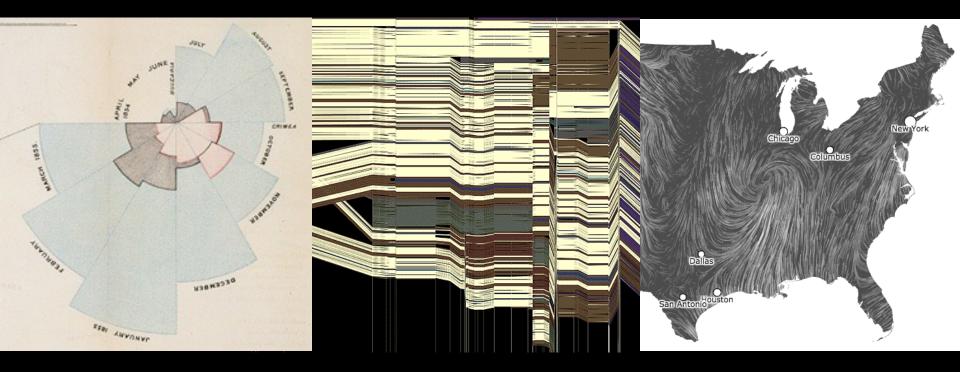
# CSE 442 - Data Visualization Using Space Effectively



Jeffrey Heer University of Washington

### Effectiveness Rankings [Mackinlay 86]

#### QUANTITATIVE

Position Length Angle Slope Area (Size) Volume Density (Value) Color Sat Color Hue Texture Connection Containment Shape

#### ORDINAL

Position Density (Value) Color Sat Color Hue Texture Connection Containment Length Angle Slope Area (Size) Volume Shape

NOMINAL Position Color Hue Texture Connection Containment Density (Value) Color Sat Shape Length Angle Slope Area Volume

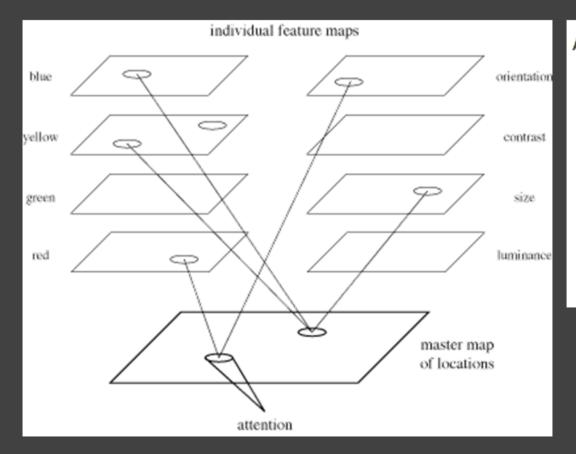
### Effectiveness Rankings [Mackinlay 86]

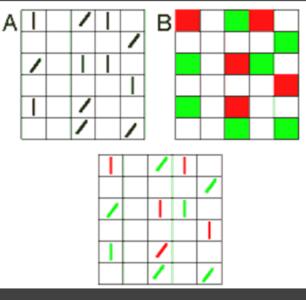
QUANTITATIVE Position · · · · · · Position · · · · · Position Length Angle Slope Area (Size) Volume Density (Value) Color Sat Color Hue Texture Connection Containment Shape

ORDINAL Density (Value) Color Sat Color Hue Texture Connection Containment Length Angle Slope Area (Size) Volume Shape

NOMINAL Color Hue Texture Connection Containment Density (Value) Color Sat Shape Length Angle Slope Area Volume

#### **Feature Integration Theory**





Feature maps for orientation & color [Green]

Treisman's feature integration model [Healey 04]

Indexed by Position!

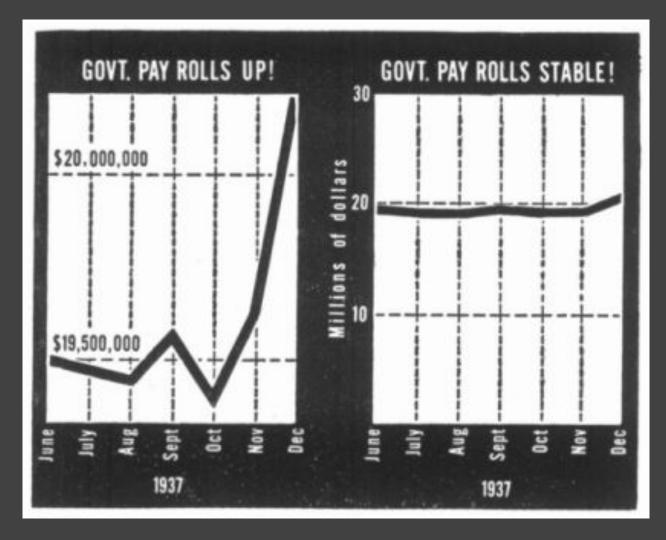
Space (x, y) is the most important encoding channel. **But are you in the** *right space?* 

# Topics

Scales & Axes Data Space, Model Space Optimizing Chart Design Zooming Focus + Context Dimensionality Reduction

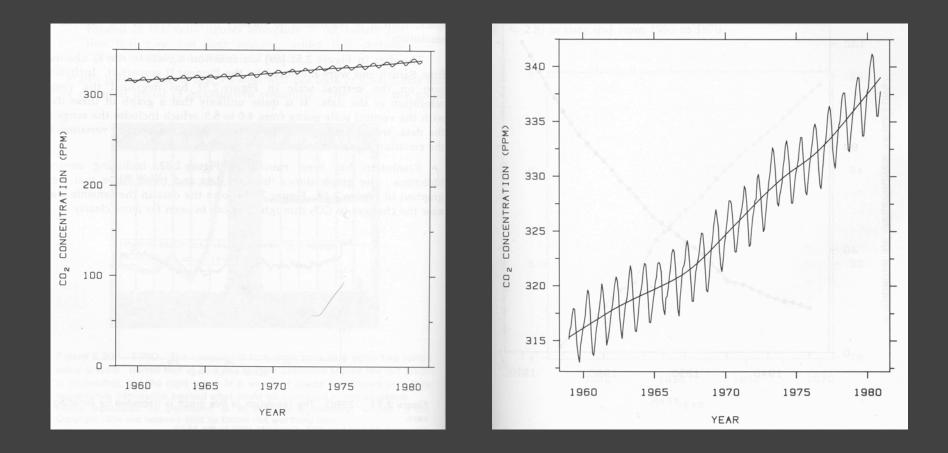
# Scales & Axes

#### Include Zero in Axis Scale?



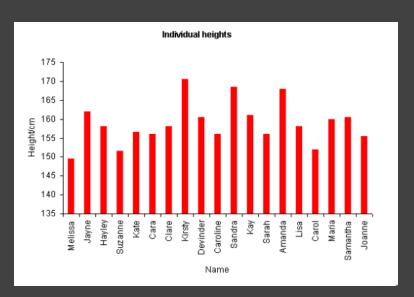
Government payrolls in 1937 [How To Lie With Statistics. Huff]

#### Include Zero in Axis Scale?



Yearly CO<sub>2</sub> concentrations [Cleveland 85]

#### Include Zero in Axis Scale?



Kristy Sandra Amanda Jayne Compare Kay Devinder Samantha Proportions Maria Havley (Q-Ratio) Clare Lisa Kate Cara

Caroline

Sarah Joanne

Carol Suzanne Melissa

Compare

Relative

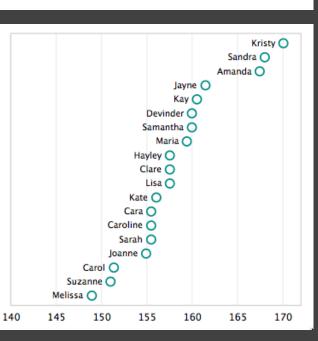
Position

(Q-Interval)

0

20

40 60 80

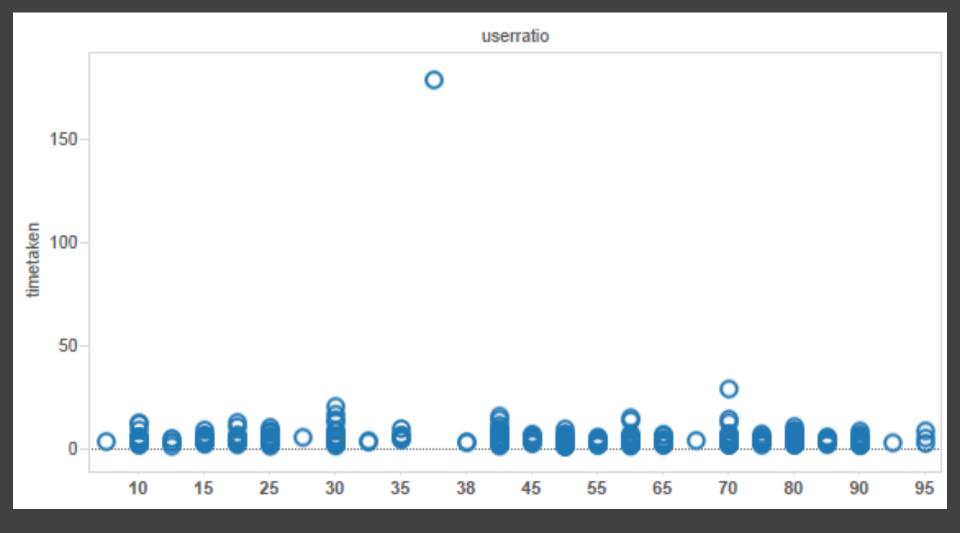


100 120 140

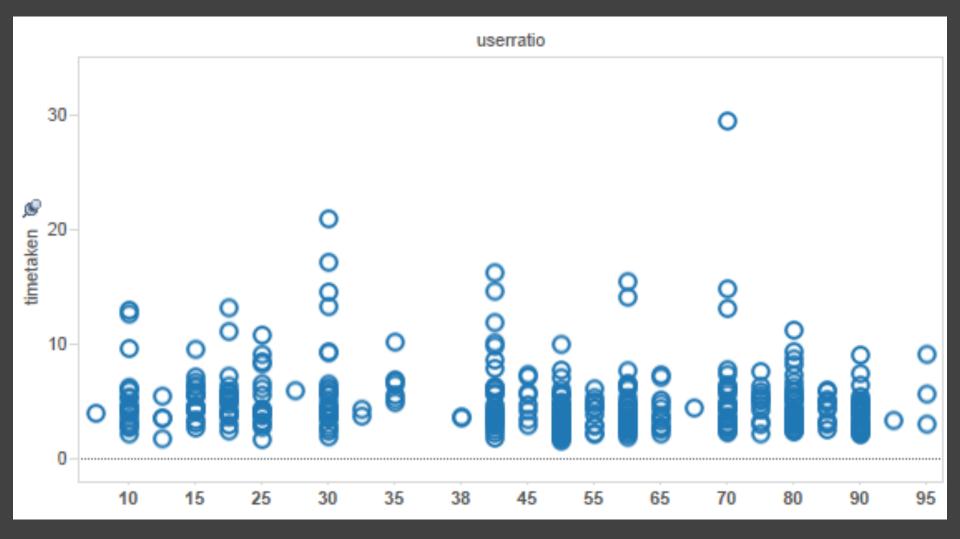
160 180

Violates Expressiveness Principle!

