1. Data Importing In [295 import pandas as pd import matplotlib.pylab as plt import seaborn as sns
In [296 df=pd.read_csv('placement.csv') 2. Data Inspecting In [297 df.head() Out[297]: StudentiD CGPA Internships Projects Workshops/Certifications AptitudeTestScore SoftSkillsRating ExtracurricularActivities PlacementTraining SSC_Marks HSC_Marks PlacementStatus 0 1 7.5 1 1 1 65 4.4 No No No 61 79 NotPlaced
1 2 8.9 0 3 2 90 4.0 Yes Yes 78 82 Placed 2 3 7.3 1 2 2 82 4.8 Yes No 79 80 NotPlaced 3 4 7.5 1 1 2 85 4.4 Yes Yes 81 80 Placed 4 5 8.3 1 2 2 86 4.5 Yes Yes 74 88 Placed
Out[298]: StudentID CGPA Internships Projects Workshops/Certifications AptitudeTestScore SoftSkillsRating ExtracurricularActivities PlacementTraining SSC_Marks HSC_Marks PlacementStatus 9995 9996 7.5 1 1 1 2 72 3.9 Yes No 85 66 NotPlaced 9996 9997 7.4 0 1 3 0 70 4.8 Yes Yes 79 81 Placed 9998 9999 8.9 0 3 2 87 4.8 Yes Yes 71 85 Placed
999 1000 8.4 0 1 1 66 3.8 No No No 62 66 NotPlaced In [299 df.info() <class 'pandas.core.frame.dataframe'=""> RangeIndex: 10000 entries, 0 to 9999 Data columns (total 12 columns):</class>
O StudentID 10000 non-null int64 CGPA 10000 non-null float64 Internships 10000 non-null int64 Projects 10000 non-null int64 Workshops/Certifications 10000 non-null int64 Workshops/Certifications 10000 non-null int64 AptitudeTestScore 10000 non-null int64 SoftSkillsRating 10000 non-null float64
7 ExtracurricularActivities 10000 non-null object 8 PlacementTraining 10000 non-null object 9 SSC_Marks 10000 non-null int64 10 HSC_Marks 10000 non-null int64 11 PlacementStatus 10000 non-null object dtypes: float64(2), int64(7), object(3) memory usage: 937.6+ KB
3. Data Preprocessing In [300 df.columns Out[300]: Index(['StudentID', 'CGPA', 'Internships', 'Projects', 'Workshops/Certifications', 'AptitudeTestScore', 'SoftSkillsRating', 'ExtracurricularActivities', 'PlacementTraining', 'SSC_Marks', 'HSC_Marks', 'PlacementStatus'], dtype='object')
3.1 Removing StudentID column: In [301 df=df.drop(columns='StudentID') df Out[301]: CGPA Internships Projects Workshops/Certifications AptitudeTestScore SoftSkillsRating ExtracurricularActivities PlacementTraining SSC_Marks HSC_Marks PlacementStatus
0 7.5 1 1 1 65 4.4 No No 61 79 NotPlaced 1 8.9 0 3 2 90 4.0 Yes Yes 78 82 Placed 2 7.3 1 2 2 82 4.8 Yes No 79 80 NotPlaced 3 7.5 1 1 2 85 4.4 Yes Yes 81 80 Placed 4 8.3 1 2 2 86 4.5 Yes Yes 74 88 Placed
9995 7.5 1 1 1 2 72 3.9 Yes No 85 66 NotPlaced 9996 7.4 0 1 0 90 4.8 No No 84 67 Placed 9997 8.4 1 3 0 70 4.8 Yes Yes 71 85 Placed 9998 8.9 0 3 2 87 4.8 Yes Yes 71 85 Placed 9999 8.4 0 1 66 3.8 No No 62 66 NotPlaced
10000 rows × 11 columns 3.2 Check for null values In [302 df.isnull().sum() Out[302]: CGPA
Internships 0 Projects 0 Workshops/Certifications 0 AptitudeTestScore 0 SoftSkillsRating 0 ExtracurricularActivities 0 PlacementTraining 0 SSC_Marks 0 HSC_Marks 0 PlacementStatus 0
atype: int64 3.3 Check for duplicate values In [303 df.duplicated().sum() Out[303]: 72
3.4 Check for outliers In [304 df.describe() Out[304]: CGPA Internships Projects Workshops/Certifications AptitudeTestScore SoftSkillsRating SSC_Marks HSC_Marks count 10000.000000 10000.000000 10000.000000 10000.000000 10000.000000 10000.000000 10000.000000
mean 7.698010 1.049200 2.026600 1.013200 79.449900 4.323960 69.159400 74.501500 std 0.640131 0.665901 0.867968 0.904272 8.159997 0.411622 10.430459 8.919527 min 6.500000 0.000000 0.000000 60.00000 3.000000 55.000000 57.000000 25% 7.40000 1.00000 0.00000 73.00000 4.40000 70.00000 73.00000 50% 7.70000 1.00000 2.00000 80.00000 4.40000 70.00000 73.00000 75% 8.20000 1.00000 2.00000 87.00000 4.70000 78.00000 83.00000
max 9.100000 3.000000 3.000000 90.000000 4.800000 90.000000 88.000000 4. Exploratory Data Analysis 4.1 Placed vs Unplaced
<pre>In [305 df['PlacementStatus'].value_counts() Out[305]:</pre>
Out[306]: Text(0.5, 1.0, 'Placement Distribution') Placement Distribution NotPlaced
Placed
4.2 Impact of cgpa over placement In [307 sns.countplot (data=df, x='CGPA') plt.xticks (rotation=90) plt.title('CGPA Analysis') Out[307]: Text(0.5, 1.0, 'CGPA Analysis') CGPA Analysis
800 - 700 - 600 -
500 - 9 400 - 300 - 200 -
In [310 sns.countplot(data=df.loc[(df.PlacementStatus=='Placed')], x='CGPA')
plt.xticks(rotation=90) plt.title('CGPA wise Placement') Out[310]: Text(0.5, 1.0, 'CGPA wise Placement') CGPA wise Placement
300 -
100 -
4.3 Impact of HSC_Marks over placement In [311 plt.figure(figsize=(5,5)) sns.displot(df['HSC_Marks'])
sns.displot(df['HSC_Marks']) C:\Users\DELL\anaconda3\Lib\site-packages\seaborn\axisgrid.py:118: UserWarning: The figure layout has changed to tight selffigure.tight_layout(*args, **kwargs) <pre></pre>
800 -
400 -
In [313 sns.countplot(data=df.loc[(df.PlacementStatus=='Placed')], x='HSC Marks')
In [313 sns.countplot(data=df.loc[(df.PlacementStatus=='Placed')],x='HSC_Marks') plt.xticks(rotation=90) plt.title('HSC Marks wise Placement') Out[313]: Text(0.5, 1.0, 'HSC Marks wise Placement') HSC Marks wise Placement 400
350 - 300 - 250 - tg 200 -
150 - 100 - 50 -
4.4 Impact of SSC_Marks over placement In [314 sns.displot(df['SSC_Marks'])
C:\Users\DELL\anaconda3\Lib\site-packages\seaborn\axisgrid.py:118: UserWarning: The figure layout has changed to tight selffigure.tight_layout(*args, **kwargs) Out[314]: 1750 -
1500 - 1250 - 1000 -
750 - 500 - 250 -
In [316 sns.countplot(data=df.loc[(df.PlacementStatus=='Placed')],x='SSC_Marks') plt.xticks(rotation=90) plt.title('SSC_Marks wise_Placement') Out[316]: Text(0.5, 1.0, 'SSC_Marks wise_Placement')
SSC Marks wise Placement
150 -
SSC_Warks 0 0 0 0 0 0 0 0 0 0 0 0 0
4.5 Impact of ExtracurricularActivities over placement In [320 sns.countplot(data=df.loc[(df.PlacementStatus=='Placed')],x='ExtracurricularActivities') plt.xticks(rotation=90) plt.title('ExtracurricularActivities wise Placement') Out[320]: Text(0.5, 1.0, 'ExtracurricularActivities wise Placement') ExtracurricularActivities wise Placement
3500 - 3000 - 2500 -
1500 - 1000 - 500 -
Extracurricular Activities 4.6 Impact of Softskills over placement
<pre>In [322 sns.countplot(data=df.loc[(df.PlacementStatus=='Placed')], x='SoftSkillsRating') plt.xticks(rotation=90) plt.title('Softskills wise Placement') Out[322]: Text(0.5, 1.0, 'Softskills wise Placement') Softskills wise Placement 1600 -</pre>
1400 - 1200 - 1000 - 1000 - 1000 -
600 - 400 - 200 -
4.7 Impact of Projects over placement In [324 sns.countplot (data=df.loc[(df.PlacementStatus=='Placed')], x='Projects')
plt.xticks(rotation=90) plt.title('Projects wise Placement') Out[324]: Text(0.5, 1.0, 'Projects wise Placement') Projects wise Placement 2500 -
2000 - = 1500 -
4.8 Impact of AptitudeTestScore over placements In [326 sns.countplot(data=df.loc[(df.PlacementStatus=='Placed')], x='AptitudeTestScore') plt.xticks(rotation=90) plt.title('AptitudeTestScore wise Placement')
AptitudeTestScore wise Placement 1400 - 1200 -
1000 - ## 800 - 600 -
400 - 200 - 0 - 0 - 0 - 0 - 0 - 0 - 0 - 0
5. Data Encoding 5.1 Label_Encoding In [327 from sklearn.preprocessing import LabelEncoder labelencoder=LabelEncoder() object_cols = df.select_dtypes(include=['object']).columns from object_cols = df.select_dtypes(include=['object']).columns
Sobject_cols = df.select_dtypes(include=['object']).columns for column in object_cols: df[column]=labelencoder.fit_transform(df[column]) In [328 df.head() Out[328]: CGPA Internships Projects Workshops/Certifications AptitudeTestScore SoftSkillsRating ExtracurricularActivities PlacementTraining SSC_Marks HSC_Marks PlacementStatus O 7.5
2 7.3 1 2 2 82 4.8 1 0 79 80 0 3 7.5 1 1 2 85 4.4 1 1 81 80 1 4 8.3 1 2 2 86 4.5 1 1 74 88 1 5.2 Correlation Matrix
In [329 correlation_matrix=df.corr() plt.figure(figsize=(12,6)) sns.heatmap(correlation_matrix,annot=True,cmap='coolwarm') Out[329]: Axxes: > CGPA - 1
Internships - 0.28
SoftSkillsRating - 0.38
HSC_Marks - 0.46
PlacementTrainin Workshops/Certification Workshops/Certification PlacementTrainin PlacementTrainin PlacementStatu Placem
<pre>In [330 X=df.drop(columns='PlacementStatus') y=df.PlacementStatus In [331 from sklearn.model_selection import train_test_split In [332 X_train,X_test,y_train,y_test=train_test_split(X,y,test_size=0.25,random_state=0) print('X_train_shape_is', X_train.shape) print('X_test_shape_is', X_test.shape)</pre>
print('X_test shape is ', X_test.shape) print('y_train shape is ', y_train.shape) print('y_test shape is ', y_test.shape) X_train shape is (7500, 10) X_test shape is (2500, 10) y_train shape is (7500,) y_test shape is (2500,) 7.Model Training and Evaluation
7.1 Logistic Regression > Accuracy Score> Classification Report> Confusion Matrix In [333 from sklearn.linear_model import LogisticRegression
In [334 from sklearn.metrics import accuracy_score, classification_report, confusion_matrix, ConfusionMatrixDisplay In [335 lr=LogisticRegression(max_iter=50000, penalty=None) lr.fit(X_train, y_train) prediction=lr.predict(X_test) scorel=accuracy_score(prediction, y_test) In [336 print('Accuracy Score is:', scorel)
Accuracy Score is: 0.7932 In [337 target_names=['NotPlaced', 'Placed'] In [338 print('Classification Report is: \n',classification_report(y_test,prediction,target_names=target_names)) Classification Report is: precision recall f1-score support NotPlaced 0.83 0.81 0.82 1471
Placed 0.74 0.77 0.75 1029 accuracy 0.79 2500 macro avg 0.79 0.79 0.79 2500 weighted avg 0.79 0.79 0.79 2500 In [339 cm=confusion_matrix(y_test, prediction) ConfusionMatrixDisplay(confusion_matrix=cm, display_labels=target_names).plot(cmap='Blues')
Out[339]: <sklearn.metricsplot.confusion_matrix.confusionmatrixdisplay 0x177d7f68550="" at=""></sklearn.metricsplot.confusion_matrix.confusionmatrixdisplay>
NotPlaced - 1190 281 - 1000
NotPlaced - 1190 281 - 1000 - 800 - 600
NotPlaced - 1190 281 - 1000 - 800
NotPlaced - 1190 281 - 1000 Placed - 236 793 - 600 NotPlaced Predicted label 7.2 Random Forest
NetPlaced 1190 281 1000 1000 1000 1000 1000 1000 1000
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