Adult Cencus Data Analysis

June 16, 2022

1 Import Basic Libraries

```
[34]: import pandas as pd
      import numpy as np
      import matplotlib.pyplot as plt
[35]: #Load the dataset
      df=pd.read_csv('adult.csv')
[36]: df.head()
[36]:
                                            education
                                                       education-num
         age
                       workclass
                                  fnlwgt
      0
          39
                       State-gov
                                   77516
                                            Bachelors
                                                                   13
      1
          50
               Self-emp-not-inc
                                   83311
                                            Bachelors
                                                                   13
      2
          38
                         Private
                                  215646
                                              HS-grad
                                                                    9
      3
          53
                         Private
                                  234721
                                                 11th
                                                                    7
      4
                                                                   13
          28
                         Private
                                  338409
                                            Bachelors
              marital-status
                                        occupation
                                                      relationship
                                                                       race
                                                                                  sex
      0
               Never-married
                                      Adm-clerical
                                                     Not-in-family
                                                                      White
                                                                                 Male
          Married-civ-spouse
                                  Exec-managerial
                                                            Husband
                                                                      White
                                                                                 Male
      1
                                                     Not-in-family
                                                                                 Male
      2
                     Divorced
                                Handlers-cleaners
                                                                      White
      3
          Married-civ-spouse
                                Handlers-cleaners
                                                            Husband
                                                                      Black
                                                                                 Male
          Married-civ-spouse
                                   Prof-specialty
                                                               Wife
                                                                      Black
                                                                               Female
         capital-gain capital-loss
                                      hours-per-week
                                                               country
                                                                        salary
      0
                 2174
                                   0
                                                         United-States
                                                                          <=50K
      1
                     0
                                   0
                                                   13
                                                         United-States
                                                                          <=50K
      2
                     0
                                   0
                                                   40
                                                         United-States
                                                                          <=50K
                                                                          <=50K
      3
                     0
                                   0
                                                   40
                                                         United-States
      4
                     0
                                   0
                                                   40
                                                                  Cuba
                                                                          <=50K
```

2 Data cleaning

```
[37]: #check for null values
      df.isnull().sum().sum()
[37]: 0
[38]: df.columns
[38]: Index(['age', 'workclass', 'fnlwgt', 'education', 'education-num',
             'marital-status', 'occupation', 'relationship', 'race', 'sex',
             'capital-gain', 'capital-loss', 'hours-per-week', 'country', 'salary'],
            dtype='object')
[39]: df['salary'].unique()
[39]: array([' <=50K', ' >50K'], dtype=object)
      df.groupby('salary').mean()
[40]:
                                        education-num capital-gain capital-loss \
                                fnlwgt
                    age
      salary
       <=50K
              36.783738
                         190340.86517
                                             9.595065
                                                          148.752468
                                                                         53.142921
       >50K
              44.249841
                         188005.00000
                                                         4006.142456
                                                                        195.001530
                                            11.611657
              hours-per-week
      salary
       <=50K
                   38.840210
       >50K
                   45.473026
[41]: df.describe().T
[41]:
                        count
                                                                   min
                                                                             25% \
                                         mean
                                                          std
                      32561.0
                                    38.581647
                                                   13.640433
                                                                  17.0
                                                                            28.0
      age
                                                               12285.0
                                                                        117827.0
      fnlwgt
                      32561.0
                                189778.366512
                                              105549.977697
                                    10.080679
      education-num
                      32561.0
                                                    2.572720
                                                                   1.0
                                                                             9.0
                                                                   0.0
                                                                             0.0
      capital-gain
                      32561.0
                                  1077.648844
                                                 7385.292085
      capital-loss
                      32561.0
                                    87.303830
                                                  402.960219
                                                                   0.0
                                                                             0.0
      hours-per-week
                                                                   1.0
                                                                            40.0
                      32561.0
                                    40.437456
                                                   12.347429
                            50%
                                      75%
                                                 max
                           37.0
                                     48.0
                                                90.0
      age
      fnlwgt
                      178356.0
                                 237051.0
                                           1484705.0
      education-num
                           10.0
                                     12.0
                                                16.0
      capital-gain
                            0.0
                                      0.0
                                             99999.0
      capital-loss
                            0.0
                                      0.0
                                              4356.0
     hours-per-week
                           40.0
                                     45.0
                                                99.0
```

```
[42]: df['workclass'].value_counts()
[42]: 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
     Maximum people are working in Private sector
[43]: df['education'].value_counts()
[43]:
      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
     Maximum people has done their High School
[44]: df['marital-status'].value_counts()
[44]: Married-civ-spouse
                                 14976
       Never-married
                                 10683
       Divorced
                                  4443
       Separated
                                  1025
                                   993
       Widowed
       Married-spouse-absent
                                   418
       Married-AF-spouse
      Name: marital-status, dtype: int64
```

Maximum people are married with civilian spouse

```
[45]: df['relationship'].value_counts()
[45]:
      Husband
                         13193
       Not-in-family
                          8305
       Own-child
                          5068
       Unmarried
                          3446
       Wife
                          1568
                           981
       Other-relative
      Name: relationship, dtype: int64
     Maximum people are Husband who are working
[46]: df['sex'].value_counts()
[46]:
      Male
                 21790
                 10771
       Female
      Name: sex, dtype: int64
     No. of males is twice than females
[47]: df['occupation'].value_counts()
[47]: Prof-specialty
                            4140
       Craft-repair
                            4099
       Exec-managerial
                            4066
       Adm-clerical
                            3770
       Sales
                            3650
       Other-service
                            3295
       Machine-op-inspct
                            2002
                            1843
       Transport-moving
                            1597
       Handlers-cleaners
                            1370
       Farming-fishing
                             994
                             928
       Tech-support
       Protective-serv
                             649
       Priv-house-serv
                             149
       Armed-Forces
      Name: occupation, dtype: int64
     Maximum people has occupation as prof-speciality(professor in a perticular subject )
     There are very less people who are in Armed forces
[48]: df['country'].value_counts()
[48]:
      United-States
                                      29170
       Mexico
                                        643
                                        583
       Philippines
                                        198
       Germany
                                        137
```

```
121
Canada
Puerto-Rico
                                  114
El-Salvador
                                  106
 India
                                  100
Cuba
                                   95
England
                                   90
 Jamaica
                                   81
South
                                   80
China
                                   75
Italy
                                   73
Dominican-Republic
                                   70
Vietnam
                                   67
Guatemala
                                   64
                                   62
 Japan
Poland
                                   60
Columbia
                                   59
Taiwan
                                   51
Haiti
                                   44
Iran
                                   43
                                   37
Portugal
Nicaragua
                                   34
Peru
                                   31
Greece
                                   29
France
                                   29
Ecuador
                                   28
Ireland
                                   24
Hong
                                   20
Cambodia
                                   19
                                   19
Trinadad&Tobago
Thailand
                                   18
Laos
                                   18
Yugoslavia
                                   16
 Outlying-US(Guam-USVI-etc)
                                   14
Hungary
                                   13
Honduras
                                   13
Scotland
                                   12
Holand-Netherlands
                                    1
Name: country, dtype: int64
```

No. of employment in united satate is maximum than other countries

```
[49]: #Filling ?
    df['workclass']=df['workclass'].replace(' ?','private')
    df['country']=df['country'].replace(' ?','United-States')
    df['occupation']=df['occupation'].replace(' ?','prof-spaciality')
```

3 Feature engineering

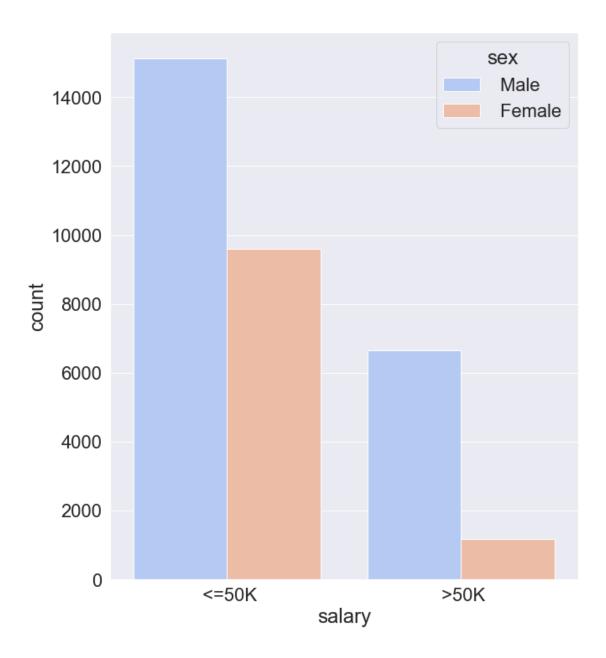
```
[50]: df.education=df.education.replace(['Preschool', '1st-4th', '5th-6th', 'L
      →7th-8th',' 9th',' 10th',' 11th',' 12th',],'School')
      df.education=df.education.replace(' HS-grad', 'High-School')
      df.education=df.education.replace([' Assoc-acdm',' Assoc-voc',' Some-college','_
      →Prof-school'], 'Higher')
      df.education=df.education.replace(' Bachelors','Graduates')
      df.education=df.education.replace(' Doctorate', 'Doc')
[51]: df['education'].unique()
[51]: array(['Graduates', 'High-School', 'School', 'Masters', 'Higher', 'Doc'],
            dtype=object)
[52]: df['marital-status']=df['marital-status'].replace([' Married-spouse-absent','|

→Married-civ-spouse',' Married-AF-spouse'],'married')
      df['marital-status']=df['marital-status'].replace([' Divorced',' Separated','_
       →Widowed'], 'others')
[53]: df['marital-status'].unique()
[53]: array([' Never-married', 'married', 'others'], dtype=object)
[54]: #what is the marital status whose working hour per week is maximum
      df.groupby(['marital-status'])['hours-per-week'].mean()
[54]: marital-status
      Never-married
                       36.939998
                       43.183628
     married
      others
                       39.667544
     Name: hours-per-week, dtype: float64
     4 Visualization
[55]: import matplotlib.pyplot as plt
      import seaborn as sns
     5 Which sex category is earning greater than 50k
```

```
[56]: plt.figure(figsize=(10,12))
sns.set(font_scale=2)
sns.countplot(df['salary'],palette='coolwarm',hue='sex',data=df)
plt.show()
```

C:\Users\hp\AppData\Roaming\Python\Python38\sitepackages\seaborn_decorators.py:36: FutureWarning:

Pass the following variable as a keyword arg: x. From version 0.12, the only valid positional argument will be `data`, and passing other arguments without an explicit keyword will result in an error or misinterpretation.



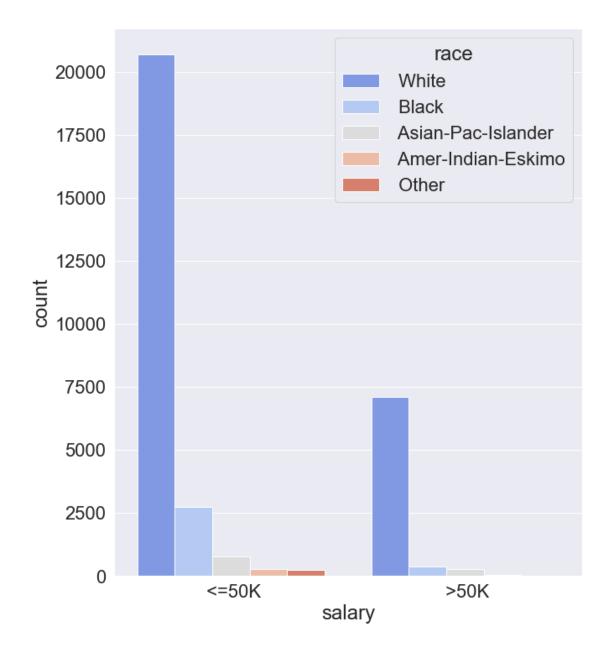
No. of males are earning more than 50k than Females

6 Which type of people in race are earning more than others

```
[57]: plt.figure(figsize=(10,12))
sns.set(font_scale=2)
sns.countplot(df['salary'],palette='coolwarm',hue='race',data=df)
plt.show()
```

C:\Users\hp\AppData\Roaming\Python\Python38\sitepackages\seaborn_decorators.py:36: FutureWarning:

Pass the following variable as a keyword arg: x. From version 0.12, the only valid positional argument will be `data`, and passing other arguments without an explicit keyword will result in an error or misinterpretation.



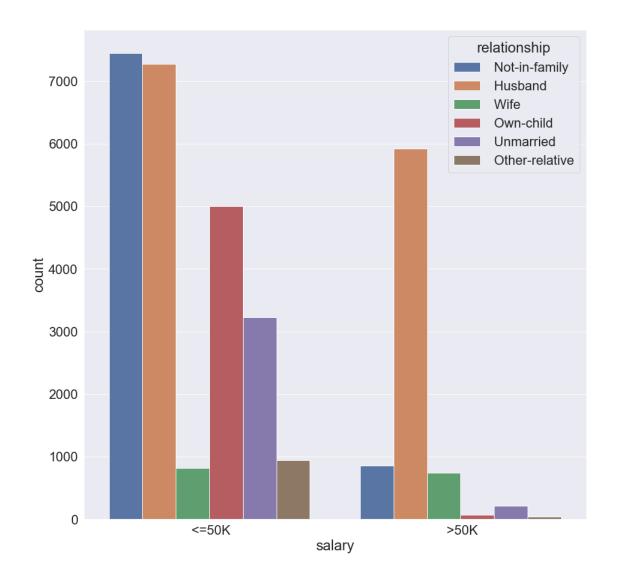
No. of White people are earning more salary than others

7 What is the relationship of people who is earning more than 50k

```
[58]: plt.figure(figsize=(15,15))
sns.set(font_scale=2)
sns.countplot(df['salary'],hue='relationship',data=df);
```

C:\Users\hp\AppData\Roaming\Python\Python38\sitepackages\seaborn_decorators.py:36: FutureWarning:

Pass the following variable as a keyword arg: x. From version 0.12, the only valid positional argument will be `data`, and passing other arguments without an explicit keyword will result in an error or misinterpretation.

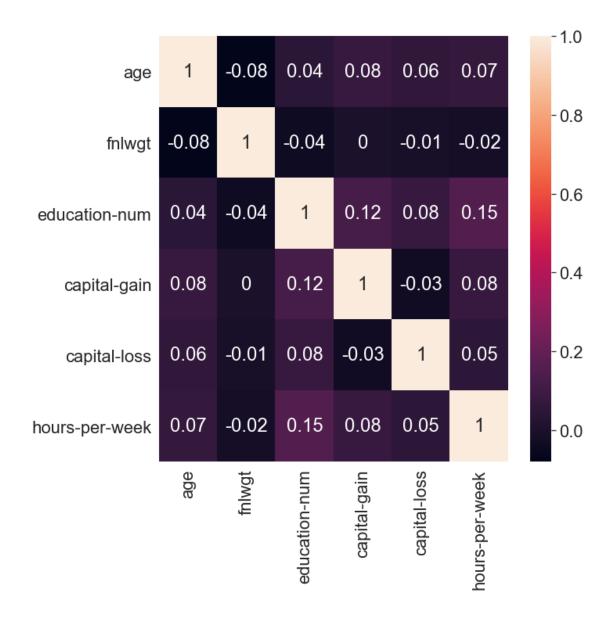


People who are Not-in-family are earning more

8 finding the correlation among all the numerical variables

| [59]: | df.corr() | | | | | | |
|-------|-----------|-----------|-----------|---------------|--------------|--------------|---|
| [59]: | | age | fnlwgt | education-num | capital-gain | capital-loss | \ |
| | age | 1.000000 | -0.076646 | 0.036527 | 0.077674 | 0.057775 | |
| | fnlwgt | -0.076646 | 1.000000 | -0.043195 | 0.000432 | -0.010252 | |

```
0.036527 -0.043195
                                               1.000000
                                                             0.122630
                                                                           0.079923
      education-num
      capital-gain
                      0.077674 0.000432
                                               0.122630
                                                             1.000000
                                                                          -0.031615
      capital-loss
                      0.057775 -0.010252
                                               0.079923
                                                            -0.031615
                                                                           1.000000
     hours-per-week 0.068756 -0.018768
                                               0.148123
                                                             0.078409
                                                                           0.054256
                     hours-per-week
                           0.068756
      age
      fnlwgt
                           -0.018768
      education-num
                            0.148123
      capital-gain
                            0.078409
      capital-loss
                            0.054256
     hours-per-week
                            1.000000
[60]: plt.figure(figsize=(10,10))
      sns.set(font_scale=2)
     sns.heatmap(np.round(df.corr(),2),annot=True)
      plt.show()
```



9 What is the % of education background in the dataset

[61]: px.pie(df,values='education-num',names='education',title='% of Education')

<IPython.core.display.Javascript object>

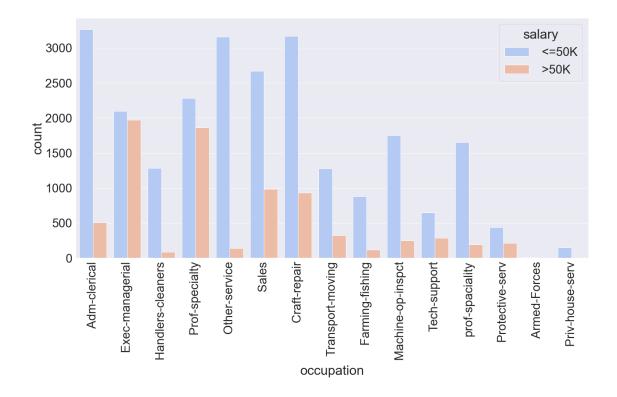
There are maximum people who had done their higher education

10 Which occupation has maximum salary than other occupations

```
[62]: plt.figure(figsize=(25,12))
    sns.set(font_scale=3)
    sns.countplot(df['occupation'],hue='salary',data=df,palette='coolwarm')
    plt.xticks(rotation=90);
```

C:\Users\hp\AppData\Roaming\Python\Python38\sitepackages\seaborn_decorators.py:36: FutureWarning:

Pass the following variable as a keyword arg: x. From version 0.12, the only valid positional argument will be `data`, and passing other arguments without an explicit keyword will result in an error or misinterpretation.



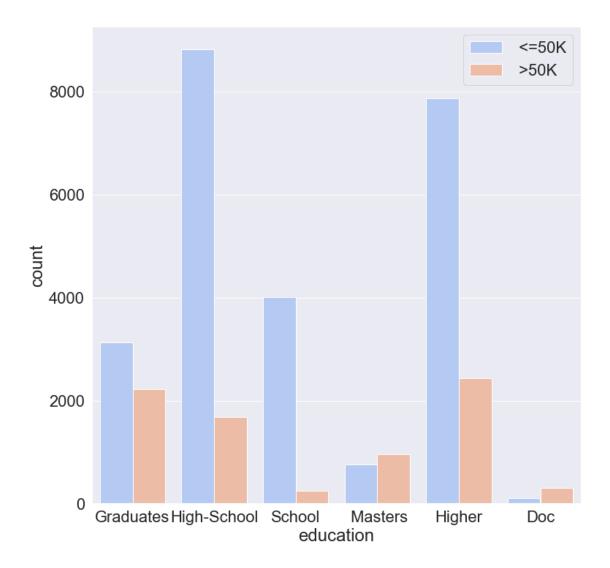
There are more People whose job role is Executive-manager is earning more than 50k

11 How the educational background is related with salary

```
[63]: plt.figure(figsize=(12,12))
    sns.set(font_scale=2)
    sns.countplot(df['education'],hue='salary',data=df,palette='coolwarm')
    plt.legend(loc='upper right', bbox_to_anchor=(1, 1.0));
```

C:\Users\hp\AppData\Roaming\Python\Python38\sitepackages\seaborn_decorators.py:36: FutureWarning:

Pass the following variable as a keyword arg: x. From version 0.12, the only valid positional argument will be `data`, and passing other arguments without an explicit keyword will result in an error or misinterpretation.



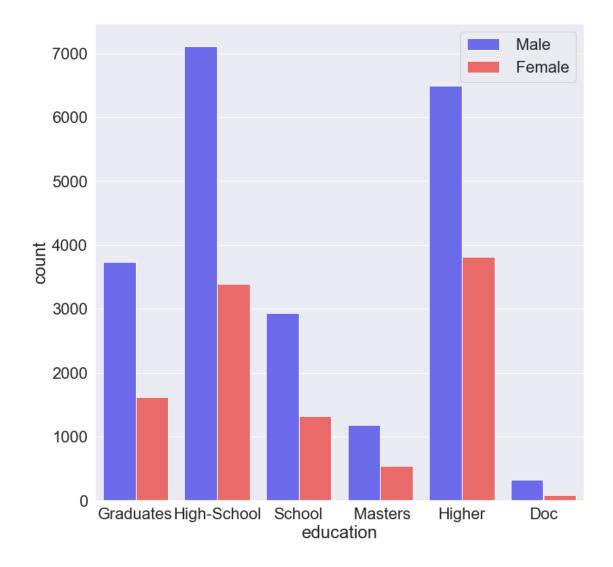
People who are from Higher educational background are earning more people who are from high school educational background are earning less

12 Which gender is more educated

```
[64]: plt.figure(figsize=(12,12))
    sns.set(font_scale=2)
    sns.countplot(df['education'],hue='sex',data=df,palette='seismic')
    plt.legend(loc='upper right', bbox_to_anchor=(1, 1.0));
```

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Pass the following variable as a keyword arg: x. From version 0.12, the only valid positional argument will be `data`, and passing other arguments without an explicit keyword will result in an error or misinterpretation.



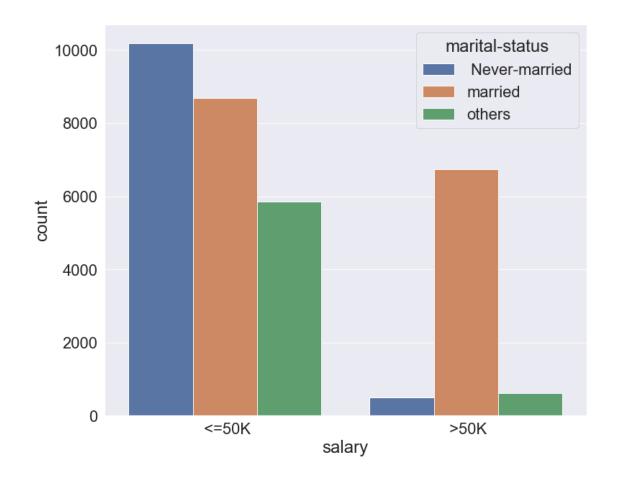
Males are more educated than Females

13 Which marital status people are earning more

```
[65]: #Salary based on martial status
plt.figure(figsize=(12,10))
sns.countplot('salary',data=df,hue='marital-status')
plt.show()
```

C:\Users\hp\AppData\Roaming\Python\Python38\sitepackages\seaborn_decorators.py:36: FutureWarning:

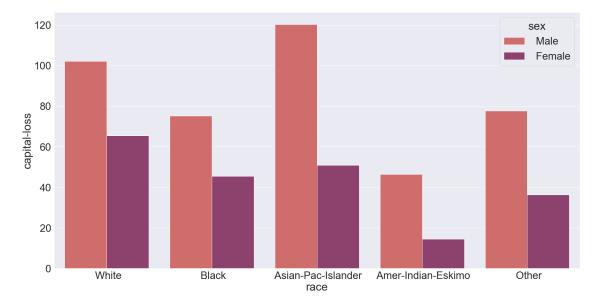
Pass the following variable as a keyword arg: x. From version 0.12, the only valid positional argument will be `data`, and passing other arguments without an explicit keyword will result in an error or misinterpretation.



Married people are earning more

14 Which kind of people have maximum capital loss

```
[67]: plt.figure(figsize=(30,15))
    sns.set(font_scale=3)
    sns.barplot(x='race',y='capital-loss',data=df,hue='sex',palette='flare',ci=None)
    plt.show()
```



Asian-Pac-islander people have more capital loss than others

15 Thank You