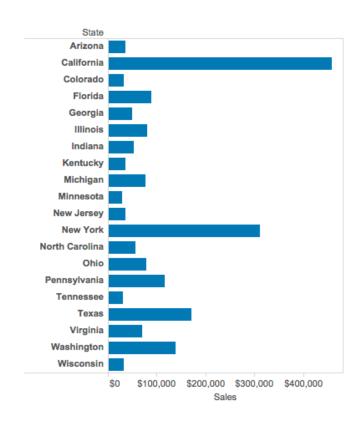
# Tips for Effective Data Visualization

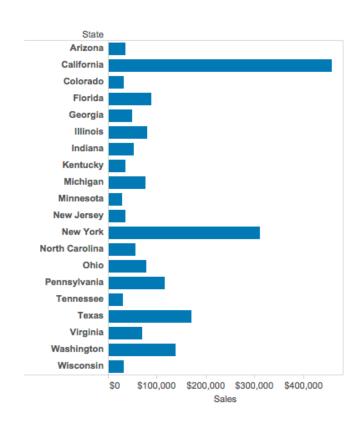
Angela Zoss · Eric Monson
Data and Visualization Services

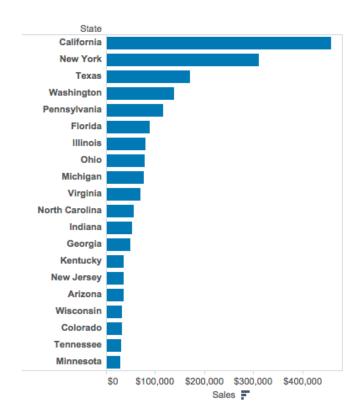
## Common missteps

### Default ordering hides patterns

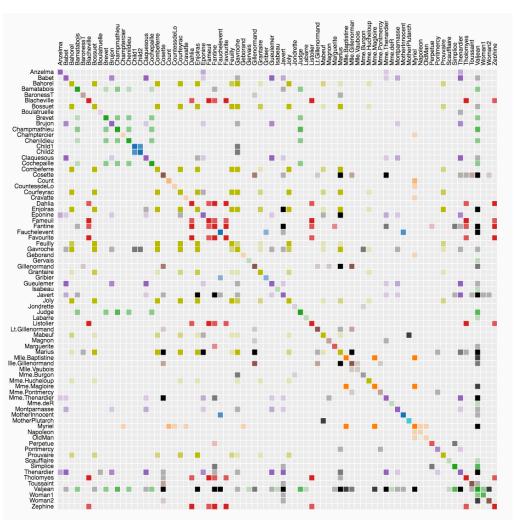


### Sorting reveals patterns



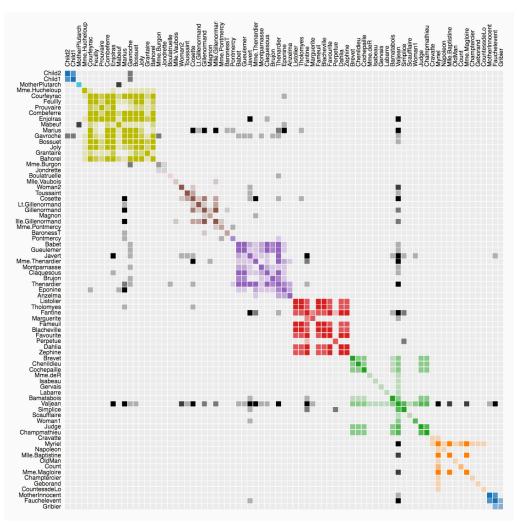


### Default ordering hides patterns



https://bost.ocks.org/mike/miserables/

### Cluster ordering reveals patterns



https://bost.ocks.org/mike/miserables/

Trust	Index			Number of
rank	rank	Borough	Amount approved (£)	grants
1	3	Tower Hamlets	£9,692,642	269
2	2	Hackney	£7,809,608	225
3	12	Southwark	£7,266,118	232
4	14	Camden	£6,140,419	136
5	4	Islington	£5,424,137	156
6	8	Lambeth	£5,257,941	156
7	2	Newham	£5,217,075	154
8	13	Hammersmith and Fulham	£4,085,708	109
9	29	Merton	£3,656,112	113
10	20	Croydon	£3,629,066	127
11	9	Lewisham	£3,537,049	144
12	17	Westminster	£3,357,911	100
13	15	Ealing	£3,057,709	84
14	30	Bromley	£3,038,621	131
15	19	Kensington and Chelsea	£2,979,468	74
16	11	Brent	£2,898,224	85
17	10	Greenwich	£2,837,658	87
18	24	Barnet	£2,796,587	99
19	21	Wandsworth	£2,592,453	89
20	5	Waltham Forest	£2,505,730	131
21	28	Sutton	£2,468,511	87
22	18	Hounslow	£2,383,393	75
23	7	Haringey	£2,360,290	101
24	22	Redbridge	£2,285,173	75
25	33	Rechmond upon Thames	£2,249,983	133
26	23	Hullingdon	£2,181,566	103
27	16	Enfield	£2,145,800	86
28	6	Barking and Dagenham	£1,943,597	68
29	25	Havering	£1,934,424	95
30	26	Bexley	£1,631,415	103
31	27	Harrow	£1,516,193	62
32	31	Kingston upon Thames	£1,353,125	55
33	32	City of London	£402,060	11
		Several Additional Inner		
		Bouroughs	£18,704,677	481
		Several Additional Outer		
		Boroughs	£6,392,100	164
		Other	£28,566,830	566
		London-wide	£86,583,750	1214
		Total	£252,883,123	6180

# Tables easily hide patterns

Total grants spend by London Borough September 1995 to March 2011

http://www.storytellingwithdata.com/blog/2012/02/grables-and-taphs

#### Total grant spend by London Borough

September 1995 - March 2011

Borough	Trust rank	Index rank		Amount approved (£)
Tower Hamlets	1	3	269	9.692.642
Hackney	2	2	225	7,809,608
Southwark	3	12	232	7.266.118
Camden	4	14	136	6,140,419
Islington	5	4	156	5,424,137
Lambeth	6	8	156	5,257,941
Newham	7	2	154	5.217.075
Hammersmith and Fulham	8	13	109	4,085,708
Merton	9	29	113	3,656,112
Croydon	10	20	127	3,629,066
Lewisham	11	9	144	3,537,049
Westminster	12	17	100	3,357,911
Ealing	13	15	84	3.057,709
Bromley	14	30	131	3,038,621
Kensington and Chelsea	15	19	74	2,979,468
Brent	16	11	85	2.898.224
Greenwich	17	10	87	2.837.658
Barnet	18	24	99	2,796,587
Wandsworth	19	21	89	2,592,453
Waltham Forest	20	5	131	2.505,730
Sutton	21	28	87	2,468,511
Hounslow	22	18	75	2.383,393
Haringey	23	7	101	2.360,290
Redbridge	24	22	75	2.285.173
Rechmond upon Thames	25	33	133	2.249.983
Hullingdon	26	23	103	2.181.566
Enfield	27	16	86	2.145,800
Barking and Dagenham	28	6	68	1,943,597
Havering	29	25	95	1.934.424
Bexley	30	26	103	1,631,415
Harrow	31	27	62	1.516.193
Kingston upon Thames	32	31	55	1.353 125
City of London	33	32	11	02,060
Several Additional Inner Bouroughs			481	18,704,677
Several Additional Outer Boroughs			164	6,392,100
Other			566	28,566,830
London-wide			1,214	86,583,750

Help people see patterns in tables

http://www.storytellingwithdata.com/blog/2012/02/grables-and-taphs

### Help viewers interpret tables

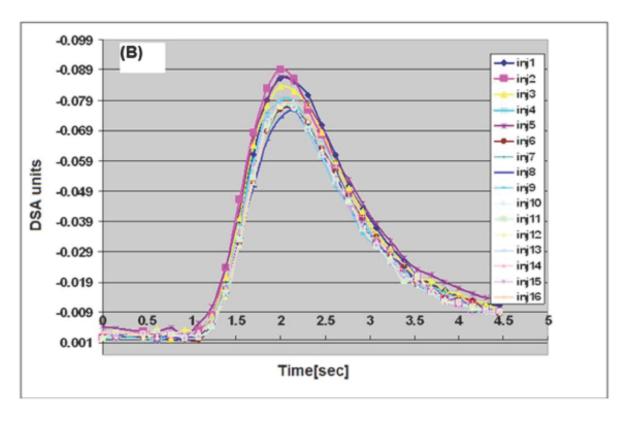
- Limit and standardize decimal places
- Emphasize important values with color, bold text and annotations
- Sort rows by values
- Turn table into another chart or a handout

### MDG 4 Progress 1990-2013: Low Income Countries

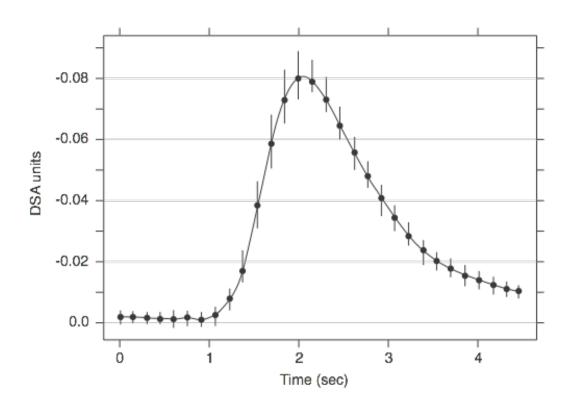
Country	1990 (Per 1,000)	2013 (per 1,000)	Percent Change	2015 Target	Percent of Target Met
Afghanistan	17	9 97	46%	60.9	69%
Bangladesh	14	4 41	. 72%	49	108%
Benin	17	9 85	53%	60.9	80%
Burkina Faso	20	2 98	51%	68.7	78%
Burundi	17	1 83	51%	58.1	78%
Cambodia	11	8 38	68%	40.1	103%
Central African Republic	17	7 139	21%	60.2	33%
Chad	21	5 148	31%	73.1	47%
Comoros	12	5 78	38%	42.5	57%
Congo, Democratic Republic of	17	6 119	32%	59.8	49%
Eritrea	15	1 50	67%	51.3	101%
Ethiopia	20	5 64	69%	69.7	104%
Gambia, The	17	0 74	56%	57.8	86%
Guinea	23	8 101	. 58%	80.9	87%
Guinea-Bissau	22	5 124	45%	76.5	68%
Haiti	14	5 73	50%	49.3	75%
Kenya	9	9 71	. 28%	33.7	43%
Korea, DPR	4	3 27	37%	14.6	56%
Liberia	24	8 71	. 71%	84.3	108%
Madagascar	16	1 56	65%	54.7	99%
Malawi	24	5 68	72%	83.3	109%
Mali	25	4 123	52%	86.4	78%
Mozambique	23	7 87	63%	80.6	96%
Myanmar	10	9 51	. 53%	37.1	81%
Nepal	14	2 40	72%	48.3	109%
Niger	32	7 104	68%	111.2	103%
Rwanda	15	2 52	66%	51.7	99.70%
Sierra Leone	26	8 161	. 40%	91.1	60%
Somalia	18	0 146	19%	61.2	29%
Tajikistan	10	8 48	56%	39.8	84%
Tanzania	16	7 52	69%	56.8	104%
Togo	14	6 85	42%	49.6	63%
Uganda	17	9 66	63%	60.9	96%
Zimbabwe	7	5 89	-19%	25.5	-28%

https://dpt.duhs.duke.edu/files/Group%2011.pdf

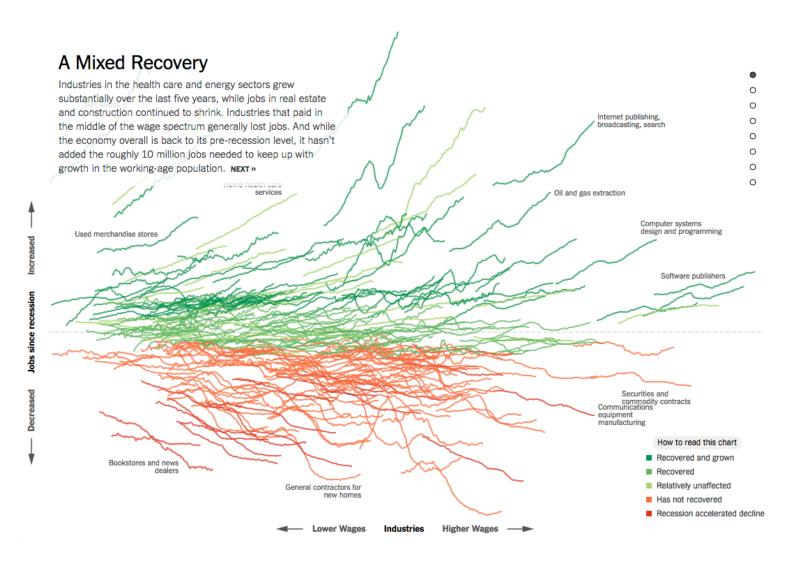
### Leave out non-story details



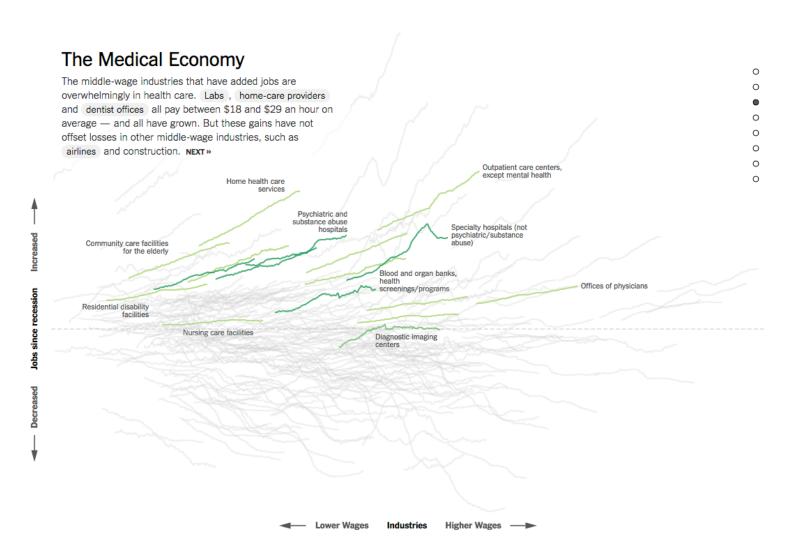
### Leave out non-story details



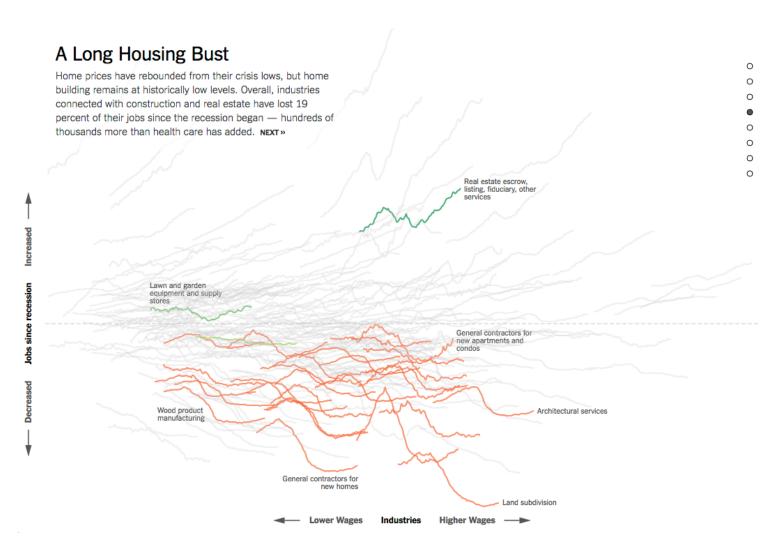
### All the data doesn't tell a story



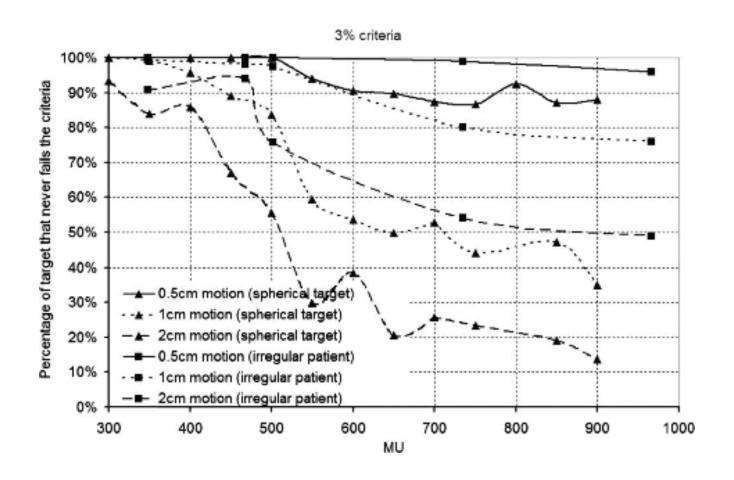
### All the data doesn't tell a story



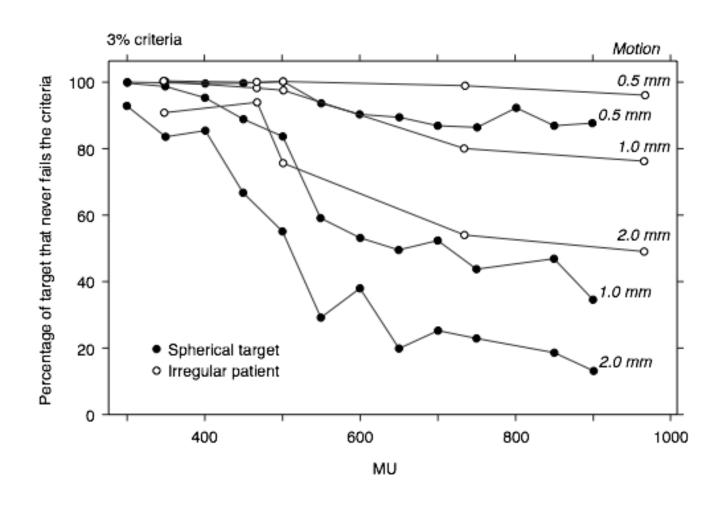
### All the data doesn't tell a story



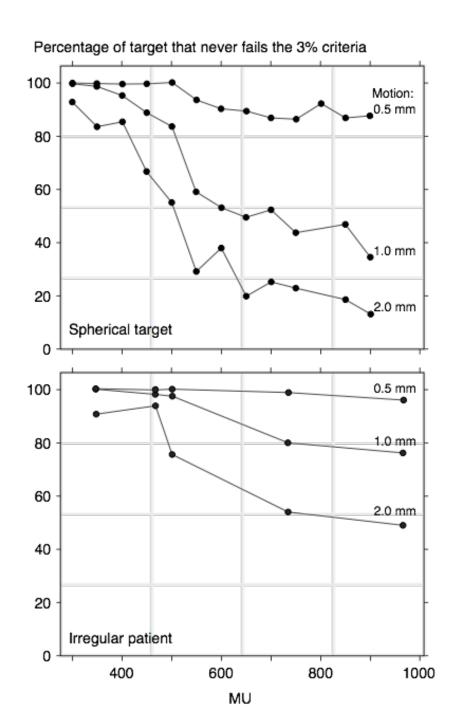
### Original



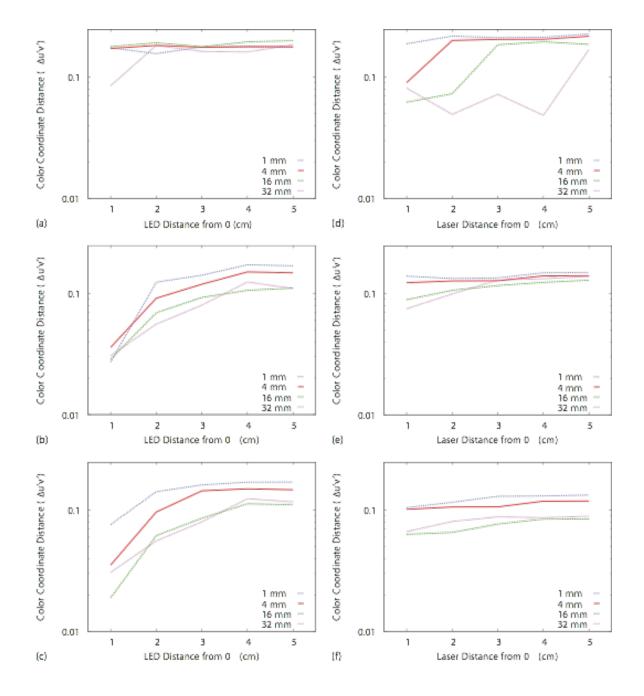
