

▼ ASSIGNMENT / TASK 9

Predict retention of an employee within an organization such that whether the employee will leave the company or continue with it. An organization is only as good as its employees, and these people are the true source of its competitive advantage. Dataset is downloaded from Kaggle.

First do data exploration and visualization, after this create a logistic regression model to predict Employee Attrition Using Machine Learning & Python.

```
import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
from sklearn.linear_model import LogisticRegression
from sklearn.metrics import classification_report, confusion_matrix
df=pd.read_csv("/content/archive (1).zip")
df.head()
```

	satisfaction_level	last_evaluation	number_project	average_monthly_hours	time_spend_company
0	0.38	0.53	2	157	1
1	0.80	0.86	5	262	3
2	0.11	0.88	7	272	2
3	0.72	0.87	5	223	2
4	0.37	0.52	2	159	1

```
df.shape
```

```
(14999, 10)
```

```
df.keys()
```

```
Index(['satisfaction_level', 'last_evaluation', 'number_project',
      'average_monthly_hours', 'time_spend_company', 'Work_accident', 'left',
      'promotion_last_5years', 'Department', 'salary'],
      dtype='object')
```

```
dummies=pd.get_dummies(df.Department)
merge=pd.concat([df,dummies],axis='columns')
dummies_new=pd.get_dummies(merge.salary)
merge_new=pd.concat([merge,dummies_new],axis="columns")
merge_new.head()
```

	satisfaction_level	last_evaluation	number_project	average_monthly_hours	time_spent
0	0.38	0.53	2	157	
1	0.80	0.86	5	262	
2	0.11	0.88	7	272	
3	0.72	0.87	5	223	
4	0.37	0.52	2	159	

```
merge_new=merge_new.drop(["salary"],axis="columns")
merge_new.head()
```

↗

	satisfaction_level	last_evaluation	number_project	average_monthly_hours	time_spent
0	0.38	0.53	2	157	
1	0.80	0.86	5	262	
2	0.11	0.88	7	272	
3	0.72	0.87	5	223	
4	0.37	0.52	2	159	

+ Code

+ Text

```
x=merge_new.loc[:,["satisfaction_level","last_evaluation","number_project","average_monthly_ho
print(x,"\n")
y=merge_new["left"]
print(y)
```

	satisfaction_level	last_evaluation	number_project	...	high	low	medium
0	0.38	0.53	2	...	0	1	0
1	0.80	0.86	5	...	0	0	1
2	0.11	0.88	7	...	0	0	1
3	0.72	0.87	5	...	0	1	0
4	0.37	0.52	2	...	0	1	0
...
14994	0.40	0.57	2	...	0	1	0
14995	0.37	0.48	2	...	0	1	0
14996	0.37	0.53	2	...	0	1	0
14997	0.11	0.96	6	...	0	1	0
14998	0.37	0.52	2	...	0	1	0

[14999 rows x 20 columns]

0	1
1	1
2	1
3	1
4	1
..	

```
14994    1
14995    1
14996    1
14997    1
14998    1
Name: left, Length: 14999, dtype: int64
```

```
from sklearn.model_selection import train_test_split
x_train, x_test, y_train, y_test=train_test_split(x,y,test_size=0.25, random_state=5)
mymodel=LogisticRegression()
mymodel.fit(x_train, y_train)
```

```
/usr/local/lib/python3.7/dist-packages/sklearn/linear_model/_logistic.py:940: Convergence
STOP: TOTAL NO. of ITERATIONS REACHED LIMIT.
```

Increase the number of iterations (max_iter) or scale the data as shown in:

<https://scikit-learn.org/stable/modules/preprocessing.html>

Please also refer to the documentation for alternative solver options:

https://scikit-learn.org/stable/modules/linear_model.html#logistic-regression

```
extra_warning_msg=_LOGISTIC_SOLVER_CONVERGENCE_MSG)
LogisticRegression(C=1.0, class_weight=None, dual=False, fit_intercept=True,
                    intercept_scaling=1, l1_ratio=None, max_iter=100,
                    multi_class='auto', n_jobs=None, penalty='l2',
                    random_state=None, solver='lbfgs', tol=0.0001, verbose=0,
                    warm_start=False)
```

```
y_pred=mymodel.predict(x_test)#y_test
print(y_pred)
```

```
[0 0 1 ... 0 0 0]
```

```
from sklearn.metrics import confusion_matrix
cm=confusion_matrix(y_test,y_pred)
cm
```

```
array([[2602,  229],
       [ 599,  320]])
```

```
from sklearn.metrics import accuracy_score
print(accuracy_score(y_test,y_pred))
```

```
0.7792
```

```
print(mymodel.score(x_train,y_train))
print(mymodel.score(x_train,mymodel.predict(x_train)))
```

```
0.8008711885500933
1.0
```

```
print(mymodel.score(y_test,y_test))
```

```
print(mymodel.score(x_test,y_test))  
print(mymodel.score(x_test,mymodel.predict(x_test)))
```

0.7792

1.0

```
print(classification_report(y, mymodel.predict(x)))
```

	precision	recall	f1-score	support
0	0.83	0.93	0.87	11428
1	0.62	0.37	0.46	3571
accuracy			0.80	14999
macro avg	0.72	0.65	0.67	14999
weighted avg	0.78	0.80	0.78	14999

