

# Lec 09 - Visualization with ggplot2

Statistical Programming

Sem 1, 2020

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Why do we visualize?

# Asncombe's Quartet

```
datasets::anscombe %>% as_tibble()
```

```
## # A tibble: 11 x 8
##       x1    x2    x3    x4    y1    y2    y3    y4
##   <dbl> <dbl> <dbl> <dbl> <dbl> <dbl> <dbl> <dbl>
## 1     10     10     10      8  8.04  9.14  7.46  6.58
## 2      8      8      8      8  6.95  8.14  6.77  5.76
## 3     13     13     13      8  7.58  8.74 12.7   7.71
## 4      9      9      9      8  8.81  8.77  7.11  8.84
## 5     11     11     11      8  8.33  9.26  7.81  8.47
## 6     14     14     14      8  9.96  8.1   8.84  7.04
## 7      6      6      6      8  7.24  6.13  6.08  5.25
## 8      4      4      4     19  4.26  3.1   5.39 12.5
## 9     12     12     12      8 10.8   9.13  8.15  5.56
## 10     7      7      7      8  4.82  7.26  6.42  7.91
## 11     5      5      5      8  5.68  4.74  5.73  6.89
```

# Tidy anscombe

```
(tidy_anscombe = datasets::anscombe %>%  
  pivot_longer(everything(), names_sep = 1, names_to = c("var", "group")) %>%  
  pivot_wider(id_cols = group, names_from = var,  
              values_from = value, values_fn = list(value = list)) %>%  
  unnest(cols = c(x,y)))
```

```
## # A tibble: 44 x 3  
##   group      x      y  
##   <chr> <dbl> <dbl>  
## 1 1      10  8.04  
## 2 1       8  6.95  
## 3 1      13  7.58  
## 4 1       9  8.81  
## 5 1      11  8.33  
## 6 1      14  9.96  
## 7 1       6  7.24  
## 8 1       4  4.26  
## 9 1      12 10.8  
## 10 1       7  4.82  
## # ... with 34 more rows
```

```
tidy_anscombe %>%  
  group_by(group) %>%  
  summarize(mean_x = mean(x), mean_y = mean(y), sd_x = sd(x), sd_y = sd(y), cor = cor(x,y))
```

```
## `summarise()` ungrouping output (override with `.groups` argument)  
## # A tibble: 4 x 6
```

```
ggplot(tidy_anscombe, aes(x = x, y = y, color = as.factor(group))) +  
  geom_point(size=2) +  
  facet_wrap(vars(group)) +  
  geom_smooth(method="lm", se=FALSE, fullrange=TRUE) +  
  guides(color=FALSE)
```

```
## `geom_smooth()` using formula 'y ~ x'
```

# DatasauRus

```
datasauRus::datasaurus_dozen
```

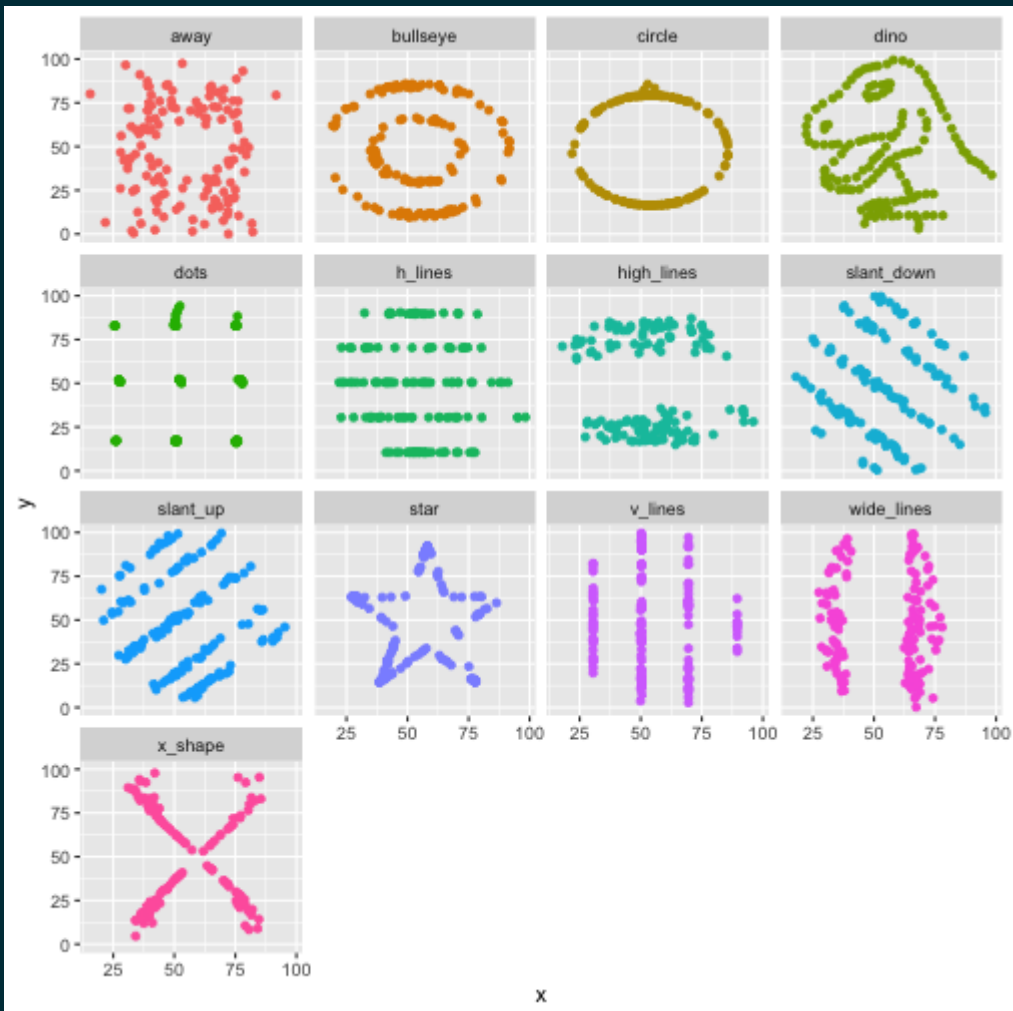
```
## # A tibble: 1,846 x 3
##   dataset      x      y
##   <chr>    <dbl> <dbl>
## 1 dino      55.4  97.2
## 2 dino      51.5  96.0
## 3 dino      46.2  94.5
## 4 dino      42.8  91.4
## 5 dino      40.8  88.3
## 6 dino      38.7  84.9
## 7 dino      35.6  79.9
## 8 dino      33.1  77.6
## 9 dino      29.0  74.5
## 10 dino     26.2  71.4
## # ... with 1,836 more rows
```

```
datasauRus::datasaurus_dozen %>%
  group_by(dataset) %>%
  summarize(mean_x = mean(x), mean_y = mean(y),
            sd_x = sd(x), sd_y = sd(y),
            cor = cor(x,y))
```

```
## `summarise()` ungrouping output (override with `.groups` argument)
```

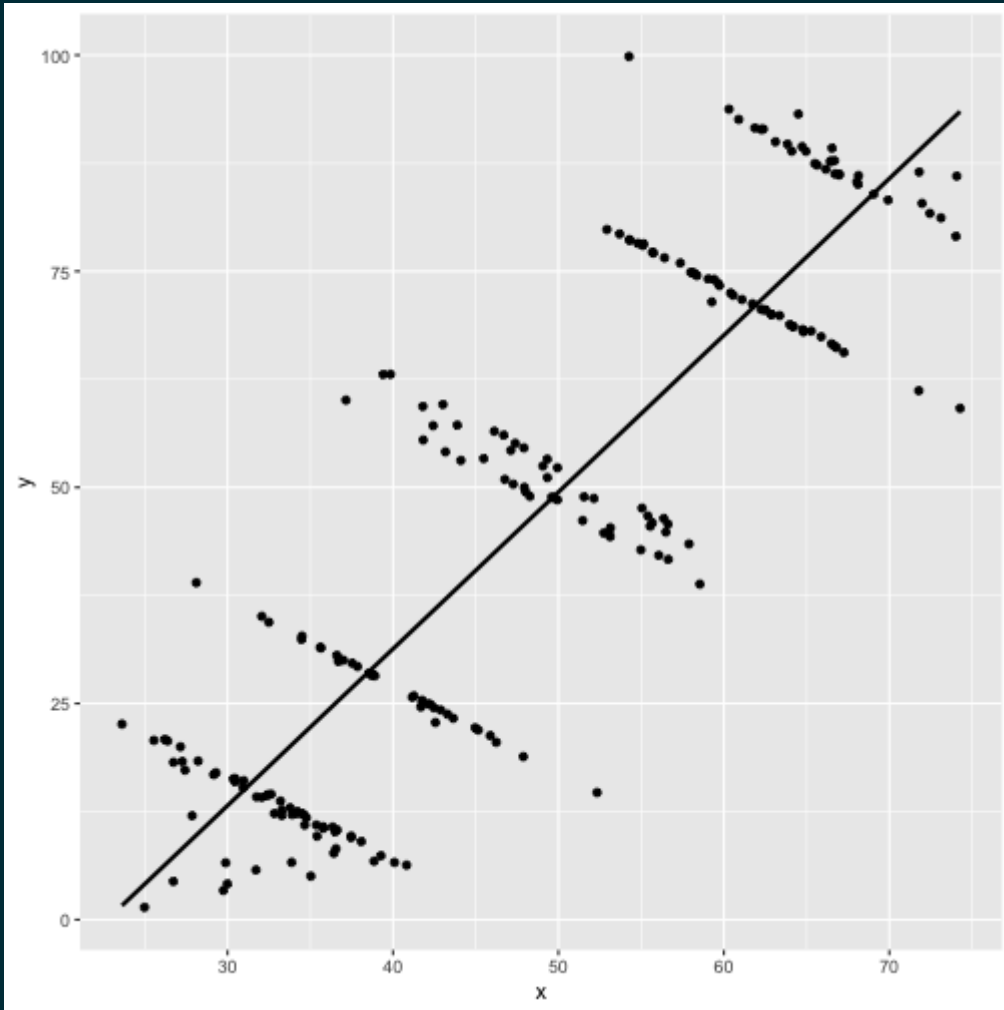
```
## # A tibble: 13 x 6
##   dataset mean_x mean_y sd_x sd_y cor
##   <chr>    <dbl> <dbl> <dbl> <dbl> <dbl>
```

```
ggplot(datasauRus::datasaurus_dozen, aes(x = x, y = y, color = dataset)) +  
  geom_point() +  
  facet_wrap(vars(dataset)) +  
  guides(color=FALSE)
```

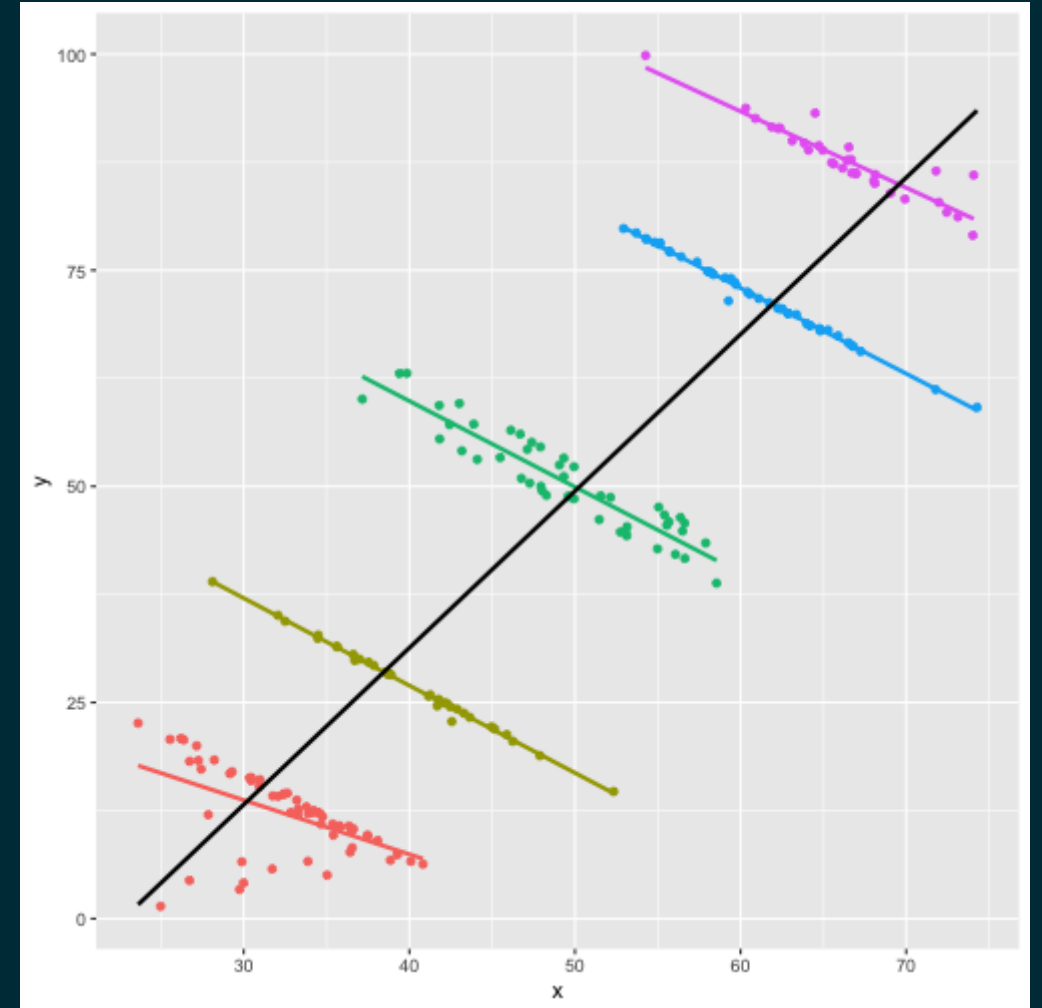


# Simpson's Paradox

```
## `geom_smooth()` using formula 'y ~ x'
```



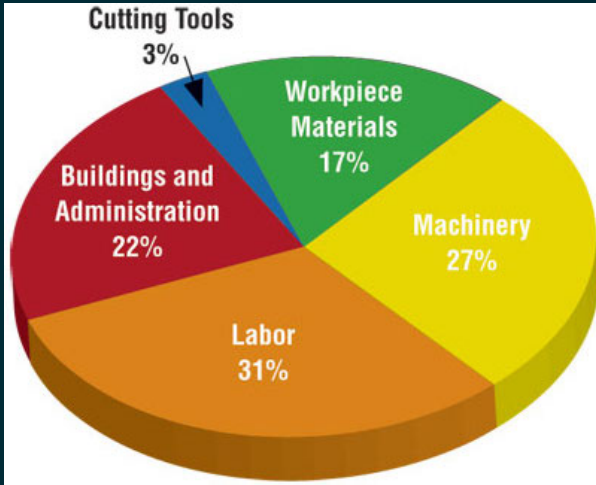
```
## `geom_smooth()` using formula 'y ~ x'  
## `geom_smooth()` using formula 'y ~ x'
```





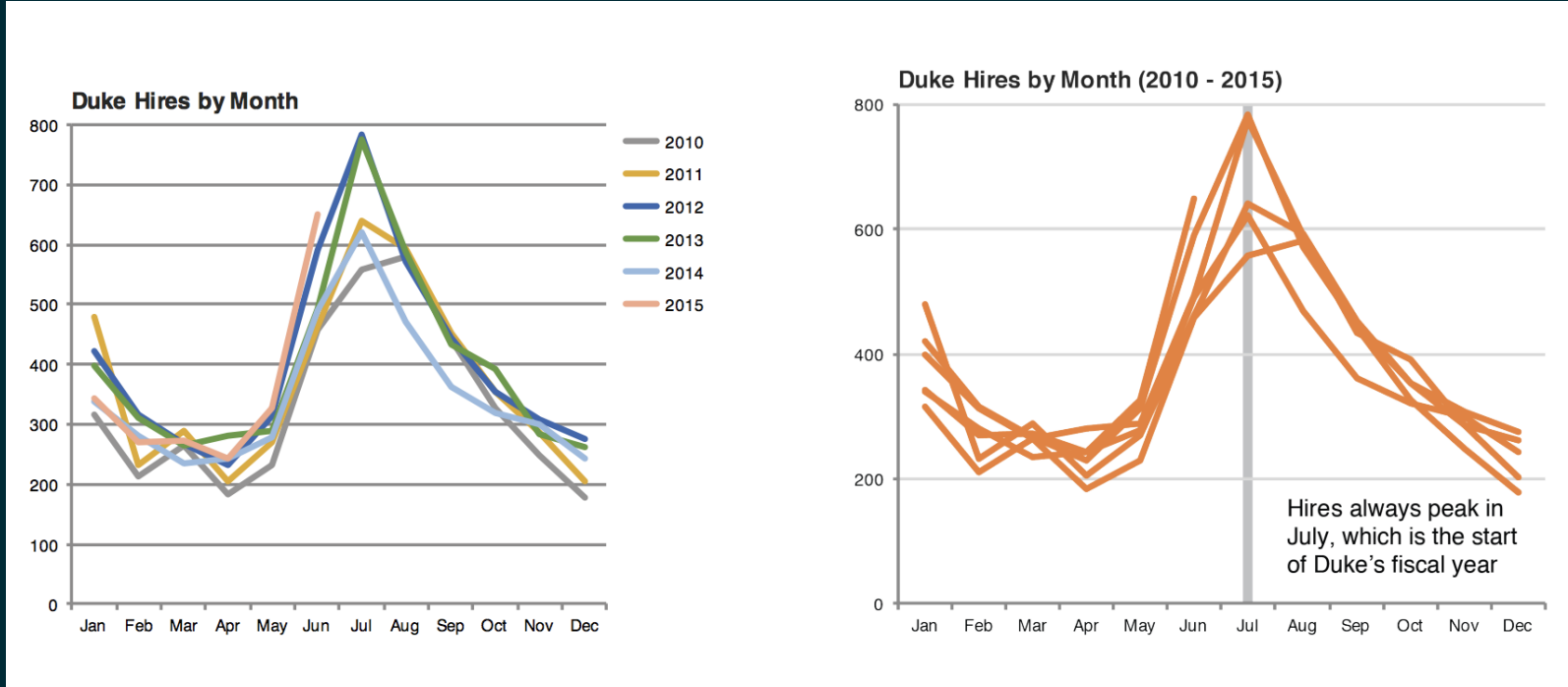
# Designing effective visualizations

# Keep it simple



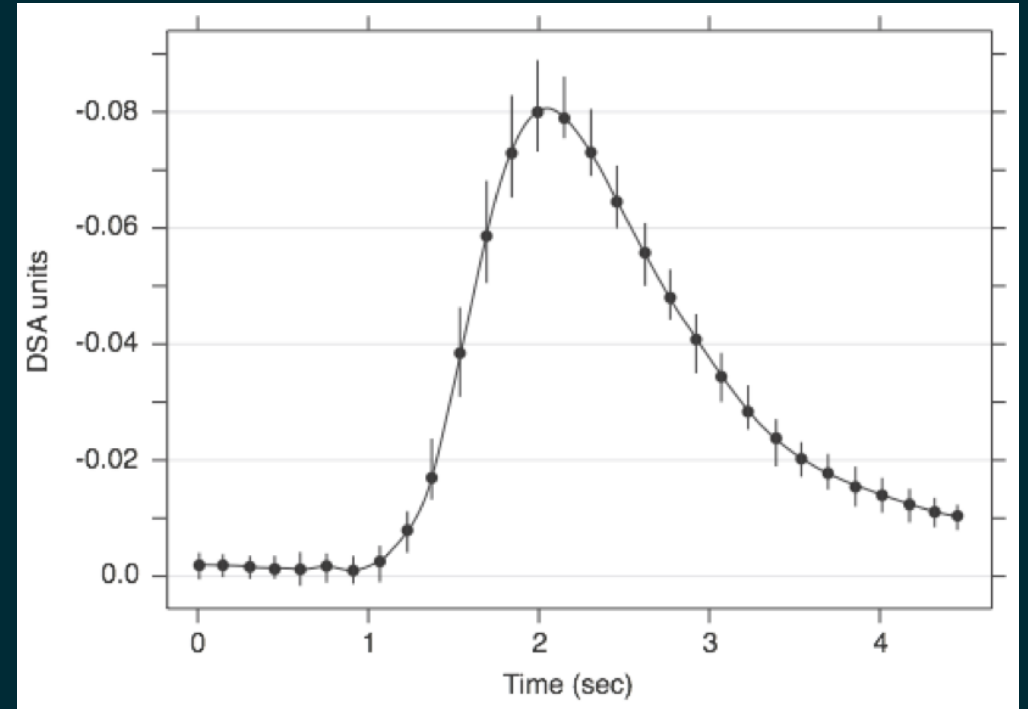
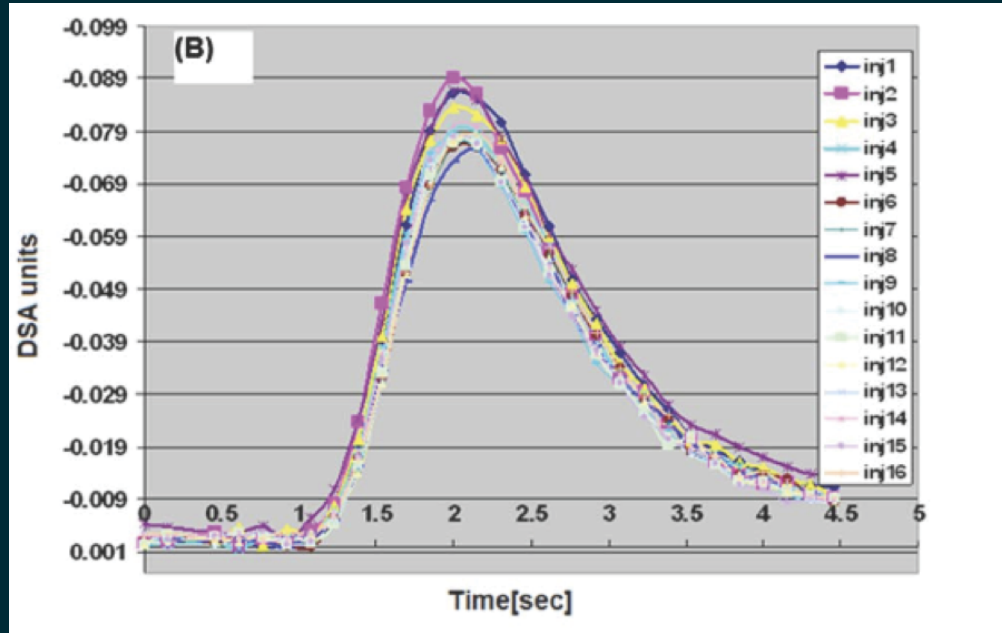
Use color to draw attention

# Tell a story



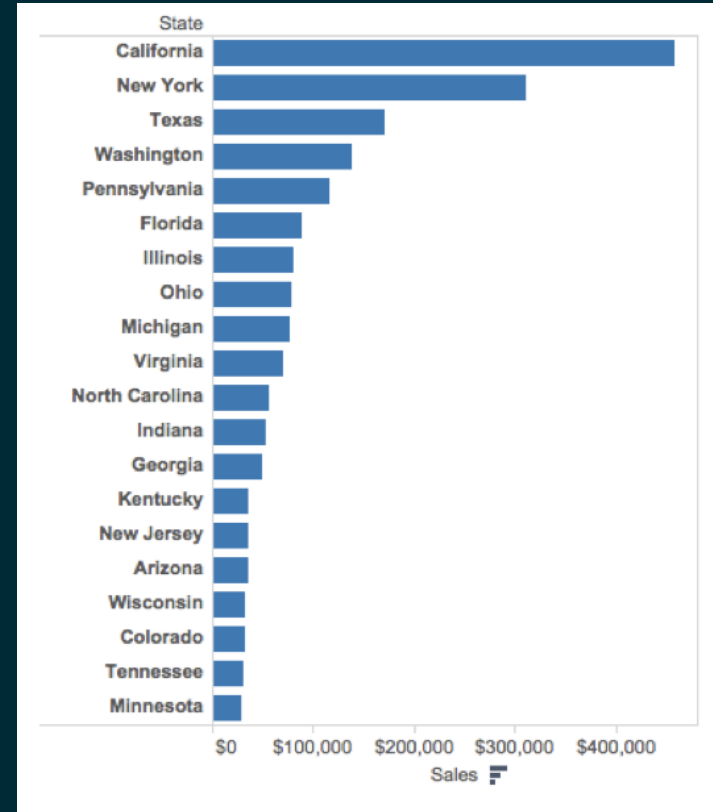
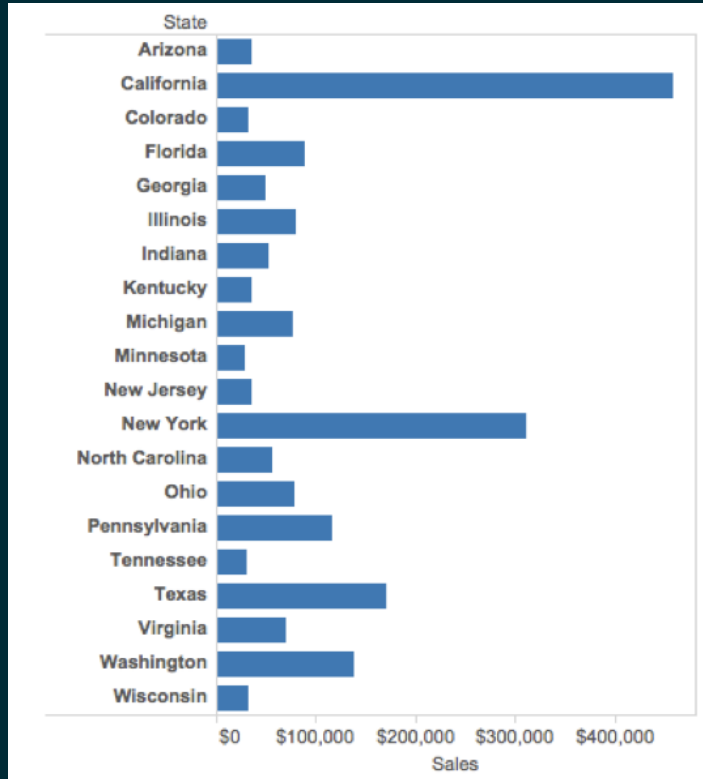
*Credit: Angela Zoss and Eric Monson, Duke DVS*

# Leave out non-story details



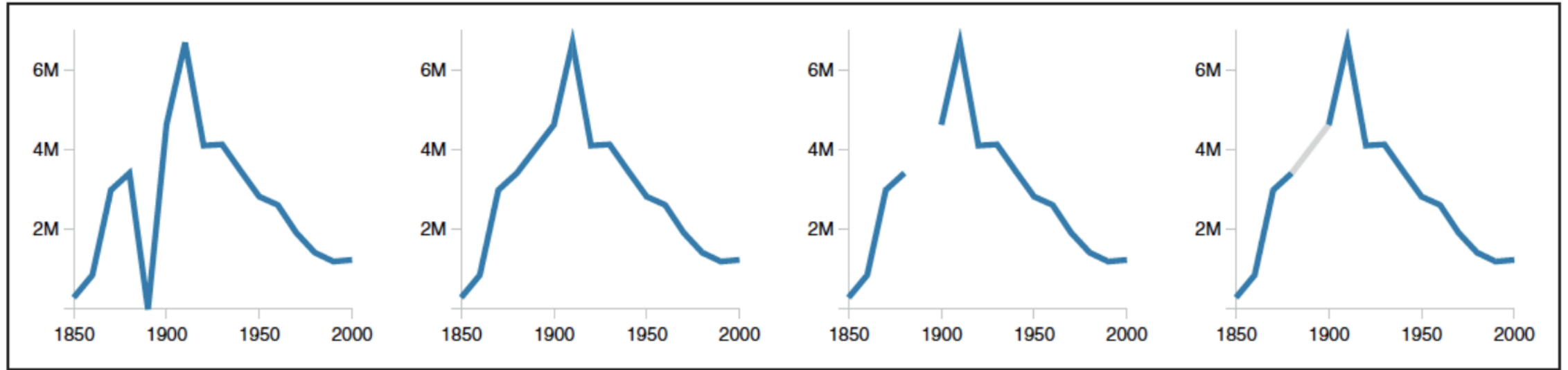
Credit: Angela Zoss and Eric Monson, Duke DVS

# Order / usage matters



*Credit: Angela Zoss and Eric Monson, Duke DVS*

# Be clear about missing data

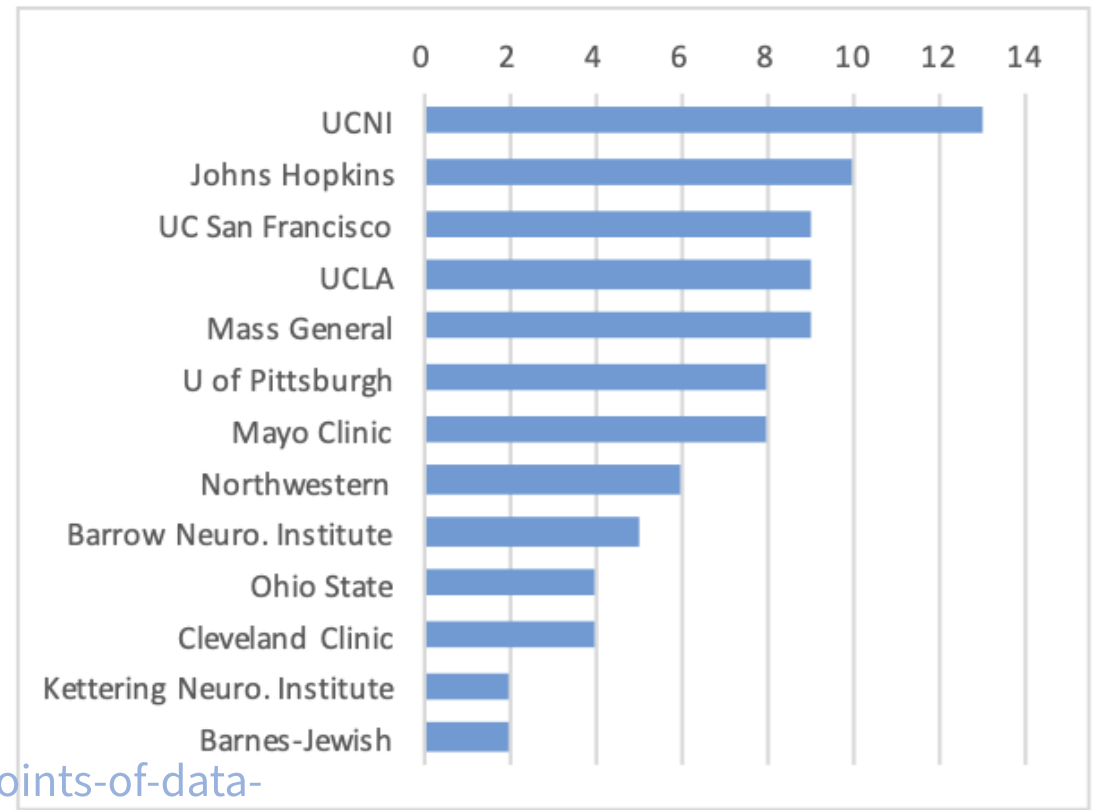
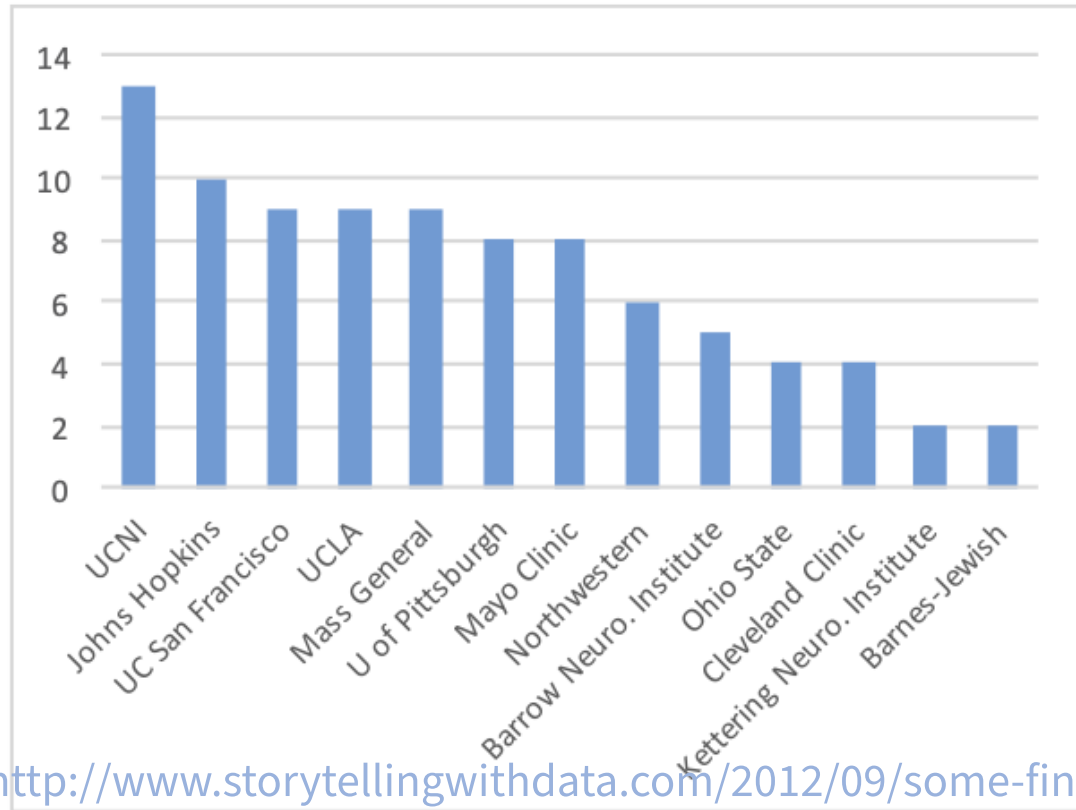


