

# Using Space Effectively

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**CS 448B: Visualization**  
**Fall 2021**

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## Reading Response Questions/Thoughts

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How do we know which type of visualization to use? Are there some general principals that lead us to choose a bar chart over a pie chart? What is the psychology of different mark types and visual encodings?

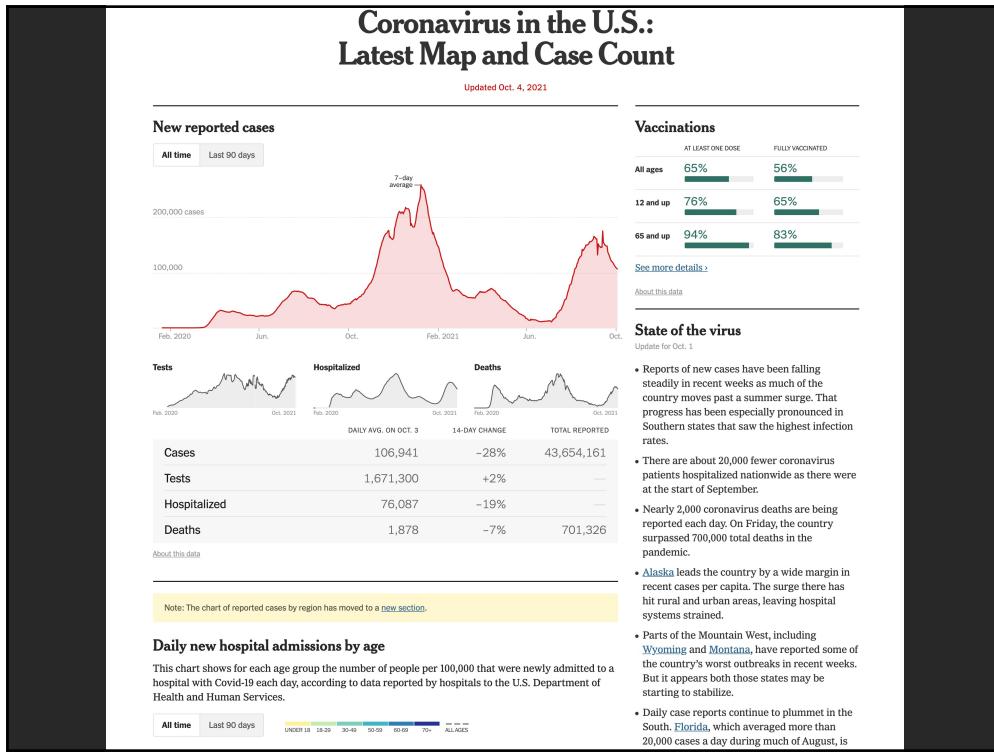
Is there a standard/scientific method of sorts by which graphic designers are supposed to explore, iterate, and finalize their designs?

In reference to the social network graph from Wednesday lecture with the node-link, linkage-sorted matrix, and non-sorted matrix views, "Are there other algorithms that can help bring out specific patterns in your data?"

In reference to public (Twitter) vs. private (academic) data visualization critiques and how people have paid more attention to data visualizations during the COVID-19 pandemic: "Do readers' goals align with designers' goals and if they don't how does that impact the insights that users walk away with as well as the redesign process?"

Is it fair to leave it solely up to the experts? Furthermore, how do authors communicate their goals to users?"

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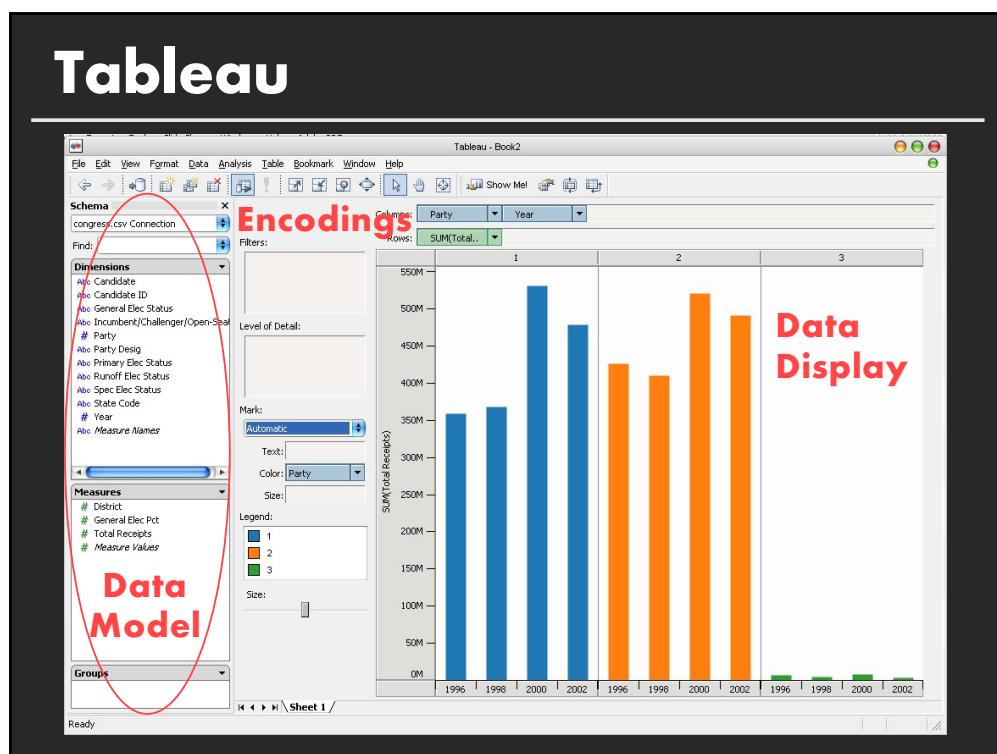


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## Last Time: EDA

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## Specifying Table Configurations

**Operands are names of database fields**  
 Each operand interpreted as a set {...}  
 Data is either O or Q and treated differently

**Three operators:**

- concatenation (+)
- cross product (x)
- nest (/)

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## Table Algebra

The operators (+,x,/) and operands (O,Q) provide an algebra for tabular visualization

Algebraic statements are mapped to

**Visualizations** – trellis partitions, visual encodings

**Queries** – selection, projection, group-by

In Tableau, users make statements via drag-and-drop

Users specify operands NOT operators!

Operators are inferred by data type (O,Q)

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## Table Algebra: Operands

**Ordinal fields:** interpret domain as a set that partitions table into rows and columns

**Quarter** =  $\{(Qtr1), (Qtr2), (Qtr3), (Qtr4)\} \rightarrow$

Qtr1	Qtr2	Qtr3	Qtr4
95892	101760	105282	98225

**Quantitative fields:** treat domain as single element set and encode spatially as axes

**Profit** =  $\{(\text{Profit}[-410,650])\} \rightarrow$



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# Concatenation (+) Operator

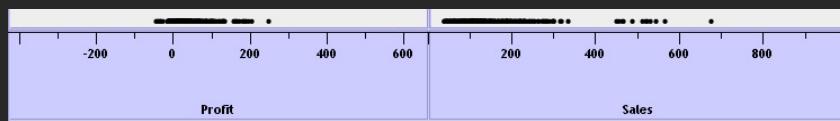
## Ordered union of sets

### Quarter + Product Type

$$\begin{aligned} &= \{(\text{Qtr1}), (\text{Qtr2}), (\text{Qtr3}), (\text{Qtr4})\} + \{(\text{Coffee}), (\text{Espresso})\} \\ &= \{(\text{Qtr1}), (\text{Qtr2}), (\text{Qtr3}), (\text{Qtr4}), (\text{Coffee}), (\text{Espresso})\} \end{aligned}$$

Qtr1	Qtr2	Qtr3	Qtr4	Coffee	Espresso
48	59	57	53	151	21

$$\text{Profit} + \text{Sales} = \{(\text{Profit}[-310, 620]), (\text{Sales}[0, 1000])\}$$



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# Cross (x) Operator

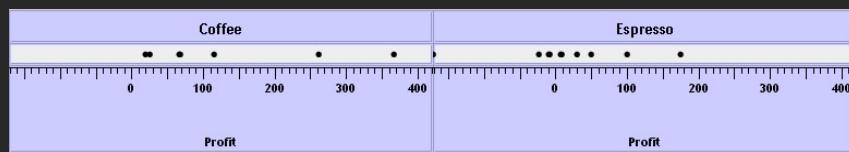
## Cross-product of sets

### Quarter x Product Type

$$= \{(\text{Qtr1}, \text{Coffee}), (\text{Qtr1}, \text{Tea}), (\text{Qtr2}, \text{Coffee}), (\text{Qtr2}, \text{Tea}), (\text{Qtr3}, \text{Coffee}), (\text{Qtr3}, \text{Tea}), (\text{Qtr4}, \text{Coffee}), (\text{Qtr4}, \text{Tea})\}$$

