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In [1]: import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
from sklearn.model_selection import train_test_split
from sklearn.metrics import accuracy_score, classification_report
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In [2]: hrattr_data = pd.read_csv("C:/Users/USER/Downloads/WA_Fn-UseC_-HR-Employee-Attrition.csv")
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```
In [3]: hrattr_data
```

Out[3]:

	Age	Attrition	BusinessTravel	DailyRate	Department	DistanceFromHome	Education	EducationField	EmployeeCount	Employ
0	41	Yes	Travel_Rarely	1102	Sales	1	2	Life Sciences	1	
1	49	No	Travel_Frequently	279	Research & Development	8	1	Life Sciences	1	
2	37	Yes	Travel_Rarely	1373	Research & Development	2	2	Other	1	
3	33	No	Travel_Frequently	1392	Research & Development	3	4	Life Sciences	1	
4	27	No	Travel_Rarely	591	Research & Development	2	1	Medical	1	
...
1465	36	No	Travel_Frequently	884	Research & Development	23	2	Medical	1	
1466	39	No	Travel_Rarely	613	Research & Development	6	1	Medical	1	
1467	27	No	Travel_Rarely	155	Research & Development	4	3	Life Sciences	1	
1468	49	No	Travel_Frequently	1023	Sales	2	3	Medical	1	
1469	34	No	Travel_Rarely	628	Research & Development	8	3	Medical	1	

1470 rows × 35 columns

```
In [4]: print (hrattr_data.head())
```

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   Age Attrition   BusinessTravel DailyRate   Department \
0   41      Yes   Travel_Rarely    1102      Sales
1   49      No  Travel_Frequently    279  Research & Development
2   37      Yes   Travel_Rarely    1373  Research & Development
3   33      No  Travel_Frequently    1392  Research & Development
4   27      No   Travel_Rarely    591   Research & Development

   DistanceFromHome Education EducationField EmployeeCount EmployeeNumber \
0                1         2 Life Sciences             1             1
1                8         1 Life Sciences             1             2
2                2         2      Other              1             4
3                3         4 Life Sciences             1             5
4                2         1      Medical              1             7

   ... RelationshipSatisfaction StandardHours StockOptionLevel \
0   ...                      1             80              0
1   ...                      4             80              1
2   ...                      2             80              0
3   ...                      3             80              0
4   ...                      4             80              1

   TotalWorkingYears TrainingTimesLastYear WorkLifeBalance YearsAtCompany \
0                8                0                1             6
1               10                3                3            10
2                7                3                3             0
3                8                3                3             8
4                6                3                3             2

   YearsInCurrentRole YearsSinceLastPromotion YearsWithCurrManager
0                4                0                5
1                7                1                7
2                0                0                0
3                7                3                0
4                2                2                2

[5 rows x 35 columns]
```

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In [5]: hrattr_data['Attrition_ind'] = 0
hrattr_data.loc[hrattr_data['Attrition']=='Yes', 'Attrition_ind'] = 1
```

```
In [6]: dummy_busnstrvl = pd.get_dummies(hrattr_data['BusinessTravel'], prefix='busns_trvl')
dummy_dept = pd.get_dummies(hrattr_data['Department'], prefix='dept')
dummy_edufield = pd.get_dummies(hrattr_data['EducationField'], prefix='edufield')
dummy_gender = pd.get_dummies(hrattr_data['Gender'], prefix='gend')
dummy_jobrole = pd.get_dummies(hrattr_data['JobRole'], prefix='jobrole')
dummy_maritstat = pd.get_dummies(hrattr_data['MaritalStatus'], prefix='maritalstat')
dummy_overtime = pd.get_dummies(hrattr_data['OverTime'], prefix='overtime')
```

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In [7]: continuous_columns = ['Age', 'DailyRate', 'DistanceFromHome', 'Education', 'EnvironmentSatisfaction',
'HourlyRate', 'JobInvolvement', 'JobLevel', 'JobSatisfaction', 'MonthlyIncome', 'MonthlyRate', 'NumCompaniesWo
'PercentSalaryHike', 'PerformanceRating', 'RelationshipSatisfaction', 'StockOptionLevel', 'TotalWorkingYears'
'TrainingTimesLastYear', 'WorkLifeBalance', 'YearsAtCompany', 'YearsInCurrentRole', 'YearsSinceLastPromotion'
'YearsWithCurrManager']
```

```
In [8]: hrattr_continuous = hrattr_data[continuous_columns]
```

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In [9]: hrattr_continuous['Age'].describe()
hrattr_data['BusinessTravel'].value_counts()
```

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Out[9]: Travel_Rarely    1043
Travel_Frequently      277
Non-Travel             150
Name: BusinessTravel, dtype: int64
```

```
In [10]: hrattr_data_new = pd.concat([dummy_busnstrvl, dummy_dept, dummy_edufield, dummy_gender, dummy_jobrole,
dummy_maritstat, dummy_overtime, hrattr_continuous, hrattr_data['Attrition_ind']], axis=1)
```

```
In [11]: # Train & Test split
x_train,x_test,y_train,y_test = train_test_split(hrattr_data_new.drop(['Attrition_ind'],axis=1),
                                                hrattr_data_new['Attrition_ind'],train_size = 0.7,random_st
```

```
In [12]: # Decision Tree Classifier
from sklearn.tree import DecisionTreeClassifier
dt_fit = DecisionTreeClassifier(criterion="gini",max_depth=5,min_samples_split=2,min_samples_leaf=1,random_s
dt_fit.fit(x_train,y_train)
```

Out[12]: DecisionTreeClassifier(max_depth=5, random_state=42)

```
In [13]: print ("\nDecision Tree - Train Confusion Matrix\n\n",pd.crosstab(y_train,dt_fit.predict(x_train),rownames =
print ("\nDecision Tree - Train accuracy:",round(accuracy_score(y_train,dt_fit.predict(x_train)),3))
print ("\nDecision Tree - Train Classification Report\n",classification_report(y_train,dt_fit.predict(x_train))
```

Decision Tree - Train Confusion Matrix

	Predicted	0	1
Actual			
0		844	9
1		98	78

Decision Tree - Train accuracy: 0.896

Decision Tree - Train Classification Report				
	precision	recall	f1-score	support
0	0.90	0.99	0.94	853
1	0.90	0.44	0.59	176
accuracy			0.90	1029
macro avg	0.90	0.72	0.77	1029
weighted avg	0.90	0.90	0.88	1029

```
In [14]: print ("\nDecision Tree - Test Confusion Matrix\n\n",pd.crosstab(y_test,dt_fit.predict(x_test),rownames =
print ("\nDecision Tree - Test accuracy:",round(accuracy_score(y_test,dt_fit.predict(x_test)),3))
print ("\nDecision Tree - Test Classification Report\n",classification_report(y_test,dt_fit.predict(x_test))
```

Decision Tree - Test Confusion Matrix

	Predicted	0	1
Actual			
0		361	19
1		49	12

Decision Tree - Test accuracy: 0.846

Decision Tree - Test Classification Report				
	precision	recall	f1-score	support
0	0.88	0.95	0.91	380
1	0.39	0.20	0.26	61
accuracy			0.85	441
macro avg	0.63	0.57	0.59	441
weighted avg	0.81	0.85	0.82	441

```

In [16]: # Tuning class weights to analyze accuracy, precision & recall
dummyarray = np.empty((6,10))
dt_wttune = pd.DataFrame(dummyarray)

dt_wttune.columns = ["zero_wght", "one_wght", "tr_accuracy", "tst_accuracy", "prec_zero", "prec_one",
                    "prec_ovll", "recl_zero", "recl_one", "recl_ovll"]

zero_clwghts = [0.01, 0.1, 0.2, 0.3, 0.4, 0.5]

for i in range(len(zero_clwghts)):
    clwght = {0: zero_clwghts[i], 1: 1.0 - zero_clwghts[i]}
    dt_fit = DecisionTreeClassifier(criterion="gini", max_depth=5, min_samples_split=2,
                                   min_samples_leaf=1, random_state=42, class_weight = clwght)
    dt_fit.fit(x_train, y_train)
    dt_wttune.loc[i, 'zero_wght'] = clwght[0]
    dt_wttune.loc[i, 'one_wght'] = clwght[1]
    dt_wttune.loc[i, 'tr_accuracy'] = round(accuracy_score(y_train, dt_fit.predict(x_train)), 3)
    dt_wttune.loc[i, 'tst_accuracy'] = round(accuracy_score(y_test, dt_fit.predict(x_test)), 3)

    clf_sp = classification_report(y_test, dt_fit.predict(x_test), output_dict=True)

    dt_wttune.loc[i, 'prec_zero'] = clf_sp['0']['precision']
    dt_wttune.loc[i, 'prec_one'] = clf_sp['1']['precision']
    dt_wttune.loc[i, 'prec_ovll'] = clf_sp['macro avg']['precision']

    dt_wttune.loc[i, 'recl_zero'] = clf_sp['0']['recall']
    dt_wttune.loc[i, 'recl_one'] = clf_sp['1']['recall']
    dt_wttune.loc[i, 'recl_ovll'] = clf_sp['macro avg']['recall']

print ("\nClass Weights", clwght, "Train accuracy:", round(accuracy_score(y_train, dt_fit.predict(x_train))),
print ("Test Confusion Matrix\n\n", pd.crosstab(y_test, dt_fit.predict(x_test), rownames = ["Actual"], coln

```

Class Weights {0: 0.01, 1: 0.99} Train accuracy: 0.342 Test accuracy: 0.272
Test Confusion Matrix

	Predicted 0	Predicted 1
Actual 0	65	315
Actual 1	6	55

Class Weights {0: 0.1, 1: 0.9} Train accuracy: 0.806 Test accuracy: 0.732
Test Confusion Matrix

	Predicted 0	Predicted 1
Actual 0	282	98
Actual 1	20	41

Class Weights {0: 0.2, 1: 0.8} Train accuracy: 0.871 Test accuracy: 0.83
Test Confusion Matrix

	Predicted 0	Predicted 1
Actual 0	341	39
Actual 1	36	25

Class Weights {0: 0.3, 1: 0.7} Train accuracy: 0.881 Test accuracy: 0.837
Test Confusion Matrix

	Predicted 0	Predicted 1
Actual 0	345	35
Actual 1	37	24

Class Weights {0: 0.4, 1: 0.6} Train accuracy: 0.894 Test accuracy: 0.832
Test Confusion Matrix

	Predicted 0	Predicted 1
Actual 0	346	34
Actual 1	40	21

Class Weights {0: 0.5, 1: 0.5} Train accuracy: 0.896 Test accuracy: 0.846
Test Confusion Matrix

	Predicted 0	Predicted 1
Actual 0	361	19
Actual 1	49	12

In []: