TUGAS UAS SAINS DATA

NAMA: DWI AHMAD DZULHIJJAH

EMAIL: 1818101@scholar.itn.ac.id (mailto:1818101@scholar.itn.ac.id)

Kampus asal: ITN Malang

Kampus ICEI: Universitas Diponegoro

Matkul: Sains Data (Data Science)

BUSINESS UNDESTANDING

MEMPERSIAPKAN LIBRARY

```
In [16]:
          1 # data analysis and wrangling
          2 import numpy as np
           3 import pandas as pd
          4 from scipy import stats
          5 from scipy.stats import norm, skew
          6 # visualization
          7 import seaborn as sns
          8 import matplotlib. pyplot as plt
          9 import plotly.figure_factory as ff
          10 import plotly.graph_objects as go
          11 # machine Learning
          12 from sklearn.preprocessing import LabelEncoder
          13 from sklearn.preprocessing import MinMaxScaler
          14  from sklearn.model_selection import train_test_split
          15 from sklearn.linear_model import LogisticRegression
          16 from sklearn.svm import SVC
          17 from sklearn.ensemble import RandomForestClassifier
          18 from sklearn.neighbors import KNeighborsClassifier
          19 from sklearn.naive_bayes import GaussianNB
          20 from sklearn.linear_model import Perceptron
          21 from sklearn.linear_model import SGDClassifier
          22 from sklearn.tree import DecisionTreeClassifier
          23 from sklearn.metrics import accuracy_score,classification_report
          24 | from xgboost import XGBClassifier
          25 | from sklearn.feature_selection import SelectKBest
          26 from sklearn.feature_selection import chi2
          27 from sklearn.metrics import roc_curve
          28 from sklearn.metrics import roc auc score
          29 | from sklearn.feature_selection import f_regression, mutual_info_regression
          30 from xgboost import XGBRegressor
          31 from xgboost import plot_importance
```

DATA (DATA WRANGLING)

MEMBACA DATA

```
In [4]: 1 Data = pd.read_csv("Placement_Data_Full_Class.csv")
In [99]: 1 Data.head(5)
Out[99]:
              sl_no gender ssc_p hsc_p degree_p workex etest_p specialisation mba_p status
                            67.00
                                  91.00
                                                            55.0
                                                                                58.80
                                                                                          1
                            79.33
                                  78.33
                                            77.48
                                                            86.5
                                                                           0
                                                                                66.28
                                                                                          1
                            65.00
                                  68.00
                                            64.00
                                                            75.0
                                                                                57.80
                                                       0
                                                                                          1
                 4
                            56.00
                                  52.00
                                            52.00
                                                      0
                                                            66.0
                                                                                59.43
                                                                                          0
                 5
                         1 85.80 73.60
                                            73.30
                                                      0
                                                            96.8
                                                                               55.50
                                                                                          1
In [20]: 1 Data.shape
Out[20]: (215, 15)
```

Telaah TIPE DATA

```
In [6]: 1 Data.info()
         <class 'pandas.core.frame.DataFrame'>
          RangeIndex: 215 entries, 0 to 214
         Data columns (total 15 columns):
                              Non-Null Count Dtype
          #
              Column
          0
              sl_no
                              215 non-null
                                               int64
          1
              gender
                              215 non-null
                                               object
                               215 non-null
              ssc_p
                                               float64
          3
                              215 non-null
                                               object
              ssc b
          4
                              215 non-null
                                               float64
              hsc p
          5
                              215 non-null
              hsc_b
                                               object
          6
              hsc_s
                              215 non-null
                                               object
              degree_p
                              215 non-null
                                               float64
           8
                              215 non-null
              degree_t
                                               object
              workex
                              215 non-null
                                               object
          10
              etest_p
                              215 non-null
                                               float64
          11 specialisation
                              215 non-null
                                               object
          12 mba_p
                              215 non-null
                                               float64
                              215 non-null
                                               object
          13
              status
          14 salary
                              148 non-null
                                               float64
         dtypes: float64(6), int64(1), object(8)
         memory usage: 25.3+ KB
         TELAAH JENIS DATA
In [10]:
           1 s = (Data.dtypes == 'object')
              object_cols = list(s[s].index)
             print("Categorical variables:")
           4 print(object_cols)
         Categorical variables:
         ['gender', 'ssc_b', 'hsc_b', 'hsc_s', 'degree_t', 'workex', 'specialisation', 'status']
In [11]:
          1 x = (Data.dtypes == ('int64'))
           2 integer_cols = list(x[x].index)
           3 print("integer variables:")
           4 print(integer_cols)
         integer variables:
         ['sl_no']
In [12]:
           1 x = (Data.dtypes == ('float64'))
           2 float_cols = list(x[x].index)
           3 print("float variables:")
           4 print(float_cols)
         float variables:
         ['ssc_p', 'hsc_p', 'degree_p', 'etest_p', 'mba_p', 'salary']
           1 print("numerical variables:")
In [13]:
           2 numerical_cols=float_cols+integer_cols
           3 print(numerical_cols)
         numerical variables:
         ['ssc_p', 'hsc_p', 'degree_p', 'etest_p', 'mba_p', 'salary', 'sl_no']
In [24]: 1 Data.columns
dtype='object')
In [18]:
           1 Data.describe()
Out[18]:
                                                           etest_p
                              ssc_p
                                                                     mba_p
                    sl no
                                        hsc_p
                                               degree_p
                                                                                  salary
          count 215.000000
                          215.000000
                                    215.000000
                                              215.000000
                                                        215.000000
                                                                  215.000000
                                                                               148.000000
                                               66.370186
           mean
                108.000000
                           67.303395
                                     66.333163
                                                         72.100558
                                                                   62.278186 288655.405405
                 62.209324
                           10.827205
                                                7.358743
                                                                   5.833385
            std
                                     10.897509
                                                         13.275956
                                                                            93457.452420
                 1.000000
                           40.890000
                                     37.000000
                                               50.000000
                                                         50.000000
                                                                   51.210000 200000.000000
            min
           25%
                 54.500000
                           60.600000
                                     60.900000
                                               61.000000
                                                         60.000000
                                                                   57.945000 240000.000000
                108.000000
                           67.000000
                                     65.000000
                                               66.000000
                                                         71.000000
                                                                   62.000000
                                                                            265000.000000
            50%
           75%
                161.500000
                           75.700000
                                     73.000000
                                               72.000000
                                                         83.500000
                                                                   66.255000
                                                                           300000.000000
```

89.400000

97.700000

91.000000

98.000000

77.890000 940000.000000

max 215.000000

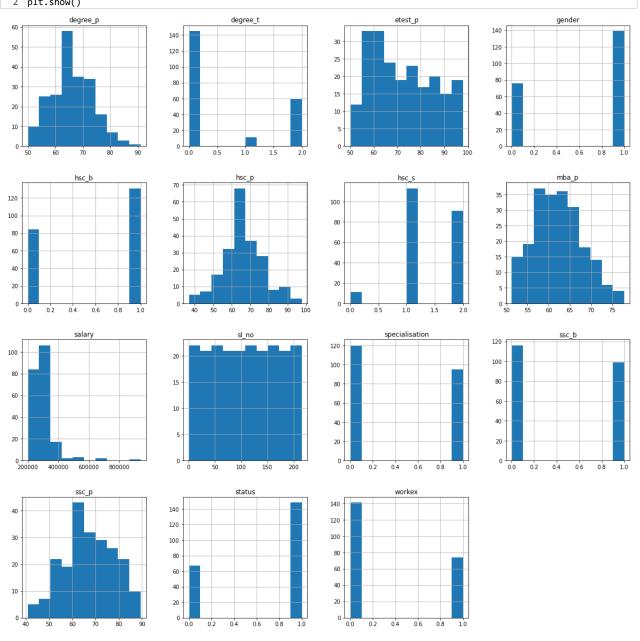
```
In [15]: 1 Data.isnull().sum()
Out[15]: sl_no
      gender
      ssc_p
                  0
      ssc_b
                  0
      hsc_p
                  0
      hsc b
      hsc s
      degree_p
      degree_t
      workex
      etest_p
                 0
      specialisation
      mba p
      status
                 a
      salary
      dtype: int64
In [26]: 1 # getting the object columns
       2 object_columns = Data.select_dtypes(include=['object']).columns
       4 # iterating over each object type column
       5 for col in object_columns:
           print('-' * 40 + col + '-' * 40 , end='-')
           display(Data[col].value_counts())
      -----gender-----
      М
         139
      Name: gender, dtype: int64
      -----ssc_b------
      Central
            116
      Others
             99
      Name: ssc_b, dtype: int64
      -----hsc_b------hsc_b------
      Central
      Name: hsc_b, dtype: int64
      Commerce
             113
      Science 91
              11
      Name: hsc_s, dtype: int64
      -----degree_t------
              145
      Comm&Mgmt
      Sci&Tech
              59
      Others
              11
      Name: degree_t, dtype: int64
      ------workex------
      No
          141
      Yes
          74
      Name: workex, dtype: int64
                  -----specialisation-----
      Mkt&Fin
            120
      Name: specialisation, dtype: int64
      -----status-----status-----
               148
      Placed
      Not Placed
               67
      Name: status, dtype: int64
      Data Preprocessing
      Mengatasi Data Null
```

```
In [36]: 1 Data['salary'].fillna((Data['salary'].mean()), inplace=True)
```

```
In [37]: 1 Data.isnull().sum()
Out[37]: sl_no
           gender
                                 0
           ssc_p
                                 0
           ssc b
                                 0
           hsc_p
           hsc_b
                                 0
           hsc_s
                                 0
                                 0
           degree_p
           degree_t
                                 0
           workex
                                 0
           etest_p
                                 0
           specialisation
                                 a
           mba_p
                                 0
           status
                                 0
                                 0
           salary
           dtype: int64
           Label Encoding
In [41]:
                label = LabelEncoder()
             2 Data["gender"] = label.fit_transform(Data["gender"])
                Data["ssc_b"] = label.fit_transform(Data["ssc_b"])
Data["hsc_b"] = label.fit_transform(Data["hsc_b"])
                Data["hsc_s"] = label.fit_transform(Data["hsc_s"])
             Data["degree_t"] = label.fit_transform(Data["degree_t"])
Data["workex"] = label.fit_transform(Data["workex"])
             8 Data["specialisation"] = label.fit_transform(Data["specialisation"])
             9 Data["status"] = label.fit_transform(Data["status"])
In [42]:
            1 Data.head(10)
Out[42]:
               sl_no gender ssc_p ssc_b
                                            hsc_p hsc_b
                                                           hsc s
                                                                   degree_p degree_t workex etest_p specialisation mba_p
                                                                                                                              status
                                                                                                                                              salarv
            0
                               67.00
                                              91.00
                                                                                    2
                                                                                            0
                                                                                                  55.00
                                                                                                                        58.80
                                                                                                                                      270000.000000
                                                         1
                                                                       58.00
                                                                                                                   1
                                          1
            1
                   2
                               79.33
                                          0
                                              78.33
                                                         1
                                                                2
                                                                       77.48
                                                                                    2
                                                                                                  86.50
                                                                                                                   0
                                                                                                                        66.28
                                                                                                                                   1 200000.000000
                   3
                               65.00
                                              68.00
                                                        0
                                                                0
                                                                       64.00
                                                                                    0
                                                                                                  75.00
                                                                                                                   0
                                                                                                                        57.80
                                                                                                                                      250000.000000
                                          0
            3
                               56.00
                                              52.00
                                                         0
                                                                2
                                                                       52.00
                                                                                    2
                                                                                            0
                                                                                                  66.00
                                                                                                                        59.43
                                                                                                                                      288655.405405
                               85.80
                                          0
                                              73.60
                                                         0
                                                                       73.30
                                                                                    0
                                                                                            0
                                                                                                  96.80
                                                                                                                   0
                                                                                                                        55.50
                                                                                                                                      425000.000000
                   6
                               55.00
                                              49.80
                                                         1
                                                                2
                                                                       67.25
                                                                                    2
                                                                                             1
                                                                                                  55.00
                                                                                                                   0
                                                                                                                        51.58
                                                                                                                                   0 288655.405405
                   7
                           0
                               46 00
                                                         1
                                                                       79.00
                                                                                    Ω
                                                                                            0
                                                                                                  74 28
                                                                                                                   0
                                                                                                                        53 29
                                                                                                                                   O
                                                                                                                                      288655.405405
                                              49 20
                                                        0
                                                                2
                                                                                    2
                                                                                                                   0
                   8
                               82.00
                                          0
                                              64.00
                                                                       66.00
                                                                                            1
                                                                                                  67.00
                                                                                                                        62.14
                                                                                                                                   1 252000.000000
                   9
                                              79.00
                                                         0
                                                                       72.00
                                                                                    0
                                                                                            0
                                                                                                  91.34
                                                                                                                   0
                                                                                                                        61.29
                                                                                                                                   1 231000.000000
                               73.00
                                          0
                                                                1
                  10
                               58.00
                                              70.00
                                                         0
                                                                       61.00
                                                                                    0
                                                                                                  54.00
                                                                                                                        52.21
                                                                                                                                     288655.405405
```

Eksplorasi Data (EDA)

In [43]: 1 Data.hist(figsize = (20, 20))
2 plt.show()



```
In [96]: 1 cor=Data.corr()
2 plt.figure(figsize=(14,6))
3 sns.heatmap(cor,annot=True)
```

Out[96]: <matplotlib.axes._subplots.AxesSubplot at 0x1718283adc0>



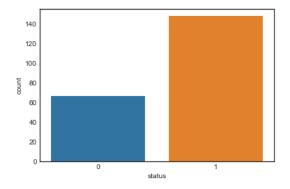
Analisis Pengalaman Kerja dan Status Penempatan

```
In [47]: 1 sns.countplot('status', data=Data)
```

C:\Users\dwiah\anaconda3\lib\site-packages\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.

warnings.warn(

Out[47]: <matplotlib.axes._subplots.AxesSubplot at 0x17182cce2e0>



```
In [49]: 1 Data['gender'].value_counts()
Out[49]: 1 139
```

Out[49]: 1 139 0 76 Name: gender, dtype: int64

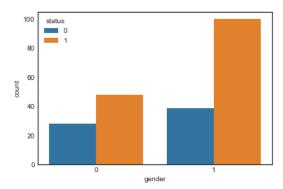
In [52]: 1 df = pd.DataFrame(Data.groupby(['gender', 'status'])['status'].count())

Out[52]:

		status
gender	status	
0	0	28
	1	48
1	0	39
	1	100

```
In [54]: 1 sns.countplot(x='gender', hue='status', data=Data)
```

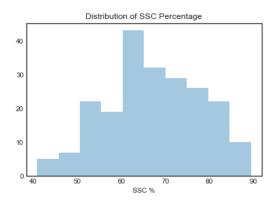
Out[54]: <matplotlib.axes._subplots.AxesSubplot at 0x171828e26d0>



C:\Users\dwiah\anaconda3\lib\site-packages\seaborn\distributions.py:2619: FutureWarning: `distplot` is a deprecated function and will be removed in a future version. Please adapt your code to use either `displot` (a figure-level function with similar flexibility) or `histplot` (an axes-level function for histograms).

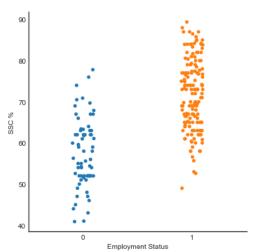
warnings.warn(msg, FutureWarning)

Out[56]: Text(0.5, 0, 'SSC %')



```
In [58]: 1 sns.catplot(y='ssc_p', x='status', data=Data)
2 plt.xlabel('Employment Status')
3 plt.ylabel('SSC %')
```

Out[58]: Text(9.54999999999997, 0.5, 'SSC %')

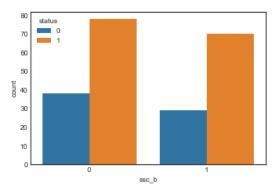


Out[60]:

		status
ssc_b	status	
0	0	38
	1	78
1	0	29
	1	70

```
In [62]: 1 sns.countplot(x='ssc_b', hue='status', data=Data)
```

Out[62]: <matplotlib.axes._subplots.AxesSubplot at 0x171808a1250>

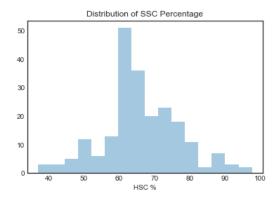


```
In [63]: 1 sns.distplot(Data['hsc_p'], kde=False)
2 plt.title('Distribution of SSC Percentage')
3 plt.xlabel('HSC %')
```

C:\Users\dwiah\anaconda3\lib\site-packages\seaborn\distributions.py:2619: FutureWarning: `distplot` is a deprecated function and will be removed in a future version. Please adapt your code to use either `displot` (a figure-level function with similar flexibility) or `histplot` (an axes-level function for histograms).

warnings.warn(msg, FutureWarning)

Out[63]: Text(0.5, 0, 'HSC %')



```
1 sns.catplot(y='hsc_p', x='status', data=Data)
In [65]:
             plt.xlabel('Employment Status')
plt.ylabel('HSC %')
Out[65]: Text(3.9249999999997, 0.5, 'HSC %')
               100
               90
               80
            HSC %
               70
               60
               50
               40
                                  Employment Status
```

```
In [66]: 1 Data['hsc_b'].value_counts()
Out[66]: 1
              131
         Name: hsc_b, dtype: int64
In [67]:
          1 df = pd.DataFrame(Data.groupby(['hsc_b', 'status'])['status'].count())
```

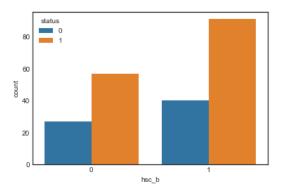
Out[67]:

hsc_b	status	
0	0	27
	1	57
1	0	40
	1	91

status

```
In [68]: 1 sns.countplot(x='hsc_b', hue='status', data=Data)
```

Out[68]: <matplotlib.axes._subplots.AxesSubplot at 0x171822753d0>



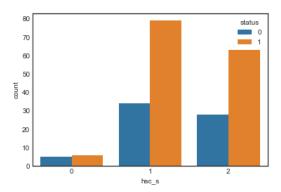
Out[70]:

		status
hsc_s	status	
0	0	5
	1	6
1	0	34
	1	79
2	0	28

```
In [71]: 1 sns.countplot(x='hsc_s', hue='status', data=Data)
```

63

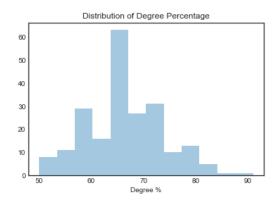
Out[71]: <matplotlib.axes._subplots.AxesSubplot at 0x17180876e50>



C:\Users\dwiah\anaconda3\lib\site-packages\seaborn\distributions.py:2619: FutureWarning: `distplot` is a deprecated function and will be removed in a future version. Please adapt your code to use either `displot` (a figure-level function with similar flexibility) or `histplot` (an axes-level function for histograms).

warnings.warn(msg, FutureWarning)

Out[72]: Text(0.5, 0, 'Degree %')



Employment Status

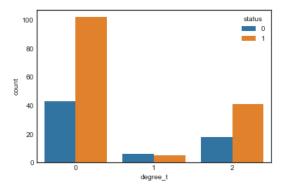
Out[75]:

degree_t	status	
0	0	43
	1	102
1	0	6
	1	5
2	0	18
	1	41

status

```
In [76]: 1 sns.countplot(x='degree_t', hue='status', data=Data)
```

Out[76]: <matplotlib.axes._subplots.AxesSubplot at 0x1718089cc70>



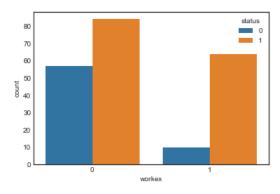
Out[78]:

Status		
	status	workex
57	0	0
84	1	
10	0	1
64	1	

etatue

```
In [79]: 1 sns.countplot(x='workex', hue='status', data=Data)
```

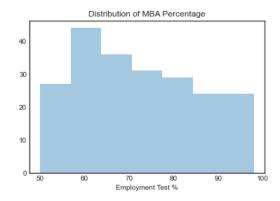
Out[79]: <matplotlib.axes._subplots.AxesSubplot at 0x17180765b80>



C:\Users\dwiah\anaconda3\lib\site-packages\seaborn\distributions.py:2619: FutureWarning: `distplot` is a deprecated function and will be removed in a future version. Please adapt your code to use either `displot` (a figure-level function with similar flexibility) or `histplot` (an axes-level function for histograms).

warnings.warn(msg, FutureWarning)

Out[80]: Text(0.5, 0, 'Employment Test %')



```
In [81]:

1 sns.catplot(y='etest_p', x='status', data=Data)
2 plt.xlabel('Employment Status')
3 plt.ylabel('Employment Test %')

Out[81]: Text(3.92499999999997, 0.5, 'Employment Test %')

90

80

90

60

Employment Status

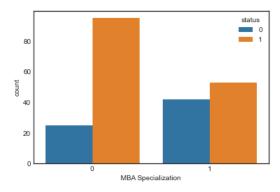
1
```

Out[83]:

Status		
	status	specialisation
25	0	0
95	1	
42	0	1
53	1	

```
In [84]: 1 sns.countplot(x='specialisation', hue='status', data=Data)
2 plt.xlabel('MBA Specialization')
```

Out[84]: Text(0.5, 0, 'MBA Specialization')

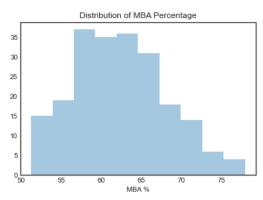


```
In [85]: 1 sns.distplot(Data['mba_p'], kde=False)
2 plt.title('Distribution of MBA Percentage')
3 plt.xlabel('MBA %')
```

C:\Users\dwiah\anaconda3\lib\site-packages\seaborn\distributions.py:2619: FutureWarning: `distplot` is a deprecated function and will be removed in a future version. Please adapt your code to use either `displot` (a figure-level function with similar flexibility) or `histplot` (an axes-level function for histograms).

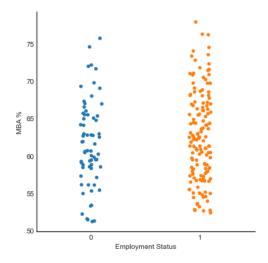
warnings.warn(msg, FutureWarning)

Out[85]: Text(0.5, 0, 'MBA %')



```
In [86]: 1 sns.catplot(y='mba_p', x='status', data=Data)
2 plt.xlabel('Employment Status')
3 plt.ylabel('MBA %')
```

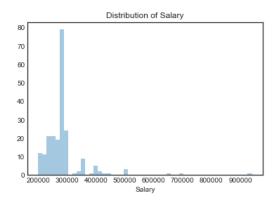
Out[86]: Text(9.54999999999997, 0.5, 'MBA %')



C:\Users\dwiah\anaconda3\lib\site-packages\seaborn\distributions.py:2619: FutureWarning: `distplot` is a deprecated function and will be removed in a future version. Please adapt your code to use either `displot` (a figure-level function with similar flexibility) or `histplot` (an axes-level function for histograms).

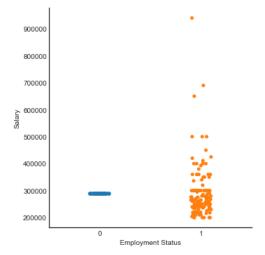
warnings.warn(msg, FutureWarning)

Out[87]: Text(0.5, 0, 'Salary')



```
In [88]: 1 sns.catplot(y='salary', x='status', data=Data)
2 plt.xlabel('Employment Status')
3 plt.ylabel('Salary')
```

Out[88]: Text(-12.950000000000003, 0.5, 'Salary')



In [89]: sns.pairplot(data=Data[['ssc_p','hsc_p','degree_p', 'etest_p','mba_p','salary', 'status']], hue="status", diag_ki Out[89]: <seaborn.axisgrid.PairGrid at 0x17181a5c940> hsc_p

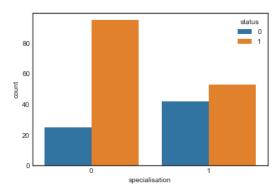
8 600000 600000

```
In [92]: 1 sns.countplot("specialisation", hue="status", data=Data)
```

C:\Users\dwiah\anaconda3\lib\site-packages\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.

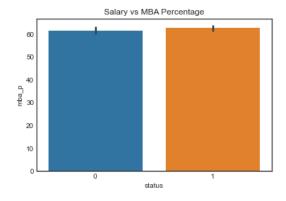
warnings.warn(

Out[92]: <matplotlib.axes._subplots.AxesSubplot at 0x171828371f0>



```
In [95]: 1 sns.barplot(x="status", y="mba_p", data=Data)
2 plt.title("Salary vs MBA Percentage")
```

Out[95]: Text(0.5, 1.0, 'Salary vs MBA Percentage')



```
In [101]: 1 Data.dtypes
Out[101]: sl_no
                              int64
          gender
                              int32
          ssc_p
                            float64
          hsc_p
                            float64
          degree p
                            float64
          workex
                              int32
          etest_p
                            float64
          specialisation
                              int32
          mba_p
                            float64
          status
                              int32
          dtype: object
In [102]:
           1 data_clf = Data.copy()
            2 data_reg = Data.copy()
```

Data Modelling

DECISSION TREE CLASSIFIER

```
In [103]:

1 from sklearn.tree import DecisionTreeClassifier
2 from sklearn.ensemble import RandomForestClassifier
3 from sklearn.model_selection import train_test_split
4 from sklearn.metrics import accuracy_score, classification_report
```

```
In [105]:
            1  # Seperating Features and Target
            2 X = data_clf[['gender', 'ssc_p', 'hsc_p', 'degree_p', 'workex', 'etest_p', 'specialisation', 'mba_p',]]
            3 y = data_clf['status']
In [106]: 1 X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.3)
           1 dtree = DecisionTreeClassifier(criterion='entropy')
In [107]:
            2 dtree.fit(X_train, y_train)
            3 y_pred = dtree.predict(X_test)
In [108]: | 1 | accuracy_score(y_test, y_pred)
Out[108]: 0.8307692307692308
In [109]: 1 print(classification_report(y_test, y_pred))
                        precision
                                     recall f1-score
                                                        support
                     0
                             0.88
                                       0.62
                                                 0.73
                                                             24
                             0.81
                                       0.95
                                                 0.88
                                                             41
                                                 0.83
                                                             65
              accuracy
                             0.85
                                       0.79
             macro avg
                                                 0.80
                                                             65
          weighted avg
                             0.84
                                       0.83
                                                 0.82
                                                             65
```

RANDOM FOREST ALGORITHM

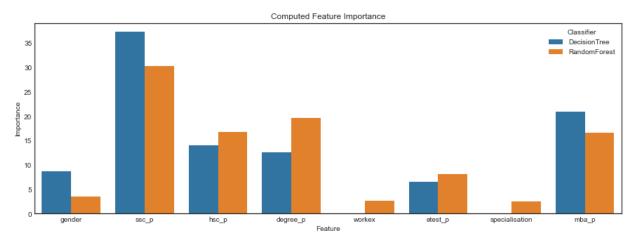
```
In [114]:
           1 #Using Random Forest Algorithm
            2 random forest = RandomForestClassifier(n estimators=100)
            3 random_forest.fit(X_train, y_train)
            4 y_pred = random_forest.predict(X_test)
In [115]: 1 | accuracy_score(y_test, y_pred)
Out[115]: 0.8615384615384616
In [116]: 1 print(classification_report(y_test, y_pred))
                                     recall f1-score
                        precision
                                                        support
                     0
                             1.00
                                       0.62
                                                 0.77
                                                             24
                             0.82
                                       1.00
                                                 0.90
                                                             41
                                                 0.86
                                                             65
              accuracy
                                                 0.84
                             0.91
                                       0.81
             macro avg
                                                             65
          weighted avg
                             0.89
                                       0.86
                                                 0.85
                                                             65
```

Feature Selection untuk Regresi Linear

```
In [117]: 1
    rows = list(X.columns)
    imp = pd.DataFrame(np.zeros(6*len(rows)).reshape(2*len(rows), 3))
    imp.columns = ["Classifier", "Feature", "Importance"]
    #Add Rows
    for index in range(0, 2*len(rows), 2):
        imp.iloc[index] = ["DecisionTree", rows[index//2], (100*dtree.feature_importances_[index//2])]
        imp.iloc[index + 1] = ["RandomForest", rows[index//2], (100*random_forest.feature_importances_[index//2])]
```

C:\Users\dwiah\anaconda3\lib\site-packages\seaborn_decorators.py:36: FutureWarning: Pass the following variables as keyword args: x, y. From version 0.12, the only valid positional argument will be `data`, and passing other argument s without an explicit keyword will result in an error or misinterpretation.

warnings.warn(



In [134]: 1 X.head(10)

Out[134]:

	ssc_p	hsc_p	degree_p	etest_p	mba_p	gender_F	gender_M	workex_No	workex_Yes	specialisation_Mkt&Fin	specialisation_Mkt&HR
(67.00	91.00	58.00	55.00	58.80	1	0	1	0	1	0
	79.33	78.33	77.48	86.50	66.28	1	0	0	1	0	1
:	65.00	68.00	64.00	75.00	57.80	1	0	1	0	0	1
;	56.00	52.00	52.00	66.00	59.43	1	0	1	0	1	0
4	85.80	73.60	73.30	96.80	55.50	1	0	1	0	0	1
,	55.00	49.80	67.25	55.00	51.58	1	0	0	1	0	1
(46.00	49.20	79.00	74.28	53.29	0	1	1	0	0	1
7	82.00	64.00	66.00	67.00	62.14	1	0	0	1	0	1
8	73.00	79.00	72.00	91.34	61.29	1	0	1	0	0	1
9	58.00	70.00	61.00	54.00	52.21	1	0	1	0	0	1

One-Hot Encoding

```
In [126]: 1  X = pd.get_dummies(X)
2  colmunn_names = X.columns.to_list()

In [127]: 1  from sklearn.preprocessing import MinMaxScaler
2  scaler = MinMaxScaler()
3  X_scaled = scaler.fit_transform(X)

In [128]: 1  #Train Test Split
2  X_train, X_test, y_train, y_test = train_test_split(X_scaled, y, test_size=0.3)
```

Logistic Regression

```
In [129]: 1 from sklearn.linear_model import LogisticRegression
            1 logistic_reg = LogisticRegression()
2 logistic_reg.fit(X_train, y_train)
In [130]:
            3 y_pred = logistic_reg.predict(X_test)
In [131]: 1 accuracy_score(y_test, y_pred)
Out[131]: 0.8615384615384616
In [133]: 1 print(classification_report(y_test, y_pred))
                         precision
                                      recall f1-score
                                                           support
                                         0.53
                                                    0.69
                                                                19
                                                   0.91
                                                                46
                                                    0.86
                                                                65
               accuracy
                               0.92
                                         0.76
                                                    0.80
                                                                65
              macro avg
                                                    0.85
           weighted avg
                               0.88
                                         0.86
                                                                65
```