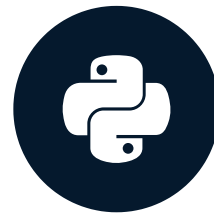


# Generating bootstrap replicates

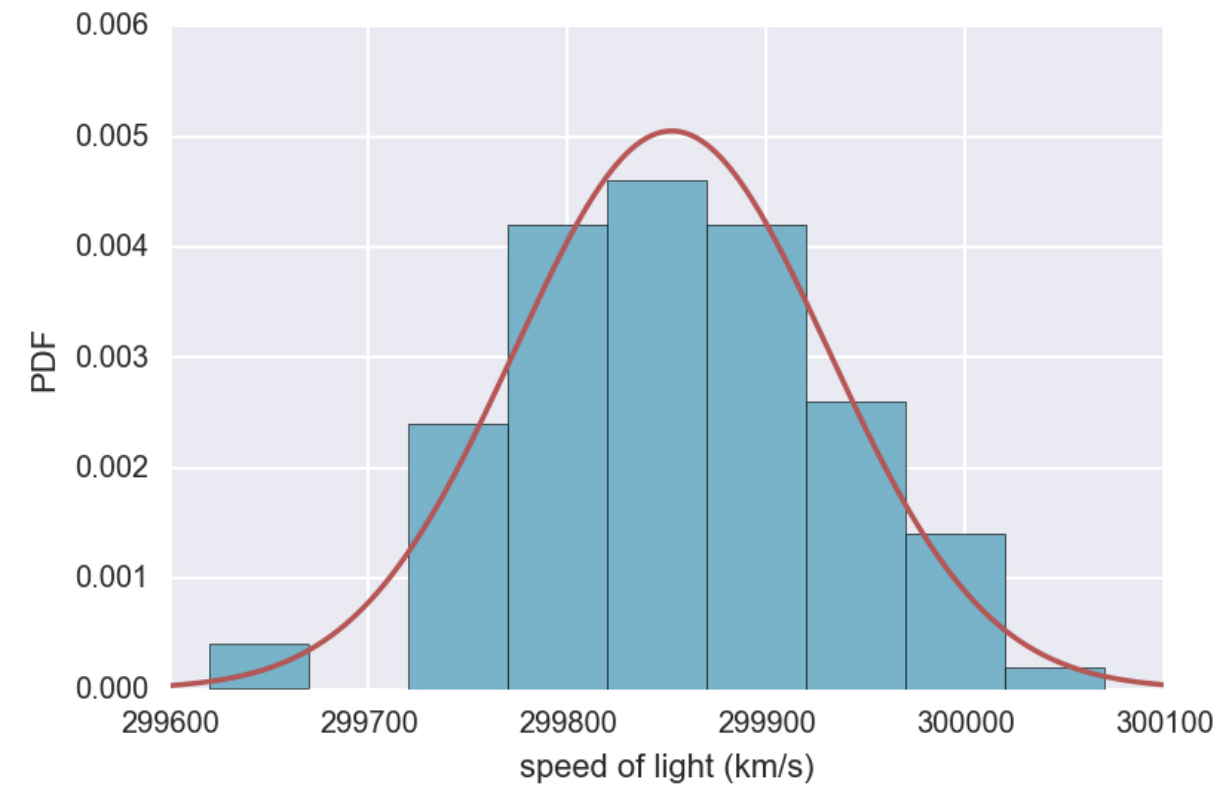
STATISTICAL THINKING IN PYTHON (PART 2)



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# Michelson's speed of light measurements



<sup>1</sup> Data: Michelson, 1880

# Resampling an array

Data:

[ 23.3 , 27.1 , 24.3 , 25.3 , 26.0 ]

Mean = 25.2

Resampled data:

[ , , , , ]

# Resampling an array

Data:

[ 23.3 , 27.1 , 24.3 , 25.3 , 26.0 ]

Mean = 25.2

Resampled data:

[ , , , , ]

# Resampling an array

Data:

[ 23.3, , 24.3, 25.3, 26.0 ]

Mean = 25.2

Resampled data:

[ 27.1, , , ]

# Resampling an array

Data:

[ 23.3, 27.1, 24.3, 25.3, 26.0 ]

Mean = 25.2

Resampled data:

[ 27.1, , , , ]

# Resampling an array

Data:

```
[23.3, 27.1, 24.3, 25.3, 26.0]
```

Mean = 25.2

Resampled data:

```
[27.1, 26.0, , , ]
```

# Resampling an array

Data:

[23.3, 27.1, 24.3, 25.3, 26.0]

Mean = 25.2

Resampled data:

[27.1, 26.0, , , ]



# Resampling an array

Data:

```
[23.3, 27.1, 24.3, 25.7, 26.0]
```

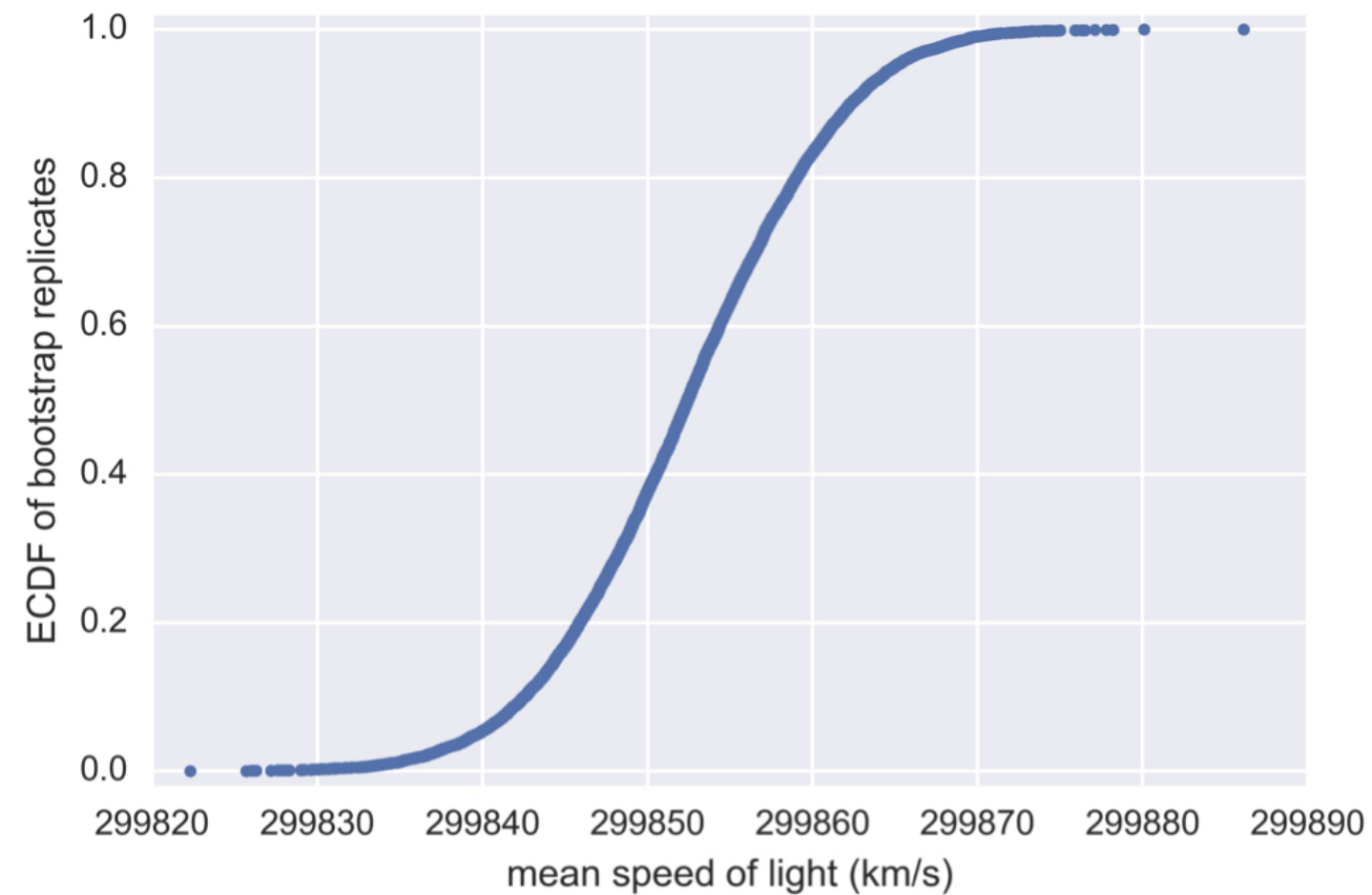
Mean = 25.2

Resampled data:

```
[27.1, 26.0, 23.3, 25.7, 23.3]
```

Mean = 25.08

# Mean of resampled Michelson measurements



# Bootstrapping

- The use of resampled data to perform statistical inference

# Bootstrap sample

- A resampled array of the data

# Bootstrap replicate

- A statistic computed from a resampled array

# Resampling engine: np.random.choice()

```
import numpy as np  
np.random.choice([1,2,3,4,5], size=5)
```

```
array([5, 3, 5, 5, 2])
```

# Computing a bootstrap replicate

```
bs_sample = np.random.choice(michelson_speed_of_light,  
                              size=100)  
  
np.mean(bs_sample)
```

```
299847.79999999999
```

```
np.median(bs_sample)
```

```
299845.0
```

```
np.std(bs_sample)
```

```
83.564286025729331
```

# Let's practice!

STATISTICAL THINKING IN PYTHON (PART 2)



# Bootstrap confidence intervals

STATISTICAL THINKING IN PYTHON (PART 2)



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# Bootstrap replicate function

```
def bootstrap_replicate_1d(data, func):  
    """Generate bootstrap replicate of 1D data."""  
    bs_sample = np.random.choice(data, len(data))  
    return func(bs_sample)  
bootstrap_replicate_1d(michelson_speed_of_light, np.mean)
```

```
299859.20000000001
```

```
bootstrap_replicate_1d(michelson_speed_of_light, np.mean)
```

```
299855.70000000001
```

```
bootstrap_replicate_1d(michelson_speed_of_light, np.mean)
```

```
299850.29999999999
```

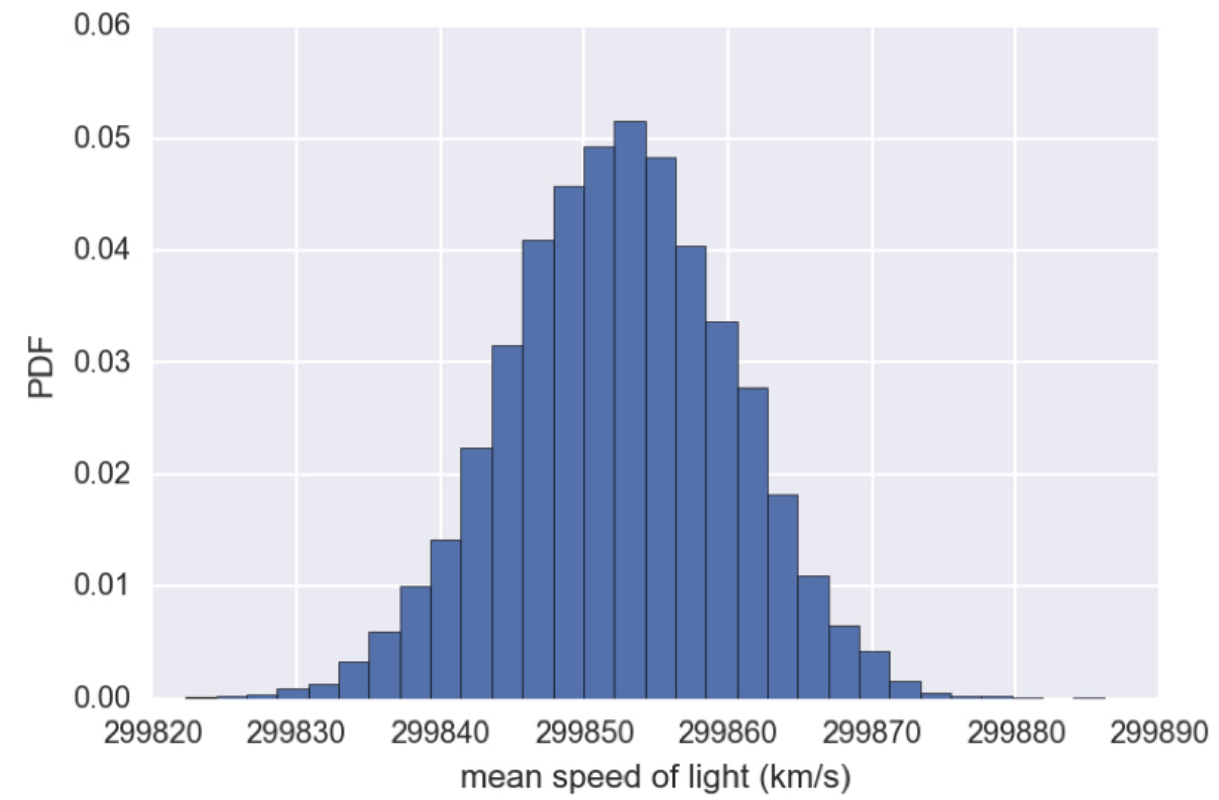
# Many bootstrap replicates

```
bs_replicates = np.empty(10000)
for i in range(10000):
    bs_replicates[i] = bootstrap_replicate_1d(
        michelson_speed_of_light, np.mean)
```

# Plotting a histogram of bootstrap replicates

```
_ = plt.hist(bs_replicates, bins=30, normed=True)
_ = plt.xlabel('mean speed of light (km/s)')
_ = plt.ylabel('PDF')
plt.show()
```

# Bootstrap estimate of the mean



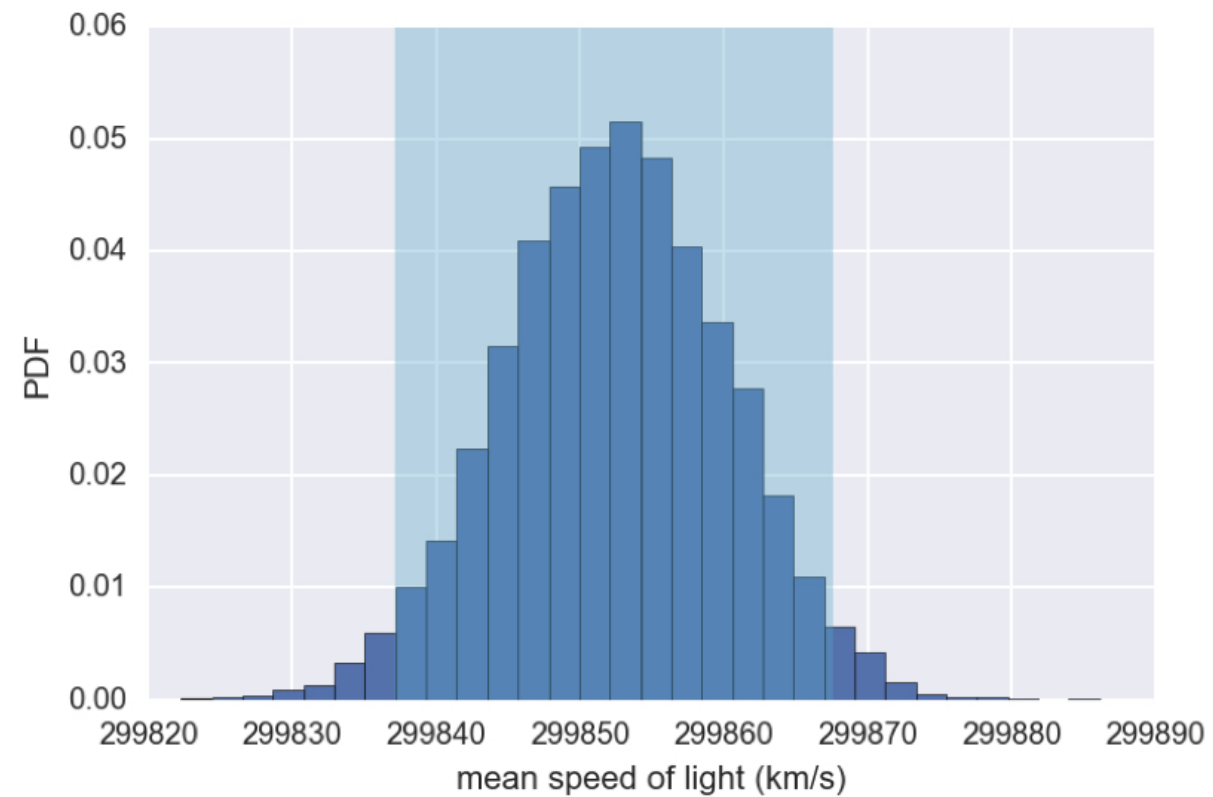
# Confidence interval of a statistic

- If we repeated measurements over and over again,  $p\%$  of the observed values would lie within the  $p\%$  confidence interval.

# Bootstrap confidence interval

```
conf_int = np.percentile(bs_replicates, [2.5, 97.5])
```

```
array([ 299837., 299868.])
```



# Let's practice!

STATISTICAL THINKING IN PYTHON (PART 2)



# Pairs bootstrap

STATISTICAL THINKING IN PYTHON (PART 2)



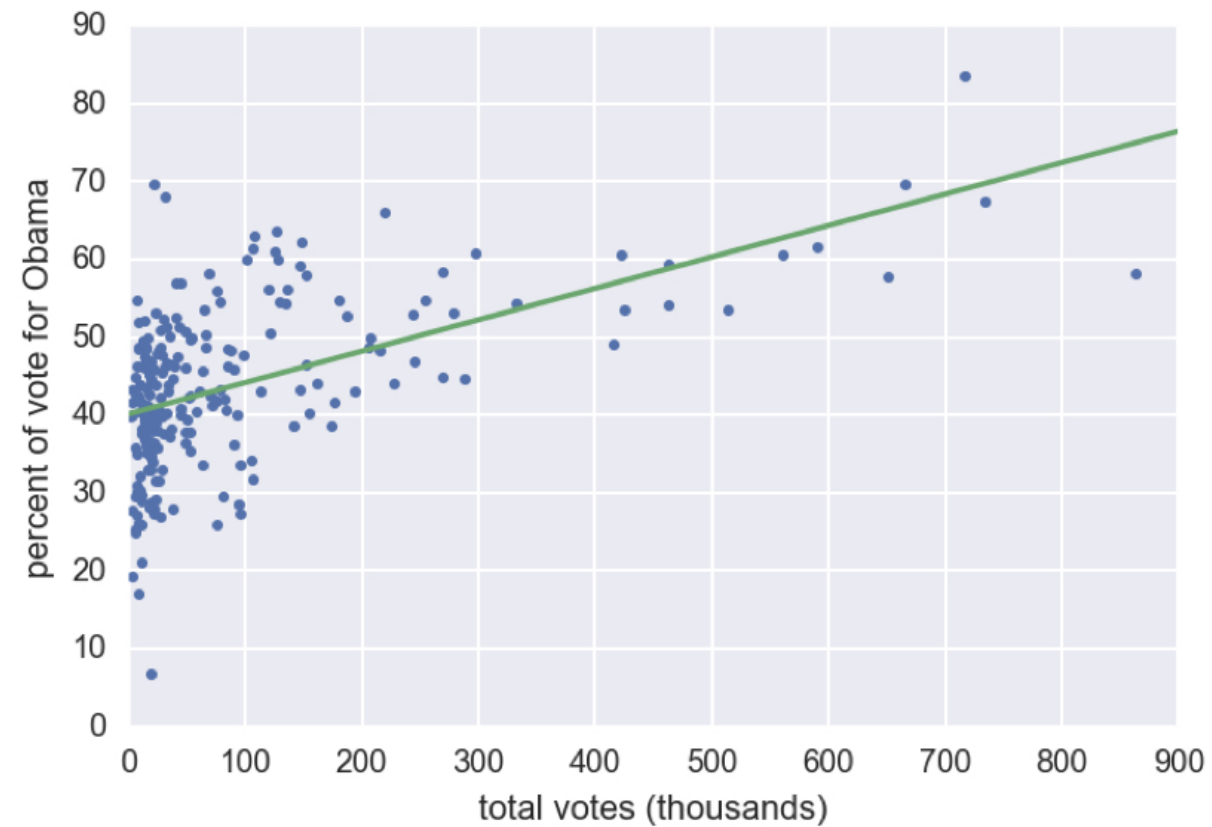
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# Nonparametric inference

- Make no assumptions about the model or probability distribution underlying the data

# 2008 US swing state election results



<sup>1</sup> Data retrieved from Data.gov (<https://www.data.gov/>)

# Pairs bootstrap for linear regression

- Resample data in pairs
- Compute slope and intercept from resampled data
- Each slope and intercept is a bootstrap replicate
- Compute confidence intervals from percentiles of bootstrap replicates

# Generating a pairs bootstrap sample

```
np.arange(7)
```

```
array([0, 1, 2, 3, 4, 5, 6])
```

```
inds = np.arange(len(total_votes))  
bs_inds = np.random.choice(inds, len(inds))  
bs_total_votes = total_votes[bs_inds]  
bs_dem_share = dem_share[bs_inds]
```

# Computing a pairs bootstrap replicate

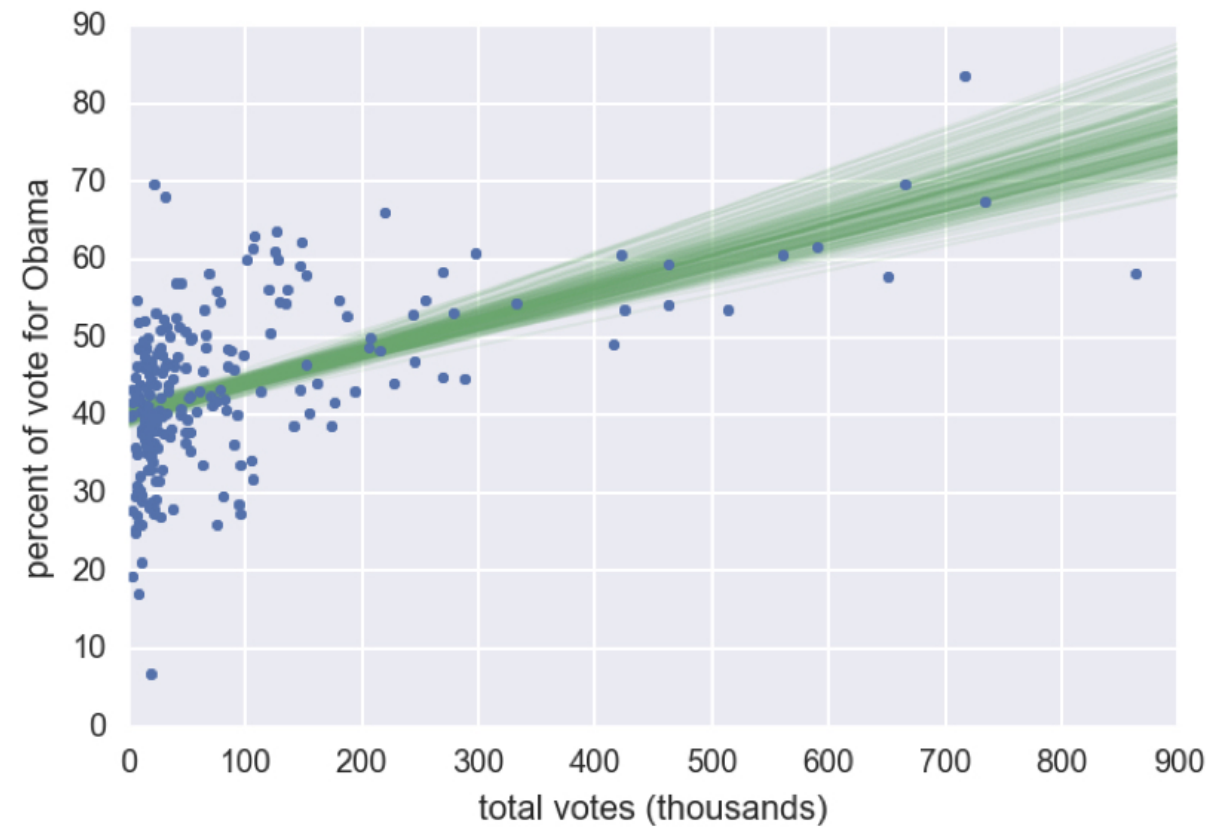
```
bs_slope, bs_intercept = np.polyfit(bs_total_votes,  
                                     bs_dem_share, 1)  
  
bs_slope, bs_intercept
```

```
(3.9053605692223672e-05, 40.387910131803025)
```

```
np.polyfit(total_votes, dem_share, 1) # fit of original
```

```
array([ 4.03707170e-05,  4.01139120e+01])
```

# 2008 US swing state election results



<sup>1</sup> Data retrieved from Data.gov (<https://www.data.gov/>)

# Let's practice!

STATISTICAL THINKING IN PYTHON (PART 2)