

Assnment_3_LR,DT,RFT

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The image displays two screenshots of a Google Colab notebook titled "Assnment 3 LR,DT,RFT.ipynb". The notebook is open to the "Code" tab, showing the initial data loading and preprocessing steps.

Top Screenshot: The code cell shows the import of necessary libraries (numpy, pandas, matplotlib, seaborn, sklearn) and the loading of the "Hr_Employee_Attrition.csv" dataset. The dataset is then displayed as a table with 5 rows and 35 columns.

| | Age | Attrition | BusinessTravel | DailyRate | Department | DistanceFromHome | Education | EducationField | EmployeeCount | EmployeeNumber | ... | RelationshipSatisfaction | StandardHours | StockOptionLevel | TotalWorkingYears |
|---|-----|-----------|-------------------|-----------|------------------------|------------------|-----------|----------------|---------------|----------------|-----|--------------------------|---------------|------------------|-------------------|
| 0 | 41 | Yes | Travel_Rarely | 1102 | Sales | 1 | 2 | Life Sciences | 1 | 1 | ... | 1 | 80 | 0 | |
| 1 | 49 | No | Travel_Frequently | 279 | Research & Development | 8 | 1 | Life Sciences | 1 | 2 | ... | 4 | 80 | 1 | |
| 2 | 37 | Yes | Travel_Rarely | 1373 | Research & Development | 2 | 2 | Other | 1 | 4 | ... | 2 | 80 | 0 | |
| 3 | 33 | No | Travel_Frequently | 1392 | Research & Development | 3 | 4 | Life Sciences | 1 | 5 | ... | 3 | 80 | 0 | |
| 4 | 27 | No | Travel_Rarely | 551 | Research & Development | 2 | 1 | Medical | 1 | 7 | ... | 4 | 80 | 1 | |

Bottom Screenshot: The code cell shows the calculation of the value counts for the "DailyRate" column and the display of the dataset's information.

```
df.DailyRate.value_counts()
```

| DailyRate | count |
|-----------|-------|
| 691 | 6 |
| 480 | 5 |
| 530 | 5 |
| 1329 | 5 |
| 1092 | 5 |
| ... | ... |
| 650 | 1 |
| 279 | 1 |
| 316 | 1 |
| 314 | 1 |
| 628 | 1 |

```
df.info()
```

<class 'pandas.core.frame.DataFrame'>
RangeIndex: 1470 entries, 0 to 1469
Data columns (total 35 columns):
Column Non-Null Count Dtype

0 Age 1470 non-null int64
1 Attrition 1470 non-null object
2 BusinessTravel 1470 non-null object
3 DailyRate 1470 non-null int64
4 Department 1470 non-null object
5 DistanceFromHome 1470 non-null int64
6 Education 1470 non-null int64
7 EducationField 1470 non-null object
8 EmployeeCount 1470 non-null int64
9 EmployeeNumber 1470 non-null int64
10 EnvironmentSatisfaction 1470 non-null int64
11 Gender 1470 non-null object
12 HourlyRate 1470 non-null int64
13 JobInvolvement 1470 non-null int64
14 JobLevel 1470 non-null int64
15 JobRole 1470 non-null object
16 JobSatisfaction 1470 non-null int64
17 MaritalStatus 1470 non-null object
18 MonthlyIncome 1470 non-null int64
19 MonthlyRate 1470 non-null int64
20 NumCompaniesWorked 1470 non-null int64
21 Over18 1470 non-null object
22 OverTime 1470 non-null object
23 PercentSalaryHike 1470 non-null int64

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Assignment 3 LR.DT.RFT.ipynb

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```
[ ] df.describe()
```

| | Age | DailyRate | DistanceFromHome | Education | EmployeeCount | EmployeeNumber | EnvironmentSatisfaction | HourlyRate | JobInvolvement | JobLevel | ... | RelationshipSatisfaction | StandardHours | StockId |
|-------|-------------|-------------|------------------|-------------|---------------|----------------|-------------------------|-------------|----------------|-------------|-----|--------------------------|---------------|---------|
| count | 1470.000000 | 1470.000000 | 1470.000000 | 1470.000000 | 1470.0 | 1470.000000 | 1470.000000 | 1470.000000 | 1470.000000 | 1470.000000 | ... | 1470.000000 | 1470.0 | 5 |
| mean | 36.923810 | 802.485714 | 9.192517 | 2.912925 | 1.0 | 1024.865306 | 2.721769 | 65.891156 | 2.729932 | 2.063946 | ... | 2.712245 | 80.0 | |
| std | 9.135373 | 403.509100 | 8.106864 | 1.024165 | 0.0 | 602.024335 | 1.093082 | 20.329428 | 0.711561 | 1.106940 | ... | 1.081209 | 0.0 | |
| min | 18.000000 | 102.000000 | 1.000000 | 1.000000 | 1.0 | 1.000000 | 1.000000 | 30.000000 | 1.000000 | 1.000000 | ... | 1.000000 | 80.0 | |
| 25% | 30.000000 | 465.000000 | 2.000000 | 2.000000 | 1.0 | 491.250000 | 2.000000 | 40.000000 | 2.000000 | 1.000000 | ... | 2.000000 | 80.0 | |
| 50% | 36.000000 | 802.000000 | 7.000000 | 3.000000 | 1.0 | 1020.500000 | 3.000000 | 66.000000 | 3.000000 | 2.000000 | ... | 3.000000 | 80.0 | |
| 75% | 43.000000 | 1167.000000 | 14.000000 | 4.000000 | 1.0 | 1556.750000 | 4.000000 | 83.750000 | 3.000000 | 3.000000 | ... | 4.000000 | 80.0 | |
| max | 60.000000 | 1499.000000 | 29.000000 | 5.000000 | 1.0 | 2068.000000 | 4.000000 | 100.000000 | 4.000000 | 5.000000 | ... | 4.000000 | 80.0 | |

8 rows x 26 columns

```
[ ] #Checking for Null Values.  
df.isnull().any()
```

| | Age | Attrition | BusinessTravel | DailyRate | Department | DistanceFromHome | Education | EducationField | EmployeeCount | EmployeeNumber | EnvironmentSatisfaction | Gender | HourlyRate | JobInvolvement | JobLevel | JobRole | JobSatisfaction | MaritalStatus | MonthlyIncome | MonthlyRate |
|--|-------|-----------|----------------|-----------|------------|------------------|-----------|----------------|---------------|----------------|-------------------------|--------|------------|----------------|----------|---------|-----------------|---------------|---------------|-------------|
| | False | False | False | False | False | False | False | False | False | False | False | False | False | False | False | False | False | False | False | False |

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```
[ ] YearsSinceLastPromotion  
YearsWithCurrManager  
dtype: bool
```

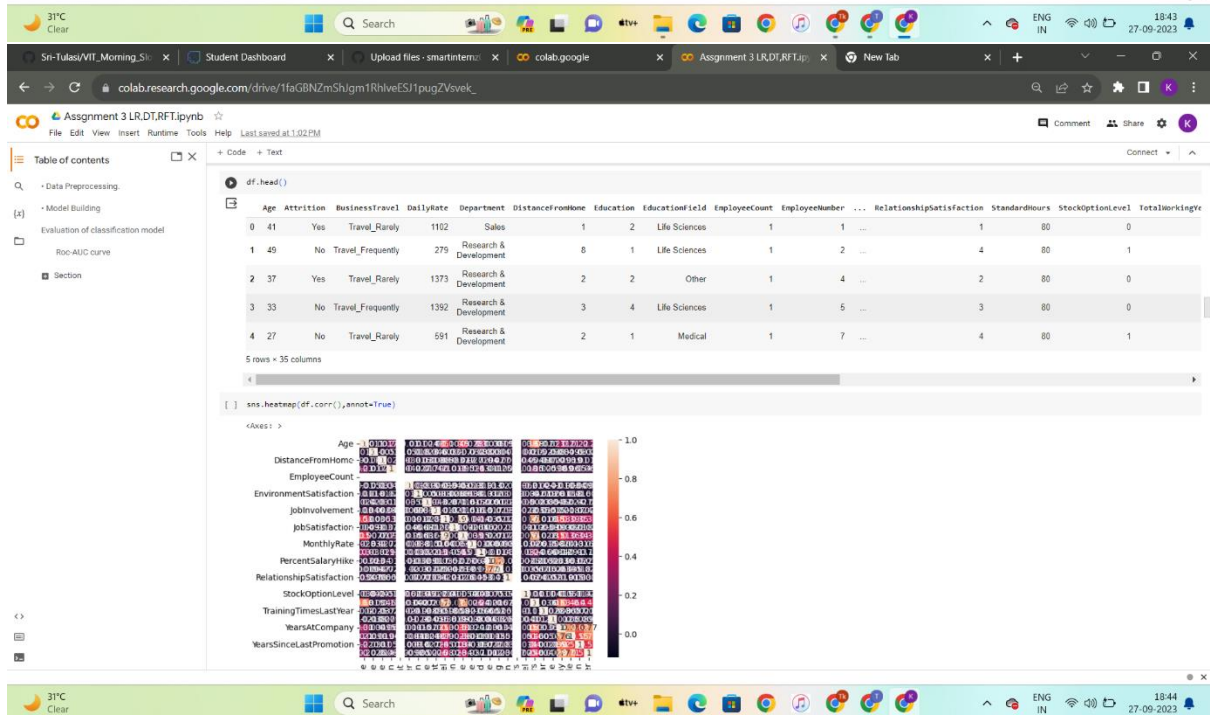
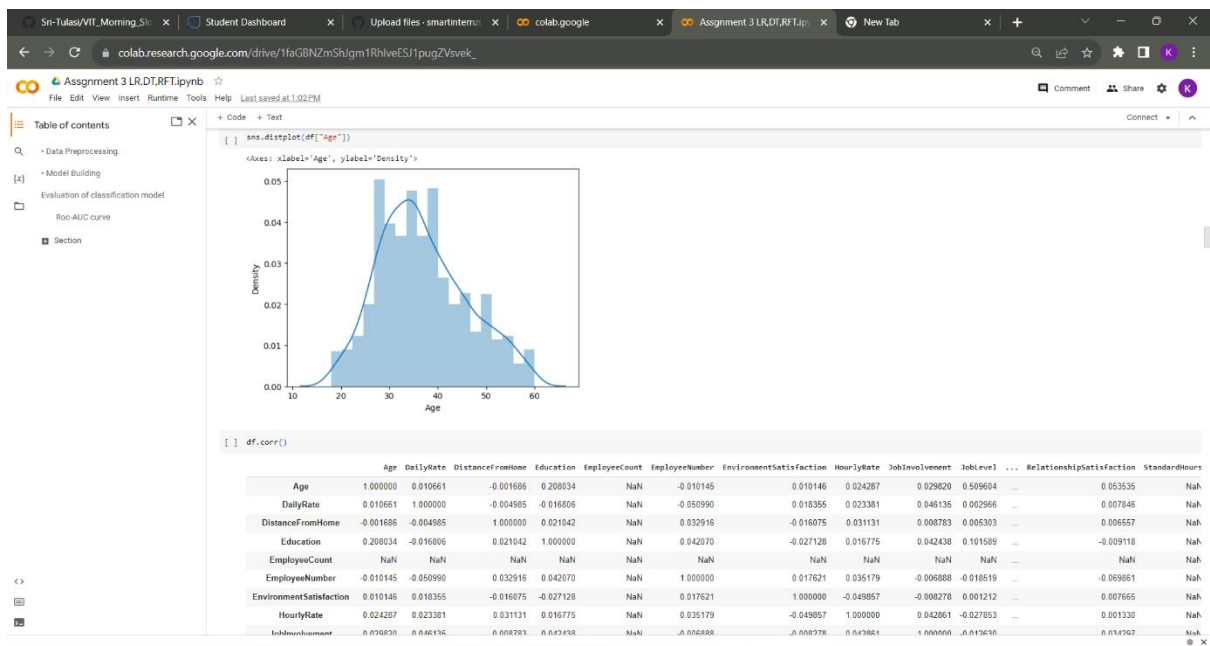
```
[ ] df.isnull().sum()
```

