

adult_data_Classification

April 24, 2018

```
In [64]: # ensemble model KN, naive(gaussiannb) , decisiontree on voting classifier
import warnings
warnings.filterwarnings('ignore')
%matplotlib inline
import matplotlib.pyplot as plt
import pandas as pd
import numpy as np
from sklearn.model_selection import train_test_split
from sklearn.naive_bayes import GaussianNB
from sklearn.tree import DecisionTreeClassifier
from sklearn.neighbors import KNeighborsClassifier
from sklearn.ensemble import VotingClassifier
```

```
In [65]: indexes = ['age', 'workclass', 'fnlwgt', 'education', 'education-num', 'marital-status', 'occu
    'relationship', 'race', 'sex', 'capital-gain', 'capital-loss', 'hours-per-week', 'native-coun
```

```
In [66]: df = pd.read_csv('https://archive.ics.uci.edu/ml/machine-learning-databases/adult/adult
    dff = df
```

```
In [67]: df.head(3)
```

```
Out[67]:
```

	age	workclass	fnlwgt	education	education-num	\
0	39	State-gov	77516	Bachelors	13	
1	50	Self-emp-not-inc	83311	Bachelors	13	
2	38	Private	215646	HS-grad	9	

	marital-status	occupation	relationship	race	sex	\
0	Never-married	Adm-clerical	Not-in-family	White	Male	
1	Married-civ-spouse	Exec-managerial	Husband	White	Male	
2	Divorced	Handlers-cleaners	Not-in-family	White	Male	

	capital-gain	capital-loss	hours-per-week	native-country	earned
0	2174	0	40	United-States	<=50K
1	0	0	13	United-States	<=50K
2	0	0	40	United-States	<=50K

```
In [68]: df['workclass'] = df['workclass'].replace(' ?',df['workclass'].max())
df['occupation'] = df['occupation'].replace(' ?',df['occupation'].max())
```

```
In [69]: earned = df['earned']
```

```
In [70]: df = df.drop('earned',axis=1)
```

```
In [71]: df.head(3)
```

```
Out[71]:
```

	age	workclass	fnlwgt	education	education-num	\
0	39	State-gov	77516	Bachelors	13	
1	50	Self-emp-not-inc	83311	Bachelors	13	
2	38	Private	215646	HS-grad	9	

	marital-status	occupation	relationship	race	sex	\
0	Never-married	Adm-clerical	Not-in-family	White	Male	
1	Married-civ-spouse	Exec-managerial	Husband	White	Male	
2	Divorced	Handlers-cleaners	Not-in-family	White	Male	

	capital-gain	capital-loss	hours-per-week	native-country
0	2174	0	40	United-States
1	0	0	13	United-States
2	0	0	40	United-States

```
In [72]: mylist = list(df.select_dtypes(include=['object']).columns)
df = pd.get_dummies(df, prefix= mylist)
df.head(3)
```

```
Out[72]:
```

	age	fnlwgt	education-num	capital-gain	capital-loss	hours-per-week	\
0	39	77516	13	2174	0	40	
1	50	83311	13	0	0	13	
2	38	215646	9	0	0	40	

	workclass_ Federal-gov	workclass_ Local-gov	workclass_ Never-worked	\
0	0	0	0	
1	0	0	0	
2	0	0	0	

	workclass_ Private	...	native-country_ Portugal	\
0	0	...	0	
1	0	...	0	
2	1	...	0	

	native-country_ Puerto-Rico	native-country_ Scotland	\
0	0	0	
1	0	0	
2	0	0	

	native-country_ South	native-country_ Taiwan	native-country_ Thailand	\
0	0	0	0	
1	0	0	0	
2	0	0	0	

	native-country_ Trinidad&Tobago	native-country_ United-States \
0	0	1
1	0	1
2	0	1

	native-country_ Vietnam	native-country_ Yugoslavia
0	0	0
1	0	0
2	0	0

[3 rows x 106 columns]

```
In [73]: earned = pd.get_dummies(earned)
```

```
In [74]: earned_less_50 = earned[' <=50K']
```

```
In [75]: train_x,test_x, train_y,test_y = train_test_split(df,earned_less_50)
```

```
In [76]: estimators = []
```

```
In [77]: clf1 = GaussianNB()
```

```
In [78]: estimators.append(('gaussiannb', clf1))
```

```
In [79]: clf2 = DecisionTreeClassifier()
```

```
In [80]: estimators.append(('decisiontree', clf2))
```

```
In [81]: clf3 = KNeighborsClassifier()
```

```
In [82]: estimators.append(('kneighbors',clf3))
```

```
In [83]: clf = VotingClassifier(estimators)
```

```
In [84]: clf.fit(train_x,train_y)
```

```
Out[84]: VotingClassifier(estimators=[('gaussiannb', GaussianNB(priors=None)), ('decisiontree',
max_features=None, max_leaf_nodes=None,
min_impurity_decrease=0.0, min_impurity_split=None,
min_samples_leaf=1, min_samples_split=2,
metric_params=None, n_jobs=1, n_neighbors=5, p=2,
weights='uniform'))],
flatten_transform=None, n_jobs=1, voting='hard', weights=None)
```

```
In [85]: score = clf.score(test_x,test_y)
```

```
In [86]: predicted = clf.predict(test_x)
mse = np.mean((predicted-test_y)**2)
```

```
In [87]: print('Score: ',score, 'MSE: ',mse)
```

Score: 0.8164844613683823 MSE: 0.18351553863161774

```
In [88]: # for i,j in zip(predicted,test_y):
#         if i:
#             print('>50K Predicted: %d Real: %d'%(i,j))
#         else:
#             print('<=50K Predicted: %d Real: %d'%(i,j))
```

```
In [89]: test_x['predicted'] = predicted.copy()
test_x.head()
```

```
Out[89]:
```

	age	fnlwgt	education-num	capital-gain	capital-loss	hours-per-week	\
13739	21	121889	10	0	0	20	
28826	21	205844	10	0	0	25	
20154	34	321787	10	0	0	40	
16438	59	140957	11	0	0	35	
24512	28	177955	7	0	0	40	

	workclass_ Federal-gov	workclass_ Local-gov	workclass_ Never-worked	\
13739	0	0	0	
28826	0	0	0	
20154	0	0	0	
16438	0	0	0	
24512	0	0	0	

	workclass_ Private	...	native-country_ Puerto-Rico	\
13739	1	...	0	
28826	1	...	0	
20154	1	...	0	
16438	0	...	0	
24512	1	...	0	

	native-country_ Scotland	native-country_ South	\
13739	0	0	
28826	0	0	
20154	0	0	
16438	0	0	
24512	0	0	

	native-country_ Taiwan	native-country_ Thailand	\
13739	0	0	
28826	0	0	
20154	0	0	
16438	0	0	
24512	0	0	

	native-country_ Trinidad&Tobago	native-country_ United-States	\
--	---------------------------------	-------------------------------	---

13739	0	1
28826	0	1
20154	0	1
16438	0	1
24512	0	0

	native-country_ Vietnam	native-country_ Yugoslavia	predicted
13739	0	0	1
28826	0	0	1
20154	0	0	1
16438	0	0	1
24512	0	0	1

[5 rows x 107 columns]

```
In [90]: import seaborn as sns
sns.set(color_codes=True)
```

```
In [91]: sns.stripplot(x="predicted", y="capital-loss", data=test_x,jitter=True)
```

```
Out[91]: <matplotlib.axes._subplots.AxesSubplot at 0x7f7b0ca92780>
```

