

Intro to comparing distributions

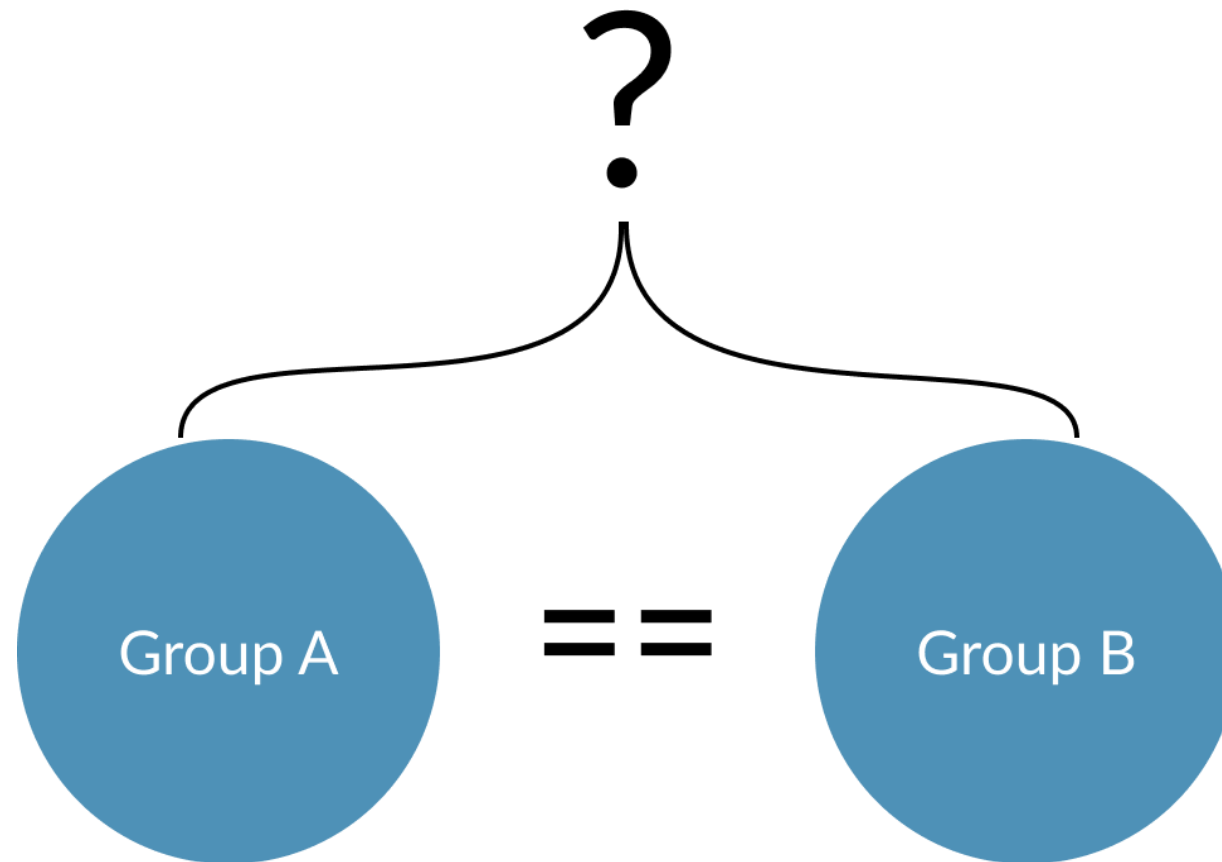
VISUALIZATION BEST PRACTICES IN R



Nick Strayer
Instructor

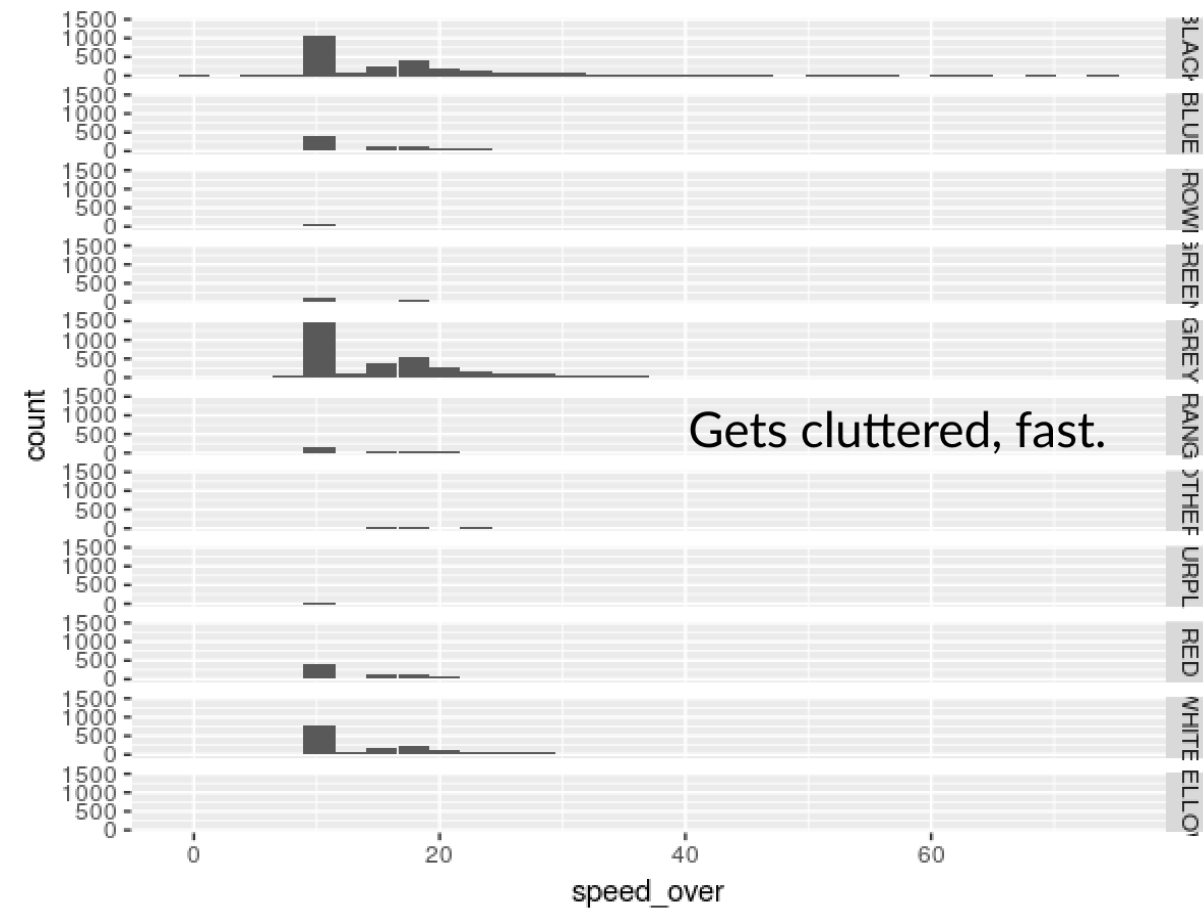
Why compare distributions?

- Verify balanced groups
- For comparison's sake

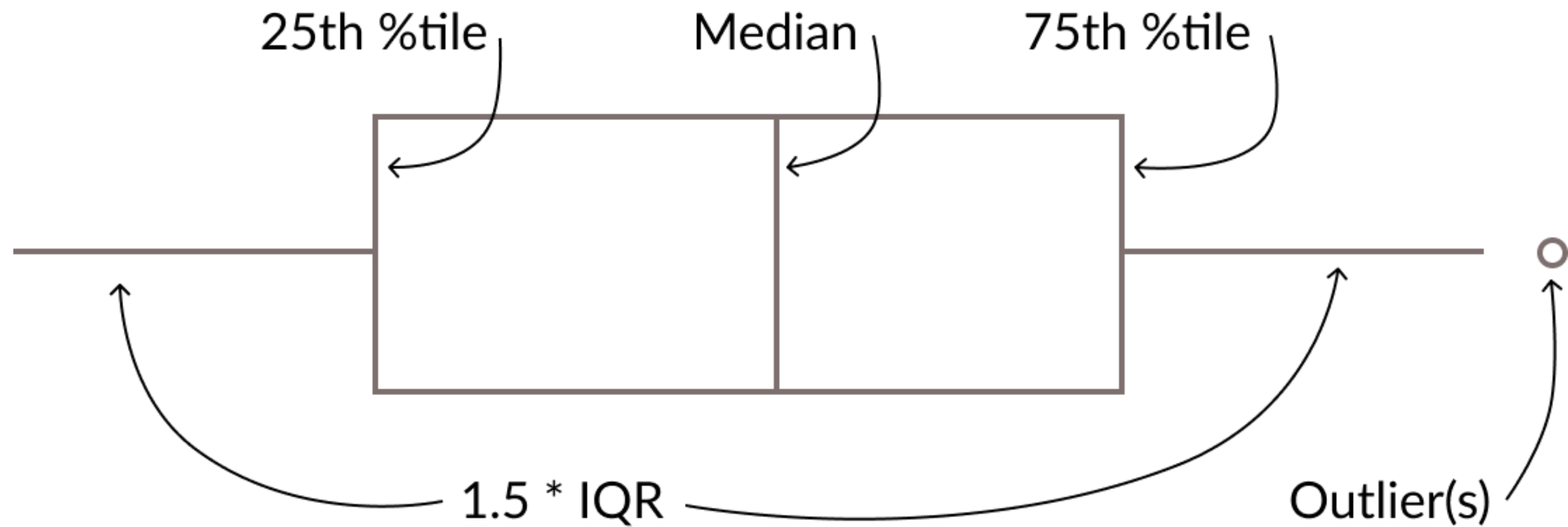


Why not facet histograms?

```
ggplot(md_speeding, aes(x = speed_over)) +  
  geom_histogram() +  
  facet_grid(vehicle_color ~ .)
```

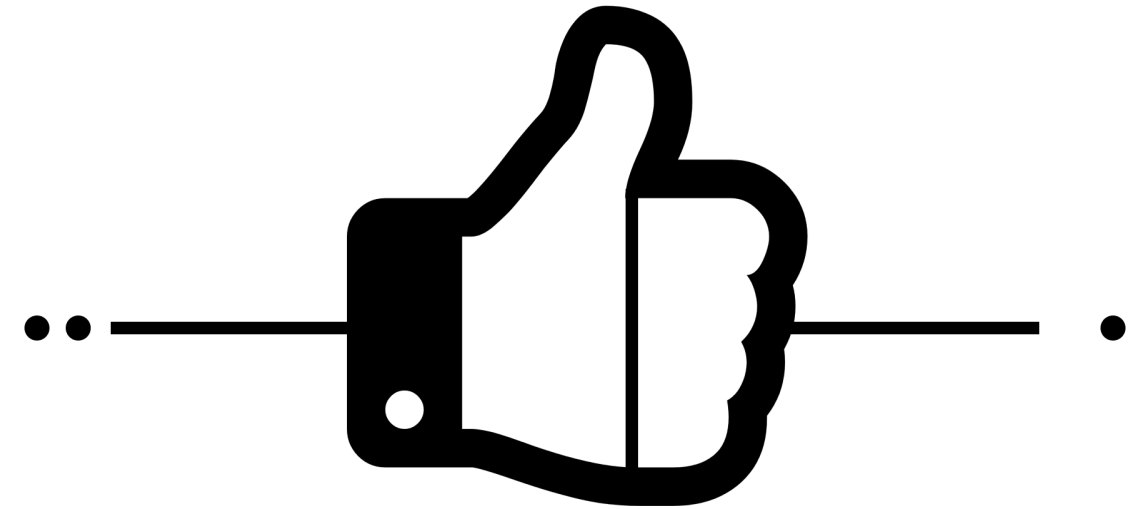


The boxplot



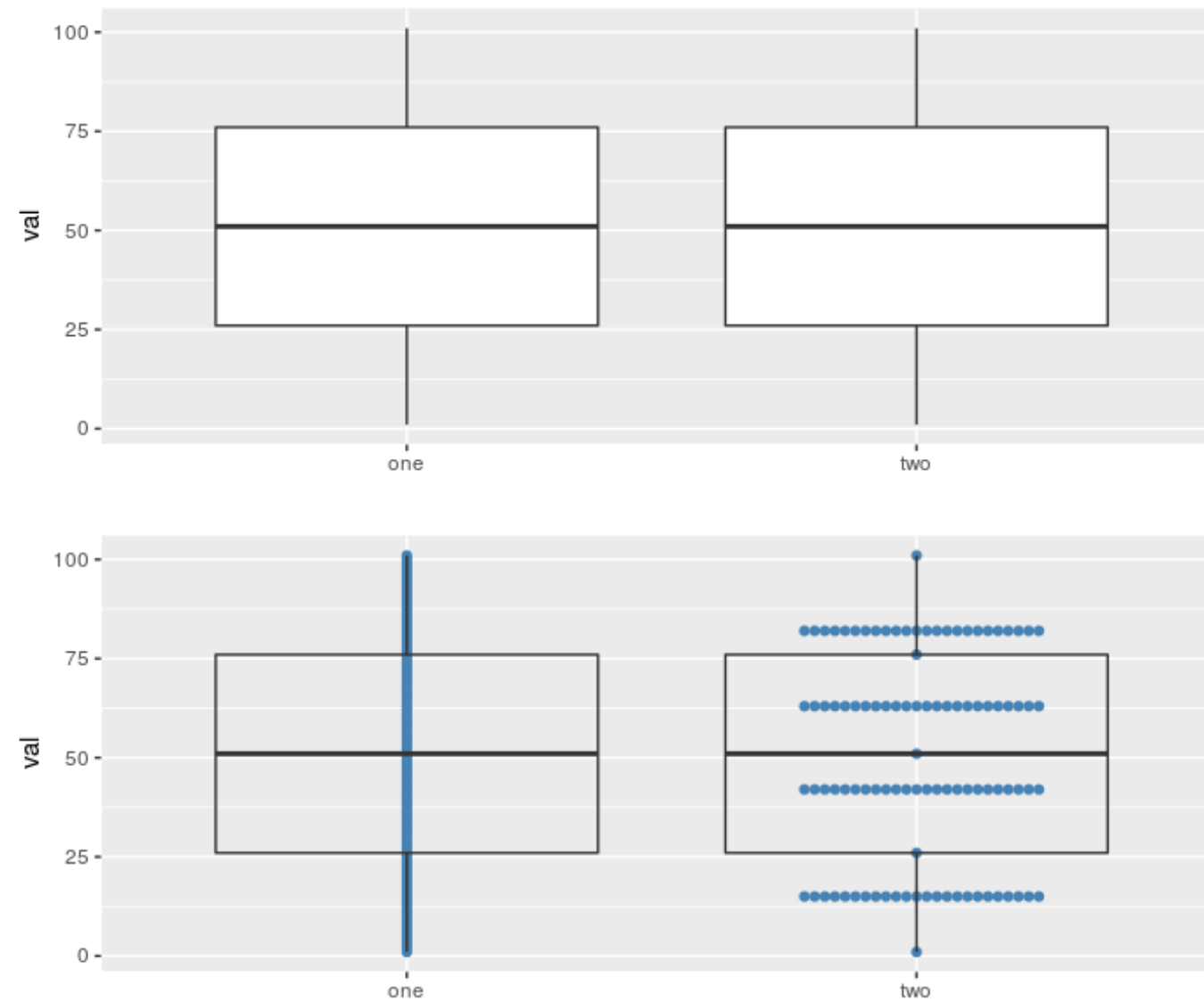
Boxplot pros

- Familiar
- Lots of good summary statistics



boxplot cons

- Show me the data!

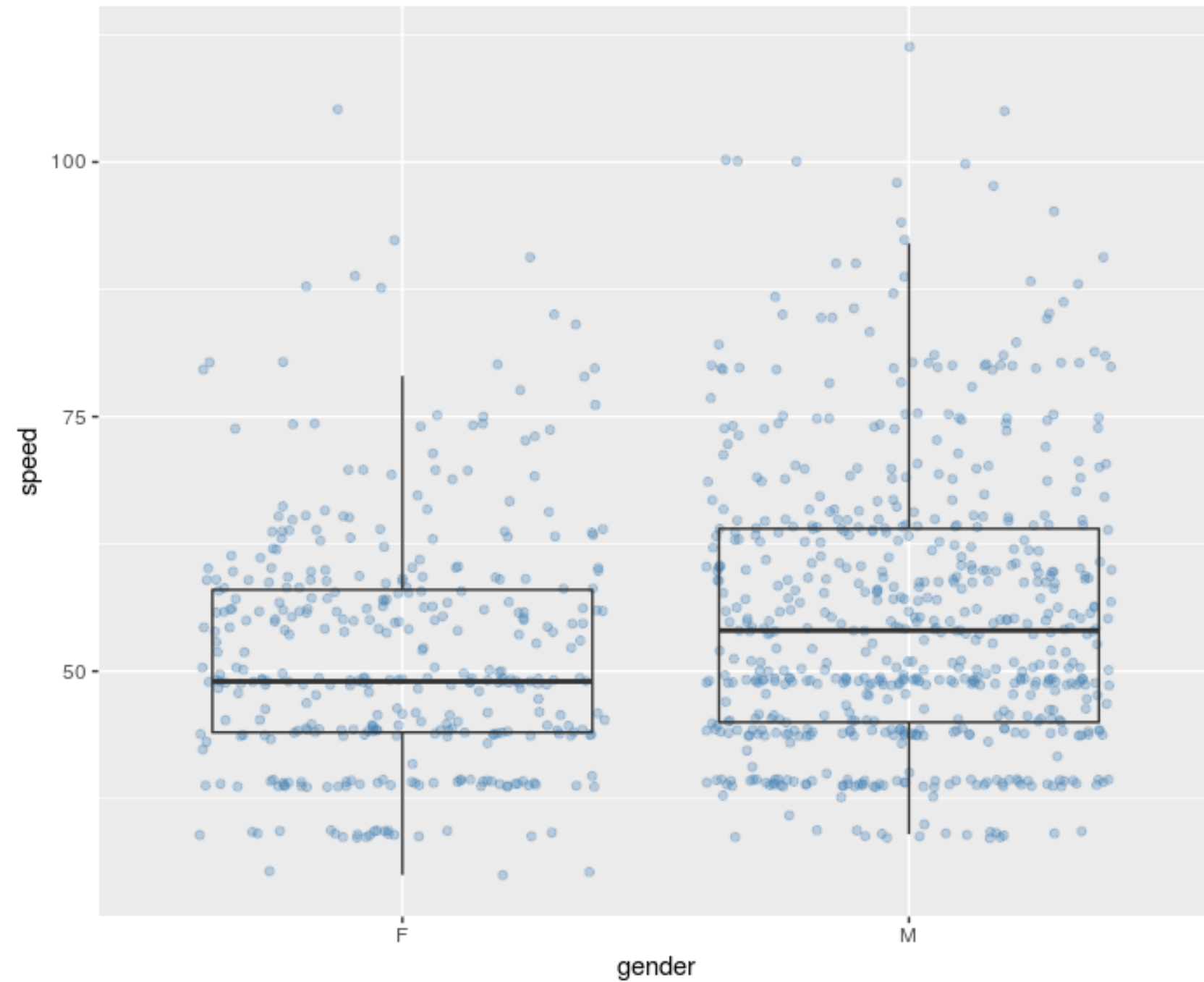


A simple addition

- `geom_jitter()` shows raw points jostled to avoid overlap.
- Layer under your `geom_boxplot()`.

```
md_speeding %>%  
  filter(vehicle_color == 'BLUE') %>%  
  ggplot(aes(x = gender, y = speed)) +  
    # Draw points behind  
    geom_jitter(alpha = 0.3, color = 'steelblue') +  
    # Make transparent  
    geom_boxplot(alpha = 0) +  
    labs(title = 'Distribution of speed for blue cars by gender')
```

Distribution of speed for blue cars by gender



Let's compare some distributions!

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Boxplot alternatives

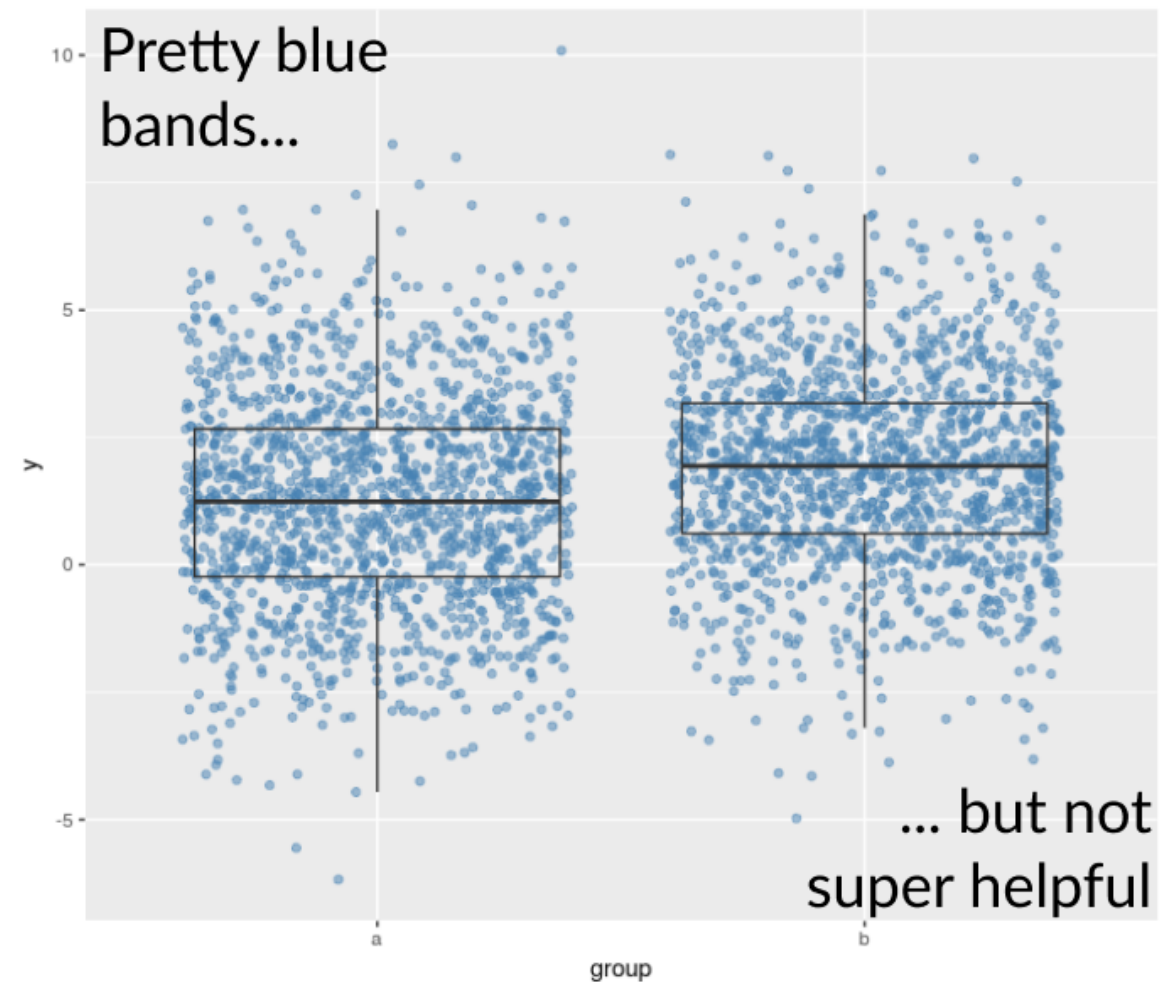
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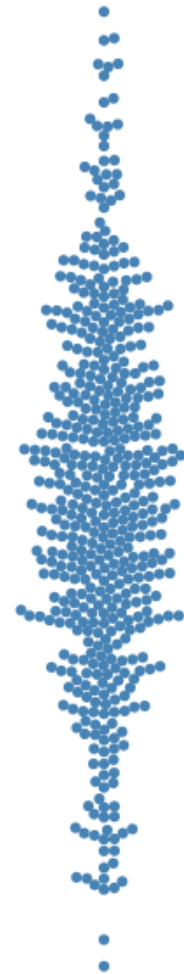
Limitations of the boxplot with jitter

- Jostling points can only deal with so much overlap
- Hard to get an idea of data density



What are some other options?

Beeswarm plots



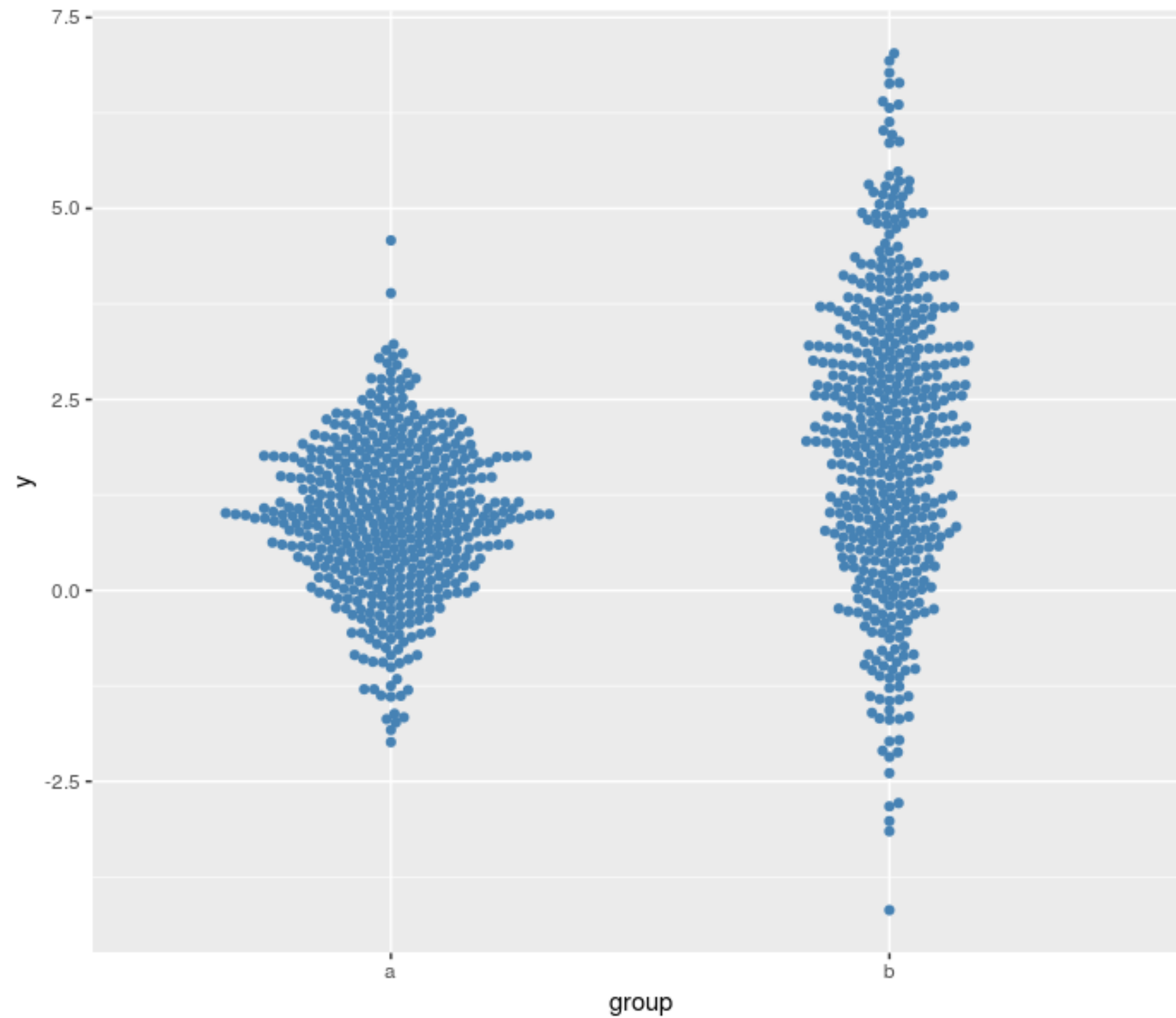
Violin plots



Beeswarm plots

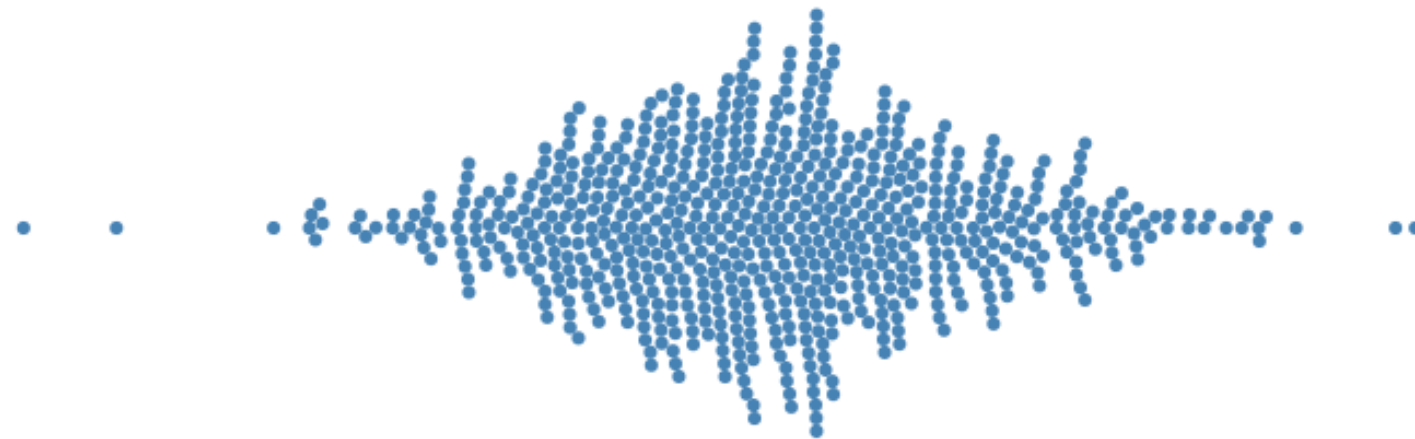
- 'Smart' jittering
- Individual points are clumped together as close to the axis as possible
- Handily included as `geom_beeswarm()` in the `ggbeeswarm` package.

```
library(ggbeeswarm)
ggplot(data, aes(y = y, x = group)) +
  geom_beeswarm(color = 'steelblue')
```



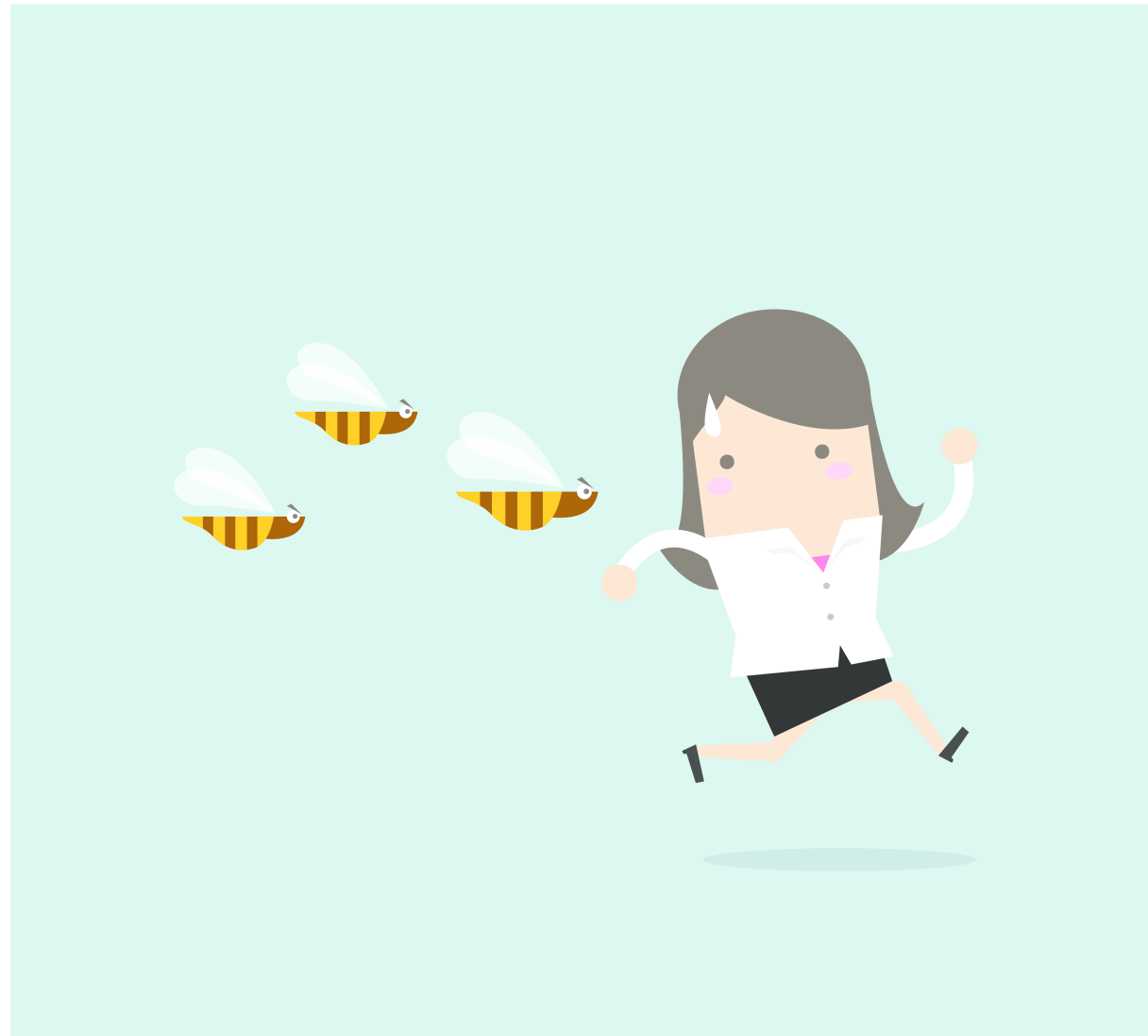
Beeswarm pros

- Individual data points
- Distributional shape



Beeswarm cons

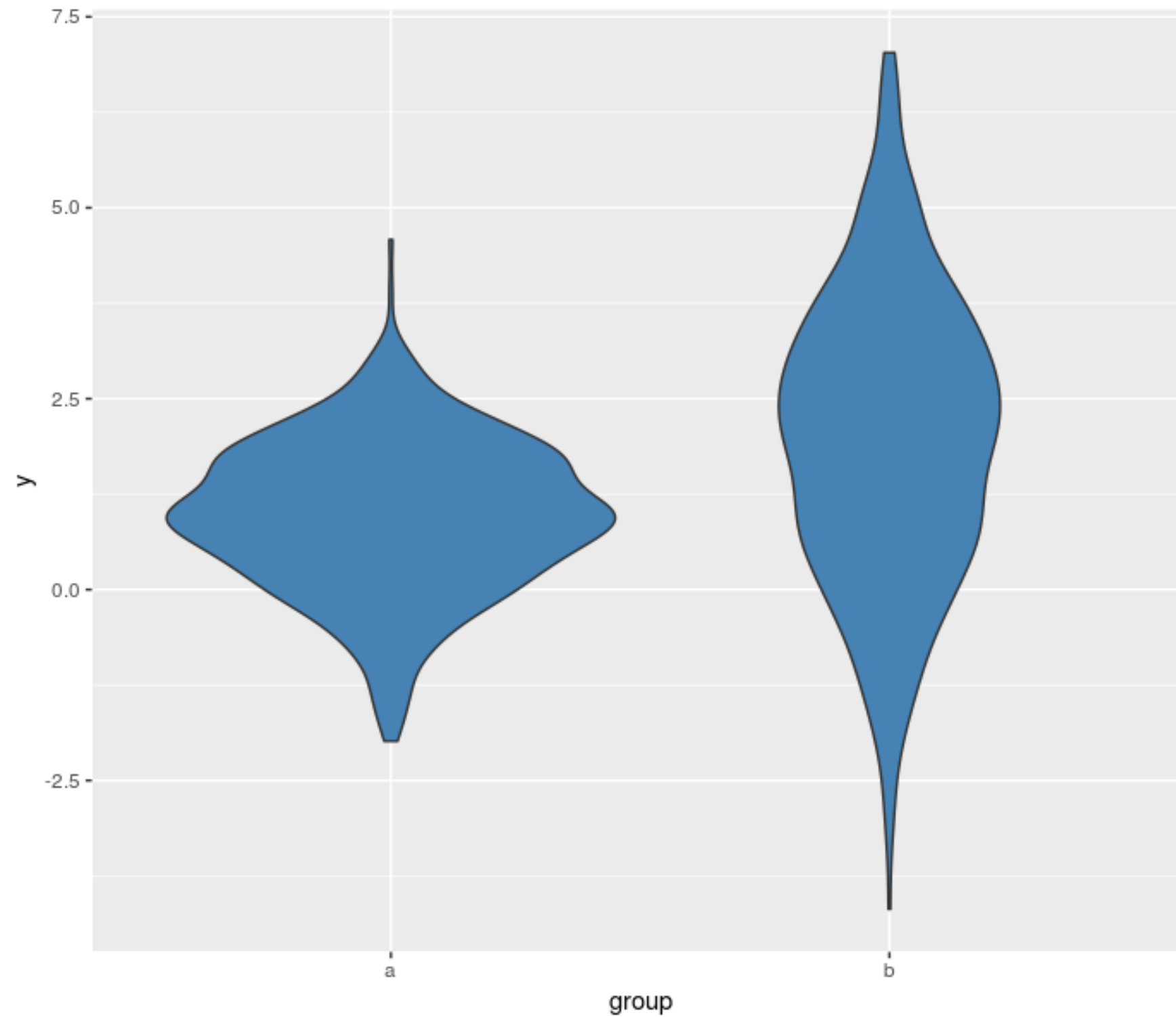
- Get hard with lots of data
- Arbitrary stacking



Violin plots

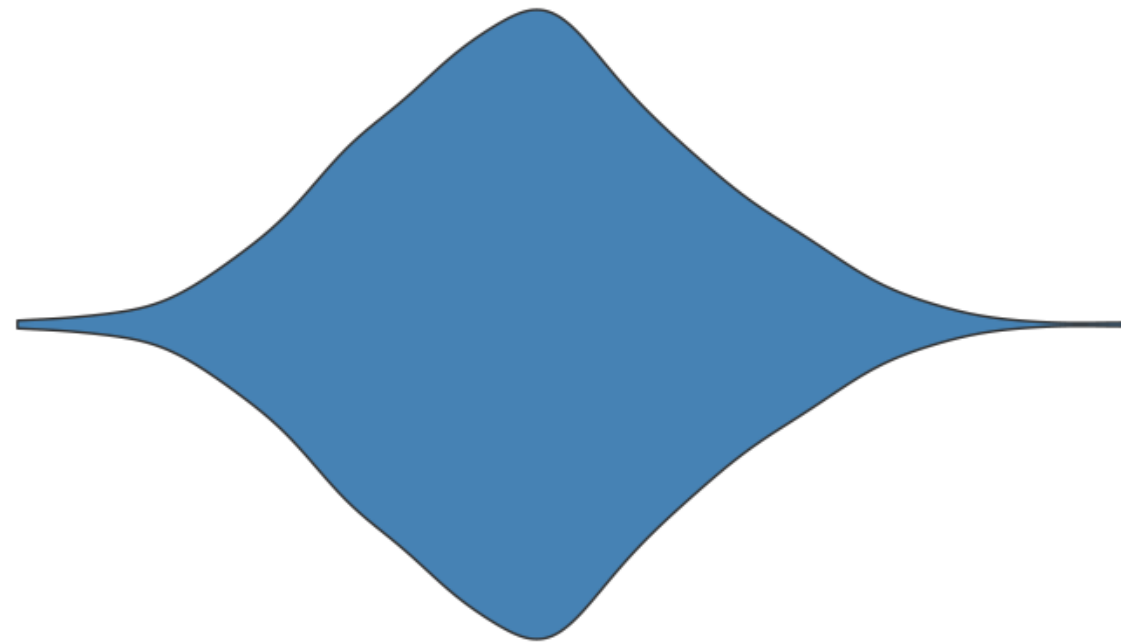
- KDE reflected to be symmetric
- Just replace `geom_boxplot()` with `geom_violin()` .

```
ggplot(data, aes(y = y, x = group)) +  
  geom_violin(fill = 'steelblue')
```



Violin pros

- Every data point is heard
- Not every data point is seen, so good for lots of data.



Violin cons

- Kernel width choice
- Not every data point is seen



**Let's try some more
advanced
comparisons!**

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Comparing spatially-related distributions

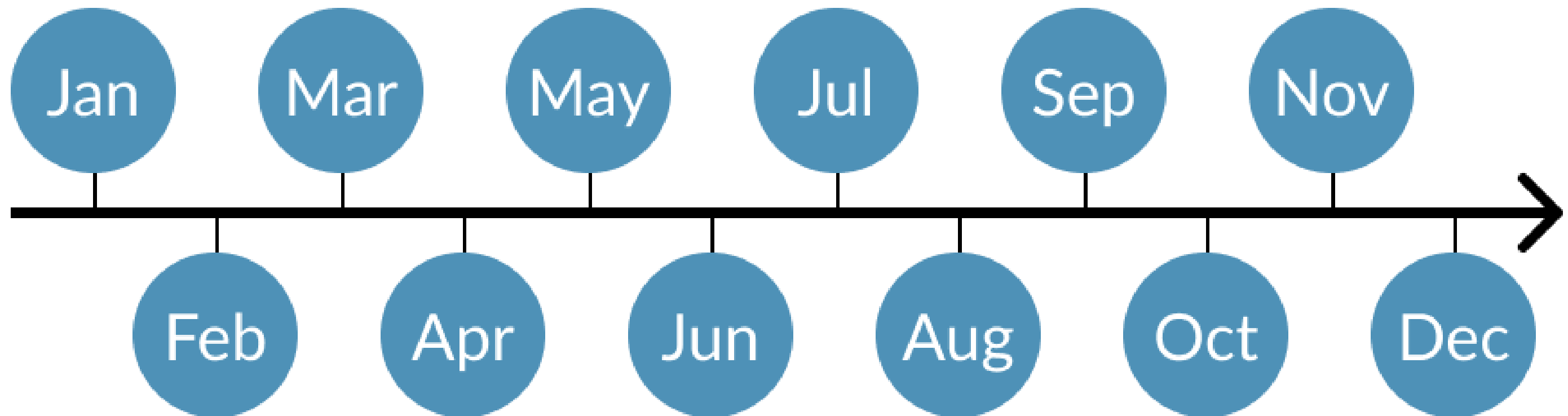
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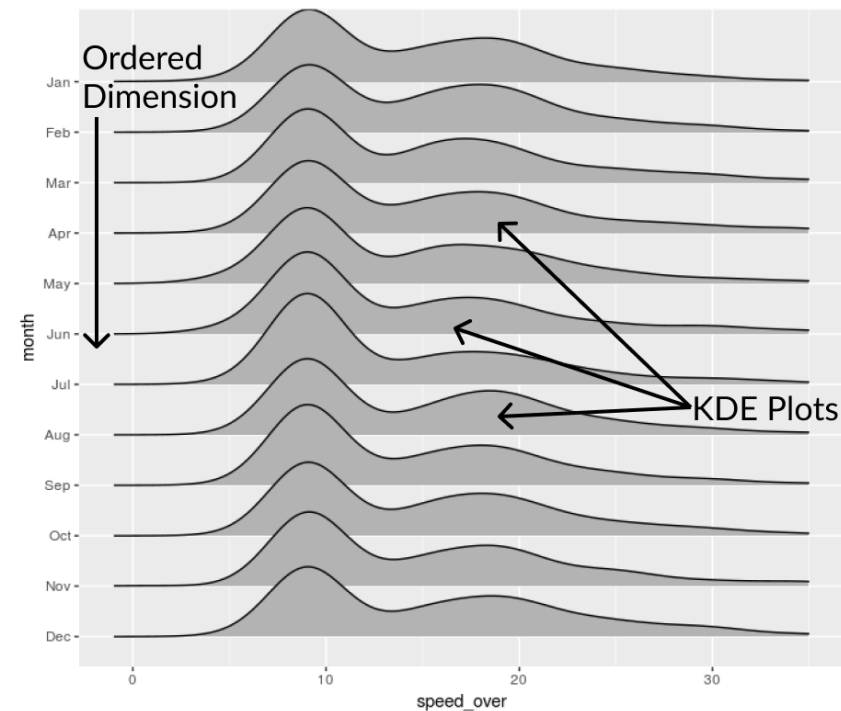
What are 'spatially connected axes'?

- There is an underlying ordering of the classes.
- E.g. months of the year: Jan < Feb < Mar < ...

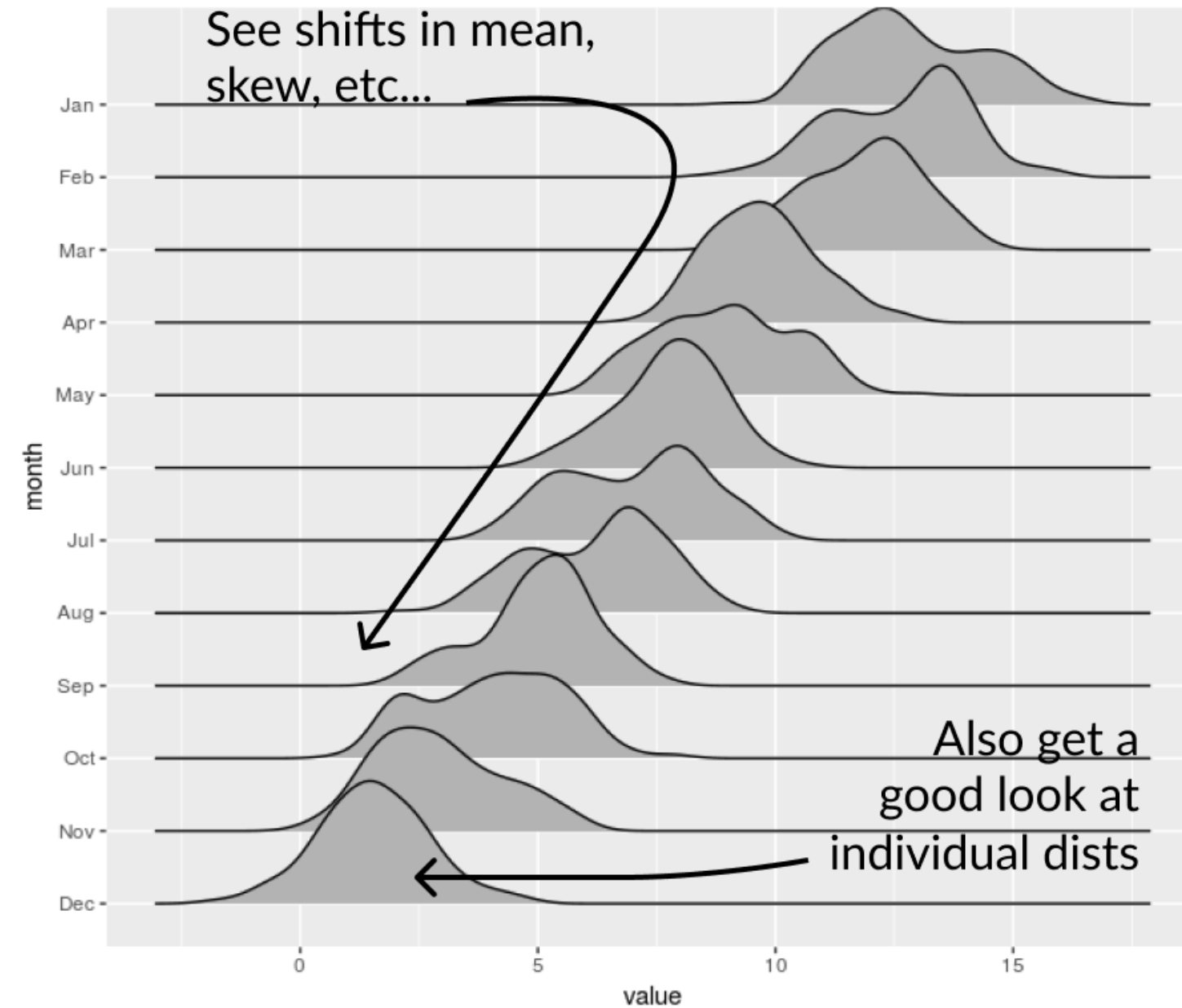


The ridgeline plot

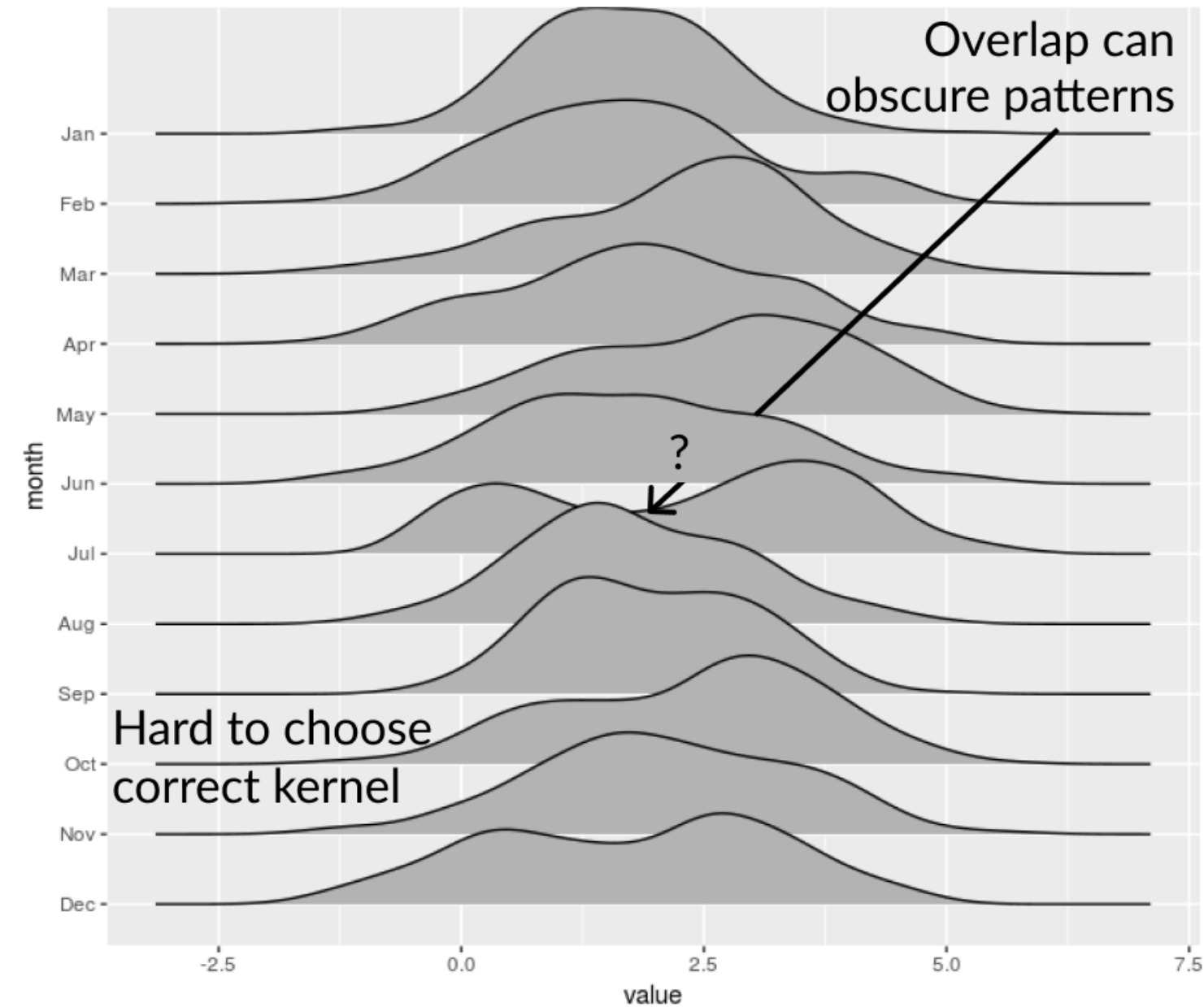
```
library(ggribes) # Gives us geom_density_ridges()  
ggplot(md_speeding, aes(x = speed_over, y = month)) +  
  geom_density_ridges(bandwidth = 2) +  
  xlim(1, 35)
```



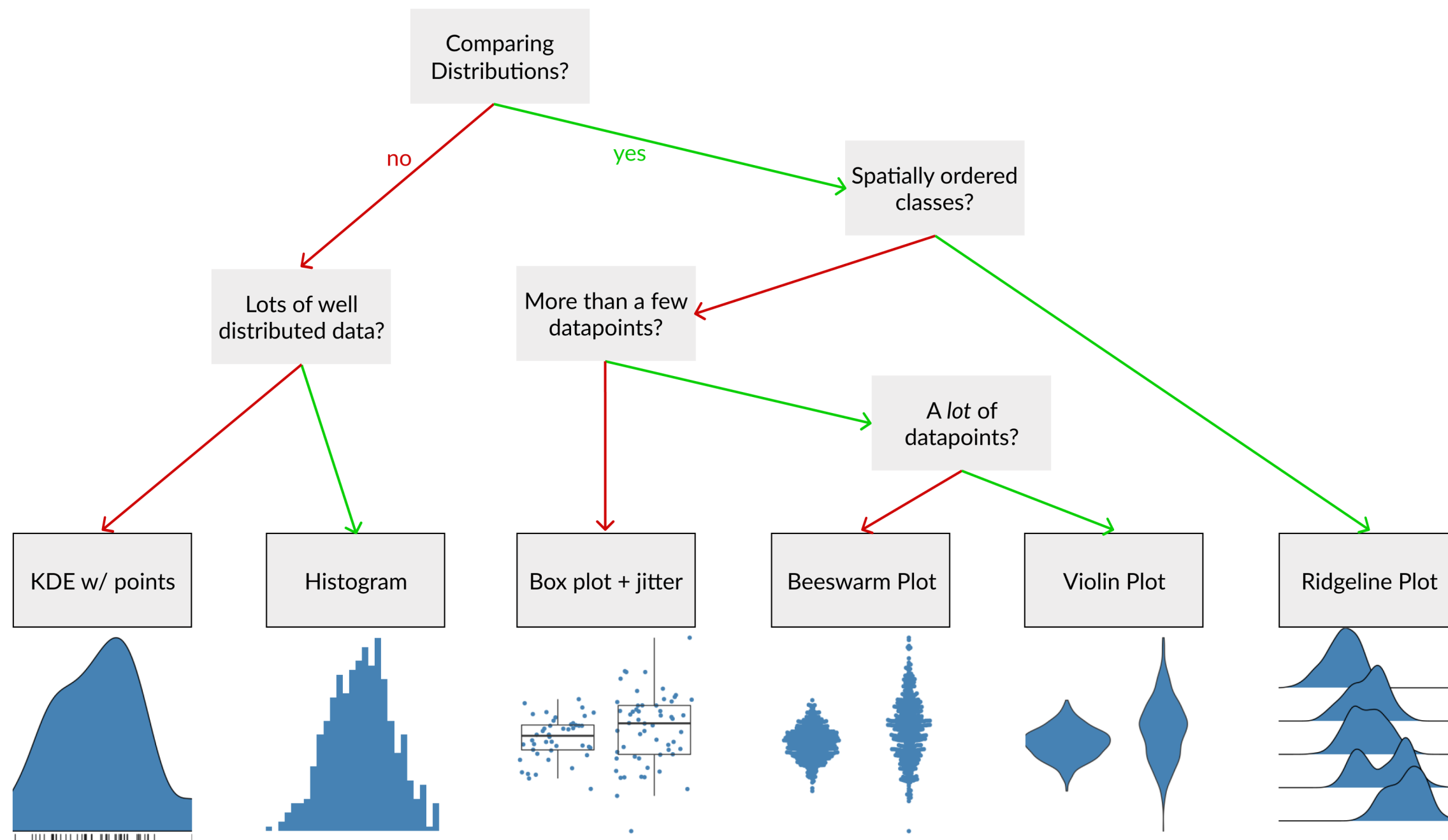
Ridgeline pros



Ridgeline cons



Overview of distribution visualizations



Let's make some ridgelines!

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Congratulations!

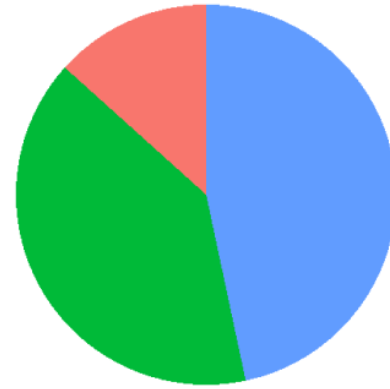
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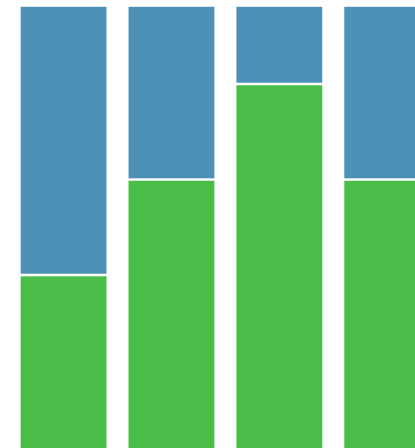
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Chapter 1: Proportions

Three or less classes and precision not important?



Good for comparing values
across populations....



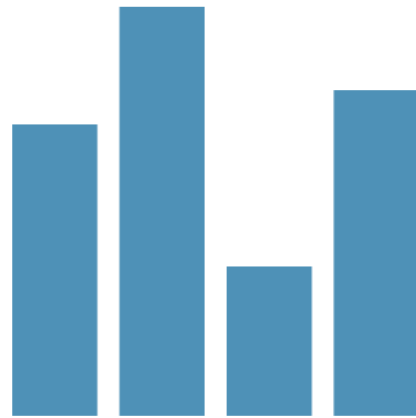
Need more precision and
have the space?



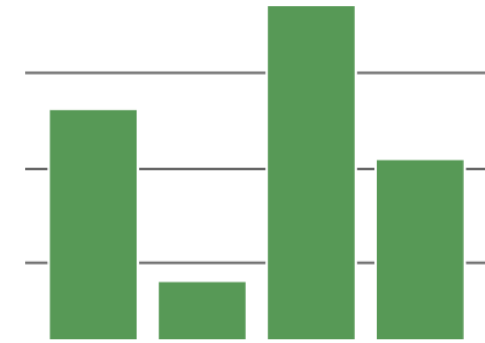
... bad for comparing
values *within* populations

Chapter 2: Point data

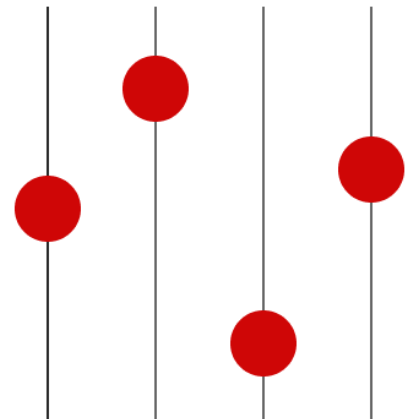
Data is a stackable quantity? E.g. dollars, counts...



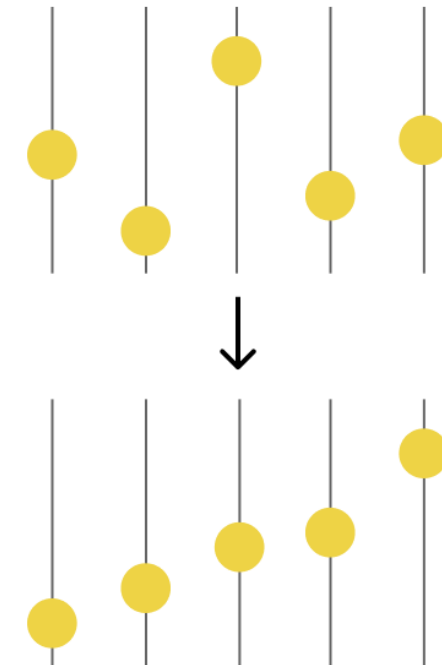
No need for vertical grid lines on bars.



Not stackable? E.g. percents, ratio...

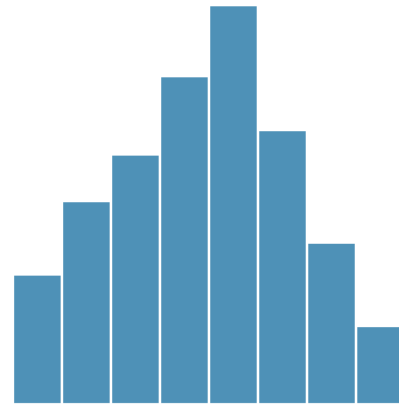


Ordering is almost always a good idea.

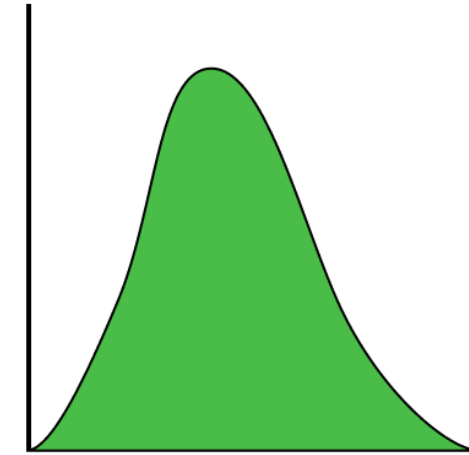


Chapter 3: Single distributions

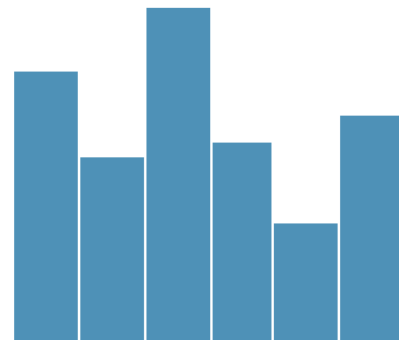
Histograms are simple and intuitive...



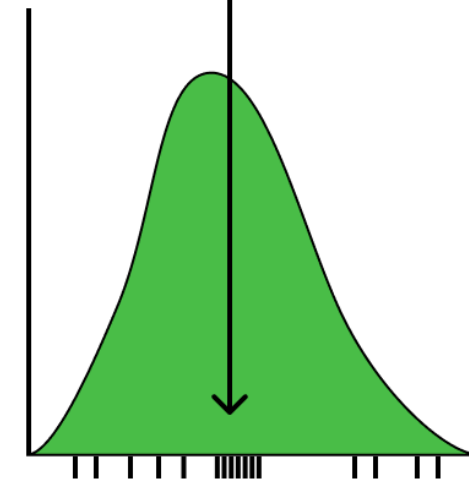
Kernel Density Estimators work with small data and produce nice smooth plots.



... but are sensitive to bin width and placement choices.

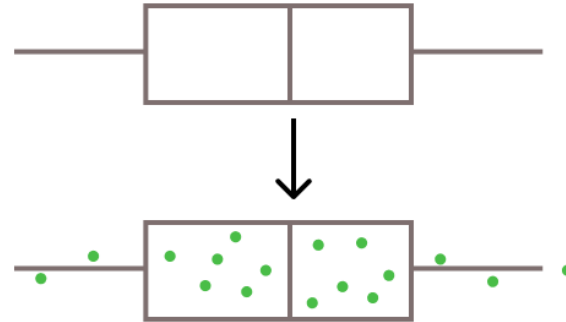


But always try and show raw data if possible to see gaps the KDE may be filling.

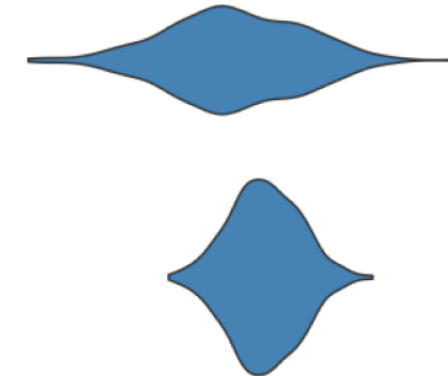


Chapter 4: Comparing distributions

Boxplots hide a lot of data, so augment them with jittered points



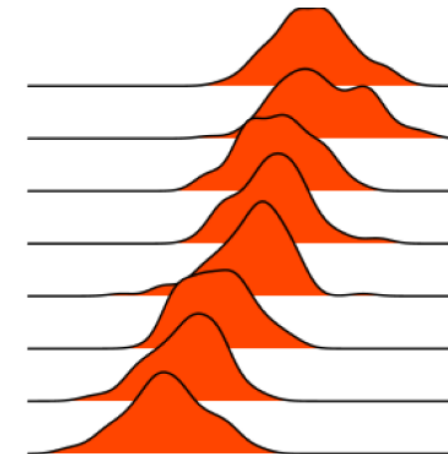
Violin plots are symmetric KDEs that work well when you have lots of points.



Beeswarm plots are an alternative 'smart' jittering that shows density.



If you have spatially ordered groups, consider the ridgeline plot.



Going further

Flowing data

Curated list of data visualizations and R-based tutorials.

Twitter (#datavis)

An ongoing stream of cool projects and inspiration.

Datawrapper Blog

Articles that dig deep into visualization techniques and mistakes.

Books!

- **Data Visualization**, Andy Kirk
- **The Functional Art and The Truthful Art** by Alberto Cairo

Thank you!

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