#### How to Scale the Axis?

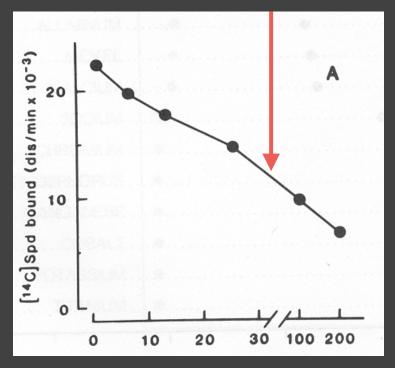


# **One Option: Clip Outliers**



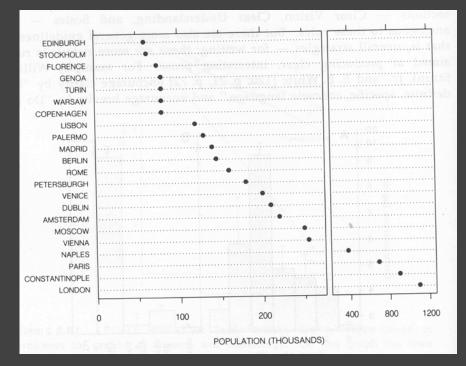
## **Clearly Mark Scale Breaks**

#### Violates Expressiveness Principle!

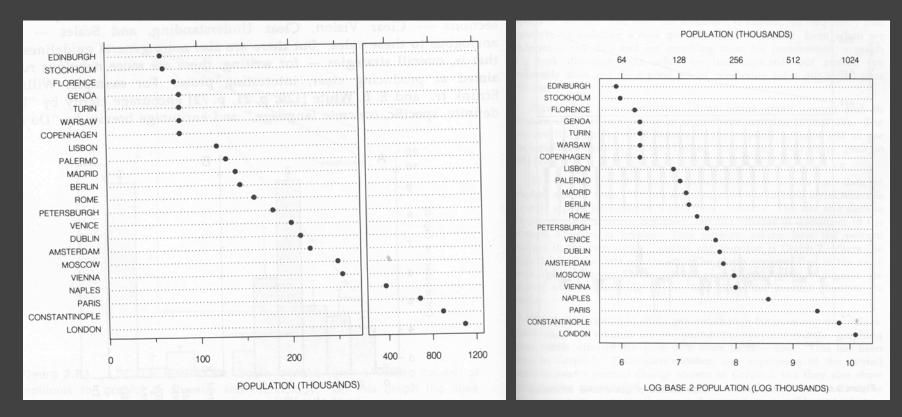




Well-marked scale break [Cleveland 85]



#### Scale Break vs. Log Scale

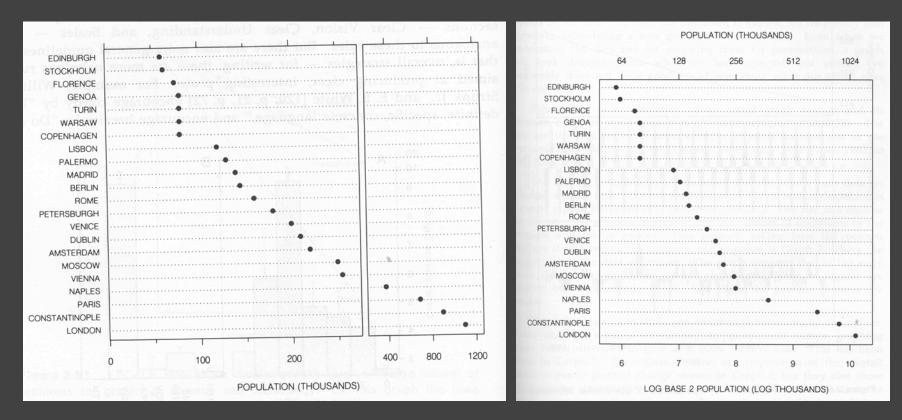


#### Scale Break

#### Log Scale

[Cleveland 85]

#### Scale Break vs. Log Scale

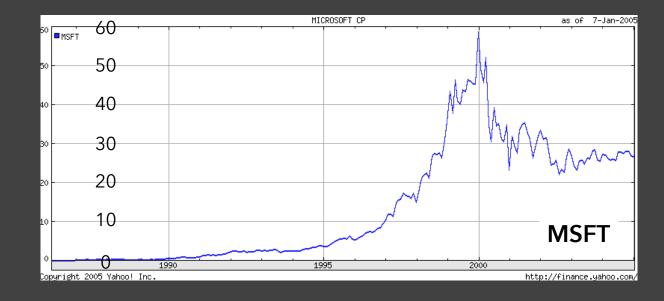


#### Both increase visual resolution

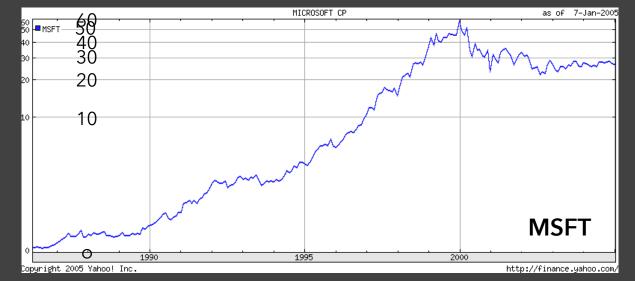
Scale break: difficult to compare (*cognitive* – not *perceptual* – work) Log scale: direct comparison of all data

#### Linear Scale vs. Log Scale

#### Linear Scale

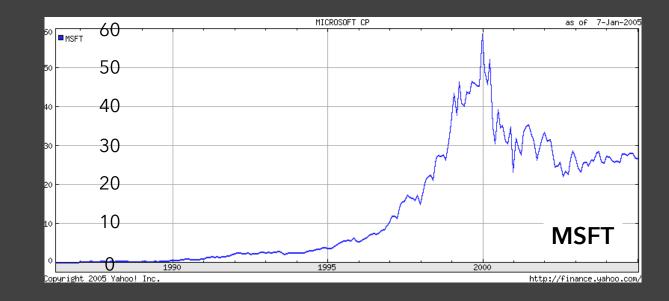


#### Log Scale

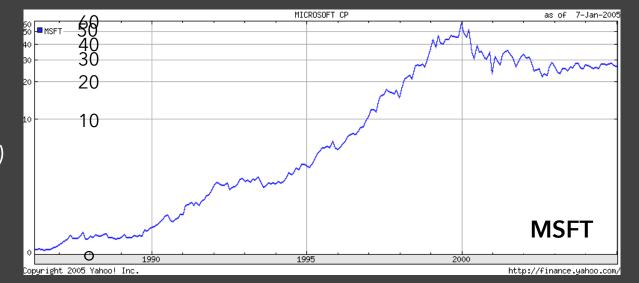


#### Linear Scale vs. Log Scale

Linear Scale Absolute change



**Log Scale** Small fluctuations Percent change d(10,20) = d(30,60)



# When To Apply a Log Scale?

**Address data skew** (e.g., dominant outliers) Enables comparison within and across multiple orders of magnitude.

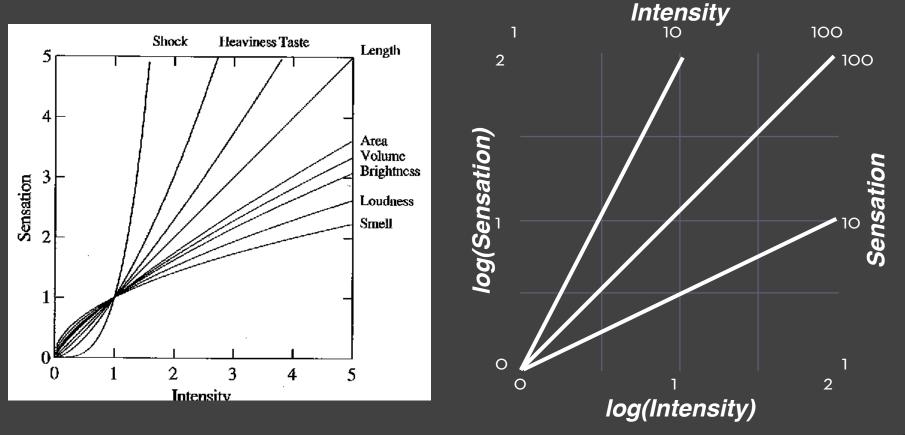
**Focus on multiplicative factors** (not additive) Recall that the logarithm transforms **×** to **+**! Percentage change, not absolute value.

Constraint: **positive, non-zero values** Constraint: **audience familiarity?** 

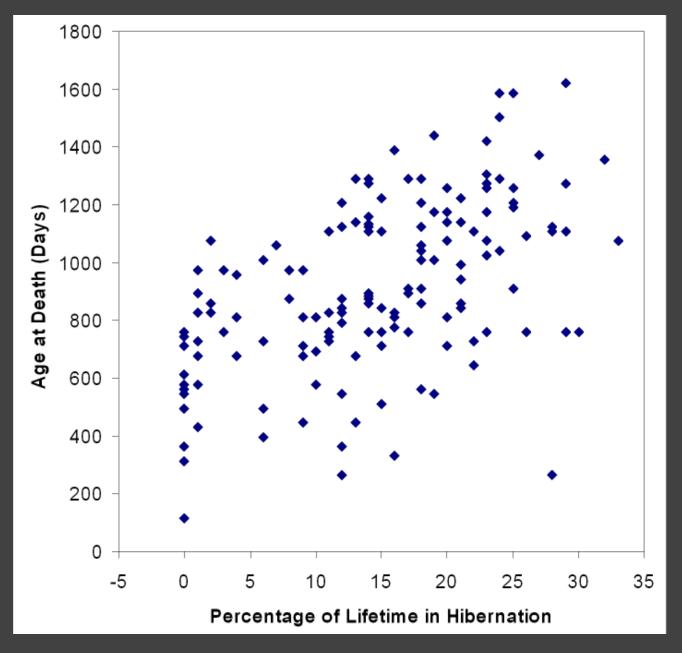
# Log-Log Plot

Power functions (**y = kx**<sup>a</sup>) transform into lines Example - Steven's Power Law:

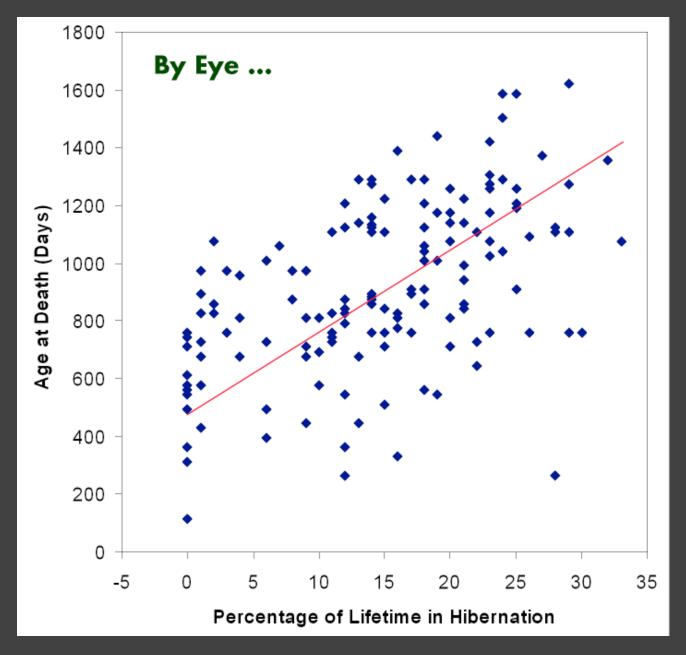
 $S = kI^p \rightarrow \log S = \log k + p \log I$ 



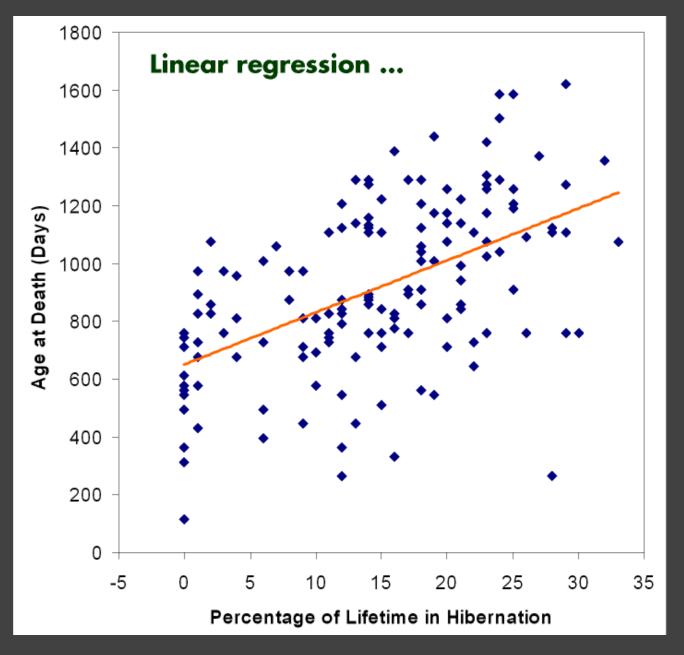
Data Space, Model Space



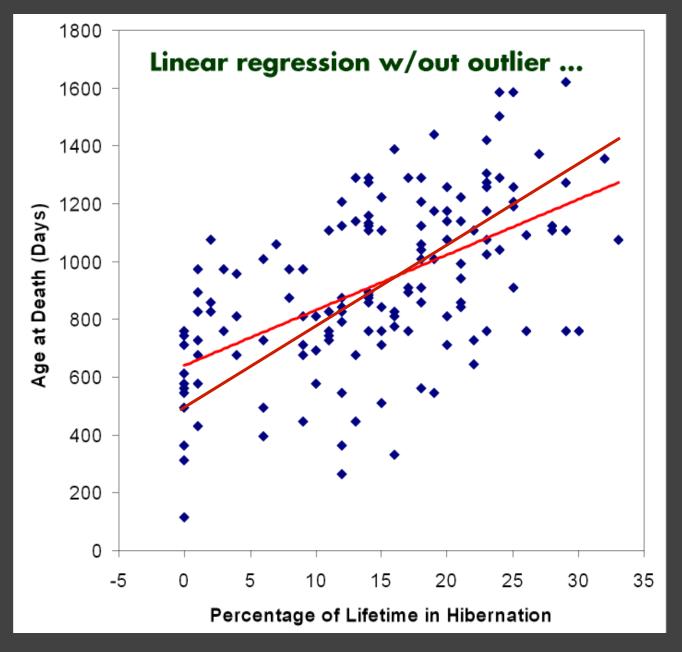
[The Elements of Graphing Data. Cleveland 94]



[The Elements of Graphing Data. Cleveland 94]



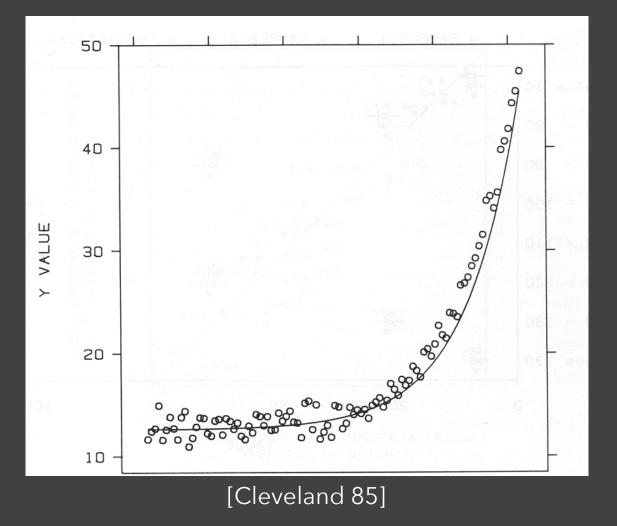
[The Elements of Graphing Data. Cleveland 94]



[The Elements of Graphing Data. Cleveland 94]

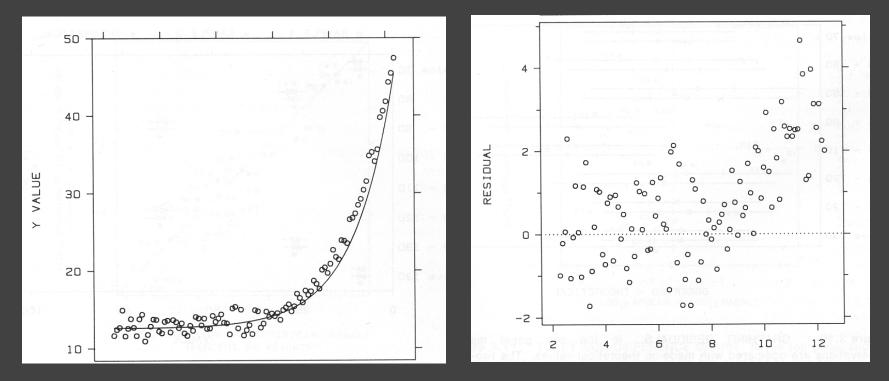
### **Transforming Data**

#### How well does the curve fit the data?



### **Plot the Residuals**

Plot vertical distance from best fit curve Residual graph shows accuracy of fit

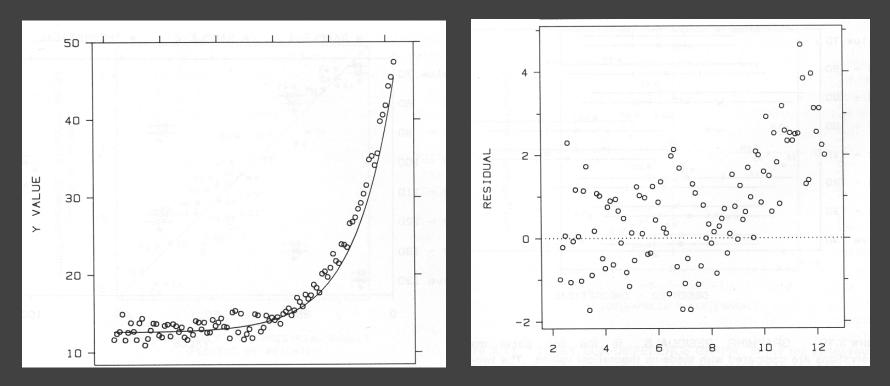


[Cleveland 85]

# **Multiple Plotting Options**

Plot model in data space

#### Plot data in model space



[Cleveland 85]

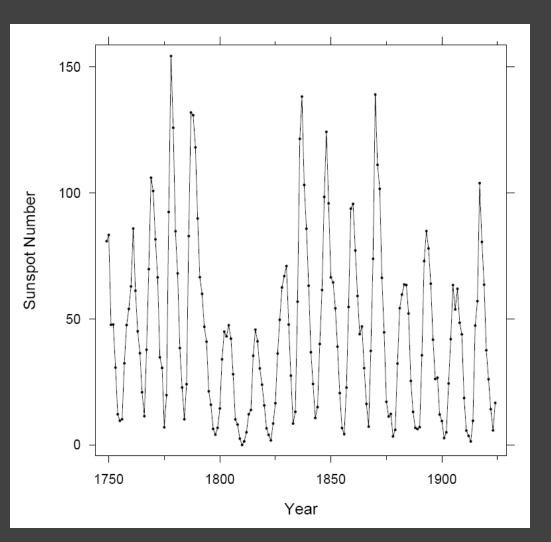
# **Optimizing Chart Design**

#### **Chart Design Parameters**

Given a visual encoding (e.g., line chart), what aspects might affect graphical perception?

Physical Size Aspect Ratio Ticks, Labels, Gridlines Line Width Data Points (e.g., dots)

How might we determine optimized choices?



#### **Optimization-Based Design**

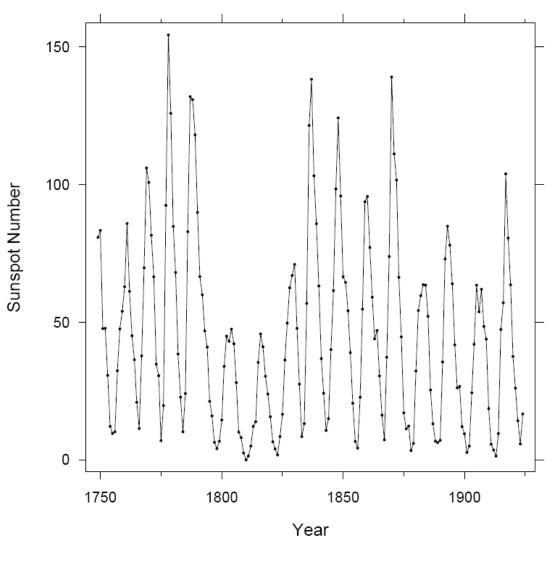
Determine *error* or *energy* functions for measuring the "quality" of a visualization.

Treat as an optimization *objective* and then *solve* (or *search*) for better chart parameters.

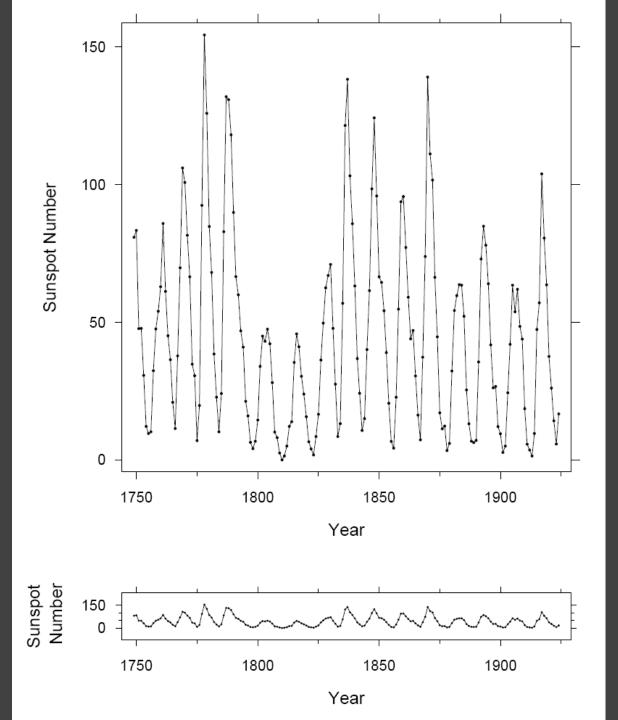
#### **Examples:**

Determining chart aspect ratio Selecting axis ticks Streamgraph layout

# Aspect Ratio



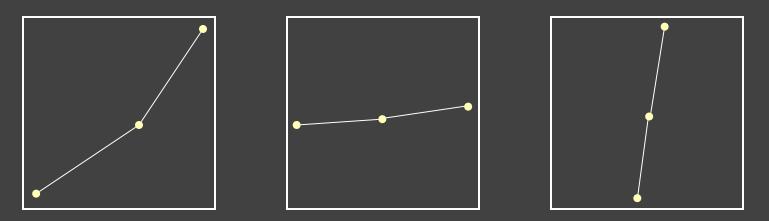
William S. Cleveland The Elements of Graphing Data



William S. Cleveland The Elements of <u>Grap</u>hing Data

## Banking to 45° [Cleveland]

To facilitate perception of trends, maximize the discriminability of line segment orientations



Two line segments are maximally discriminable when their average absolute angle is 45° Insight: to optimize the aspect ratio, bank to 45°

## **Aspect Ratio Banking Methods**

#### Median-Absolute-Slope $\alpha = \text{median} | s_i | R_x / R_y$

#### Avg-Absolute-Orientation Unweighted

$$\sum_{i} \frac{|\theta_i(\alpha)|}{n} = 45^{\circ}$$

Weighted

$$\frac{\sum_{i} |\theta_{i}(\alpha)| l_{i}(\alpha)}{\sum_{i} l_{i}(\alpha)} = 45^{\circ}$$

Average-Absolute-Slope  $\alpha = \text{mean} | s_i | R_x / R_y$ 

Max-Orientation-Resolution Global (over all i, j s.t. i≠j)

$$\sum_{i}\sum_{j}|\theta_{i}(\alpha)-\theta_{j}(\alpha)|^{2}$$

Local (over adjacent segments)

$$\sum_{i} |\theta_{i}(\alpha) - \theta_{i+1}(\alpha)|^{2}$$

#### **Closed Form Solutions**

#### Median-Absolute-Slope $\alpha = \text{median} | s_i | R_x / R_y$

Avg-Absolute-Orientation Unweighted

$$\sum_{i} \frac{|\theta_i(\alpha)|}{n} = 45^{\circ}$$

Weighted

$$\frac{\sum_{i} |\theta_{i}(\alpha)| l_{i}(\alpha)}{\sum_{i} l_{i}(\alpha)} = 45^{\circ}$$

Average-Absolute-Slope  $\alpha = \text{mean} | s_i | R_x / R_y$ 

Max-Orientation-Resolution Global (over all i, j s.t. i≠j)

$$\sum_{i}\sum_{j}|\theta_{i}(\alpha)-\theta_{j}(\alpha)|^{2}$$

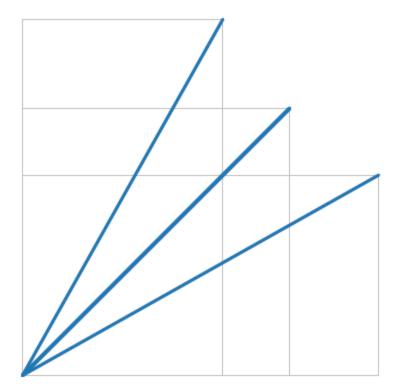
Local (over adjacent segments)

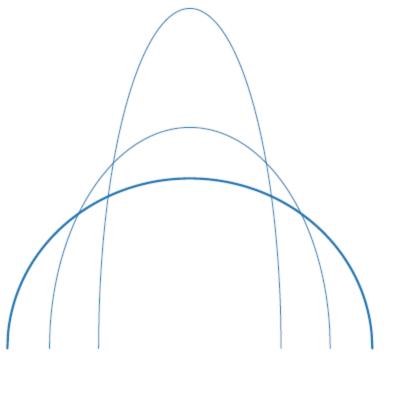
$$\sum_{i} |\theta_{i}(\alpha) - \theta_{i+1}(\alpha)|^{2}$$

**Requires Iterative Optimization** 

## **Alternative: Minimize Arc Length**

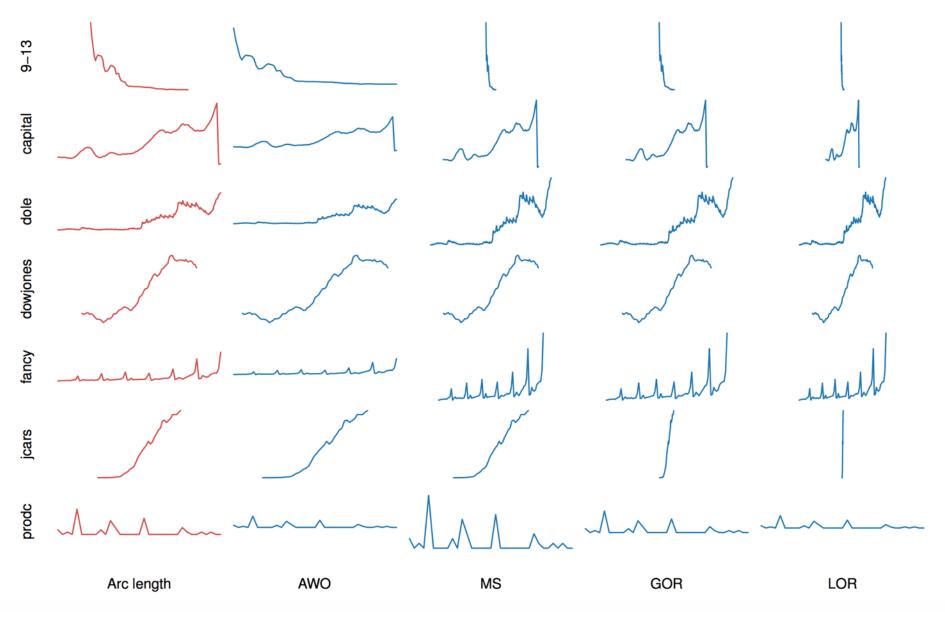
while holding area constant [Talbot et al. 2011]





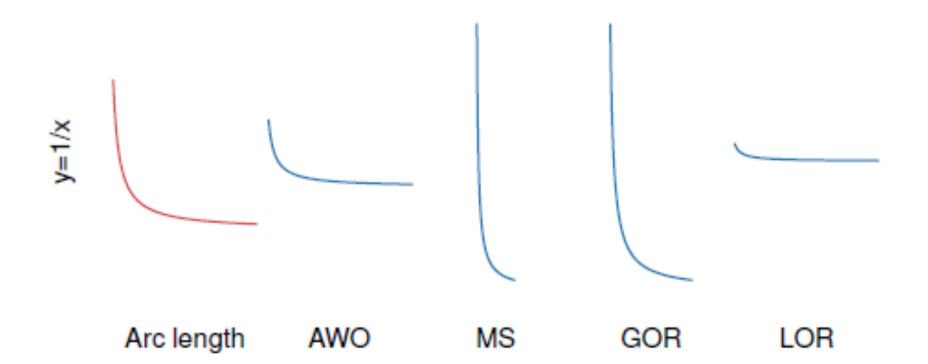
Straight line -> 45°

Ellipse -> Circle



[Talbot et al. 2011]

### Robustness: Banking y = 1/x

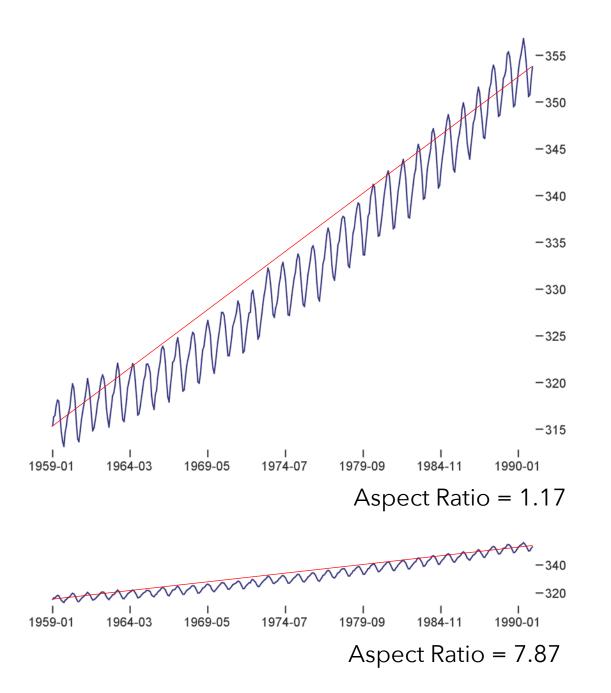


[Talbot et al. 2011]

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9-4	0 +					
lynx	0		<b>O</b>			
computer	0 +					
bankdata	0 +		<b>`</b>			
elec	0 +					
wagesuk	0 +					
9–17b	0 +					
schizo		00				
9-9	0 +					
labour	00					
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### A Good Compromise Arc-length banking produces aspect ratios in-between those produced by other methods.

[Talbot et al. 2011]



### Trends may occur at different scales!

Apply banking to the original data *or* to fitted trend lines. [Heer & Agrawala '06]

### CO<sub>2</sub> Measurements

William S. Cleveland *Visualizing Data* 

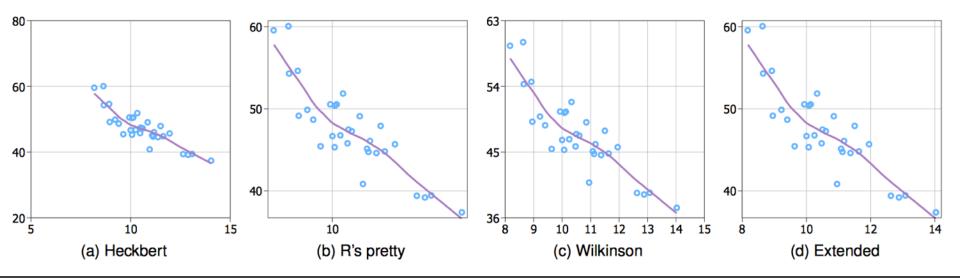
### Discussion

Arc-length banking preferable to prior methods Parameterization invariant Robust (corner cases, compromise results) Applicable to plotted curves and contour lines Fast-converging iterative optimization

**But what about perceptual effectiveness?** We lack theory to motivate aspect ratio selection. Perceptual experiments needed to assess?

# Axis Ticks

### **Tick Mark Selection**



### What are some properties of "good" tick marks?

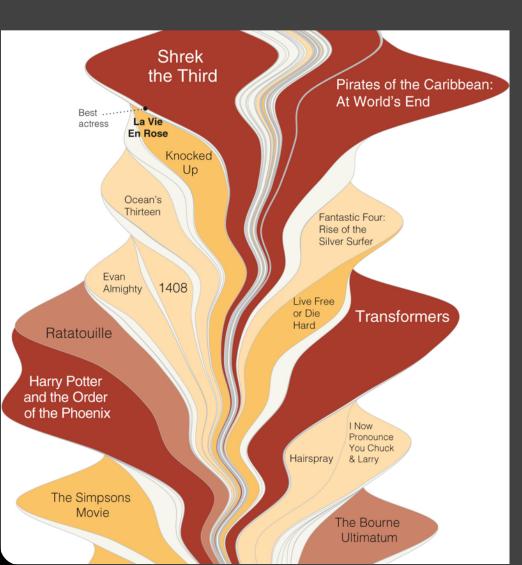
## Tick Mark Criteria

Simplicity - numbers are multiples of 10, 5, 2
Coverage - ticks near the ends of the data
Density - not too many, nor too few
Legibility - whitespace, horizontal text, size

### Optimization

Talbot et al '10 use a search procedure that optimizes criteria in turn (e.g., find simple numbers first, then adjust coverage, etc.). *S* = 0.2 simplicity + 0.25 coverage + 0.5 density + 0.05 legibility Streamgraph Layout

### Streamgraph Layout [Byron & Wattenberg '08]



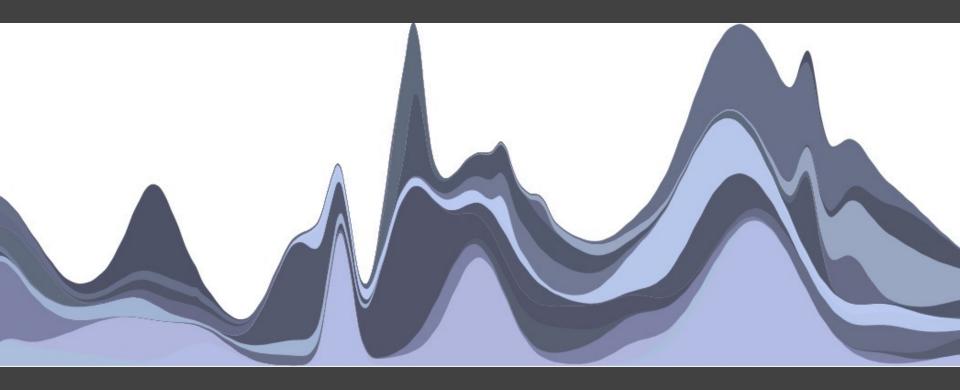
Thickness encodes values, stack extent encodes sum of values.

### Issues

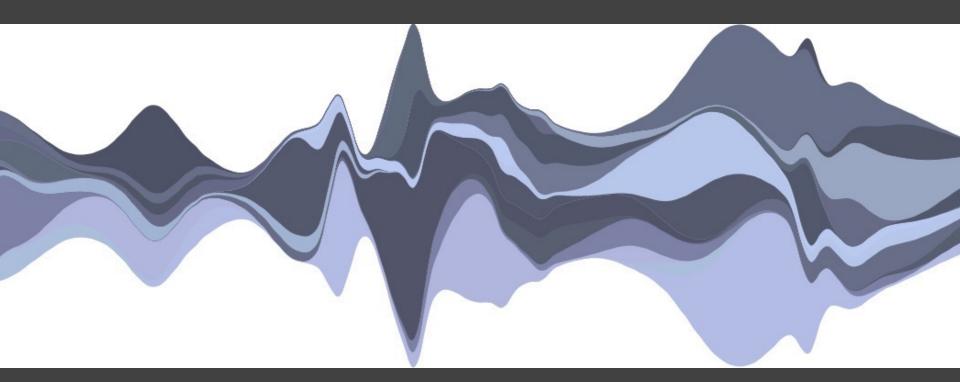
Layers bias each other Perceived thickness Slope (bank to 45?)

**Design Parameters** Layer baseline Layer order

### Zero Baseline

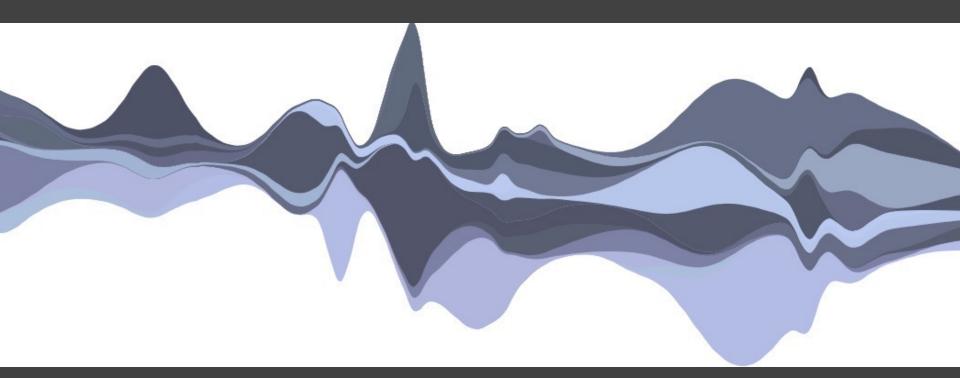


### **Centered Baseline**



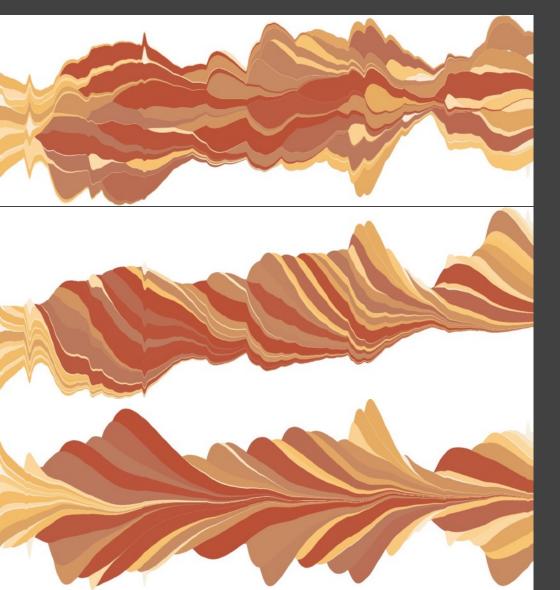
**Pros**: reduces "spikiness", minimizes slope (in the least squares sense) for bottom and top layers. **Cons**: removes fixed baseline for sum comparison.

### "Wiggle" Baseline [Byron & Wattenberg '08]



**Pros**: minimizes *overall* slope (in the least squares sense), weighted by layer thickness. **Cons**: chart may require more total area.

### Layer Sorting



Unsorted

#### Sort by First Non-Zero Value

**Sort "Inside-Out"**: Place layers alternately at top or bottom, in a thickness balancing manner.

### Streamgraphs Redux [Di Bartolomeo & Hu '16]

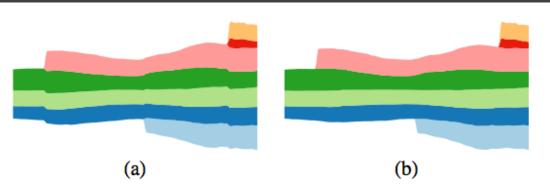


Figure 2: Different baselines: (a) with 2-norm, distortions are present; (b) with 1-norm, all layers are smooth.

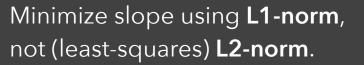
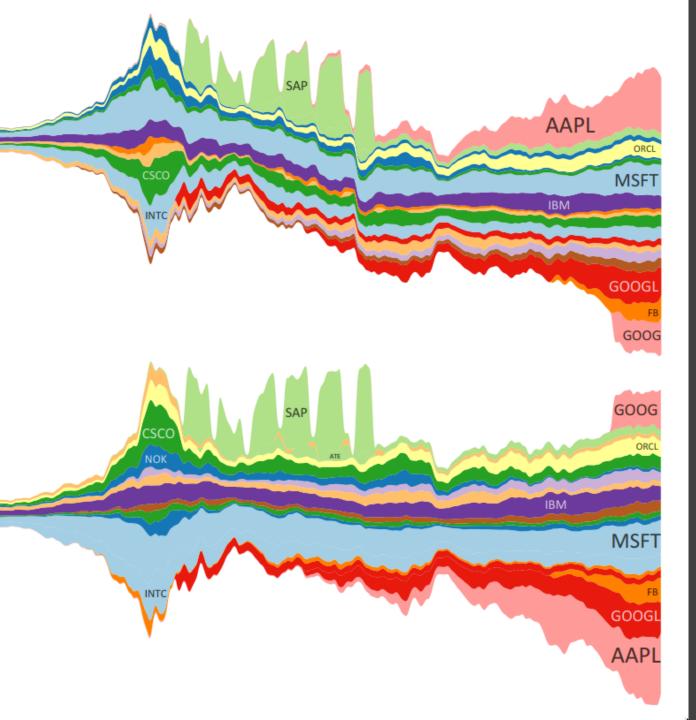




Figure 4: Two orderings of two layers on a flat baseline. Case (b) produces a distortion, case (a) is preferred.

Perform **iterative optimization**, swapping adjacent layers if it improves the overall layout.



### Results

Byron & Wattenberg '08 (via D3's stacked layout)

Di Bartolomeo & Hu '16

# Administrivia

### **FP: Interactive Prototype**

Create an interactive visualization. Choose a driving question in your topic area and develop an appropriate visualization + interaction techniques, then deploy your visualization on the web. Due by *5pm* on **Wednesday, May 3**.

We will assign your GitHub repos shortly!



### Requirements

**Interactive.** You must implement interaction methods! However, this is not only selection / filtering / tooltips. Also consider annotations or other narrative features to draw attention and provide additional context

**Web-based.** D3 is encouraged, but not required. Deploy your visualization using GitHub pages.

Write-up. Provide design rationale on your web page.



## Interactive Prototype Tips

**Start now.** It will take longer than you think.

**Keep it simple.** Choose a minimal set of interactions that enables users to explore and generate interesting insights. Do not feel obligated to convey *everything* about the data: focus on a compelling subset.

**Promote engagement**. How do your chosen interactions reveal interesting observations?



### **Team Member Roles**

We encourage you to structure team responsibilities! **Coordinator**: Organize meetings, track deadlines, *etc.* **Data Lead**: Data wrangling, management, distillation **Tech Lead**: Manage code integration, GitHub repo **UX Lead**: Visualization/interaction design & evaluation *One may have multiple roles, share work across roles...* 



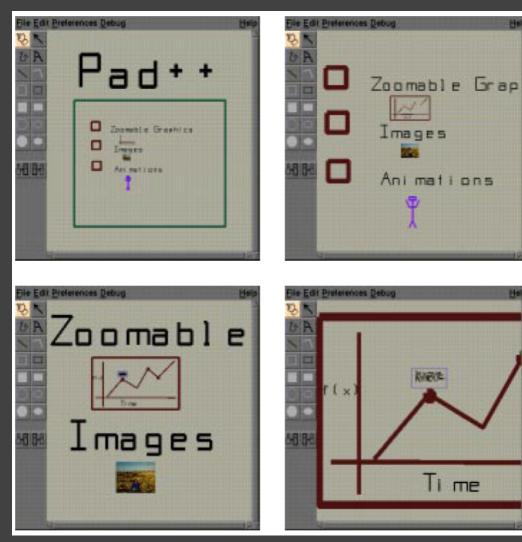


## Zooming



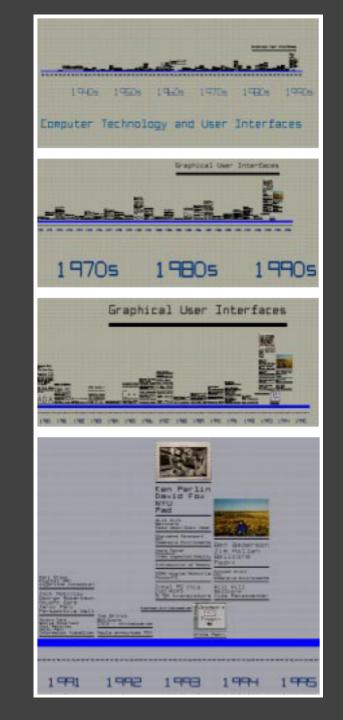
#### Eames' Powers of Ten

## Interactive Zooming



### Pad++ [Bederson and Hollan 94]

Ti me



## **Semantic Zooming**

### Change representations as zoom level changes

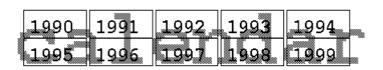
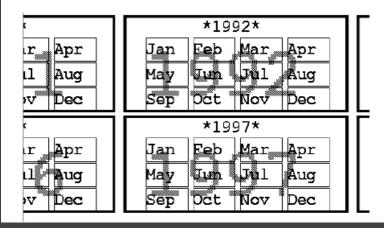
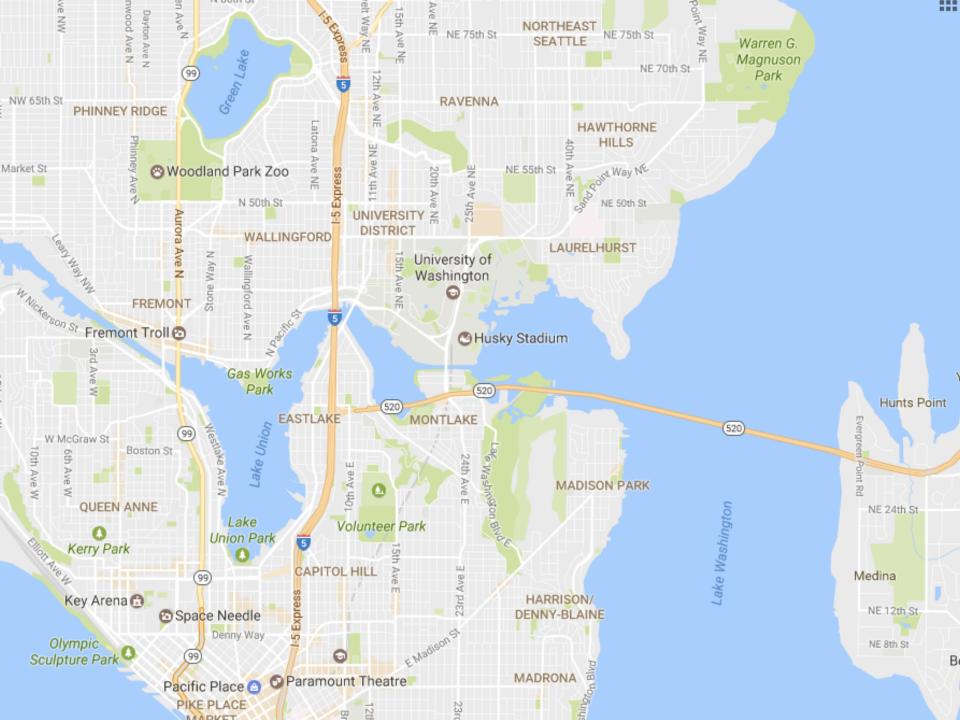


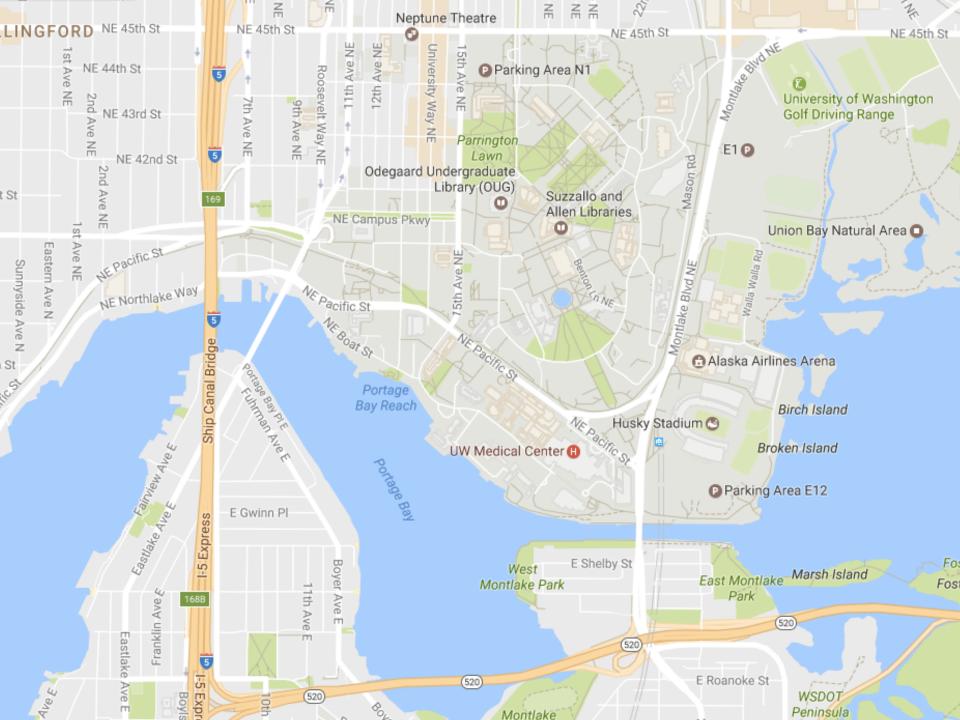
Figure 2: As you approach the calendar object the large scale display items fade out and disappear.



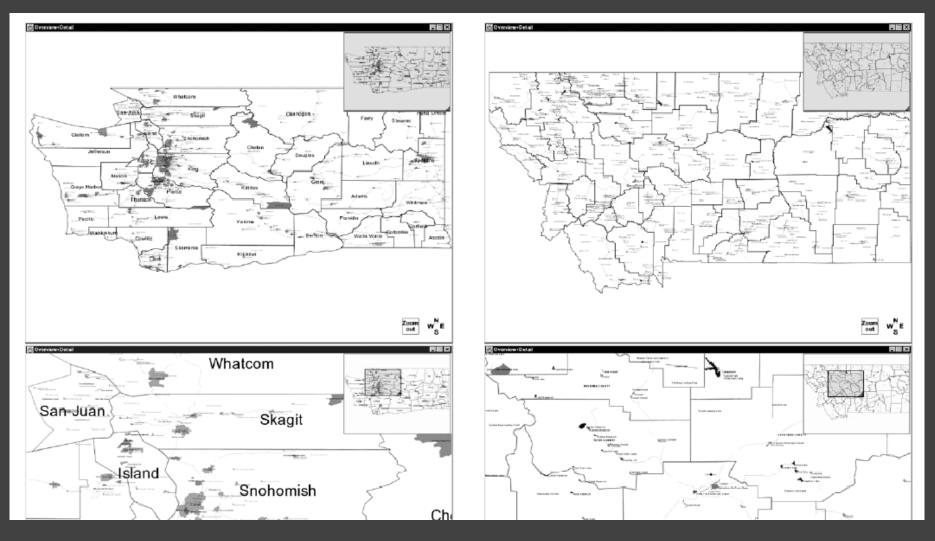
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6		
1992	Monday Dec 14 1992	Tuesday
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PAD [Perlin and Fox 93]





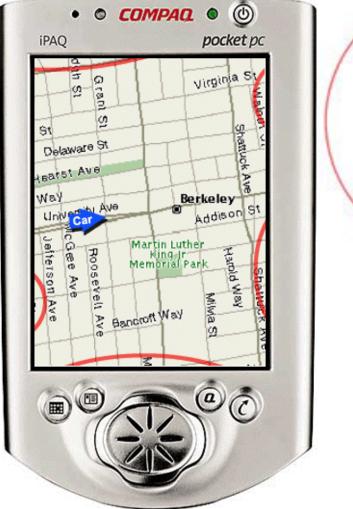
### **Overview + Detail**



[Hornbaek et al. 2002]

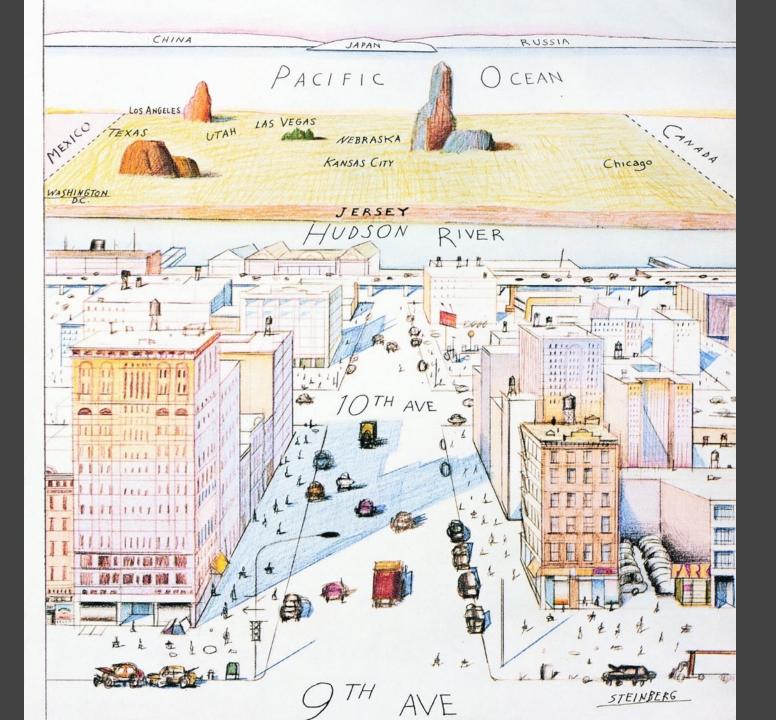
### Halo [Baudisch et al '03]





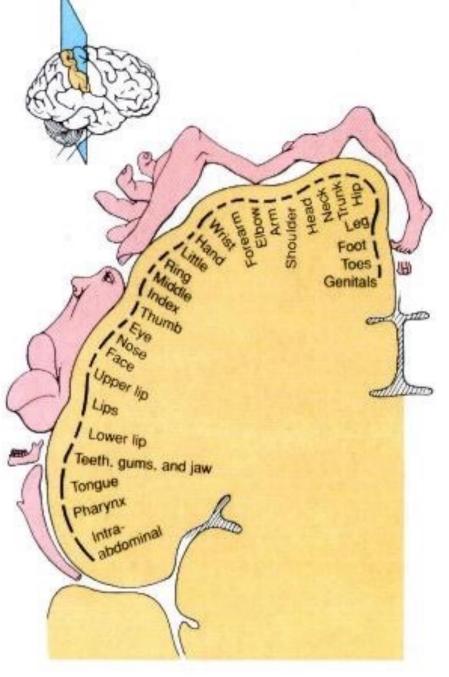


# Focus + Context





This model shows what a man's body would look like if each part grew in proportion to the area of the cortex of the brain concerned with its sensory perception. The hands and lips dominate — but the feet are also disproportionately large, indicating their sensory importance.



(a) Somatosensory cortex in right cerebral hemisphere

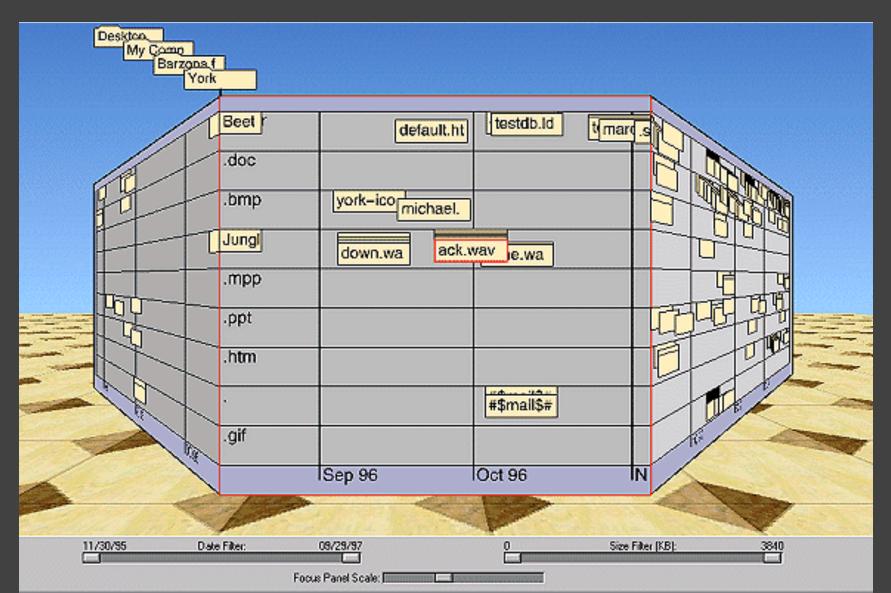
### Degree-of-Interest [Furnas '81, '06]

Estimate the saliency of information to display. Can affect **what** to show and/or **how** to show it!

DOI ~ f (Current Focus, A Priori Importance)

Example: Google Search Current Focus = Query Hits (e.g., text matching) A Priori Importance = PageRank What to show: Top N results How to show it: List

### Perspective Wall [Mackinlay et al. '91]



### TableLens [Rao & Card '94]

Baseball.txt - TLDemo								
Leag	jue		Players	At Bats	Hits	Home Runs	Runs	Rbi
N								<u> </u>
	5	52	Andres	321	87	10	39	42
	5	53	Jose Cruz	479	133	10	48	72
	-	54	Bo Diaz	474	129	10	50	56
	5	55	Tony Pena	510	147	10	56	52
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	1	91	Reggie J	419	101	18	65	58
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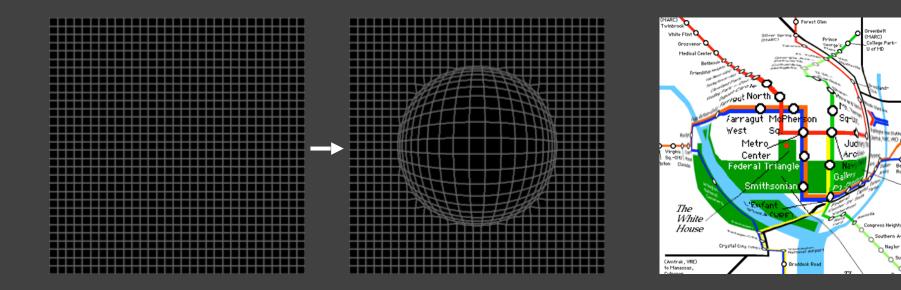
#### **DateLens** [Bederson et al. '04]

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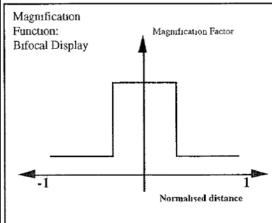
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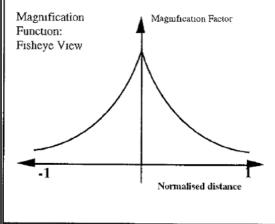
#### Focus + Context Distortion

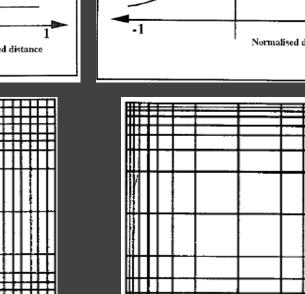
Focus area - local details De-magnified area - surrounding context Like a rubber sheet with borders tacked down

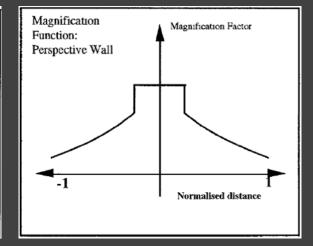


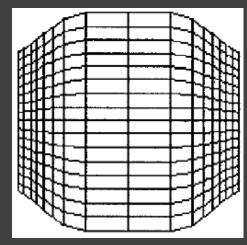
#### **Distortion Functions**











**Perspective Wall** 

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**Bifocal** 



#### Uses (and Abuses) of Distortion

#### Often more harm than help...

Pan & zoom more familiar – and visually stable – than "rubber sheet" navigation.

Consider F+C over **data** rather than just **view** 

## **Dimensionality Reduction**

**Project from nD down to 2D.** Spatial proximity –> data points are (approximately) more similar. File Options

# Dimensionality Reduction

2:-0.157,0.106(47.74) 3:-0.251,-0.178(9.00) 4:-0.442,0.723(1.00) 5:0.016,0.222(1.00) 6:0.726,0.461(3.00) 7:0.424,-0.195(1.00)

1:0.099,0.367(243.00)

<u> – – ×</u>

6

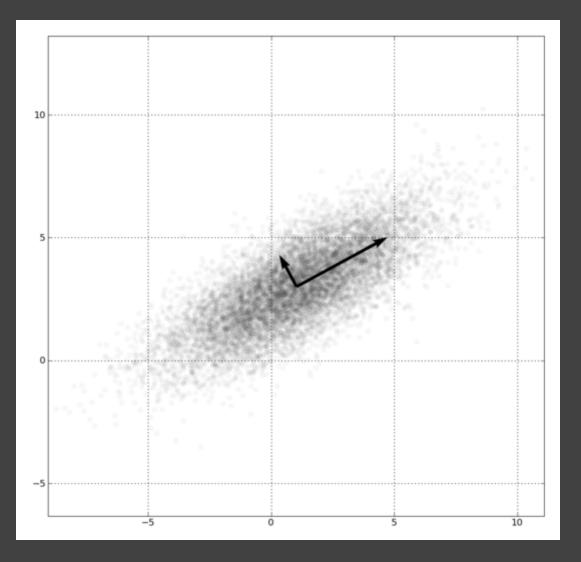
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http://www.ggobi.org/

#### **Principal Components Analysis**

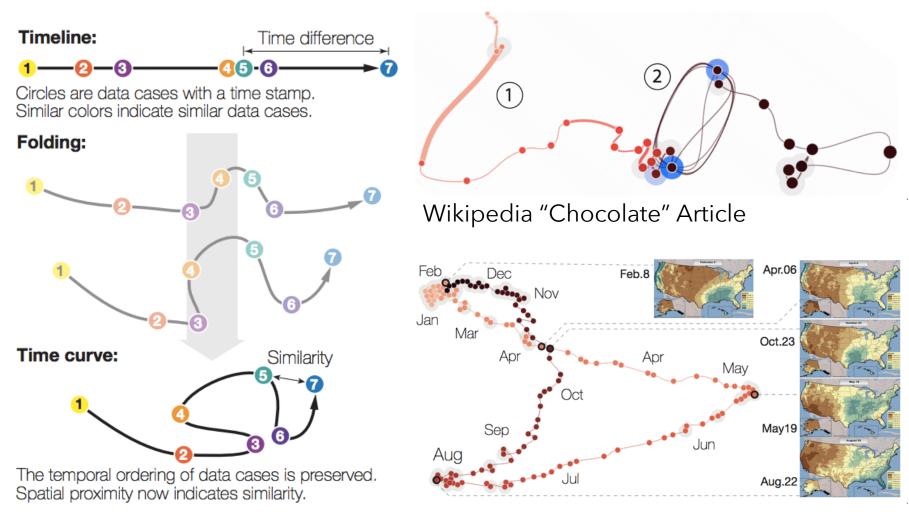


1. Mean-center the data. 2. Find  $\perp$  basis vectors that maximize the data variance. 3. Plot the data using the top vectors.

#### PCA of Genomes [Demiralp et al. '13]



#### Time Curves [Bach et al. '16]



(a) Folding time

U.S. Precipitation over 1 Year

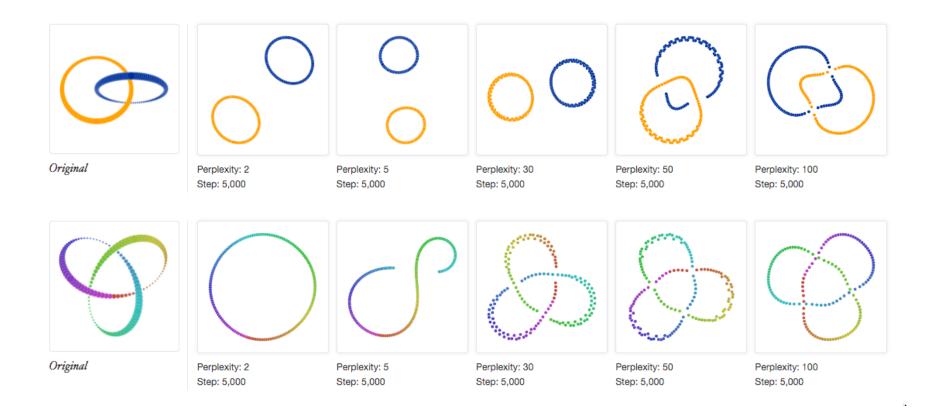
#### Many Reduction Techniques!

Principal Components Analysis (PCA) Multidimensional Scaling (MDS) Locally Linear Embedding (LLE) t-Dist. Stochastic Neighbor Embedding (t-SNE) Isomap Auto-Encoder Neural Networks Topological methods

#### Many Reduction Techniques!

Principal Components Analysis (PCA) Multidimensional Scaling (MDS) Locally Linear Embedding (LLE) t-Dist. Stochastic Neighbor Embedding (t-SNE) Isomap Auto-Encoder Neural Networks Topological methods

#### Visualizing t-SNE [Wattenberg et al. '16]



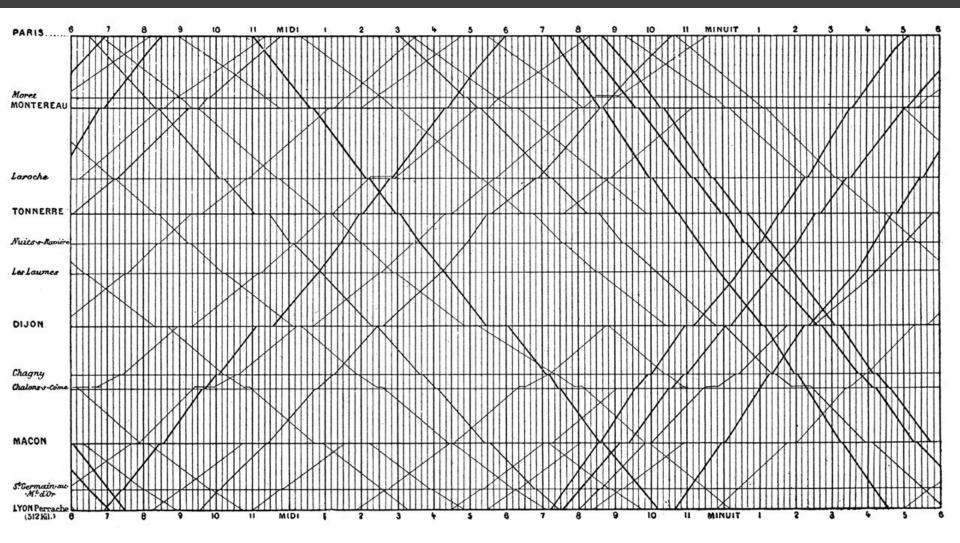
### Summary

Spatial layout is the most important encoding ... but you need to be in the right space.

Geometric properties of spatial transforms support geometric reasoning

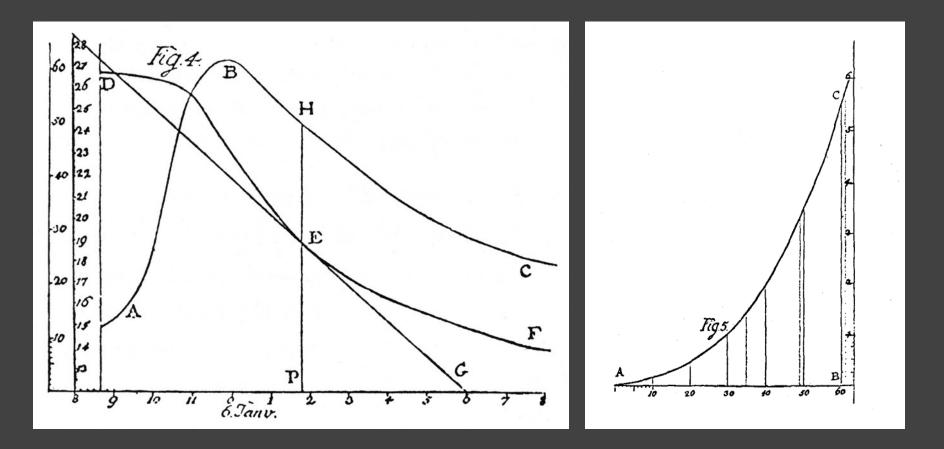
Emphasize important information Consider *what* to show, not just *how* 

# **Graphical Calculation**



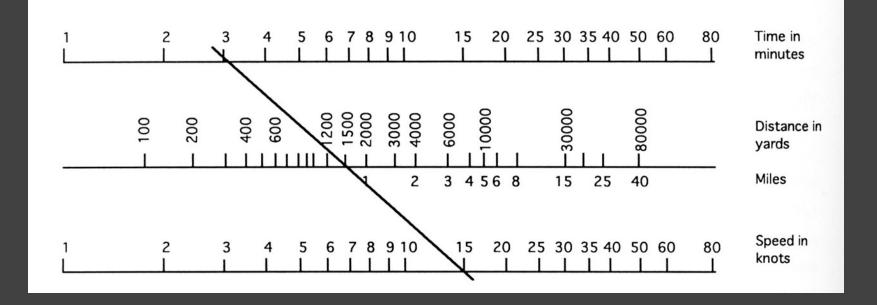
E. J. Marey, *La Méthode Graphique* (Paris, 1885), p. 20. The method is attributed to the French engineer, Ibry.

#### Lambert's Graphical Construction



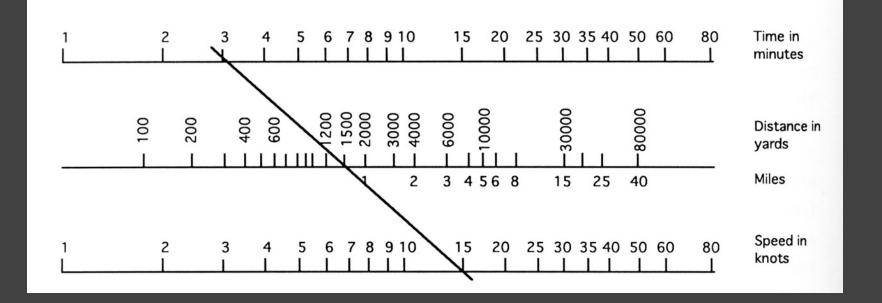
Johannes Lambert used graphs to study the rate of water evaporation as function of temperature [from Tufte 83].

#### Nomograms



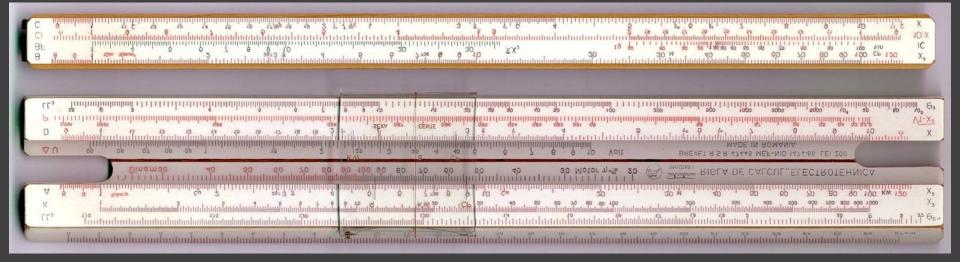
Sailing: The Rule of Three

#### Nomograms



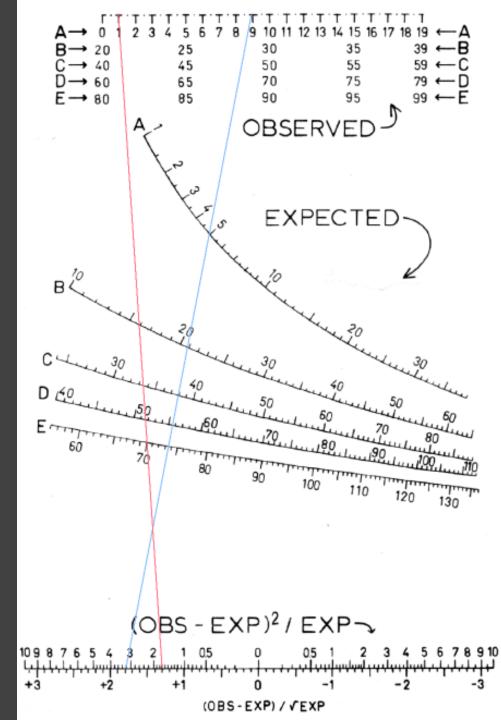
Compute in any direction; fix n-1 params and read nth param
 Illustrate sensitivity to perturbation of inputs
 Clearly show domain of validity of computation

## Slide Rule



#### Model 1474-66 Electrotechnica 18 Scales

Tehnolemn Timisoara Slide Rule Archive http://pubpages.unh.edu/~jwc/tehnolemn/



**Chi-Square Test** (Obs - Exp)<sup>2</sup> / Exp

Blue line:  $(9 - 5)^2 / 5 = 3.2$ 

Red line:  $(81 - 70)^2 / 70 = 1.7$ 

