

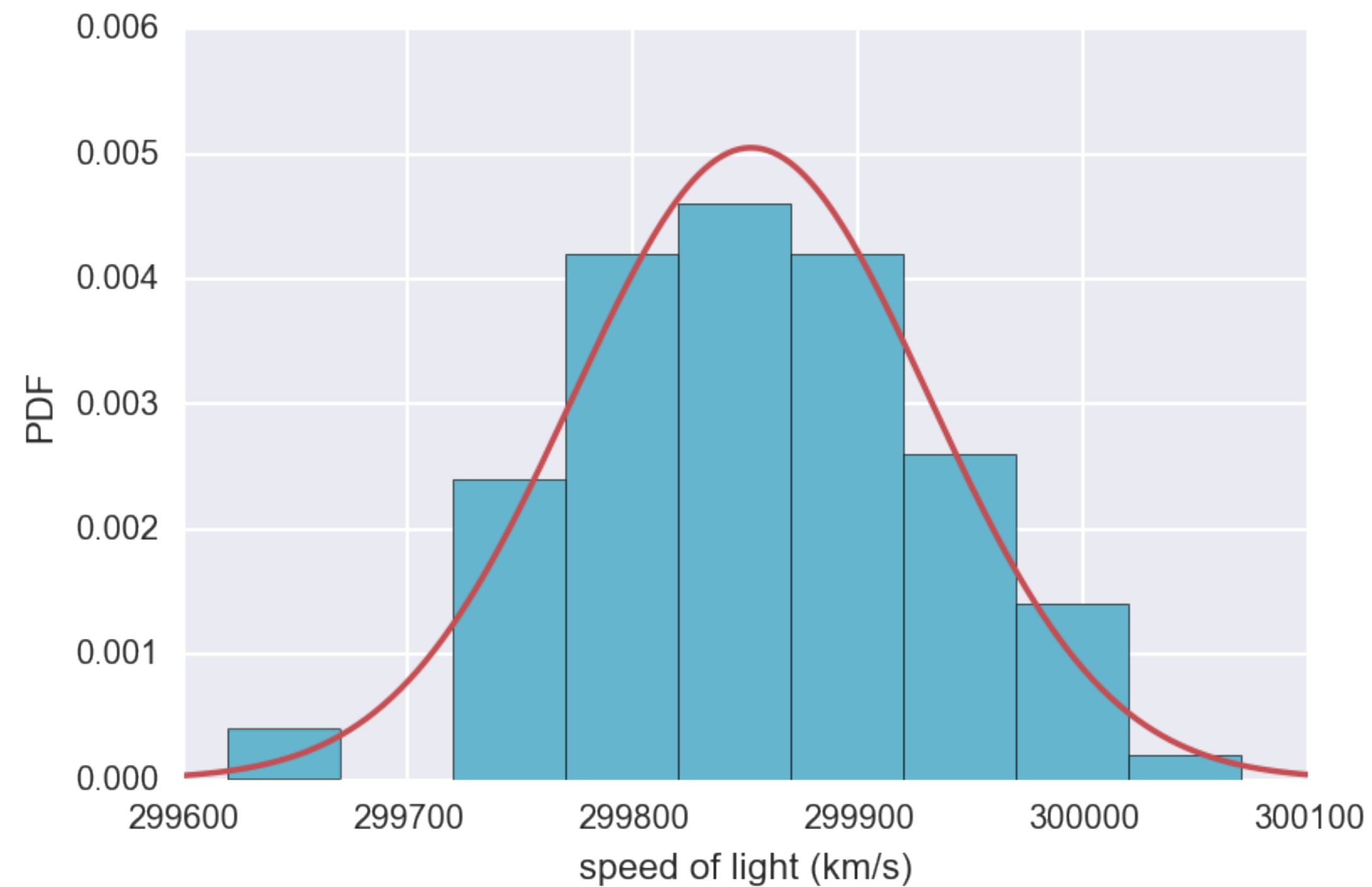


STATISTICAL THINKING IN PYTHON II

# **Generating bootstrap replicates**



# Michelson's speed of light measurements





# 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, 25.7, 26.0]
```

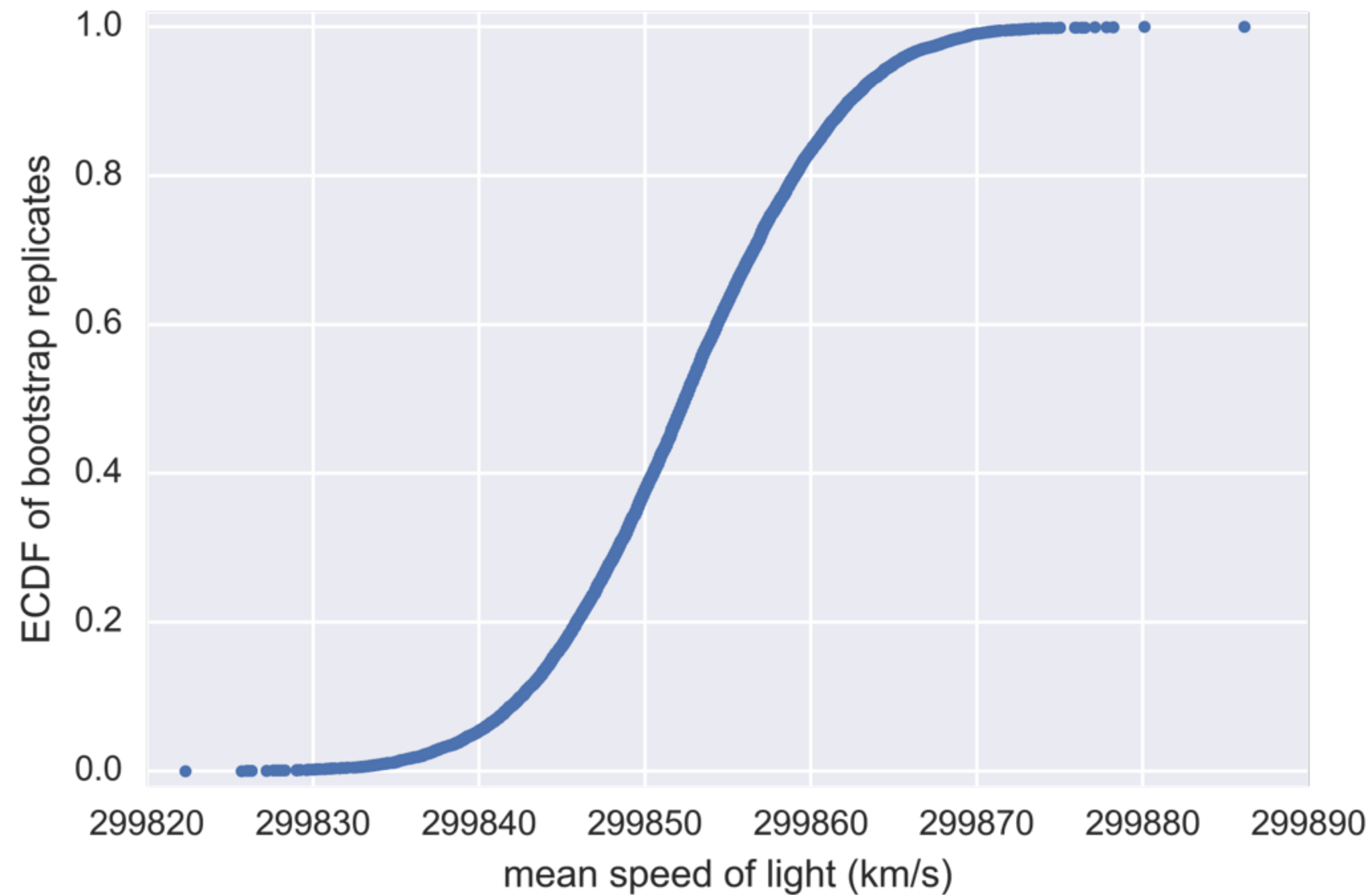
Mean = 25.2

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()`

```
In [1]: import numpy as np
```

```
In [2]: np.random.choice([1,2,3,4,5], size=5)
```

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



# Computing a bootstrap replicate

```
In [1]: bs_sample = np.random.choice(michelson_speed_of_light,  
....:                                size=100)
```

```
In [2]: np.mean(bs_sample)  
Out[2]: 299847.79999999999
```

```
In [3]: np.median(bs_sample)  
Out[3]: 299845.0
```

```
In [4]: np.std(bs_sample)  
Out[4]: 83.564286025729331
```



STATISTICAL THINKING IN PYTHON II

**Let's practice!**



STATISTICAL THINKING WITH PYTHON II

# **Bootstrap confidence intervals**



# Bootstrap replicate function

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

```
In [2]: bootstrap_replicate_1d(michelson_speed_of_light, np.mean)  
Out[2]: 299859.200000000001
```

```
In [3]: bootstrap_replicate_1d(michelson_speed_of_light, np.mean)  
Out[3]: 299855.700000000001
```

```
In [4]: bootstrap_replicate_1d(michelson_speed_of_light, np.mean)  
Out[4]: 299850.29999999999
```





# Many bootstrap replicates

```
In [1]: bs_replicates = np.empty(10000)

In [2]: for i in range(10000):
...:     bs_replicates[i] = bootstrap_replicate_1d(
...:         michelson_speed_of_light, np.mean)
...:
```



# Plotting a histogram of bootstrap replicates

```
In [1]: _ = plt.hist(bs_replicates, bins=30, normed=True)
```

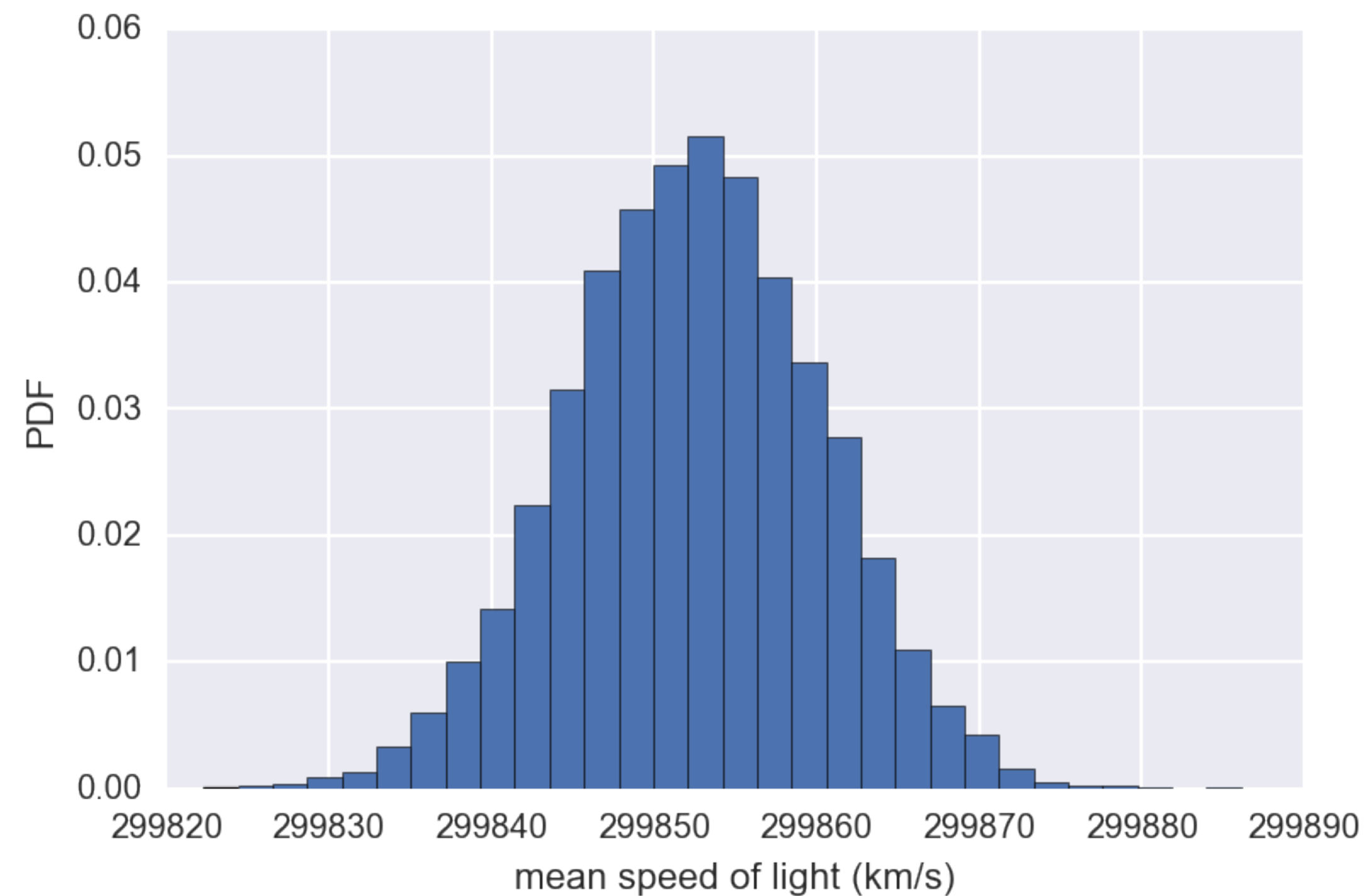
```
In [2]: _ = plt.xlabel('mean speed of light (km/s)')
```

```
In [3]: _ = plt.ylabel('PDF')
```

```
In [4]: plt.show()
```



# Bootstrap estimate of the mean



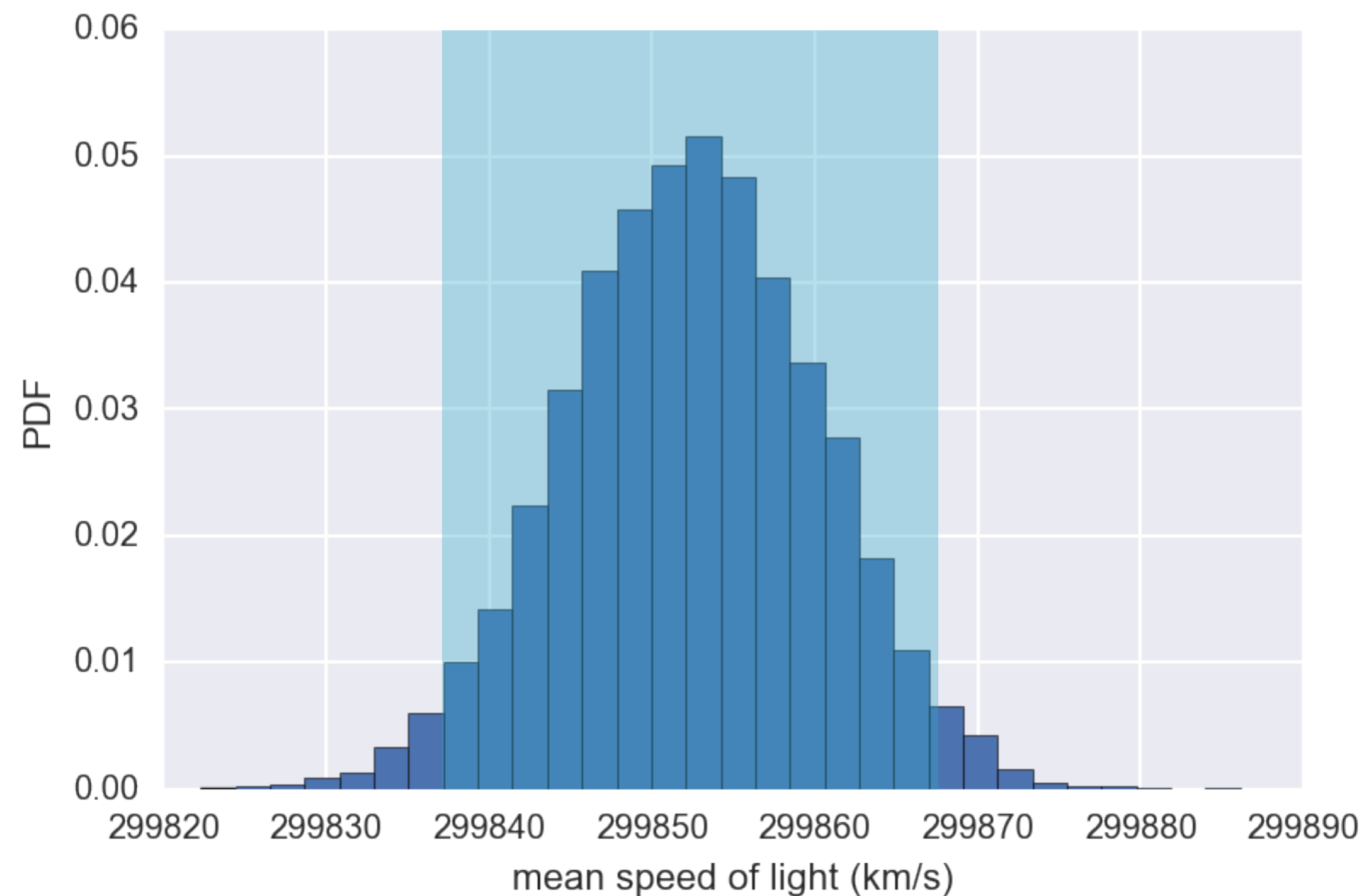
# Confidence interval of a statistic

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



# Bootstrap confidence interval

```
In [1]: conf_int = np.percentile(bs_replicates, [2.5, 97.5])  
Out[1]: array([ 299837.,  299868.])
```





STATISTICAL THINKING WITH PYTHON II

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STATISTICAL THINKING IN PYTHON II

# Pairs bootstrap

# Nonparametric inference

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





# 2008 US swing state election results



# 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

```
In [1]: np.arange(7)
```

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

```
In [1]: inds = np.arange(len(total_votes))
```

```
In [2]: bs_inds = np.random.choice(inds, len(inds))
```

```
In [3]: bs_total_votes = total_votes[bs_inds]
```

```
In [4]: bs_dem_share = dem_share[bs_inds]
```



# Computing a pairs bootstrap replicate

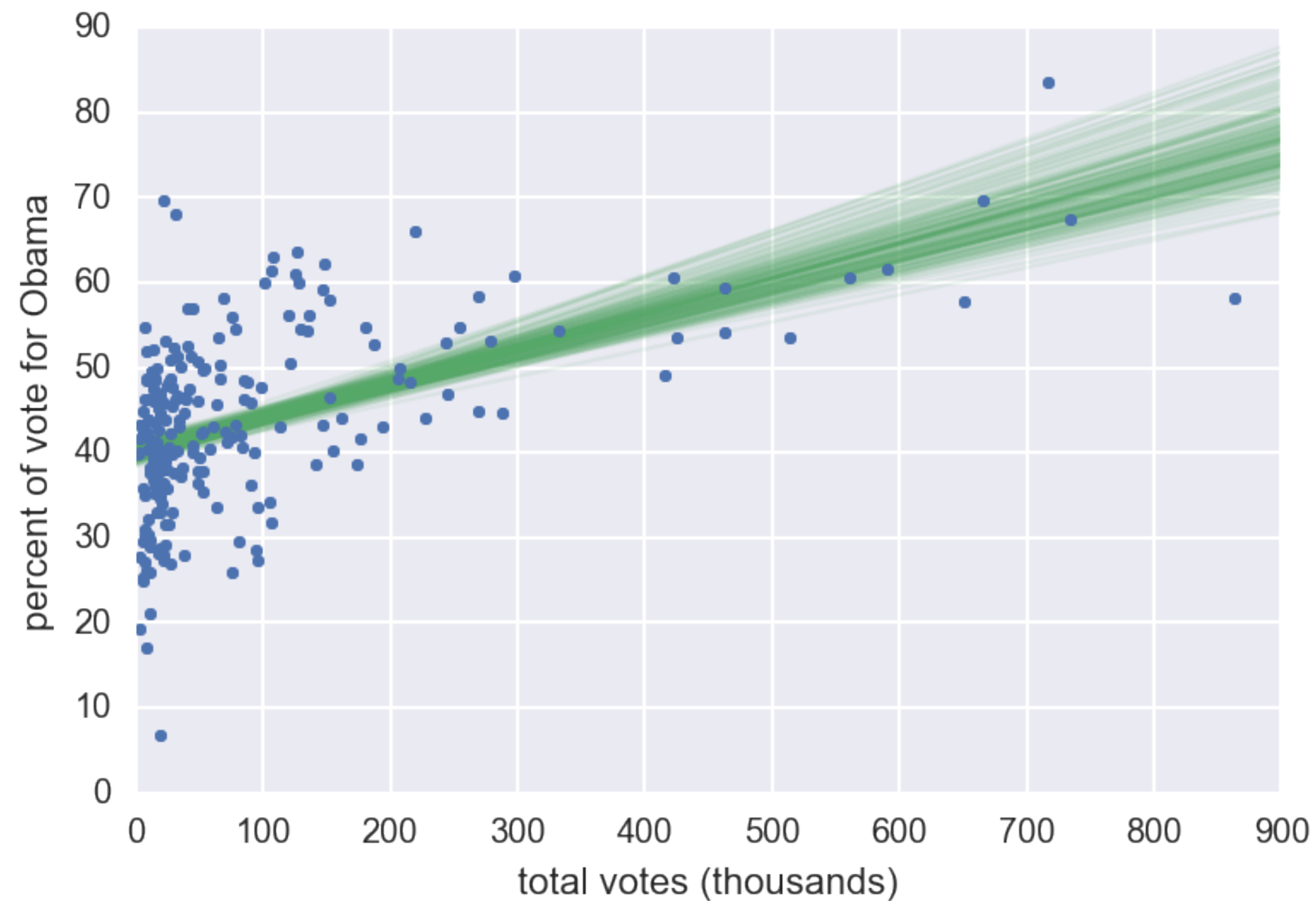
```
In [1]: bs_slope, bs_intercept = np.polyfit(bs_total_votes,  
.....:                                     bs_dem_share, 1)
```

```
In [2]: bs_slope, bs_intercept  
Out[2]: (3.9053605692223672e-05, 40.387910131803025)
```

```
In [3]: np.polyfit(total_votes, dem_share, 1) # fit of original  
Out[3]: array([ 4.03707170e-05,  4.01139120e+01])
```



# 2008 US swing state election results





STATISTICAL THINKING IN PYTHON II

**Let's practice!**