

Visualization Research and the Human Mind

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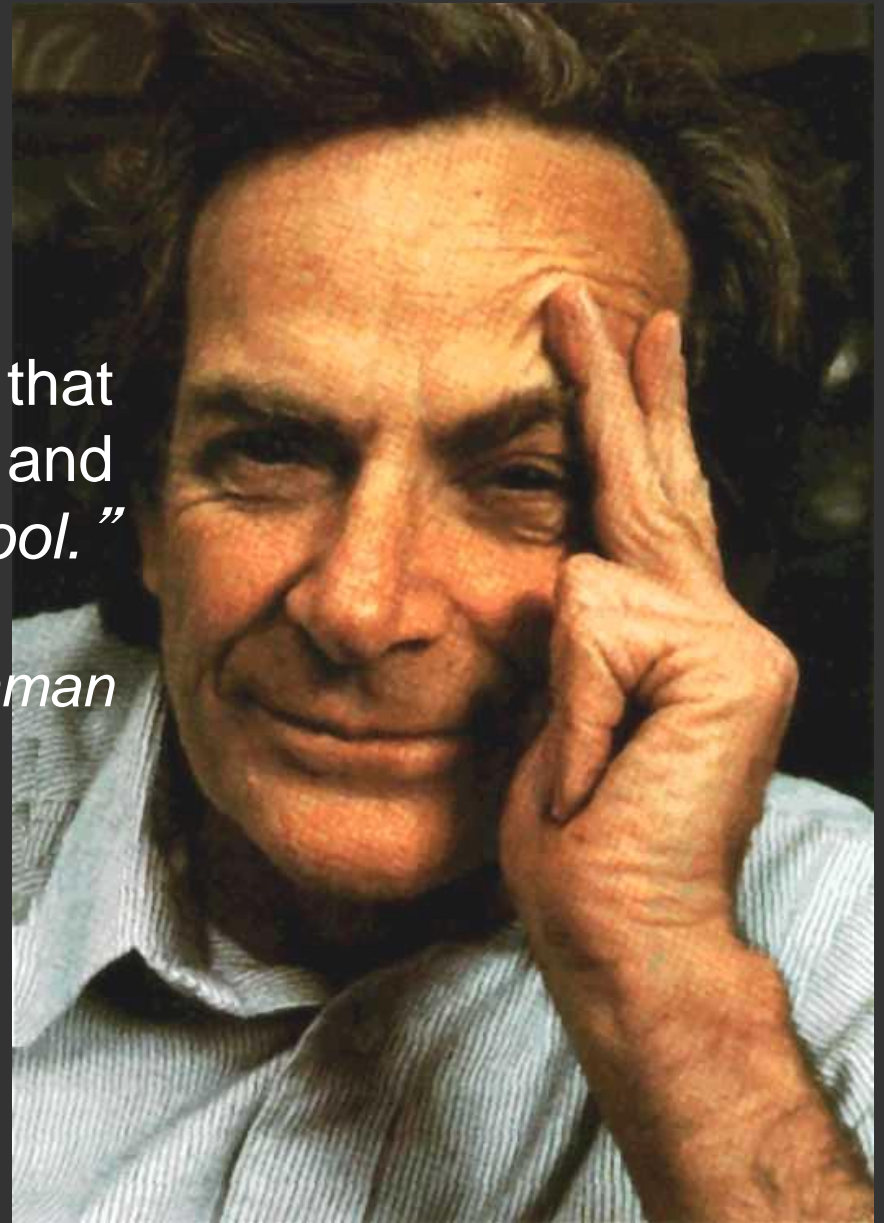
How can empirical research in visualization be further improved?

- *more knowledge of high-level cognition*
- *more knowledge of visual perception*

High-level cognition (i.e., thinking)...

“The first principle is that you must not fool yourself, and *you are the easiest person to fool.*”

–*Richard Feynman*



1. *High-level Cognition: Finding Answers*

Human minds are not that logical (e.g., *Kahneman, 2011*)

- *tend to be “quick and dirty”*
- *typically follow particular heuristics*
- *we don’t notice how prevalent these are*

E.g.: Confirmation bias

Does system/method/theory X work?



But – are you fooling yourself?

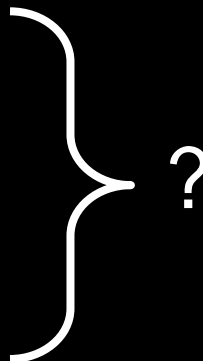
Advocacy?

E.g.: Confirmation bias

~~Does system/method/theory X work?~~

Is system/method/theory X *better than* Y, Z, etc?

System X	Test 1
System Y	Test 1
System Z	Test 1



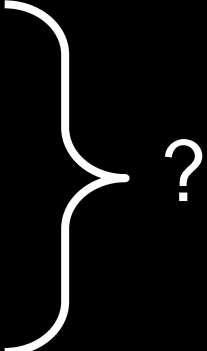
- ~~Compare your data to yourself?~~ *Cherry picking?*
- multiple hypotheses (*competing hypotheses*)

E.g.: Confirmation bias

~~Does system/method/theory X work?~~

Is system/method/theory X **better than** Y, Z, etc?

System X	Test 1	Test 2	...
System Y	Test 1	Test 2	...
System Z	Test 1	Test 2	...



- Obtain **different kinds of evidence**
 - e.g., *control condition: varies only in the key factor; (everything else is held constant)*

High-level Cognition: Finding Answers

Recommendations: *(don't fool yourself!)*

- *be alert to the possibility your belief may be wrong*
- *design tests to distinguish between alternatives*
- *design tests that use multiple sources of evidence*

Look at books / articles on experimental methodology

2. *High-level Cognition: Finding Questions*

Human minds are not that imaginative

- *have trouble dealing with alternatives*
- *have trouble dealing with uncertainty*
- *imagination is often limited to variations of the known*

This can interfere with creating new issues / new designs.

High-level Cognition: Finding Questions

Recommendations: *(don't fool yourself!)*

- *don't immediately reject any new idea / proposal*
 - *learn to be open to new ideas; give them consideration*
- *don't immediately reject strange findings / phenomena*
 - *learn to be open to feedback, esp. negative feedback*
- *try to focus on **differences that make a difference***
 - *try to evolve a feel for what the interesting issues might be*

All of these are skills; can be improved by training

(see e.g. Rensink, IEEE CG&A, Mar-Apr 2015)

What about issues specific to visualization?

Look at vision science...

1. Vision Science: *Mechanisms*

Knowledge of visual perception can help design effective visualizations (e.g., Card, Mackinlay, & Shneiderman, 1999)

Introductory books are okay as a start.

But - our knowledge of how we see is

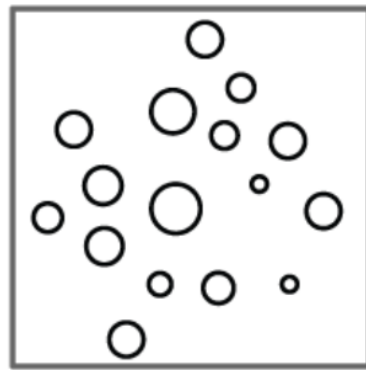
- *increasing at a rapid rate*
- *becoming increasingly counterintuitive*
- *becoming increasingly relevant for visualization*

E.g.: Ensemble coding

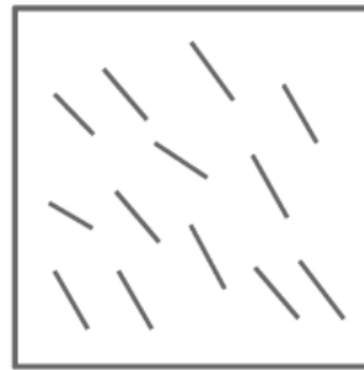
Within **100-150 ms**, humans can perceive average



position



size



orientation



color

(e.g., Safir, Haroz, Gleicher, & Franconeri, 2016)

Vision Science: Mechanisms

Recommendations: Stay in touch with this community

- *look at textbooks that are recent (≤ 10 yrs old)*
- *look at review articles that are recent (≤ 10 yrs old)*
- *talk to your friendly neighborhood vision scientist 😊*

2. Vision Science: *Methodology*

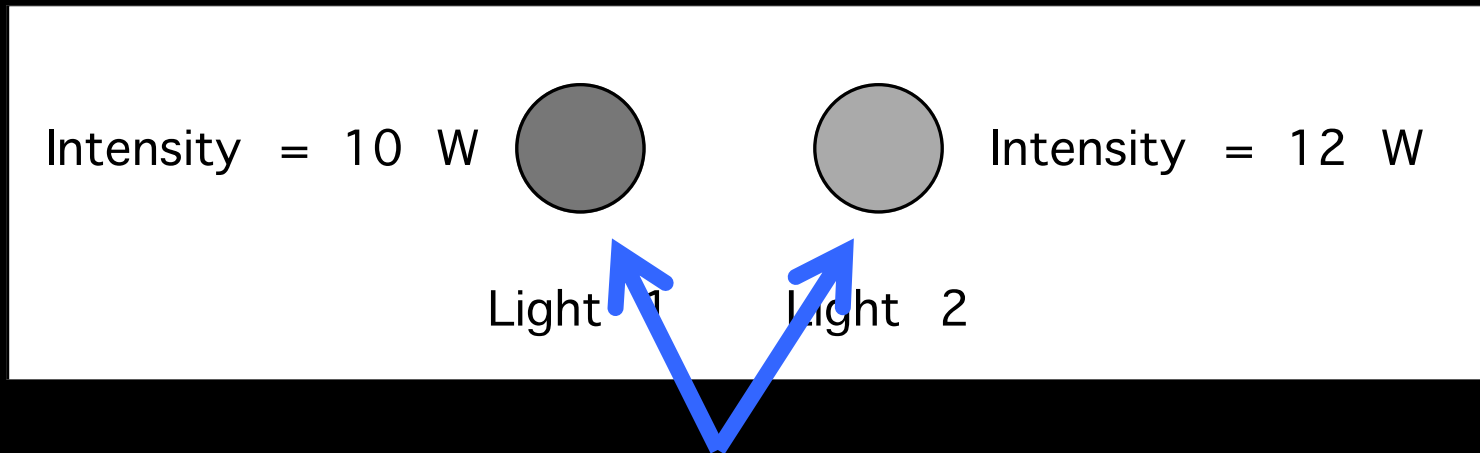
Many possible measures of performance of a design

Consider adapting those developed in vision science

- *highly sensitive assessment of visual performance*
 - *quantities measured are often not obvious*
- *result of long experience in avoiding “booby traps”*
- *can often be adapted to complex situations*

E.g.: Just noticeable difference (JND)

Q: Which **light** has the greater **intensity**?



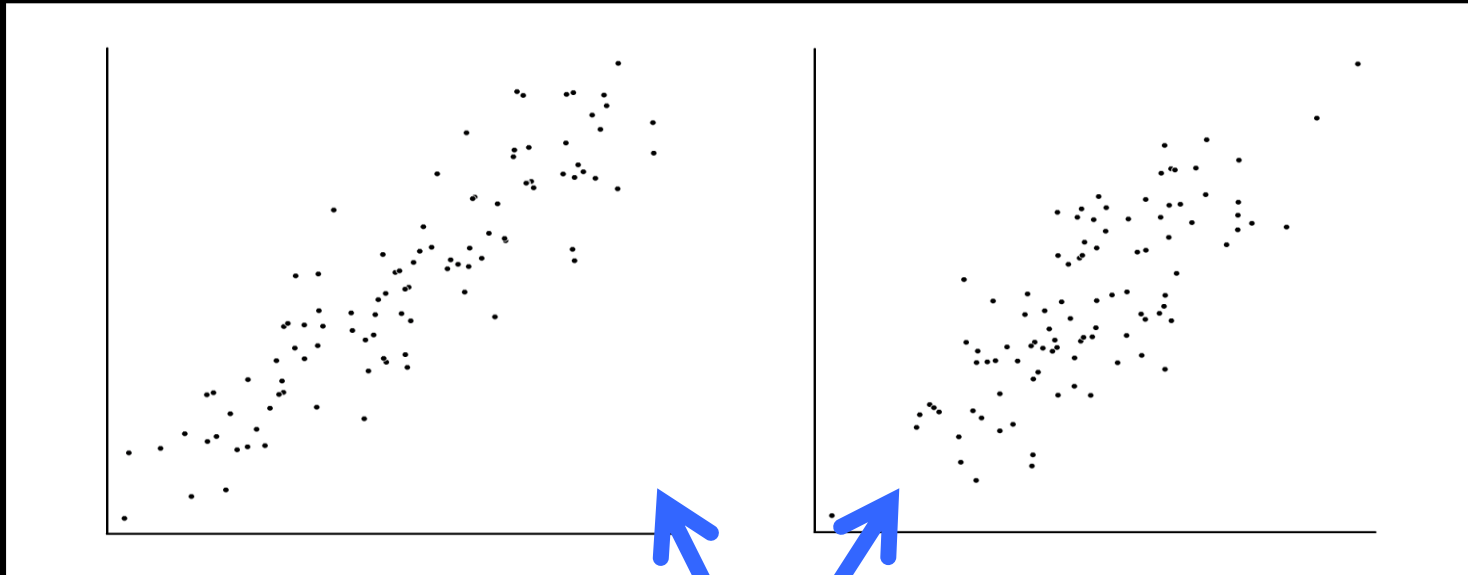
JND = the difference in intensity needed to choose correctly 75% of the time.

A measure of the **precision** (variability) of estimates

$$\text{JND} \propto 1 / \text{SNR} = \text{noise} / \text{signal}$$

E.g.: Just noticeable difference (JND)

Q: Which scatterplot has the greater correlation?



JND(r) = separation (in r) for 75% correct

Results (n=20)

$$\square r = k(1/b - r)$$

k: variability

$$= 0.21 [.17, .24]$$

b: bias

$$= 0.90 [.84, .94]$$

(Let $u = 1 - br$)

$$\square u = ku$$

$$\frac{\square u}{u} = k$$

Weber's Law

$\Delta r = \text{JND}$ (75% correct)

Vision Science: Methodology

Recommendations:

- *try for measures with high sensitivity & robustness*
- *look at handbook chapters in vision science*
- *talk to your friendly neighborhood vision scientist 😊*

3. Vision Science: **Goal**

Goal: Discover **why** a visualization works

1. measure performance under *various conditions*
2. look for *laws* that describe the results
3. look for *mechanisms* to account for these laws / results
4. connect these—if possible—to *vision science*

If we knew the laws or mechanisms, might be possible to

- determine the conditions in which the design will / won't work
- simplify some parts of evaluation
- inspire new, more effective designs

E.g.: Correlation perception in scatterplots

Precision

$$JND(r) = k(1/b_{disc} - r)$$

Accuracy

$$g(r) = \frac{\ln(1 - \frac{b_{est}}{r})}{\ln(1 - b_{est})}$$

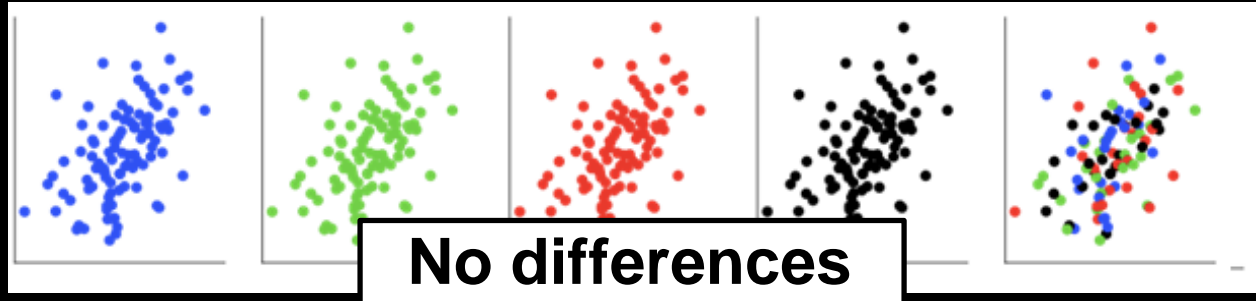
Only two measurements needed to get k, b ;
- describes both precision and accuracy(!)

(Rensink, in press)

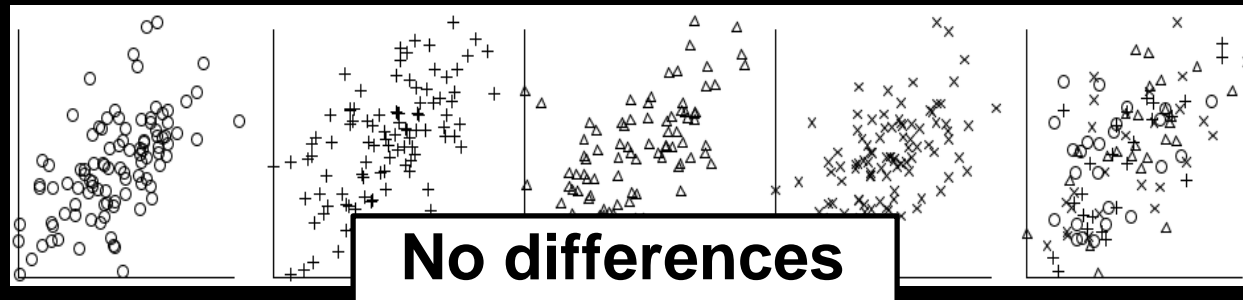
$$b_{disc} = b_{est}$$

(common bias b)

Color



Shape
(symmetric)



Size

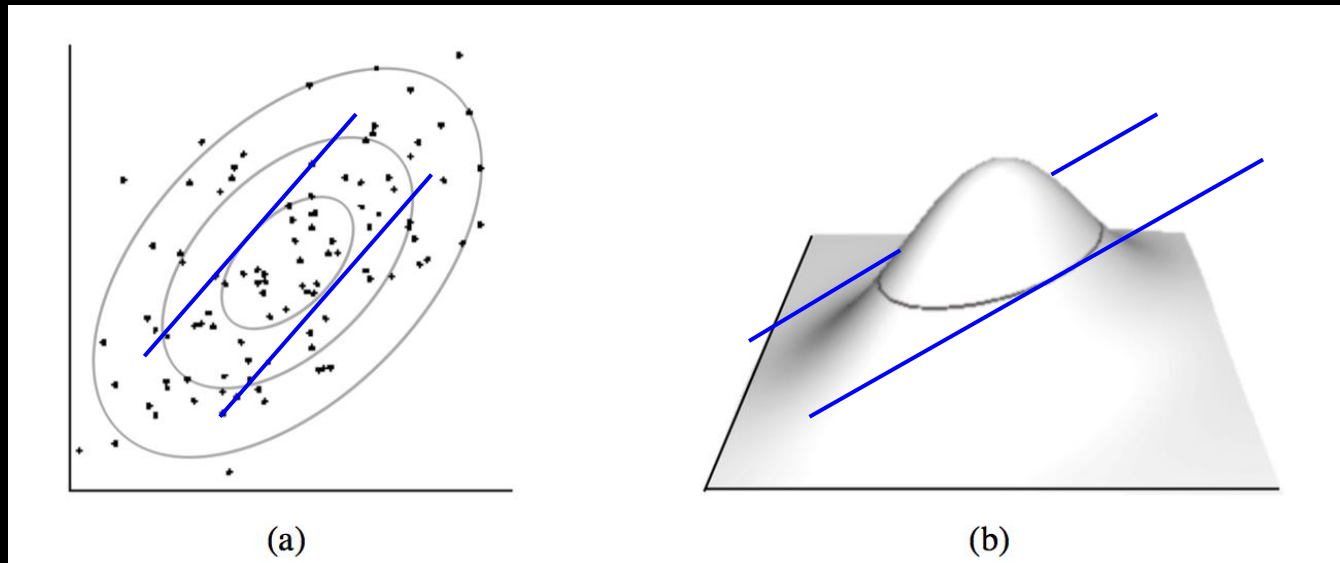


What underlies all this?

Proposal (*Rensink, in press*):

Our visual system perceives **entropy**

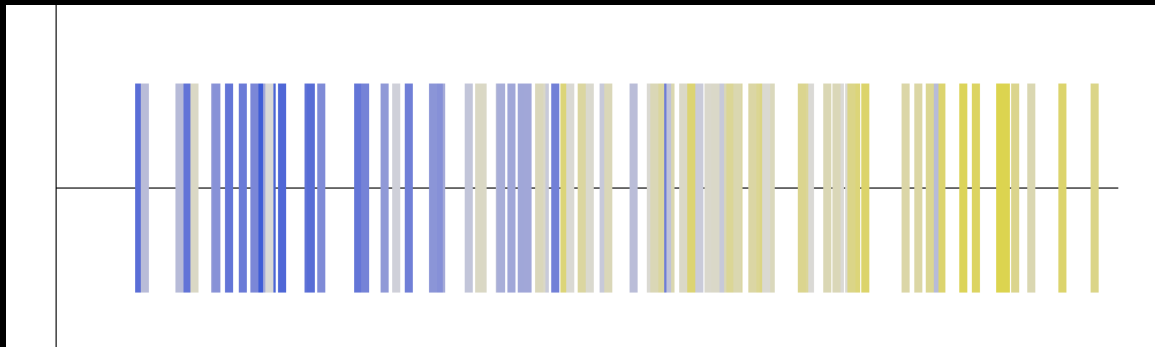
- Visual system infers a **probability distribution** from the dot cloud via **ensemble coding** (at early levels of visual processing)



- Logarithm of **the width of this distribution** approximates entropy
- This approximation is used as a proxy for Pearson correlation

Application: Other kinds of possible visualizations

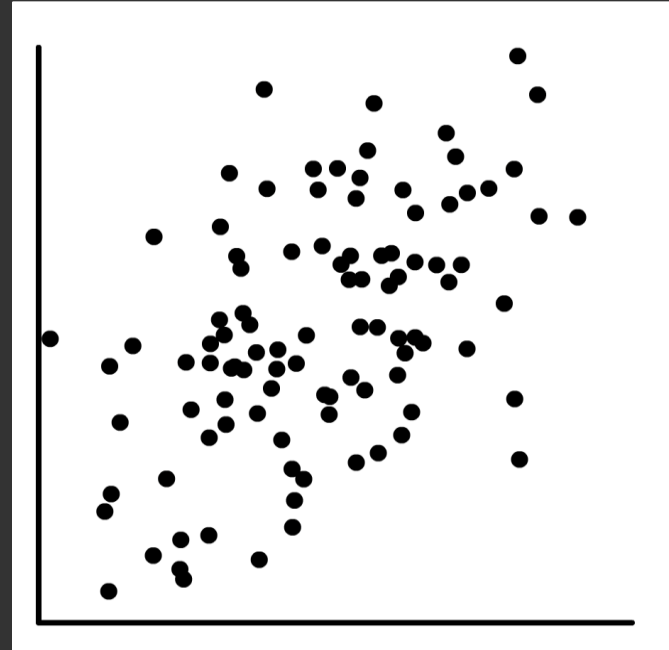
- may work as well as scatterplots



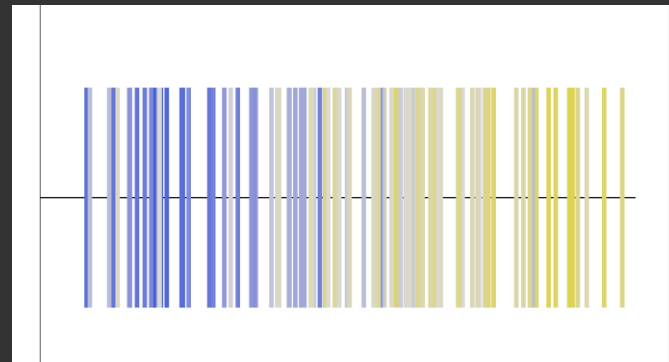
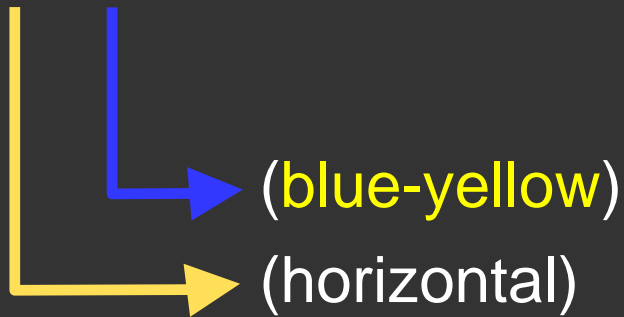
Second dimension is represented by **color** (position on blue-yellow axis)

Nonspatial carriers

$\langle x_1, x_2 \rangle$

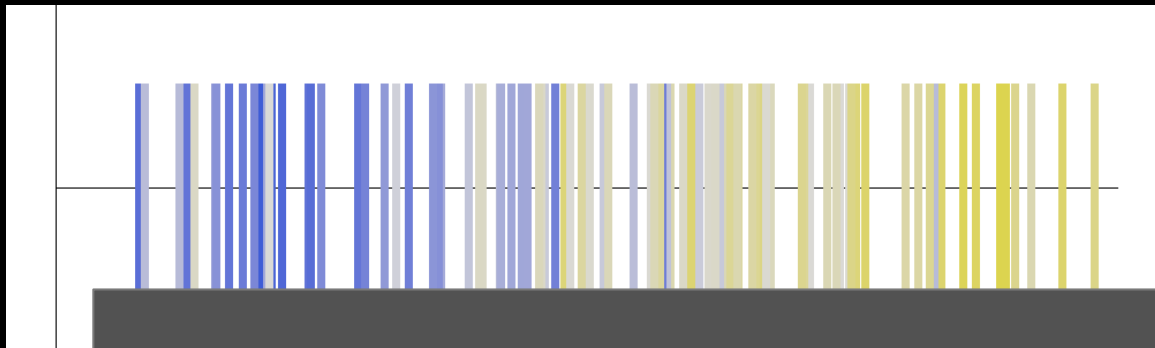


$\langle x_1, x_2 \rangle$



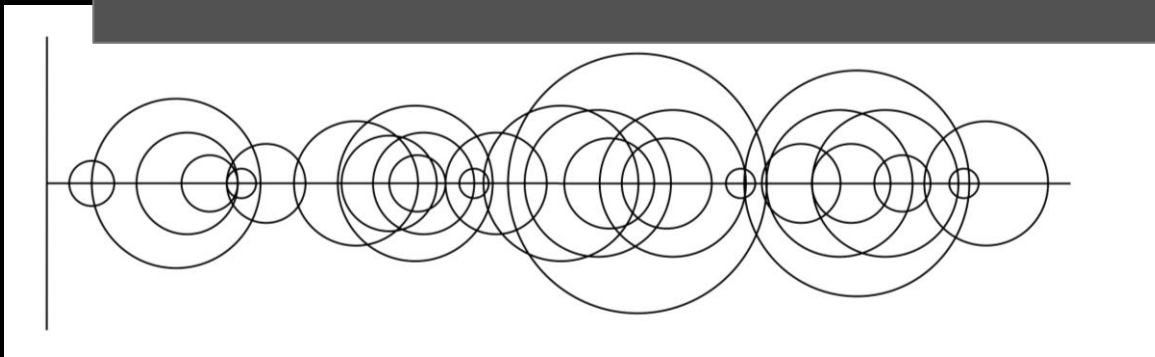
Application: Other kinds of possible visualizations

- may work as well as scatterplots



Second dimension is represented by **color** (position on blue-yellow axis)

Same laws apply! (Rensink, 2014, 2015)



Second dimension is represented by **size** (diameter) of the disk

Vision Science: Goal

Recommendations

- *research into “why” should be a distinct part of vis*
- *begin with **simple versions** of visualizations*
 - *cf . the use of fruit flies in biology (see Rensink, 2014)*
- *focus on **particular aspects** of these*

Possible development: **a science of visualization**

- *at least for some parts of vis (“islands”)*
- *may simplify some parts of evaluation (?)*
- *may inspire new designs(?)*