Qtr1		Qtr2		Qtr3		Qtr4	
Coffee	Espresso	Coffee	Espresso	Coffee	Espresso	Coffee	Espresso
131	19	160	20	178	12	134	33

$$\text{Product Type} \times \text{Profit} =$$



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# Nest (/) Operator

**Cross-product filtered by existing records**

**Quarter x Month**

creates 12 entries for each qtr. i.e., (Qtr1, Dec)

**Quarter / Month**

creates three entries per quarter based on tuples in database (not semantics)

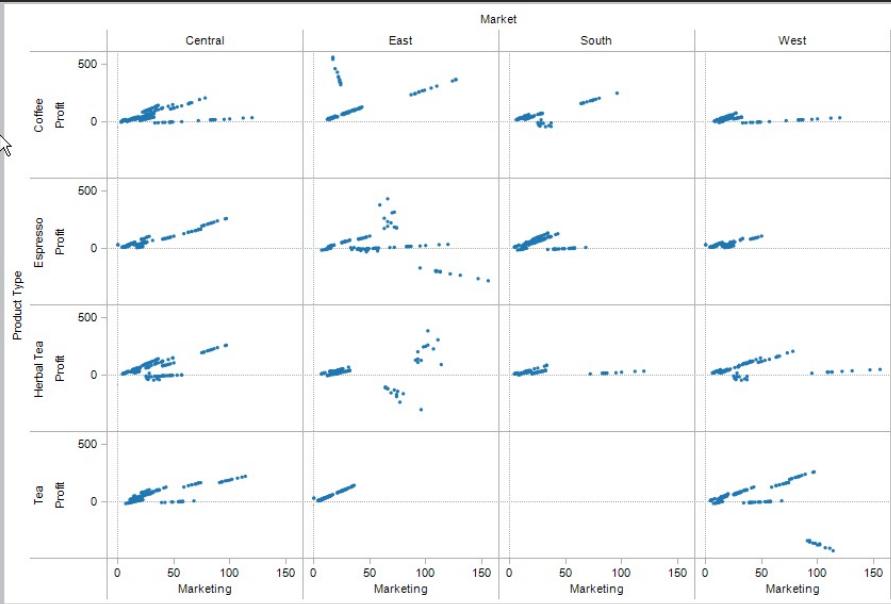
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# Ordinal - Ordinal

State	Product Type			
	Coffee	Espresso	Herbal Tea	Tea
Colorado	•	•	•	•
Connecticut	•	•	•	•
Florida	•	•	•	•
Illinois	•	•	•	•
Iowa	•	•	•	•
Louisiana	•	•	•	•
Massachusetts	•	•	•	•
Missouri	•	•	•	•
Nevada	•	•	•	•
New Hampshire	•	•	•	•
New Mexico	•	•	•	•
New York	•	•	•	•
Ohio	•	•	•	•
Oklahoma	•	•	•	•
Oregon	•	•	•	•
Texas	•	•	•	•
Utah	•	•	•	•
Washington	•	•	•	•
Wisconsin	•	•	•	•

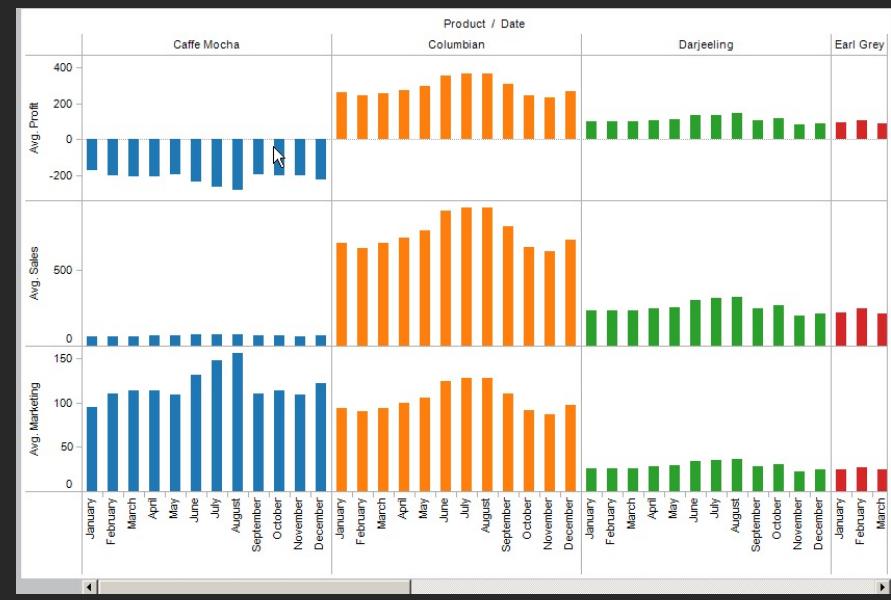
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## Quantitative - Quantitative



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## Ordinal - Quantitative



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## **Summary**

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**Exploratory analysis may combine graphical methods, and statistics**

**Use questions to uncover more questions**

**Interaction is essential for exploring large multidimensional datasets**

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## **Announcements**

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## A2: Exploratory Data Analysis

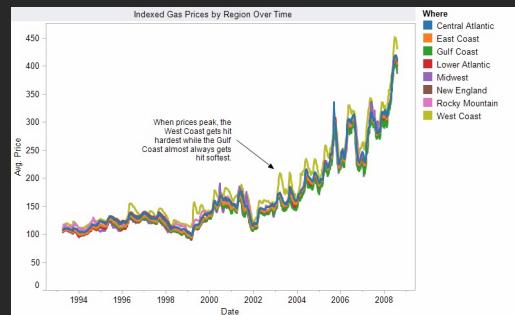
Use **Tableau** or **Vega-Lite** to formulate & answer questions

### 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

**Due before class on Oct 11, 2021**

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## Using Space Effectively

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## **Topics**

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**Graphs and lines**

**Selecting aspect ratio**

**Fitting data and depicting residuals**

**Sorting**

**Graphical calculations**

**Cartographic distortion**

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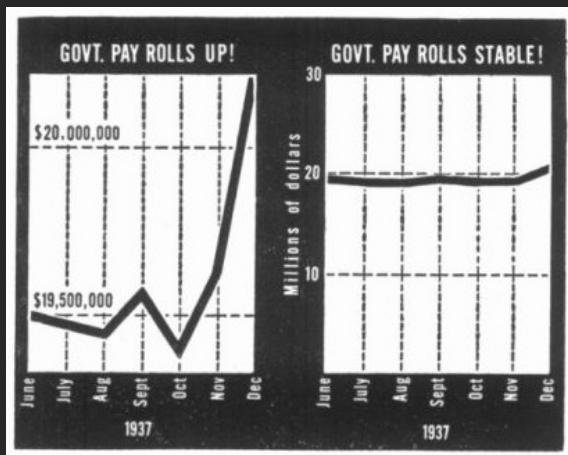
**Graphs and Lines**

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## Effective use of space

Which graph is better?

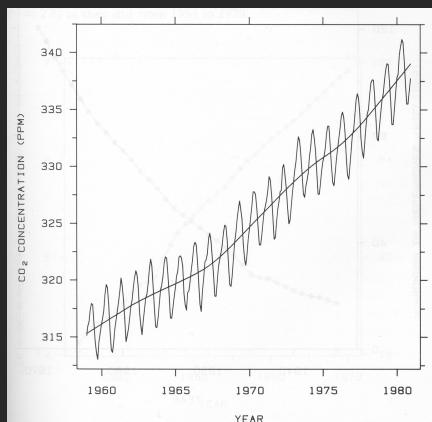
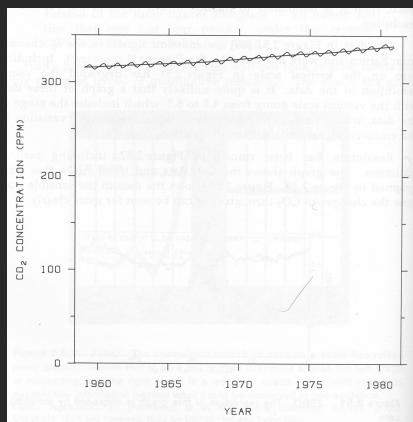


Government payrolls in 1937 [Huff 93]

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## Fill space

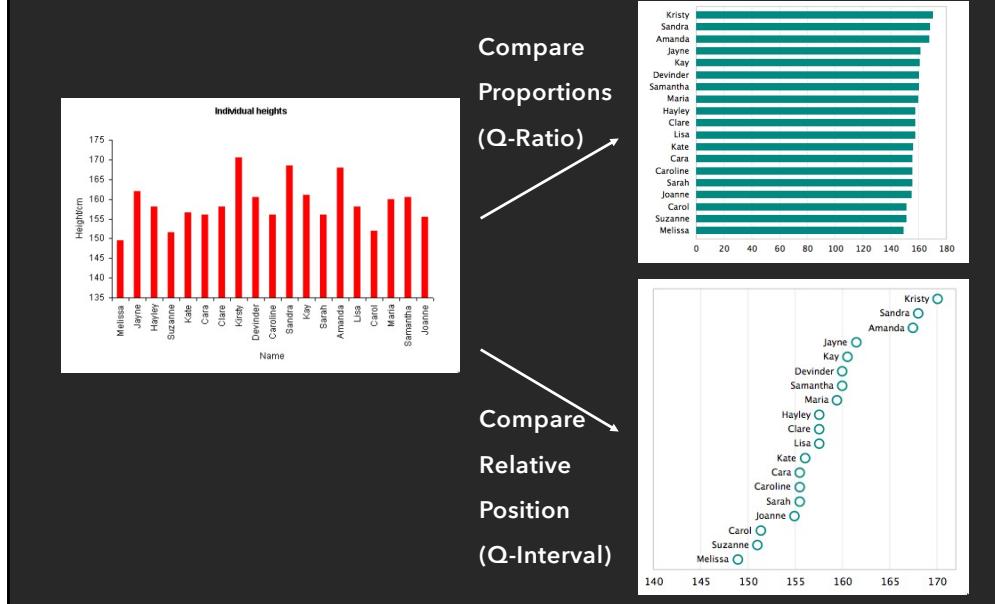
Show data with as much resolution as possible  
Don't worry about showing zero



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

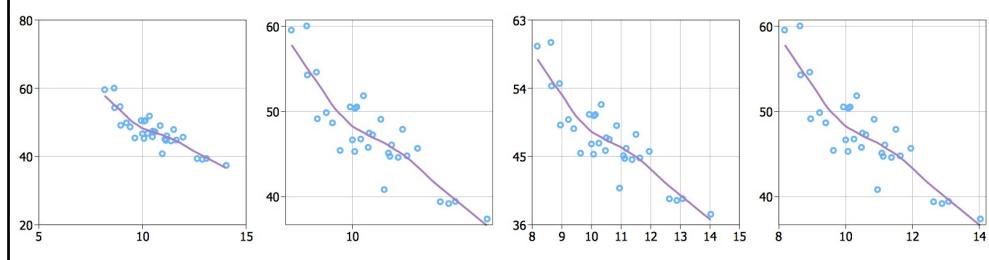
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# Include zero in axis scale?



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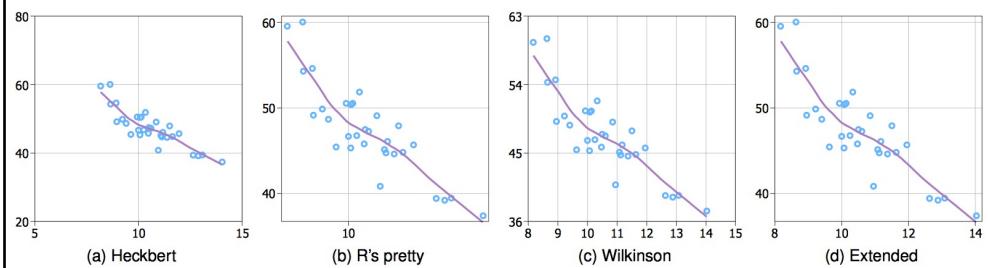
# Axis Tick Mark Selection



What are some properties of “good” tick marks?

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## Axis Tick Mark Selection



**Simplicity** - numbers are multiples of 10, 5, 2

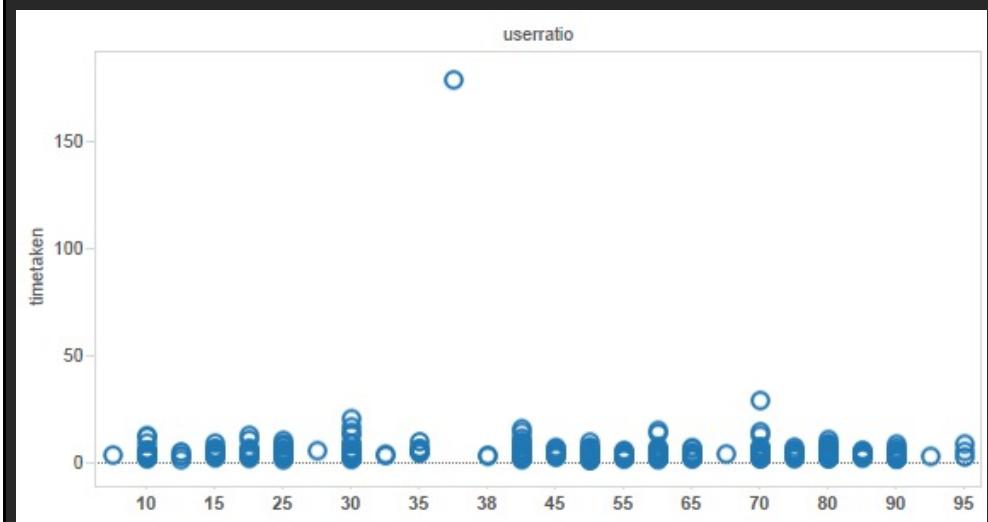
**Coverage** - ticks near the ends of the data

**Density** - not too many, nor too few

**Legibility** - whitespace, horizontal text, size

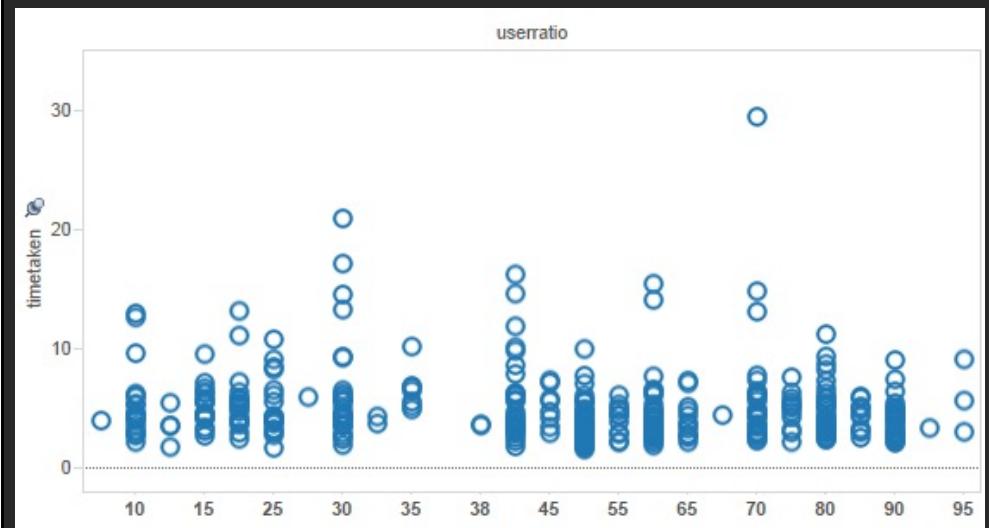
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## How to Scale the Axis?



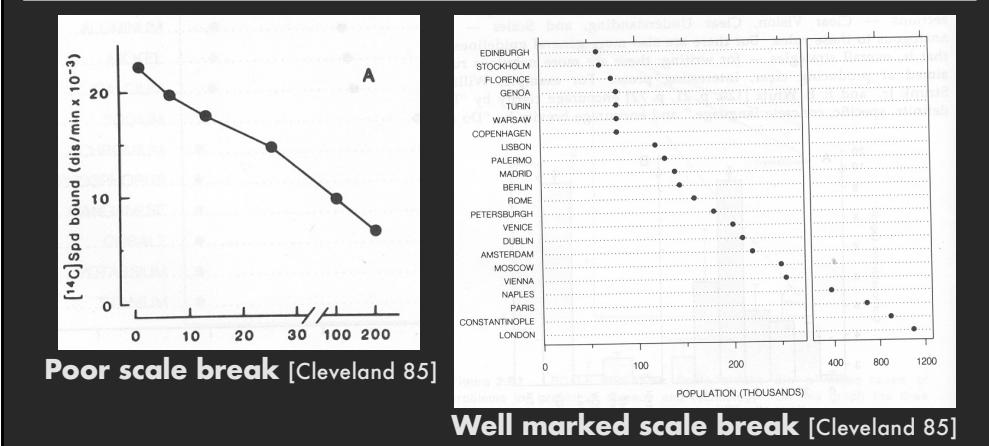
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## One Option: Clip Outliers



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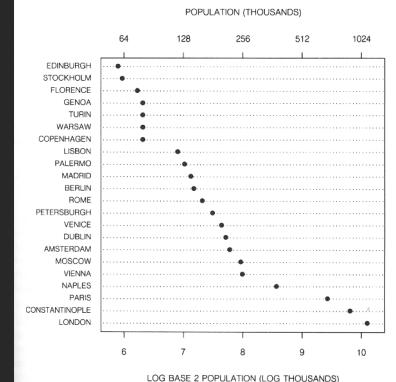
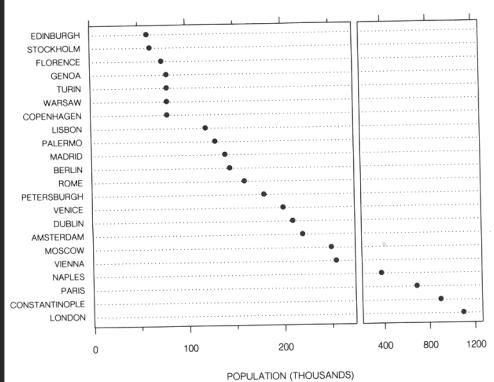
## Clearly mark scale breaks



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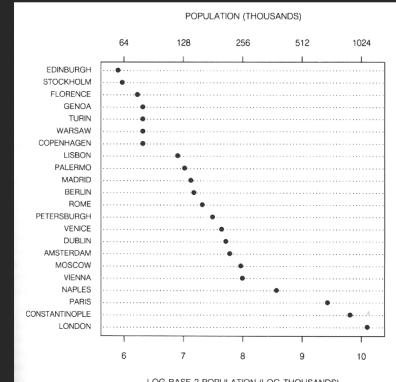
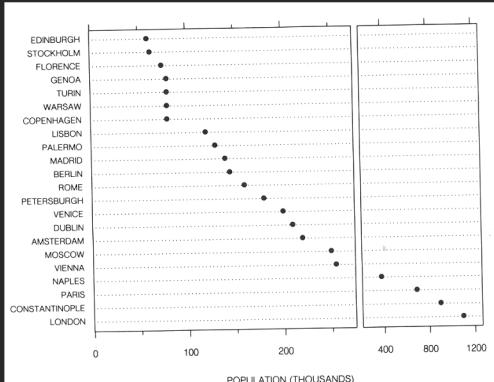
## Scale break vs. Log scale



[Cleveland 85]

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## Scale break vs. Log scale



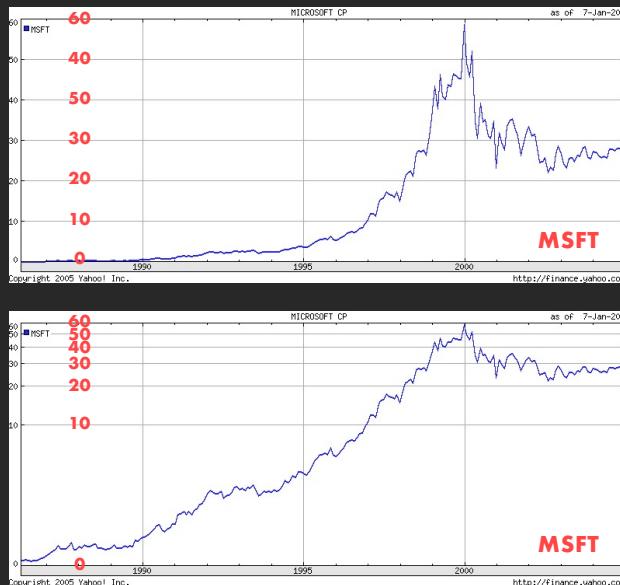
[Cleveland 85]

**Both increase visual resolution**

- **Log scale - easy comparisons of all data**
- **Scale break – more difficult to compare across break**

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## Linear scale vs. Log scale

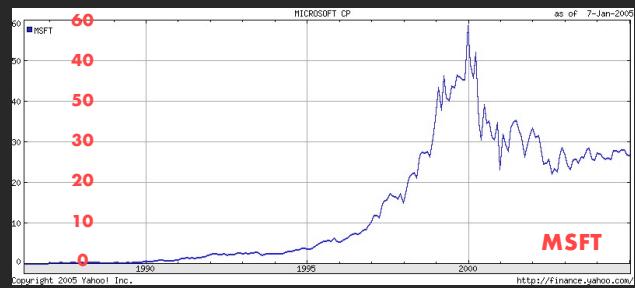


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## Linear scale vs. Log scale

### Linear scale

- Absolute change

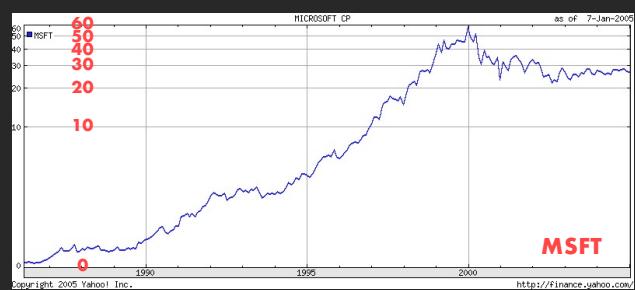


### Log scale

- Small fluctuations

- Percent change

$$d(10,20) = d(30,60)$$



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## Log scales

Logarithms turn multiplication into addition

$$\log(xy) = \log(x) + \log(y)$$

Equal steps on a log scale correspond to equal changes to a multiplicative scale factor

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## When to apply log scale?

Address data skew (e.g., long tails, outliers)

Enables comparison across multiple orders of magnitude

Focus on multiplicative factors (not additive)

Recall that the logarithm transforms  $\times$  to  $+$ !

Percentage change, not linear difference.

Constraint: positive, non-zero values

Constraint: audience familiarity?

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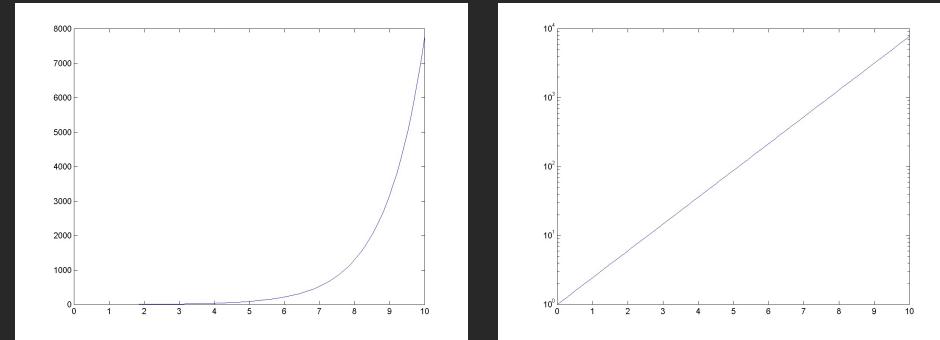
## Semilog graph: Exponential growth

**Exponential functions ( $y = ka^{mx}$ ) transform into lines**

$$\log(y) = \log(k) + \log(a)mx$$

**Intercept:**  $\log(k)$

**Slope:**  $\log(a)m$



$$y = 6^{0.5x}, \text{slope in semilog space: } \log(6) * 0.5 = 0.3891$$

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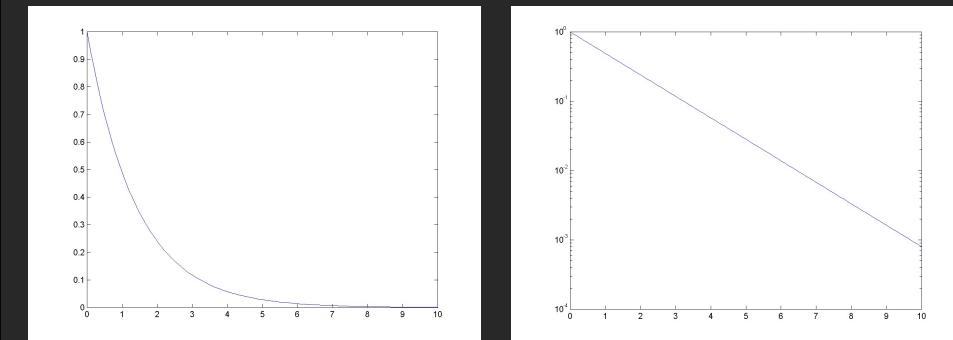
## Semilog graph: Exponential decay

**Exponential functions ( $y = ka^{mx}$ ) transform into lines**

$$\log(y) = \log(k) + \log(a)mx$$

**Intercept:**  $\log(k)$

**Slope:**  $\log(a)m$

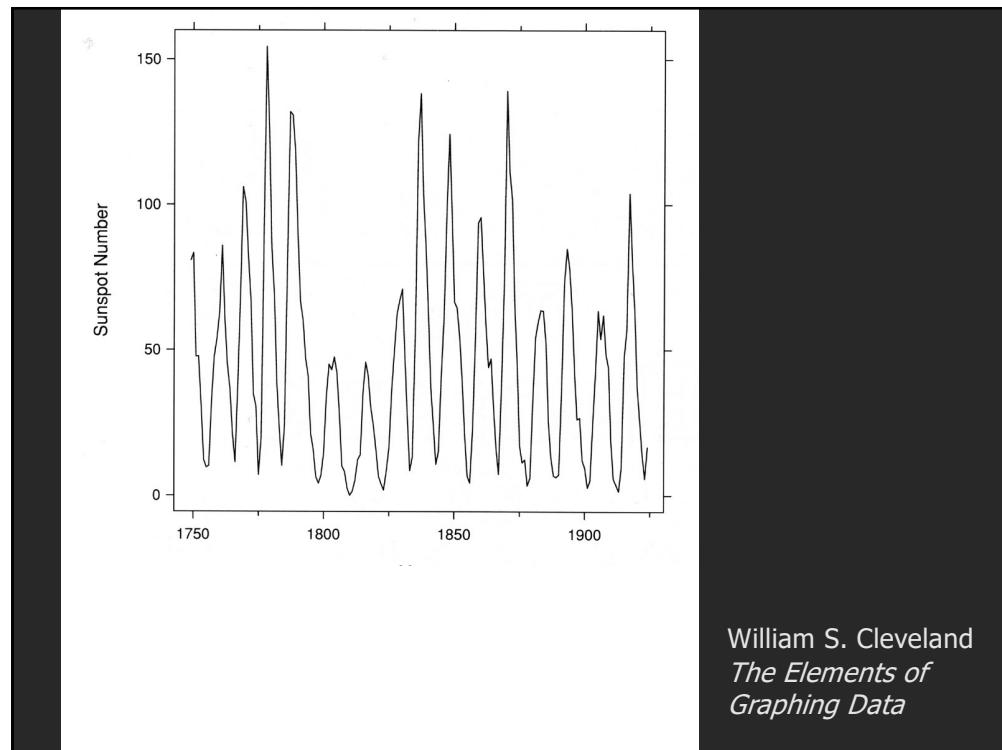


$$y = 0.5^{2x}, \text{slope in semilog space: } \log(0.5) * 2 = -0.602$$

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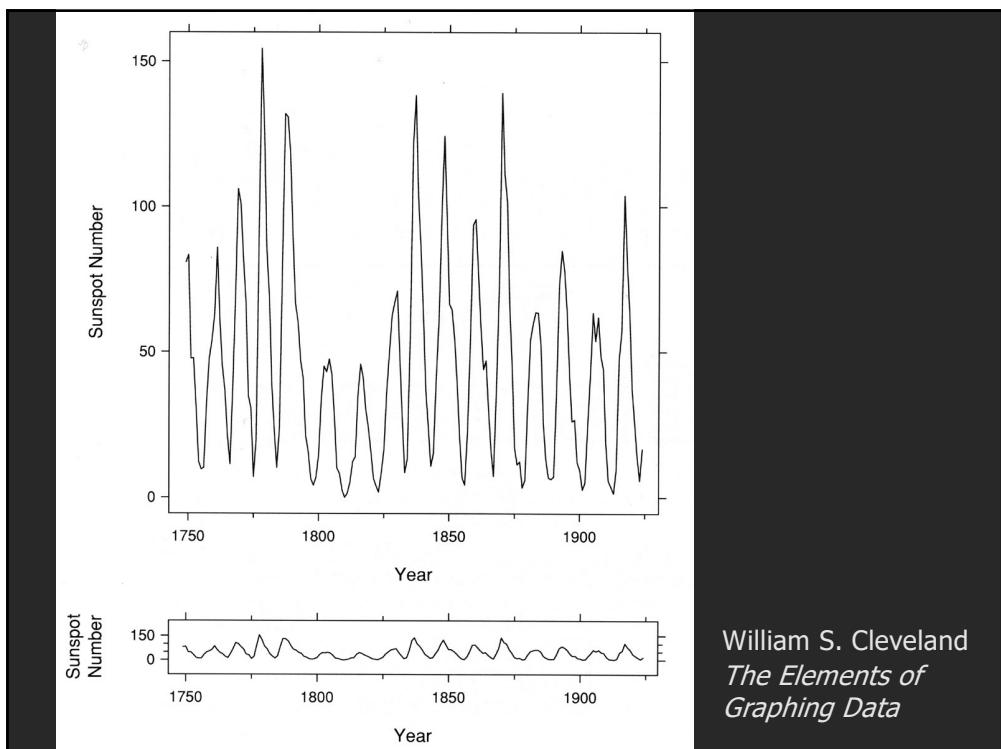
# Selecting Aspect Ratio

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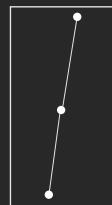
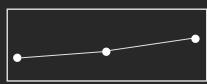
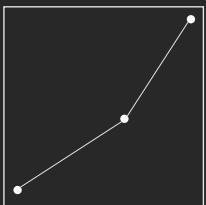
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## Banking to $45^\circ$ [Cleveland]

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

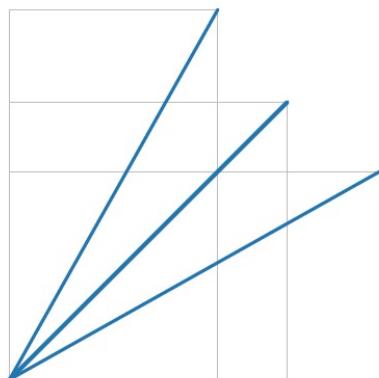


Two line segments are maximally discriminable when the absolute angle between them is  $45^\circ$

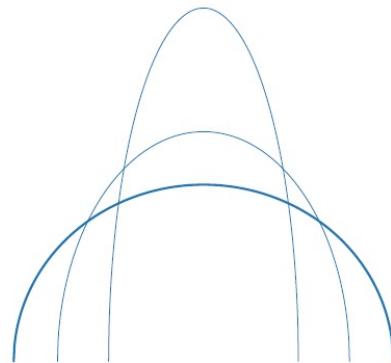
**Method:** Optimize the aspect ratio such that the average absolute angle between all segments is  $45^\circ$

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An alternate approach:  
**Minimize arc length** (hold area constant)



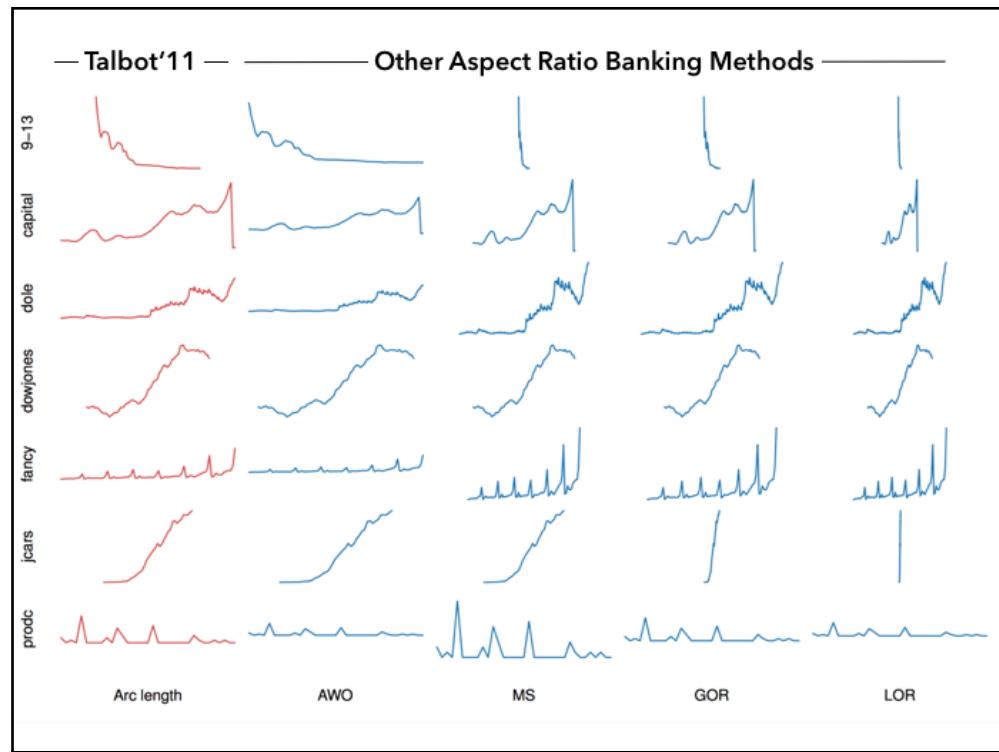
Straight line  $\rightarrow 45$  deg



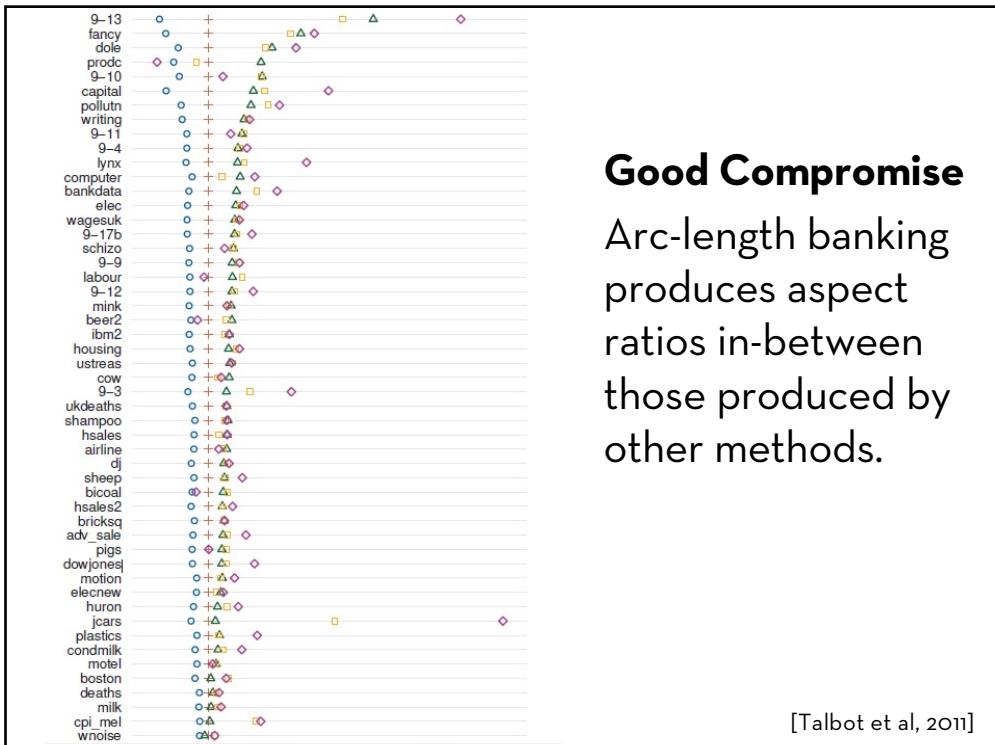
Ellipse  $\rightarrow$  Circle

[Talbot et al, 2011]

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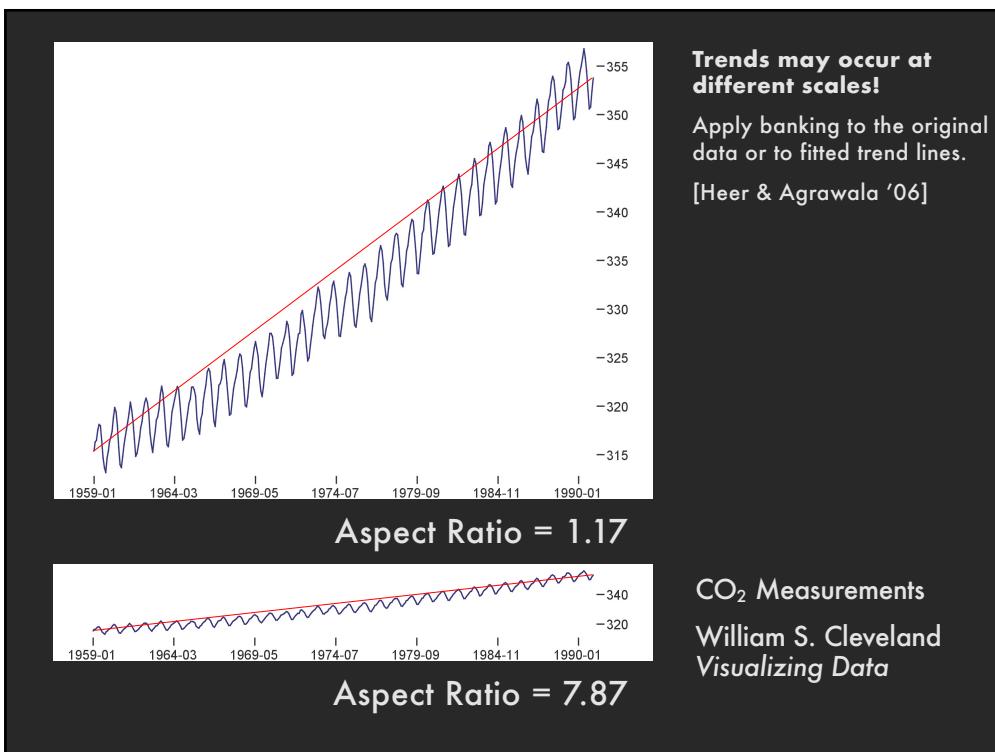
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## Good Compromise

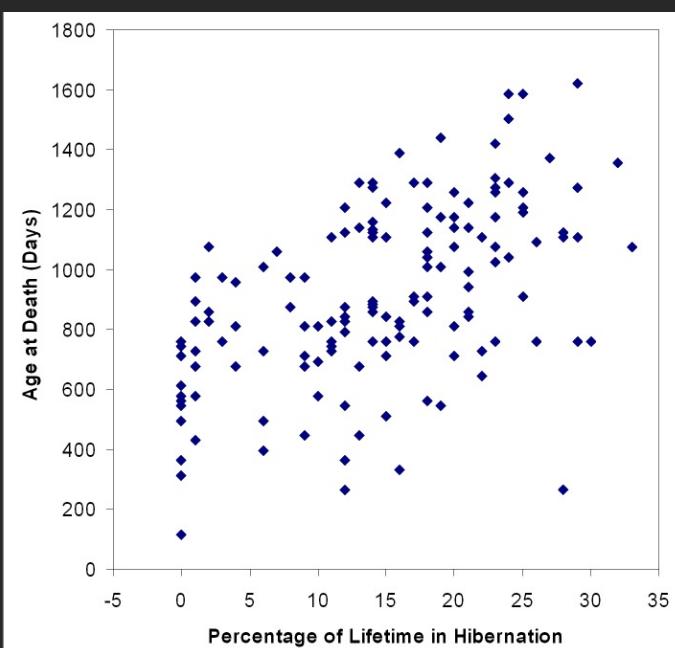
Arc-length banking produces aspect ratios in-between those produced by other methods.



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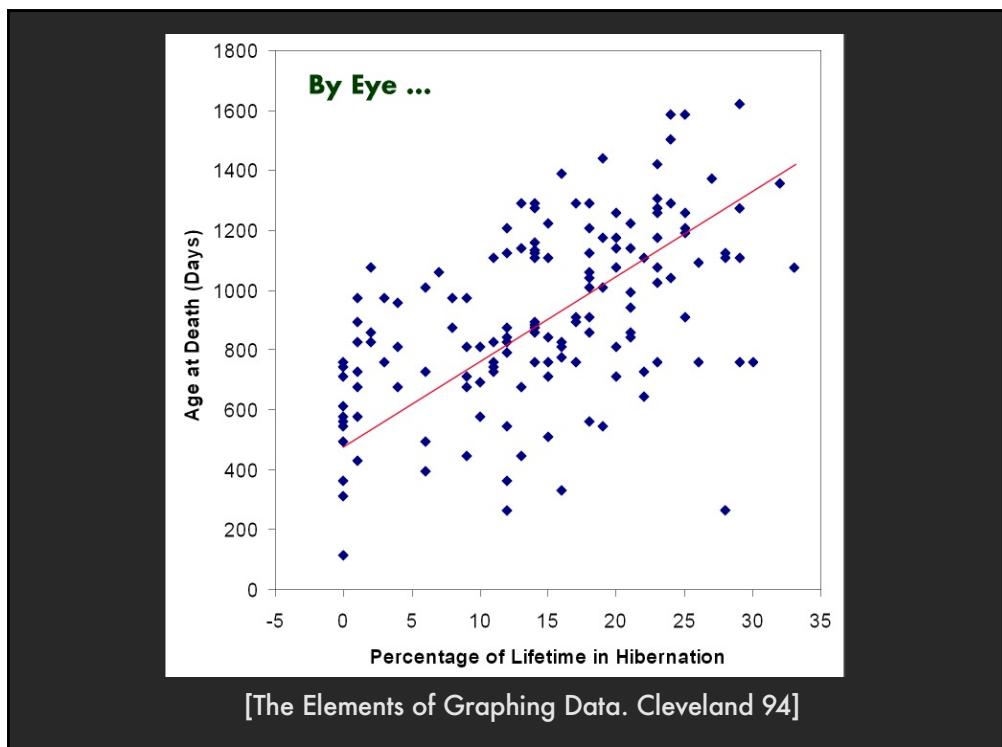
# Fitting the Data

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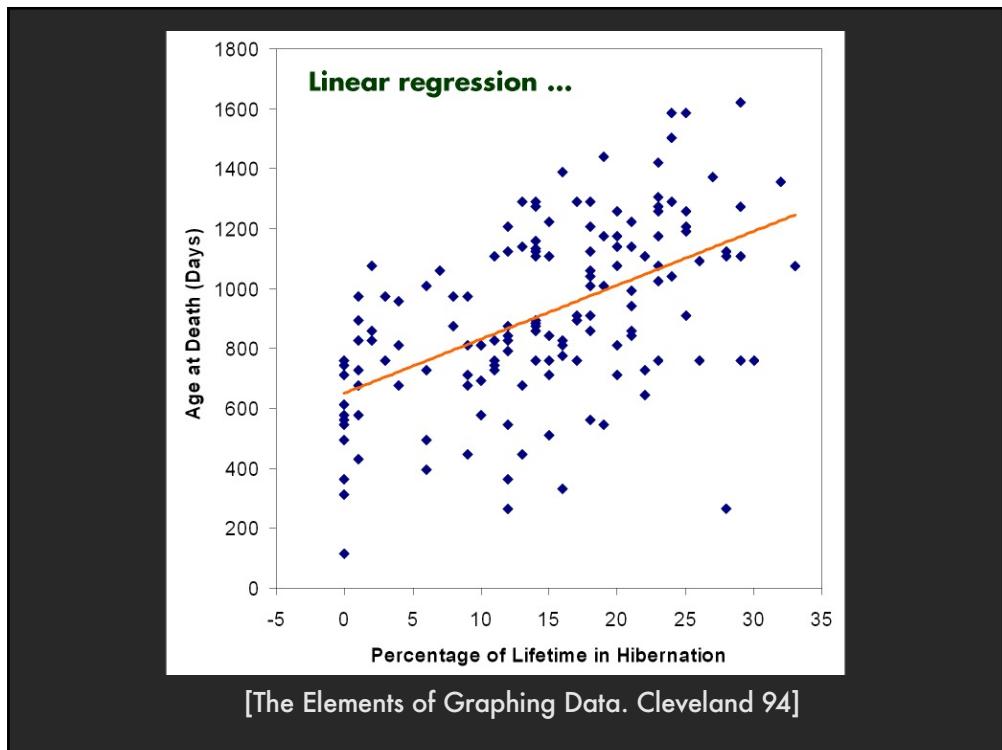


[The Elements of Graphing Data. Cleveland 94]

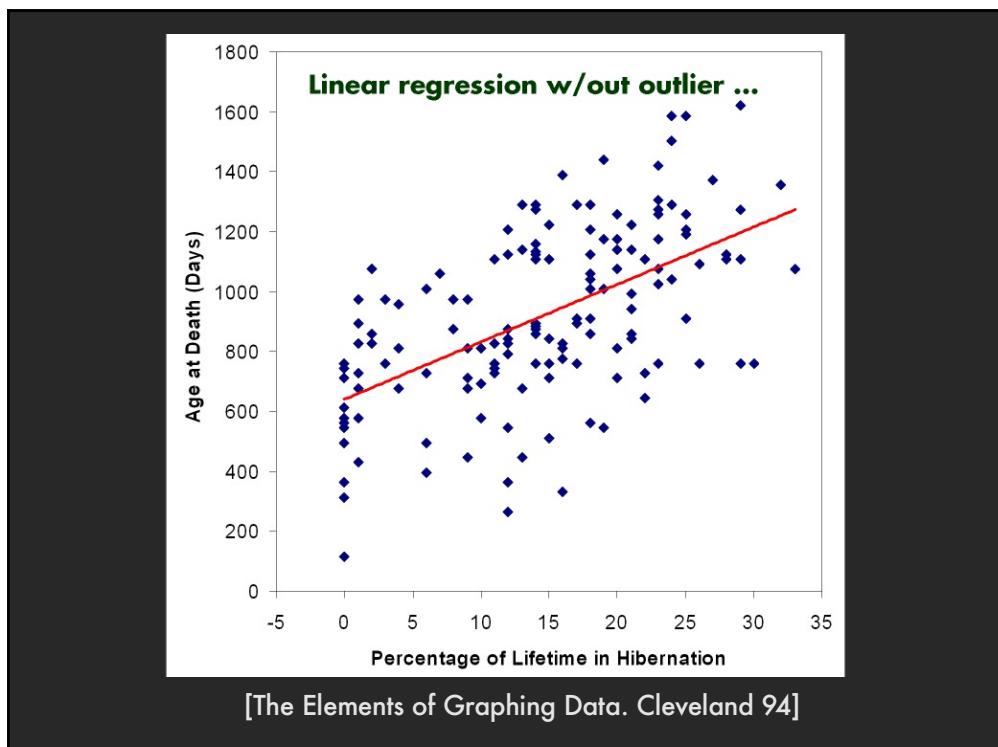
80



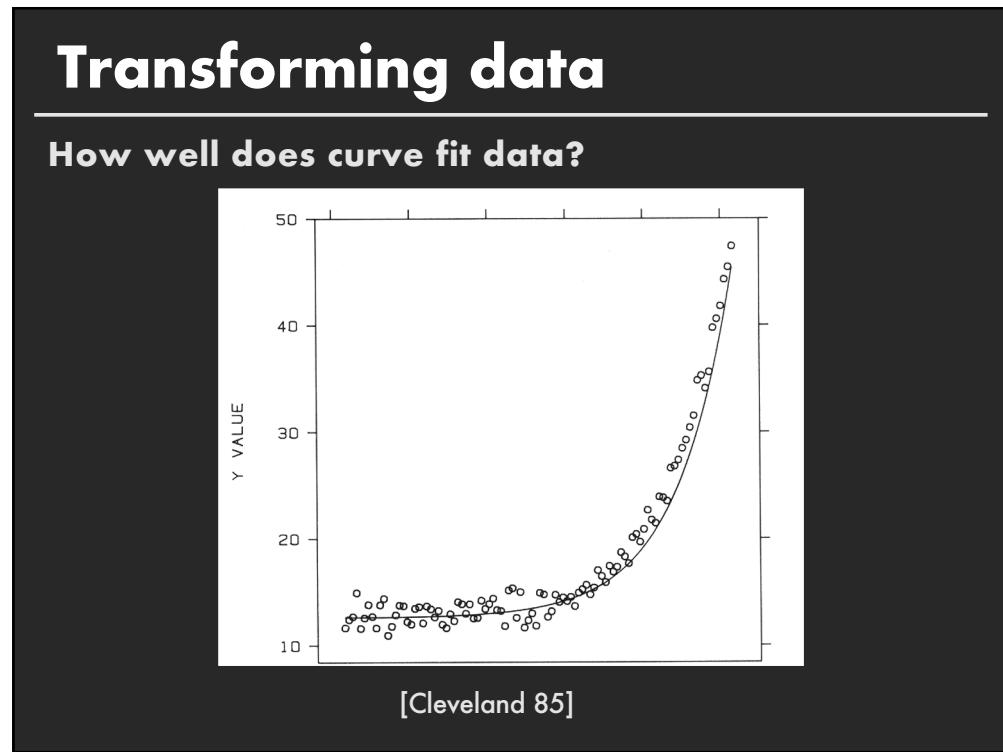
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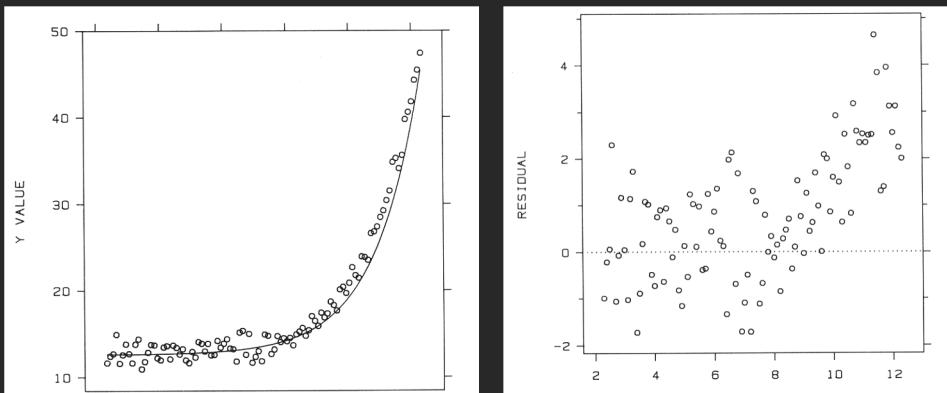


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# Transforming data

## Residual graph

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



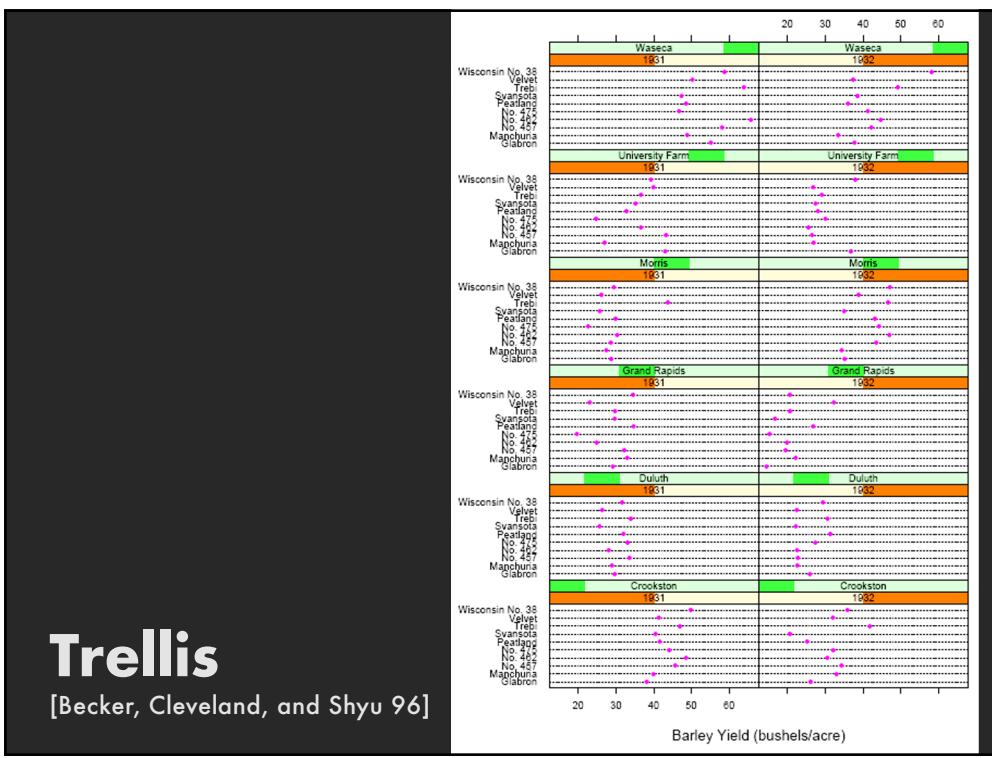
[Cleveland 85]

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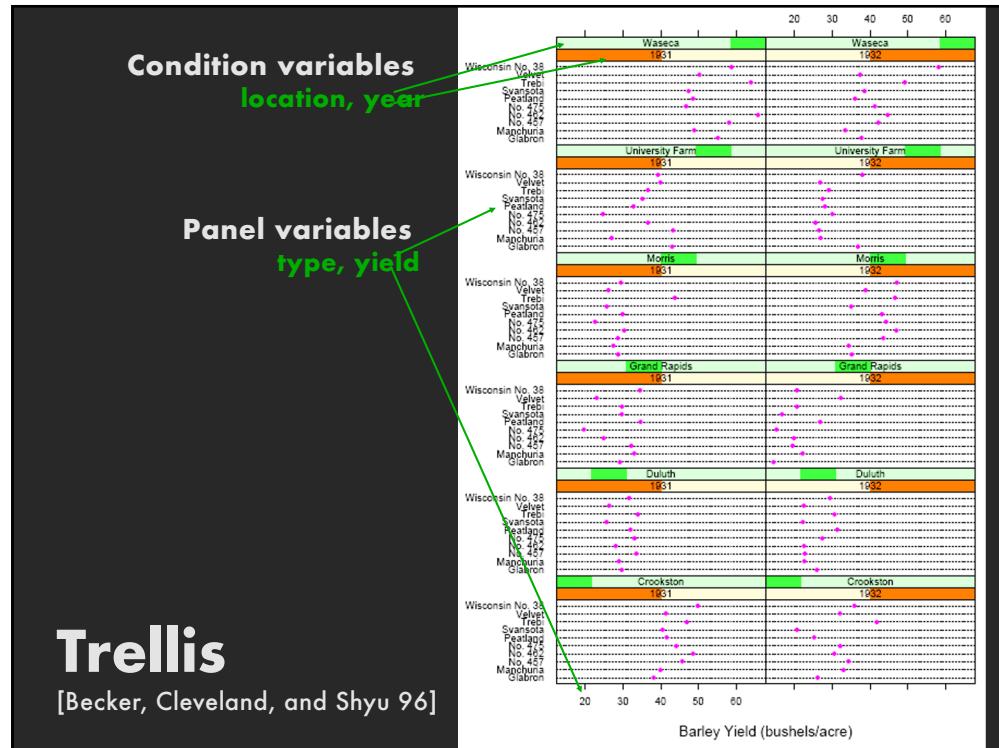
# Sorting

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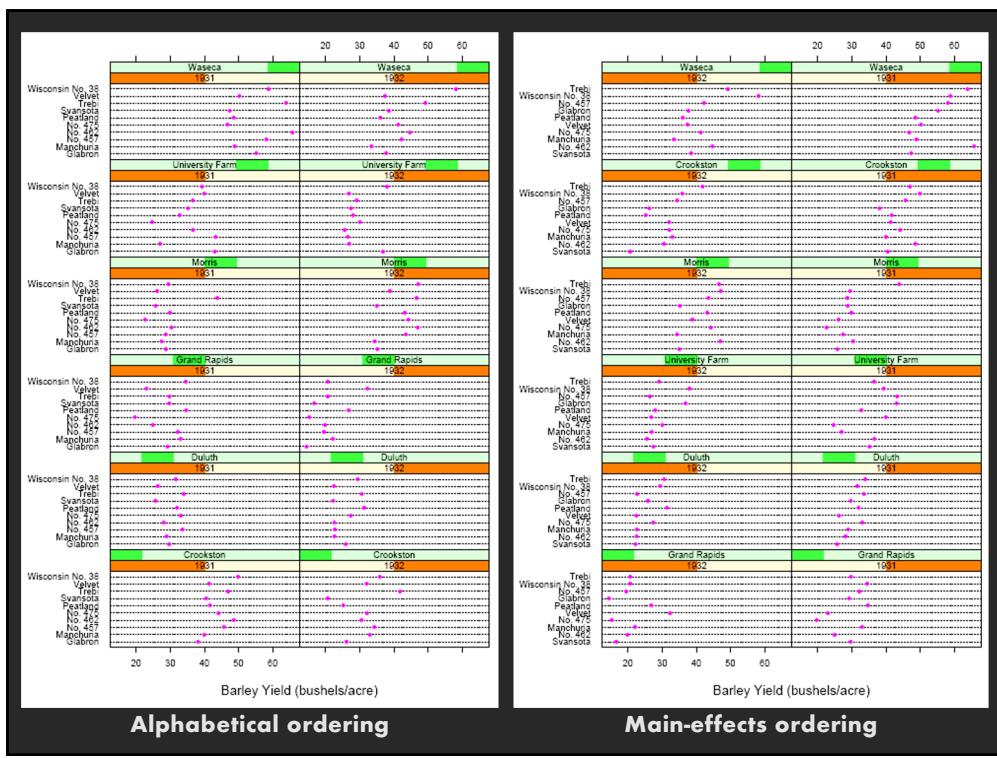
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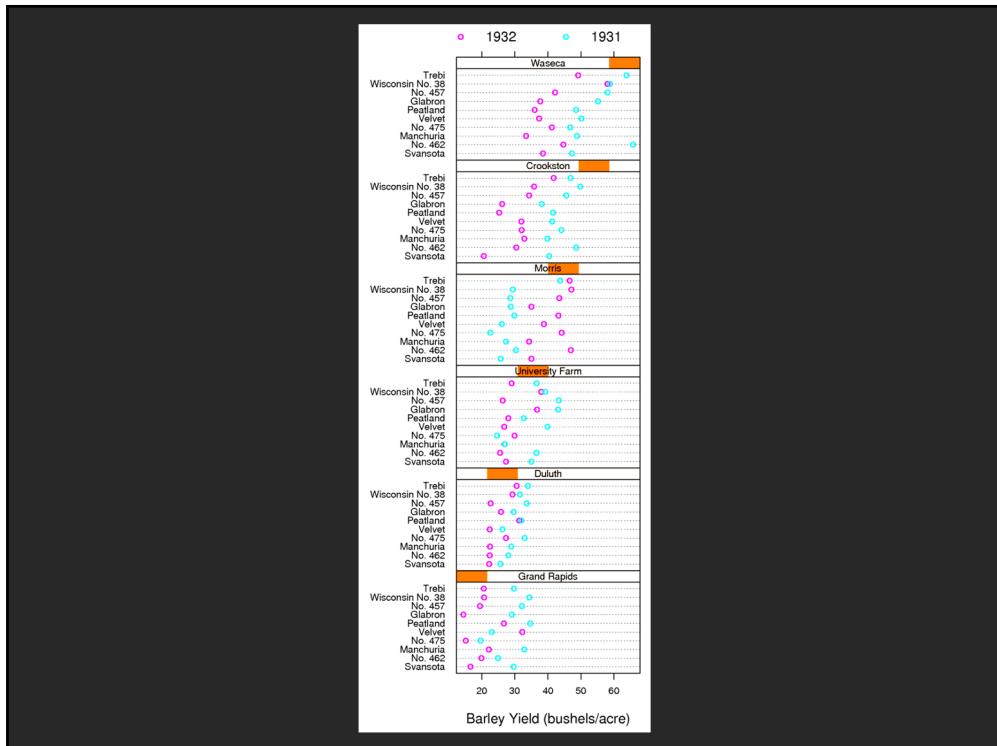
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