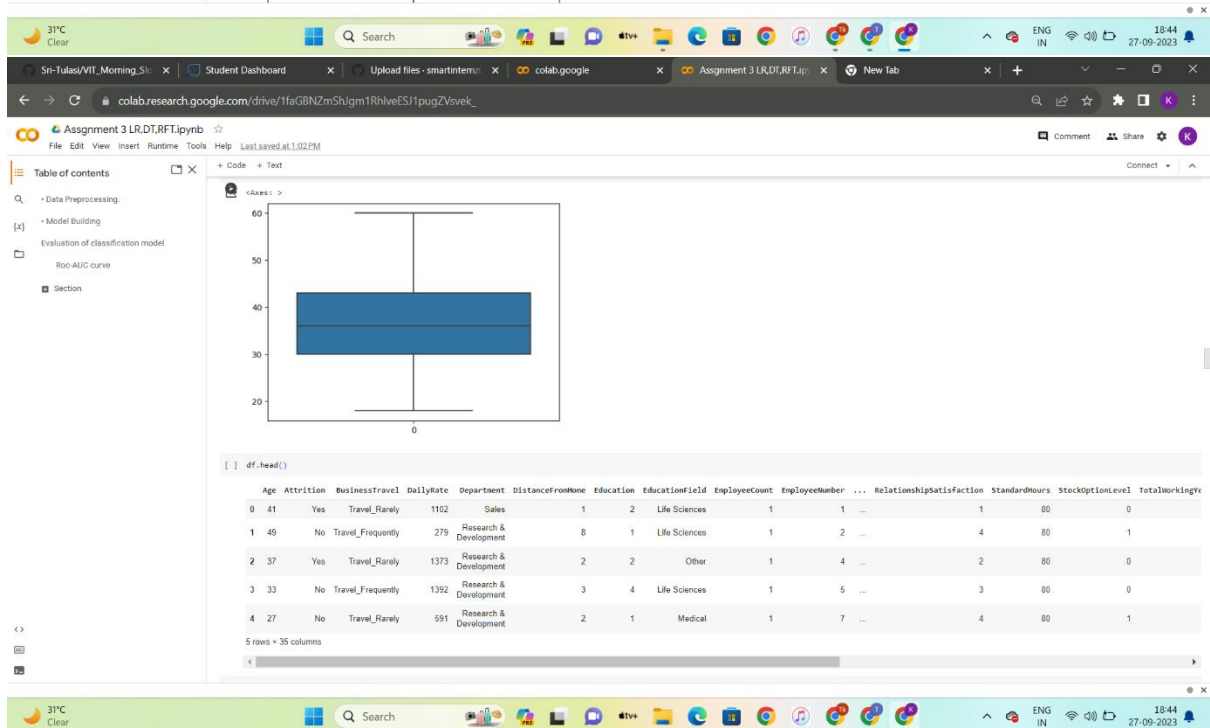
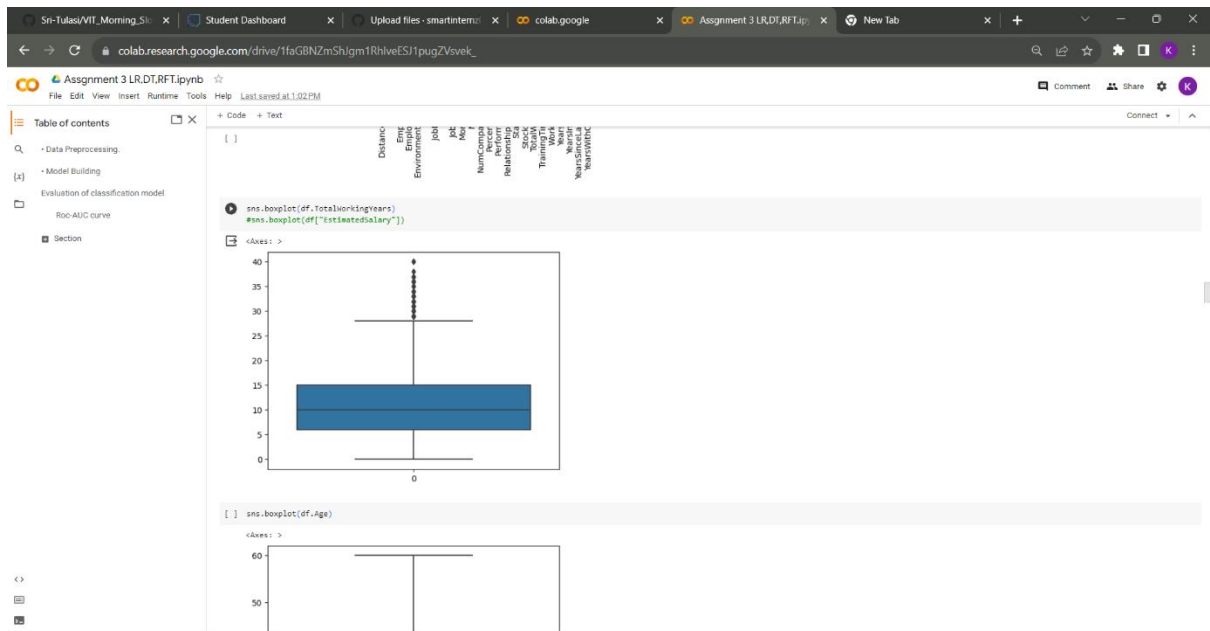
| | Age | Attrition | BusinessTravel | DailyRate | Department | DistanceFromHome | Education | EducationField | EmployeeCount | EmployeeNumber | EnvironmentSatisfaction | Gender | HourlyRate | JobInvolvement | JobLevel | JobRole | JobSatisfaction | MaritalStatus | MonthlyIncome | MonthlyRate | NumCompaniesWorked | OverTime | PercentSalaryHike | PerformanceRating | RelationshipSatisfaction | StandardHours | StockOptionLevel | TotalWorkingYears | TrainingTimesLastYear | WorkLifeBalance | YearsatCompany | YearsinCurrentRole | YearsSinceLastPromotion | YearsWithCurrManager |
|--|-----|-----------|----------------|-----------|------------|------------------|-----------|----------------|---------------|----------------|-------------------------|--------|------------|----------------|----------|---------|-----------------|---------------|---------------|-------------|--------------------|----------|-------------------|-------------------|--------------------------|---------------|------------------|-------------------|-----------------------|-----------------|----------------|--------------------|-------------------------|----------------------|
| | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 |

```
[ ] #Data Visualization.  
sns.distplot(df["Age"])
```

<x>: 'Age', y-label: 'Density'

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```
#Splitting Dependent and Independent variables
x=df.iloc[:,4]
y=df.DailyRate
y.head()
```

| | Age | Attrition | BusinessTravel | DailyRate |
|---|-----|-----------|-------------------|-----------|
| 0 | 41 | Yes | Travel_Rarely | 1102 |
| 1 | 49 | No | Travel_Frequently | 279 |
| 2 | 37 | Yes | Travel_Rarely | 1373 |
| 3 | 33 | No | Travel_Frequently | 1392 |
| 4 | 27 | No | Travel_Rarely | 591 |

```
[ ] y=df.DailyRate
y.head()
```

| | DailyRate |
|---|-----------|
| 0 | 1102 |
| 1 | 279 |
| 2 | 1373 |
| 3 | 1392 |
| 4 | 591 |

Name: DailyRate, dtype: int64

```
[ ] #label encoding
from sklearn.preprocessing import LabelEncoder
le=LabelEncoder()
x.BusinessTravel=le.fit_transform(x.BusinessTravel)
x.head()
```

| | Age | Attrition | BusinessTravel | DailyRate |
|---|-----|-----------|----------------|-----------|
| 0 | 41 | Yes | 2 | 1102 |
| 1 | 49 | No | 1 | 279 |
| 2 | 37 | Yes | 2 | 1373 |
| 3 | 33 | No | 1 | 1392 |
| 4 | 27 | No | 2 | 591 |

```
[ ] #label encoding
from sklearn.preprocessing import LabelEncoder
le=LabelEncoder()
x.Attrition=le.fit_transform(x.Attrition)
x.head()
```

| | Age | Attrition | BusinessTravel | DailyRate |
|---|-----|-----------|----------------|-----------|
| 0 | 41 | 1 | 2 | 1102 |
| 1 | 49 | 0 | 1 | 279 |
| 2 | 37 | 1 | 2 | 1373 |
| 3 | 33 | 0 | 1 | 1392 |
| 4 | 27 | 0 | 2 | 591 |

```
[ ] #feature scaling
from sklearn.preprocessing import MinMaxScaler
ms=MinMaxScaler()
x_scaled=pd.DataFrame(ms.fit_transform(x),columns=x.columns)
```

```
[ ] x_scaled
```

| | Age | Attrition | BusinessTravel | DailyRate |
|------|----------|-----------|----------------|-----------|
| 0 | 0.547619 | 1.0 | 1.0 | 0.715820 |
| 1 | 0.738095 | 0.0 | 0.5 | 0.126700 |
| 2 | 0.452381 | 1.0 | 1.0 | 0.909807 |
| 3 | 0.357143 | 0.0 | 0.5 | 0.923407 |
| 4 | 0.214286 | 0.0 | 1.0 | 0.350036 |
| ... | ... | ... | ... | ... |
| 1465 | 0.428571 | 0.0 | 0.5 | 0.559771 |
| 1466 | 0.500000 | 0.0 | 1.0 | 0.365784 |
| 1467 | 0.214286 | 0.0 | 1.0 | 0.037938 |
| 1468 | 0.738095 | 0.0 | 0.5 | 0.659270 |

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```
[ ] #Splitting Data into Train and Test.
from sklearn.model_selection import train_test_split
x_train,x_test,y_train,y_test=train_test_split(x_scaled,y,test_size=0.2,random_state=0)

[ ] x_train.shape,x_test.shape,y_train.shape,y_test.shape
((1176, 4), (294, 4), (1176,), (294,))

[ ] x_train.head()

      Age  Attrition  BusinessTravel  DailyRate
1374  0.952381      0.0             1.0  0.360057
1092  0.642057      0.0             1.0  0.607015
768   0.523810      0.0             1.0  0.141732
569   0.420571      0.0             0.0  0.953472
911   0.166667      1.0             0.5  0.355762
```

Model Building

- Import the model building libraries
- Initializing the model
- Training and testing the model
- Evaluation of Model
- Save the Model

```
[ ] from sklearn.tree import DecisionTreeClassifier
dtc=DecisionTreeClassifier()

[ ] dtc.fit(x_train,y_train)

DecisionTreeClassifier()

[ ] pred=dtc.predict(x_test)

array([ 635, 575, 663, 1490, 458, 169, 791, 1283, 142, 439, 1376,
        359, 995, 654, 1037, 1309, 618, 616, 326, 1107, 658, 824,
        499, 781, 1442, 567, 720, 1282, 591, 816, 401, 1212, 2023,
        588, 1457, 982, 720, 571, 669, 567, 194, 1239, 201, 119,
        913, 238, 591, 771, 495, 813, 1214, 1361, 143, 1498, 592,
        367, 688, 563, 1369, 161, 243, 1372, 488, 1476, 458, 638,
        1084, 827, 933, 1102, 1179, 691, 310, 672, 1375, 883, 155,
        715, 703, 444, 168, 810, 394, 330, 573, 1003, 692, 1254,
        1041, 1084, 1434, 1309, 125, 110, 697, 439, 136, 109, 1394,
        1005, 1366, 1179, 332, 1179, 804, 1305, 917, 1396, 584, 1442,
        1416, 902, 593, 448, 894, 179, 829, 896, 796, 1112, 1336,
        758, 374, 304, 111, 359, 1091, 1111, 1001, 200, 581, 679,
        1380, 974, 1400, 630, 249, 969, 1125, 720, 776, 1174, 442,
        852, 1435, 680, 294, 512, 539, 1354, 979, 634, 959, 1450,
        1117, 277, 1098, 303, 267, 1419, 482, 1320, 691, 377, 392,
        917, 974, 1130, 759, 520, 556, 439, 706, 1208, 1161, 959,
        1075, 515, 1004, 265, 1246, 1002, 1420, 1169, 1226, 147, 1179,
        587, 559, 305, 467, 1256, 499, 181, 1277, 248, 304, 1441,
        1476, 1117, 1309, 381, 381, 706, 658, 1189, 486, 855, 219,
        483, 921, 1403, 681, 1498, 725, 1495, 1094, 881, 1060, 1427,
        855, 202, 458, 179, 140, 1084, 548, 977, 638, 1103, 672,
        371, 852, 557, 1318, 515, 471, 1482, 1305, 104, 429, 581,
        1219, 652, 536, 452, 1109, 1569, 852, 1232, 400, 500, 205,
        589, 1440, 1220, 1249, 908, 1130, 1049, 672, 1141, 739, 598,
        1018, 802, 684, 1208, 1451, 1089, 196, 654, 976, 492, 588,
        573, 852, 1457, 374, 991, 1366, 1181, 887, 704, 448, 735,
        1179, 330, 1318, 559, 256, 251, 1266, 370])

[ ] y_test

442 635
1091 575
981 662
785 1492
1332 459
...
1439 557
403 254
124 249
198 1261
1229 309
Name: DailyRate, Length: 294, dtype: int64
```

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```
+ Code + Text
[] df
1229 369
Name: DailyRate, Length: 294, dtype: int64

[] df
Age Attrition BusinessTravel DailyRate Department DistanceFromHome Education EducationField EmployeeCount EmployeeNumber ... RelationshipSatisfaction StandardHours StockOptionLevel TotalWorkin
0 41 Yes Travel_Rarely 1102 Sales 1 2 Life Sciences 1 1 ... 1 80 0
1 49 No Travel_Frequently 279 Research & Development 8 1 Life Sciences 1 2 ... 4 80 1
2 37 Yes Travel_Rarely 1373 Research & Development 2 2 Other 1 4 ... 2 80 0
3 33 No Travel_Frequently 1392 Research & Development 3 4 Life Sciences 1 5 ... 3 80 0
4 27 No Travel_Rarely 591 Research & Development 2 1 Medical 1 7 ... 4 80 1
... ..
1465 36 No Travel_Frequently 884 Research & Development 23 2 Medical 1 2061 ... 3 80 1
1466 39 No Travel_Rarely 613 Research & Development 6 1 Medical 1 2062 ... 1 80 1
1467 27 No Travel_Rarely 155 Research & Development 4 3 Life Sciences 1 2064 ... 2 80 1
1468 49 No Travel_Frequently 1923 Sales 2 3 Medical 1 2065 ... 4 80 0
1469 34 No Travel_Rarely 628 Research & Development 8 3 Medical 1 2068 ... 1 80 0
1470 rows x 35 columns

[] dtc.predict(mn.transform([[1,19,1900,19000]]))
array([1499])
```

~ Evaluation of classification model

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```
+ Code + Text
~ Evaluation of classification model

[] accuracy_score
from sklearn.metrics import accuracy_score, confusion_matrix, classification_report, roc_auc_score, roc_curve

[] accuracy_score(y_test, pred)
0.3435374149659864

[] confusion_matrix(y_test, pred)
array([[0, 0, 1, ..., 0, 0, 0],
       [0, 0, 1, ..., 0, 0, 0],
       [0, 0, ..., 0, 0, 0],
       ...,
       [0, 0, ..., 1, 0, 0],
       [0, 0, ..., 0, 0, 1],
       [0, 0, ..., 0, 0, 0]])

[] pd.crosstab(y_test, pred)
col_0 104 109 111 119 140 142 143 147 155 160 ... 1441 1442 1450 1457 1462 1476 1482 1490 1495 1498
DailyRate
102 1 0 0 0 0 0 0 0 0 0 0 ... 0 0 0 0 0 0 0 0 0
103 1 0 0 0 0 0 0 0 0 0 0 ... 0 0 0 0 0 0 0 0 0
106 0 1 0 0 0 0 0 0 0 0 0 ... 0 0 0 0 0 0 0 0 0
111 0 0 1 0 0 0 0 0 0 0 0 ... 0 0 0 0 0 0 0 0 0
119 0 0 0 1 0 0 0 0 0 0 0 ... 0 0 0 0 0 0 0 0 0
... ..
1485 0 0 0 0 0 0 0 0 0 0 0 ... 0 0 0 0 0 1 1 0 0
1488 0 0 0 0 0 0 0 0 0 0 0 ... 0 0 0 0 0 0 0 1 0
1492 0 0 0 0 0 0 0 0 0 0 0 ... 0 0 0 0 0 0 0 1 0
1495 0 0 0 0 0 0 0 0 0 0 0 ... 0 0 0 0 0 0 0 0 1
1496 0 0 0 0 0 0 0 0 0 0 0 ... 0 0 0 0 0 0 0 0 1
```


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predicted no predicted yes

Actual No 58-TN 0-FP Actual yes 6-FN 16-TP

```
[ ] (58+16)/88 #accuracy
0.925
```

```
print(classification_report(y_test,pred))
```

| | precision | recall | f1-score | support |
|-----|-----------|--------|----------|---------|
| 192 | 0.00 | 0.00 | 0.00 | 1 |
| 193 | 0.00 | 0.00 | 0.00 | 1 |
| 194 | 0.00 | 0.00 | 0.00 | 0 |
| 196 | 0.00 | 0.00 | 0.00 | 1 |
| 199 | 0.00 | 0.00 | 0.00 | 0 |
| 111 | 1.00 | 1.00 | 1.00 | 1 |
| 113 | 1.00 | 1.00 | 1.00 | 1 |
| 140 | 0.00 | 0.00 | 0.00 | 0 |
| 141 | 0.00 | 0.00 | 0.00 | 1 |
| 142 | 0.00 | 0.00 | 0.00 | 1 |
| 143 | 0.00 | 0.00 | 0.00 | 0 |
| 145 | 0.00 | 0.00 | 0.00 | 1 |
| 147 | 1.00 | 1.00 | 1.00 | 1 |
| 155 | 1.00 | 1.00 | 1.00 | 1 |
| 168 | 1.00 | 1.00 | 1.00 | 1 |
| 168 | 1.00 | 1.00 | 1.00 | 1 |
| 177 | 0.00 | 0.00 | 0.00 | 1 |
| 179 | 0.00 | 0.00 | 0.00 | 0 |
| 181 | 0.00 | 0.00 | 0.00 | 0 |
| 182 | 0.00 | 0.00 | 0.00 | 1 |
| 185 | 0.00 | 0.00 | 0.00 | 1 |
| 194 | 0.00 | 0.00 | 0.00 | 0 |
| 195 | 0.00 | 0.00 | 0.00 | 1 |
| 196 | 0.00 | 0.00 | 0.00 | 0 |
| 200 | 0.00 | 0.00 | 0.00 | 1 |
| 201 | 1.00 | 1.00 | 1.00 | 1 |
| 202 | 1.00 | 1.00 | 1.00 | 1 |
| 218 | 0.00 | 0.00 | 0.00 | 1 |
| 219 | 0.00 | 0.00 | 0.00 | 0 |
| 238 | 1.00 | 1.00 | 1.00 | 1 |
| 240 | 1.00 | 1.00 | 1.00 | 1 |
| 243 | 0.00 | 0.00 | 0.00 | 0 |
| 244 | 0.00 | 0.00 | 0.00 | 1 |

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image.png

```
[ ] probability=ctc.predict_proba(x_test)
```

```
[ ] probability
array([[0., 0., ..., 0., 0., 0.],
       [0., 0., ..., 0., 0., 0.],
       [0., 0., ..., 0., 0., 0.],
       ...,
       [0., 0., ..., 0., 0., 0.],
       [0., 0., ..., 0., 0., 0.],
       [0., 0., ..., 0., 0., 0.]])
```

```
y_test
442    635
1091    575
981     662
785    1492
1332    459
...
1439    557
481     254
124     249
198    1261
1229     809
Name: DailyRate, Length: 294, dtype: int64
```

```
[ ] X, y = make_classification(n_samples=1500, n_features=10, n_classes=2, random_state=42)
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.2, random_state=42)
```

```
[ ] y_train_bin = label_binarize(y_train, classes=np.unique(y))
y_test_bin = label_binarize(y_test, classes=np.unique(y))
```

```
[ ] classifier = OneVsRestClassifier(LogisticRegression())
classifier.fit(X_train, y_train_bin)
```


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```

classfier = OneVsRestClassifier(LogisticRegression())
classfier.fit(X_train, y_train_bin)

OneVsRestClassifier
- estimator: LogisticRegression
  - LogisticRegression

[] y_probs = classfier.predict_proba(X_test)

fpr = dict()
tpr = dict()
roc_auc = dict()
for i in range(y_train_bin.shape[1]):
    fpr[i], tpr[i], _ = roc_curve(y_test_bin[:, i], y_probs[:, i])
    roc_auc[i] = roc_auc_score(y_test_bin[:, i], y_probs[:, i])

plt.figure(figsize=(8,6))
for i in range(y_train_bin.shape[1]):
    plt.plot(fpr[i], tpr[i], lw=2, label=f'ROC curve (Class {i}, AUC = {roc_auc[i]:.2f})')

```

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```

plt.figure(figsize=(8,6))
plt.plot([0, 1], [0, 1], color='gray', linestyle='--')
plt.xlim([0.0, 1.0])
plt.ylim([0.0, 1.05])
plt.xlabel('FPR')
plt.ylabel('TPR')
plt.title('Receiver Operating Characteristic (ROC) Curve')
plt.legend(loc='lower right')
plt.show()

```

WARNING:matplotlib.legend:No artists with labels found to put in legend. Note that artists whose label start with an underscore are ignored when legend() is called with no argument.

