US Election

September 9, 2020

[5]: US_election_data_set

```
[5]:
              state state_abbreviation
                                                  county
                                                                 fips
                                                                             party \
     0
            Alabama
                                                               1001.0
                                                                         Democrat
                                      AL
                                                 Autauga
     1
            Alabama
                                      ΑL
                                                 Autauga
                                                               1001.0
                                                                         Democrat
     2
            Alabama
                                     ΑL
                                                               1003.0
                                                                         Democrat
                                                 Baldwin
     3
            Alabama
                                     AL
                                                 Baldwin
                                                               1003.0
                                                                         Democrat
     4
            Alabama
                                     AL
                                                 Barbour
                                                               1005.0
                                                                         Democrat
                                     WY
                                          Teton-Sublette 95600028.0
                                                                       Republican
     24606
            Wyoming
                                     WY
     24607
            Wyoming
                                           Uinta-Lincoln 95600027.0
                                                                       Republican
                                     WY
                                           Uinta-Lincoln
                                                                       Republican
     24608
            Wyoming
                                                           95600027.0
     24609
            Wyoming
                                     WY
                                           Uinta-Lincoln 95600027.0
                                                                       Republican
                                           Uinta-Lincoln 95600027.0
                                                                       Republican
     24610
            Wyoming
                  candidate
                              votes
                                     fraction_votes
     0
             Bernie Sanders
                                544
                                               0.182
     1
            Hillary Clinton
                               2387
                                               0.800
     2
             Bernie Sanders
                               2694
                                               0.329
     3
            Hillary Clinton
                               5290
                                               0.647
             Bernie Sanders
                                222
                                               0.078
```

```
Ted Cruz
                            0
                                         0.000
24606
          Donald Trump
                                         0.000
24607
                            0
           John Kasich
                            0
                                         0.000
24608
24609
           Marco Rubio
                            0
                                         0.000
              Ted Cruz
                           53
24610
                                         1.000
```

[24611 rows x 8 columns]

```
[6]: def data_shape(data, label):
    print('Rows number of ' + label + " is: ", data.shape[0])
    print('Columns number of ' + label + ' is: ', data.shape[1])

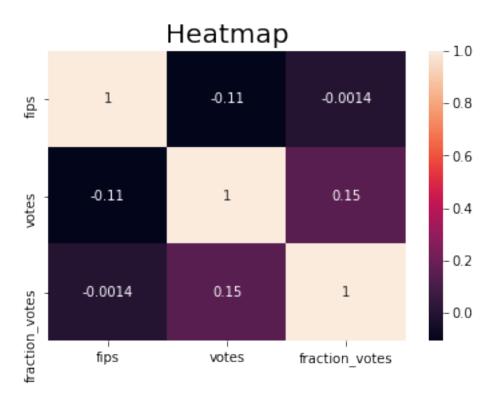
def data_columns(data):
    return list(data.columns)

def describe_data(data):
    return data.describe()

data_shape(US_election_data_set, 'US election data set')
    data_columns(US_election_data_set)
    describe_data(US_election_data_set)

sns.heatmap(US_election_data_set.corr(), annot = True)
plt.title('Heatmap', fontsize = 20)
plt.show()
```

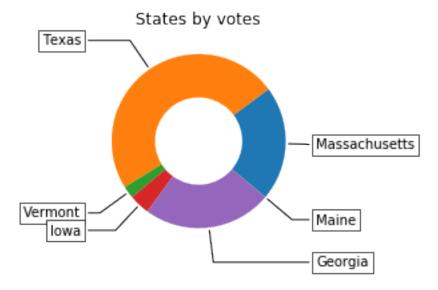
Rows number of US election data set is: 24611 Columns number of US election data set is: 8



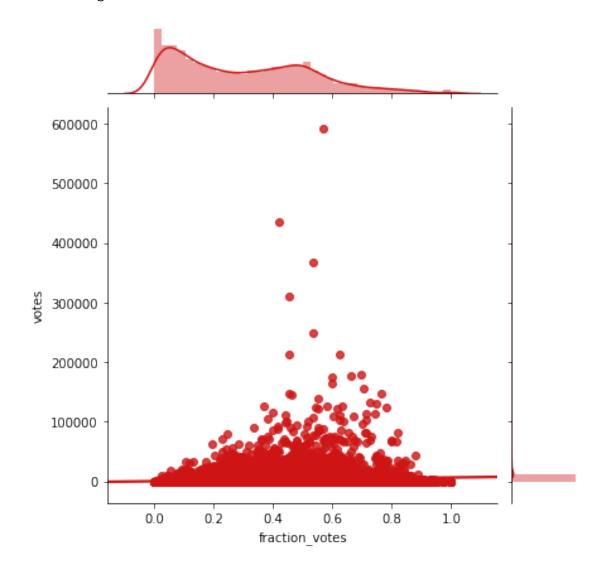
```
[7]: Massachusetts
                      2808
     Texas
                      1778
     Vermont
                      1722
     Iowa
                      1485
     Georgia
                      1113
     Maine
                       994
     Name: state, dtype: int64
[8]: def visualize_data_by_votes(data_frame, label):
         Massachusetts = data_frame[data_frame['state'] == 'Massachusetts']['votes'].
      ⇒sum()
         Texas = data_frame[data_frame['state'] == 'Texas']['votes'].sum()
         Vermont = data_frame[data_frame['state'] == 'Vermont']['votes'].sum()
         Iowa = data_frame[data_frame['state'] == 'Iowa']['votes'].sum()
         Georgia = data_frame[data_frame['state'] == 'Georgia']['votes'].sum()
         Maine = data_frame[data_frame['state'] == 'Maine']['votes'].sum()
         fig, ax = plt.subplots(figsize=(6, 3), subplot_kw=dict(aspect="equal"))
         months = ['Massachusetts',
                   'Texas',
```

[7]: US_election_data_set['state'].value_counts().head(6)

```
'Vermont',
              'Iowa',
              'Georgia',
              'Maine']
   data = [Massachusetts, Texas, Vermont, Iowa, Georgia, Maine]
   wedges, texts = ax.pie(data, wedgeprops=dict(width=0.5), startangle=-40)
   bbox_props = dict(boxstyle="square,pad=0.3", fc="w", ec="k", lw=0.72)
   kw = dict(arrowprops=dict(arrowstyle="-"),
              bbox=bbox_props, zorder=0, va="center")
   for i, p in enumerate(wedges):
        ang = (p.theta2 - p.theta1) / 2. + p.theta1
        y = np.sin(np.deg2rad(ang))
        x = np.cos(np.deg2rad(ang))
       horizontalalignment = {-1: "right", 1: "left"}[int(np.sign(x))]
        connectionstyle = "angle,angleA=0,angleB={}".format(ang)
       kw["arrowprops"].update({"connectionstyle": connectionstyle})
        ax.annotate(months[i], xy=(x, y), xytext=(1.35 * np.sign(x), 1.4 * y),
                    horizontalalignment=horizontalalignment, **kw)
   ax.set_title(label)
   plt.show()
visualize_data_by_votes(US_election_data_set, "States by votes")
```

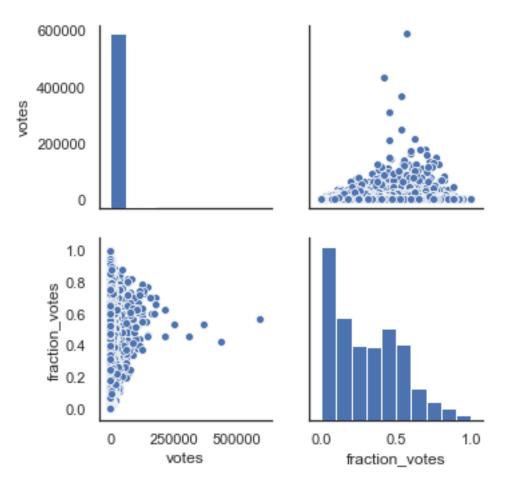


[9]: <seaborn.axisgrid.JointGrid at 0x2d4fb147198>



```
[10]: sns.set(style="white")
sns.pairplot(US_election_data_set[['votes','fraction_votes']])
```

[10]: <seaborn.axisgrid.PairGrid at 0x2d4fc59f978>



```
[11]: from sklearn import preprocessing import pandas as pd
```

```
import pandas as pd
import numpy as np

def pre_processing(original_data, supervised_or_unsupervised):
    def remove_columns(data_frame, col_name):
        try:
        columns = list(data_frame.columns)
        include_columns = [x for x in columns if x not in col_name]
        new_data_frame = data_frame[include_columns]
```

```
return new_data_frame
       except:
           print('Something got wrong - remove_columns')
   name = original_data.isnull().sum().where(lambda x: x > 2500).dropna().
→keys().to_list()
   original_data = remove_columns(original_data, name)
   original_data = remove_columns(original_data, 'state')
   original_data = remove_columns(original_data, 'fips')
   if supervised_or_unsupervised == 'supervised':
       data_frame_for_supervised = original_data
       def encoders(data_frame, label):
           try:
               encoders = {
                   label: preprocessing.LabelEncoder()
               data_frame[label] = encoders[label].
→fit_transform(data_frame[label].astype(str))
           except:
               print('Something got wrong - encoders')
       encoders(data_frame_for_supervised, 'party')
       encoders(data_frame_for_supervised, 'candidate')
       encoders(data frame for supervised, 'county')
       encoders(data_frame_for_supervised, 'state_abbreviation')
       return data_frame_for_supervised
   if supervised_or_unsupervised == 'unsupervised':
       data_frame_for_unsupervised = original_data
       def get_dummies(data_frame):
           try:
               data_frame = pd.get_dummies(data_frame)
               return data_frame
           except:
               print('Something got wrong - get_dummies')
       data_frame_for_unsupervised = get_dummies(data_frame_for_unsupervised)
       return data_frame_for_unsupervised
   return original_data
```

[12]: US_election_data_set_for_supervised

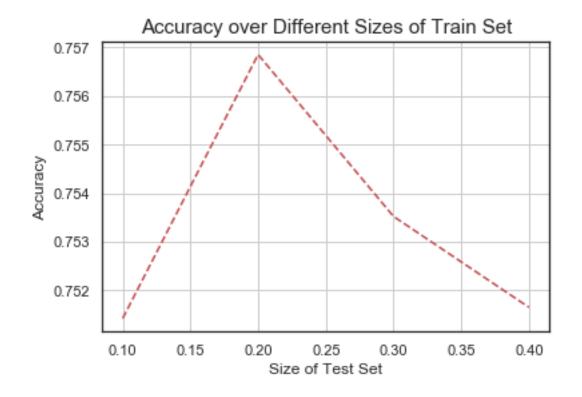
[12]:	state_abbreviation	county	party	candidate	votes	fraction_votes
0	1	103	0	3	544	0.182
1	1	103	0	7	2387	0.800
2	1	114	0	3	2694	0.329
3	1	114	0	7	5290	0.647
4	1	127	0	3	222	0.078
•••	•••		•••	•••		•••
24606	48	2312	1	15	0	0.000
24607	48	2383	1	6	0	0.000
24608	48	2383	1	9	0	0.000
24609	48	2383	1	10	0	0.000
24610	48	2383	1	15	53	1.000

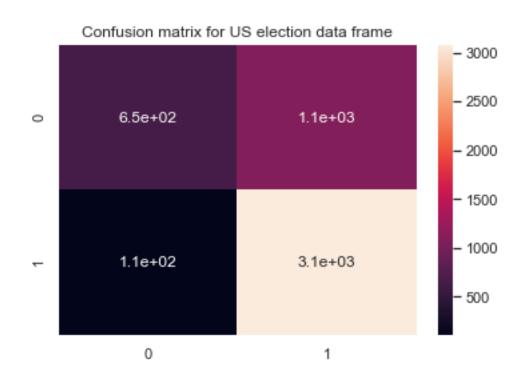
[24611 rows x 6 columns]

```
[13]: import numpy as np
      from sklearn.naive_bayes import GaussianNB
      from sklearn.model_selection import train_test_split
      from sklearn.metrics import accuracy_score, confusion_matrix
      import matplotlib.pyplot as plt
      import seaborn as sns
      def naive_bayes_algorithm(data_frame):
          cols = list(data_frame.columns)
          cols.remove('party')
          X = data_frame[cols].copy()
          y = data_frame['party'].copy()
          def split_test_train(X, y, test_size):
              try:
                  return train_test_split(X, y, test_size=test_size, random_state=0)
              except:
                 print('Something got wrong - split_test_train')
          def create_naive_bayes_classifier(X, y):
              try:
                  model = GaussianNB()
```

```
model.fit(X, y)
            return model
        except:
            print('Something got wrong - create_naive_bayes_classifier')
    accuracy = []
    for ratio in np.arange(0.1, 0.5, 0.1):
        X_train, X_test, y_train, y_test = split_test_train(X, y, test_size =_
 →ratio)
        model = create_naive_bayes_classifier(X_train, y_train)
        y_pred = model.predict(X_test)
        accuracy.append(accuracy_score(y_test, y_pred))
    best_accuracy = 0
    x = 0
    split = None
    for i in accuracy:
        x += 1
        if i > best_accuracy:
            best_accuracy = i
            split = "0." + str(x)
    print('The best accuracy is: ' + str(best_accuracy) +'\nThe size of the
 →test team is: ' + str(split))
    ratios = np.arange(0.1, 0.5, 0.1)
    plt.grid(True)
    plt.plot(ratios, accuracy, 'r--')
    plt.xlabel('Size of Test Set')
    plt.ylabel('Accuracy')
    plt.title('Accuracy over Different Sizes of Train Set', fontsize=15)
    plt.show()
    X_train, X_test, y_train, y_test = split_test_train(X, y, test_size=0.2)
    model = create_naive_bayes_classifier(X_train, y_train)
    y_pred = model.predict(X_test)
    confusion_matrix(y_test, y_pred)
    sns.heatmap(confusion_matrix(y_test, y_pred), annot=True)
    plt.title('Confusion matrix for US election data frame')
    plt.show()
naive_bayes_algorithm(US_election_data_set_for_supervised)
```

The best accuracy is: 0.7568555758683729
The size of the test team is: 0.2





[14]: US_election_data_set_for_unsupervised

[14]:		votes	fraction_votes	state_abbreviation_AK	${\tt state_abbreviation_AL}$	\
	0	544	0.182	0	1	
	1	2387	0.800	0	1	
	2	2694	0.329	0	1	
	3	5290	0.647	0	1	
	4	222	0.078	0	1	
		•••	•••	•••		
	24606	0	0.000	0	0	
	24607	0	0.000	0	0	
	24608	0	0.000	0	0	
	24609	0	0.000	0	0	
	24610	53	1.000	0	0	
		state_	abbreviation_AR	${\tt state_abbreviation_AZ}$	${\tt state_abbreviation_CA}$	\
	0		0	0	0	
	1		0	0	0	
	2		0	0	0	
	3		0	0	0	
	4		0	0	0	
			•••			
	24606		0	0	0	
	24607		0	0	0	
	24608		0	0	0	
	24609		0	0	0	
	24610		0	0	0	
		state_	abbreviation_CO	${\tt state_abbreviation_CT}$	${\tt state_abbreviation_DE}$	\
	0		0	0	0	
	1		0	0	0	
	2		0	0	0	
	3		0	0	0	
	4		0	0	0	
			•••	•••	•••	
	24606		0	0	0	
	24607		0	0	0	
	24608		0	0	0	
	24609		0	0	0	
	24610		0	0	0	
		can	didate_Donald Tr	ump candidate_Hillary		
	0	•••		0	0	
	1	•••		0	1	
	2	•••		0	0	
	3	•••		0	1	
	4	•••		0	0	

24606 24607 24608 24609 24610	 	 0 1 0 0 0	 0 0 0 0 0	
0 1 2 3 4 24606 24607 24608 24609 24610	candidate_Jeb Bush	candidate_John Kasich 0 0 0 0 0 0 0 1 0 0 0 0 0 0 0 0 0		Rubio \
0 1 2 3 4 24606 24607 24608 24609 24610	candidate_Martin O'N	Malley candidate_Mike 0 0 0 0 0 0 0 0 0 0 0 0 0	Huckabee \	
0 1 2 3 4 24606 24607 24608 24609 24610	candidate_Rand Paul 0 0 0 0 0 0 0 0 0 0 0 0 0 0		rum candidate_Ted 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0	Cruz 0 0 0 0 0 1 0 0 1

[24611 rows x 2702 columns]

```
[15]: import pandas as pd
      import numpy as np
      import matplotlib.pyplot as plt
      import seaborn as sns
      %matplotlib inline
      from sklearn.model_selection import train_test_split
      X = US_election_data_set_for_supervised.drop('party',axis=1)
      y = US_election_data_set_for_supervised['party']
      X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.30)
[16]: from sklearn.tree import DecisionTreeClassifier
      dtree = DecisionTreeClassifier()
      dtree.fit(X_train,y_train)
[16]: DecisionTreeClassifier(class_weight=None, criterion='gini', max_depth=None,
                             max_features=None, max_leaf_nodes=None,
                             min_impurity_decrease=0.0, min_impurity_split=None,
                             min_samples_leaf=1, min_samples_split=2,
                             min_weight_fraction_leaf=0.0, presort=False,
                             random_state=None, splitter='best')
[17]: | predictions = dtree.predict(X_test)
[18]: from sklearn.metrics import classification_report,confusion_matrix
[19]: print(classification_report(y_test,predictions))
                   precision
                                recall f1-score
                                                    support
                0
                        1.00
                                   1.00
                                             1.00
                                                       2591
                1
                         1.00
                                   1.00
                                             1.00
                                                       4793
                                                       7384
                                             1.00
         accuracy
        macro avg
                         1.00
                                   1.00
                                             1.00
                                                       7384
                                             1.00
     weighted avg
                         1.00
                                   1.00
                                                       7384
[20]: print(confusion_matrix(y_test,predictions))
     [[2588
               3]
          5 4788]]
[23]: from IPython.display import Image
      from sklearn.externals.six import StringIO
```

```
from sklearn.tree import export_graphviz
      import pydot
      features = list(US_election_data_set_for_supervised.columns[1:])
      features
[23]: ['county', 'party', 'candidate', 'votes', 'fraction_votes']
```

```
[24]: dot_data = StringIO()
      export_graphviz(dtree,_
      →out_file=dot_data,feature_names=features,filled=True,rounded=True)
      graph = pydot.graph_from_dot_data(dot_data.getvalue())
      Image(graph[0].create_png())
```

[24]:

