

PSY 3307

## Central Tendency

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# Agenda

- Greek Terms
- Central Tendency
- Measures of Central Tendency
- The Mode
- The Median
- The Mean
- When to Use Each Measure
- Using The Mean in Research
  - Deviation
- Describing Population Mean

# Greek Terms

$$\Sigma = \text{Sum of scores}$$

```
21+ 21+ 25+ 23+ 24+ 23+ 20+ 19+ 24+ 25+ 21+ 21+ 26+ 22+ 22+ 24+ 20+ 19+
```

```
## [1] 696
```

$$\Sigma X = \text{Sum Of } X$$

```
x <- c(21, 21, 25, 23, 24, 23, 20, 19, 24, 25, 21, 21, 26, 22, 22, 24, 20, 19)
sum(x)
```

```
## [1] 696
```

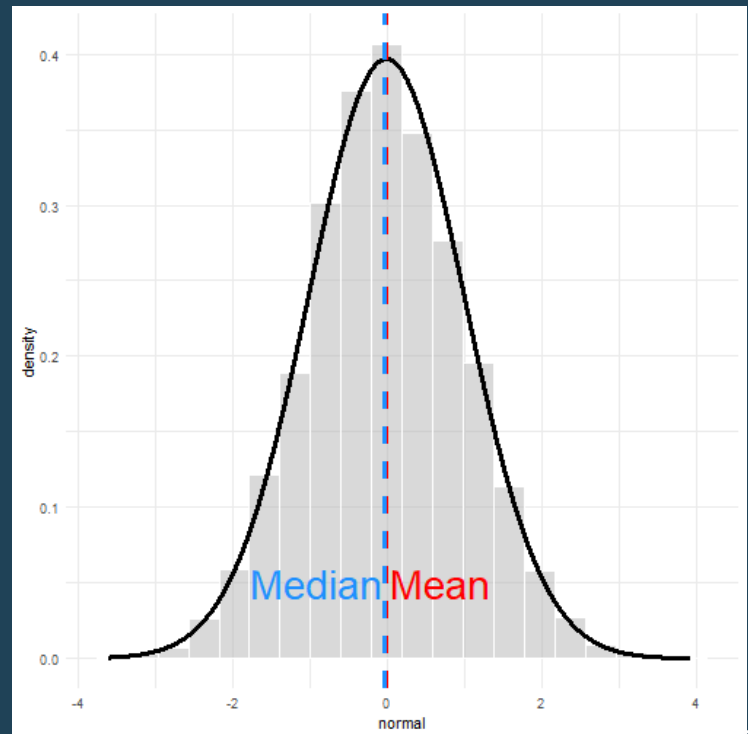
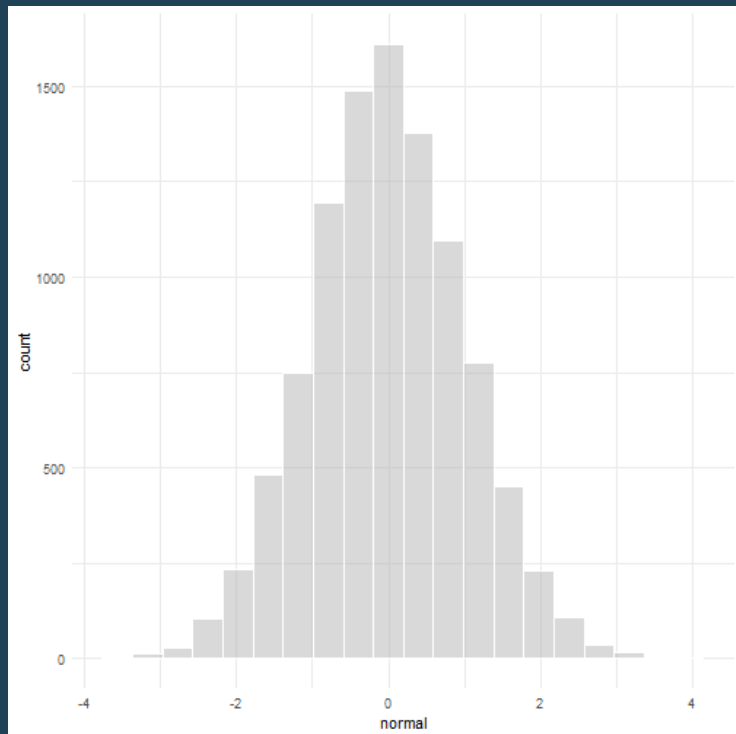
- Both of these are stating that we are adding all of the data points together.
- **Sum of X** is the sum of the scores in a sample
  - X is just another way to say all the data points for the variable.

# Central Tendency

- Concept that as statisticians every person is just a data point
- We are interested in the central score
  - We are interested in how much a person is away from that central score
  - But when it comes to statistics we like to group together our participants' scores/values
- **Measures of Central Tendency** are statistics that summarize the location of a distribution on a variable by indicating where the center is
- A Normal distribution will have the central point right down the center
- A skewed distribution will have the central point where the frequency of scores is the highest

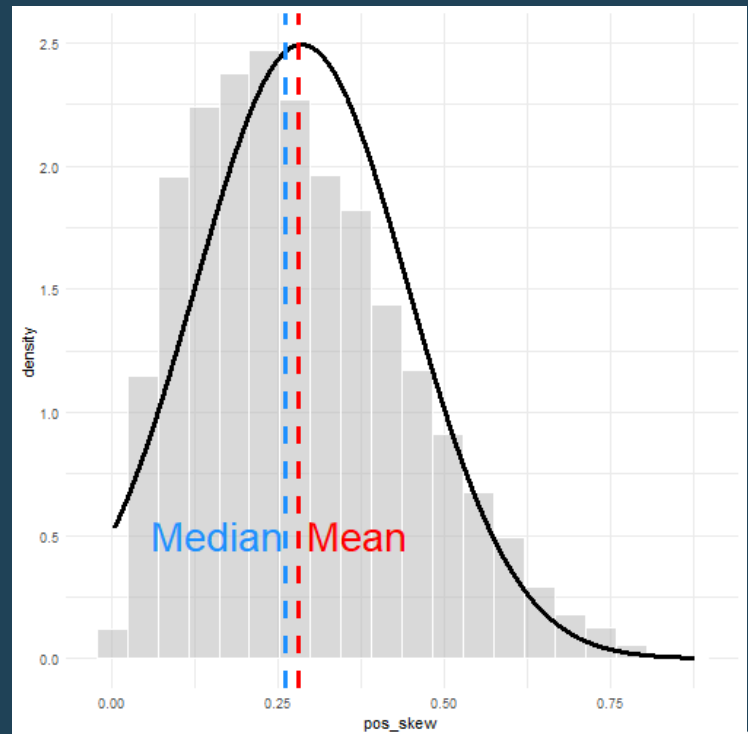
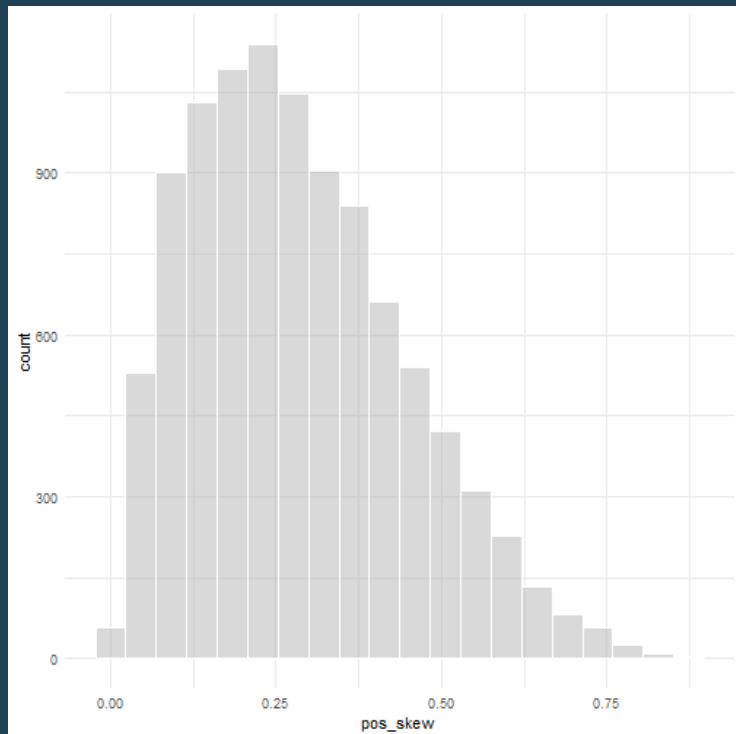
# Normal Distribution Example

```
##          mean    sd median
## normal   -0.01  1.01  -0.03
## pos_skew  0.29  0.16   0.26
## neg_skew  0.71  0.16   0.74
```



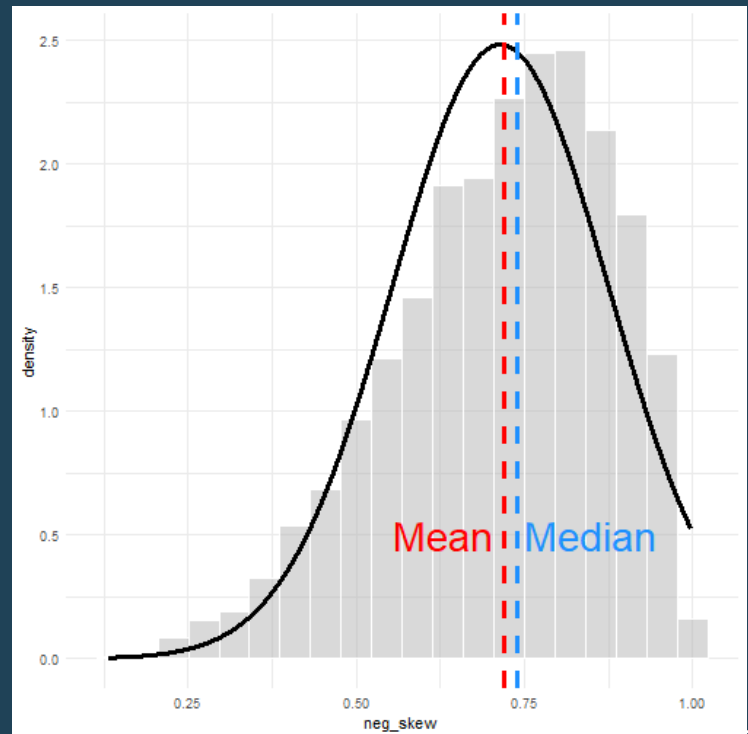
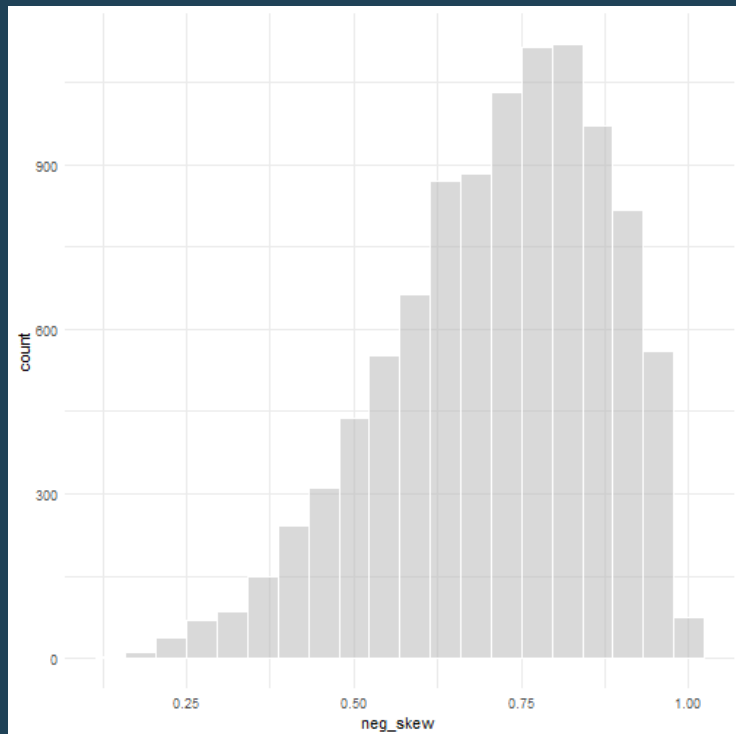
# Positively Skewed Data

```
##          mean    sd median
## normal  -0.01  1.01  -0.03
## pos_skew  0.29  0.16   0.26
## neg_skew  0.71  0.16   0.74
```



# Negatively Skewed Data

```
##          mean    sd median
## normal  -0.01  1.01  -0.03
## pos_skew  0.29  0.16   0.26
## neg_skew  0.71  0.16   0.74
```



# The Mode



- Value/score with the highest frequency
- Essentially useless in statistics
- **Unimodal distributions**  
distribution with only one hump;  
one value is the mode
- **Bimodal distributions**  
distribution with two humps; two  
values with the highest frequency;  
two modes

##	x										
##	19	20	21	22	23	24	25	26	33	34	37
##	4	4	6	2	4	3	2	2	1	1	1



# The Median



- **Median** is the middle value/score; the 50th percentile
- Unlike the mode, it will always be close to the middle of a distribution
- You'll only ever have one median
- The symbol is:

$$Mdn = Median$$

# The Median

- Preferred for ordinal/ordered data
- Not the best option for normally distributed interval & ratio scores
- Is more reliable when dealing with skewed data
- *Important Note* If you have an even number of scores/values, then you will add the two middle values and divide by 2

```
odd <- c(1, 6, 3, 8, 9, 8, 3)
even <- c(1, 6, 3, 4, 9, 8, 5, 6)

sort(odd)
```

```
## [1] 1 3 3 6 8 8 9
```

```
sort(even)
```

```
## [1] 1 3 4 5 6 6 8 9
```

```
get_median <- (5+6)/2
get_median
```

```
## [1] 5.5
```

```
median(odd)
```

```
## [1] 6
```

```
median(even)
```

```
## [1] 5.5
```

# The Mean/Average

- **Mean** is the score located at the **mathematical** center of a distribution
- Xbar is often used for the mean

- $\bar{X}$

- $$\bar{X} = \frac{\sum X}{N}$$

is the formula to calculate the mean.

- Xbar is the Sum of X/Scores divided by the total number of observations/scores/values
- Calculate the mean for interval and ratio scales
  - The mean of ordinal/ordered data makes no sense
- Basis for most **inferential statistics**

```
x <- c(21, 21, 25, 23, 24, 23, 20, 19, 24, 25, 21, 21, 26, 22, 22, 24, 2
sigma_x <- sum(x)
N <- 30

x_bar <- sigma_x/N

x_bar
```

```
## [1] 23.2
```

```
mean(x)
```

```
## [1] 23.2
```

# When to Use Each Measure

- The Median more accurately describes/summarizes skewed data compared to the mean
  - The mean will be pulled toward the extreme tail of the distribution
- Normal distribution then use the mean as the best measure to describe/summarize data

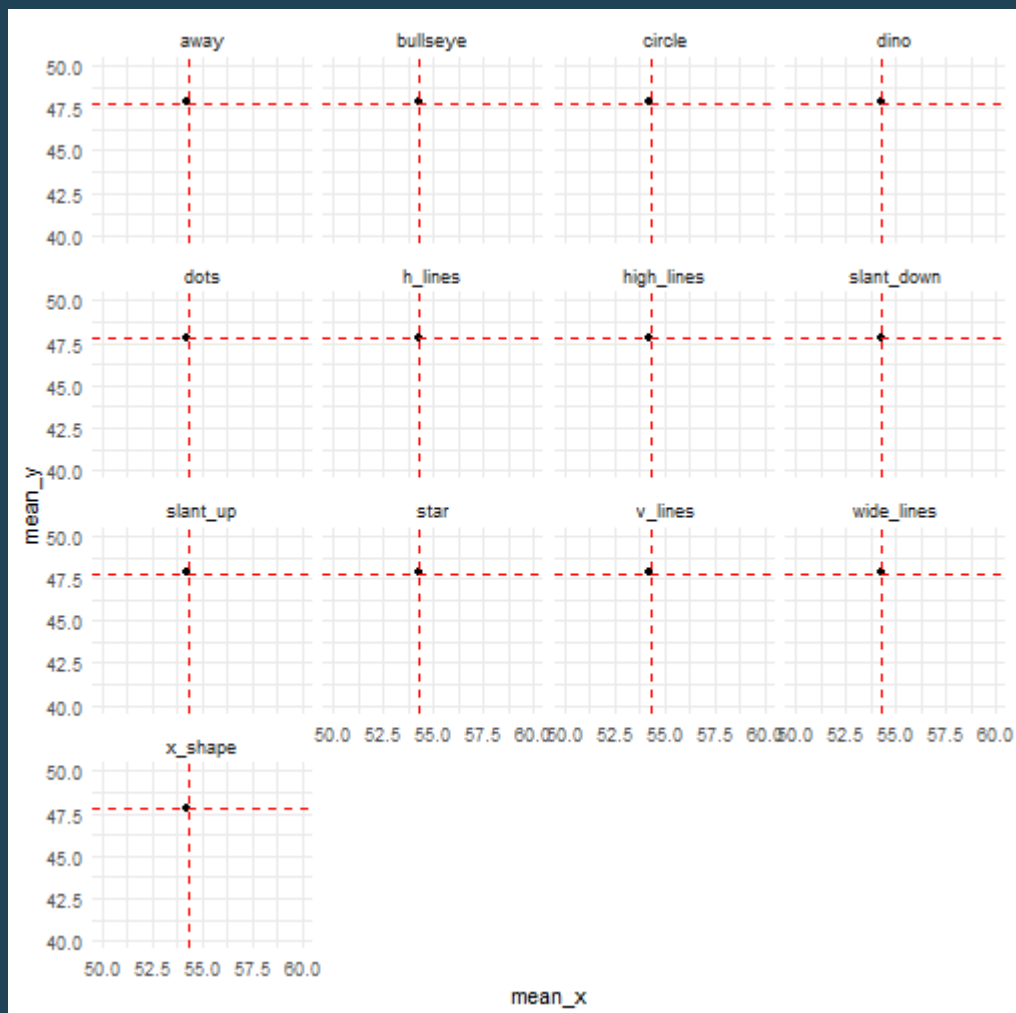
# The Mean in Research

- Most statistics revolve around the mean
- Can't just trust descriptive statistics like the mean

# Descriptive Statistics Are Never Enough

```
## Warning: package 'datasauRus' was built under R version 4.0.5
```





# Deviation

- The distance a participant's score/value is from the mean/average
- Deviations can be positive or negative
  - participants can score lower (negative) than the mean and higher (positive) than the mean
- To get the deviation, you subtract the mean from each participant's score
- $$X - \bar{X}$$
- The larger the value the farther away from the mean the score/value is
- **Sum of the deviations around the mean** is the sum of all differences between the scores and the mean

# Example

```
x
```

```
## [1] 21 21 25 23 24 23 20 19 24 25 21 21 26 22 22 24 20 19 19 20 20 23 21 2  
## [26] 23 19 26 34 33
```

```
one_part <- 37  
another_part <- 19  
  
x_bar
```

```
## [1] 23.2
```

```
one_part - x_bar
```

```
## [1] 13.8
```

```
another_part - x_bar
```

```
## [1] -4.2
```

```
all_deviations <- x - x_bar
all_deviations
```

```
## [1] -2.2 -2.2 1.8 -0.2 0.8 -0.2 -3.2 -4.2 0.8 1.8 -2.2 -2.2 2.8 -1.2
## [16] 0.8 -3.2 -4.2 -4.2 -3.2 -3.2 -0.2 -2.2 -2.2 13.8 -0.2 -4.2 2.8 10.8
```

```
sum_deviations <- sum(all_deviations)
sum_deviations
```

```
## [1] 0.000000000000002131628
```

# Looking to the Future

- Deviations is the start for upcoming lectures and statistical tests, especially the sum of the deviations

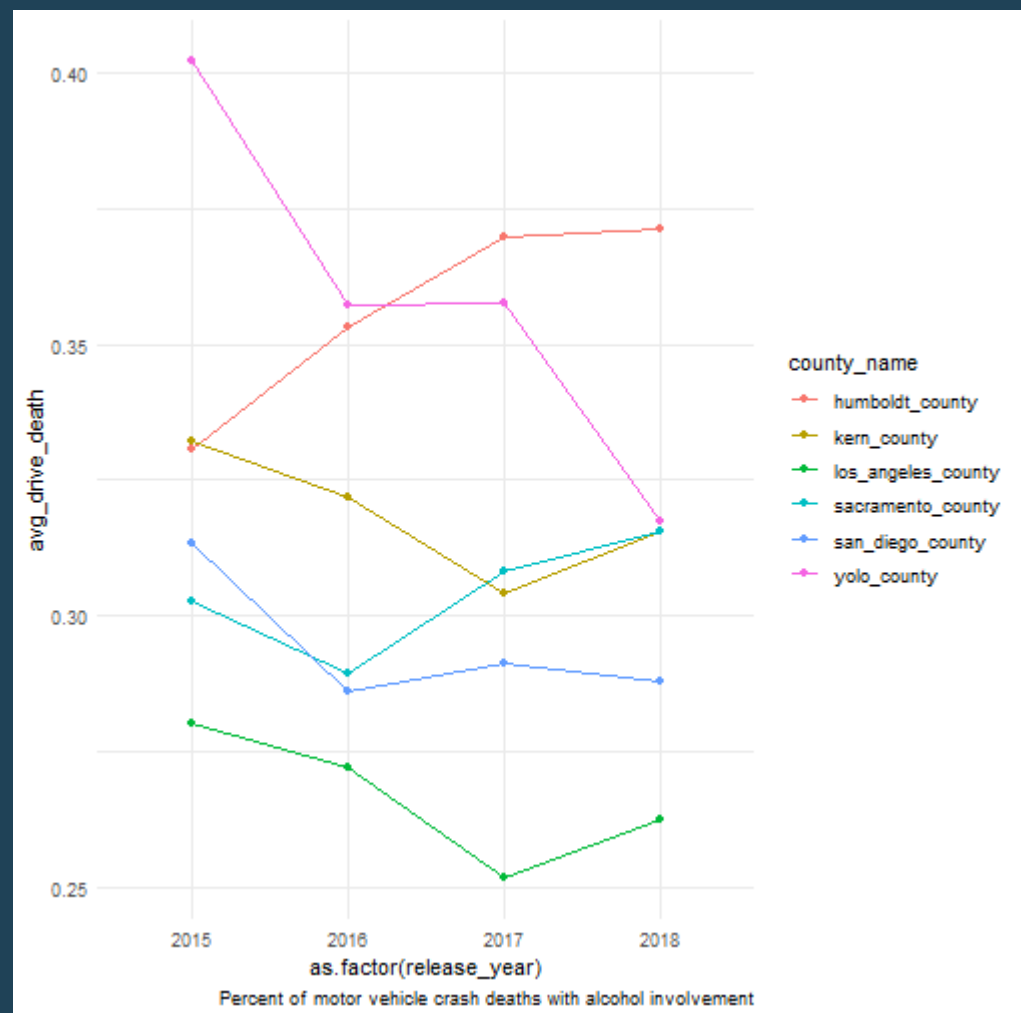
- $$\Sigma(X - \bar{X})$$

- If the sum of the deviations is 0 then that means your math is good
- Deviation of each score/value from the mean is often referred to as error/residual in statistical tests
- Correlational designs use the mean of IV and the mean of DV to look for a relationship between the two variables
- Experiments compare the two or more groups (IV) and the relationship with the mean value/score of the DV

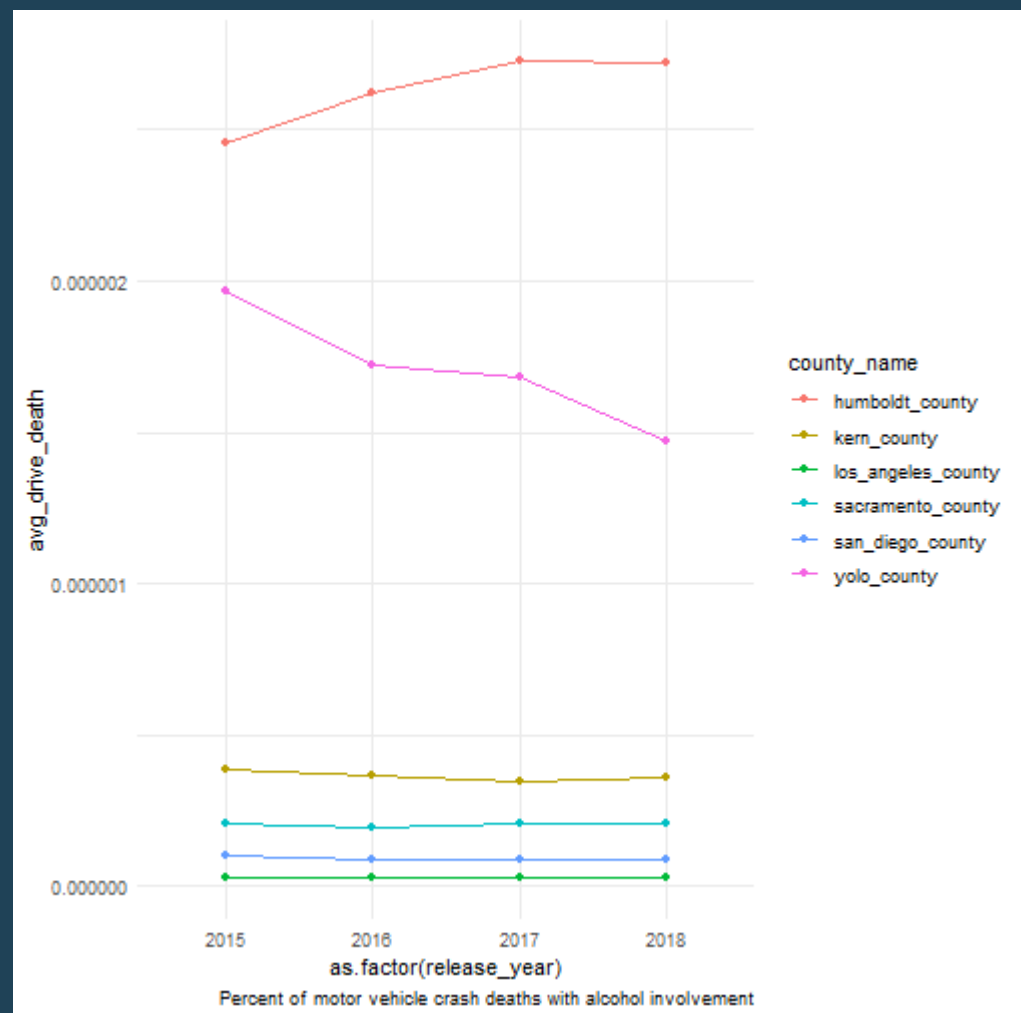
# Visuals

- Line Graphs are good for showing progress or change in time
- To show group differences (nominal or ordinal IV), bar graphs are the norm
- Scatterplots are best for continuous IV and continuous DV (interval or ratio)

## `summarise()` has grouped output by 'release\_year'. You can override using



## `summarise()` has grouped output by 'release\_year'. You can override using





# Describing the Population Mean

- $\mu = \text{Population Mean}$
- If you test a population then you would use mu instead of xbar

- $$\mu = \frac{\sum X}{N}$$