

```
In [14]: 1 import numpy as np
         2 import pandas as pd
```

```
In [15]: 1 csv_df = pd.read_csv("Melbourne_housing_FULL.csv")
```

```
In [16]: 1 from sklearn.ensemble import RandomForestRegressor
```

```
In [17]: 1 rf = RandomForestRegressor()
```

```
In [28]: 1 csv_df.head()
```

Out[28]:

|   | Suburb     | Address            | Rooms | Type | Price     | Method | SellerG | Date      |
|---|------------|--------------------|-------|------|-----------|--------|---------|-----------|
| 0 | Abbotsford | 68 Studley St      | 2     | h    | NaN       | SS     | Jellis  | 3/09/2016 |
| 1 | Abbotsford | 85 Turner St       | 2     | h    | 1480000.0 | S      | Biggin  | 3/12/2016 |
| 2 | Abbotsford | 25 Bloomburg St    | 2     | h    | 1035000.0 | S      | Biggin  | 4/02/2016 |
| 3 | Abbotsford | 18/659 Victoria St | 3     | u    | NaN       | VB     | Rounds  | 4/02/2016 |
| 4 | Abbotsford | 5 Charles St       | 3     | h    | 1465000.0 | SP     | Biggin  | 4/03/2017 |

5 rows × 21 columns



```
In [19]: 1 feat_df = csv_df.drop('Price', axis=1)
```

```
In [20]: 1 csv_df.shape
```

Out[20]: (34857, 21)

```
In [21]: 1 feat_df.shape
```

Out[21]: (34857, 20)

```
In [24]: 1 y = csv_df['Price'].values
```

```
In [25]: 1 y
```

Out[25]: array([ nan, 1480000., 1035000., ..., 705000., 1140000., 1020000.])

```
In [26]: 1 y.shape
```

```
Out[26]: (34857,)
```

```
In [34]: 1 rows_labeled_na = csv_df.isnull().any(axis=1)
```

```
In [37]: 1 rows_with_na = csv_df[rows_labeled_na]
```

```
In [38]: 1 rows_with_data = csv_df[~rows_labeled_na]
```

```
In [40]: 1 csv_df.shape, rows_with_na.shape, rows_with_data.shape
```

```
Out[40]: ((34857, 21), (25970, 21), (8887, 21))
```

```
In [43]: 1 feat_df = rows_with_data.drop('Price', axis=1)
```

```
In [44]: 1 feat_df.shape
```

```
Out[44]: (8887, 20)
```

```
In [45]: 1 y = rows_with_data['Price'].values  
2 y.shape
```

```
Out[45]: (8887,)
```

```
In [51]: 1 suburbs = {}  
2 for s in feat_df['Suburb'].values:  
3     if s not in suburbs:  
4         suburbs[s] = len(suburbs)
```

```
In [53]: 1 len(suburbs)
```

```
Out[53]: 315
```

```
In [60]: 1 feat_df['Suburb'] = feat_df['Suburb'].replace(suburbs)
```

```
In [61]: 1 feat_df.head()
```

```
Out[61]:
```

|    | Suburb | Address               | Rooms | Type | Method | SellerG | Date      | Distance | I |
|----|--------|-----------------------|-------|------|--------|---------|-----------|----------|---|
| 2  | 0      | 25<br>Bloomburg<br>St | 2     | h    | S      | Biggin  | 4/02/2016 | 2.5      |   |
| 4  | 0      | 5 Charles<br>St       | 3     | h    | SP     | Biggin  | 4/03/2017 | 2.5      |   |
| 6  | 0      | 55a Park<br>St        | 4     | h    | VB     | Nelson  | 4/06/2016 | 2.5      |   |
| 11 | 0      | 124 Yarra<br>St       | 3     | h    | S      | Nelson  | 7/05/2016 | 2.5      |   |
| 14 | 0      | 98 Charles<br>St      | 2     | h    | S      | Nelson  | 8/10/2016 | 2.5      |   |

```
In [72]: 1 feat_df['Type'] = feat_df['Type'].astype('category').cat.codes
```

```
In [74]: 1 feat_df['Address'] = feat_df['Address'].astype('category').cat.codes  
2 feat_df['Method'] = feat_df['Method'].astype('category').cat.codes  
3 feat_df['SellerG'] = feat_df['SellerG'].astype('category').cat.codes  
4 feat_df['CouncilArea'] = feat_df['CouncilArea'].astype('category').cat.codes  
5 feat_df['Regionname'] = feat_df['Regionname'].astype('category').cat.codes
```

```
In [85]: 1 feat_df.head()
```

```
Out[85]:
```

|    | Suburb | Address | Rooms | Type | Method | SellerG | Date                 | I |
|----|--------|---------|-------|------|--------|---------|----------------------|---|
| 2  | 0      | 3922    | 2     | 0    | 1      | 22      | 14595552000000000000 |   |
| 4  | 0      | 6458    | 3     | 0    | 3      | 22      | 14911776000000000000 |   |
| 6  | 0      | 6960    | 4     | 0    | 4      | 147     | 14599008000000000000 |   |
| 11 | 0      | 1374    | 3     | 0    | 1      | 147     | 14676768000000000000 |   |
| 14 | 0      | 8740    | 2     | 0    | 1      | 147     | 14707872000000000000 |   |

```
In [80]: 1 feat_df['Date'] = pd.to_datetime(feat_df['Date'],  
2 infer_datetime_format=True)
```

```
In [84]: 1 feat_df['Date'] = feat_df['Date'].astype(np.int64)
```

```
In [86]: 1 feat_df.head()
```

```
Out[86]:
```

|    | Suburb | Address | Rooms | Type | Method | SellerG | Date                 | C |
|----|--------|---------|-------|------|--------|---------|----------------------|---|
| 2  | 0      | 3922    | 2     | 0    | 1      | 22      | 14595552000000000000 |   |
| 4  | 0      | 6458    | 3     | 0    | 3      | 22      | 14911776000000000000 |   |
| 6  | 0      | 6960    | 4     | 0    | 4      | 147     | 14599008000000000000 |   |
| 11 | 0      | 1374    | 3     | 0    | 1      | 147     | 14676768000000000000 |   |
| 14 | 0      | 8740    | 2     | 0    | 1      | 147     | 14707872000000000000 |   |

```
In [88]: 1 rf = RandomForestRegressor()  
2 rf.fit(feat_df, y)
```

```
/Users/kalininaalex/miniconda3/envs/py36/lib/python3.6/site-package  
s/sklearn/ensemble/forest.py:248: FutureWarning: The default value  
of n_estimators will change from 10 in version 0.20 to 100 in 0.22.  
"10 in version 0.20 to 100 in 0.22.", FutureWarning)
```

```
Out[88]: RandomForestRegressor(bootstrap=True, criterion='mse', max_depth=None,  
                                max_features='auto', 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, n_estimators=10, n_jobs=None,  
                                oob_score=False, random_state=None, verbose=0, warm_start=False)
```

```
In [92]: 1 from sklearn.model_selection import train_test_split
```

```
In [93]: 1 ? train_test_split
```

```
In [106]: 1 X_train, X_test, y_train, y_test = train_test_split(feat_df, y, ra
```

```
In [100]: 1 feat_df.shape, X_train.shape, X_test.shape, y_train.shape, y_test.
```

```
Out[100]: ((8887, 20), (6665, 20), (2222, 20), (6665,), (2222,))
```

```
In [124]: 1 rf = RandomForestRegressor(n_estimators=100 , random_state=17)
          2 %time rf.fit(X_train, y_train)
```

CPU times: user 5.64 s, sys: 139 ms, total: 5.77 s  
Wall time: 6.39 s

```
Out[124]: RandomForestRegressor(bootstrap=True, criterion='mse', max_depth=None,
                                max_features='auto', 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, n_estimators=100, n_jobs=None,
                                oob_score=False, random_state=17, verbose=0, warm_start=False)
```

```
In [125]: 1 rf.score(X_train, y_train)
```

Out[125]: 0.9748081366886574

```
In [126]: 1 rf.score(X_test, y_test)
```

Out[126]: 0.8306171578843503

```
In [115]: 1 rf.predict(X_test)
```

Out[115]: array([1178050., 1010750., 865900., ..., 1120000., 620400., 2322600.])

```
In [ ]: 1 rf = RandomForestRegressor(random_state=17)
        2 rf.fit(X_train, y_train)
```

```
In [129]: 1 rf.estimators_[0].predict(X_test)
```

Out[129]: array([1350000., 786000., 890000., ..., 854000., 623500., 270000.])

```
In [130]: 1 rf.estimators_[1].predict(X_test)
```

Out[130]: array([ 910000., 1220000., 795000., ..., 1120000., 560000., 185000.])

```
In [131]: 1 rf.estimators_[2].predict(X_test)
```

Out[131]: array([1250000., 1140000., 875000., ..., 1115000., 680500., 220000.])

```
In [132]: 1 rf.predict(X_test)
```

Out[132]: array([1317370. , 1030150. , 861895. , ..., 1075386.28, 611038.64, 2238520. ])

```
In [134]: 1 y_hat = rf.predict(X_test)
```

```
In [135]: 1 y_hat
```

```
Out[135]: array([1317370. , 1030150. , 861895. , ..., 1075386.28, 61103
      8.64,
      2238520.  ])
```

```
In [136]: 1 y_test
```

```
Out[136]: array([ 981000., 875000., 700000., ..., 932000., 572000., 22000
      00.])
```

```
In [138]: 1 y_hat.shape, y_test.shape
```

```
Out[138]: ((2222,), (2222,))
```

```
In [150]: 1 mse = ((y_hat - y_test) ** 2).mean()
      2 mse
```

```
Out[150]: 67775704512.086395
```

```
In [142]: 1 rmse = np.sqrt(mse)
```

```
In [143]: 1 mse, rmse
```

```
Out[143]: (67775704512.086395, 260337.6740160486)
```

```
In [151]: 1 v = ((y_test - y_test.mean()) ** 2).mean()
      2 v
```

```
Out[151]: 400133234662.6414
```

```
In [153]: 1 score = 1 - mse / v
      2 score
```

```
Out[153]: 0.8306171578843503
```