

## ass2

February 13, 2024

```
[2]: import numpy as np
import pandas as pd
import seaborn as sns

df = pd.read_csv(r"C:\Users\abhib\Desktop\ml_project\Untitled_
↳Folder\Academic_Performance.csv")
```

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

```
[3]:
```

	STUDENT_ID	GENDER	PLACEMENT	HONOR_OPTED_OR_NOT	EDUCATION_TYPE	\
0	SB11201210000129	F	Yes	Yes	ACADEMIC	
1	SB11201210000137	F	Yes	Yes	ACADEMIC	
2	SB11201210005154	M	No	Yes	ACADEMIC	
3	SB11201210007504	F	Yes	Yes	ACADEMIC	
4	SB11201210007548	M	Yes	Yes	ACADEMIC	

	ACADEMIC_PROGRAM	COURSE 1 MARKS	COURSE 2 MARKS	COURSE 3 MARKS	\
0	INDUSTRIAL ENGINEERING	71.0	93.0	71.0	
1	INDUSTRIAL ENGINEERING	97.0	38.0	86.0	
2	ELECTRONIC ENGINEERING	17.0	1.0	18.0	
3	INDUSTRIAL ENGINEERING	65.0	35.0	76.0	
4	INDUSTRIAL ENGINEERING	94.0	94.0	98.0	

	COURSE 4 MARKS	COURSE 5 MARKS	PERCENTILE	OVEARLL_GRADE
0	93.0	79.0	91	FIRST CLASS
1	98.0	78.0	92	THIRD CLASS
2	43.0	22.0	7	DISTINCTION
3	80.0	48.0	67	FIRST CLASS
4	100.0	71.0	98	FIRST CLASS

```
[4]: df.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 12411 entries, 0 to 12410
Data columns (total 13 columns):
#   Column                Non-Null Count  Dtype
---  -
0   STUDENT_ID            12411 non-null  object
```

```

1  GENDER                12389 non-null object
2  PLACEMENT             12396 non-null object
3  HONOR_OPTED_OR_NOT    12397 non-null object
4  EDUCATION_TYPE        12396 non-null object
5  ACADEMIC_PROGRAM      12377 non-null object
6  COURSE 1 MARKS        12400 non-null float64
7  COURSE 2 MARKS        12403 non-null float64
8  COURSE 3 MARKS        12397 non-null float64
9  COURSE 4 MARKS        12397 non-null float64
10 COURSE 5 MARKS        12389 non-null float64
11 PERCENTILE            12411 non-null int64
12 OVEARLL_GRADE         12411 non-null object
dtypes: float64(5), int64(1), object(7)
memory usage: 1.2+ MB

```

```
[5]: missing_values=df.isnull().sum()
      print(missing_values)
```

```

STUDENT_ID      0
GENDER          22
PLACEMENT       15
HONOR_OPTED_OR_NOT 14
EDUCATION_TYPE  15
ACADEMIC_PROGRAM 34
COURSE 1 MARKS  11
COURSE 2 MARKS   8
COURSE 3 MARKS  14
COURSE 4 MARKS  14
COURSE 5 MARKS  22
PERCENTILE      0
OVEARLL_GRADE   0
dtype: int64

```

```
[6]: df.dropna(subset=['GENDER'], inplace=True)
      df.isnull().sum()
```

```
[6]: STUDENT_ID      0
      GENDER         0
      PLACEMENT     15
      HONOR_OPTED_OR_NOT 14
      EDUCATION_TYPE  14
      ACADEMIC_PROGRAM 28
      COURSE 1 MARKS  11
      COURSE 2 MARKS   8
      COURSE 3 MARKS  13
      COURSE 4 MARKS  12
      COURSE 5 MARKS  21

```

```
PERCENTILE          0
OVEARLL_GRADE       0
dtype: int64
```

```
[7]: df['COURSE 1 MARKS']=df['COURSE 1 MARKS'].replace(np.NaN,df['COURSE 1 MARKS'].
    ↪median())
df['COURSE 2 MARKS']=df['COURSE 2 MARKS'].replace(np.NaN,df['COURSE 2 MARKS'].
    ↪median())
df['COURSE 3 MARKS']=df['COURSE 3 MARKS'].replace(np.NaN,df['COURSE 3 MARKS'].
    ↪median())
df['COURSE 4 MARKS']=df['COURSE 4 MARKS'].replace(np.NaN,df['COURSE 4 MARKS'].
    ↪median())
df['COURSE 5 MARKS']=df['COURSE 5 MARKS'].replace(np.NaN,df['COURSE 5 MARKS'].
    ↪median())

df.isnull().sum()
```

```
[7]: STUDENT_ID          0
GENDER                 0
PLACEMENT             15
HONOR_OPTED_OR_NOT    14
EDUCATION_TYPE        14
ACADEMIC_PROGRAM      28
COURSE 1 MARKS        0
COURSE 2 MARKS        0
COURSE 3 MARKS        0
COURSE 4 MARKS        0
COURSE 5 MARKS        0
PERCENTILE            0
OVEARLL_GRADE         0
dtype: int64
```

```
[8]: from sklearn.impute import SimpleImputer
imputer = SimpleImputer(strategy='constant', fill_value='missing')
x = imputer.fit_transform(df[['ACADEMIC_PROGRAM']])
print(x)

pd.isnull(x).sum()
```

```
['INDUSTRIAL ENGINEERING']
['INDUSTRIAL ENGINEERING']
['ELECTRONIC ENGINEERING']
...
['INDUSTRIAL ENGINEERING']
['missing']
['INDUSTRIAL ENGINEERING']]
```

[8]: 0

```
[9]: imputer = SimpleImputer(strategy='most_frequent')
y = imputer.fit_transform(df[['EDUCATION_TYPE']])
print(y)

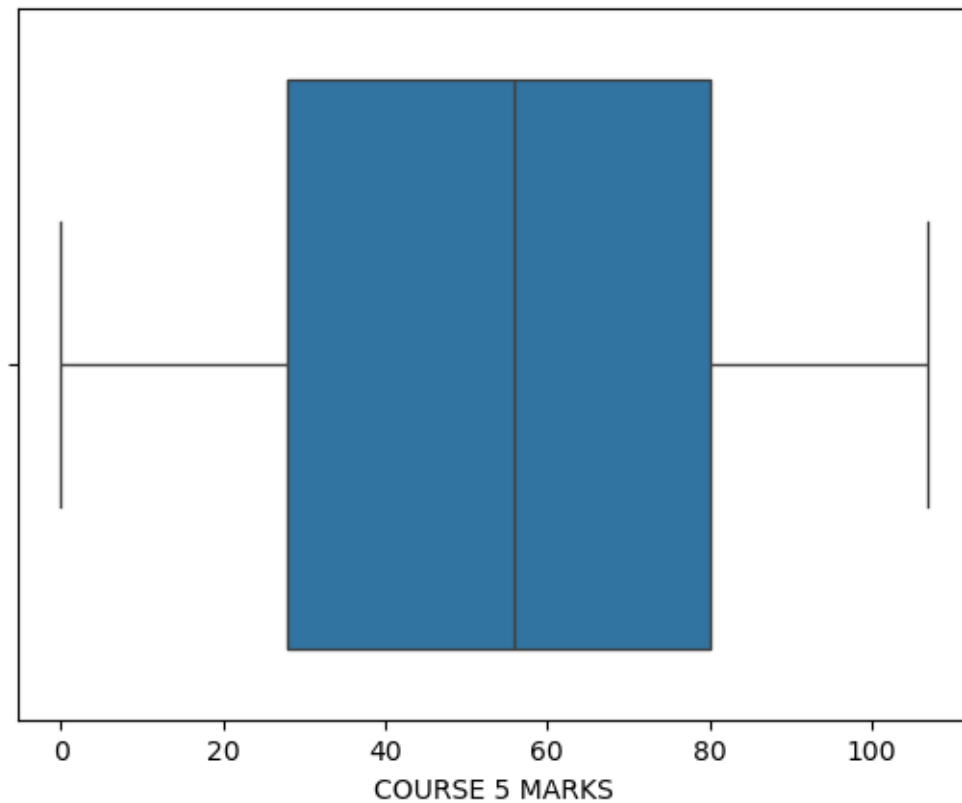
pd.isnull(y).sum()
```

```
['ACADEMIC']
['ACADEMIC']
['ACADEMIC']
...
['ACADEMIC']
['ACADEMIC']
['ACADEMIC']
```

[9]: 0

```
[10]: import seaborn as sns
import matplotlib.pyplot as plt
sns.boxplot(data=df, x=df['COURSE 5 MARKS'])

plt.show()
```



```
[11]: #Detecting Outliers with Z-scores
import numpy as np
outliers = []

def outliers_zscore(data):
    thres = 3
    mean = np.mean(data)
    std = np.std(data)
    for i in data:
        z_score = (i-mean)/std
        if (np.abs(z_score) > thres):
            outliers.append(i)
    return outliers

col_outliers = outliers_zscore(df['COURSE 1 MARKS'])
print("Outliers from Z-scores method: ", col_outliers)
```

```
Outliers from Z-scores method: [6.0, 3.0, 1.0, 5.0, 2.0, 8.0, 7.0, 2.0, 8.0,
6.0, 9.0, 2.0, 9.0, 8.0, 1.0, 9.0, 2.0, 2.0, 1.0, 6.0, 7.0, 4.0, 5.0, 9.0, 7.0,
9.0, 1.0, 2.0, 8.0, 5.0, 2.0, 8.0, 8.0, 1.0, 4.0, 7.0, 4.0, 7.0, 8.0, 3.0, 8.0,
5.0, 9.0, 7.0, 8.0, 7.0, 1.0, 9.0, 2.0, 7.0, 5.0, 3.0, 7.0, 3.0, 8.0, 6.0, 9.0,
8.0, 9.0, 6.0, 1.0, 7.0, 8.0, 1.0, 9.0, 1.0, 7.0, 8.0, 9.0, 6.0, 7.0, 7.0, 8.0,
4.0, 6.0, 6.0, 5.0, -1.0, 8.0, 8.0, 3.0, 1.0, 3.0, 3.0, 2.0, 9.0, 8.0, 3.0, 6.0,
3.0, 2.0, 7.0, 8.0, 4.0, 8.0, 3.0, 7.0, 9.0, 9.0, 3.0, 7.0, 6.0, 1.0, 1.0, 1.0,
-1.0, 9.0, 4.0, 8.0, 7.0, 1.0, 6.0]
```

```
[12]: #Detecting Outliers with IQR
outliers = []
def outliers_iqr(data):
    data = sorted(data)
    q1 = np.percentile(data, 25)
    q3 = np.percentile(data, 75)
    IQR = q3-q1
    lwr_bound = q1-(1.5*IQR)
    upr_bound = q3+(1.5*IQR)
    print(lwr_bound, upr_bound)
    for i in data:
        if (i<lwr_bound or i>upr_bound):
            outliers.append(i)
    return outliers

marks_outliers = outliers_iqr(df['COURSE 2 MARKS'])
print("Outliers from IQR method: ", marks_outliers)
```

```
-24.0 152.0
```

Outliers from IQR method: []

```
[13]: categorical = df.select_dtypes(exclude=[np.number])
categorical
```

```
[13]:
```

	STUDENT_ID	GENDER	PLACEMENT	HONOR_OPTED_OR_NOT	EDUCATION_TYPE	\
0	SB11201210000129	F	Yes	Yes	ACADEMIC	
1	SB11201210000137	F	Yes	Yes	ACADEMIC	
2	SB11201210005154	M	No	Yes	ACADEMIC	
3	SB11201210007504	F	Yes	Yes	ACADEMIC	
4	SB11201210007548	M	Yes	Yes	ACADEMIC	
...	...	...	...	...	...	
12406	SB11201420568705	M	Yes	Yes	ACADEMIC	
12407	SB11201420573045	M	Yes	Yes	ACADEMIC	
12408	SB11201420578809	M	Yes	No	ACADEMIC	
12409	SB11201420578812	F	Yes	Yes	ACADEMIC	
12410	SB11201420583232	M	No	No	ACADEMIC	

	ACADEMIC_PROGRAM	OVEARLL_GRADE
0	INDUSTRIAL ENGINEERING	FIRST CLASS
1	INDUSTRIAL ENGINEERING	THIRD CLASS
2	ELECTRONIC ENGINEERING	DISTINCTION
3	INDUSTRIAL ENGINEERING	FIRST CLASS
4	INDUSTRIAL ENGINEERING	FIRST CLASS
...	...	...
12406	MECHATRONICS ENGINEERING	FIRST CLASS
12407	INDUSTRIAL ENGINEERING	FIRST CLASS
12408	INDUSTRIAL ENGINEERING	FIRST CLASS
12409	NaN	FIRST CLASS
12410	INDUSTRIAL ENGINEERING	THIRD CLASS

[12389 rows x 7 columns]

```
[14]: categorical.PLACEMENT.value_counts()
```

```
[14]: PLACEMENT
Yes    9720
No     2654
Name: count, dtype: int64
```

```
[15]: categorical.PLACEMENT.replace({"Yes":1, "No":0}, inplace= True)
categorical.head()
```

```
[15]:
```

	STUDENT_ID	GENDER	PLACEMENT	HONOR_OPTED_OR_NOT	EDUCATION_TYPE	\
0	SB11201210000129	F	1.0	Yes	ACADEMIC	
1	SB11201210000137	F	1.0	Yes	ACADEMIC	
2	SB11201210005154	M	0.0	Yes	ACADEMIC	

3	SB11201210007504	F	1.0	Yes	ACADEMIC
4	SB11201210007548	M	1.0	Yes	ACADEMIC

	ACADEMIC_PROGRAM	OVEARLL_GRADE
0	INDUSTRIAL ENGINEERING	FIRST CLASS
1	INDUSTRIAL ENGINEERING	THIRD CLASS
2	ELECTRONIC ENGINEERING	DISTINCTION
3	INDUSTRIAL ENGINEERING	FIRST CLASS
4	INDUSTRIAL ENGINEERING	FIRST CLASS

```
[16]: #Label Encoding
categorical = categorical.drop('STUDENT_ID',axis=1)
categorical.head()
```

```
[16]:  GENDER  PLACEMENT  HONOR_OPTED_OR_NOT  EDUCATION_TYPE  ACADEMIC_PROGRAM \
0      F          1.0                Yes      ACADEMIC  INDUSTRIAL ENGINEERING
1      F          1.0                Yes      ACADEMIC  INDUSTRIAL ENGINEERING
2      M          0.0                Yes      ACADEMIC  ELECTRONIC ENGINEERING
3      F          1.0                Yes      ACADEMIC  INDUSTRIAL ENGINEERING
4      M          1.0                Yes      ACADEMIC  INDUSTRIAL ENGINEERING

OVEARLL_GRADE
0  FIRST CLASS
1  THIRD CLASS
2  DISTINCTION
3  FIRST CLASS
4  FIRST CLASS
```

```
[17]: column_category = categorical.select_dtypes(exclude=[np.number]).columns
column_category
```

```
[17]: Index(['GENDER', 'HONOR_OPTED_OR_NOT', 'EDUCATION_TYPE', 'ACADEMIC_PROGRAM',
        'OVEARLL_GRADE'],
        dtype='object')
```

```
[18]: from sklearn.preprocessing import LabelEncoder
label_encoder = LabelEncoder()
for i in column_category:
    categorical[i] = label_encoder.fit_transform(categorical[i])
print("Label Encoded Data: ")
categorical.head()
```

Label Encoded Data:

```
[18]:  GENDER  PLACEMENT  HONOR_OPTED_OR_NOT  EDUCATION_TYPE  ACADEMIC_PROGRAM \
0      0          1.0                1                0                13
1      0          1.0                1                0                13
```

2	1	0.0	1	0	10
3	0	1.0	1	0	13
4	1	1.0	1	0	13

OVEARLL_GRADE	
0	1
1	3
2	0
3	1
4	1

```
[19]: #One Hot Encoding
from sklearn.preprocessing import OneHotEncoder
onehot_encoder = OneHotEncoder(sparse_output=False)
onehot_encoded = onehot_encoder.fit_transform(categorical[column_category])

onehot_encoded
```

```
[19]: array([[1., 0., 0., ..., 1., 0., 0.],
        [1., 0., 0., ..., 0., 0., 1.],
        [0., 1., 0., ..., 0., 0., 0.],
        ...,
        [0., 1., 1., ..., 1., 0., 0.],
        [1., 0., 0., ..., 1., 0., 0.],
        [0., 1., 1., ..., 0., 0., 1.]])
```

```
[21]: onehot_encoded_frame = pd.DataFrame(onehot_encoded, columns = onehot_encoder.
    ↪get_feature_names_out(column_category))

onehot_encoded_frame.head()
```

```
[21]:  GENDER_0  GENDER_1  HONOR_OPTED_OR_NOT_0  HONOR_OPTED_OR_NOT_1  \
0         1.0        0.0                   0.0                   1.0
1         1.0        0.0                   0.0                   1.0
2         0.0        1.0                   0.0                   1.0
3         1.0        0.0                   0.0                   1.0
4         0.0        1.0                   0.0                   1.0

    HONOR_OPTED_OR_NOT_2  EDUCATION_TYPE_0  EDUCATION_TYPE_1  EDUCATION_TYPE_2  \
0                      0.0                1.0                0.0                0.0
1                      0.0                1.0                0.0                0.0
2                      0.0                1.0                0.0                0.0
3                      0.0                1.0                0.0                0.0
4                      0.0                1.0                0.0                0.0

    EDUCATION_TYPE_3  EDUCATION_TYPE_4  ...  ACADEMIC_PROGRAM_16  \
0                 0.0                0.0  ...                0.0
```



1	0.0	0.0	...	0.0
2	0.0	0.0	...	0.0
3	0.0	0.0	...	0.0
4	0.0	0.0	...	0.0

	ACADEMIC_PROGRAM_17	ACADEMIC_PROGRAM_18	ACADEMIC_PROGRAM_19	\
0	0.0	0.0	0.0	
1	0.0	0.0	0.0	
2	0.0	0.0	0.0	
3	0.0	0.0	0.0	
4	0.0	0.0	0.0	

	ACADEMIC_PROGRAM_20	ACADEMIC_PROGRAM_21	OVEARLL_GRADE_0	OVEARLL_GRADE_1	\
0	0.0	0.0	0.0	1.0	
1	0.0	0.0	0.0	0.0	
2	0.0	0.0	1.0	0.0	
3	0.0	0.0	0.0	1.0	
4	0.0	0.0	0.0	1.0	

	OVEARLL_GRADE_2	OVEARLL_GRADE_3
0	0.0	0.0
1	0.0	1.0
2	0.0	0.0
3	0.0	0.0
4	0.0	0.0

[5 rows x 36 columns]

[ ]: