```
import pandas as pd
In [52]:
         import numpy as np
         import pandas profiling
         import matplotlib.pyplot as plt
         from sklearn.model_selection import train_test_split
         from sklearn.metrics import accuracy score, classification report
         from sklearn.ensemble import RandomForestClassifier, GradientBoostingClassifie
         from sklearn.tree import DecisionTreeClassifier
         from sklearn.model_selection import cross_val_score
         from sklearn.externals.six import StringIO
         from IPython.display import Image
         from sklearn.tree import export_graphviz
         import pydotplus
In [53]: | #data is the variable name for the Pandas default data structure called
         # a "Dataframe" in this case its a 2D Array
         data = pd.read csv('Demographic Data.csv')
In [54]:
         #data.head()
         clean data = data.drop duplicates()
In [55]: clean data.to csv(r'D Data Clean.csv', index=False)
In [56]: data = pd.read_csv('D_Data_Clean.csv')
In [57]: | data.duplicated().sum()
Out[57]: 0
In [58]: | from sklearn.datasets import load_digits
         digits = load_digits()
         print(digits.data)
         digits.target
         [[ 0. 0. 5. ... 0. 0. 0.]
          [0. 0. 0. ... 10. 0. 0.]
          [ 0. 0. 0. ... 16.
                                9. 0.1
          [0. 0. 1. ... 6. 0. 0.]
          [ 0. 0. 2. ... 12. 0. 0.]
          [ 0. 0. 10. ... 12. 1. 0.]]
Out[58]: array([0, 1, 2, ..., 8, 9, 8])
```

```
In [59]:
         data.info()
         data.describe()
         <class 'pandas.core.frame.DataFrame'>
         RangeIndex: 79979 entries, 0 to 79978
         Data columns (total 5 columns):
         in-store
                     79979 non-null int64
         age
                     79979 non-null int64
         items
                     79979 non-null int64
                     79979 non-null float64
         amount
         region
                     79979 non-null int64
         dtypes: float64(1), int64(4)
         memory usage: 3.1 MB
```

Out[59]:

	in-store	age	items	amount	region
count	79979.000000	79979.000000	79979.000000	79979.000000	79979.000000
mean	0.500006	45.758512	4.505133	835.825727	2.674915
std	0.500003	15.715158	2.061250	721.263650	1.126642
min	0.000000	18.000000	1.000000	5.004700	1.000000
25%	0.000000	33.000000	3.000000	285.120000	2.000000
50%	1.000000	45.000000	4.000000	582.140000	3.000000
75%	1.000000	56.000000	6.000000	1233.400000	4.000000
max	1.000000	85.000000	8.000000	3000.000000	4.000000

```
In [60]: #Define bins for age ranges 18-29 (young), 29-59 (middle aged), 59-85 (older)
bins = [18, 29, 59, 85]
```

```
In [61]: group_names = ['Young', 'Middle', 'Older']
```

```
In [62]: data['age_bins'] = pd.cut(data['age'], bins, labels=group_names)
```

In [63]: data

Out[63]:

	in-store	age	items	amount	region	age_bins
0	0	37	4	281.03	2	Middle
1	0	35	2	219.51	2	Middle
2	1	45	3	1525.70	4	Middle
3	1	46	3	715.25	3	Middle
4	1	33	4	1937.50	1	Middle
79974	1	71	3	558.82	1	Older
79975	0	59	7	1932.00	3	Middle
79976	0	54	1	414.16	2	Middle
79977	1	49	4	335.32	1	Middle
79978	1	30	1	527.12	3	Middle

79979 rows × 6 columns

```
In [64]: #drop rows of customers under 60 and put into a new dataframe
    df2 = data.drop(index=data[data['age_bins'] == 'Middle'].index)
    data_older = df2.drop(index=df2[df2['age_bins'] == 'Young'].index)
    older_online = data_older.drop(index=data_older[data_older['in-store'] == 1].i
    ndex)
```

```
In [65]: old_sum = older_online['amount'].sum()
    print('Online Customers 60+ spent: $','%.2f'%old_sum)
```

Online Customers 60+ spent: \$ 5004192.13

```
In [66]: #show new dataframe data_older
```

Out[66]:

	in-store	age	items	amount	region	age_bins
7	1	66	5	819.9800	3	Older
32	1	74	6	652.5700	3	Older
36	1	66	6	429.1300	3	Older
51	0	74	3	473.6200	2	Older
59	0	66	1	65.6190	2	Older
79956	0	61	7	251.4200	2	Older
79968	0	69	1	404.4200	2	Older
79969	0	71	7	6.9172	2	Older
79971	0	77	4	170.0200	2	Older
79974	1	71	3	558.8200	1	Older

16473 rows × 6 columns

```
In [67]: #drop rows of customers under 30 and put into a new dataframe
    df3 = data.drop(index=data[data['age_bins'] == 'Middle'].index)
    data_young = df3.drop(index=df3[df3['age_bins'] == 'Older'].index)
    older_online = data_older.drop(index=data_older[data_older['in-store'] == 1].i
    ndex)
    young_online = data_young.drop(index=data_young[data_young['in-store'] == 1].i
    ndex)
    temp2 = data_older['amount'].sum()
```

```
In [68]: young_online['amount']
```

1314.200

```
29
          623.050
39
          261.100
46
          592.340
78
           39.771
79948
         2906.400
79958
          428.840
79963
         1672.700
79964
          491.770
79973
          698.540
```

Out[68]: 5

Name: amount, Length: 5610, dtype: float64

```
In [71]: temp = 0
    for index in range(len(older_online)):
        temp += older_online['items'].iloc[index]

temp2 = 0
    for index in range(len(young_online)):
        temp2 += young_online['items'].iloc[index]

print('older group online items bought: ',temp)
print('younger group online items bought: ',temp2)
```

older group online items bought: 48895 younger group online items bought: 25394

```
In [72]: young_sum = young_online['amount'].sum()
    older_sum = older_online['amount'].sum()

temp2 = data_older['amount'].sum()
    temp = data_young['amount'].sum()

olderInstore = temp2 - older_sum
    youngInstore = temp - young_sum
```

In [73]: older_online

Out[73]:

	in-store	age	items	amount	region	age_bins
51	0	74	3	473.6200	2	Older
59	0	66	1	65.6190	2	Older
64	0	69	4	213.6200	2	Older
65	0	81	4	324.9800	2	Older
67	0	65	2	204.6700	2	Older
79952	0	76	1	62.3100	2	Older
79956	0	61	7	251.4200	2	Older
79968	0	69	1	404.4200	2	Older
79969	0	71	7	6.9172	2	Older
79971	0	77	4	170.0200	2	Older

10786 rows × 6 columns

```
In [22]: #Sum total of online customers based on the age ranges 60-85 and 18-30
         older_online = (data_older['in-store'] == 0)
         young online = (data young['in-store'] == 0)
         print('Older Online Customers: ', older online.sum())
         print('Young Online Customers: ', young online.sum())
         print('')
         print('Amount In-store Older: $', end="")
         print('%.2f'%olderInstore)
         print('Amount Online Older: $', end="")
         print('%.2f'%old_sum)
         print('')
         print('Amount In-store Young: $', end="")
         print('%.2f' %youngInstore)
         print('Amount Online Young: $', end="")
         print('%.2f'%young sum)
         print('Customers under 30 spent $', '%.2f'%(young_sum - old_sum), ' more than
          older customers online')
         print('Customers under 30 spent $', '%.2f'%(youngInstore - olderInstore), ' mo
         re than older customers in-store')
         Older Online Customers: 10786
         Young Online Customers:
         Amount In-store Older: $2952619.67
         Amount Online Older: $5004192.13
         Amount In-store Young: $7140473.06
         Amount Online Young: $7993186.83
         Customers under 30 spent $ 2988994.70 more than older customers online
         Customers under 30 spent $ 4187853.39 more than older customers in-store
         X = data.iloc[:, 1:4]
In [23]:
         print('Summary of feature sample')
         X.head()
         Summary of feature sample
Out[23]:
             age items amount
          0
             37
                       281.03
          1
             35
                       219.51
          2
             45
                    3 1525.70
          3
             46
                    3
                       715.25
                    4 1937.50
             33
In [40]: features = ["in-store", "region", "items", "amount"]
In [25]: y = data['in-store']
         #region1 = (older_online['region'] == 1)
         #region1.sum()
         # iloc[sr:ed, sc:ed]
```

```
In [26]: | X_train, X_test, y_train, y_test = train_test_split(X, y, test_size = .30, ran
          dom state = 123)
In [27]: X train.head()
Out[27]:
                age items amount
          65254
                 46
                        2 1912.70
          11430
                           231.78
                 60
                        7
          46451
                        7 1692.70
                 41
          20822
                 41
                        8
                            390.01
          52818
                 26
                        5 2399.80
In [28]:
         y_train
Out[28]: 65254
                   1
         11430
                   0
         46451
                   0
         20822
                   0
         52818
                   0
         63206
                   1
         61404
                   1
         17730
                   1
         28030
                   1
         15725
         Name: in-store, Length: 55985, dtype: int64
In [29]: | from sklearn.model_selection import cross_val_score
In [30]: | algos Class = []
          #algos_Class.append(('Random Forest Classifier', RandomForestClassifier()))
          algos Class.append(('Decision Tree Classifier', DecisionTreeClassifier()))
In [68]: | #instanciate
          #algo = RandomForestClassifier()
In [69]: #information for the group to build the model
          #model = algo.fit(X train, y train)
In [36]: #compare preds to ground truth (y_test - known values)
In [31]:
         #classification
          results = []
          names = []
          for name, model in algos_Class:
               result = cross_val_score(model, X,y, cv=3, scoring='accuracy')
               names.append(name)
               results.append(result)
```

```
In [32]: for i in range(len(names)):
              print(names[i],results[i].mean())
          Decision Tree Classifier 0.6524462893922213
 In [33]: #Modeling (Classification)
          algo = DecisionTreeClassifier(max_depth=4)
          model = algo.fit(X_train,y_train)
In [34]:
          print(cross_val_score(model, X, y, cv=3))
          [0.70712678 0.71856714 0.70996662]
In [115]:
          #Predictions
          preds = model.predict(X_test)
In [116]: | accuracy_score(y_test, preds)
Out[116]: 0.7230974410269234
In [117]: print(classification_report(y_test, preds))
                         precision
                                      recall f1-score
                                                         support
                     0
                              0.71
                                        0.75
                                                  0.73
                                                           11874
                      1
                              0.74
                                        0.70
                                                  0.72
                                                           12120
                                                  0.72
                                                           23994
              accuracy
             macro avg
                              0.72
                                        0.72
                                                  0.72
                                                           23994
          weighted avg
                              0.72
                                        0.72
                                                  0.72
                                                           23994
 In [35]: region_values = ['0','1']
```

