```
In [1]: import numpy as np
         import pandas as pd
         import matplotlib.pyplot as plt
         import seaborn as sns
         import sklearn as sk
         %matplotlib inline
In [2]: df = pd.read_csv("pharmacy_with_pop.csv")
         df.head()
Out[2]:
                               Pharmacy
             Delivery Pharmacy
                                                                                                                           Age
                                                                                                                                                                                                       Total
                                                                                       ATC Units Price Contribution Age
                                         YOB Gender
                                                         CNK
                                                                      Product Name
                                                                                                                                Province
                                                                                                                                          ATC_Classification Medication_Type year
                                                                                                                                                                                                  Population
               Date
                           No
                                   Post
               2017-
          0
                       7341765
                                                   1 5520523
                                                               WACHTHONORARIUM
                                                                                                0 4.90
                                                                                                                      95
                                                                                                                         95-99
                                                                                                                                                   Unknown
                                                                                                                                                                                  5274.0
                                                                                                                                                                                          16548.0
                                                                                                                                                                                                     21822.0
                                     21 1922
                                                                                                                0.00
                                                                                                                                                                  Unknown 2017
                                                                                                                                 Antwerp
               01-01
               2017-
                                                                ZALDIAR 37,5 MG/325
                       7341765
                                     21 1925
                                                   1 1799931
                                                                                    N02AJ13
                                                                                               20 9.26
                                                                                                               3.62
                                                                                                                      92
                                                                                                                         90-94
                                                                                                                                 Antwerp
                                                                                                                                              Nervous system
                                                                                                                                                             Analgesic drugs 2017
                                                                                                                                                                                 28060.0
                                                                                                                                                                                          66339.0
                                                                                                                                                                                                     94399.0
                                                               MG FILMOMH TABL 20
               01-01
               2017-
                                                               VASEXTEN CAPS BLIST
                                                                                                                                  Flemish
                                                                                                                                              Cardiovascular
                                                                                                                                                             Calcium channel
          2
                                                                                   C08CA12
                       8272695
                                     16 1932
                                                   2 1719400
                                                                                               28 19.22
                                                                                                                         85-89
                                                                                                                                                                           2017
                                                                                                                                                                                 73868.0 134533.0
                                                                                                                                                                                                    208401.0
                                                                                                                4.98
                                                                                                                      85
              01-01
                                                                        28 X 10 MG
                                                                                                                                  Brabant
                                                                                                                                                    system
                                                                                                                                                                   blockers
               2017-
                                                                                                                                  Flemish
          3
                       8272695
                                                   2 5520523
                                                               WACHTHONORARIUM
                                                                                                                         80-84
                                                                                                                                                                  Unknown 2017 130079.0 192247.0
                                                                                                                                                                                                    322326.0
                                     16 1933
                                                                                                0 4.90
                                                                                                                0.00
                                                                                                                      84
                                                                                                                                                   Unknown
               01-01
                                                                                                                                  Brabant
                                                                                                                                           Systemic hormonal
               2017-
                                                                 AACIDEXAM 5MG/ML
                                                                                                                                                              Corticosteroids
                                                   1 1750132
                                                                                   H02AB02
                       9111423
                                     10 1931
                                                                                                1
                                                                                                  6.15
                                                                                                                0.39
                                                                                                                      86
                                                                                                                         85-89
                                                                                                                                 Brussels
                                                                                                                                                preparations,
                                                                                                                                                                           2017
                                                                                                                                                                                 73868.0 134533.0
                                                                                                                                                                                                    208401.0
                                                              OPL INJ FL INJ 1 X 1ML
               01-01
                                                                                                                                                                   systemic
                                                                                                                                              excluding repr...
In [3]: df['Age Group'] = df['Age Group'].apply(lambda x: x.replace('100 or older', '100+'))
In [4]: MedicationType = 'Psychoanaleptics'
         Frequency= 'M'
In [5]: is_Med = df['Medication_Type'] == MedicationType
         df2 = df[is\_Med]
         df2.head()
Out[5]:
```

	Delivery Date	Pharmacy No	Pharmacy Post	ΥОВ	Gender	CNK	Product Name	ATC	Units	Price	Contribution	Age	Age Group	Province	ATC_Classification	Medication_Type	year	М	F	Total Population
4	0 2017- 01-01	7641438	40	1969	2	3183092	CYMBALTA 60 MG MAAGSAPRESIST. CAPS 98 X 60 MG	N06AX21	98	53.30	13.67	48	45-49	Liege	Nervous system	Psychoanaleptics	2017	400752.0	384630.0	785382.0
83	8 2017- 01-01	7084071	86	1899	0	126987	REDOMEX DIFFUCAPS CAPS 40 X 25 MG	N06AA09	40	7.22	0.91	118	100+	West Flanders	Nervous system	Psychoanaleptics	2017	NaN	NaN	NaN
83	9 2017- 01-01	7056201	89	1899	0	127019	REDOMEX DIFFUCAPS CAPS 40 X 50 MG	N06AA09	40	8.83	0.89	118	100+	West Flanders	Nervous system	Psychoanaleptics	2017	NaN	NaN	NaN
97	8 2017- 01-01	7067208	20	1899	0	1390343	SERLAIN 50 MG COMP PELL 30 X 50 MG	N06AB06	30	14.96	2.14	118	100+	Antwerp	Nervous system	Psychoanaleptics	2017	NaN	NaN	NaN
126	9 2017- 01-01	7122399	30	1899	0	1625672	FLUOXETINE EG CAPS 56 X 20 MG	N06AB03	56	25.74	4.26	118	100+	Flemish Brabant	Nervous system	Psychoanaleptics	2017	NaN	NaN	NaN

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D:\Anaconda\lib\site-packages\ipykernel_launcher.py:1: SettingWithCopyWarning:

A value is trying to be set on a copy of a slice from a DataFrame.

Try using .loc[row_indexer,col_indexer] = value instead

See the caveats in the documentation: http://pandas.pydata.org/pandas-docs/stable/user_guide/indexing.html#returning-a-view-versus-a-copy """Entry point for launching an IPython kernel.

Out[7]:

	Delivery Date	Pharmacy No	Pharmacy Post	УОВ	Gender	CNK	Product Name	ATC	Units	Price	Contribution	Age	Age Group	Province	ATC_Classification	Medication_Type	year	М	F	Total Population
40	2017- 01-01	7641438	40	1969	2	3183092	CYMBALTA 60 MG MAAGSAPRESIST. CAPS 98 X 60 MG	N06AX21	98	53.30	13.67	48	45-49	Liege	Nervous system	Psychoanaleptics	2017	400752.0	384630.0	785382.0
838	2017- 01-01	7084071	86	1899	0	126987	REDOMEX DIFFUCAPS CAPS 40 X 25 MG	N06AA09	40	7.22	0.91	118	100+	West Flanders	Nervous system	Psychoanaleptics	2017	NaN	NaN	NaN
839	2017- 01-01	7056201	89	1899	0	127019	REDOMEX DIFFUCAPS CAPS 40 X 50 MG	N06AA09	40	8.83	0.89	118	100+	West Flanders	Nervous system	Psychoanaleptics	2017	NaN	NaN	NaN
978	2017- 01-01	7067208	20	1899	0	1390343	SERLAIN 50 MG COMP PELL 30 X 50 MG	N06AB06	30	14.96	2.14	118	100+	Antwerp	Nervous system	Psychoanaleptics	2017	NaN	NaN	NaN
1269	2017- 01-01	7122399	30	1899	0	1625672	FLUOXETINE EG CAPS 56 X 20 MG	N06AB03	56	25.74	4.26	118	100+	Flemish Brabant	Nervous system	Psychoanaleptics	2017	NaN	NaN	NaN

In [8]: res = df2.set_index('Delivery Date').groupby([pd.Grouper(freq=Frequency), 'Total Population', 'Age Group'])['Units'].sum().reset_index()
print(res)

	Delivery Date	Total Population	Age Group	Units
0	2017-01-31	21822.0	95-99	5188
1	2017-01-31	94399.0	90-94	31220
2	2017-01-31	208401.0	85-89	91037
3	2017-01-31	322326.0	80-84	139935
4	2017-01-31	379764.0	75-79	152454
• •		•••		
701	2019-12-31	751517.0	40-44	131912
702	2019-12-31	754578.0	35-39	107362
703	2019-12-31	772635.0	45-49	178354
704	2019-12-31	799615.0	55-59	256272
705	2019-12-31	802276.0	50-54	237066

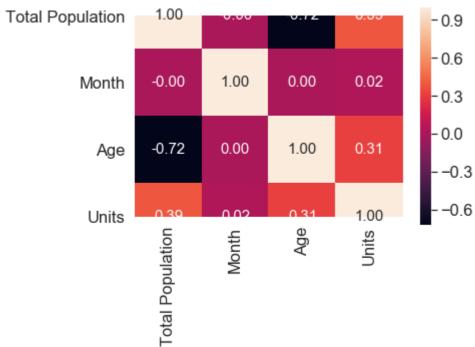
[706 rows x 4 columns]

```
In [9]: res['Month'] = 0
          res['Month'] = pd.DatetimeIndex(res['Delivery Date']).month
          res.head()
 Out[9]:
             Delivery Date Total Population Age Group
                                                   Units Month
               2017-01-31
          0
                                            95-99
                                                    5188
                                21822.0
                                                             1
               2017-01-31
                                94399.0
                                            90-94
                                                   31220
               2017-01-31
                               208401.0
                                            85-89
                                                   91037
                                                             1
               2017-01-31
                               322326.0
                                            80-84
                                                  139935
               2017-01-31
                               379764.0
                                            75-79 152454
                                                             1
In [10]: | res['Age'] = 0
          res['Age'] = res['Age Group'].map( {'0-4': 1, '5-9': 2, '10-14': 3, '15-19': 4, '20-24': 5, '25-29': 6, '30-34': 7, '35-39': 8, '40-44': 9, '45-49': 10, '50-54': 11, '55-59': 12,
          '60-64': 13, '65-69': 14, '70-74': 15, '75-79': 16, '80-84': 17, '85-89': 18, '90-94': 19, '95-99': 20, '100+': 20, } ).astype(int)
          res.head(3)
Out[10]:
             Delivery Date Total Population Age Group Units Month Age
              2017-01-31
                                                            1 20
                                21822.0
                                            95-99
                                                  5188
               2017-01-31
                                94399.0
                                            90-94 31220
                                                            1 19
               2017-01-31
                               208401.0
                                            85-89 91037
                                                            1 18
In [11]: res['Total Population'] = res['Total Population'].astype(float)
          res.dtypes
          res = res.reset_index()
          res.head()
Out[11]:
             index Delivery Date Total Population Age Group
                                                         Units Month Age
          0
                     2017-01-31
                0
                                      21822.0
                                                  95-99
                                                          5188
                                                                   1
                                                                      20
                     2017-01-31
                                      94399.0
                                                  90-94
                                                         31220
                                                                   1 19
                     2017-01-31
                2
                                     208401.0
                                                  85-89
                                                         91037
                                                                   1 18
          3
                     2017-01-31
                                     322326.0
                                                  80-84
                                                       139935
                                                                   1 17
                     2017-01-31
                                     379764.0
                                                  75-79 152454
                                                                   1 16
In [12]: res.dtypes
Out[12]: index
                                        int64
          Delivery Date
                               datetime64[ns]
          Total Population
                                      float64
          Age Group
                                       object
          Units
                                        int64
                                        int64
          Month
                                        int32
          Age
          dtype: object
In [13]: cols = list(res.columns)
          a, b = cols.index('Age Group'), cols.index('Age')
          cols[b], cols[a] = cols[a], cols[b]
          res = res[cols]
```

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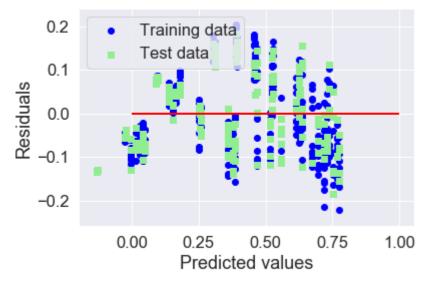
```
In [14]: from sklearn.preprocessing import MinMaxScaler
         minmax = MinMaxScaler()
         res[[i for i in list(res.columns)[2:5]]] = minmax.fit_transform(res[[i for i in list(res.columns)[2:5]]])
         print(res)
              index Delivery Date Total Population
                                                                Units Month \
                                                        Age
                      2017-01-31
         0
                 0
                                         0.000000 1.000000 0.018426
                                                                          1
                 1
                      2017-01-31
         1
                                         0.091636 0.947368 0.111238
                                                                          1
                 2
                      2017-01-31
                                         0.235574 0.894737 0.324504
                                                                          1
         3
                 3
                      2017-01-31
                                         0.379416 0.842105 0.498841
                                                                          1
                      2017-01-31
                                         0.451937 0.789474 0.543475
                                                                          1
                                              ...
                                                        . . .
                                                                         . . .
                . . .
         701
               701
                      2019-12-31
                                         0.921312 0.421053 0.470237
                                                                         12
         702
               702
                      2019-12-31
                                         0.925177 0.368421 0.382708
                                                                         12
         703
               703
                      2019-12-31
                                         0.947976 0.473684 0.635817
                                                                         12
                                                                         12
         704
               704
                      2019-12-31
                                         0.982041 0.578947 0.913620
         705
               705
                      2019-12-31
                                         0.985401 0.526316 0.845144
                                                                         12
             Age Group
                95-99
         0
         1
                90-94
         2
                85-89
                80-84
         3
         4
                75-79
                  . . .
         701
                40-44
         702
                35-39
         703
                45-49
         704
                55-59
         705
                50-54
```

[706 rows x 7 columns]



Slope: 1.193 Intercept: -1.058

```
In [17]: plt.scatter(y_train_pred, y_train_pred - y_train,
             c='blue', marker='o', label='Training data'
         plt.scatter(y test pred, y test pred - y test,
             c='lightgreen', marker='s', label='Test data'
         plt.xlabel('Predicted values')
         plt.ylabel('Residuals')
         plt.legend(loc='upper left')
         plt.hlines(y=0, xmin=-0, xmax=1, lw=2, color='red')
         plt.show()
```



```
In [18]: from sklearn.metrics import median_absolute_error
         from sklearn.metrics import mean_squared_error
         from sklearn.metrics import r2_score
         print("Mean Squared Error: ",mean_squared_error(y_test, y_test_pred))
         errors = abs(y_test_pred-y_test)
         print('Mean Absolute Error:', round(np.mean(errors), 2))
         print('R2 score: ',r2_score(y_test, y_test_pred))
         print('Median Absolute Error: ',median_absolute_error(y_test, y_test_pred))
         print('Accuracy: ', slr.score(X_test,y_test)*100)
         Mean Squared Error: 0.007453403110110695
```

Mean Absolute Error: 0.07 R2 score: 0.8953789950699395

Median Absolute Error: 0.07063054115529932

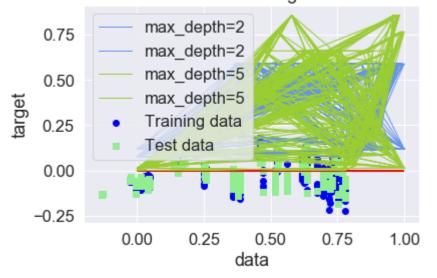
Accuracy: 89.53789950699395

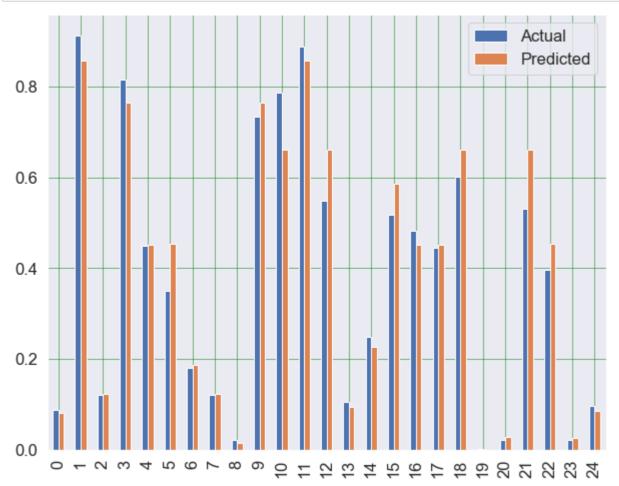
```
In [19]: | from sklearn.tree import DecisionTreeRegressor
         regr_1 = DecisionTreeRegressor(max_depth=2)
         regr 2 = DecisionTreeRegressor(max depth=5)
         regr_1.fit(X, y)
         regr_2.fit(X, y)
```

```
Out[19]: DecisionTreeRegressor(criterion='mse', max_depth=5, 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')
```

```
In [20]: # Predict
         #X_test = np.arange(0.0, 5.0, 0.01)[:, np.newaxis]
         y_1 = regr_1.predict(X_test)
         y_2 = regr_2.predict(X_test)
In [21]: plt.figure()
         plt.scatter(y_train_pred, y_train_pred - y_train,
             c='blue', marker='o', label='Training data')
         plt.scatter(y_test_pred, y_test_pred - y_test,
             c='lightgreen', marker='s', label='Test data')
         plt.plot(X_test, y_1, color="cornflowerblue",
                  label="max_depth=2", linewidth=1)
         plt.plot(X_test, y_2, color="yellowgreen", label="max_depth=5", linewidth=1)
         plt.xlabel("data")
         plt.ylabel("target")
         plt.title("Decision Tree Regression")
         plt.legend()
         plt.hlines(y=0, xmin=-0, xmax=1, lw=2, color='red')
         plt.show()
```

Decision Tree Regression





```
In [22]:
    from sklearn.metrics import median_absolute_error
    from sklearn.metrics import r2_score
    from sklearn.metrics import r2_score
    print("Mean Squared Error: ",mean_squared_error(y_test, y_1))
    errors = abs(y_1-y_test)
    print('Mean Absolute Error:', round(np.mean(errors), 2))
    print('R2 score Model 1: ',r2_score(y_test, y_1))
    print('R2 score Model 2: ',r2_score(y_test, y_2))
    print('Median Absolute Error: ',median_absolute_error(y_test, y_1))
    print('Accuracy Model 1: ', regr_1.score(X_test,y_test)*100)
    print('Accuracy Model 2: ', regr_2.score(X_test,y_test)*100)
```

Mean Squared Error: 0.016193578479481607

Mean Absolute Error: 0.1

R2 score Model 1: 0.772695984785934
R2 score Model 2: 0.9668504699790277
Median Absolute Error: 0.0758911358497309
Accuracy Model 1: 77.26959847859341
Accuracy Model 2: 96.68504699790277

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WIP

Linear SVR

Fit The plot

from sklearn.svm import SVR

svr_rbf = SVR(kernel='rbf', C=1e3, gamma=0.1) svr_lin = SVR(kernel='linear', C=1e3) svr_poly = SVR(kernel='poly', C=1e3, degree=2) y_rbf = svr_rbf.fit(X_train, y_train).predict(X_test) y_lin = svr_lin.fit(X_train, y_train).predict(X_test) y_poly = svr_poly.fit(X_train, y_train).predict(X_test)

lw = 2 plt.scatter(X, y, color='darkorange', label='data') plt.hold('on') plt.plot(X, y_rbf, color='navy', lw=lw, label='RBF model') plt.plot(X, y_lin, color='c', lw=lw, label='Linear model') plt.plot(X, y_poly, color='cornflowerblue', lw=lw, label='Polynomial model') plt.xlabel('data') plt.ylabel('target') plt.title('Support Vector Regression') plt.show()

In []: