



# Introduction to Exploratory Data Analysis





#### Exploratory data analysis

 The process of organizing, plotting, and summarizing a data set

1 (\$)

"Exploratory data analysis can never be the whole story, but nothing else can serve as the foundation stone."

—John Tukey





#### 2008\_swing\_states.csv state,county,total\_votes,dem\_votes,rep\_votes,dem\_share PA, Erie County, 127691, 75775, 50351, 60.08 PA, Bradford County, 25787, 10306, 15057, 40.64 PA, Tioga County, 17984, 6390, 11326, 36.07 PA, McKean County, 15947, 6465, 9224, 41.21 PA, Potter County, 7507, 2300, 5109, 31.04 PA, Wayne County, 22835, 9892, 12702, 43.78 PA, Susquehanna County, 19286, 8381, 10633, 44.08 PA, Warren County, 18517, 8537, 9685, 46.85 OH, Ashtabula County, 44874, 25027, 18949, 56.94 OH, Lake County, 121335, 60155, 59142, 50.46 PA, Crawford County, 38134, 16780, 20750, 44.71 OH, Lucas County, 219830, 142852, 73706, 65.99 OH, Fulton County, 21973, 9900, 11689, 45.88 OH, Geauga County, 51102, 21250, 29096, 42.23 OH, Williams County, 18397, 8174, 9880, 45.26 PA, Wyoming County, 13138, 5985, 6983, 46.15 PA, Lackawanna County, 107876, 67520, 39488, 63.10 PA, Elk County, 14271, 7290, 6676, 52.20 PA, Forest County, 2444, 1038, 1366, 43.18 PA, Venango County, 23307, 9238, 13718, 40.24 OH, Erie County, 41229, 23148, 17432, 57.01

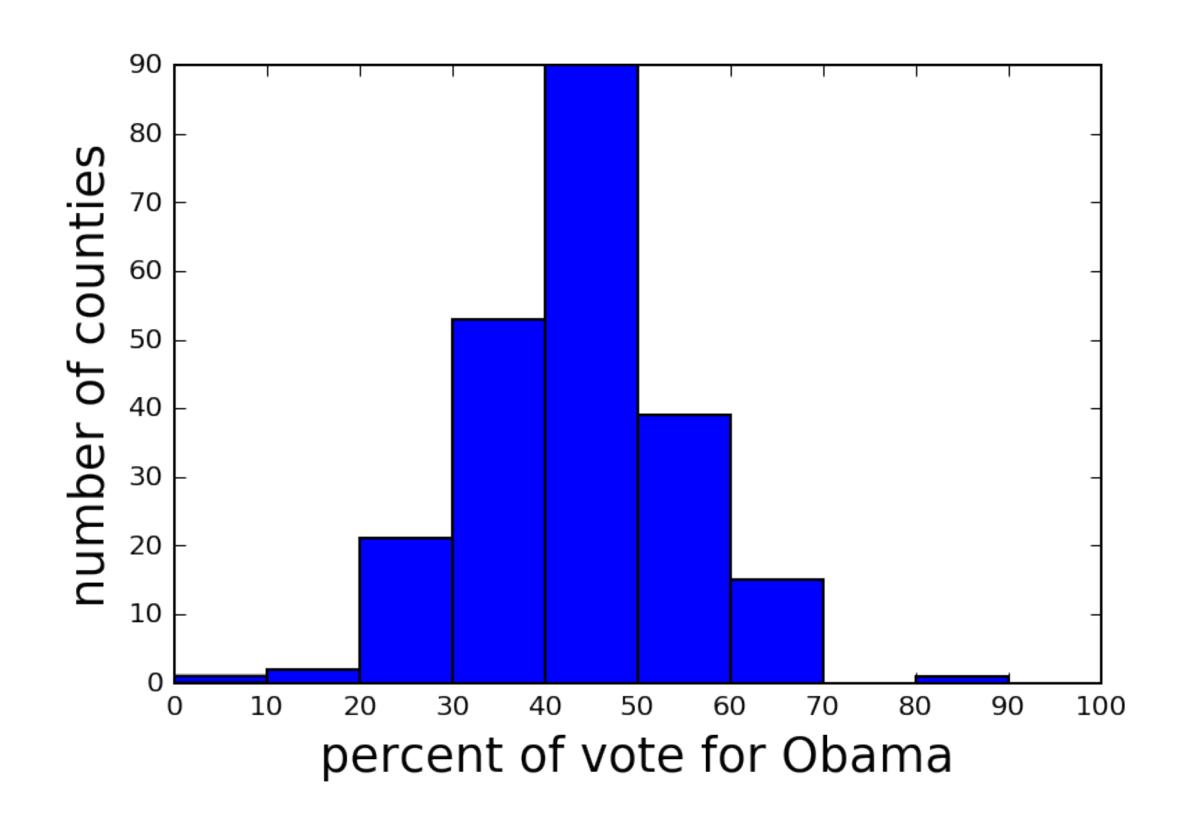




```
In [1]: import pandas as pd
In [2]: df_swing = pd.read_csv('2008_swing_states.csv')
In [3]: df_swing[['state', 'county', 'dem_share']]
Out[3]:
                            dem_share
                     county
   state
                Erie County
      PA
                                60.08
            Bradford County 40.64
      PA
              Tioga County 36.07
3
      PA
              McKean County 41.21
              Potter County
      PA
                            31.04
4
               Wayne County 43.78
      PA
          Susquehanna County
6
                                44.08
              Warren County
                                46.85
      PA
      OH
            Ashtabula County
                                56.94
```











### Let's practice!

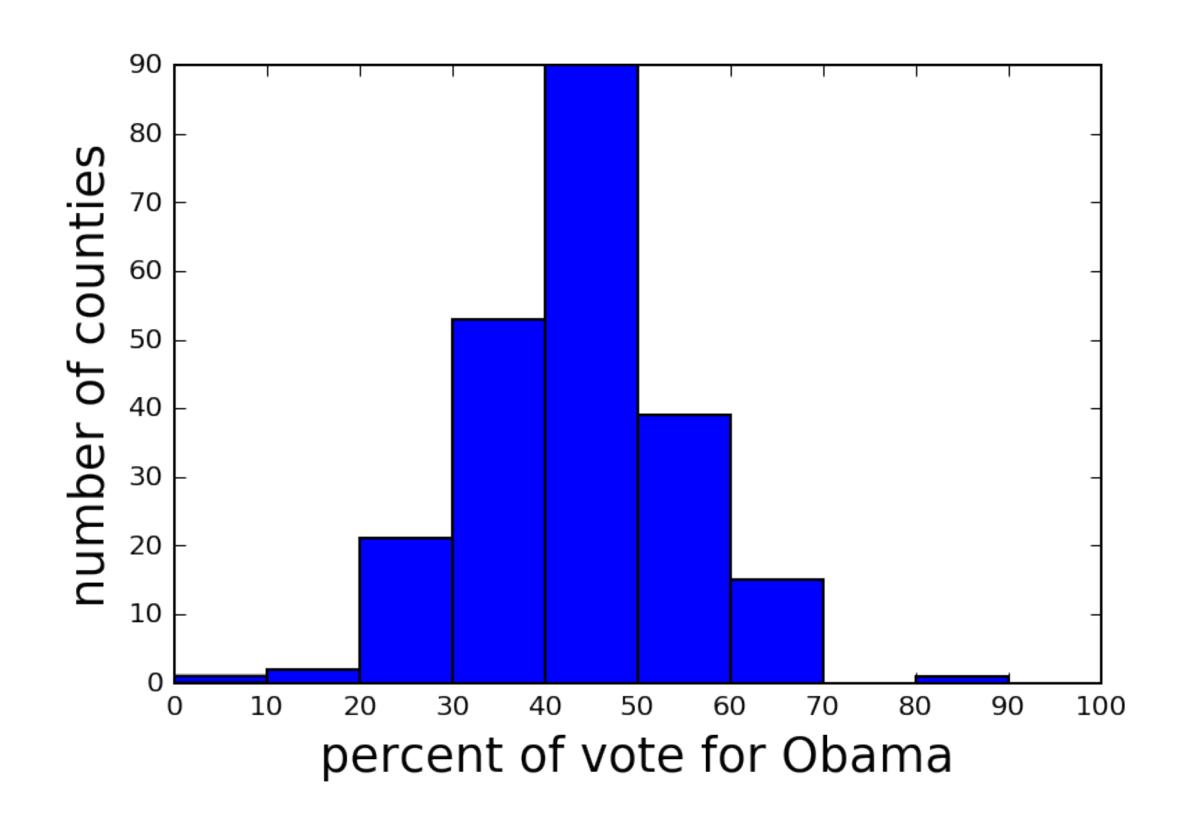




## Plotting a histogram











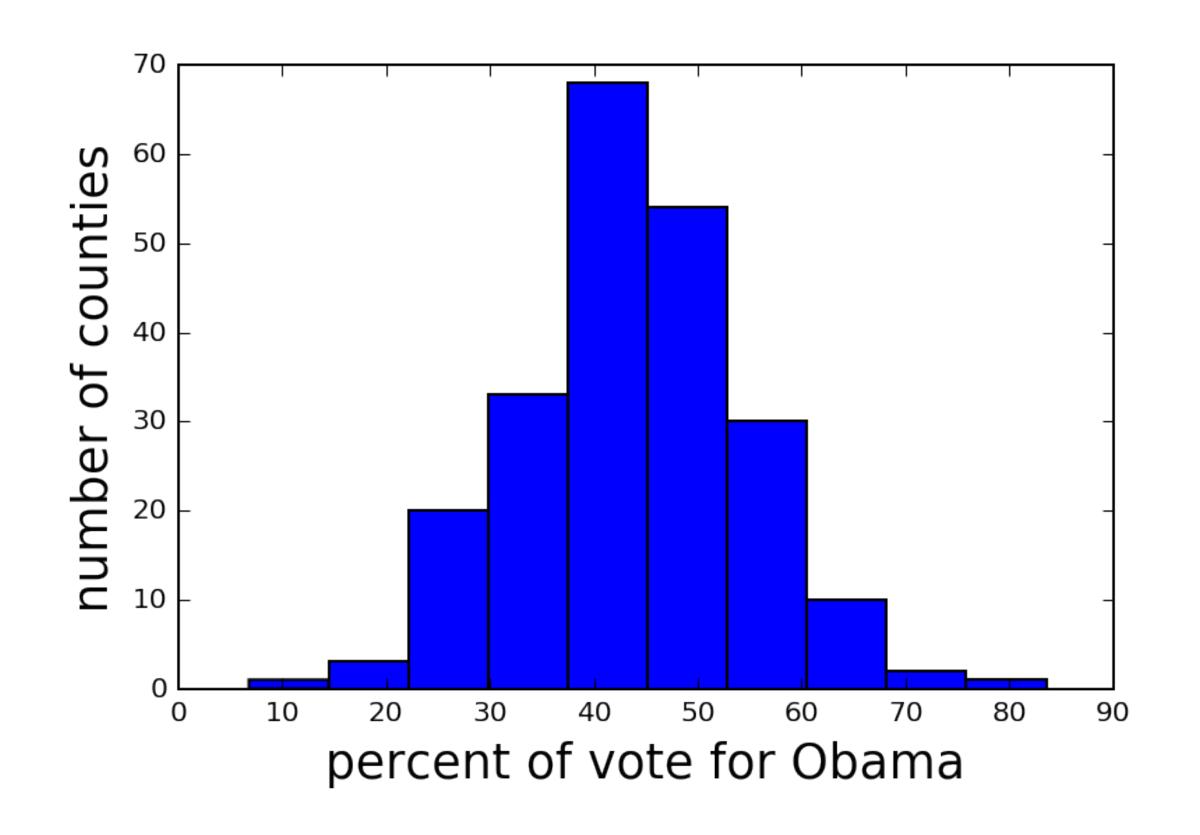
#### Generating a histogram

```
In [1]: import matplotlib.pyplot as plt
In [2]: _ = plt.hist(df_swing['dem_share'])
In [3]: _ = plt.xlabel('percent of vote for Obama')
In [4]: _ = plt.ylabel('number of counties')
In [5]: plt.show()
```

Always label your axes

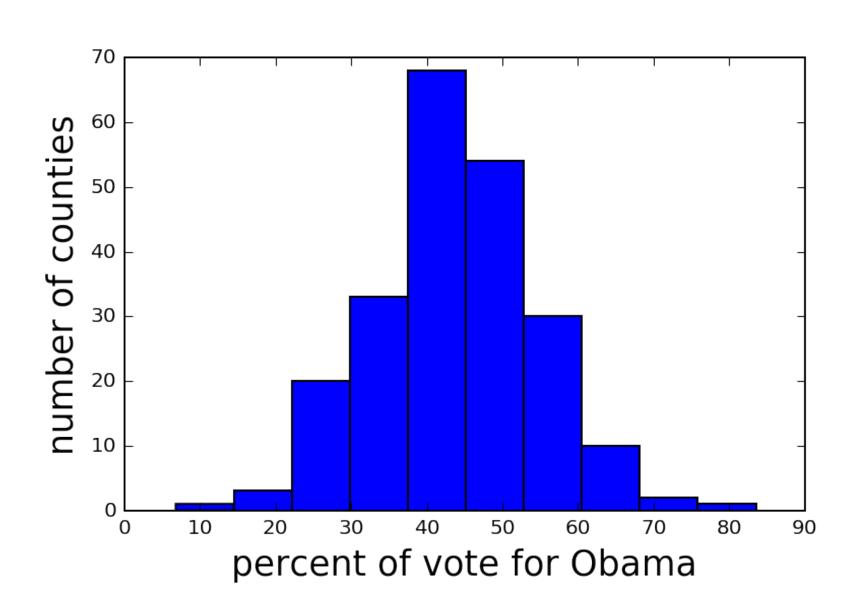


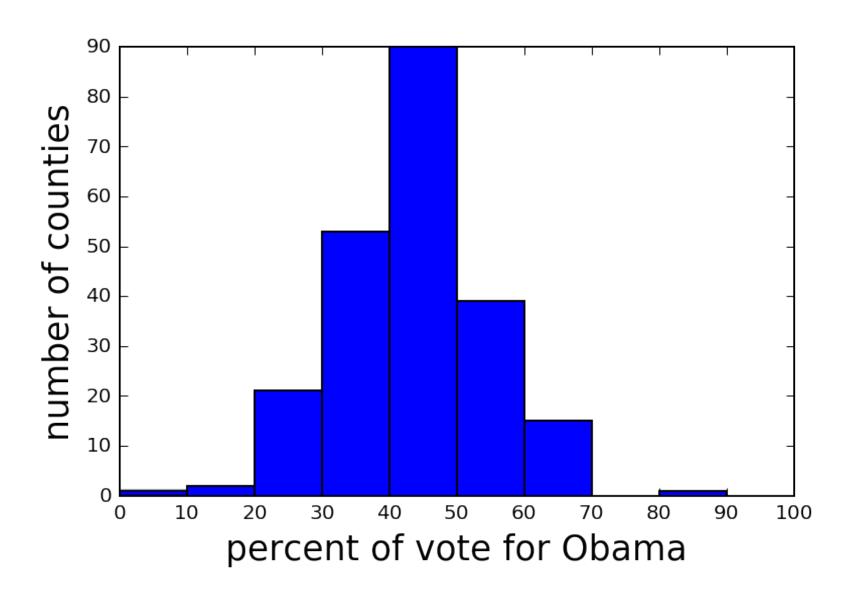






#### Histograms with different binning

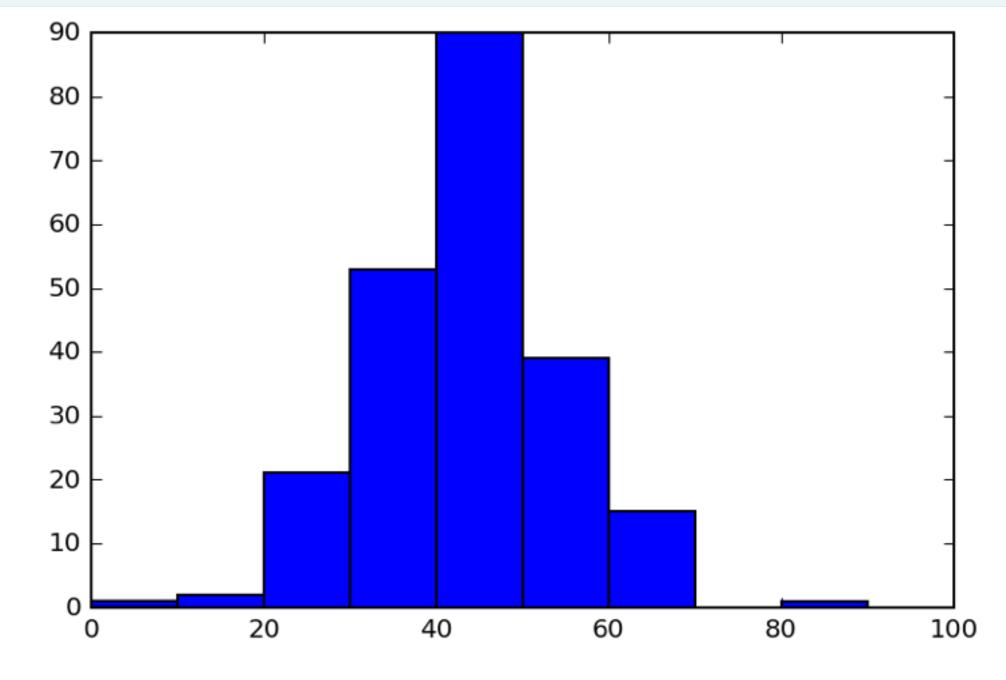








#### Setting the bins of a histogram



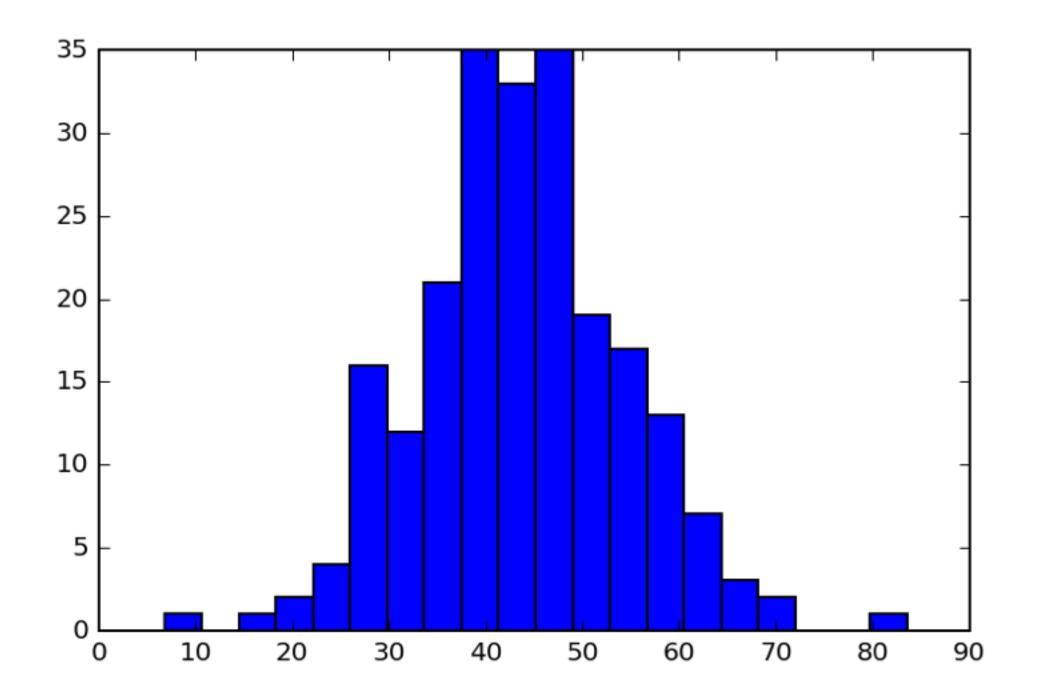




#### Setting the bins of a histogram

```
In [1]: _ = plt.hist(df_swing['dem_share'], bins=20)
```

In [2]: plt.show()







#### Seaborn

 An excellent Matplotlib-based statistical data visualization package written by Michael Waskom



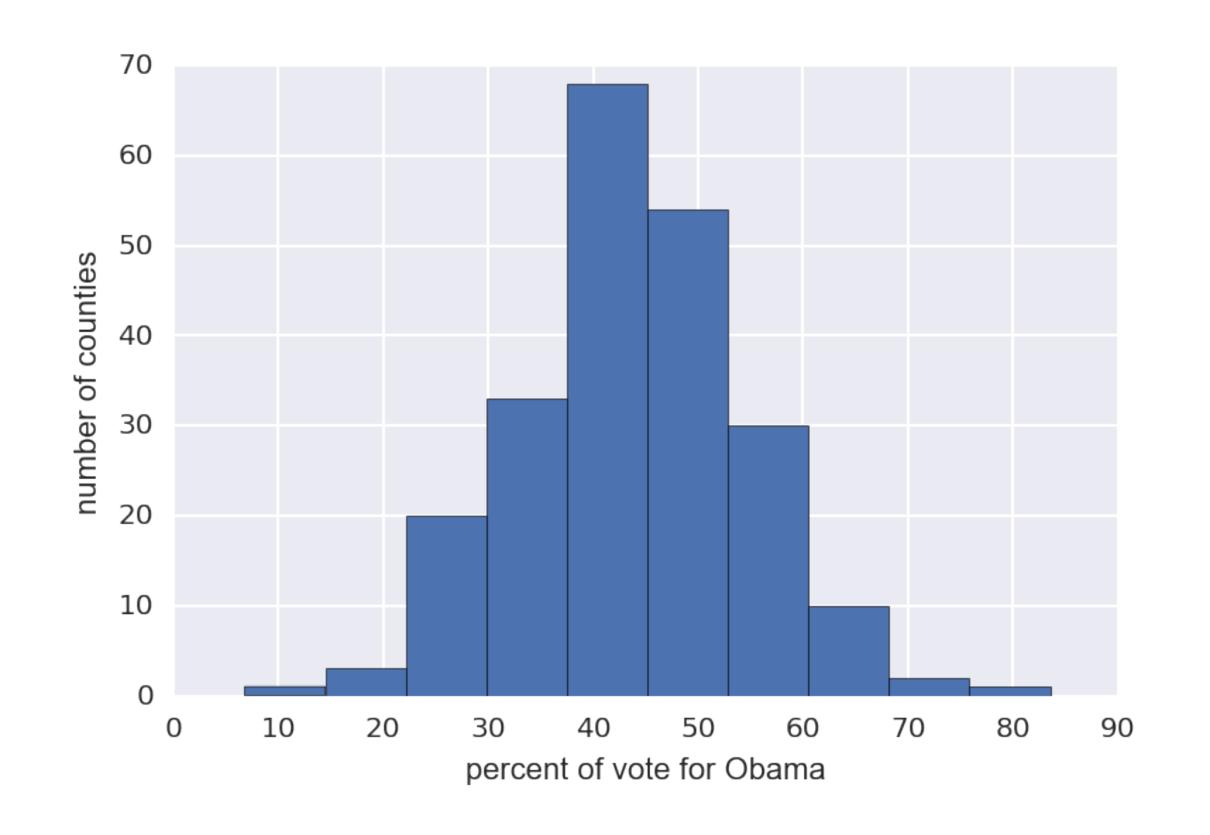
#### Setting Seaborn styling

```
In [1]: import seaborn as sns
In [2]: sns.set()
In [3]: _ = plt.hist(df_swing['dem_share'])
In [4]: _ = plt.xlabel('percent of vote for Obama')
In [5]: _ = plt.ylabel('number of counties')
In [6]: plt.show()
```





#### A Seaborn-styled histogram







### Let's practice!

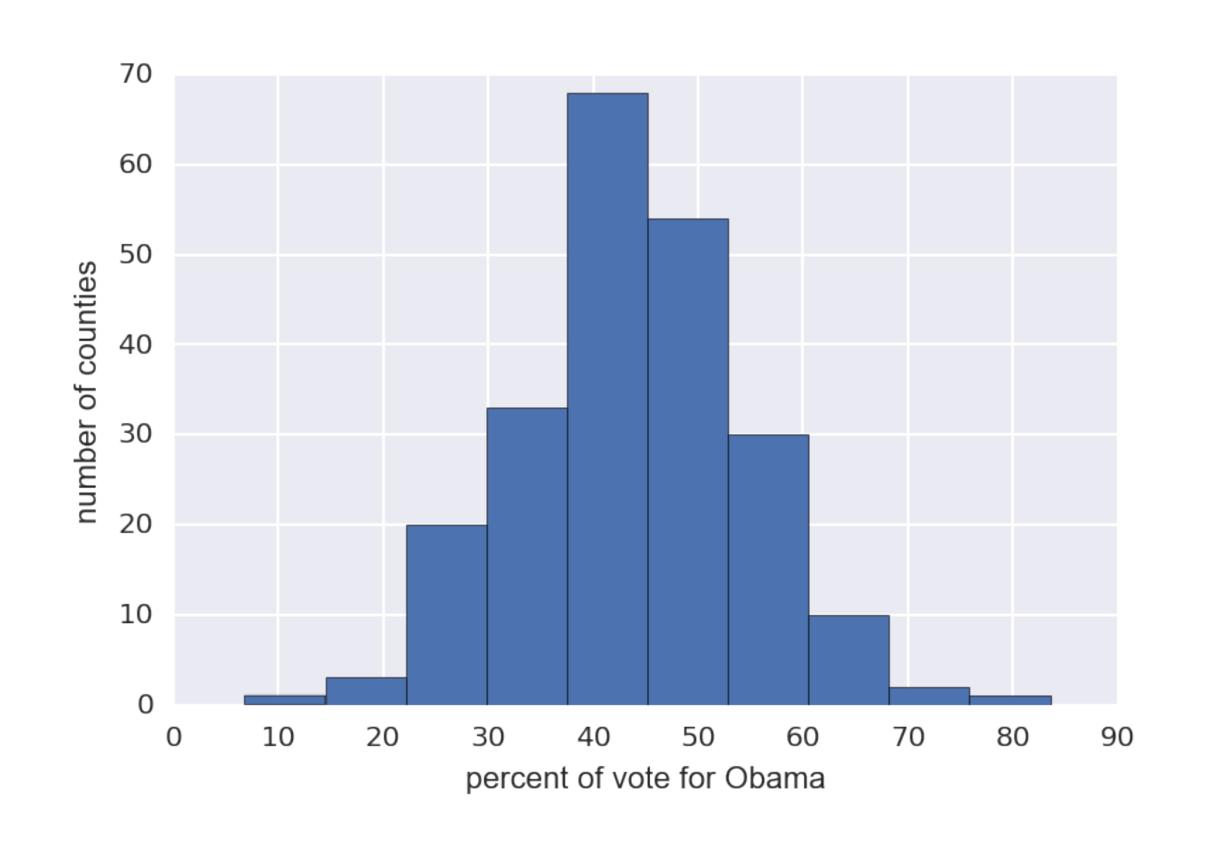




## Plot all of your data: Bee swarm plots

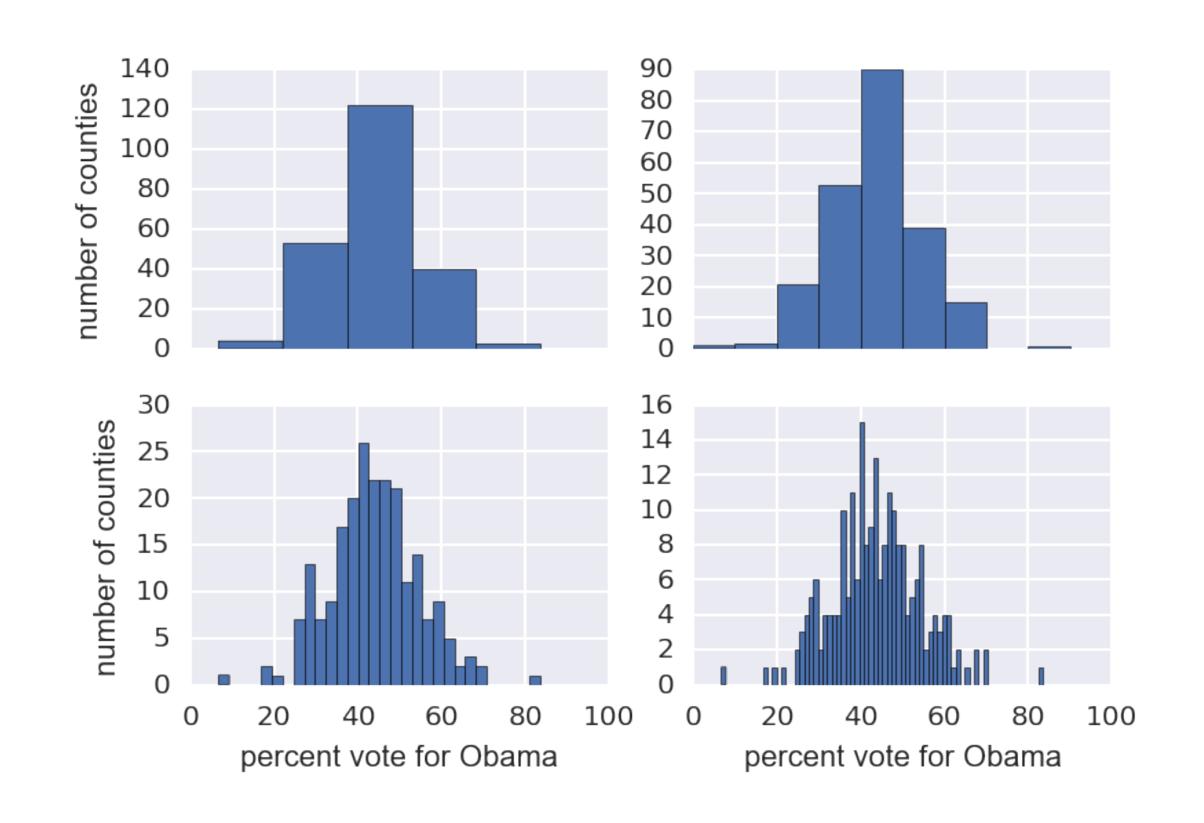
















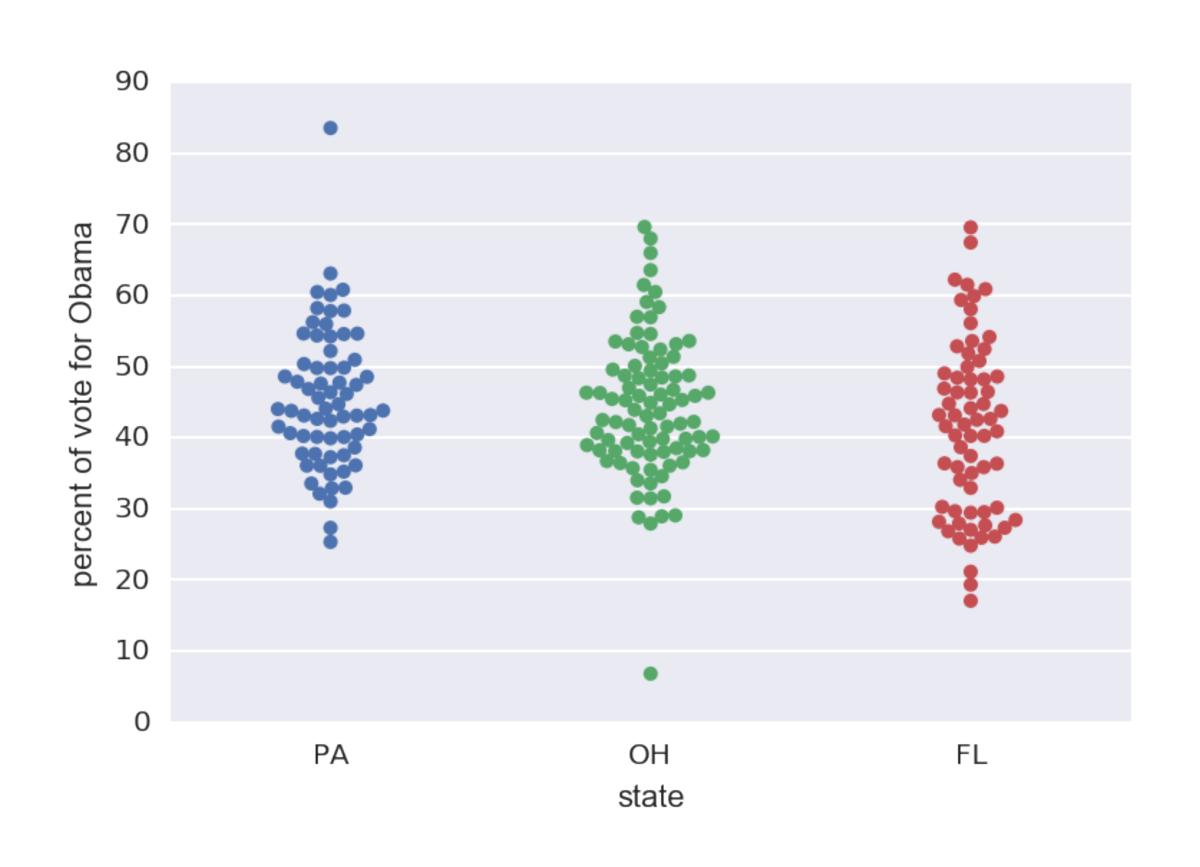
#### Binning bias

 The same data may be interpreted differently depending on choice of bins





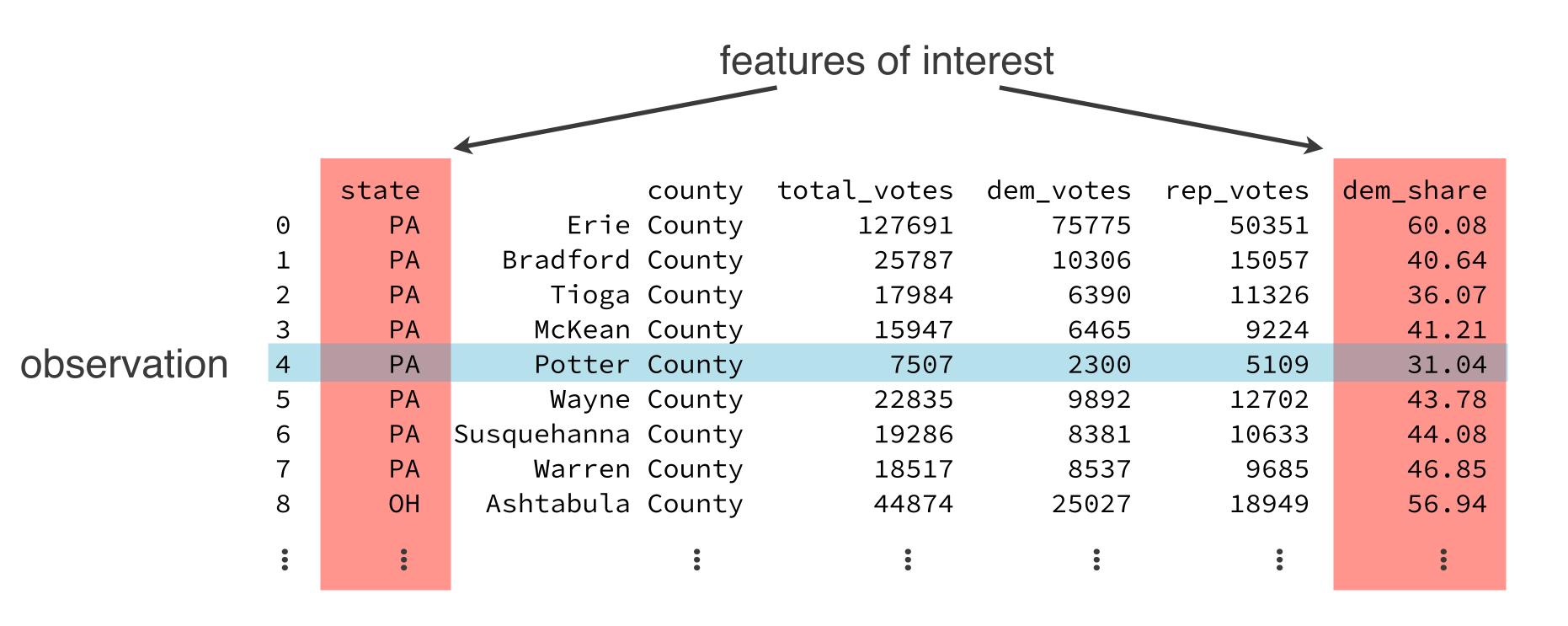
#### Bee swarm plot







#### Organization of the data frame





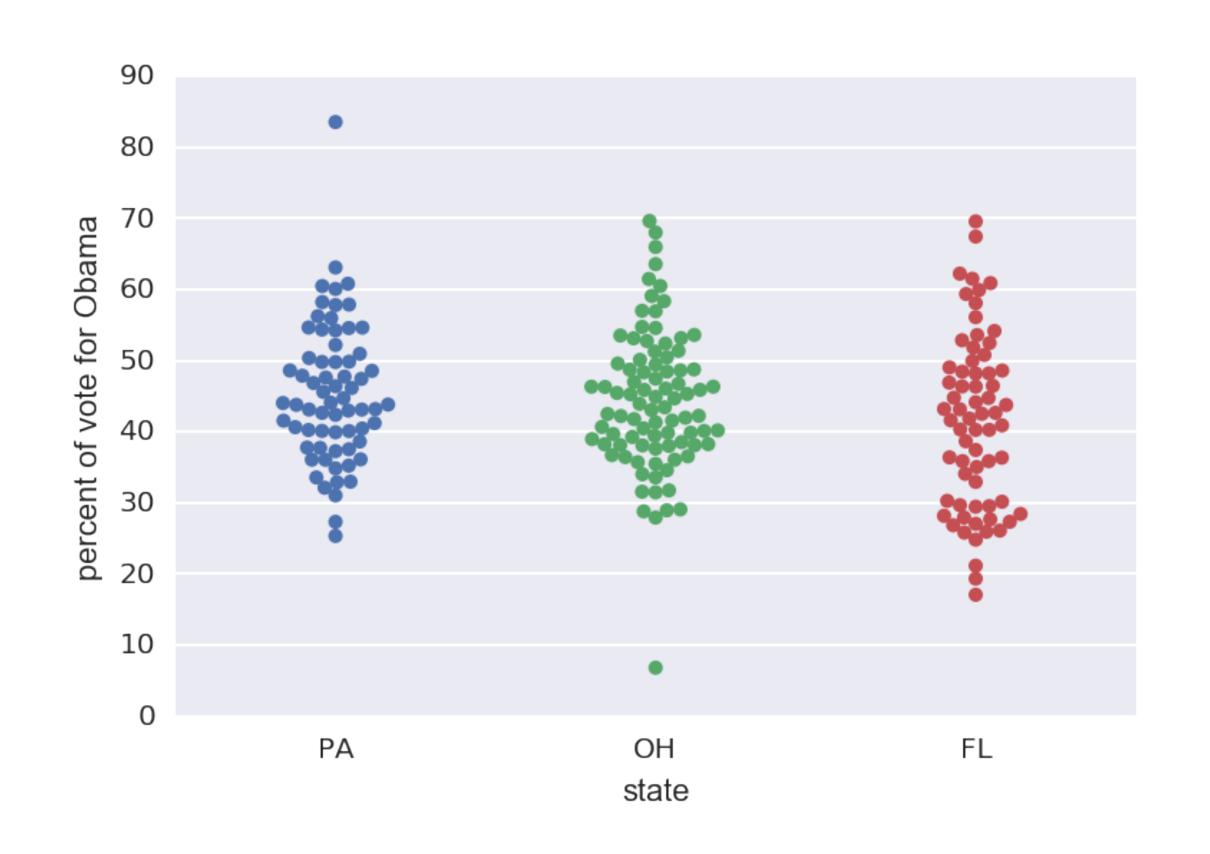


#### Generating a bee swarm plot

```
In [1]: _ = sns.swarmplot(x='state', y='dem_share', data=df_swing)
In [2]: _ = plt.xlabel('state')
In [3]: _ = plt.ylabel('percent of vote for Obama')
In [4]: plt.show()
```











### Let's practice!

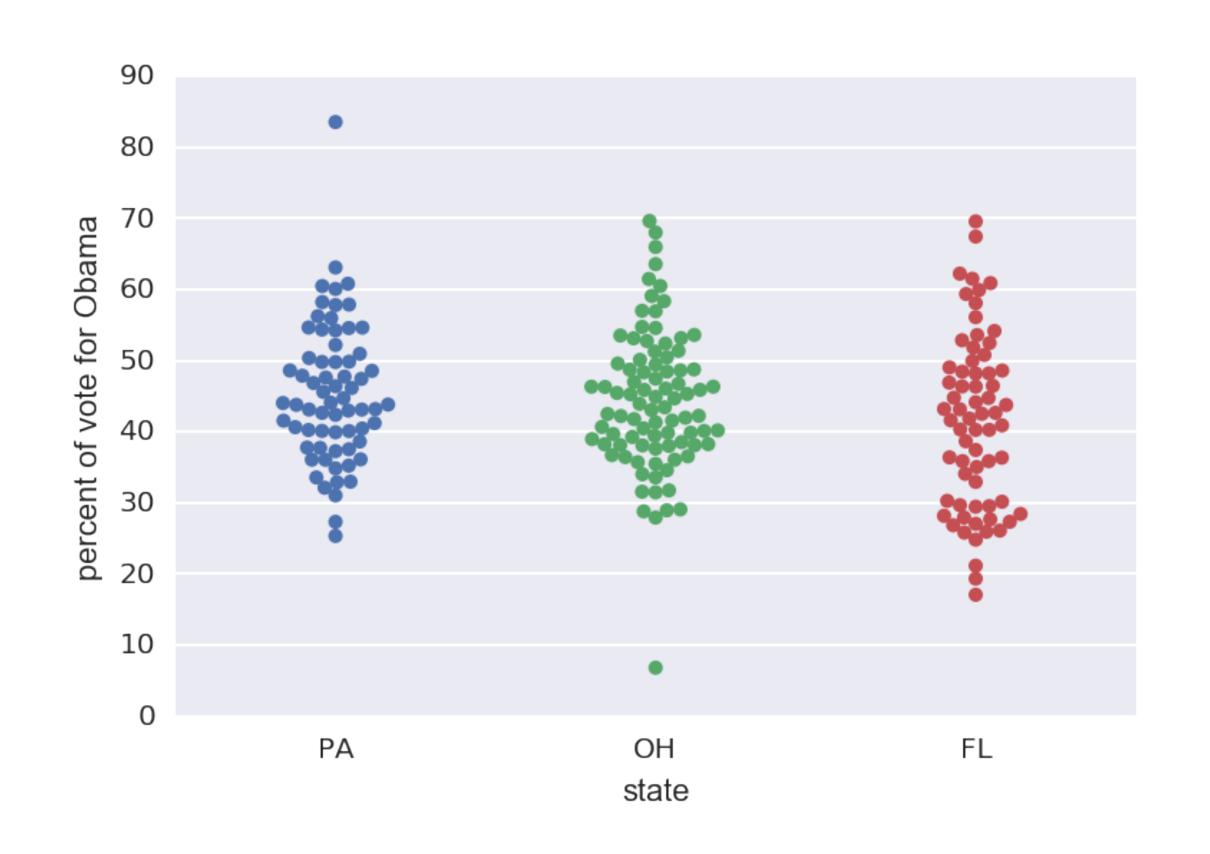




# Plot all of your data: ECDFs



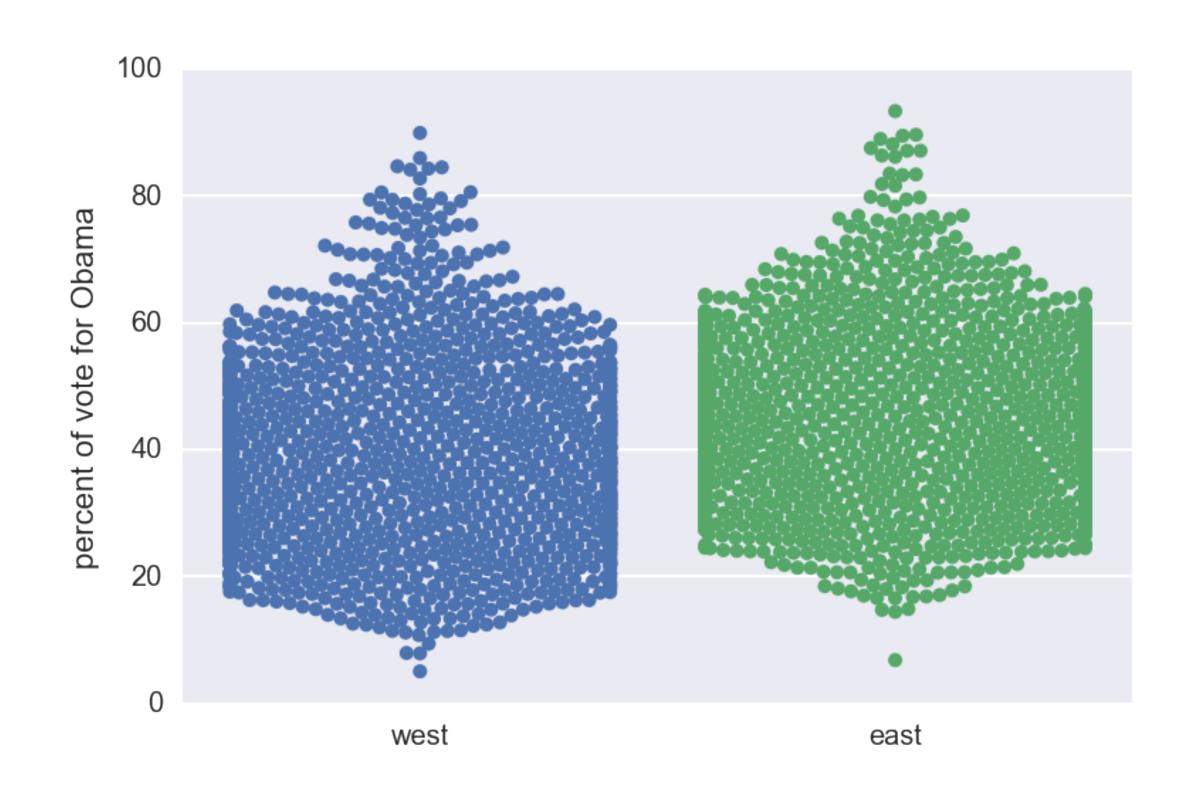








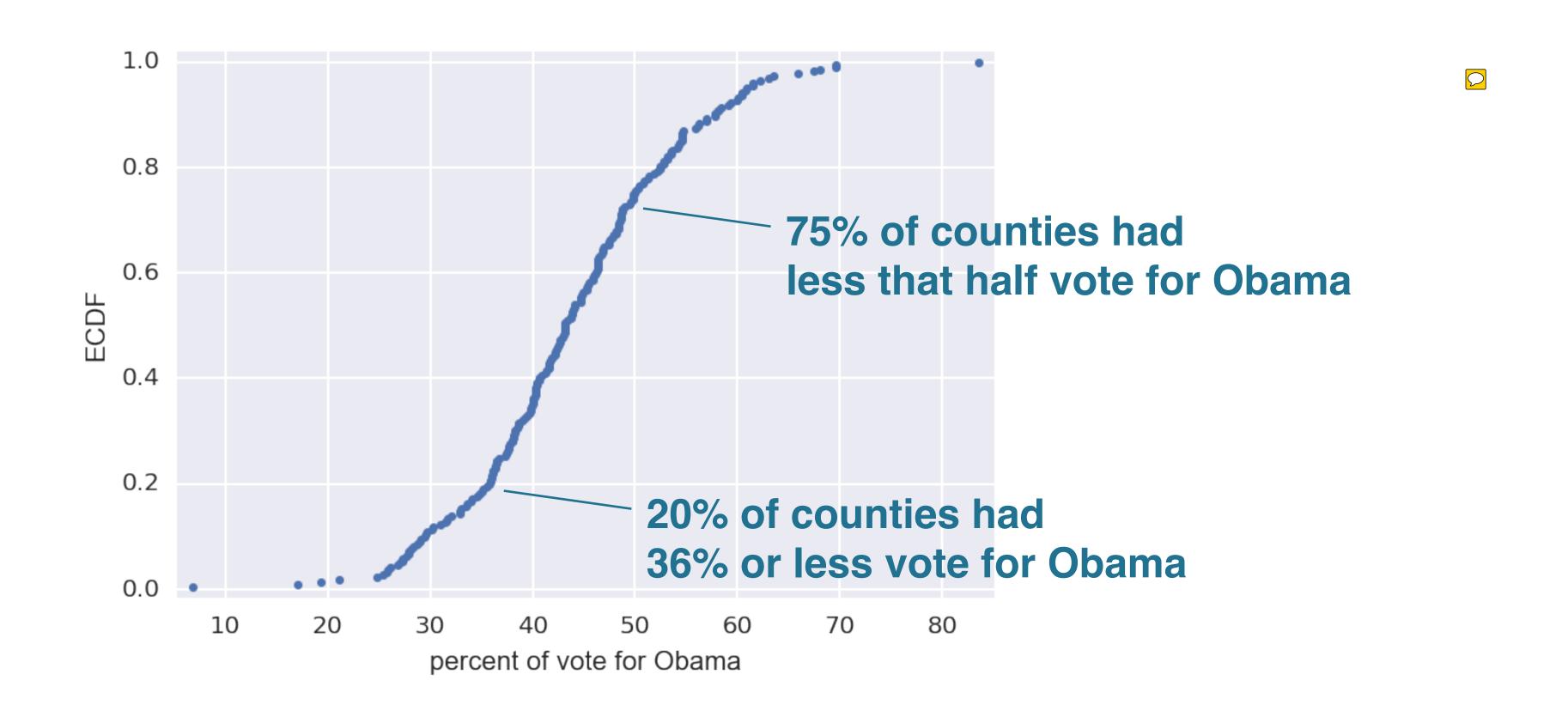
#### 2008 US election results: East and West







#### Empirical cumulative distribution function (ECDF)





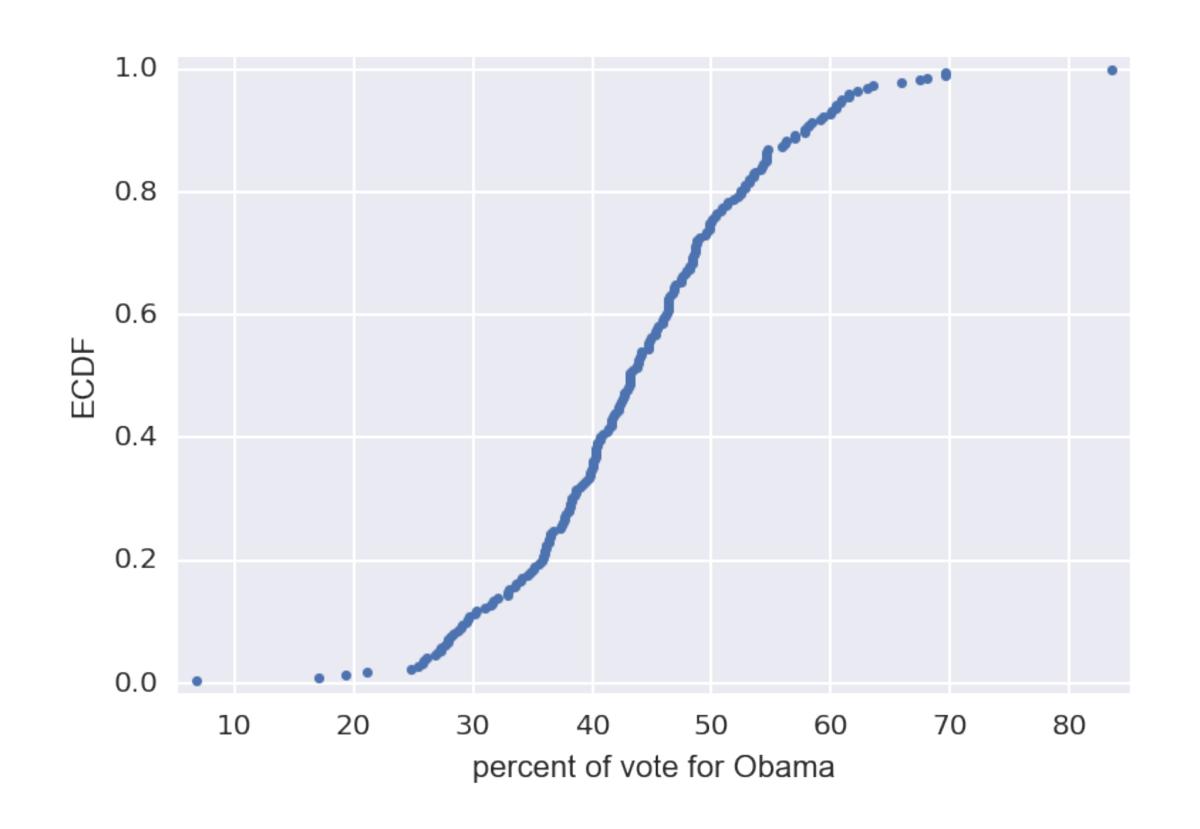
#### Making an ECDF

```
In [1]: import numpy as np
In [2]: x = np.sort(df_swing['dem_share'])
In [3]: y = np.arange(1, len(x)+1) / len(x)
In [4]: _ = plt.plot(x, y, marker='.', linestyle='none')
In [5]: _ = plt.xlabel('percent of vote for Obama')
In [6]: _ = plt.ylabel('ECDF')
In [7]: plt.margins(0.02) # Keeps data off plot edges
In [8]: plt.show()
```





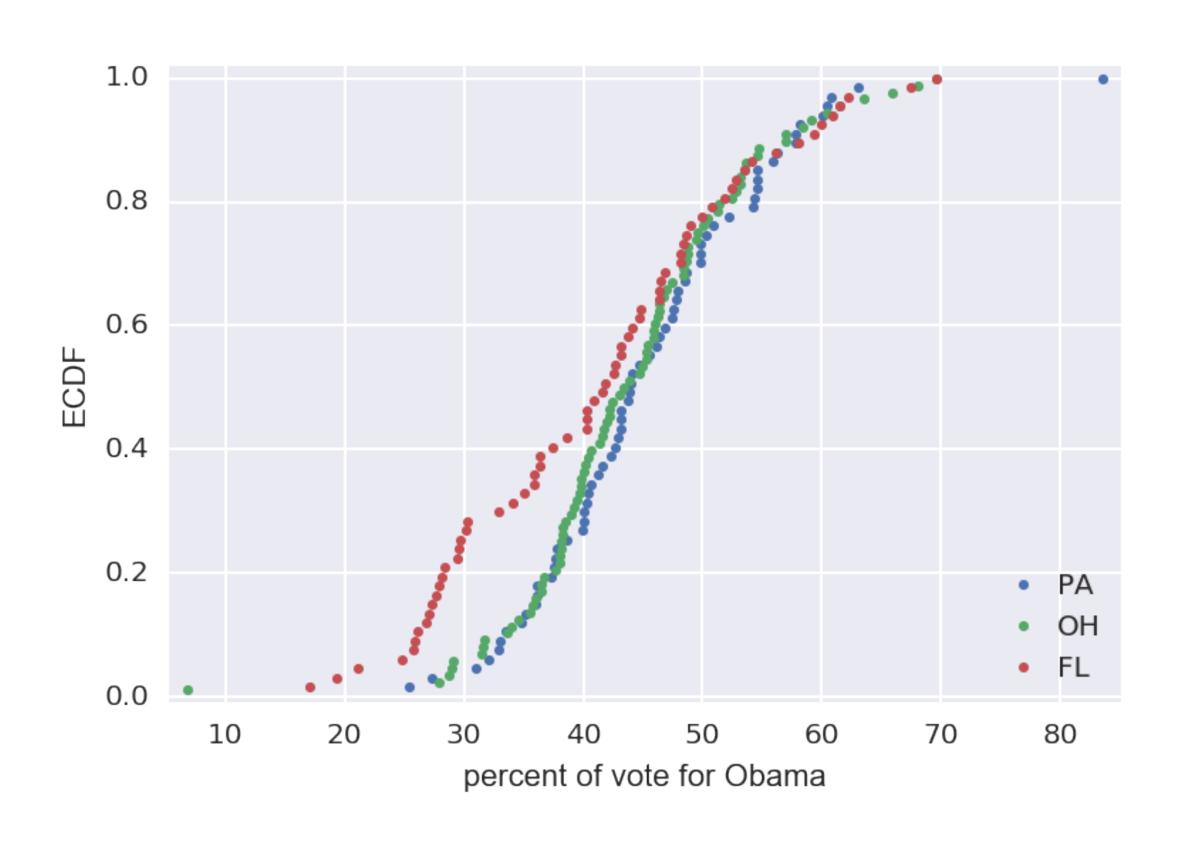
#### 2008 US swing state election ECDF







#### 2008 US swing state election ECDFs







### Let's practice!

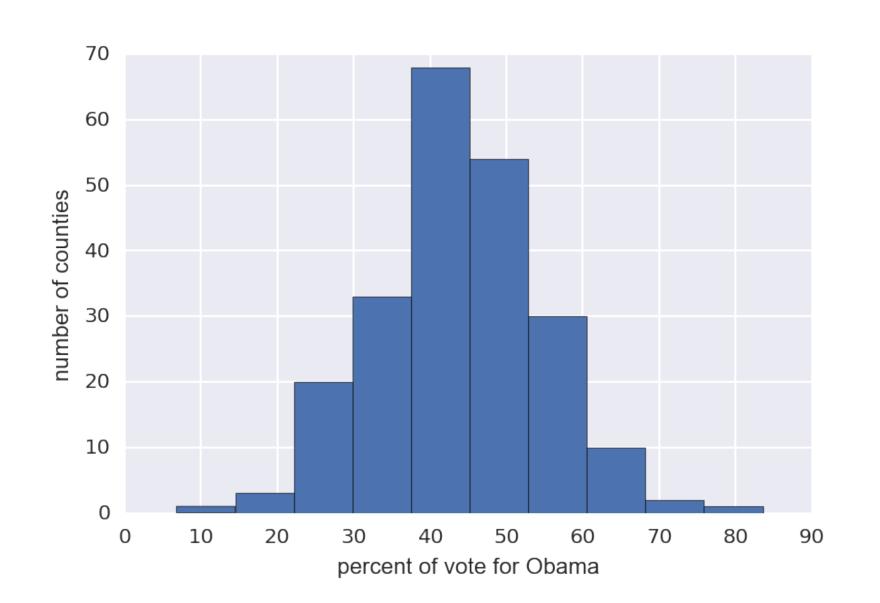


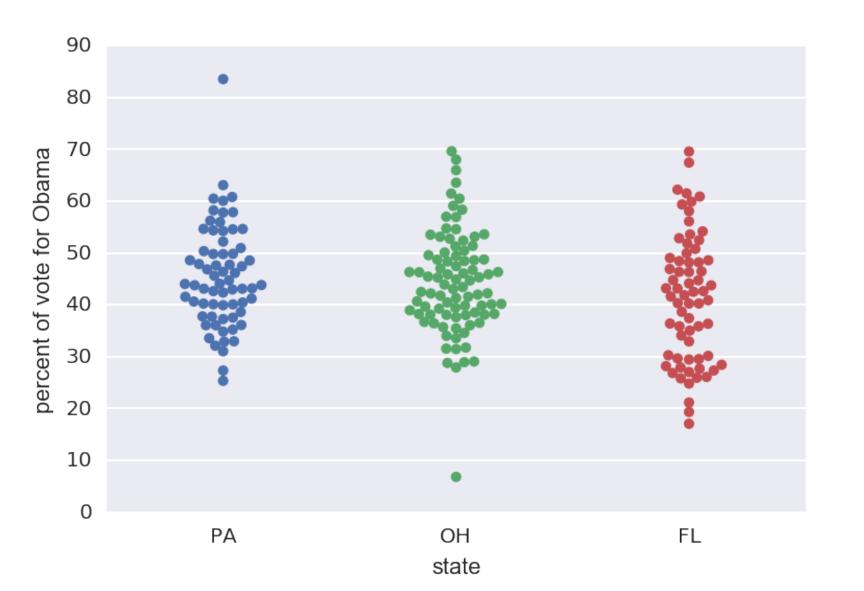


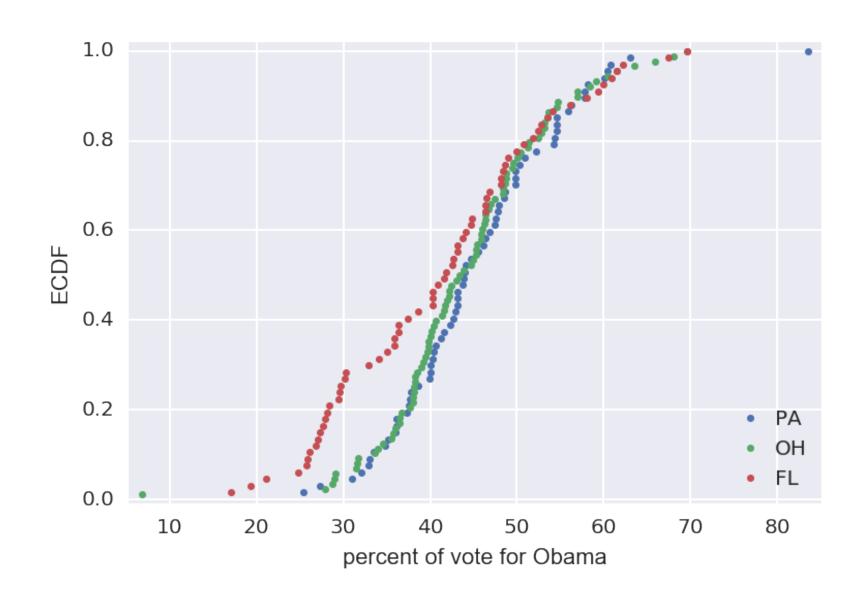
# Onward toward the whole story!











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—John Tukey



#### Coming up...

- Thinking probabilistically
- Discrete and continuous distributions
- The power of hacker statistics using np.random()





#### Let's get to work!