Receiver Operating Characteristic (ROC) Curve

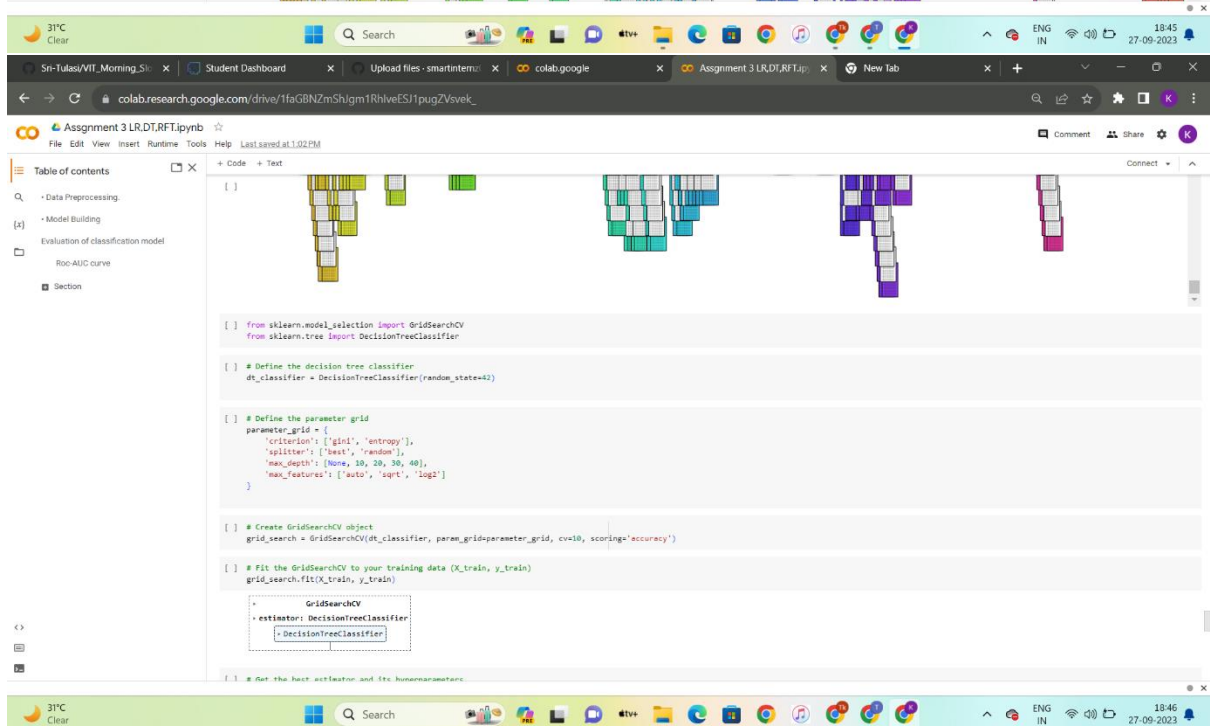
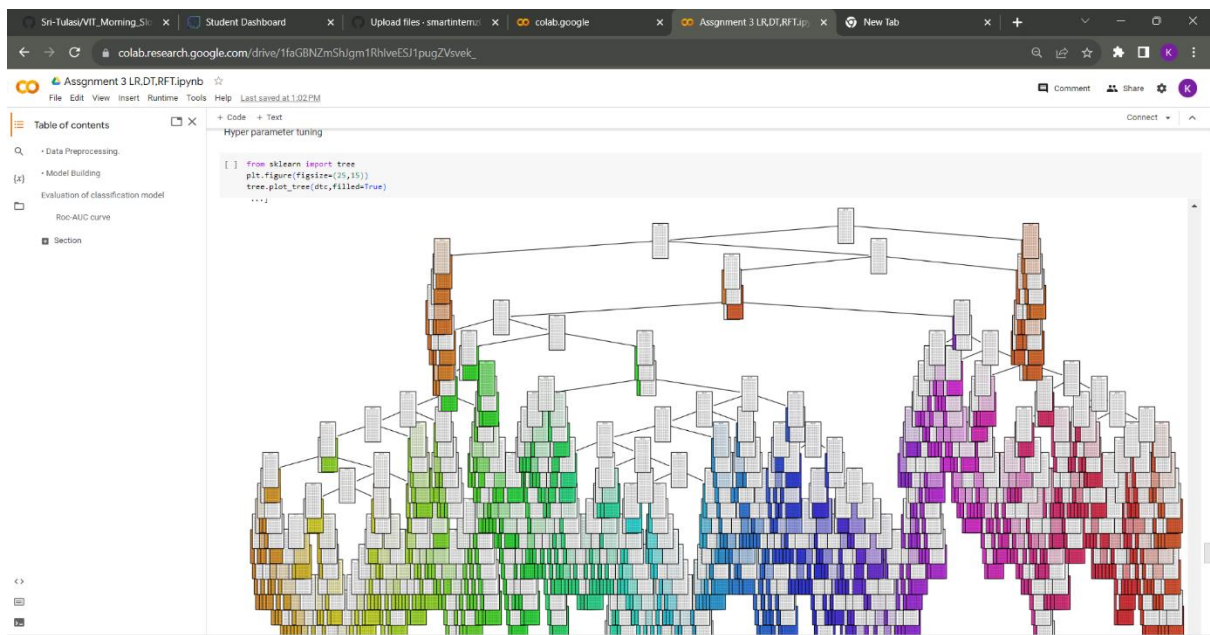
```

[] probability=dtc.predict_proba(x_test)[:,1]

Hyper parameter tuning

[] from sklearn import tree
plt.figure(figsize=(25,15))
tree.plot_tree(dtc, filled=True)

```



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Assignment 3 LR,DT,RF,TP.ipynb

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```
[ ] # Get the best estimator and its hyperparameters
best_dt_classifier = grid_search.best_estimator_
best_params = grid_search.best_params_

[ ] # Make predictions on the test set using the best estimator
pred = best_dt_classifier.predict(X_test)

[ ] # Evaluate the best classifier on the test data
accuracy = accuracy_score(y_test, pred)

[ ] best_params

{'criterion': 'gini',
 'max_depth': 10,
 'max_features': 'log2',
 'splitter': 'best'}

[ ] accuracy

0.8966666666666666

Random Forest

[ ] # Define the RandomForestClassifier and parameter grid
rfc = RandomForestClassifier(random_state=42)
param_grid = [
    {'n_estimators': [30, 50, 100],
     'max_depth': [None, 10, 20],
```

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```
[ ] # Create GridSearchCV object
rfc_cv = GridSearchCV(rfc, param_grid, cv=10, scoring='accuracy')

[ ] # Fit the GridSearchCV to the training data
rfc_cv.fit(X_train, y_train)

GridSearchCV
- estimator: RandomForestClassifier
  - RandomForestClassifier

[ ] # Get the best estimator and its hyperparameters
best_classifier = rfc_cv.best_estimator_
best_params = rfc_cv.best_params_

[ ] # Make predictions on the test set using the best estimator
pred = best_classifier.predict(X_test)

[ ] # Evaluate the best classifier on the test data
accuracy = accuracy_score(y_test, pred)

[ ] best_params

{'max_depth': 10, 'n_estimators': 100}

[ ] accuracy

0.9233333333333333
```

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