**Figure 4.** Alternative representations of missing data in a line chart. The data are U.S. census counts of people working as 'Farm Laborers'; values from 1890 are missing due to records being burned in a fire. (a) Missing data is treated as a zero value. (b) Missing data is ignored, resulting in a line segment that interpolates the missing value. (c) Missing data is omitted from the chart. (d) Missing data is explicitly interpolated and rendered in gray.

<http://ivi.sagepub.com/content/10/4/271>

Angela Zoss and Eric Monson, Duke DVS

# Reduce cognitive burden



<http://www.storytellingwithdata.com/2012/09/some-finer-points-of-data-visualization.html>



# Use descriptive titles

*Credit:* Angela Zoss and Eric Monson, Duke DVS

# Annotate figures directly

<https://bl.ocks.org/susielu/23dc3082669ee026c552b85081d90976>

# All of the data doesn't tell a story

<http://www.nytimes.com/interactive/2014/06/05/upshot/how-the-recession-reshaped-the-economy-in-255-charts.html>

# All of the data doesn't tell a story

<http://www.nytimes.com/interactive/2014/06/05/upshot/how-the-recession-reshaped-the-economy-in-255-charts.html>

# All of the data doesn't tell a story

<http://www.nytimes.com/interactive/2014/06/05/upshot/how-the-recession-reshaped-the-economy-in-255-charts.html>

# Chart Remakes / Makeovers

# The Why Axis - BLS

<http://thewhyaxis.info/defaults/>

# The Why Axis - Gender Gap

<http://thewhyaxis.info/gap-remake/>



# Acknowledgments

# Acknowledgments

Above materials are derived in part from the following sources:

- Hadley Wickham - R for Data Science & Elegant Graphics for Data Analysis
- ggplot2 website
- Visualization training materials developed by Angela Zoss and Eric Monson, Duke DVS