### Reworked as single plot



# Reworked as small multiples

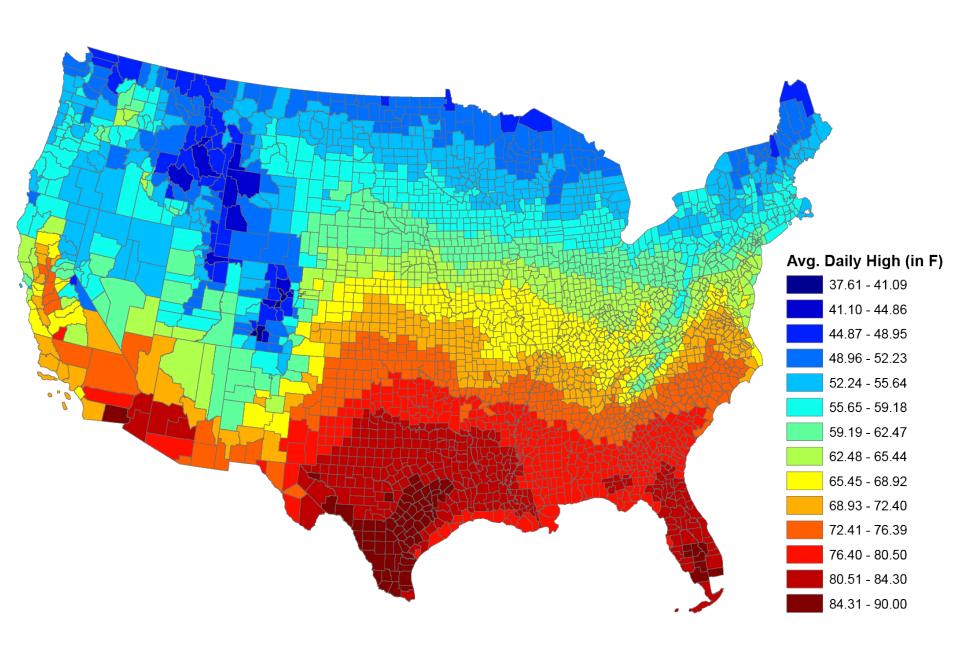


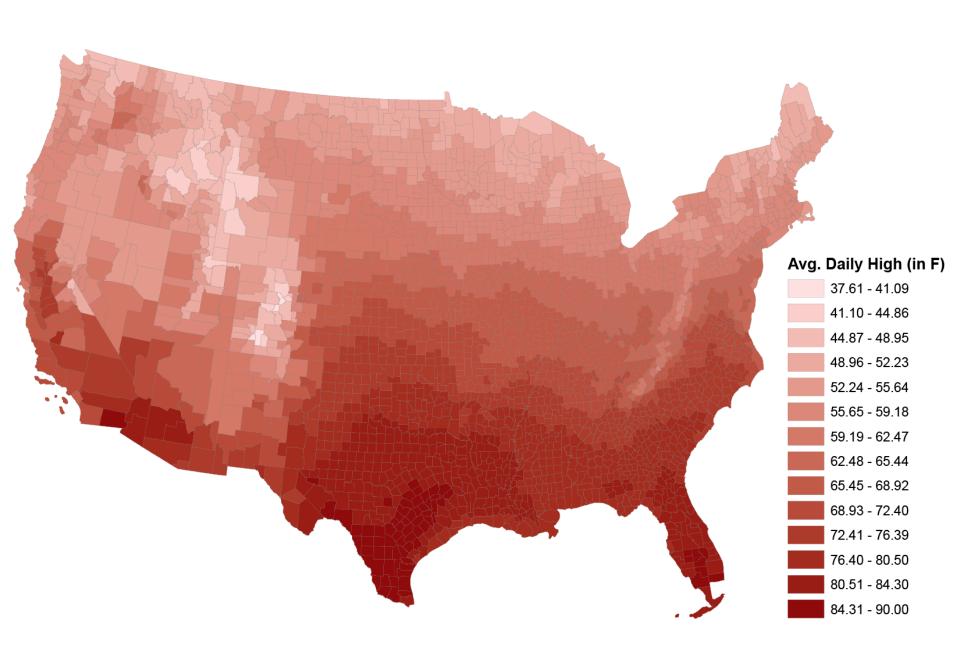
## Original



#### Reworked (a) (d) 0.01 Color Coordinate Distance $(\Delta u'v')$ 0.1 (b) (e) 0.01 0.1 1 mm 4 mm (c) (f) - 16 mm - 32 mm 0.01 2 3 5 2 3 5 LED distance from 0 cm Laser distance from 0 cm

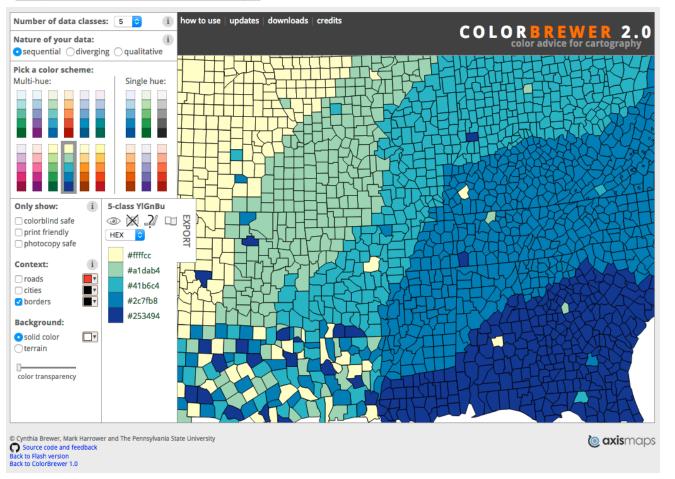
### Reworked (a) (d) 0.01 Color Coordinate Distance (∆u'v') - 0.1 (b) (e) 0.01 0.1 (f) (c) - 16 mm - 32 mm 0.01 2 3 5 2 3 5 LED distance from 0 cm Laser distance from 0 cm





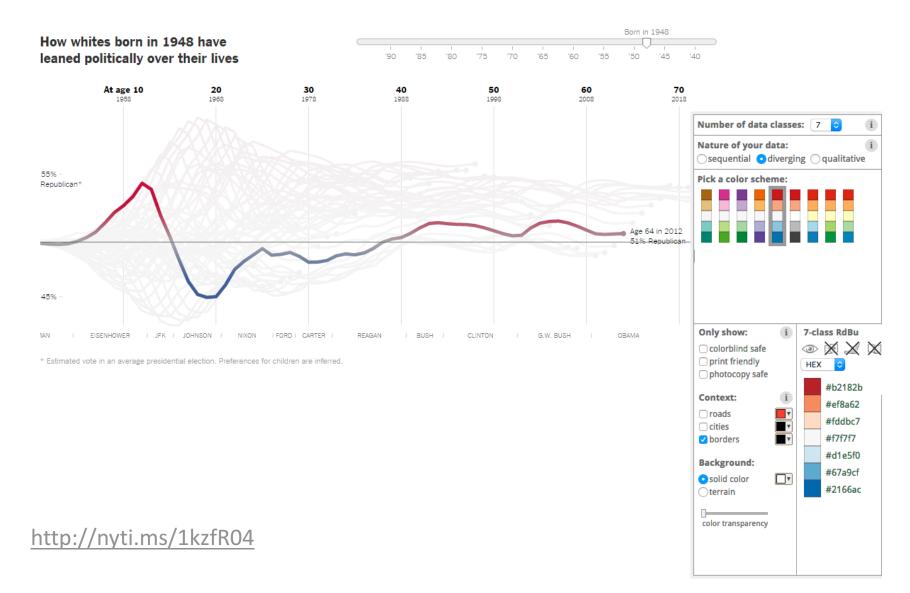
### ColorBrewer for good colormaps

http://colorbrewer2.org/



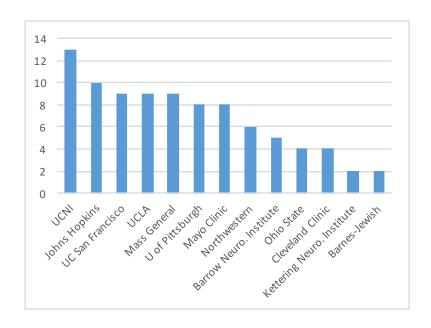


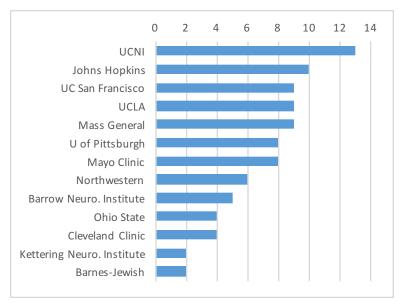
### Do data have a natural center?



## Text to clarify

### Keep text horizontal



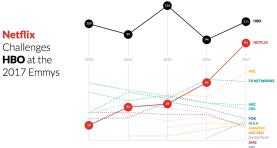


http://www.storytellingwithdata.com/2012/09/some-finer-points-of-data-visualization.html

### Annotate figures directly

AAPL stock example





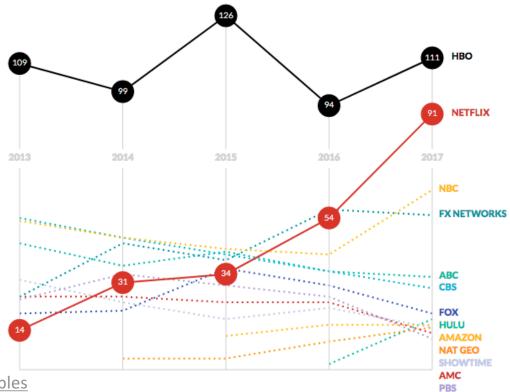
**Netflix** Challenges **HBO** at the

### Annotate figures directly



#### **Netflix**

Challenges **HBO** at the
2017 Emmys

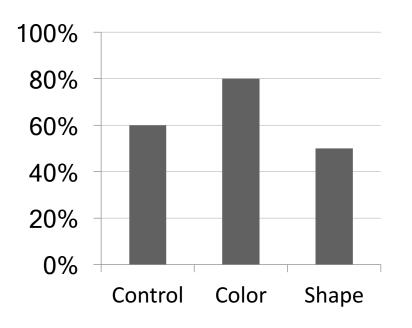


http://d3-annotation.susielu.com/#examples

### Use descriptive titles

Active titles summarize trends in the figure and reinforce your message.

# **Accuracy versus Color and Shape**



# Accuracy Improved by Color, not Shape

