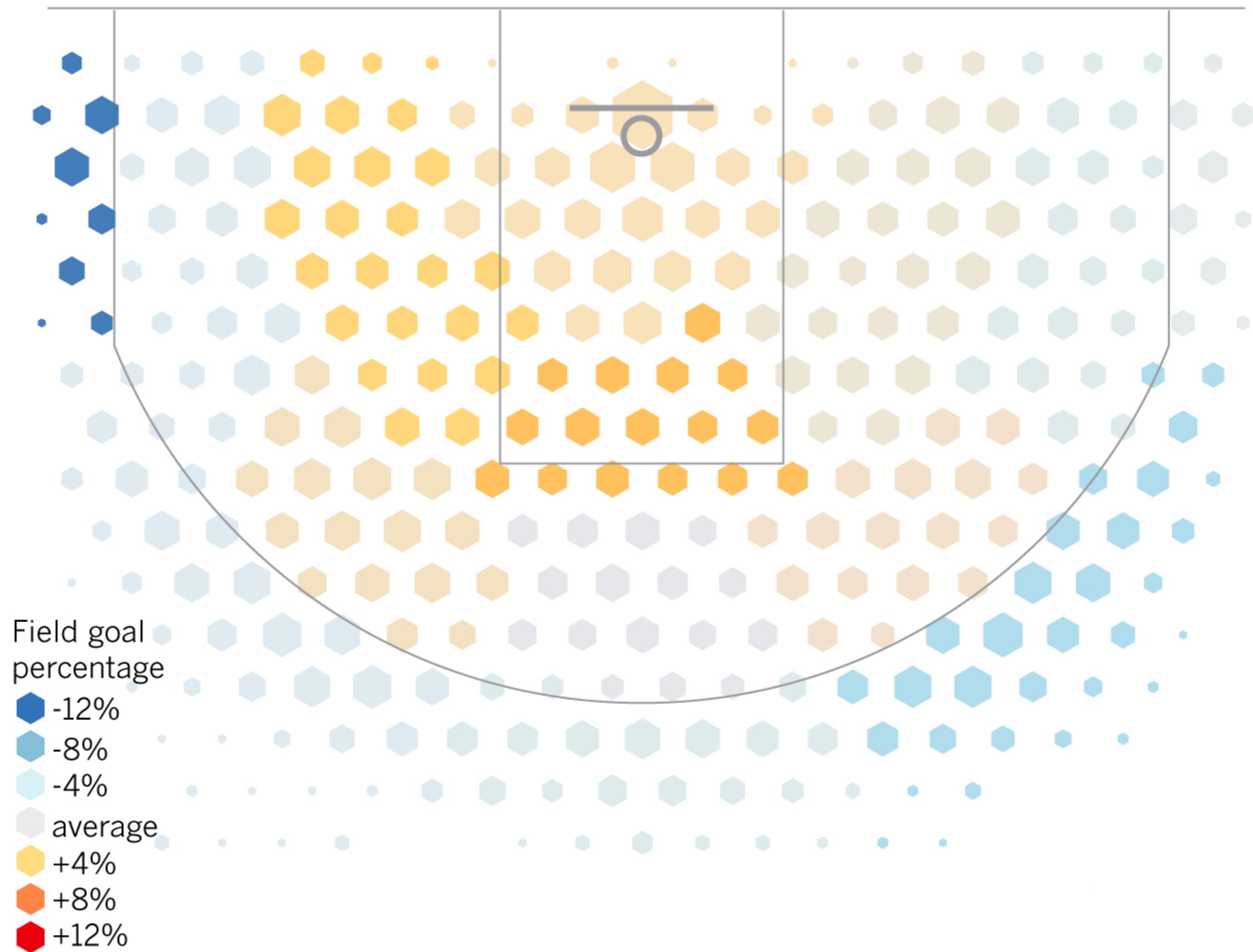
US unemployment rate by county



DIVERGING COLORMAPS

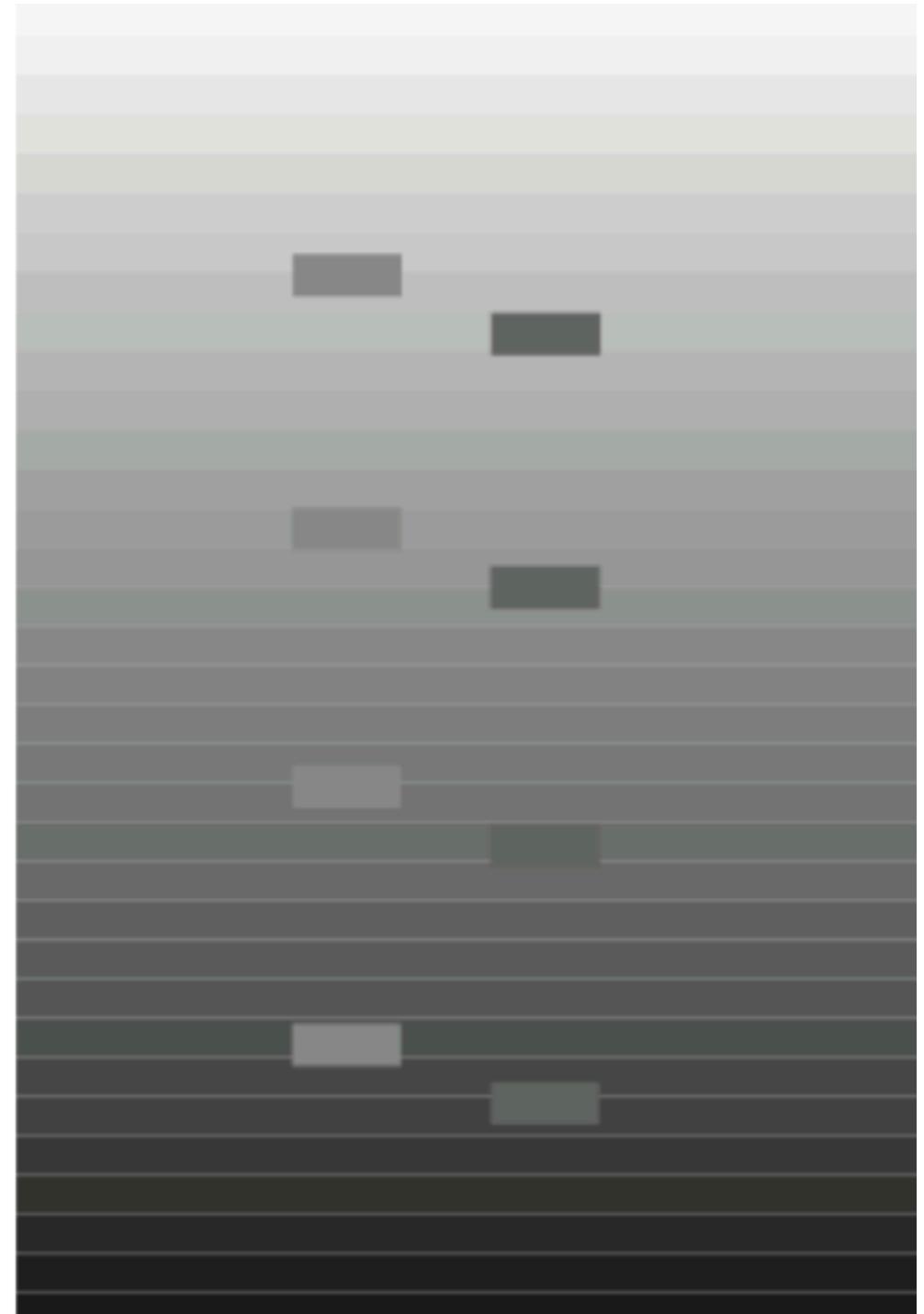


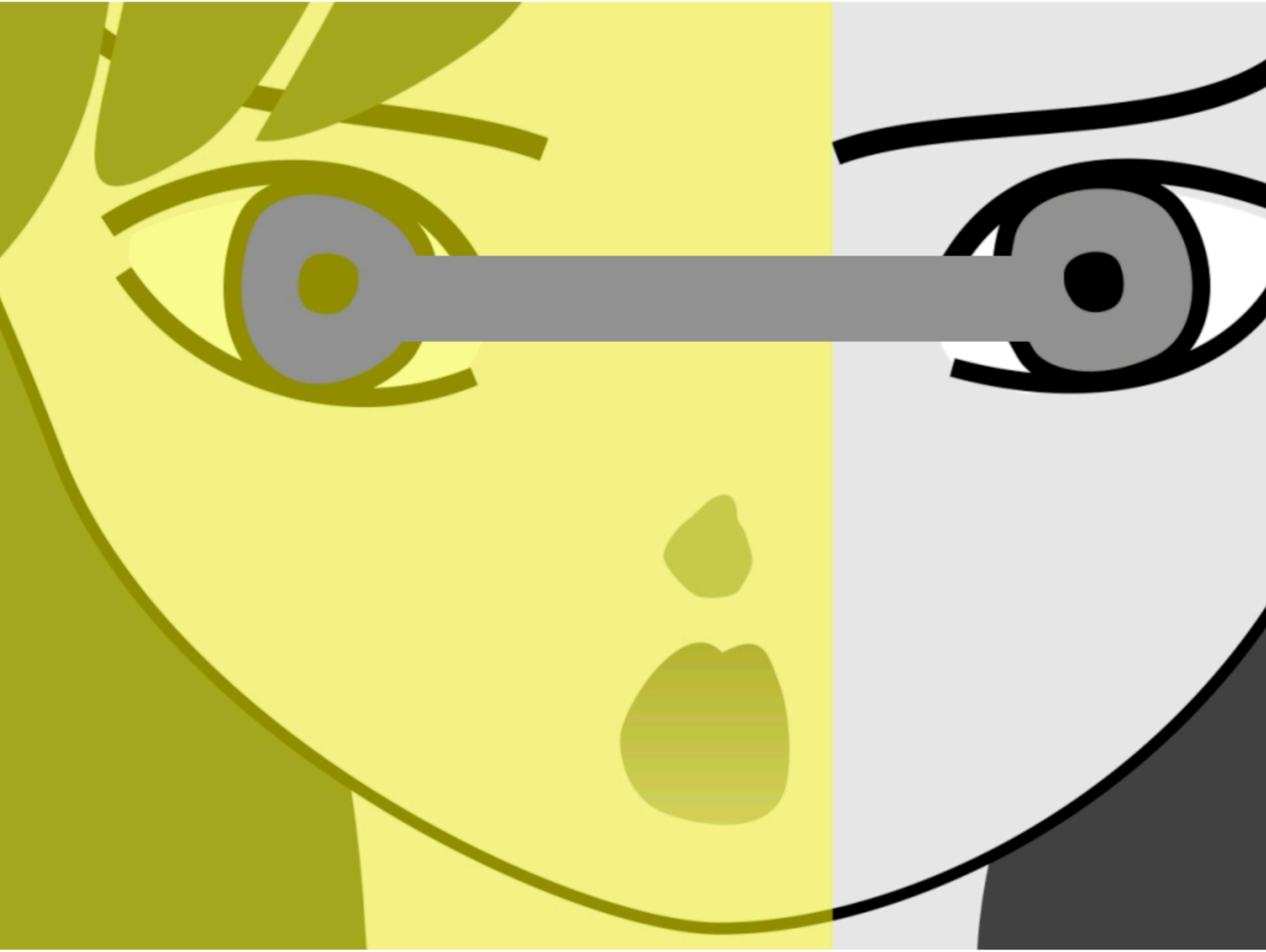
Kobe Bryant vs. league average (1996-2016)



COLOR PERCEPTION IS LOCAL

- Luminance perception is not absolute, but contrast relative ... and local!
- Discern objects(foreground) from background
- Gray color scales are not recommended for more than 5 segments... too “imprecise” to convey quantitative data



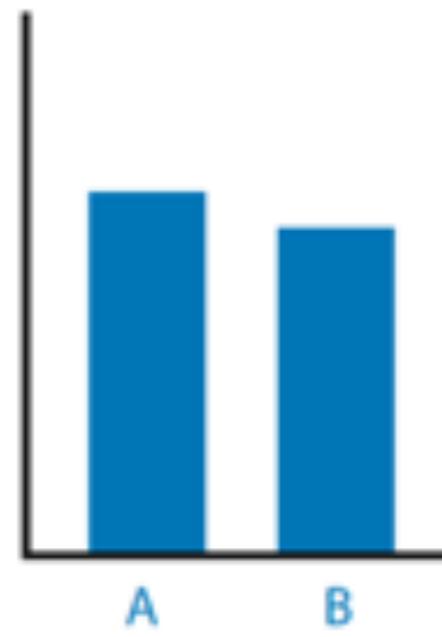


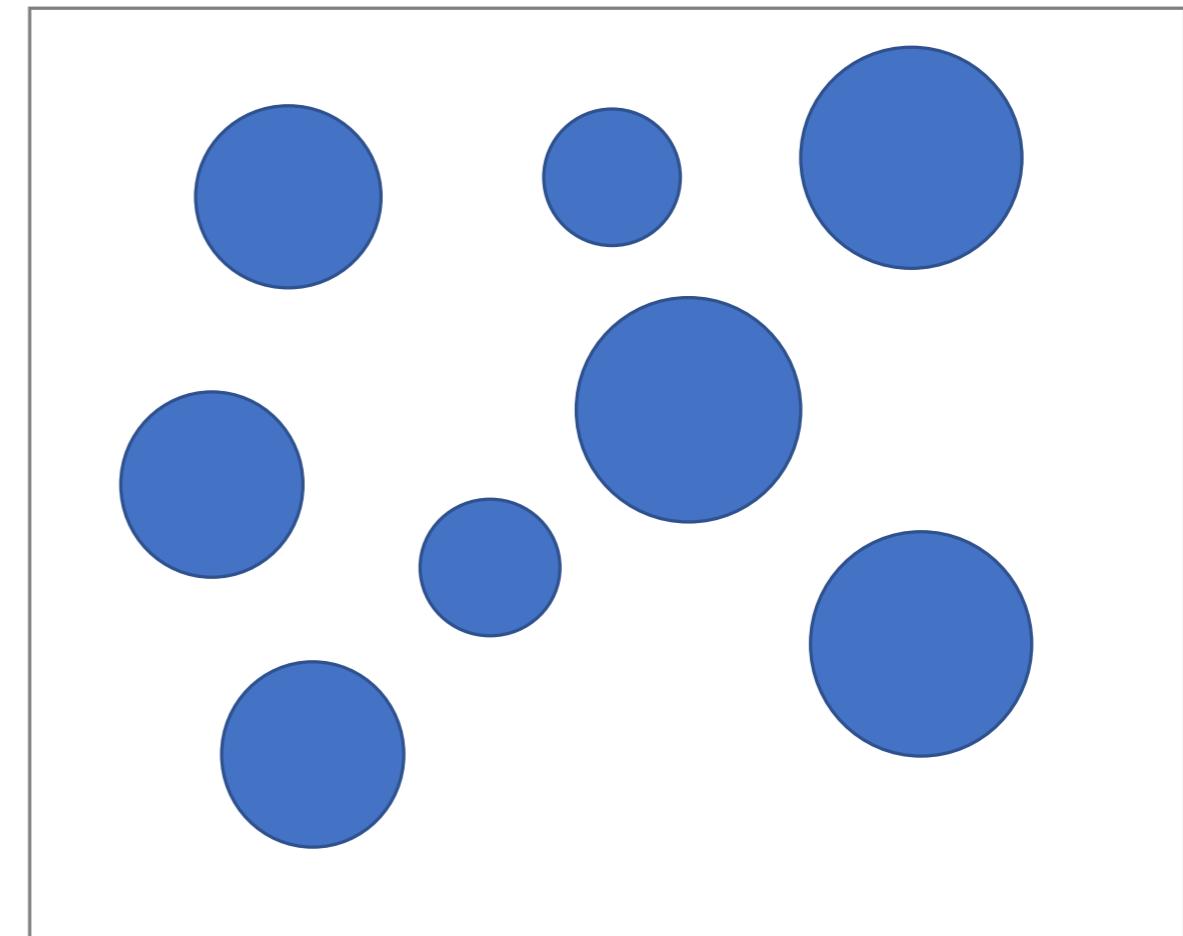
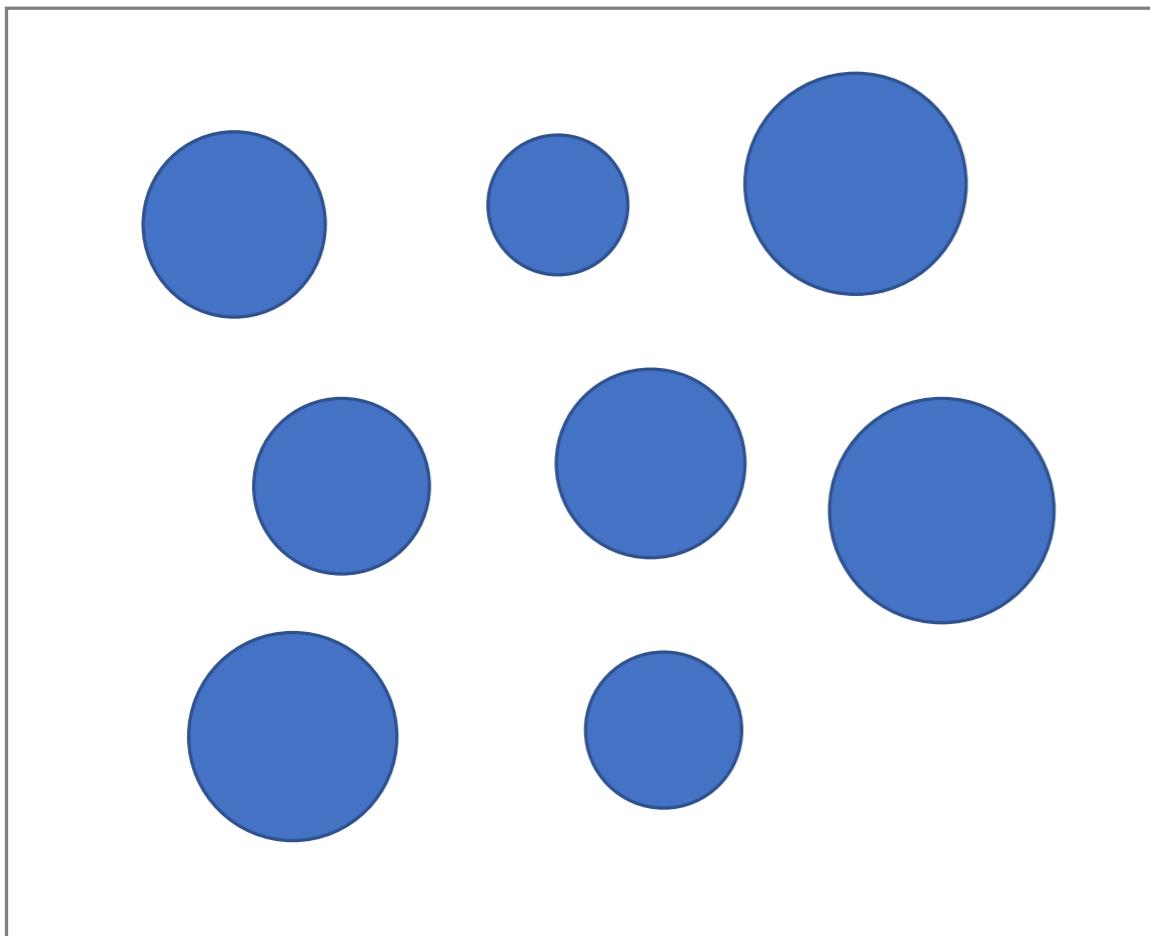
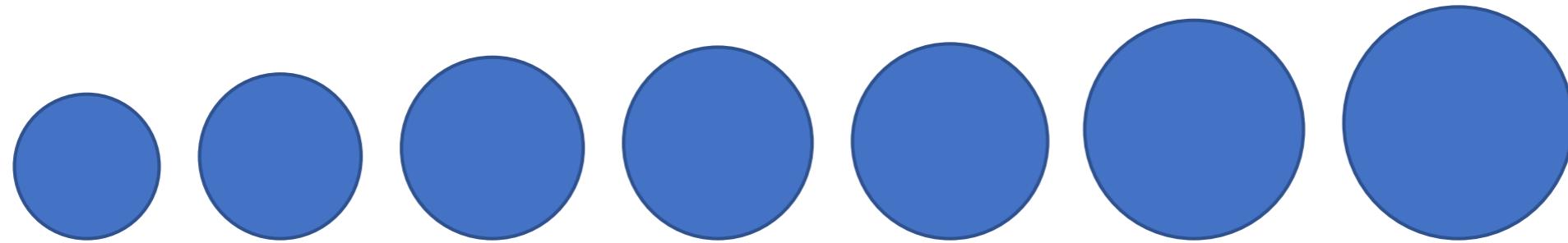


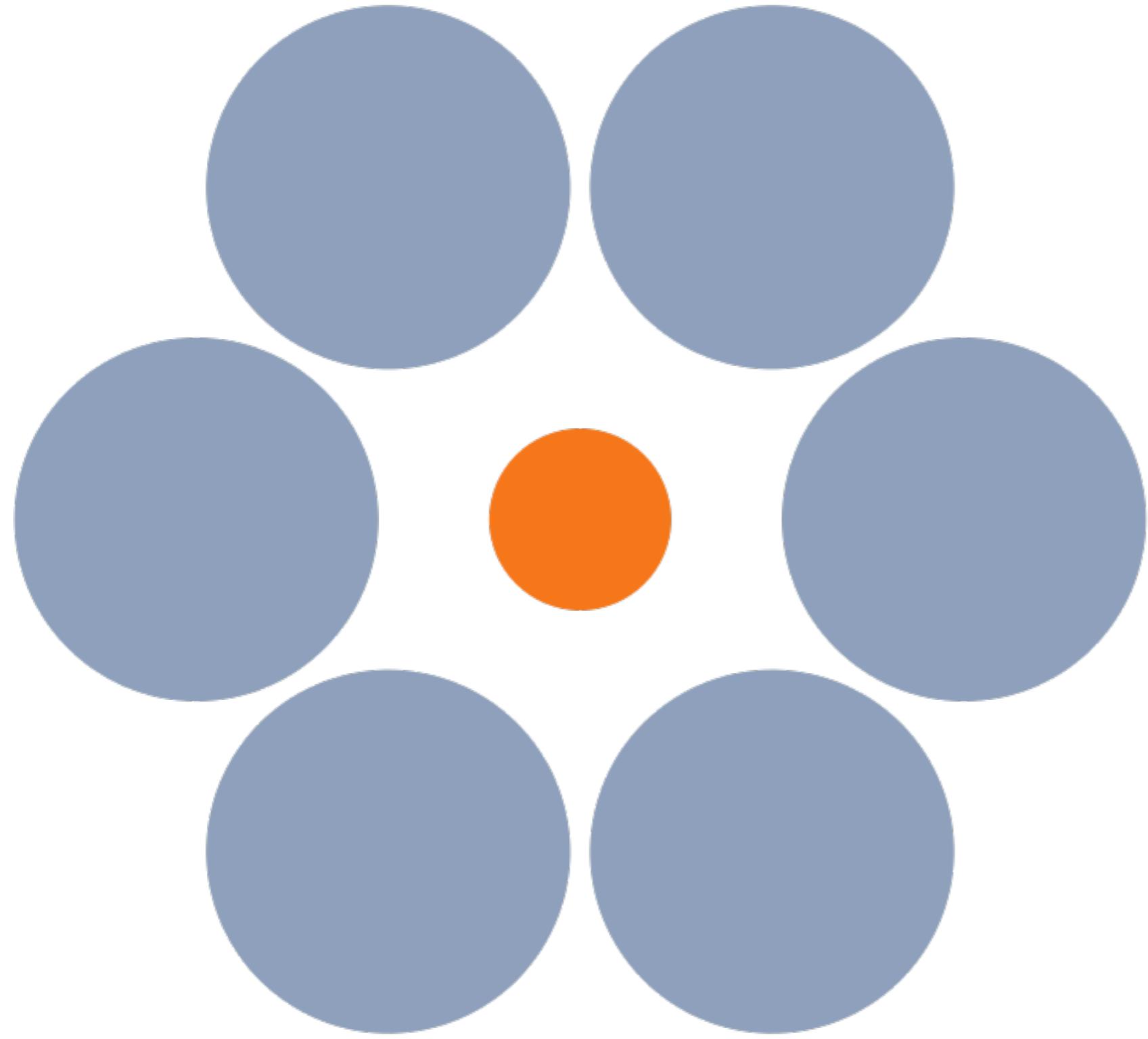


length?

SIZE?

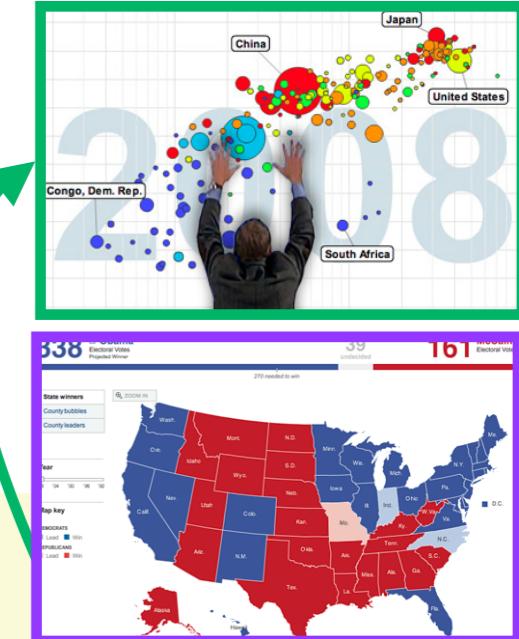
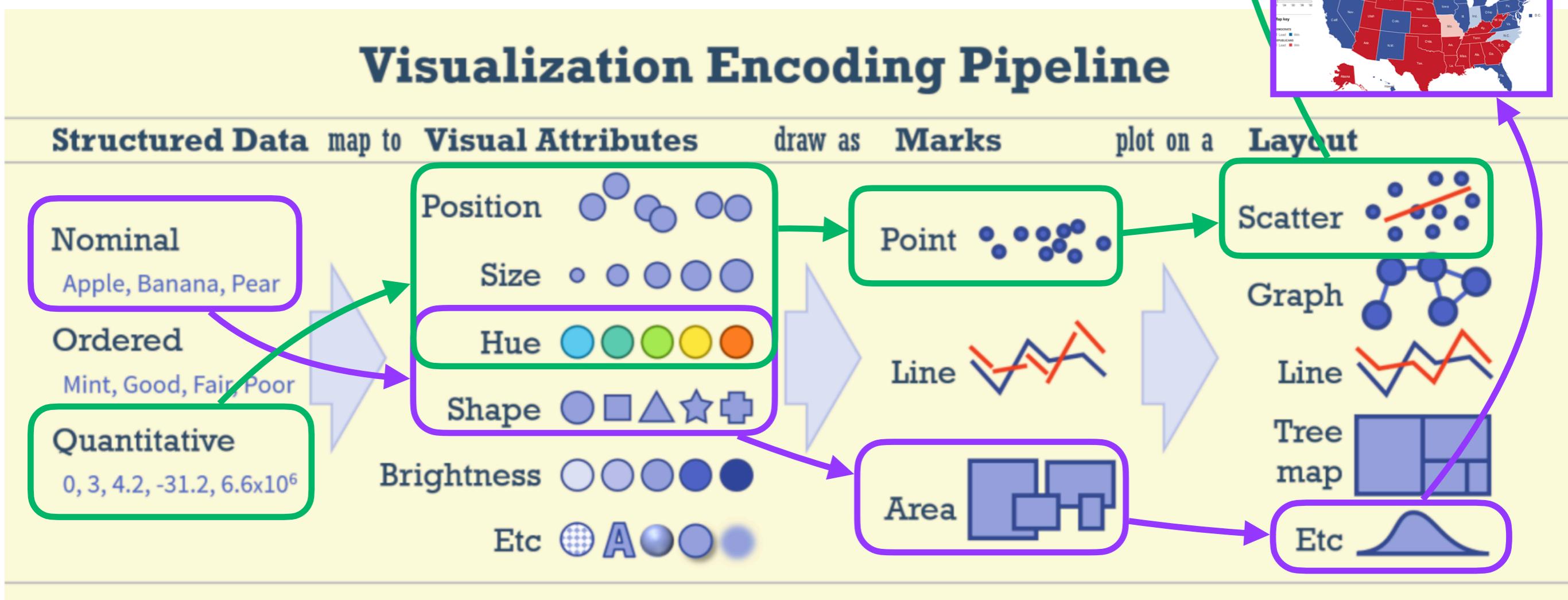






PUTTING IT ALL TOGETHER

VISUALIZATION ENCODING



KNOW YOUR USER'S TASK

“Every [designed] thing is best at something, and worst at something else”

— Bill Buxton

"A tool that serves well for one task can be poorly suited for another, for exactly the same dataset. The task of the users is an equally important constraint for a vis designer as the kind of data that the users have"

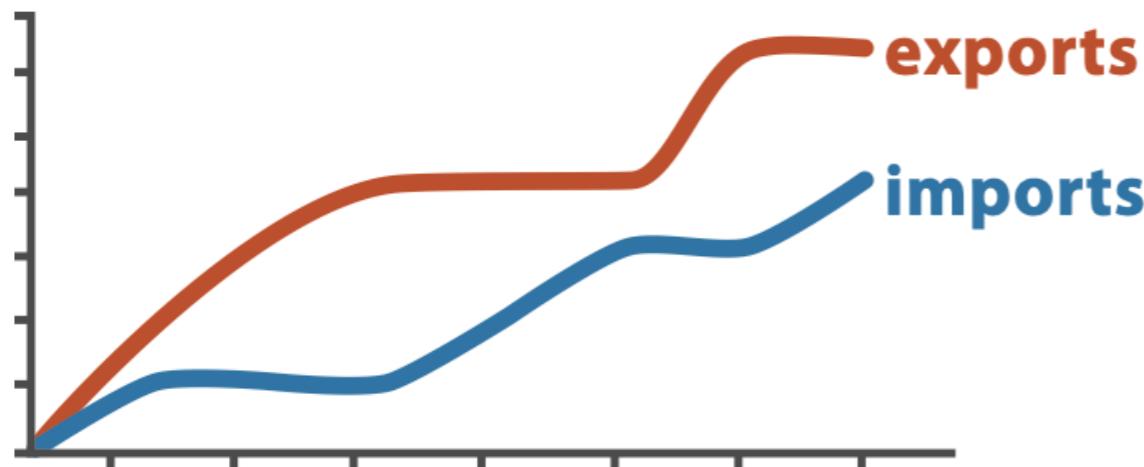
— Tamara Munzner

DON'T JUST DRAW WHAT YOU'RE GIVEN!

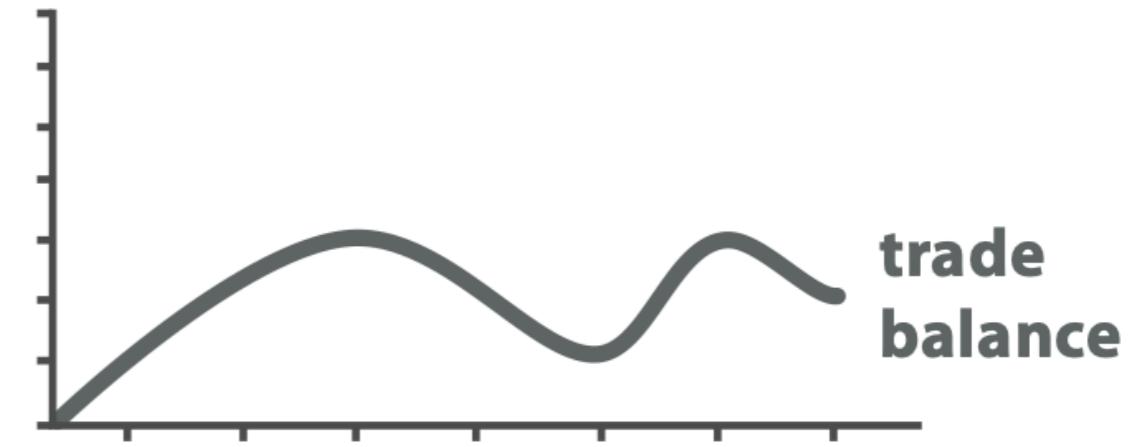
Decide what the right thing to show is

Create it with a series of transformations from the original dataset

Draw that



Original Data



$$\text{trade balance} = \text{exports} - \text{imports}$$

Derived Data



Examples, lots of
examples!

Fundamentals

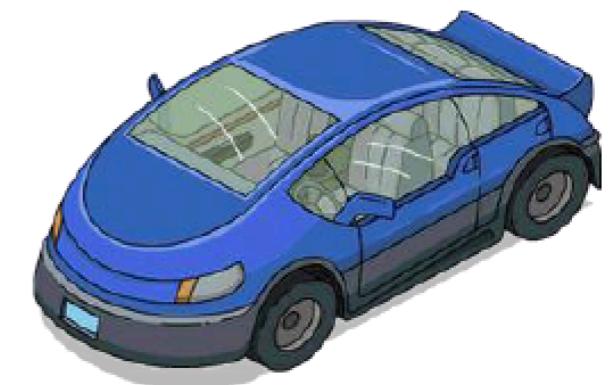
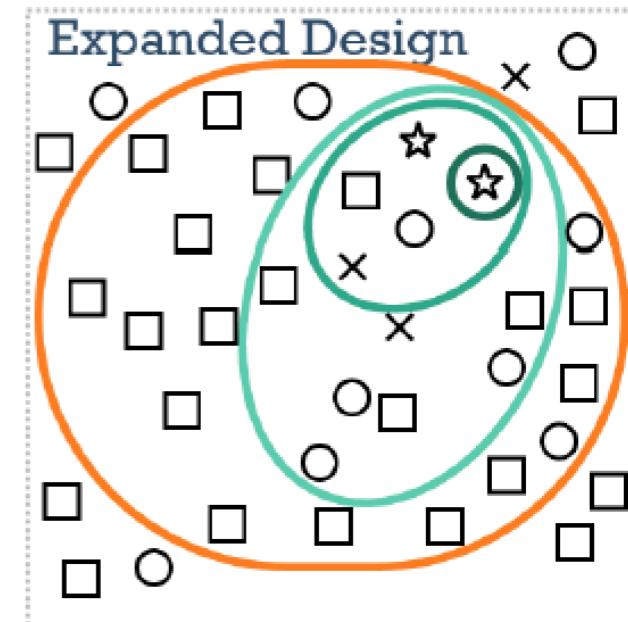
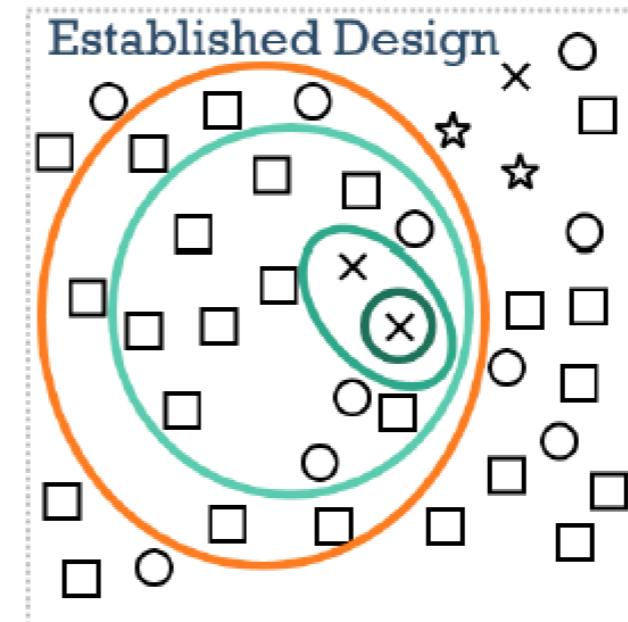
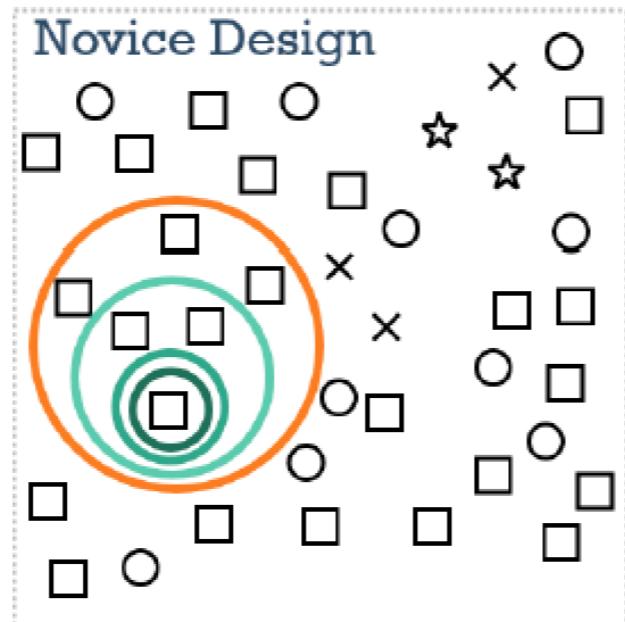
DESIGN SPACE?

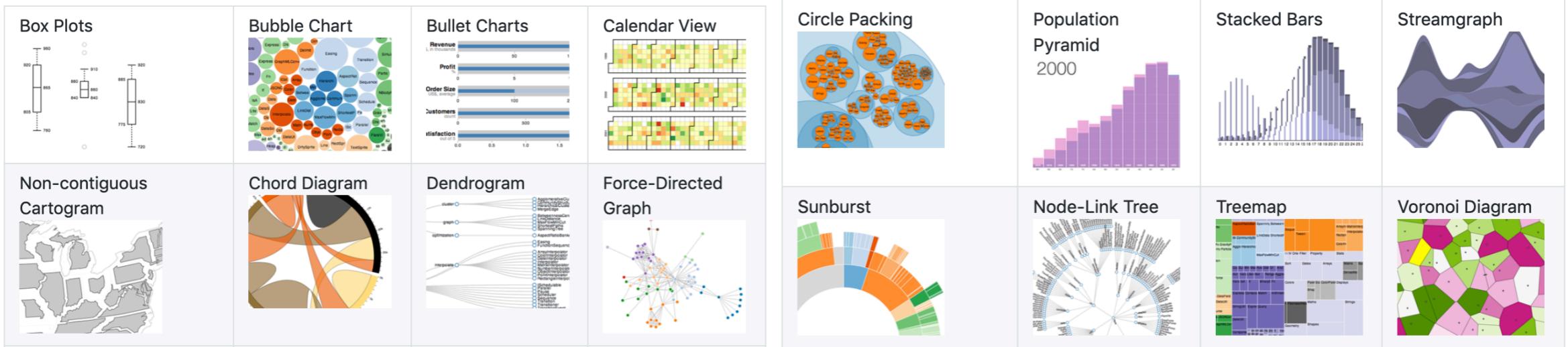
The set of possible design elements, parameters, and configurations that meet the specific application objectives.

WHY DOES THE DESIGN SPACE MATTER?

- Space of all possible solutions
- Design space
- Consideration space
- Proposal space
- Selected solution

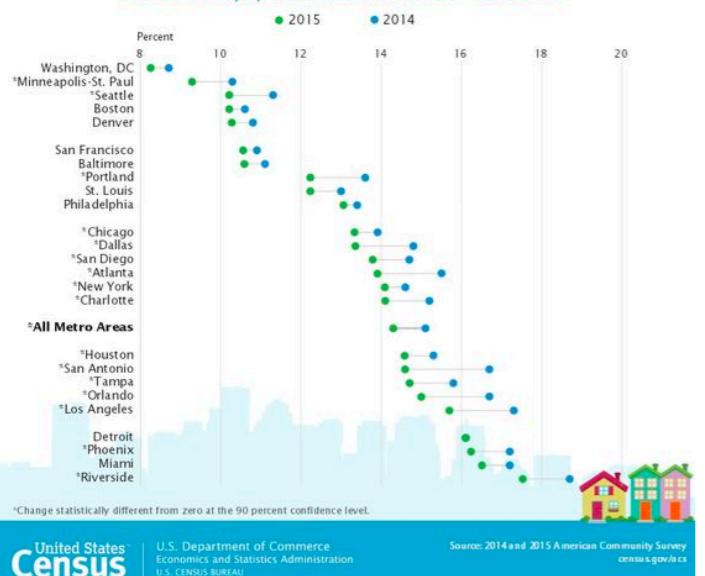
- Poor solution
- OK solution
- ✗ Good solution
- ★ Better solution



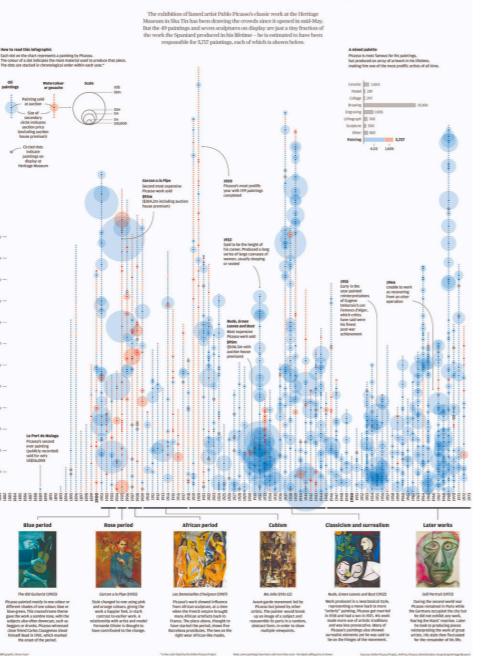


Poverty in the United States

The 25 most populous metro areas: 2014 and 2015



Picasso's paintings



In the Shadow Of Foreclosures

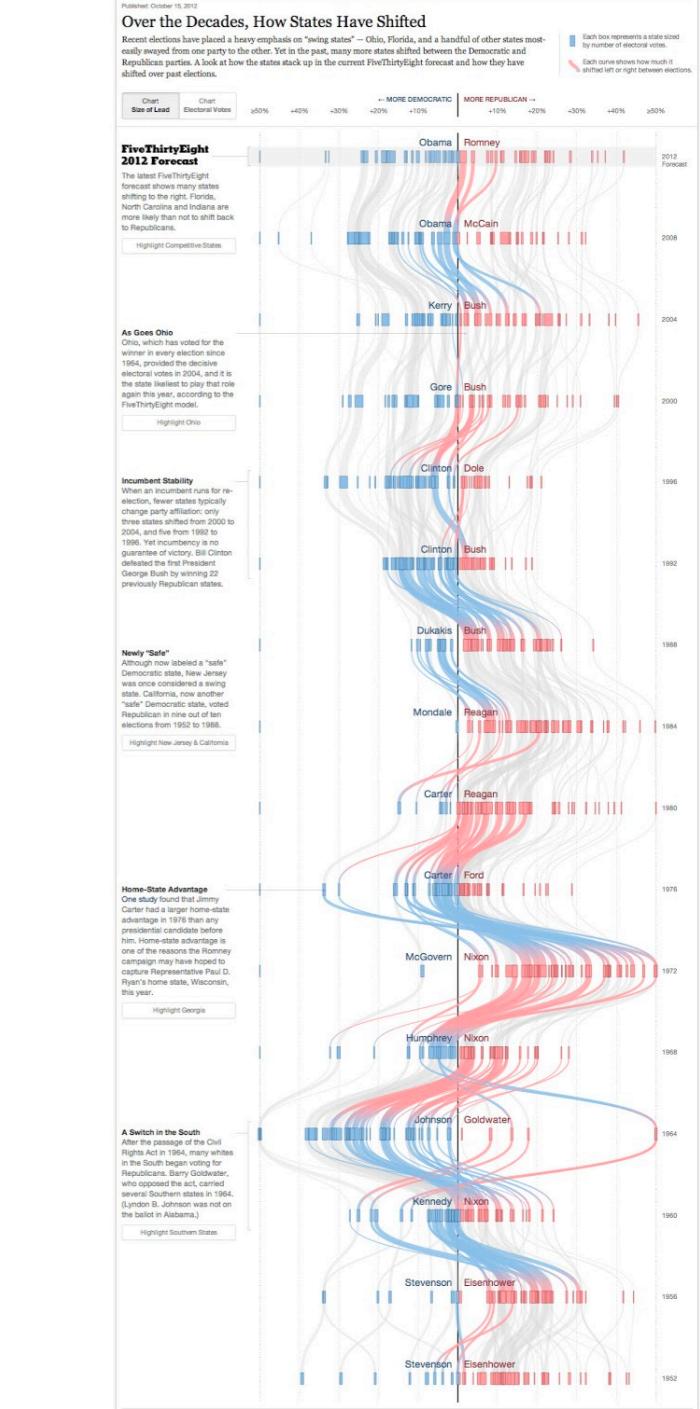
In the subprime mortgage squeeze, some regions are feeling the pain more acutely than others. Although most Southern metropolitan areas have high percentages of subprime mortgages, homeowners in those areas have been forced to pay their bills, so subprime foreclosure rates are low.

Not so in the Rust Belt, where subprime mortgages are less common but foreclosure rates are sky-high, mostly a result of overbuilding.

And overbuilding in regions of Florida, California and other states with large real estate bubbles has led to speculators buying up several homes at once, hoping that their values would rise. Getting second loans was all too easy.

But as housing prices fall, foreclosing prices are all too hard, and many economists believe that foreclosures will continue.

"The collapse will affect other markets," said New York City and D.C.-based David Baker, co-director of the Center for Economic and Policy Research. "Suburban areas near those

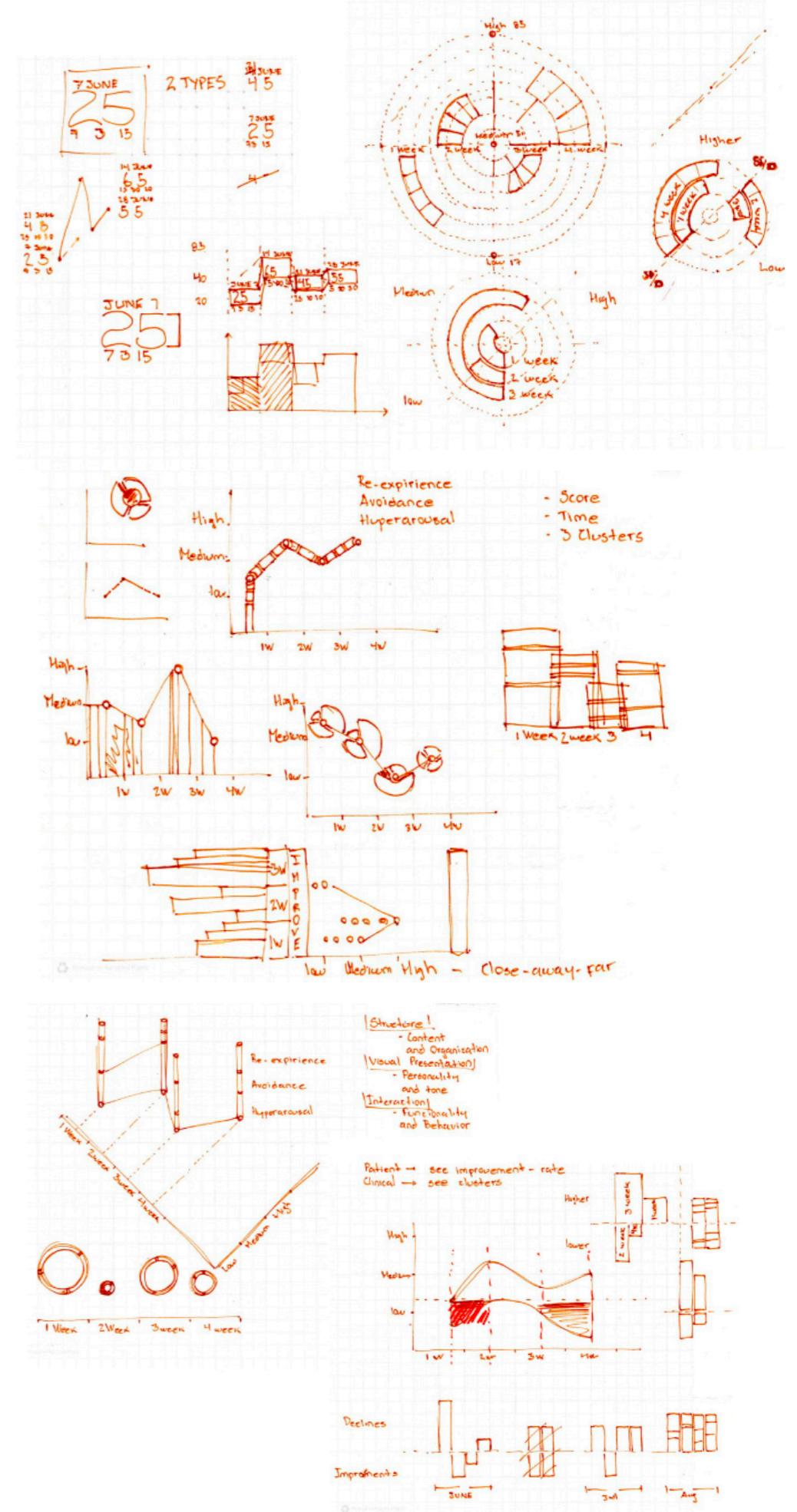


SKETCH!

Don't commit to a solution immediately; (easy to make, easy to discard)

Prototype ideas in early stages

Share and discuss your designs with other people

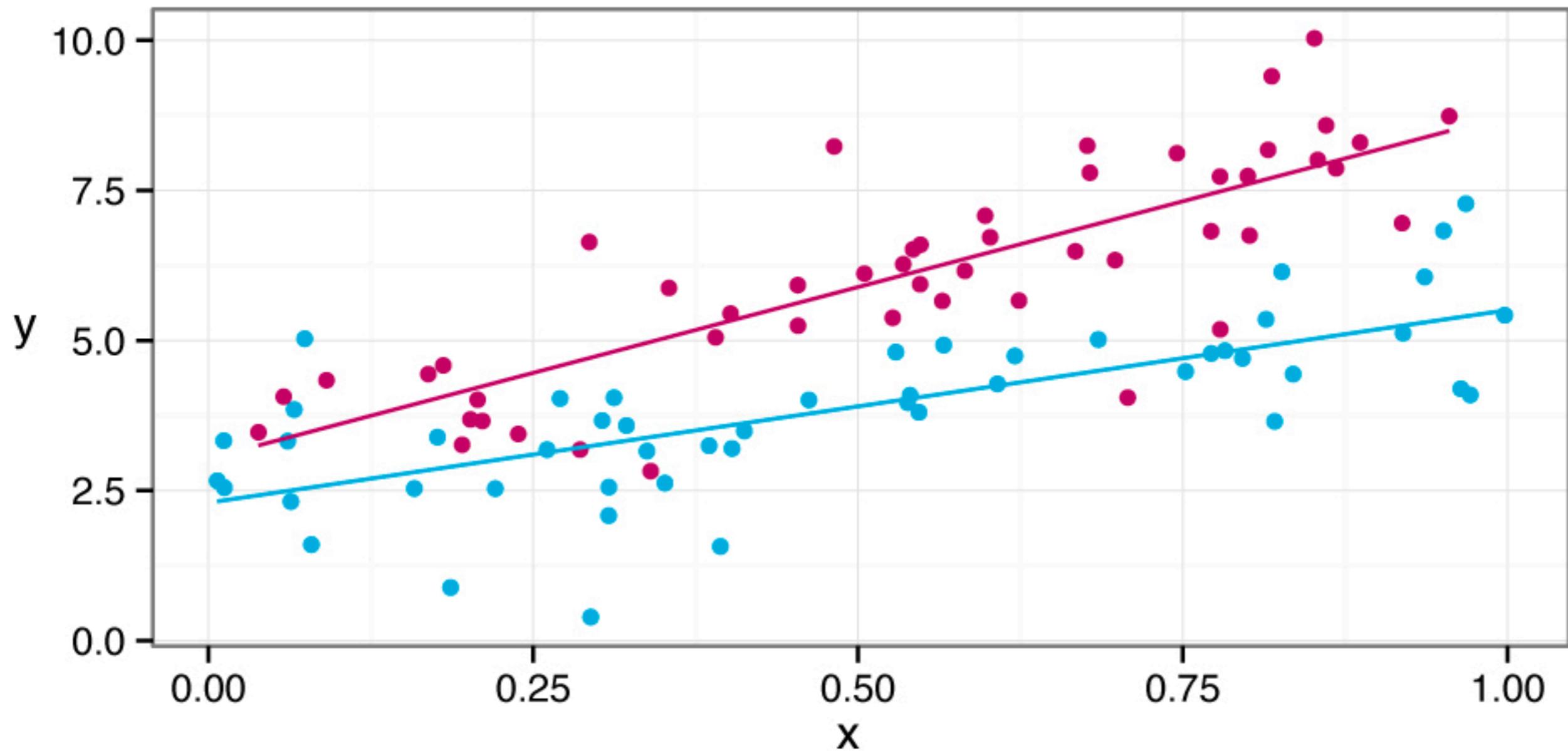


DON'T JUST GRAPH IT, FIT IT!

Graphing is needed, to see the details of the raw data

Fitting data to mathematical functions is needed too!

You can display both in the same graph...



REFERENCES

EXPAND YOUR HORIZON

Lookup

Visualization Catalogs

www.datavis.ca

<http://www.datavis.ca/milestones/>

(historical!)

www.infovis-wiki.net

http://www.visual-literacy.org/periodic_table/periodic_table.html

Color

<http://colorbrewer2.org/>

Color Brewer

<http://tools.medialab.sciences-po.fr/iwanhue/>

I Want Hue

To Reflect:

<http://hint.fm/>

F. Viégas & M. Wattenberg

www.edwardtufte.com/bboard

E. Tufte

<http://fellinlovewithdata.com/>

Fell in love with data

E. Bertini

<http://datastori.es/> (podcasts)

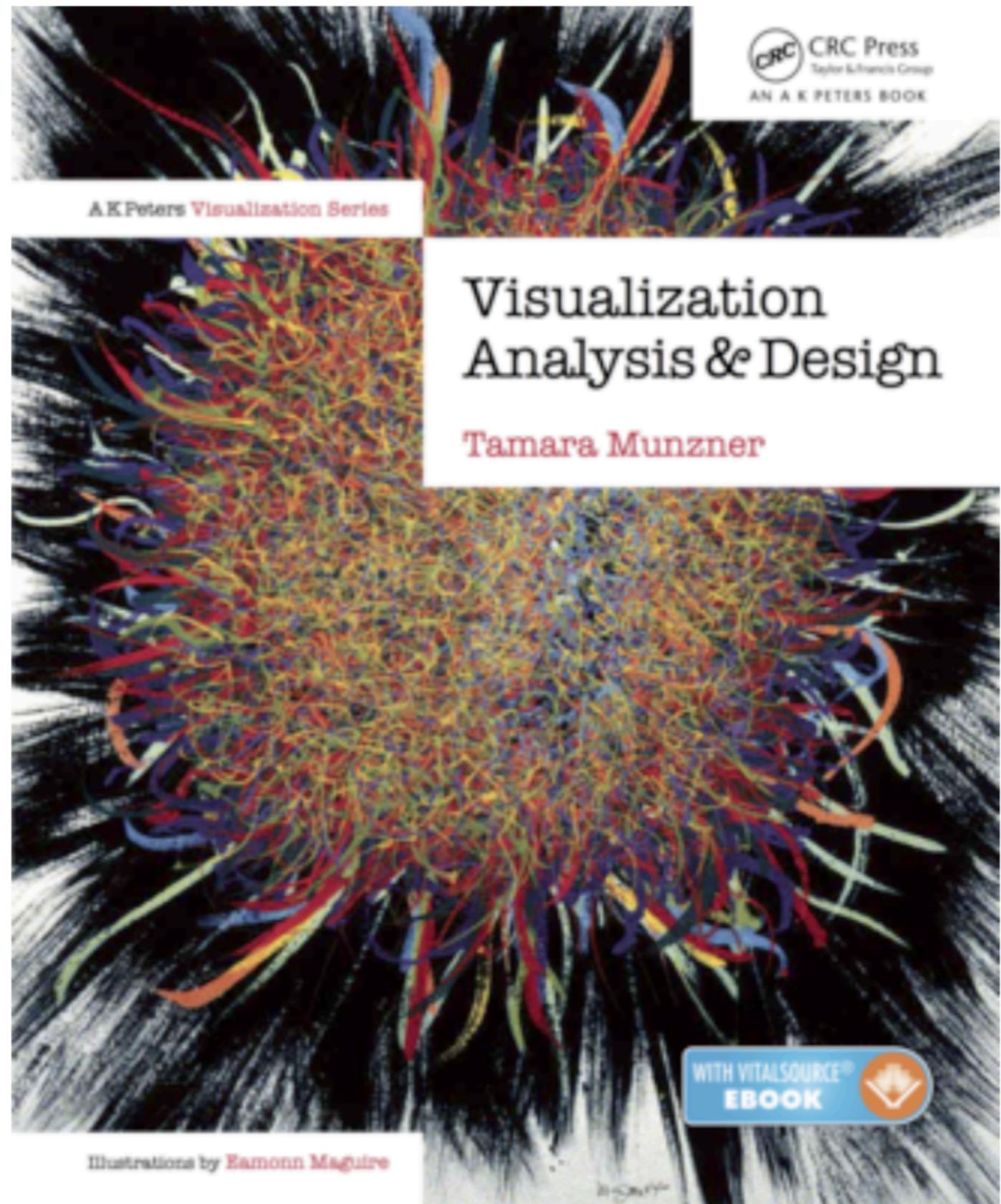
E. Bertini & M. Stefaner

<http://junkcharts.typepad.com/>

Junk Charts: learn from others (mistakes)

Text-book on data visualization
(1st of its kind)

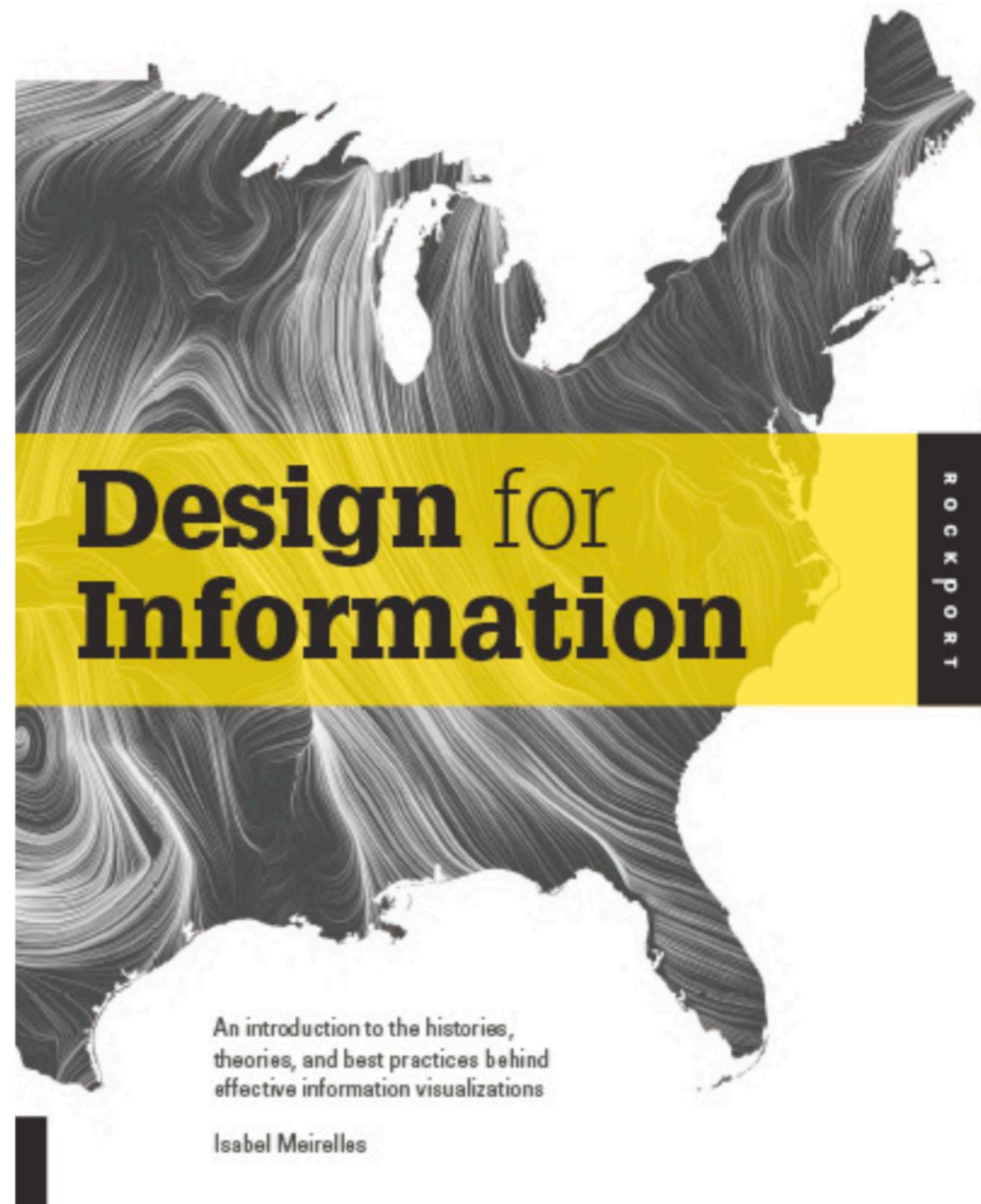
If you can only get one,
THIS is the one!



Munzner, Tamara. Visualization Analysis and Design. CRC Press, 2014.

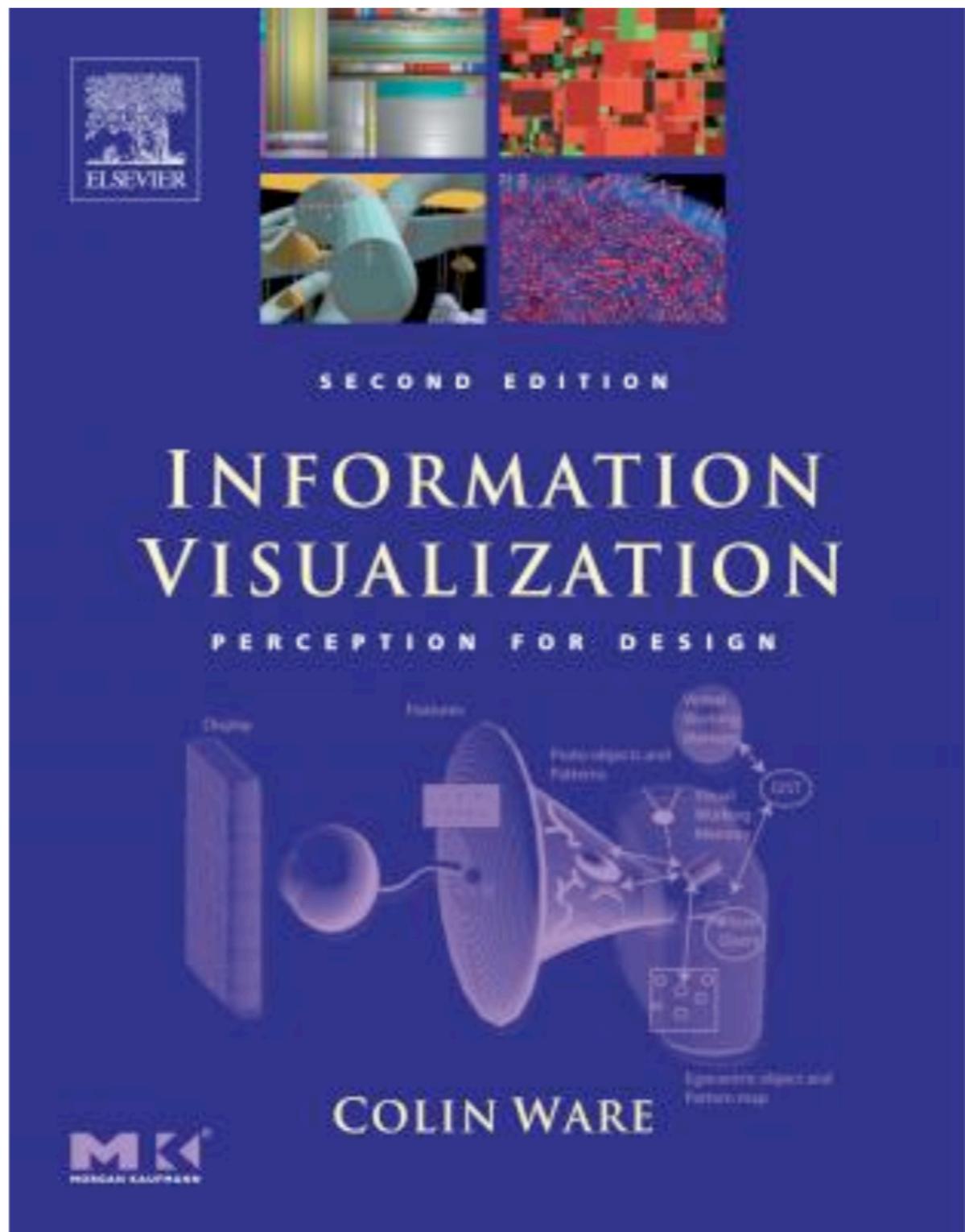
Infovis through examples, history and some free discussion.
Very visual.

Beautiful!



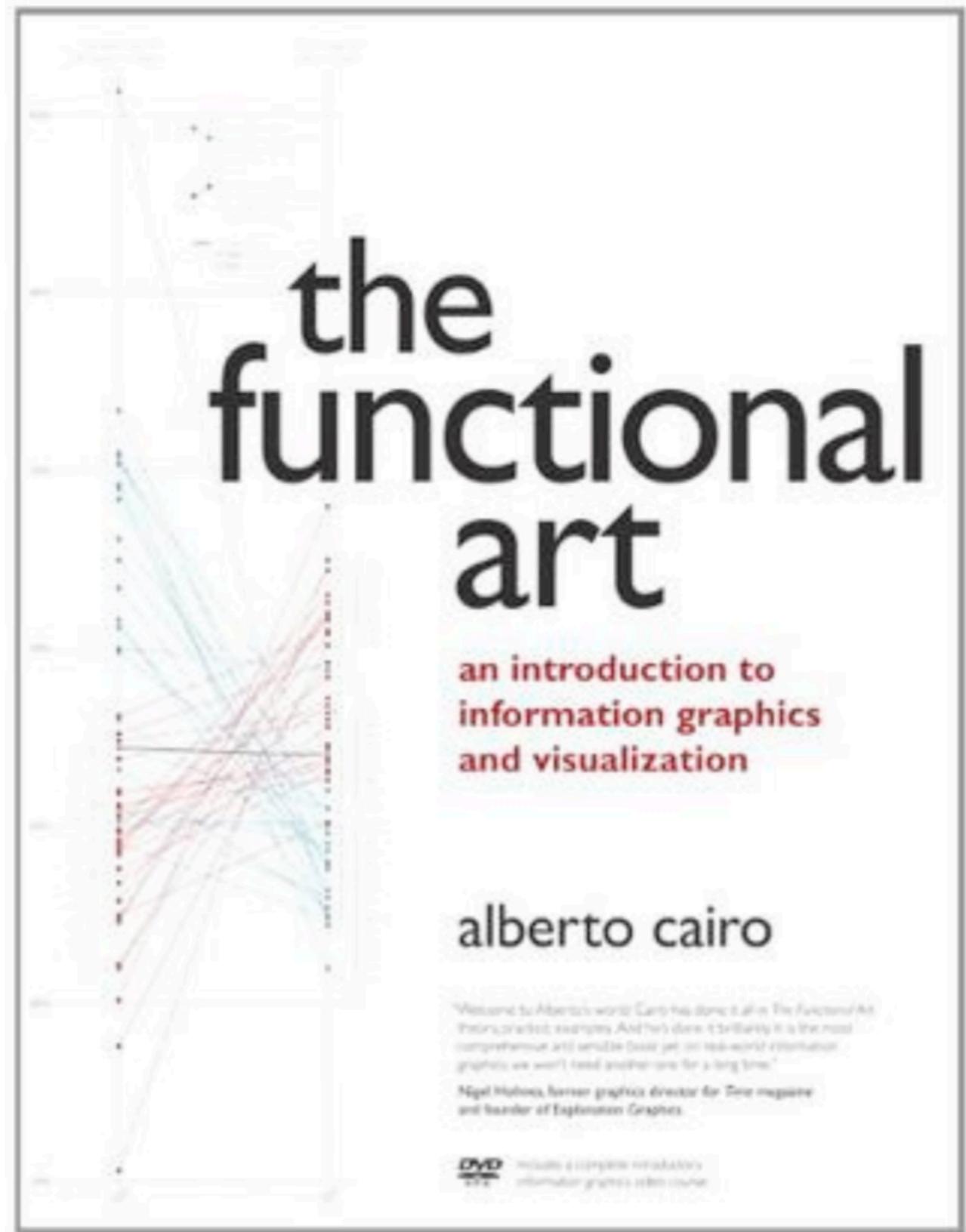
Meirelles, Isabel. Design for Information: An Introduction to the Histories, Theories, and Best Practices Behind Effective Information Visualizations. Rockport publishers, 2013.

All about perception,
and the funny tricks our
brain plays on us.



Ware, Colin. Information visualization: perception for design. Elsevier, 2013.

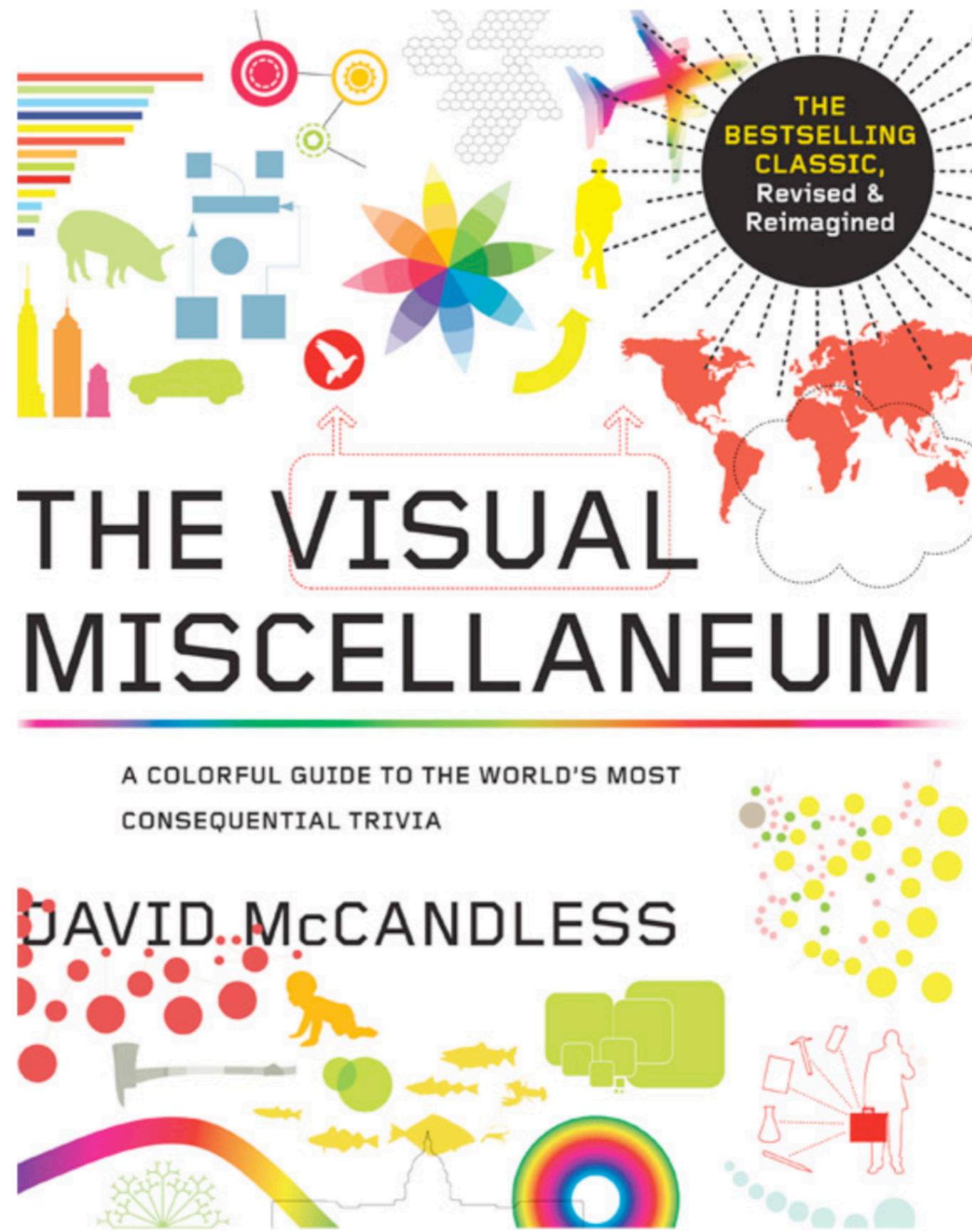
Focus on
infographics



Cairo, Alberto. *The Functional Art: An introduction to information graphics and visualization*. New Riders, 2012.

For inspiration!

Some excellent visualizations! And some so-so...



McCandless, David. *The visual miscellaneum: a colorful guide to the world's most consequential trivia*. Collins Design, 2009.