## **Assignment-15-Random Forest (Company Data)**

Random Forest

Assignment

About the data:Let's consider a company dataset with around 10 variables and 400 records.

Sales -- Unit sales (in thousands) at each location 
Competitor Price -- Price charged by competitor at each location 
Income -- Community income level (in thousands of dollars) 
Advertising -- Local advertising budget for company at each location (in thousands of dollars) 
Price -- Price company charges for car seats at each site 
Shelf Location at stores -- A factor with levels Bad, Good and Medium indicating the quality of the shelving location for the car seats at each site 
Age -- Average age of the local population 
Education -- Education level at each location 
Urban -- A factor with levels No and Yes to indicate whether the store is in an urban or rural location 
US -- A factor with levels No and Yes to indicate whether the store is in the US or not The company dataset looks like this:

Problem Statement:

A cloth manufacturing company is interested to know about the segment or attributes causes high sale. Approach - A Random Forest can be built with target variable Sales (we will first convert it in categorical variable) & all other variable will be independent in the analysis.

## **Import Libraries**

```
In [131]:
    import pandas as pd
    import numpy as np
    import matplotlib.pyplot as pd
    import seaborn as sns
    // import seaborn as sns
    // from sklearn.preprocessing import LabelEncoder # for train test splitting,
    // from sklearn.model_selection import train_test_split # for decision tree object
    // from sklearn.tree import DecisionTreeClassifier # for checking testing results
    from sklearn.tree import classification_report,confusion_matrix # for visualizing tree
    import warnings
    import warnings
    warnings.filterwarnings('ignore')
```

UsageError: unrecognized arguments: #for encoding

## **Import Data**

```
In [132]:
             1 # Pandas is used for data manipulation
               import pandas as pd
                # Read in data and display first 5 rows
             4 features = pd.read_csv('Company_Data.csv')
             5 features.head(5)
Out[132]:
              Sales CompPrice Income Advertising Population Price ShelveLoc Age Education Urban US
                9.50
                                               11
                                                                              42
                                                                                        17
            0
                           138
                                   73
                                                        276
                                                              120
                                                                        Bad
                                                                                              Yes
                                                                                                  Yes
            1 11.22
                           111
                                   48
                                               16
                                                               83
                                                                              65
                                                        260
                                                                       Good
                                                                                        10
                                                                                              Yes Yes
              10.06
                           113
                                   35
                                               10
                                                        269
                                                               80
                                                                     Medium
                                                                              59
                                                                                        12
                                                                                              Yes Yes
                           117
                                                               97
                7.40
                                   100
                                                        466
                                                                     Medium
                                                                              55
                                                                                        14
                                                                                              Yes
                                                                                                  Yes
                4.15
                           141
                                                        340
                                                                        Bad
                                                                                        13
                                                                                              Yes
```

```
In [133]:
           1 # getting information of dataset
            2 features.info()
          <class 'pandas.core.frame.DataFrame'>
          RangeIndex: 400 entries, 0 to 399
          Data columns (total 11 columns):
                            Non-Null Count
           #
               Column
                                            Dtype
               Sales
                            400 non-null
                                            float64
           0
               {\tt CompPrice}
                            400 non-null
                                            int64
               Income
                            400 non-null
                                            int64
               Advertising 400 non-null
                                            int64
                            400 non-null
               Population
                                            int64
               Price
                            400 non-null
                                            int64
               ShelveLoc
                            400 non-null
                                            object
                            400 non-null
                                            int64
               Age
           8
               Education
                            400 non-null
                                            int64
           9
               Urban
                            400 non-null
                                            object
           10 US
                            400 non-null
                                            object
          dtypes: float64(1), int64(7), object(3)
          memory usage: 34.5+ KB
In [134]: 1 print('The Shape of our features is:',features.shape)
          The Shape of our features is: (400, 11)
In [135]: 1 features.isnull().any()
Out[135]: Sales
                         False
          CompPrice
                         False
          Income
                         False
          Advertising
                         False
          Population
                         False
          Price
                         False
          ShelveLoc
                         False
          Age
                         False
          Education
                         False
          Urban
                         False
                         False
          dtype: bool
```

```
In [136]:
             1 # lets plot pair plot to visualise the attributes all at once.
             2 sns.pairplot(data=features, hue = 'ShelveLoc')
Out[136]: <seaborn.axisgrid.PairGrid at 0x227371a58b0>
            S 7.5
              5.0
                                                                                                     50 100 150 200
Price
             # Creating dummy variables dropping first dummy variable
df=pd.get_dummies(features,columns=['Urban','US'], drop_first=True)
In [137]:
            1 print(df.head())
In [138]:
                       CompPrice Income Advertising
                                                          Population
                                                                        Price ShelveLoc
                                                                                           Age
               9.50
                             138
                                       73
                                                      11
                                                                  276
                                                                          120
                                                                                     Bad
                                                                                            42
                                       48
                                                                           83
                                                                                            65
              11.22
                             111
                                                      16
                                                                  260
                                                                                     Good
               10.06
                             113
                                       35
                                                      10
                                                                  269
                                                                           80
                                                                                            59
                                                                                  Medium
                                                                           97
                                                                                            55
           3
                7.40
                             117
                                      100
                                                       4
                                                                  466
                                                                                  Medium
                4.15
                             141
                                       64
                                                                  340
                                                                          128
                                                                                      Bad
                                                                                            38
               Education Urban_Yes
                                      US Yes
           0
                       17
                       10
                      12
                                             1
           3
4
                       14
                                             1
                                    1
1
                                             0
                       13
```

```
In [139]:
            1 from sklearn.metrics import f1_score
            2 from sklearn.model_selection import train_test_split
In [140]:
            1 df['ShelveLoc']=df['ShelveLoc'].map({'Good':1,'Medium':2,'Bad':3})
In [141]:
            1 print(df.head())
              Sales CompPrice Income Advertising Population
                                                                   Price ShelveLoc
                                                                                      Age
              9.50
                                     73
                                                              276
                                                                     120
                                                                                       42
                           138
                                                  11
                                                                                   3
             11.22
                           111
                                     48
                                                   16
                                                              260
                                                                      83
                                                                                   1
                                                                                       65
          2
             10.06
                           113
                                     35
                                                   10
                                                              269
                                                                      80
                                                                                   2
                                                                                       59
                                    100
                                                                      97
                                                                                       55
              7.40
                           117
                                                   4
                                                              466
                                                                                   2
          4
                                                    3
                                                                     128
                                                                                       38
                           141
                                                              340
               4.15
                                     64
              Education Urban_Yes US_Yes
                     17
          1
                     10
                                  1
                                          1
          2
                     12
                                  1
                                          1
          3
                     14
                                  1
                                          1
          4
                     13
                                          0
In [142]:
            1 x=df.iloc[:,0:6]
               y=df['ShelveLoc']
            3
               x
Out[142]:
                Sales CompPrice Income Advertising Population Price
             0
                 9.50
                            138
                                    73
                                               11
                                                        276
                                                              120
             1
                11.22
                            111
                                    48
                                               16
                                                        260
                                                              83
             2
                10.06
                            113
                                    35
                                               10
                                                        269
                                                              80
                                                              97
                 7.40
                            117
                                   100
                                                        466
                 4.15
                            141
                                    64
                                               3
                                                        340
                                                              128
           395
                12.57
                            138
                                   108
                                               17
                                                        203
                                                              128
           396
                 6.14
                            139
                                    23
                                               3
                                                         37
                                                              120
           397
                 7.41
                            162
                                    26
                                               12
                                                        368
                                                              159
                                               7
           398
                 5.94
                            100
                                    79
                                                        284
                                                              95
           399
                 9.71
                            134
                                    37
                                               0
                                                         27
                                                              120
           400 rows × 6 columns
            1 y
In [143]:
Out[143]: 0
                  3
                  1
          2
                  2
                  2
          3
          4
                  3
           395
                  1
          396
                  2
           397
                  2
           398
           399
           Name: ShelveLoc, Length: 400, dtype: int64
In [144]: 1 df['ShelveLoc'].unique()
Out[144]: array([3, 1, 2], dtype=int64)
In [145]:
            1 df.ShelveLoc.value_counts()
Out[145]: 2
                219
                 96
                 85
          1
          Name: ShelveLoc, dtype: int64
```

```
1 colnames = list(df.columns)
In [146]:
             2 colnames
Out[146]: ['Sales',
              CompPrice',
             'Income',
             'Advertising',
             'Population',
             'Price',
             'ShelveLoc',
             'Age',
             'Education',
             'Urban_Yes',
             'US_Yes']
             1 # Descriptive statistics for each column
In [147]:
             2 df.describe()
Out[147]:
                       Sales CompPrice
                                            Income Advertising Population
                                                                               Price
                                                                                      ShelveLoc
                                                                                                            Education Urban_Yes
                                                                                                                                     US Yes
                  400.000000 400.000000
                                         400.000000
                                                    400.000000
                                                               400.000000
                                                                          400.000000
                                                                                     400.000000 400.000000
                                                                                                            400.000000
                                                                                                                      400.000000
                                                                                                                                 400.000000
                     7.496325 124.975000
                                          68.657500
                                                      6.635000 264.840000
                                                                          115.795000
                                                                                       2.027500
                                                                                                  53.322500
                                                                                                             13.900000
                                                                                                                        0.705000
                                                                                                                                    0.645000
              std
                     2.824115
                              15.334512
                                          27.986037
                                                      6.650364
                                                               147.376436
                                                                           23.676664
                                                                                       0.672961
                                                                                                  16.200297
                                                                                                             2.620528
                                                                                                                        0.456614
                                                                                                                                    0.479113
              min
                     0.000000
                              77.000000
                                          21.000000
                                                      0.000000
                                                                10.000000
                                                                           24.000000
                                                                                       1.000000
                                                                                                 25.000000
                                                                                                             10.000000
                                                                                                                        0.000000
                                                                                                                                    0.000000
              25%
                     5.390000 115.000000
                                          42.750000
                                                      0.000000 139.000000
                                                                          100.000000
                                                                                       2.000000
                                                                                                 39.750000
                                                                                                             12.000000
                                                                                                                        0.000000
                                                                                                                                    0.000000
                                                                                                                                    1.000000
              50%
                     7.490000 125.000000
                                          69.000000
                                                      5.000000 272.000000
                                                                          117.000000
                                                                                       2.000000
                                                                                                  54.500000
                                                                                                             14.000000
                                                                                                                         1.000000
              75%
                     9.320000 135.000000
                                          91.000000
                                                     12.000000 398.500000
                                                                          131.000000
                                                                                       2.000000
                                                                                                  66.000000
                                                                                                             16.000000
                                                                                                                         1.000000
                                                                                                                                    1.000000
                    16.270000 175.000000
                                                                                                                                    1.000000
                                         120.000000
                                                     29.000000 509.000000 191.000000
                                                                                       3.000000
                                                                                                  80.000000
                                                                                                             18.000000
                                                                                                                         1.000000
In [148]:
             1 df.head()
Out[148]:
               Sales CompPrice Income
                                        Advertising Population Price
                                                                    ShelveLoc Age Education Urban_Yes US_Yes
            0
                9.50
                            138
                                    73
                                                          276
                                                                120
                                                                            3
                                                                                42
                                                                                           17
                                                                                                               1
                                                11
               11.22
                            111
                                    48
                                                16
                                                          260
                                                                 83
                                                                                           10
                                                                                                               1
               10.06
                            113
                                    35
                                                10
                                                          269
                                                                 80
                                                                                           12
                                                                                                               1
                7.40
                            117
                                    100
                                                 4
                                                          466
                                                                 97
                                                                                 55
                                                                                           14
                                                                            2
                4.15
                            141
                                    64
                                                          340
                                                                128
                                                                            3
                                                                                38
                                                                                           13
                                                                                                               0
In [149]:
             1 # Labels are the values we want to predict
                labels = np.array(df['Income'])
                # Remove the labels from the features
             4 # axis 1 refers to the columns
             5 features=df.drop('Income', axis = 1)
                # Saving feature names for later use
                features_list = list(df.columns)
                # Convert to numpy array
             9 features = np.array(df)
In [150]:
                # Using Skicit-learn to split data into training and testing sets
                from sklearn.model_selection import train_test_split
                # Split the data into training and testing sets
             4 train_features, test_features, train_labels, test_labels = train_test_split(features, labels, test_size = 0.25, random_state
In [151]:
                print('Training Features Shape:', train_features.shape)
             print('Training Labels Shape:', train_labels.shape)
print('Testing Features Shape:', test_features.shape)
               print('Testing Labels Shape:', test_labels.shape)
           Training Features Shape: (300, 11)
           Training Labels Shape: (300,)
           Testing Features Shape: (100, 11)
           Testing Labels Shape: (100,)
```

## **Establish Baseline**

```
In [152]:
           1 # The baseline predictions are the historical averages
            2 baseline_preds = test_features[:, features_list.index('Sales')]
            3 # Baseline errors, and display average baseline error
            4 baseline_errors = abs(baseline_preds - test_labels)
            5 print('Average baseline error: ', round(np.mean(baseline_errors), 2))
          Average baseline error: 65.26
In [153]: | 1 | # Import the model we are using
            2 from sklearn.ensemble import RandomForestRegressor
            3 # Instantiate model with 1000 decision trees
            4 rf = RandomForestRegressor(n_estimators = 1000, random_state = 42)
            5 # Train the model on training data
            6 rf.fit(train_features, train_labels);
In [154]:
           1 # Use the forest's predict method on the test data
            predictions = rf.predict(test_features)
            3 # Calculate the absolute errors
            4 errors = abs(predictions - test_labels)
            5 # print out the mean absolute error (mae)
            6 print('Meam Absolute Error:', round(np.mean(errors), 2), 'degrees.')
          Meam Absolute Error: 0.27 degrees.
          Determine Performance Metrics
In [155]:
           1 # Calculate mean absolute percentage error (MAPE)
            2 mape = 100 * (errors / test_labels)
            3 # Calculate and display accuracy
            4 accuracy = 100 - np.mean(mape)
            5 print('Accuracy:', round(accuracy, 2), '%.')
          Accuracy: 99.58 %.
In [156]: 1 !pip install pydot
          Requirement already satisfied: pydot in c:\users\admin\anaconda3\lib\site-packages (1.4.2)
          Requirement already satisfied: pyparsing>=2.1.4 in c:\users\admin\anaconda3\lib\site-packages (from pydot) (3.0.4)
In [157]: 1 !pip install graphviz
          Requirement already satisfied: graphviz in c:\users\admin\anaconda3\lib\site-packages (0.20.1)
In [171]: 1 graph.write_png
Out[171]: <function pydot.Dot.__init__.<locals>.new_method(path, f='png', prog='dot', encoding=None)>
```

```
In [169]:
           1 # Import tools needed for visualization
            2 from sklearn.tree import export_graphviz
            3 import pydot
            4 # Pull out one tree from the forest
            5 tree = rf.estimators_[5]
            6 # Import tools needed for visualization
            7 from sklearn.tree import export graphviz
            8 import pydot
            9 # Pull out one tree from the forest
           10 tree = rf.estimators_[5]
           11 # Export the image to a dot file
           12 export_graphviz(tree, out_file = 'tree.dot', feature_names = features_list, rounded = True, precision = 1)
           13 # Use dot file to create a graph
           14 (graph, ) = pydot.graph_from_dot_file('tree.dot')
           15 # Write graph to a png file
           16 graph.write_png('tree.png')
          FileNotFoundError
                                                    Traceback (most recent call last)
          ~\anaconda3\lib\site-packages\pydot.py in create(self, prog, format, encoding)
             1922
          -> 1923
                              stdout_data, stderr_data, process = call_graphviz(
             1924
                                  program=prog,
          ~\anaconda3\lib\site-packages\pydot.py in call_graphviz(program, arguments, working_dir, **kwargs)
              131
          --> 132
                      process = subprocess.Popen(
              133
                          program_with_args,
          ~\anaconda3\lib\subprocess.py in __init__(self, args, bufsize, executable, stdin, stdout, stderr, preexec_fn, close_fds, shell,
          cwd, env, universal_newlines, startupinfo, creationflags, restore_signals, start_new_session, pass_fds, user, group, extra_grou
          ps, encoding, errors, text, umask)
              950
          --> 951
                              self._execute_child(args, executable, preexec_fn, close_fds,
                                                  pass_fds, cwd, env,
              952
          ~\anaconda3\lib\subprocess.py in _execute_child(self, args, executable, preexec_fn, close_fds, pass_fds, cwd, env, startupinfo,
          creationflags, shell, p2cread, p2cwrite, c2pread, c2pwrite, errread, errwrite, unused_restore_signals, unused_gid, unused_gids,
          unused_uid, unused_umask, unused_start_new_session)
             1419
                              try:
          -> 1420
                                  hp, ht, pid, tid = _winapi.CreateProcess(executable, args,
                                                           # no special security
             1421
          FileNotFoundError: [WinError 2] The system cannot find the file specified
          During handling of the above exception, another exception occurred:
          FileNotFoundError
                                                    Traceback (most recent call last)
          ~\AppData\Local\Temp/ipykernel_6664/2350536375.py in <module>
               14 (graph, ) = pydot.graph_from_dot_file('tree.dot')
               15 # Write graph to a png file
          ---> 16 graph.write_png('tree.png')
          ~\anaconda3\lib\site-packages\pydot.py in new_method(path, f, prog, encoding)
             1741
                                      encoding=None):
                                  """Refer to docstring of method `write.`"""
             1742
          -> 1743
                                  self.write(
                                      path, format=f, prog=prog,
             1744
             1745
                                      encoding=encoding)
          ~\anaconda3\lib\site-packages\pydot.py in write(self, path, prog, format, encoding)
             1826
                                  f.write(s)
             1827
                          else:
          -> 1828
                              s = self.create(prog, format, encoding=encoding)
             1829
                              with io.open(path, mode='wb') as f:
             1830
                                  f.write(s)
          ~\anaconda3\lib\site-packages\pydot.py in create(self, prog, format, encoding)
                                  args[1] = '"{prog}" not found in path.'.format(
             1931
             1932
                                      prog=prog)
          -> 1933
                                  raise OSError(*args)
             1934
                              else:
             1935
                                  raise
```

FileNotFoundError: [WinError 2] "dot" not found in path.

```
In [166]:
           1 # Import tools needed for visualization
           2 from sklearn.tree import export_graphviz
           3 import pydot
           4 import graphviz
           5 # Pull out one tree from the forest
           6 tree = rf.estimators_[5]
           7 # Import tools needed for visualization
           8 from sklearn.tree import export_graphviz
           9 import pydot
          10 # Pull out one tree from the forest
          11 tree = rf.estimators_[5]
          12 # Export the image to a dot file
          13 export_graphviz(tree, out_file = 'tree.dot', feature_names = features_list, rounded = True, precision = 1)
          14 # Use dot file to create a graph
          15 (graph, ) = pydot.graph_from_dot_file('tree.dot')
          16 # Write graph to a png file
          17 graph.write_png('tree.png')
```

```
FileNotFoundError
                                           Traceback (most recent call last)
~\anaconda3\lib\site-packages\pydot.py in create(self, prog, format, encoding)
-> 1923
                    stdout data, stderr data, process = call graphviz(
   1924
                        program=prog,
~\anaconda3\lib\site-packages\pydot.py in call_graphviz(program, arguments, working_dir, **kwargs)
   131
            process = subprocess.Popen(
--> 132
    133
                program_with_args,
~\anaconda3\lib\subprocess.py in __init__(self, args, bufsize, executable, stdin, stdout, stderr, preexec_fn, close_fds, shell,
cwd, env, universal_newlines, startupinfo, creationflags, restore_signals, start_new_session, pass_fds, user, group, extra_grou
ps, encoding, errors, text, umask)
    950
--> 951
                    self._execute_child(args, executable, preexec_fn, close_fds,
    952
                                         pass_fds, cwd, env,
~\anaconda3\lib\subprocess.py in _execute_child(self, args, executable, preexec_fn, close_fds, pass_fds, cwd, env, startupinfo,
creationflags, shell, p2cread, p2cwrite, c2pread, c2pwrite, errread, errwrite, unused_restore_signals, unused_gid, unused_gids,
unused_uid, unused_umask, unused_start_new_session)
   1419
                    try:
-> 1420
                        hp, ht, pid, tid = _winapi.CreateProcess(executable, args,
   1421
                                                   # no special security
FileNotFoundError: [WinError 2] The system cannot find the file specified
During handling of the above exception, another exception occurred:
FileNotFoundError
                                           Traceback (most recent call last)
~\AppData\Local\Temp/ipykernel_6664/2479579253.py in <module>
     15 (graph, ) = pydot.graph_from_dot_file('tree.dot')
     16 # Write graph to a png file
---> 17 graph.write_png('tree.png')
~\anaconda3\lib\site-packages\pydot.py in new_method(path, f, prog, encoding)
   1741
                             encoding=None):
                         """Refer to docstring of method `write.`"""
   1742
-> 1743
                        self.write(
   1744
                            path, format=f, prog=prog,
   1745
                             encoding=encoding)
~\anaconda3\lib\site-packages\pydot.py in write(self, path, prog, format, encoding)
   1826
                        f.write(s)
   1827
                else:
-> 1828
                    s = self.create(prog, format, encoding=encoding)
   1829
                     with io.open(path, mode='wb') as f:
~\anaconda3\lib\site-packages\pydot.py in create(self, prog, format, encoding)
1931 args[1] = '"{prog}" not found in path.'.format(
                            prog=prog)
   1932
-> 1933
                        raise OSError(*args)
   1934
                     else:
   1935
                        raise
FileNotFoundError: [WinError 2] "dot" not found in path.
```

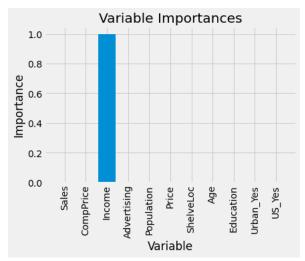
```
In [159]:
           1 # Limit depth of tree to 3 levels
            2 rf_small = RandomForestRegressor(n_estimators=10, max_depth = 3)
            3 rf_small.fit(train_features, train_labels)
            4 # Extract the small tree
            5 tree_small = rf_small.estimators_[5]
            6 # Save the tree as a png image
            7 export_graphviz(tree_small, out_file = 'small_tree.dot', feature_names = features_list, rounded = True, precision = 1)
            8 (graph, ) = pydot.graph_from_dot_file('small_tree.dot')
              graph.write_png('small_tree.png');
                                                    Traceback (most recent call last)
          ~\anaconda3\lib\site-packages\pydot.py in create(self, prog, format, encoding)
            1922
                          trv:
          -> 1923
                              stdout_data, stderr_data, process = call_graphviz(
             1924
                                  program=prog,
          ~\anaconda3\lib\site-packages\pydot.py in call_graphviz(program, arguments, working_dir, **kwargs)
              131
          --> 132
                       process = subprocess.Popen(
              133
                          program_with_args,
          ~\anaconda3\lib\subprocess.py in __init__(self, args, bufsize, executable, stdin, stdout, stderr, preexec_fn, close_fds, shell,
          cwd, env, universal_newlines, startupinfo, creationflags, restore_signals, start_new_session, pass_fds, user, group, extra_grou
          ps, encoding, errors, text, umask)
              950
          --> 951
                              self._execute_child(args, executable, preexec_fn, close_fds,
              952
                                                  pass_fds, cwd, env,
          ~\anaconda3\lib\subprocess.py in _execute_child(self, args, executable, preexec_fn, close_fds, pass_fds, cwd, env, startupinfo,
          creationflags, shell, p2cread, p2cwrite, c2pread, c2pwrite, errread, errwrite, unused_restore_signals, unused_gid, unused_gids,
          unused_uid, unused_umask, unused_start_new_session)
             1419
          -> 1420
                                  hp, ht, pid, tid = _winapi.CreateProcess(executable, args,
                                                            # no special security
             1421
          FileNotFoundError: [WinError 2] The system cannot find the file specified
          During handling of the above exception, another exception occurred:
          FileNotFoundError
                                                    Traceback (most recent call last)
          ~\AppData\Local\Temp/ipykernel_6664/1441242784.py in <module>
                7 export_graphviz(tree_small, out_file = 'small_tree.dot', feature_names = features_list, rounded = True, precision = 1)
                8 (graph, ) = pydot.graph_from_dot_file('small_tree.dot')
          ---> 9 graph.write_png('small_tree.png');
          ~\anaconda3\lib\site-packages\pydot.py in new_method(path, f, prog, encoding)
             1741
                                      encoding=None):
                                   """Refer to docstring of method `write.`"""
             1742
          -> 1743
                                  self.write(
             1744
                                      path, format=f, prog=prog,
             1745
                                      encoding=encoding)
          ~\anaconda3\lib\site-packages\pydot.py in write(self, path, prog, format, encoding)
             1826
                                  f.write(s)
             1827
                          else:
          -> 1828
                              s = self.create(prog, format, encoding=encoding)
                              with io.open(path, mode='wb') as f:
             1830
                                  f.write(s)
          ~\anaconda3\lib\site-packages\pydot.py in create(self, prog, format, encoding)
             1931
                                  args[1] = '"{prog}" not found in path.'.format(
             1932
                                      prog=prog)
          -> 1933
                                  raise OSError(*args)
             1934
                              else:
             1935
                                  raise
          FileNotFoundError: [WinError 2] "dot" not found in path.
```

```
In [161]:
           1 # Get numerical feature importances
            2 importances = list(rf.feature_importances_)
            3 # List of tuples with variable and importance
            4 | feature_importances = [(feature, round(importance, 2)) for feature, importance in zip(features_list, importances)]
              # Sort the feature importances by most important first
            6 feature_importances = sorted(feature_importances, key = lambda x: x[1], reverse = True)
            7 # Print out the feature and importances
            8 [print('Variable: {:20} Importance: {}'.format(*pair)) for pair in feature_importances];
          Variable: Income
                                         Importance: 1.0
          Variable: Sales
                                         Importance: 0.0
          Variable: CompPrice
                                         Importance: 0.0
          Variable: Advertising
                                         Importance: 0.0
          Variable: Population
                                         Importance: 0.0
          Variable: Price
                                         Importance: 0.0
          Variable: ShelveLoc
                                         Importance: 0.0
          Variable: Age
                                         Importance: 0.0
          Variable: Education
                                         Importance: 0.0
          Variable: Urban_Yes
                                         Importance: 0.0
          Variable: US_Yes
                                         Importance: 0.0
In [163]: | 1 | # New random forest with only the two most important variables
              rf_most_important = RandomForestRegressor(n_estimators= 1000, random_state=42)
            3 # Extract the two most important features
            4 important_indices = [features_list.index('Sales'), features_list.index('Income')]
            5 train_important = train_features[:, important_indices]
            6 test_important = test_features[:, important_indices]
              # Train the random forest
            8 rf_most_important.fit(train_important, train_labels)
            9 # Make predictions and determine the error
           10 predictions = rf_most_important.predict(test_important)
           11 errors = abs(predictions - test_labels)
           12 # Display the performance metrics
           print('Mean Absolute Error:', round(np.mean(errors), 2), 'degrees.')
           mape = np.mean(100 * (errors / test_labels))
           15 accuracy = 100 - mape
           16 print('Accuracy:', round(accuracy, 2), '%.')
```

Mean Absolute Error: 0.16 degrees.

Accuracy: 99.75 %.

Out[165]: Text(0.5, 1.0, 'Variable Importances')



In [ ]:	1	
In [ ]:	1	