

```
In [1]: import pandas as pd
import numpy as np
import seaborn as sns
import matplotlib.pyplot as plt
```

```
In [2]: pd.set_option("display.max_rows",None)
pd.set_option("display.max_columns",None)
```

```
In [3]: url = 'https://archive.ics.uci.edu/ml/machine-learning-databases/adult/adult.data'
```

```
In [4]: columns = ['age', 'workclass', 'fnlwgt', 'education', 'education-num', 'marital-status', 'occupation', 'relationship', 'race', 'sex', 'capital-gain', 'capital-loss', 'hours-per-week', 'native-country', 'salary', '']
```

```
In [5]: df = pd.read_csv(url, names = columns)
df.head(2)
```

Out[5]:

	age	workclass	fnlwgt	education	education-num	marital-status	occupation	relationship	race	sex	capital-gain	capital-loss	hours-per-week	native-country	salary
0	39	State-gov	77516	Bachelors	13	Never-married	Adm-clerical	Not-in-family	White	Male	2174	0	40	United-States	<=50K
1	50	Self-emp-not-inc	83311	Bachelors	13	Married-civ-spouse	Exec-managerial	Husband	White	Male	0	0	13	United-States	<=50K

```
In [6]: df.shape
```

Out[6]: (32561, 15)

```
In [7]: df.describe()
```

```
Out[7]:
```

	age	fnlwgt	education-num	capital-gain	capital-loss	hours-per-week
count	32561.000000	3.256100e+04	32561.000000	32561.000000	32561.000000	32561.000000
mean	38.581647	1.897784e+05	10.080679	1077.648844	87.303830	40.437456
std	13.640433	1.055500e+05	2.572720	7385.292085	402.960219	12.347429
min	17.000000	1.228500e+04	1.000000	0.000000	0.000000	1.000000
25%	28.000000	1.178270e+05	9.000000	0.000000	0.000000	40.000000
50%	37.000000	1.783560e+05	10.000000	0.000000	0.000000	40.000000
75%	48.000000	2.370510e+05	12.000000	0.000000	0.000000	45.000000
max	90.000000	1.484705e+06	16.000000	99999.000000	4356.000000	99.000000

```
In [8]: df.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 32561 entries, 0 to 32560
Data columns (total 15 columns):
#   Column                Non-Null Count  Dtype
---  -
0   age                    32561 non-null  int64
1   workclass              32561 non-null  object
2   fnlwgt                 32561 non-null  int64
3   education              32561 non-null  object
4   education-num          32561 non-null  int64
5   marital-status         32561 non-null  object
6   occupation             32561 non-null  object
7   relationship           32561 non-null  object
8   race                   32561 non-null  object
9   sex                    32561 non-null  object
10  capital-gain            32561 non-null  int64
11  capital-loss            32561 non-null  int64
12  hours-per-week          32561 non-null  int64
13  native-country          32561 non-null  object
14  salary                  32561 non-null  object
dtypes: int64(6), object(9)
memory usage: 3.7+ MB
```

```
In [9]: df.isnull().sum()
```

```
Out[9]: age                0
workclass                0
fnlwgt                  0
education                0
education-num           0
marital-status          0
occupation              0
relationship            0
race                    0
sex                     0
capital-gain            0
capital-loss            0
hours-per-week          0
native-country          0
salary                  0
dtype: int64
```

```
In [10]: df.workclass.unique()
```

```
Out[10]: array([' State-gov', ' Self-emp-not-inc', ' Private', ' Federal-gov',
               ' Local-gov', ' ?', ' Self-emp-inc', ' Without-pay',
               ' Never-worked'], dtype=object)
```

```
In [11]: df.workclass.value_counts()
```

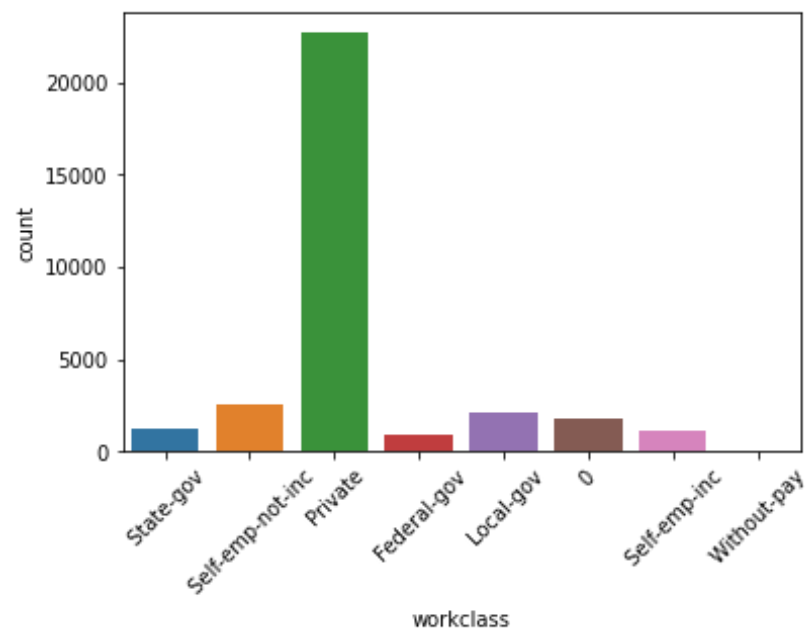
```
Out[11]: Private                22696
Self-emp-not-inc              2541
Local-gov                     2093
?                              1836
State-gov                     1298
Self-emp-inc                   1116
Federal-gov                     960
Without-pay                     14
Never-worked                     7
Name: workclass, dtype: int64
```

```
In [12]: df = df.replace(' Never-worked', ' Without-pay')
df['workclass'].value_counts()
```

```
Out[12]: Private                22696
Self-emp-not-inc             2541
Local-gov                   2093
?                           1836
State-gov                   1298
Self-emp-inc                1116
Federal-gov                 960
Without-pay                 21
Name: workclass, dtype: int64
```

```
In [13]: df.replace(' ?', np.nan, inplace= True)
df['workclass'].fillna('0', inplace=True)
```

```
In [14]: sns.countplot(x = df['workclass'])  
plt.xticks(rotation = 45)  
plt.show()
```



```
In [15]: df['salary'].unique()
```

```
Out[15]: array([' <=50K', ' >50K'], dtype=object)
```

```
In [16]: salary = {' <=50K': 0 , ' >50K': '1'}
df = df.replace(salary)
df.head(2)
```

Out[16]:

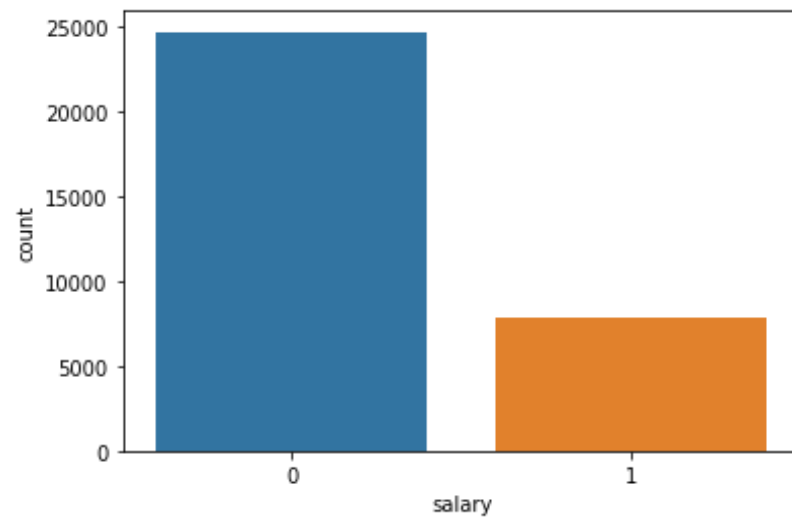
	age	workclass	fnlwgt	education	education-num	marital-status	occupation	relationship	race	sex	capital-gain	capital-loss	hours-per-week	native-country	salary
0	39	State-gov	77516	Bachelors	13	Never-married	Adm-clerical	Not-in-family	White	Male	2174	0	40	United-States	0
1	50	Self-emp-not-inc	83311	Bachelors	13	Married-civ-spouse	Exec-managerial	Husband	White	Male	0	0	13	United-States	0

```
In [17]: df['salary'].value_counts()
```

```
Out[17]: 0    24720
1     7841
Name: salary, dtype: int64
```

```
In [18]: sns.countplot(x=df['salary'])  
plt.xticks(rotation = 0)
```

```
Out[18]: (array([0, 1]), [Text(0, 0, '0'), Text(1, 0, '1')])
```

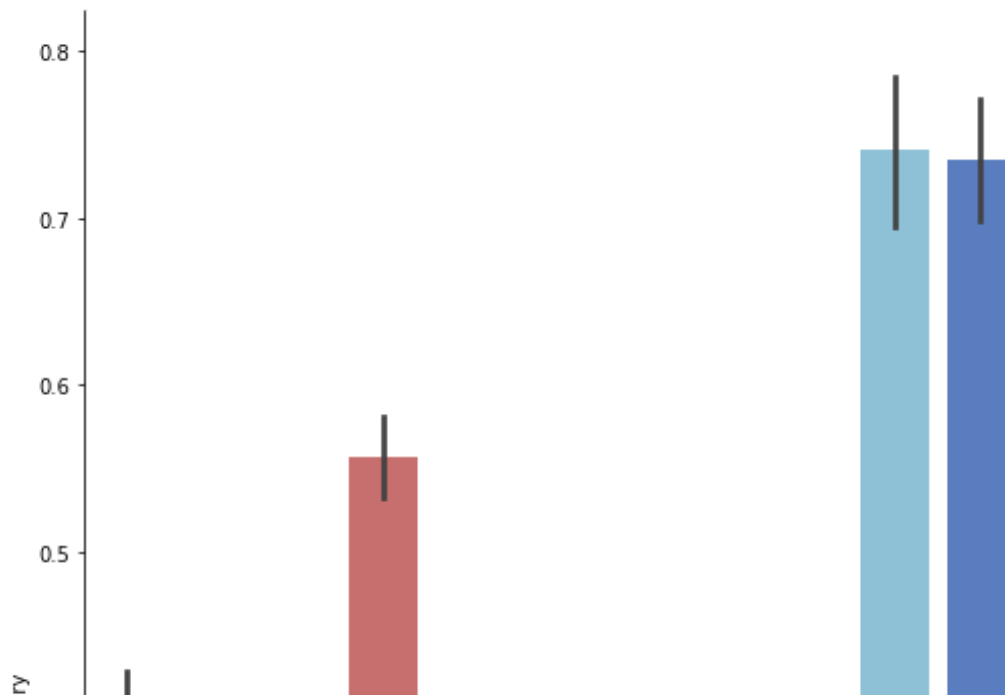


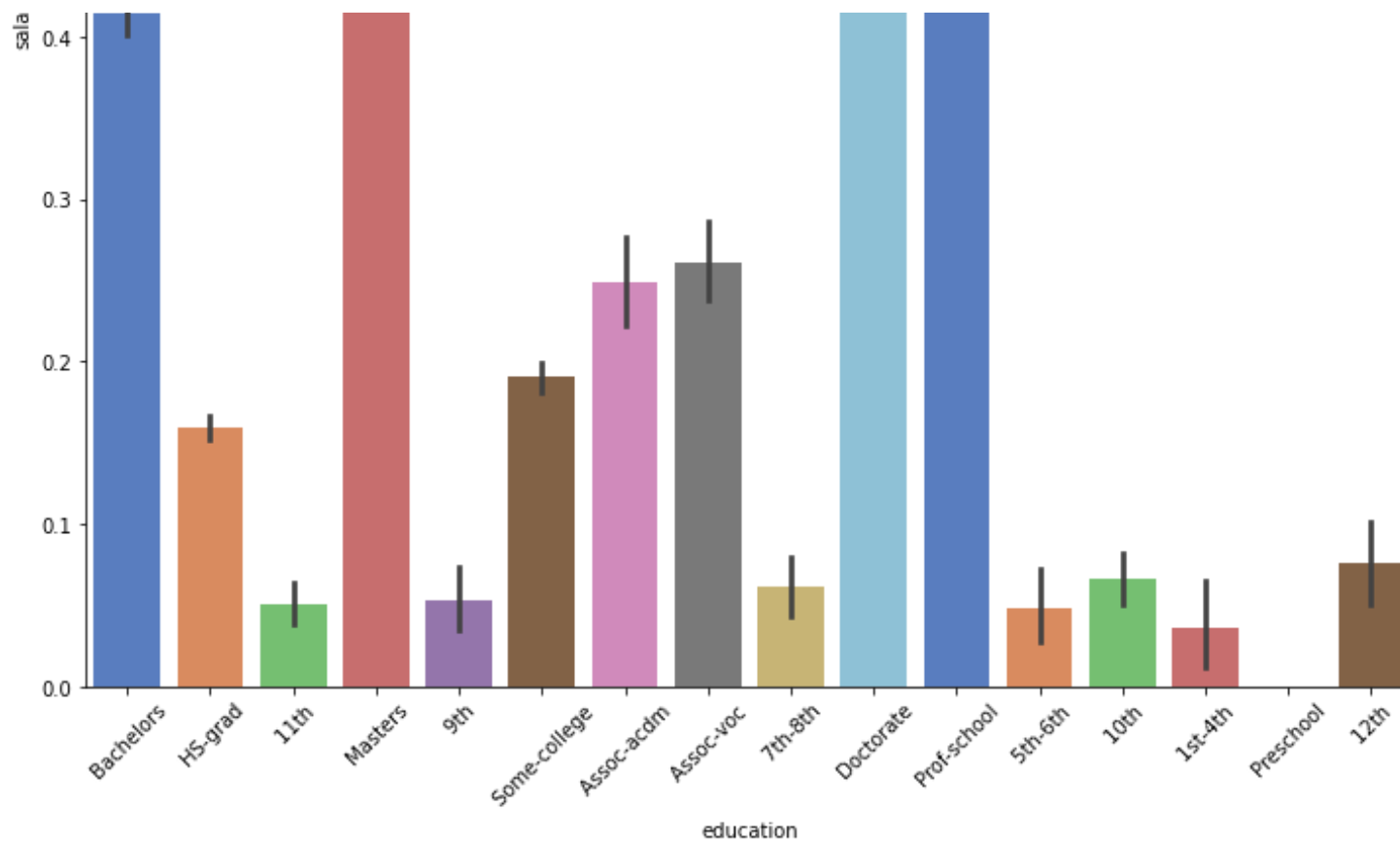

```
In [19]: df['education'].value_counts()
```

```
Out[19]: HS-grad      10501  
Some-college  7291  
Bachelors    5355  
Masters      1723  
Assoc-voc    1382  
11th         1175  
Assoc-acdm   1067  
10th         933  
7th-8th      646  
Prof-school  576  
9th          514  
12th         433  
Doctorate    413  
5th-6th      333  
1st-4th      168  
Preschool    51  
Name: education, dtype: int64
```

```
In [20]: sns.catplot(x='education',y=pd.to_numeric(df['salary']),data=df,height=10,palette='muted',kind='bar')
plt.xticks(rotation=45)
```

```
Out[20]: (array([ 0,  1,  2,  3,  4,  5,  6,  7,  8,  9, 10, 11, 12, 13, 14, 15]),
 [Text(0, 0, ' Bachelors'),
  Text(1, 0, ' HS-grad'),
  Text(2, 0, ' 11th'),
  Text(3, 0, ' Masters'),
  Text(4, 0, ' 9th'),
  Text(5, 0, ' Some-college'),
  Text(6, 0, ' Assoc-acdm'),
  Text(7, 0, ' Assoc-voc'),
  Text(8, 0, ' 7th-8th'),
  Text(9, 0, ' Doctorate'),
  Text(10, 0, ' Prof-school'),
  Text(11, 0, ' 5th-6th'),
  Text(12, 0, ' 10th'),
  Text(13, 0, ' 1st-4th'),
  Text(14, 0, ' Preschool'),
  Text(15, 0, ' 12th')])
```





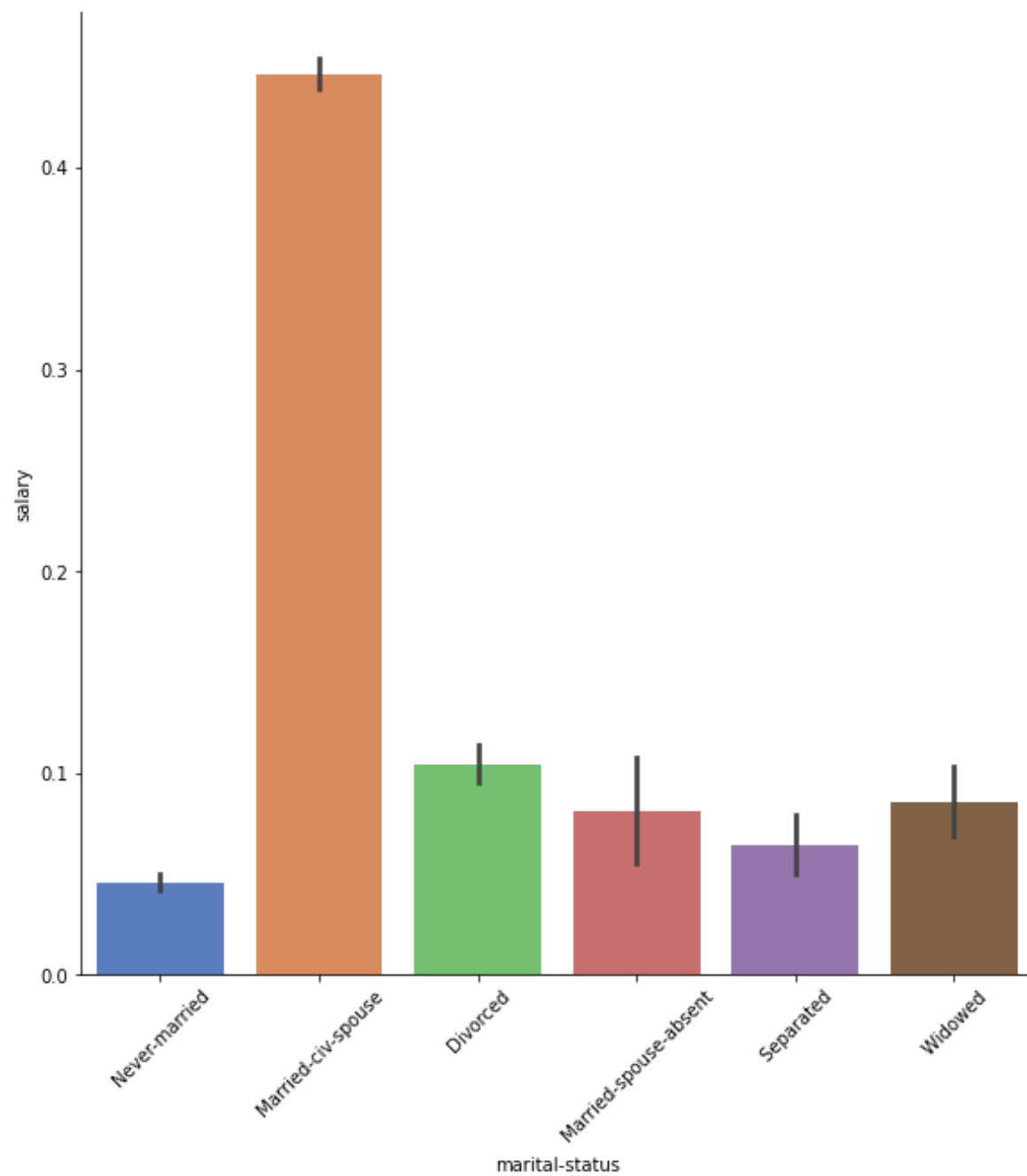
```
In [21]: df['marital-status'].value_counts()
```

```
Out[21]: Married-civ-spouse    14976
Never-married    10683
Divorced    4443
Separated    1025
Widowed    993
Married-spouse-absent    418
Married-AF-spouse    23
Name: marital-status, dtype: int64
```

```
In [22]: df['marital-status'].replace(' Married-AF-spouse', ' Married-civ-spouse',inplace=True)
```

```
In [23]: sns.catplot(x='marital-status',y=pd.to_numeric(df['salary']),data=df,palette='muted',kind='bar',height=8)
plt.xticks(rotation=45)
```

```
Out[23]: (array([0, 1, 2, 3, 4, 5]),
 [Text(0, 0, ' Never-married'),
  Text(1, 0, ' Married-civ-spouse'),
  Text(2, 0, ' Divorced'),
  Text(3, 0, ' Married-spouse-absent'),
  Text(4, 0, ' Separated'),
  Text(5, 0, ' Widowed')])
```



```
In [24]: df['occupation'].fillna('0',inplace=True)
df['occupation'].value_counts()
```

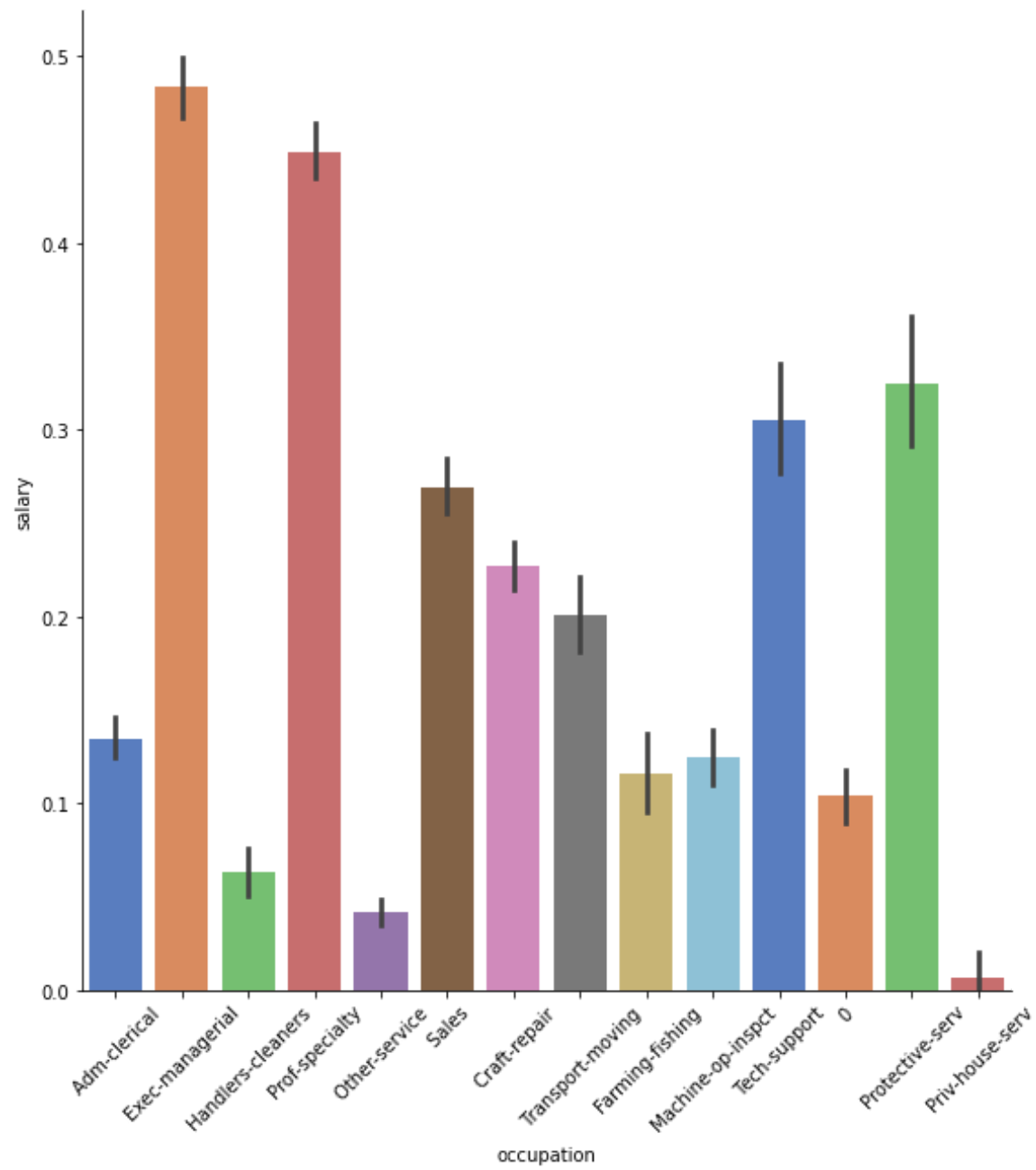
```
Out[24]: Prof-specialty      4140
Craft-repair      4099
Exec-managerial   4066
Adm-clerical      3770
Sales             3650
Other-service     3295
Machine-op-inspct 2002
0                1843
Transport-moving  1597
Handlers-cleaners 1370
Farming-fishing   994
Tech-support      928
Protective-serv   649
Priv-house-serv   149
Armed-Forces       9
Name: occupation, dtype: int64
```

```
In [25]: df['occupation'].replace(' Armed-Forces','0',inplace=True)
df['occupation'].value_counts()
```

```
Out[25]: Prof-specialty      4140
Craft-repair      4099
Exec-managerial   4066
Adm-clerical      3770
Sales             3650
Other-service     3295
Machine-op-inspct 2002
0                1852
Transport-moving  1597
Handlers-cleaners 1370
Farming-fishing   994
Tech-support      928
Protective-serv   649
Priv-house-serv   149
Name: occupation, dtype: int64
```

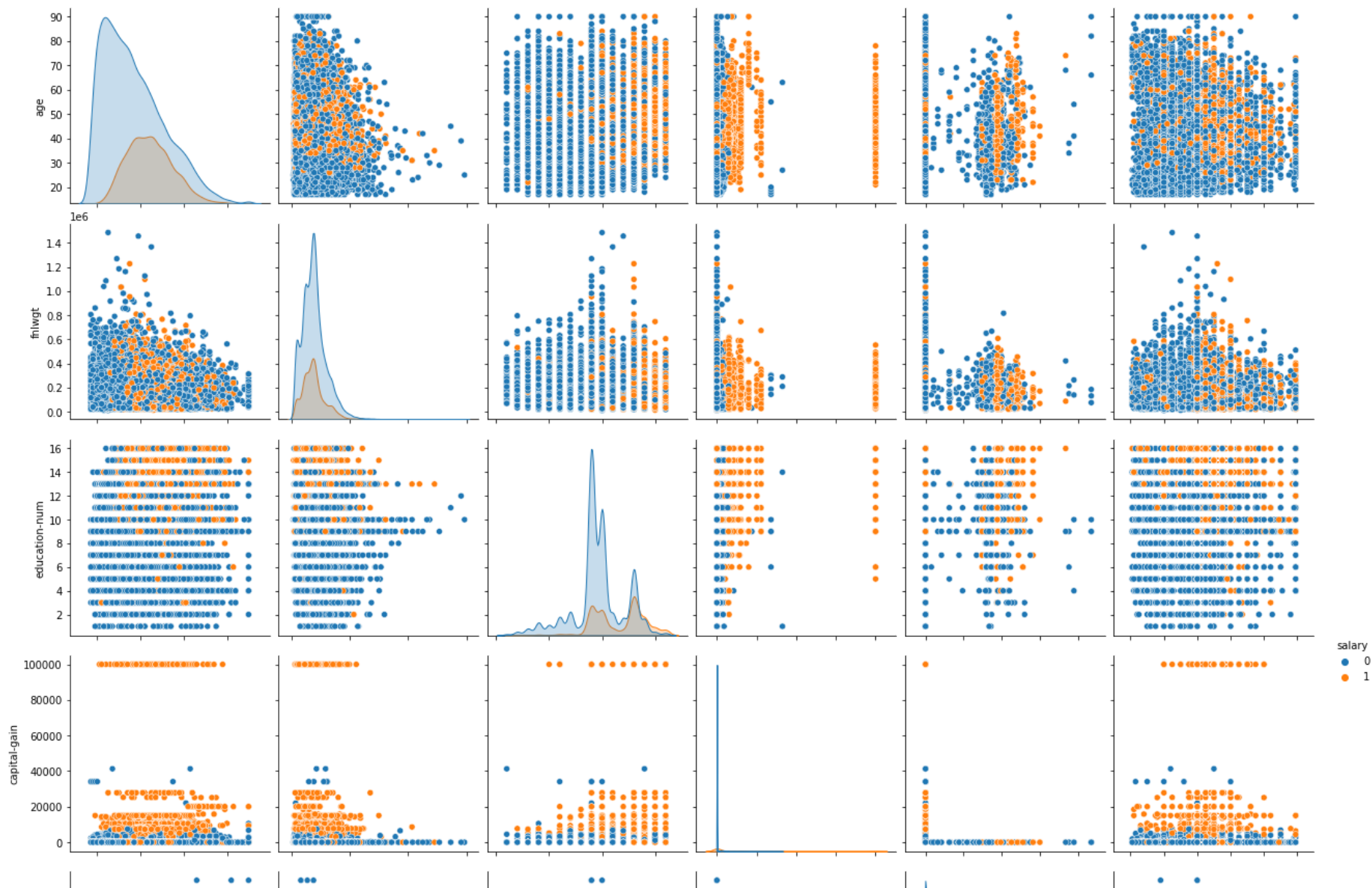
```
In [26]: sns.catplot(x='occupation',y=pd.to_numeric(df['salary']),data=df,palette='muted',kind='bar',height=8)
plt.xticks(rotation=45)
```

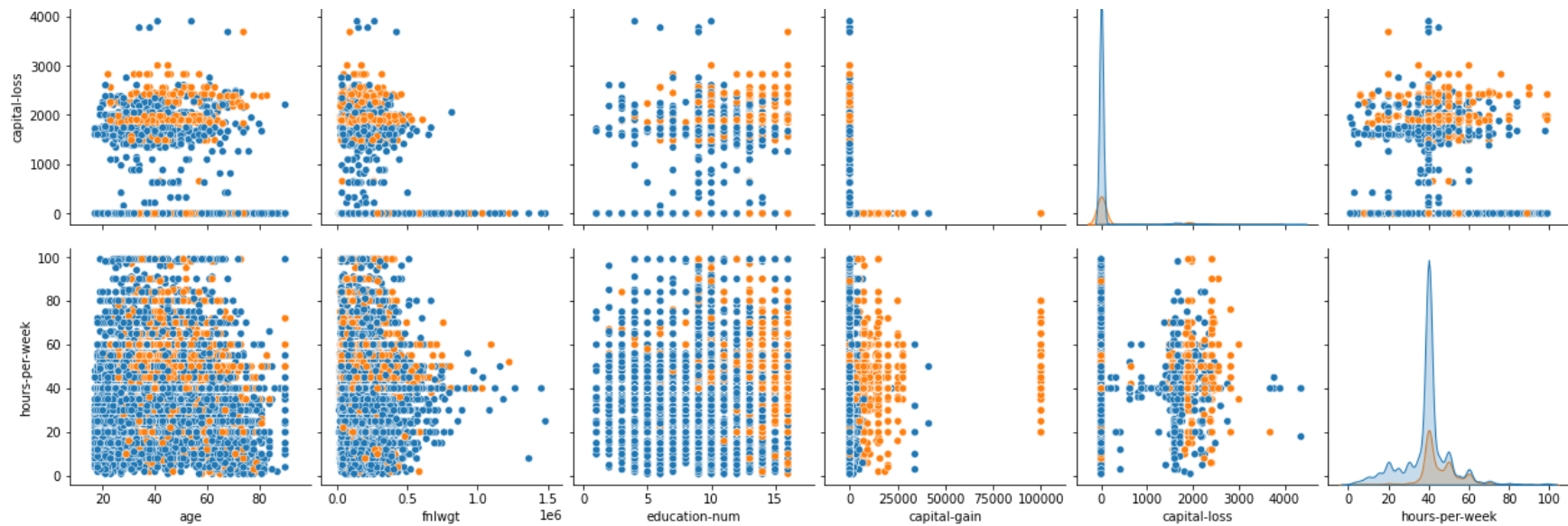
```
Out[26]: (array([ 0,  1,  2,  3,  4,  5,  6,  7,  8,  9, 10, 11, 12, 13]),
 [Text(0, 0, ' Adm-clerical'),
  Text(1, 0, ' Exec-managerial'),
  Text(2, 0, ' Handlers-cleaners'),
  Text(3, 0, ' Prof-specialty'),
  Text(4, 0, ' Other-service'),
  Text(5, 0, ' Sales'),
  Text(6, 0, ' Craft-repair'),
  Text(7, 0, ' Transport-moving'),
  Text(8, 0, ' Farming-fishing'),
  Text(9, 0, ' Machine-op-inspct'),
  Text(10, 0, ' Tech-support'),
  Text(11, 0, '0'),
  Text(12, 0, ' Protective-serv'),
  Text(13, 0, ' Priv-house-serv')])
```




```
In [28]: sns.pairplot(df,hue='salary',height=3)  
plt.plot()
```

```
Out[28]: []
```





```
In [34]: corr = df.corr()
sns.heatmap(corr, annot = True, cmap='YlGnBu')
```

Out[34]: <AxesSubplot:>



```
In [35]: df.drop('fnlwgt',axis=1,inplace=True)
```

```
In [36]: df.head(n=2)
```

Out[36]:

	age	workclass	education	education-num	marital-status	occupation	relationship	race	sex	capital-gain	capital-loss	hours-per-week	native-country	salary
0	39	State-gov	Bachelors	13	Never-married	Adm-clerical	Not-in-family	White	Male	2174	0	40	United-States	0
1	50	Self-emp-not-inc	Bachelors	13	Married-civ-spouse	Exec-managerial	Husband	White	Male	0	0	13	United-States	0

```
In [37]: X = df.drop('salary',axis=1)
y = pd.to_numeric(df['salary'])
```

```
In [38]: X_d = pd.get_dummies(X)
X_d.head(2)
```

Out[38]:

	age	education- num	capital- gain	capital- loss	hours- per- week	workclass_ Federal- gov	workclass_ Local-gov	workclass_ Private	workclass_ Self-emp- inc	workclass_ Self-emp- not-inc	workclass_ State-gov	workclass_ Without- pay	workclass_0
0	39	13	2174	0	40	0	0	0	0	0	1	0	0
1	50	13	0	0	13	0	0	0	0	1	0	0	0

```
In [39]: from sklearn.model_selection import train_test_split,GridSearchCV,StratifiedKFold
x_train,x_test,y_train,y_test = train_test_split(X_d,y,test_size=0.3,random_state=101)
```

```
In [41]: from sklearn.tree import DecisionTreeClassifier
from sklearn.ensemble import RandomForestClassifier
```

```
In [42]: classifier = [DecisionTreeClassifier(random_state=42),RandomForestClassifier(random_state=42)]
```

```
In [43]: dt_grid_param = { "min_samples_split" : range(10,500,20),
                        "max_depth": range(1,20,2)

}
```

```
In [46]: rf_grid_param = {"max_features": [1,3,10],
                        "min_samples_split":[2,3,10],
                        "min_samples_leaf":[1,3,10],
                        "bootstrap":[False],
                        "n_estimators":[100,300],
                        "criterion":["gini"]}
```

```
In [47]: classifier_param = [dt_grid_param, rf_grid_param]
```

```
In [48]: cv_result = []
best_estimators = []
for i in range(len(classifier)):
    clf = GridSearchCV(classifier[i], param_grid=classifier_param[i], cv = StratifiedKFold(n_splits = 10), scoring = "a
    clf.fit(x_train, y_train)
    cv_result.append(clf.best_score_)
    best_estimators.append(clf.best_estimator_)
print(cv_result[i])
```

Fitting 10 folds for each of 250 candidates, totalling 2500 fits
0.8584153560733778
Fitting 10 folds for each of 54 candidates, totalling 540 fits
0.8646456971740454

```
In [50]: cv_results = pd.DataFrame({"Cross Validation Means":cv_result, "ML Models":["DecisionTreeClassifier", "RandomForestClas
```

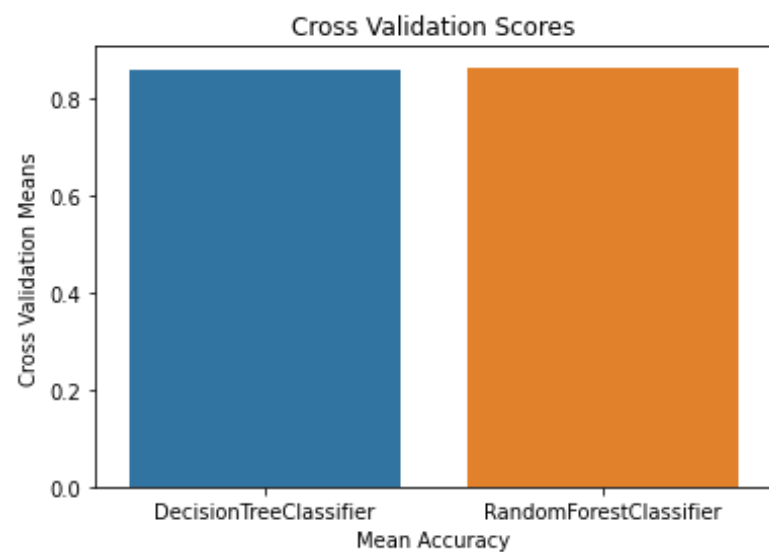
```
In [55]: cv_results
```

Out[55]:

	Cross Validation Means	ML Models
0	0.858415	DecisionTreeClassifier
1	0.864646	RandomForestClassifier

```
In [56]: g = sns.barplot(y="Cross Validation Means", x="ML Models", data = cv_results)
g.set_xlabel("Mean Accuracy")
g.set_title("Cross Validation Scores")
```

Out[56]: Text(0.5, 1.0, 'Cross Validation Scores')



In []:

