

employe-turnover-analysis-2

July 1, 2024

```
[1]: import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
import seaborn as sns
```

```
[ ]:
```

```
[2]: # Import the dataset
data = pd.read_excel('1673873196_hr_comma_sep.xlsx')
```

```
[3]: # Display the dataset
data
```

```
[3]:
```

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

	average_monthly_hours	time_spent_company	Work_accident	left	\
0	157	3	0	1	
1	262	6	0	1	
2	272	4	0	1	
3	223	5	0	1	
4	159	3	0	1	
...	
14994	151	3	0	1	
14995	160	3	0	1	
14996	143	3	0	1	
14997	280	4	0	1	

14998	158	3	0	1
-------	-----	---	---	---

	promotion_last_5years	sales	salary
0	0	sales	low
1	0	sales	medium
2	0	sales	medium
3	0	sales	low
4	0	sales	low
...
14994	0	support	low
14995	0	support	low
14996	0	support	low
14997	0	support	low
14998	0	support	low

[14999 rows x 10 columns]

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

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 14999 entries, 0 to 14998
Data columns (total 10 columns):
#   Column                Non-Null Count  Dtype
---  -
0   satisfaction_level     14999 non-null  float64
1   last_evaluation        14999 non-null  float64
2   number_project         14999 non-null  int64
3   average_monthly_hours  14999 non-null  int64
4   time_spend_company     14999 non-null  int64
5   Work_accident          14999 non-null  int64
6   left                   14999 non-null  int64
7   promotion_last_5years  14999 non-null  int64
8   sales                  14999 non-null  object
9   salary                 14999 non-null  object
dtypes: float64(2), int64(6), object(2)
memory usage: 1.1+ MB
```

```
[ ]: Perform the following steps:
1.      Perform data quality checks by checking for missing values, if any
2.      Understand what factors contributed most to employee turnover by EDA
2.1.    Draw a heatmap of the Correlation Matrix between all numerical
         features/columns in the data
2.2.    Draw the distribution plot of
         Employee Satisfaction (use column satisfaction_level)
         Employee Evaluation (use column last_evaluation)
         Employee Average Monthly Hours (use column average_monthly_hours)
```

2.3. Draw the bar plot of the Employee Project Count of both employees who left and stayed in the organization (use column number_project and hue column left), and give your inferences from the plot

```
[6]: # missing values
data.isnull().sum().any()
```

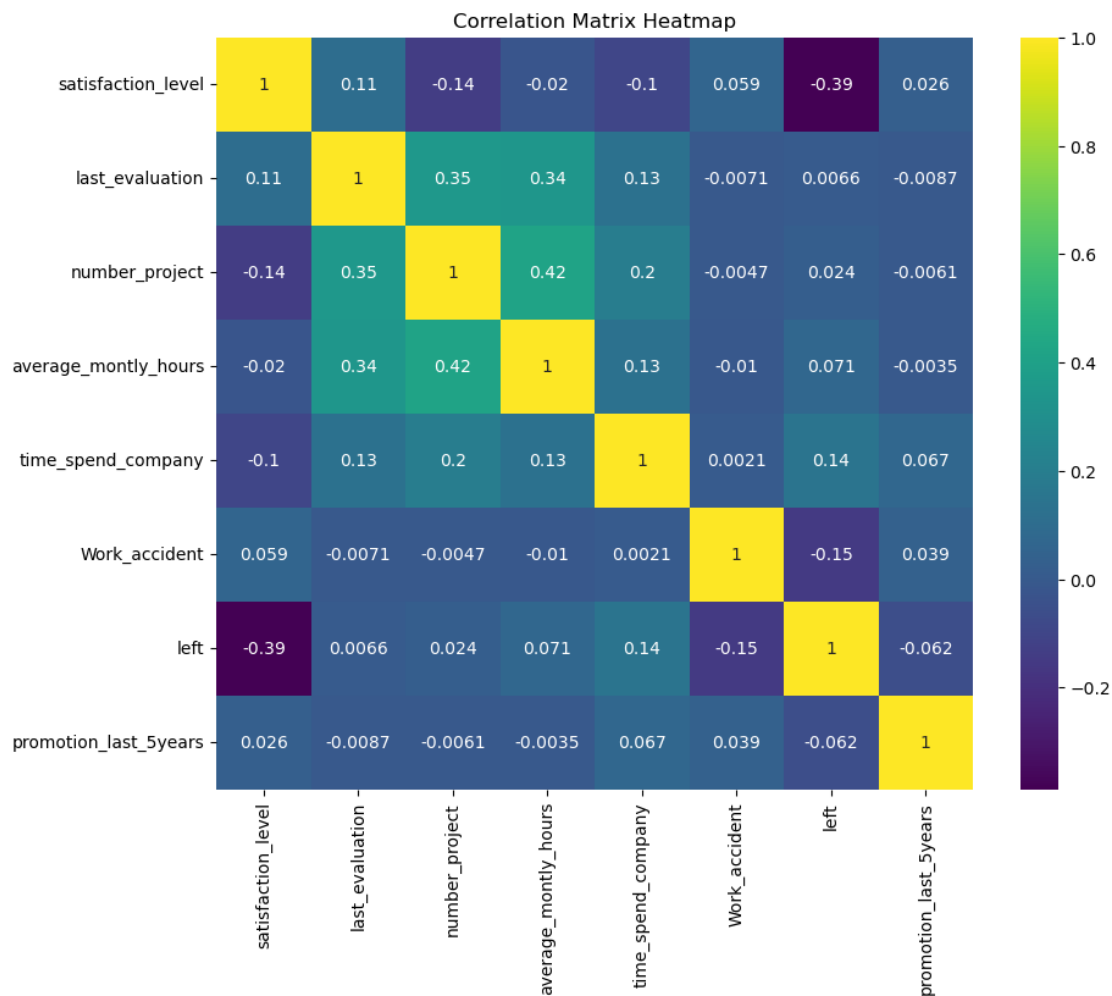
[6]: False

```
[7]: # correlation matrix
plt.figure(figsize=(10,8))
sns.heatmap(data.corr(numeric_only=True),annot=True)
plt.show()
```



```
[9]: # Generate the correlation matrix
corr_matrix = data.corr(numeric_only=True)

# Plot the heatmap with a specific color map
plt.figure(figsize=(10,8))
sns.heatmap(corr_matrix, annot=True, cmap='viridis')
plt.title('Correlation Matrix Heatmap')
plt.show()
```



```
[15]: # Create a figure with subplots
plt.figure(figsize=(18,6))

# Satisfaction level histogram
plt.subplot(1,3,1)
sns.histplot(data['satisfaction_level'], kde=True, color='red')
plt.title('Employee Satisfaction Level')
```

```

plt.xlabel('Satisfaction Level')
plt.ylabel('Frequency')

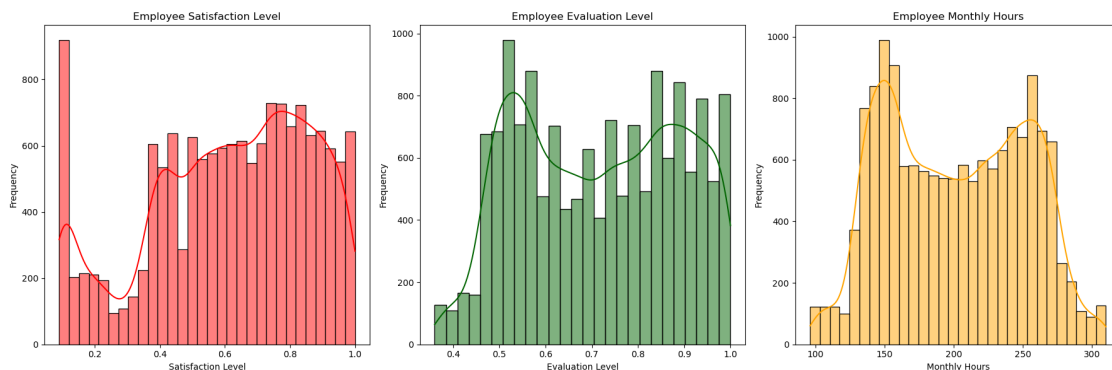
# Last evaluation histogram
plt.subplot(1,3,2)
sns.histplot(data['last_evaluation'], kde=True, color='darkgreen')
plt.title('Employee Evaluation Level')
plt.xlabel('Evaluation Level')
plt.ylabel('Frequency')

# Average monthly hours histogram
plt.subplot(1,3,3)
sns.histplot(data['average_monthly_hours'], kde=True, color='orange')
plt.title('Employee Monthly Hours')
plt.xlabel('Monthly Hours')
plt.ylabel('Frequency')

# Adjust layout for better spacing
plt.tight_layout()

# Show the plots
plt.show()

```



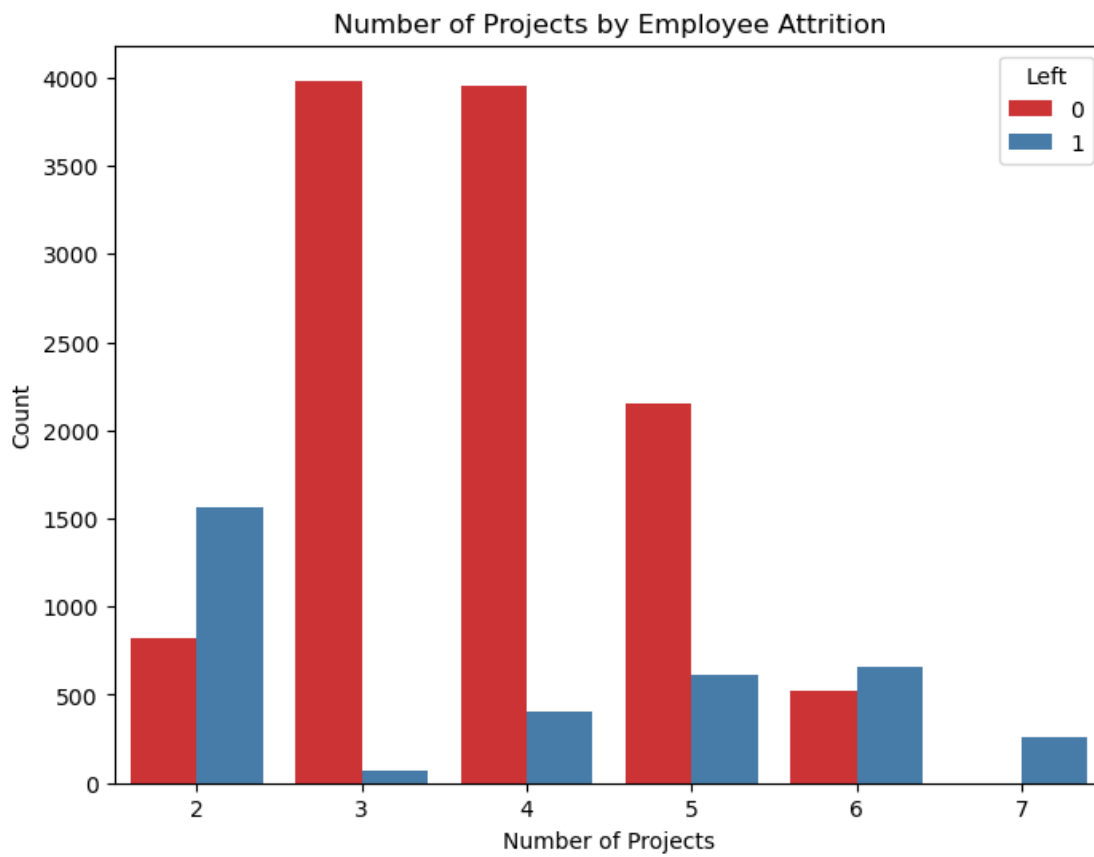
[]: Draw the bar plot of the Employee Project Count of both employees who left and stayed in the organization (use column number_project and hue column left), and give your inferences from the plot

```

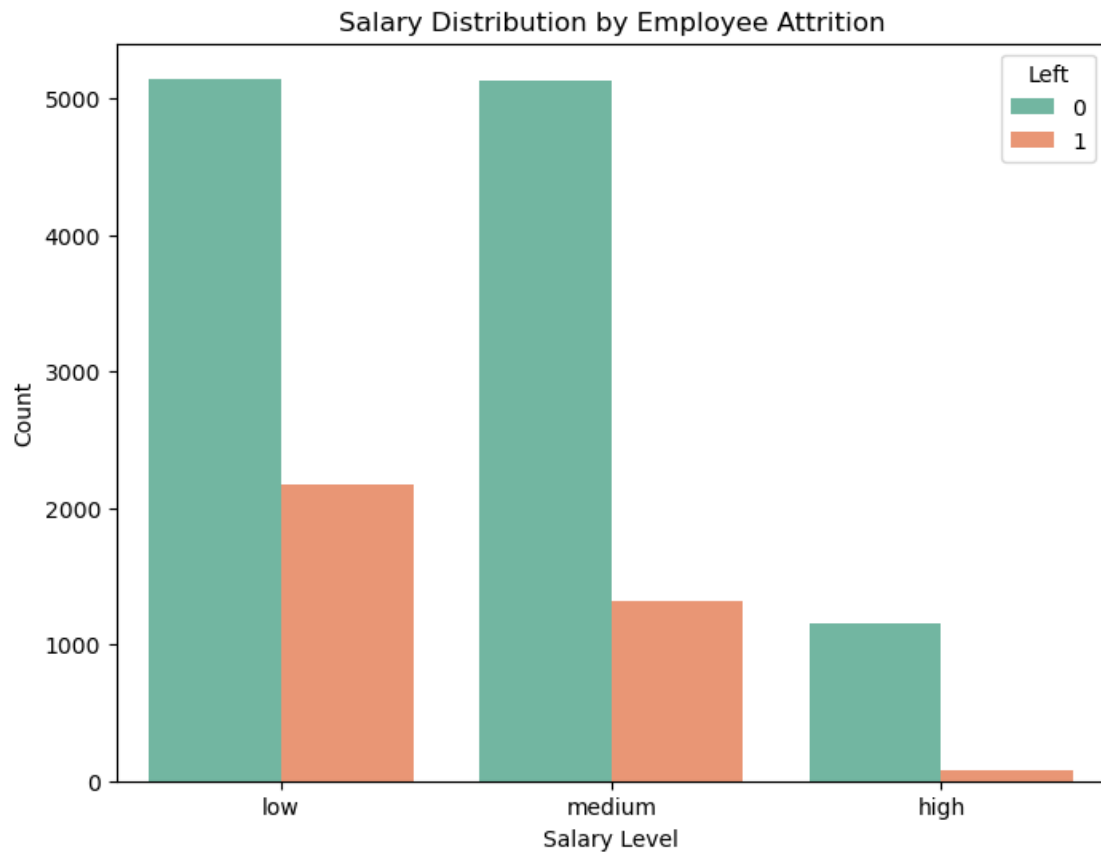
[16]: plt.figure(figsize=(8,6))
sns.countplot(x='number_project', hue='left', data=data, palette='Set1')
plt.title('Number of Projects by Employee Attrition')
plt.xlabel('Number of Projects')
plt.ylabel('Count')

```

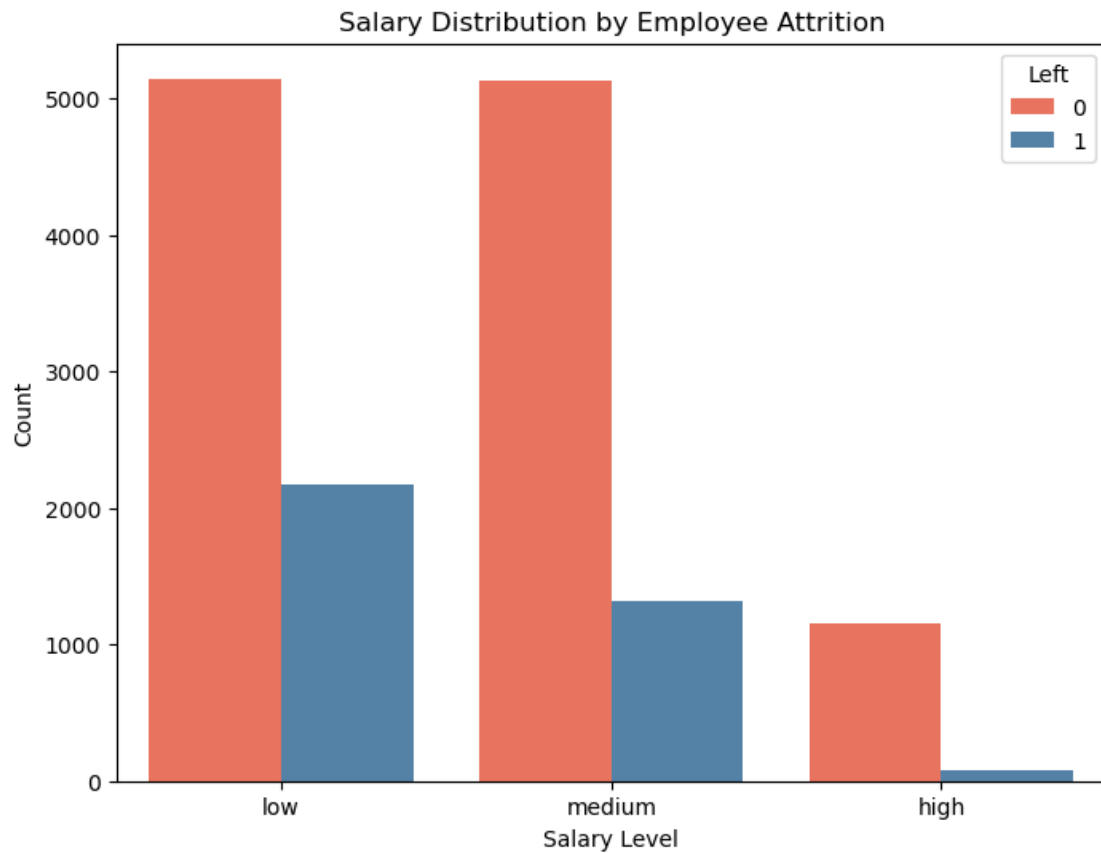
```
plt.legend(title='Left', loc='upper right')
plt.show()
```



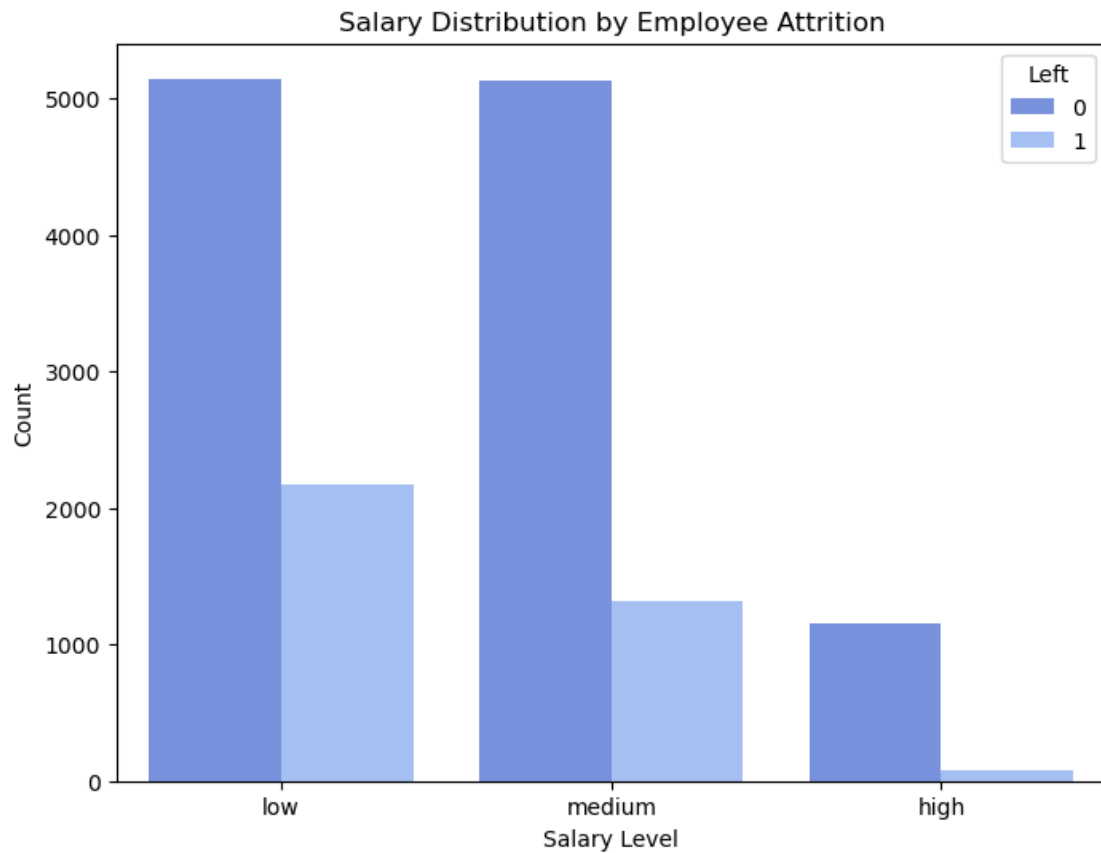
```
[17]: # Create the count plot with a custom color palette
plt.figure(figsize=(8,6))
sns.countplot(x='salary', hue='left', data=data, palette='Set2')
plt.title('Salary Distribution by Employee Attrition')
plt.xlabel('Salary Level')
plt.ylabel('Count')
plt.legend(title='Left', loc='upper right')
plt.show()
```



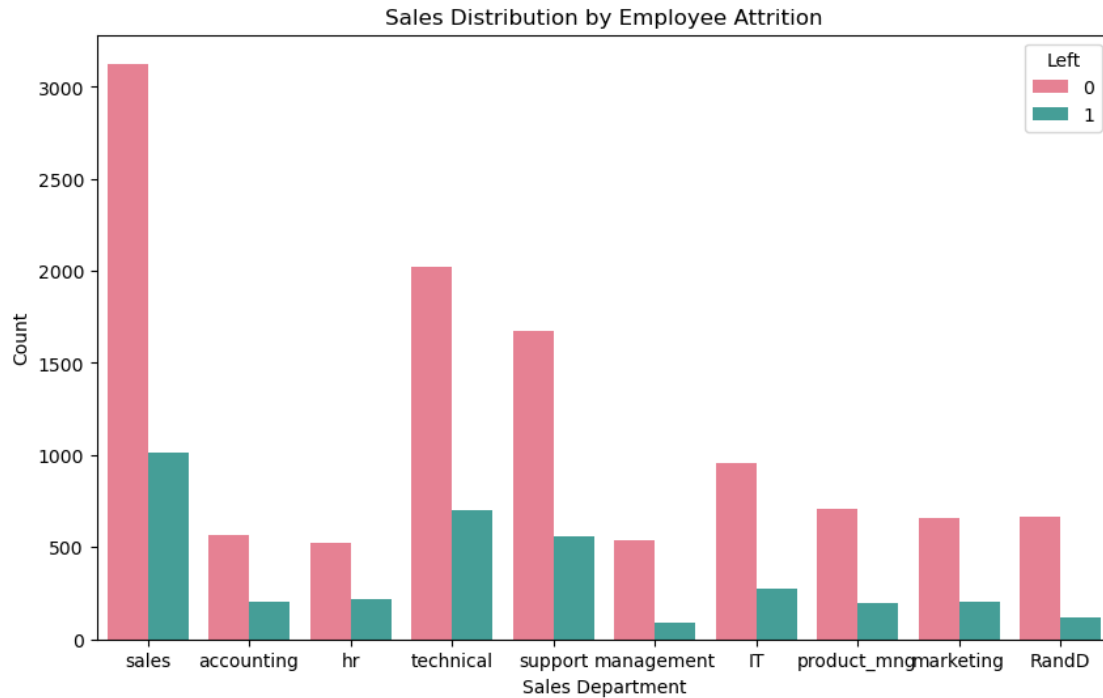
```
[18]: plt.figure(figsize=(8,6))
sns.countplot(x='salary', hue='left', data=data, palette=['#FF6347', '#4682B4'])
plt.title('Salary Distribution by Employee Attrition')
plt.xlabel('Salary Level')
plt.ylabel('Count')
plt.legend(title='Left', loc='upper right')
plt.show()
```



```
[19]: plt.figure(figsize=(8,6))
sns.countplot(x='salary', hue='left', data=data, palette=sns.
    color_palette("coolwarm"))
plt.title('Salary Distribution by Employee Attrition')
plt.xlabel('Salary Level')
plt.ylabel('Count')
plt.legend(title='Left', loc='upper right')
plt.show()
```

```
[20]: plt.figure(figsize=(10,6))
sns.countplot(x='sales', hue='left', data=data, palette='husl')
plt.title('Sales Distribution by Employee Attrition')
plt.xlabel('Sales Department')
plt.ylabel('Count')
plt.legend(title='Left', loc='upper right')
plt.show()
```



```
[ ]: 3.Perform clustering of employees who left based on their satisfaction and
      ↳evaluation
3.1. Choose columns satisfaction_level, last_evaluation and left
3.2. Do K-Means clustering of employees who left the company into 3 clusters?
3.3. Based on the satisfaction and evaluation factors, give your thoughts on
      ↳the employee clusters
```

```
[21]: # Display the column names
print(data.columns)
```

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

```
[22]: # select the relevant columns
cluster_data=data[['satisfaction_level', 'last_evaluation','left']]
left_emp_data=cluster_data[cluster_data['left']==1]
```

```
[23]: left_emp_data
```

```
[23]:      satisfaction_level  last_evaluation  left
0              0.38          0.53      1
1              0.80          0.86      1
```

2	0.11	0.88	1
3	0.72	0.87	1
4	0.37	0.52	1
...
14994	0.40	0.57	1
14995	0.37	0.48	1
14996	0.37	0.53	1
14997	0.11	0.96	1
14998	0.37	0.52	1

[3571 rows x 3 columns]

```
[ ]: # drop left label
left_emp_data.drop('left',axis=1,inplace=True)
```

```
[29]: left_emp_data
```

```
[29]:      satisfaction_level  last_evaluation
0          0.38          0.53
1          0.80          0.86
2          0.11          0.88
3          0.72          0.87
4          0.37          0.52
...
14994      0.40          0.57
14995      0.37          0.48
14996      0.37          0.53
14997      0.11          0.96
14998      0.37          0.52
```

[3571 rows x 2 columns]

```
[30]: import warnings
warnings.filterwarnings('ignore')
```

```
[31]: from sklearn.cluster import KMeans
```

```
[32]: # perform K means
kmeans=KMeans(n_clusters=3,random_state=42)
kmeans.fit(left_emp_data)
```

```
[32]: KMeans(n_clusters=3, random_state=42)
```

```
[33]: kmeans.labels_
```

```
[33]: array([0, 1, 2, ..., 0, 2, 0])
```

```
[56]: # Assuming you have already performed K-means clustering and have cluster labels
# Replace 'kmeans.labels_' with your actual cluster labels

# Check the length of cluster labels
print("Length of cluster labels:", len(kmeans.labels_))

# Check the length of the DataFrame
print("Length of DataFrame:", len(left_emp_data))

# Add cluster labels to the DataFrame if the lengths match
if len(kmeans.labels_) == len(left_emp_data):
    left_emp_data['cluster'] = kmeans.labels_
    print("Cluster labels added successfully.")
else:
    print("Length of cluster labels does not match length of DataFrame.")

# Display the DataFrame
left_emp_data.head()
```

Length of cluster labels: 3571
Length of DataFrame: 14999
Length of cluster labels does not match length of DataFrame.

```
[56]:
```

	satisfaction_level	last_evaluation	number_project	average_monthly_hours	\
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	

	time_spend_company	Work_accident	left	promotion_last_5years	sales	\
0	3	0	1	0	sales	
1	6	0	1	0	sales	
2	4	0	1	0	sales	
3	5	0	1	0	sales	
4	3	0	1	0	sales	

	salary
0	low
1	medium
2	medium
3	low
4	low

```
[61]: import pandas as pd
from sklearn.cluster import KMeans
import matplotlib.pyplot as plt
```

```

import seaborn as sns

# Load the dataset
data = pd.read_excel('1673873196_hr_comma_sep.xlsx')

# Filter employees who left the company
left_emp_data = data[data['left'] == 1]

# Select satisfaction_level and last_evaluation features
X = left_emp_data[['satisfaction_level', 'last_evaluation']]

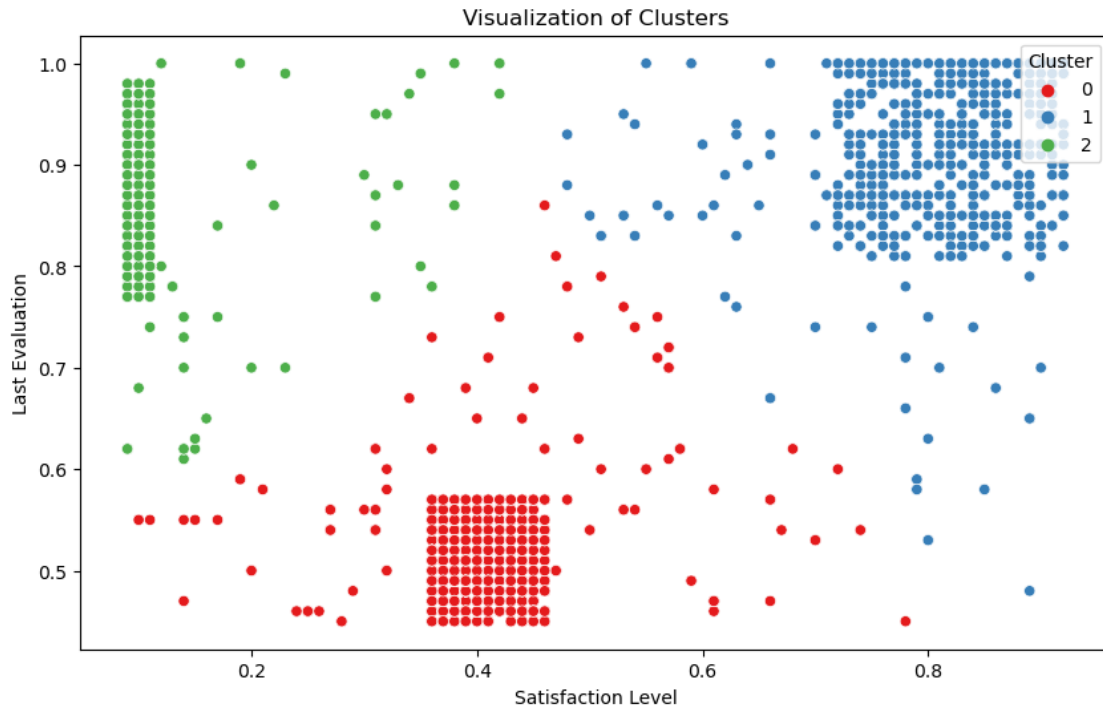
# Perform KMeans clustering
kmeans = KMeans(n_clusters=3, random_state=42)
kmeans.fit(X)

# Add cluster labels to the DataFrame
left_emp_data['cluster'] = kmeans.labels_

# Visualize the clusters
plt.figure(figsize=(10,6))
sns.scatterplot(x='satisfaction_level', y='last_evaluation', hue='cluster',
               palette='Set1', data=left_emp_data)
plt.title('Visualization of Clusters')
plt.xlabel('Satisfaction Level')
plt.ylabel('Last Evaluation')
plt.legend(title='Cluster', loc='upper right')
plt.show()

# Print the centroids of each cluster
print("Centroids of each cluster:")
print(kmeans.cluster_centers_)

```



Centroids of each cluster:

```
[[0.41014545 0.51698182]
 [0.80851586 0.91170931]
 [0.11115466 0.86930085]]
```

```
[62]: left_emp_data['cluster'].value_counts()
```

```
[62]: cluster
0    1650
1     977
2     944
Name: count, dtype: int64
```

```
[ ]: Handle the left Class Imbalance using the SMOTE technique
4.1. Pre-process the data by converting categorical columns to numerical
    ↳ columns by ↳ Separating categorical variables
    and numeric variables.
    ↳ Applying get_dummies() to the categorical variables.
    ↳ Combining categorical variables and numeric variables.
4.2. Do the stratified split of the dataset to train and test in the ratio 80:
    ↳ 20 with random_state=123
4.3. Upsample the train dataset using the SMOTE technique from the imblearn
    ↳ module
```

```
[8]: data.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 14999 entries, 0 to 14998
Data columns (total 10 columns):
#   Column                Non-Null Count  Dtype
---  -
0   satisfaction_level     14999 non-null  float64
1   last_evaluation        14999 non-null  float64
2   number_project         14999 non-null  int64
3   average_monthly_hours 14999 non-null  int64
4   time_spend_company     14999 non-null  int64
5   Work_accident          14999 non-null  int64
6   left                   14999 non-null  int64
7   promotion_last_5years  14999 non-null  int64
8   sales                  14999 non-null  object
9   salary                 14999 non-null  object
dtypes: float64(2), int64(6), object(2)
memory usage: 1.1+ MB
```

```
[9]: df_numerical=data.select_dtypes(include=['int64','float64'])
df_categorical=data.select_dtypes(include=['object'])
```

```
[10]: df_categorical
```

```
[10]:      sales  salary
0      sales    low
1      sales  medium
2      sales  medium
3      sales    low
4      sales    low
...
14994 support    low
14995 support    low
14996 support    low
14997 support    low
14998 support    low

[14999 rows x 2 columns]
```

```
[11]: # apply dummy encoding
df_converted=pd.get_dummies(data=df_categorical).astype('int')
```

```
[12]: df_converted
```

```
[12]:      sales_IT  sales_RandD  sales_accounting  sales_hr  sales_management \
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
...
14994	0	0	0	0	0	0
14995	0	0	0	0	0	0
14996	0	0	0	0	0	0
14997	0	0	0	0	0	0
14998	0	0	0	0	0	0

	sales_marketing	sales_product_mng	sales_sales	sales_support	\
0	0	0	1	0	
1	0	0	1	0	
2	0	0	1	0	
3	0	0	1	0	
4	0	0	1	0	
...	
14994	0	0	0	1	
14995	0	0	0	1	
14996	0	0	0	1	
14997	0	0	0	1	
14998	0	0	0	1	

	sales_technical	salary_high	salary_low	salary_medium
0	0	0	1	0
1	0	0	0	1
2	0	0	0	1
3	0	0	1	0
4	0	0	1	0
...
14994	0	0	1	0
14995	0	0	1	0
14996	0	0	1	0
14997	0	0	1	0
14998	0	0	1	0

[14999 rows x 13 columns]

```
[13]: # concat the data
df_new=pd.concat([df_numerical,df_converted],axis=1)
```

```
[14]: df_new.shape
```

```
[14]: (14999, 21)
```

```
[15]: df_new['left'].value_counts()
```



```
[15]: left
      0    11428
      1     3571
      Name: count, dtype: int64
```

```
[17]: import sklearn
      print(sklearn.__version__)
```

1.5.0

```
[ ]: Perform 5-fold cross-validation model training and evaluate performance
      5.1. Train a Logistic Regression model, apply a 5-fold CV, and plot the
          ↪ classification report
      5.2. Train a Random Forest Classifier model, apply the 5-fold CV, and plot the
          ↪ classification report
      5.3. Train a Gradient Boosting Classifier model, apply the 5-fold CV, and plot
          ↪ the classification report
```

```
[19]: pip install pandas numpy matplotlib seaborn scikit-learn imbalanced-learn
```

Requirement already satisfied: pandas in
c:\users\priya\appdata\roaming\python\python311\site-packages (2.2.2)Note: you
may need to restart the kernel to use updated packages.

Requirement already satisfied: numpy in
c:\users\priya\appdata\roaming\python\python311\site-packages (1.26.4)
Requirement already satisfied: matplotlib in
c:\users\priya\appdata\roaming\python\python311\site-packages (3.9.0)
Requirement already satisfied: seaborn in
c:\users\priya\appdata\roaming\python\python311\site-packages (0.13.2)
Requirement already satisfied: scikit-learn in
c:\users\priya\appdata\roaming\python\python311\site-packages (1.5.0)
Requirement already satisfied: imbalanced-learn in
c:\users\priya\appdata\roaming\python\python311\site-packages (0.12.3)
Requirement already satisfied: python-dateutil>=2.8.2 in
c:\users\priya\appdata\roaming\python\python311\site-packages (from pandas)
(2.8.2)
Requirement already satisfied: pytz>=2020.1 in
c:\users\priya\appdata\roaming\python\python311\site-packages (from pandas)
(2024.1)
Requirement already satisfied: tzdata>=2022.7 in
c:\users\priya\appdata\roaming\python\python311\site-packages (from pandas)
(2024.1)
Requirement already satisfied: contourpy>=1.0.1 in
c:\users\priya\appdata\roaming\python\python311\site-packages (from matplotlib)
(1.2.1)
Requirement already satisfied: cycler>=0.10 in

c:\users\priya\appdata\roaming\python\python311\site-packages (from matplotlib) (0.12.1)
Requirement already satisfied: fonttools>=4.22.0 in
c:\users\priya\appdata\roaming\python\python311\site-packages (from matplotlib) (4.53.0)
Requirement already satisfied: kiwisolver>=1.3.1 in
c:\users\priya\appdata\roaming\python\python311\site-packages (from matplotlib) (1.4.5)
Requirement already satisfied: packaging>=20.0 in
c:\users\priya\appdata\roaming\python\python311\site-packages (from matplotlib) (23.1)
Requirement already satisfied: pillow>=8 in
c:\users\priya\appdata\roaming\python\python311\site-packages (from matplotlib) (10.3.0)
Requirement already satisfied: pyparsing>=2.3.1 in
c:\users\priya\appdata\roaming\python\python311\site-packages (from matplotlib) (3.1.2)
Requirement already satisfied: scipy>=1.6.0 in
c:\users\priya\appdata\roaming\python\python311\site-packages (from scikit-learn) (1.13.1)
Requirement already satisfied: joblib>=1.2.0 in
c:\users\priya\appdata\roaming\python\python311\site-packages (from scikit-learn) (1.4.2)
Requirement already satisfied: threadpoolctl>=3.1.0 in
c:\users\priya\appdata\roaming\python\python311\site-packages (from scikit-learn) (3.5.0)
Requirement already satisfied: six>=1.5 in
c:\users\priya\appdata\roaming\python\python311\site-packages (from python-dateutil>=2.8.2->pandas) (1.16.0)

```
[23]: # Import necessary libraries
from sklearn.model_selection import cross_val_score
from sklearn.linear_model import LogisticRegression
from sklearn.metrics import accuracy_score, roc_auc_score, classification_report

# Apply Logistic Regression
log_reg = LogisticRegression(max_iter=1000, random_state=123)
log_reg.fit(X_train_smote, y_train_smote)

# Predict on the test set
y_pred1 = log_reg.predict(X_test)

# Print the accuracy score
print('Accuracy score:', accuracy_score(y_test, y_pred1))

# Print ROC AUC score
roc_auc = roc_auc_score(y_test, y_pred1)
```

```
print('ROC AUC score:', roc_auc)

# Print classification report
print('Classification Report:\n', classification_report(y_test, y_pred1))
```

Accuracy score: 0.7823333333333333

ROC AUC score: 0.7603810553092628

Classification Report:

	precision	recall	f1-score	support
0	0.90	0.80	0.85	2286
1	0.53	0.72	0.61	714
accuracy			0.78	3000
macro avg	0.72	0.76	0.73	3000
weighted avg	0.81	0.78	0.79	3000

C:\Users\priya\AppData\Roaming\Python\Python311\site-packages\sklearn\linear_model_logistic.py:469: ConvergenceWarning: lbfgs failed to converge (status=1):

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

```
n_iter_i = _check_optimize_result(
```

```
[25]: print(roc_auc_score(y_test,y_pred1))
```

0.7603810553092628

```
[26]: # Import necessary library
from sklearn import metrics

# Calculate predicted probabilities
y_pred_prob1 = log_reg.predict_proba(X_test)[:, 1]

# Calculate false positive rate, true positive rate, and thresholds
fpr, tpr, thresholds = metrics.roc_curve(y_test, y_pred_prob1)

# Print false positive rate, true positive rate, and thresholds
print("False Positive Rate (fpr):", fpr)
print("True Positive Rate (tpr):", tpr)
print("Thresholds:", thresholds)
```

```
# Calculate ROC AUC score
roc_auc = metrics.auc(fpr, tpr)
print("ROC AUC Score:", roc_auc)
```

False Positive Rate (fpr): [0.00000000e+00 8.74890639e-04 4.81189851e-03
5.68678915e-03

6.12423447e-03 6.99912511e-03 1.04986877e-02 1.04986877e-02
1.13735783e-02 1.18110236e-02 1.26859143e-02 1.31233596e-02
1.39982502e-02 2.01224847e-02 2.01224847e-02 2.09973753e-02
2.09973753e-02 2.36220472e-02 2.36220472e-02 2.58092738e-02
2.58092738e-02 2.66841645e-02 2.66841645e-02 2.71216098e-02
2.71216098e-02 2.75590551e-02 2.75590551e-02 2.84339458e-02
2.88713911e-02 2.88713911e-02 2.88713911e-02 3.06211724e-02
3.06211724e-02 3.06211724e-02 3.14960630e-02 3.14960630e-02
3.19335083e-02 3.19335083e-02 3.28083990e-02 3.28083990e-02
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6.80664917e-01	6.85914261e-01	6.85914261e-01	6.92038495e-01		
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8.07086614e-01	8.07524059e-01	8.07524059e-01	8.14523185e-01		
8.15398075e-01	8.23272091e-01	8.23272091e-01	8.28958880e-01		
8.28958880e-01	8.45144357e-01	8.45144357e-01	8.46019248e-01		
8.46894138e-01	8.69203850e-01	8.69203850e-01	8.77952756e-01		
8.77952756e-01	8.82327209e-01	8.83202100e-01	9.01574803e-01		
9.02449694e-01	9.13823272e-01	9.14698163e-01	9.16885389e-01		
9.16885389e-01	9.22134733e-01	9.23009624e-01	9.33945757e-01		
9.34820647e-01	9.44006999e-01	9.44881890e-01	9.49256343e-01		
9.49256343e-01	9.66754156e-01	9.66754156e-01	9.72878390e-01		
9.73753281e-01	9.84689414e-01	9.85564304e-01	9.97812773e-01		
9.98687664e-01	1.00000000e+00]				
True Positive Rate (tpr):	[0.	0.	0.	0.	0.
0.					
0.	0.00140056	0.00140056	0.00140056	0.00140056	0.00140056
0.00140056	0.00140056	0.00280112	0.00280112	0.00420168	0.00420168
0.00560224	0.00560224	0.0070028	0.0070028	0.00980392	0.00980392
0.01120448	0.01120448	0.01260504	0.01260504	0.01260504	0.0140056
0.01680672	0.01680672	0.01960784	0.02240896	0.02240896	0.02521008
0.02521008	0.02941176	0.02941176	0.03361345	0.03361345	0.03641457
0.03641457	0.03781513	0.03781513	0.04061625	0.04341737	0.04761905
0.05042017	0.05182073	0.05462185	0.05882353	0.05882353	0.06302521
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0.12605042	0.12605042	0.12885154	0.1302521	0.1302521	0.13305322
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0.15686275	0.16526611	0.16526611	0.16666667	0.16946779	0.17086835
0.17086835	0.17366947	0.17366947	0.17927171	0.17927171	0.18067227
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0.22408964	0.22689076	0.22689076	0.22829132	0.22829132	0.2394958
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```

ROC AUC Score: 0.8206608977799343

```

[27]: from sklearn import metrics

# Compute ROC curve
fpr, tpr, thresholds = metrics.roc_curve(y_test, y_pred1)

# Print FPR, TPR, and Thresholds
print("False Positive Rate (FPR):", fpr)
print("True Positive Rate (TPR):", tpr)
print("Thresholds:", thresholds)

# Calculate AUC
roc_auc = metrics.auc(fpr, tpr)

# Print AUC score
print("Area Under Curve (AUC):", roc_auc)

```

```

False Positive Rate (FPR): [0.          0.19772528 1.          ]
True Positive Rate (TPR): [0.          0.71848739 1.          ]
Thresholds: [inf  1.  0.]
Area Under Curve (AUC): 0.7603810553092628

```

```

[29]: import matplotlib.pyplot as plt

# Plot ROC curve
plt.figure(figsize=(8, 6))

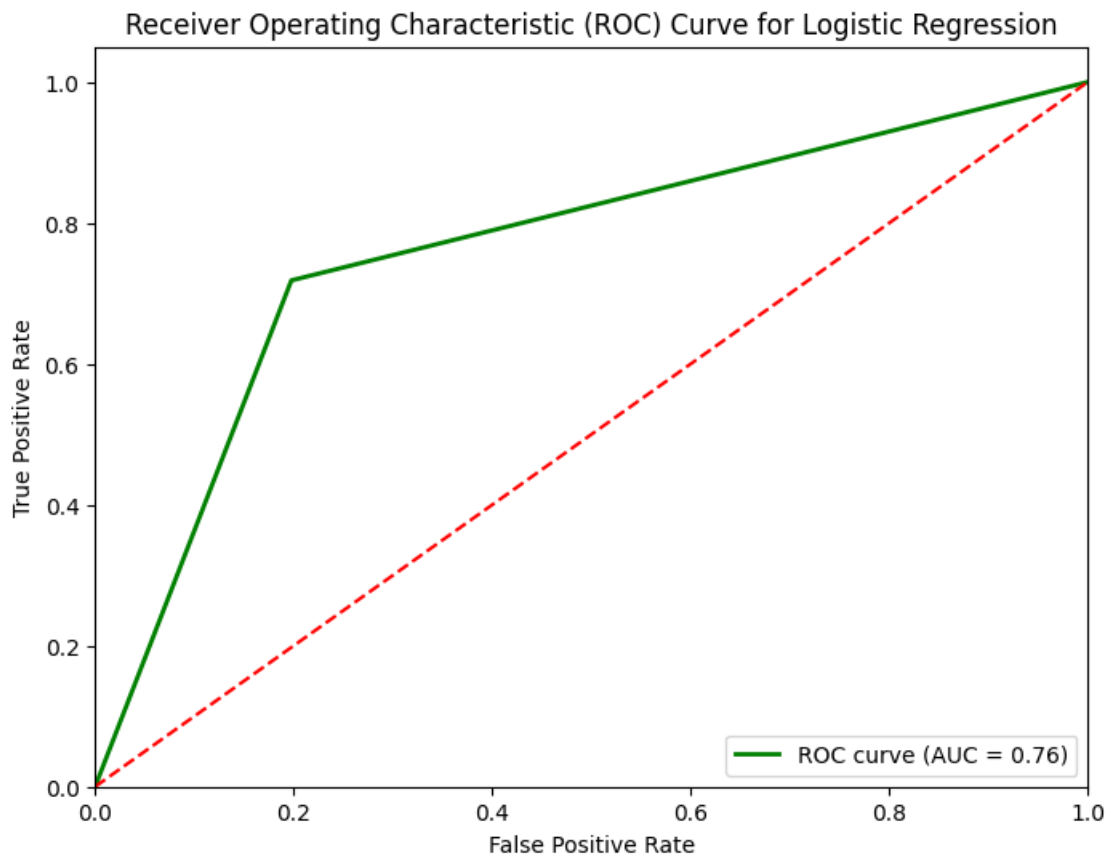
```



```

plt.plot(fpr, tpr, color='green', lw=2, label='ROC curve (AUC = %0.2f)' %
        roc_auc)
plt.plot([0, 1], [0, 1], color='red', linestyle='--')
plt.xlim([0.0, 1.0])
plt.ylim([0.0, 1.05])
plt.xlabel('False Positive Rate')
plt.ylabel('True Positive Rate')
plt.title('Receiver Operating Characteristic (ROC) Curve for Logistic
        Regression')
plt.legend(loc="lower right")
plt.show()

```



```

[36]: from sklearn.ensemble import RandomForestClassifier
      from sklearn.metrics import accuracy_score

      # Assuming X_train_resample and y_train_resample contain the resampled training
      # data using SMOTE

      # Create and train Random Forest Classifier

```

```

random_forest = RandomForestClassifier(max_depth=5)
random_forest.fit(X_train_smote, y_train_smote)

# Predict on test data
y_pred = random_forest.predict(X_test)

# Print accuracy score
print('Accuracy score:', accuracy_score(y_test, y_pred))

```

Accuracy score: 0.9536666666666667

```

[37]: from sklearn import metrics
      fpr, tpr, threshold = metrics.roc_curve(y_test, y_pred)
      print(fpr)
      print(tpr)
      print(threshold)
      roc_auc = metrics.auc(fpr, tpr)

```

```

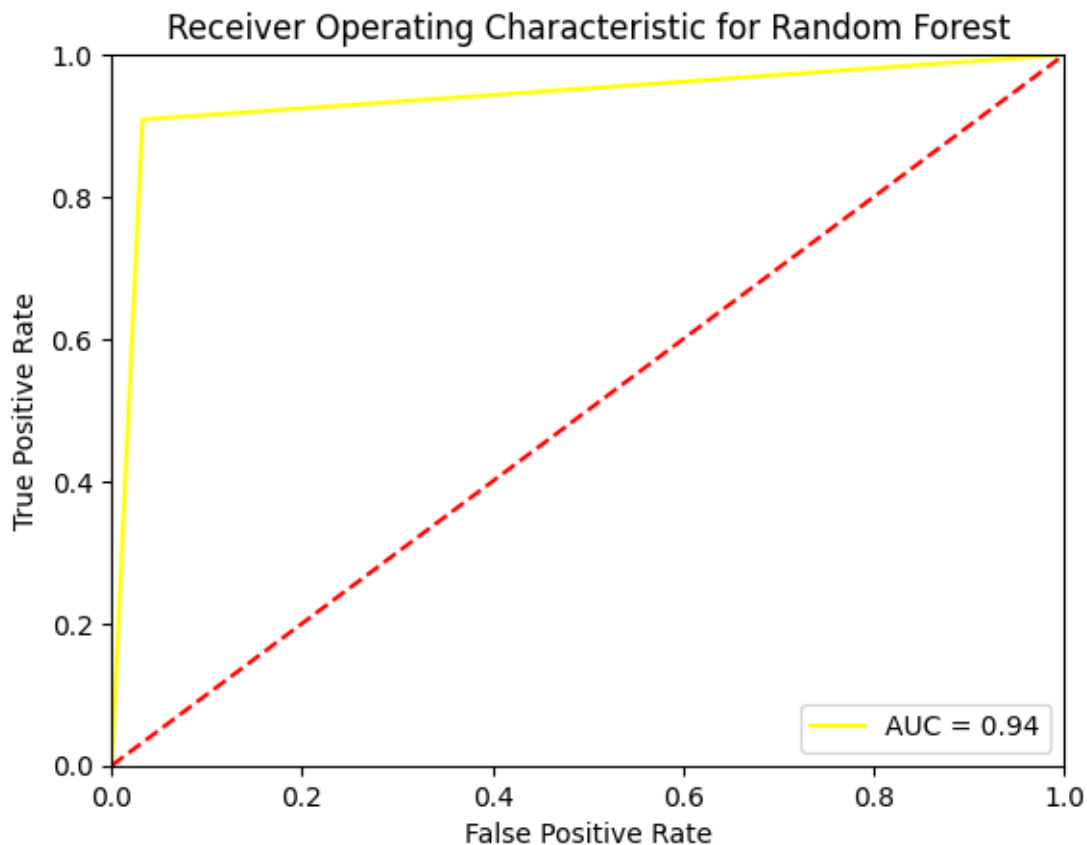
[0.          0.03237095 1.          ]
[0.          0.90896359 1.          ]
[inf  1.  0.]

```

```

[40]: # method I: plt
      plt.title('Receiver Operating Characteristic for Random Forest')
      plt.plot(fpr, tpr, 'yellow', label = 'AUC = %0.2f' % roc_auc) # Change 'b' to
      ↪ 'g' for green color
      plt.legend(loc = 'lower right')
      plt.plot([0, 1], [0, 1], 'r--')
      plt.xlim([0, 1])
      plt.ylim([0, 1])
      plt.ylabel('True Positive Rate')
      plt.xlabel('False Positive Rate')
      plt.show()

```



```
[41]: # apply Gradient Boosting
from sklearn.ensemble import GradientBoostingClassifier
gradient_boost=GradientBoostingClassifier()
```

```
[42]: gradient_boost.fit(X_train_smote,y_train_smote)
y_pred2=gradient_boost.predict(X_test)
```

```
[43]: print('Accuracy score',accuracy_score(y_test,y_pred2))
```

Accuracy score 0.9636666666666667

```
[44]: print(classification_report(y_test,y_pred2))
```

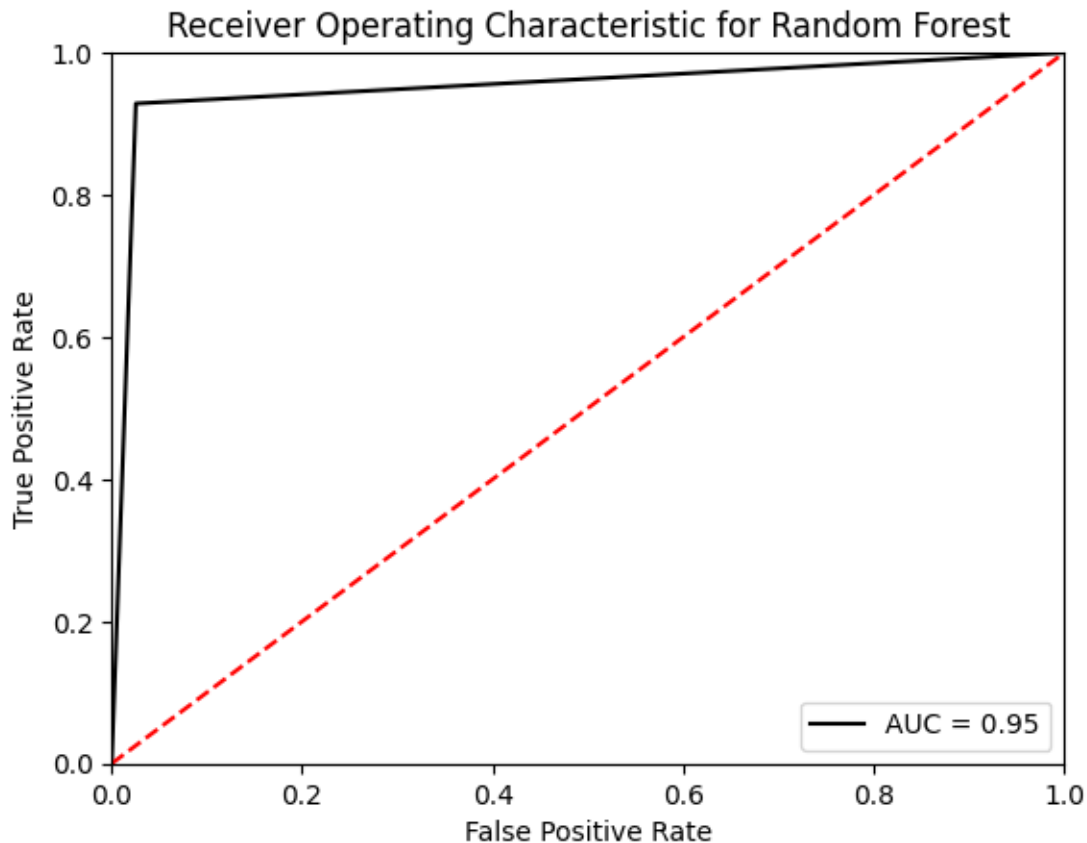
	precision	recall	f1-score	support
0	0.98	0.97	0.98	2286
1	0.92	0.93	0.92	714
accuracy			0.96	3000
macro avg	0.95	0.95	0.95	3000

weighted avg	0.96	0.96	0.96	3000
--------------	------	------	------	------

```
[45]: from sklearn import metrics
fpr, tpr, threshold = metrics.roc_curve(y_test, y_pred2)
print(fpr)
print(tpr)
print(threshold)
roc_auc = metrics.auc(fpr, tpr)
```

```
[0.          0.02537183 1.          ]
[0.          0.92857143 1.          ]
[inf  1.  0.]
```

```
[46]: # method I: plt
plt.title('Receiver Operating Characteristic for Random Forest')
plt.plot(fpr, tpr, 'black', label = 'AUC = %0.2f' % roc_auc)
plt.legend(loc = 'lower right')
plt.plot([0, 1], [0, 1], 'r--')
plt.xlim([0, 1])
plt.ylim([0, 1])
plt.ylabel('True Positive Rate')
plt.xlabel('False Positive Rate')
plt.show()
```



```
[47]: !pip install xgboost
```

Collecting xgboost

Downloading xgboost-2.0.3-py3-none-win_amd64.whl.metadata (2.0 kB)

Requirement already satisfied: numpy in

c:\users\priya\appdata\roaming\python\python311\site-packages (from xgboost) (1.26.4)

Requirement already satisfied: scipy in

c:\users\priya\appdata\roaming\python\python311\site-packages (from xgboost) (1.13.1)

Downloading xgboost-2.0.3-py3-none-win_amd64.whl (99.8 MB)

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-- ----- 5.9/99.8 MB 3.3 MB/s eta 0:00:29
-- ----- 5.9/99.8 MB 3.3 MB/s eta 0:00:29
-- ----- 6.1/99.8 MB 3.2 MB/s eta 0:00:29
-- ----- 6.4/99.8 MB 3.3 MB/s eta 0:00:29
-- ----- 6.7/99.8 MB 3.4 MB/s eta 0:00:28
-- ----- 7.0/99.8 MB 3.4 MB/s eta 0:00:27
-- ----- 7.3/99.8 MB 3.5 MB/s eta 0:00:27
-- ----- 7.4/99.8 MB 3.5 MB/s eta 0:00:27
--- ----- 7.7/99.8 MB 3.5 MB/s eta 0:00:26
--- ----- 7.8/99.8 MB 3.5 MB/s eta 0:00:27
--- ----- 8.1/99.8 MB 3.6 MB/s eta 0:00:26
--- ----- 8.5/99.8 MB 3.6 MB/s eta 0:00:26
--- ----- 8.8/99.8 MB 3.7 MB/s eta 0:00:25
--- ----- 9.1/99.8 MB 3.8 MB/s eta 0:00:25
--- ----- 9.4/99.8 MB 3.8 MB/s eta 0:00:24
--- ----- 9.5/99.8 MB 3.8 MB/s eta 0:00:24
--- ----- 9.7/99.8 MB 3.8 MB/s eta 0:00:24

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-----	10.0/99.8	MB	3.8	MB/s	eta	0:00:24
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-----	10.9/99.8	MB	4.5	MB/s	eta	0:00:20
-----	11.0/99.8	MB	4.5	MB/s	eta	0:00:20
-----	11.2/99.8	MB	4.5	MB/s	eta	0:00:20
-----	11.4/99.8	MB	4.5	MB/s	eta	0:00:20
-----	11.6/99.8	MB	4.4	MB/s	eta	0:00:20
-----	12.0/99.8	MB	4.7	MB/s	eta	0:00:19
-----	12.3/99.8	MB	4.7	MB/s	eta	0:00:19
-----	12.5/99.8	MB	4.7	MB/s	eta	0:00:19
-----	12.7/99.8	MB	4.7	MB/s	eta	0:00:19
-----	13.0/99.8	MB	4.7	MB/s	eta	0:00:19
-----	13.5/99.8	MB	4.7	MB/s	eta	0:00:19
-----	13.8/99.8	MB	4.7	MB/s	eta	0:00:19
-----	13.9/99.8	MB	4.7	MB/s	eta	0:00:19
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-----	14.7/99.8	MB	4.8	MB/s	eta	0:00:18
-----	14.7/99.8	MB	4.7	MB/s	eta	0:00:18
-----	14.9/99.8	MB	4.6	MB/s	eta	0:00:19
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-----	15.2/99.8	MB	5.0	MB/s	eta	0:00:17
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-----	15.7/99.8	MB	5.0	MB/s	eta	0:00:17
-----	15.8/99.8	MB	4.9	MB/s	eta	0:00:18
-----	16.1/99.8	MB	4.9	MB/s	eta	0:00:17
-----	16.3/99.8	MB	5.0	MB/s	eta	0:00:17
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-----	17.4/99.8	MB	5.0	MB/s	eta	0:00:17
-----	17.7/99.8	MB	5.1	MB/s	eta	0:00:17
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-----	18.1/99.8	MB	5.1	MB/s	eta	0:00:16
-----	18.3/99.8	MB	5.0	MB/s	eta	0:00:17
-----	18.4/99.8	MB	5.0	MB/s	eta	0:00:17
-----	18.7/99.8	MB	4.9	MB/s	eta	0:00:17
-----	19.0/99.8	MB	4.9	MB/s	eta	0:00:17
-----	19.3/99.8	MB	4.9	MB/s	eta	0:00:17
-----	19.4/99.8	MB	4.9	MB/s	eta	0:00:17
-----	19.7/99.8	MB	4.9	MB/s	eta	0:00:17
-----	19.8/99.8	MB	4.9	MB/s	eta	0:00:17
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-----	20.4/99.8	MB	4.9	MB/s	eta	0:00:17
-----	20.8/99.8	MB	5.0	MB/s	eta	0:00:16
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-----	-----	21.7/99.8	MB	5.1	MB/s	eta	0:00:16
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-----	-----	22.8/99.8	MB	5.0	MB/s	eta	0:00:16
-----	-----	23.0/99.8	MB	5.0	MB/s	eta	0:00:16
-----	-----	23.3/99.8	MB	5.1	MB/s	eta	0:00:16
-----	-----	23.7/99.8	MB	5.1	MB/s	eta	0:00:15
-----	-----	24.0/99.8	MB	5.1	MB/s	eta	0:00:15
-----	-----	24.2/99.8	MB	5.1	MB/s	eta	0:00:15
-----	-----	24.5/99.8	MB	5.1	MB/s	eta	0:00:15
-----	-----	24.7/99.8	MB	5.2	MB/s	eta	0:00:15
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-----	-----	25.7/99.8	MB	5.5	MB/s	eta	0:00:14
-----	-----	26.0/99.8	MB	5.5	MB/s	eta	0:00:14
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-----	-----	26.6/99.8	MB	5.4	MB/s	eta	0:00:14
-----	-----	26.9/99.8	MB	5.4	MB/s	eta	0:00:14
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-----	-----	27.7/99.8	MB	5.5	MB/s	eta	0:00:14
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-----	-----	31.8/99.8	MB	5.5	MB/s	eta	0:00:13
-----	-----	31.9/99.8	MB	5.5	MB/s	eta	0:00:13
-----	-----	32.2/99.8	MB	5.4	MB/s	eta	0:00:13
-----	-----	32.3/99.8	MB	5.4	MB/s	eta	0:00:13
-----	-----	32.4/99.8	MB	5.3	MB/s	eta	0:00:13
-----	-----	32.8/99.8	MB	5.5	MB/s	eta	0:00:13
-----	-----	33.2/99.8	MB	5.5	MB/s	eta	0:00:13
-----	-----	33.2/99.8	MB	5.4	MB/s	eta	0:00:13


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----- 33.6/99.8 MB 5.4 MB/s eta 0:00:13
----- 33.8/99.8 MB 5.4 MB/s eta 0:00:13
----- 33.8/99.8 MB 5.4 MB/s eta 0:00:13
----- 34.0/99.8 MB 5.2 MB/s eta 0:00:13
----- 34.1/99.8 MB 5.1 MB/s eta 0:00:13
----- 34.3/99.8 MB 5.2 MB/s eta 0:00:13
----- 34.4/99.8 MB 5.1 MB/s eta 0:00:13
----- 34.4/99.8 MB 5.1 MB/s eta 0:00:13
----- 34.6/99.8 MB 4.9 MB/s eta 0:00:14
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----- 36.1/99.8 MB 4.7 MB/s eta 0:00:14
----- 36.5/99.8 MB 4.7 MB/s eta 0:00:14
----- 36.6/99.8 MB 4.7 MB/s eta 0:00:14
----- 36.7/99.8 MB 4.7 MB/s eta 0:00:14
----- 36.8/99.8 MB 4.7 MB/s eta 0:00:14
----- 37.1/99.8 MB 4.7 MB/s eta 0:00:14
----- 37.3/99.8 MB 4.7 MB/s eta 0:00:14
----- 37.5/99.8 MB 4.6 MB/s eta 0:00:14
----- 37.5/99.8 MB 4.6 MB/s eta 0:00:14
----- 37.6/99.8 MB 4.5 MB/s eta 0:00:14
----- 37.8/99.8 MB 4.5 MB/s eta 0:00:14
----- 38.2/99.8 MB 4.5 MB/s eta 0:00:14
----- 38.2/99.8 MB 4.4 MB/s eta 0:00:14
----- 38.6/99.8 MB 4.4 MB/s eta 0:00:14
----- 38.9/99.8 MB 4.5 MB/s eta 0:00:14
----- 38.9/99.8 MB 4.5 MB/s eta 0:00:14
----- 39.2/99.8 MB 4.3 MB/s eta 0:00:14
----- 39.4/99.8 MB 4.4 MB/s eta 0:00:14
----- 39.5/99.8 MB 4.3 MB/s eta 0:00:14
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----- 40.6/99.8 MB 4.1 MB/s eta 0:00:15
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----- 41.4/99.8 MB 4.0 MB/s eta 0:00:15
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----- 41.5/99.8 MB 4.0 MB/s eta 0:00:15
----- 41.7/99.8 MB 3.9 MB/s eta 0:00:15

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----- 42.3/99.8 MB 4.0 MB/s eta 0:00:15
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----- 42.7/99.8 MB 3.9 MB/s eta 0:00:15
----- 42.7/99.8 MB 3.9 MB/s eta 0:00:15
----- 43.0/99.8 MB 3.9 MB/s eta 0:00:15
----- 43.3/99.8 MB 3.9 MB/s eta 0:00:15
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----- 43.6/99.8 MB 3.9 MB/s eta 0:00:15
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----- 44.5/99.8 MB 3.9 MB/s eta 0:00:15
----- 44.8/99.8 MB 4.1 MB/s eta 0:00:14
----- 44.8/99.8 MB 4.0 MB/s eta 0:00:14
----- 44.8/99.8 MB 4.0 MB/s eta 0:00:14
----- 44.9/99.8 MB 3.9 MB/s eta 0:00:14
----- 45.2/99.8 MB 3.9 MB/s eta 0:00:14
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----- 45.4/99.8 MB 3.8 MB/s eta 0:00:15
----- 45.6/99.8 MB 3.9 MB/s eta 0:00:15
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----- 46.4/99.8 MB 3.8 MB/s eta 0:00:15
----- 46.7/99.8 MB 3.8 MB/s eta 0:00:15
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----- 47.1/99.8 MB 3.9 MB/s eta 0:00:14
----- 47.4/99.8 MB 3.9 MB/s eta 0:00:14
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----- 47.7/99.8 MB 3.8 MB/s eta 0:00:14
----- 47.9/99.8 MB 3.9 MB/s eta 0:00:14
----- 48.0/99.8 MB 3.9 MB/s eta 0:00:14
----- 48.3/99.8 MB 3.9 MB/s eta 0:00:14
----- 48.3/99.8 MB 3.9 MB/s eta 0:00:14
----- 48.4/99.8 MB 3.8 MB/s eta 0:00:14
----- 48.7/99.8 MB 3.8 MB/s eta 0:00:14
----- 48.7/99.8 MB 3.7 MB/s eta 0:00:14
----- 49.0/99.8 MB 3.7 MB/s eta 0:00:14
----- 49.3/99.8 MB 3.8 MB/s eta 0:00:14
----- 49.3/99.8 MB 3.7 MB/s eta 0:00:14
----- 49.6/99.8 MB 3.7 MB/s eta 0:00:14
----- 50.0/99.8 MB 3.9 MB/s eta 0:00:13
----- 50.0/99.8 MB 3.9 MB/s eta 0:00:13
----- 50.3/99.8 MB 3.9 MB/s eta 0:00:13
----- 50.6/99.8 MB 3.9 MB/s eta 0:00:13
----- 50.8/99.8 MB 3.9 MB/s eta 0:00:13

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-----	-----	54.7/99.8	MB	4.0	MB/s	eta	0:00:12
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-----	-----	56.5/99.8	MB	4.5	MB/s	eta	0:00:10
-----	-----	56.7/99.8	MB	4.5	MB/s	eta	0:00:10
-----	-----	57.0/99.8	MB	4.5	MB/s	eta	0:00:10
-----	-----	57.3/99.8	MB	4.6	MB/s	eta	0:00:10
-----	-----	57.7/99.8	MB	4.6	MB/s	eta	0:00:10
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-----	-----	62.0/99.8	MB	5.6	MB/s	eta	0:00:07
-----	-----	62.3/99.8	MB	5.5	MB/s	eta	0:00:07
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-----	-----	62.8/99.8	MB	5.6	MB/s	eta	0:00:07

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-----	-----	63.7/99.8	MB	5.6	MB/s	eta	0:00:07
-----	-----	64.0/99.8	MB	5.6	MB/s	eta	0:00:07
-----	-----	64.3/99.8	MB	5.6	MB/s	eta	0:00:07
-----	-----	64.6/99.8	MB	5.7	MB/s	eta	0:00:07
-----	-----	64.9/99.8	MB	5.8	MB/s	eta	0:00:06
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-----	-----	65.9/99.8	MB	5.9	MB/s	eta	0:00:06
-----	-----	66.2/99.8	MB	5.8	MB/s	eta	0:00:06
-----	-----	66.5/99.8	MB	5.9	MB/s	eta	0:00:06
-----	-----	66.7/99.8	MB	5.9	MB/s	eta	0:00:06
-----	-----	66.8/99.8	MB	5.7	MB/s	eta	0:00:06
-----	-----	67.1/99.8	MB	5.8	MB/s	eta	0:00:06
-----	-----	67.3/99.8	MB	5.7	MB/s	eta	0:00:06
-----	-----	67.6/99.8	MB	5.7	MB/s	eta	0:00:06
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-----	-----	69.6/99.8	MB	6.0	MB/s	eta	0:00:06
-----	-----	70.0/99.8	MB	6.0	MB/s	eta	0:00:06
-----	-----	70.3/99.8	MB	6.0	MB/s	eta	0:00:05
-----	-----	70.6/99.8	MB	6.1	MB/s	eta	0:00:05
-----	-----	70.9/99.8	MB	6.1	MB/s	eta	0:00:05
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-----	-----	71.3/99.8	MB	6.1	MB/s	eta	0:00:05
-----	-----	71.5/99.8	MB	6.1	MB/s	eta	0:00:05
-----	-----	71.8/99.8	MB	6.0	MB/s	eta	0:00:05
-----	-----	72.1/99.8	MB	6.1	MB/s	eta	0:00:05
-----	-----	72.4/99.8	MB	6.1	MB/s	eta	0:00:05
-----	-----	72.7/99.8	MB	6.2	MB/s	eta	0:00:05
-----	-----	73.0/99.8	MB	6.2	MB/s	eta	0:00:05
-----	-----	73.3/99.8	MB	6.2	MB/s	eta	0:00:05
-----	-----	73.6/99.8	MB	6.2	MB/s	eta	0:00:05
-----	-----	73.9/99.8	MB	6.2	MB/s	eta	0:00:05
-----	-----	74.1/99.8	MB	6.2	MB/s	eta	0:00:05
-----	-----	74.4/99.8	MB	6.2	MB/s	eta	0:00:05
-----	-----	74.8/99.8	MB	6.2	MB/s	eta	0:00:05
-----	-----	75.1/99.8	MB	6.2	MB/s	eta	0:00:04
-----	-----	75.3/99.8	MB	6.2	MB/s	eta	0:00:04
-----	-----	75.5/99.8	MB	6.1	MB/s	eta	0:00:04
-----	-----	75.7/99.8	MB	6.1	MB/s	eta	0:00:04
-----	-----	75.9/99.8	MB	6.0	MB/s	eta	0:00:04
-----	-----	76.1/99.8	MB	6.0	MB/s	eta	0:00:04
-----	-----	76.5/99.8	MB	6.0	MB/s	eta	0:00:04

-----	76.7/99.8	MB	6.0	MB/s	eta	0:00:04
-----	77.1/99.8	MB	6.1	MB/s	eta	0:00:04
-----	77.4/99.8	MB	6.2	MB/s	eta	0:00:04
-----	77.7/99.8	MB	6.2	MB/s	eta	0:00:04
-----	78.0/99.8	MB	6.2	MB/s	eta	0:00:04
-----	78.3/99.8	MB	6.2	MB/s	eta	0:00:04
-----	78.6/99.8	MB	6.2	MB/s	eta	0:00:04
-----	79.0/99.8	MB	6.2	MB/s	eta	0:00:04
-----	79.3/99.8	MB	6.2	MB/s	eta	0:00:04
-----	79.6/99.8	MB	6.2	MB/s	eta	0:00:04
-----	79.9/99.8	MB	6.2	MB/s	eta	0:00:04
-----	80.2/99.8	MB	6.2	MB/s	eta	0:00:04
-----	80.4/99.8	MB	6.1	MB/s	eta	0:00:04
-----	80.6/99.8	MB	6.1	MB/s	eta	0:00:04
-----	80.8/99.8	MB	5.9	MB/s	eta	0:00:04
-----	81.0/99.8	MB	6.0	MB/s	eta	0:00:04
-----	81.4/99.8	MB	5.9	MB/s	eta	0:00:04
-----	81.7/99.8	MB	6.0	MB/s	eta	0:00:04
-----	82.0/99.8	MB	6.1	MB/s	eta	0:00:03
-----	82.3/99.8	MB	6.1	MB/s	eta	0:00:03
-----	82.6/99.8	MB	6.2	MB/s	eta	0:00:03
-----	82.9/99.8	MB	6.2	MB/s	eta	0:00:03
-----	83.1/99.8	MB	6.1	MB/s	eta	0:00:03
-----	83.5/99.8	MB	6.2	MB/s	eta	0:00:03
-----	83.8/99.8	MB	6.1	MB/s	eta	0:00:03
-----	84.1/99.8	MB	6.2	MB/s	eta	0:00:03
-----	84.4/99.8	MB	6.2	MB/s	eta	0:00:03
-----	84.6/99.8	MB	6.2	MB/s	eta	0:00:03
-----	84.9/99.8	MB	6.2	MB/s	eta	0:00:03
-----	85.1/99.8	MB	6.0	MB/s	eta	0:00:03
-----	85.2/99.8	MB	6.1	MB/s	eta	0:00:03
-----	85.5/99.8	MB	6.0	MB/s	eta	0:00:03
-----	85.8/99.8	MB	6.1	MB/s	eta	0:00:03
-----	86.2/99.8	MB	6.2	MB/s	eta	0:00:03
-----	86.5/99.8	MB	6.2	MB/s	eta	0:00:03
-----	86.7/99.8	MB	6.1	MB/s	eta	0:00:03
-----	87.0/99.8	MB	6.1	MB/s	eta	0:00:03
-----	87.3/99.8	MB	6.1	MB/s	eta	0:00:03
-----	87.5/99.8	MB	6.1	MB/s	eta	0:00:03
-----	87.9/99.8	MB	6.2	MB/s	eta	0:00:02
-----	88.3/99.8	MB	6.2	MB/s	eta	0:00:02
-----	88.6/99.8	MB	6.2	MB/s	eta	0:00:02
-----	88.7/99.8	MB	6.0	MB/s	eta	0:00:02
-----	89.0/99.8	MB	6.0	MB/s	eta	0:00:02
-----	89.2/99.8	MB	5.9	MB/s	eta	0:00:02
-----	89.5/99.8	MB	5.9	MB/s	eta	0:00:02
-----	89.8/99.8	MB	5.9	MB/s	eta	0:00:02
-----	90.1/99.8	MB	5.9	MB/s	eta	0:00:02

```

----- -- 90.4/99.8 MB 5.9 MB/s eta 0:00:02
----- -- 90.7/99.8 MB 6.1 MB/s eta 0:00:02
----- -- 91.0/99.8 MB 6.1 MB/s eta 0:00:02
----- -- 91.3/99.8 MB 6.1 MB/s eta 0:00:02
----- -- 91.6/99.8 MB 6.1 MB/s eta 0:00:02
----- -- 91.9/99.8 MB 6.1 MB/s eta 0:00:02
----- -- 92.3/99.8 MB 6.1 MB/s eta 0:00:02
----- -- 92.6/99.8 MB 6.1 MB/s eta 0:00:02
----- -- 92.9/99.8 MB 6.1 MB/s eta 0:00:02
----- -- 93.2/99.8 MB 6.1 MB/s eta 0:00:02
----- -- 93.4/99.8 MB 6.1 MB/s eta 0:00:02
----- -- 93.5/99.8 MB 6.0 MB/s eta 0:00:02
----- -- 93.6/99.8 MB 6.0 MB/s eta 0:00:02
----- -- 93.7/99.8 MB 5.8 MB/s eta 0:00:02
----- -- 93.9/99.8 MB 5.7 MB/s eta 0:00:02
----- -- 94.2/99.8 MB 5.7 MB/s eta 0:00:01
----- -- 94.6/99.8 MB 5.7 MB/s eta 0:00:01
----- - 94.9/99.8 MB 5.7 MB/s eta 0:00:01
----- - 95.2/99.8 MB 5.8 MB/s eta 0:00:01
----- - 95.5/99.8 MB 6.0 MB/s eta 0:00:01
----- - 95.9/99.8 MB 5.9 MB/s eta 0:00:01
----- - 96.2/99.8 MB 5.9 MB/s eta 0:00:01
----- - 96.4/99.8 MB 5.9 MB/s eta 0:00:01
----- - 96.7/99.8 MB 6.0 MB/s eta 0:00:01
----- - 97.0/99.8 MB 6.0 MB/s eta 0:00:01
----- 97.4/99.8 MB 6.0 MB/s eta 0:00:01
----- 97.5/99.8 MB 5.8 MB/s eta 0:00:01
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----- 98.2/99.8 MB 5.8 MB/s eta 0:00:01
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----- 99.1/99.8 MB 5.8 MB/s eta 0:00:01
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----- 99.7/99.8 MB 5.7 MB/s eta 0:00:01
----- 99.8/99.8 MB 4.7 MB/s eta 0:00:00

```

Installing collected packages: xgboost
Successfully installed xgboost-2.0.3

```
[48]: # apply xgboost
import xgboost as xgb
model=xgb.XGBClassifier()
model.fit(X_train_smote,y_train_smote)
```

```
[48]: XGBClassifier(base_score=None, booster=None, callbacks=None,
                  colsample_bylevel=None, colsample_bynode=None,
                  colsample_bytree=None, device=None, early_stopping_rounds=None,
                  enable_categorical=False, eval_metric=None, feature_types=None,
                  gamma=None, grow_policy=None, importance_type=None,
                  interaction_constraints=None, learning_rate=None, max_bin=None,
                  max_cat_threshold=None, max_cat_to_onehot=None,
                  max_delta_step=None, max_depth=None, max_leaves=None,
                  min_child_weight=None, missing=nan, monotone_constraints=None,
                  multi_strategy=None, n_estimators=None, n_jobs=None,
                  num_parallel_tree=None, random_state=None, ...)
```

```
[49]: y_pred3=model.predict(X_test)
```

```
[50]: print('Accuracy score',accuracy_score(y_test,y_pred3))
```

Accuracy score 0.983

```
[ ]: Suggest various retention strategies for targeted employees
7.1. Using the best model, predict the probability of employee turnover in the
    ↳test data
7.2. Based on the probability score range below, categorize the employees into
    ↳four zones and suggest your thoughts on
the retention strategies for each zone
[ ] Safe Zone (Green) (Score < 20%)
[ ] Low Risk Zone (Yellow) (20% < Score < 60%)
[ ] Medium Risk Zone (Orange) (60% < Score < 90%)
[ ] High Risk Zone (Red) (Score > 90%).
```

```
[51]: predicted_prob=model.predict_proba(X_test)
```

```
[52]: predicted_prob[:,1]
```

```
[52]: array([2.0239866e-04, 8.9765954e-01, 3.8354285e-02, ..., 1.6143646e-02,
          1.4798146e-03, 9.8090887e-01], dtype=float32)
```

```
[53]: zone=[]
prob=[]

for i in predicted_prob[:,1]:
    prob.append(i)
    if (i<=0.2):
        zone.append("Safe Zone")
```

```

elif (i>0.2 and i<=0.6):
    zone.append("Low Risk Zone")
elif (i>0.6 and i<=0.9):
    zone.append("Medium Risk Zone ")
else:
    zone.append("High Risk Zone ")

```

```

[54]: categories = ["Safe Zone","Low Risk Zone","Medium Risk Zone ","High Risk Zone "]
      color = ["blue","green","yellow","Red"]

```

```

[56]: colordict = dict(zip(categories, color))
      clr = pd.DataFrame({"zone":zone,"probability":prob})
      clr["Color"] = clr["zone"].apply(lambda x: colordict[x])

```

```

[57]: clr['zone'] = clr['zone'].astype(str)

```

```

[59]: color= clr["Color"].tolist()
      c = ["blue","Red","Orange","Yellow"]

```

```

[60]: plt.figure(figsize=(7,7))
      sns.countplot(x=zone,palette=c)

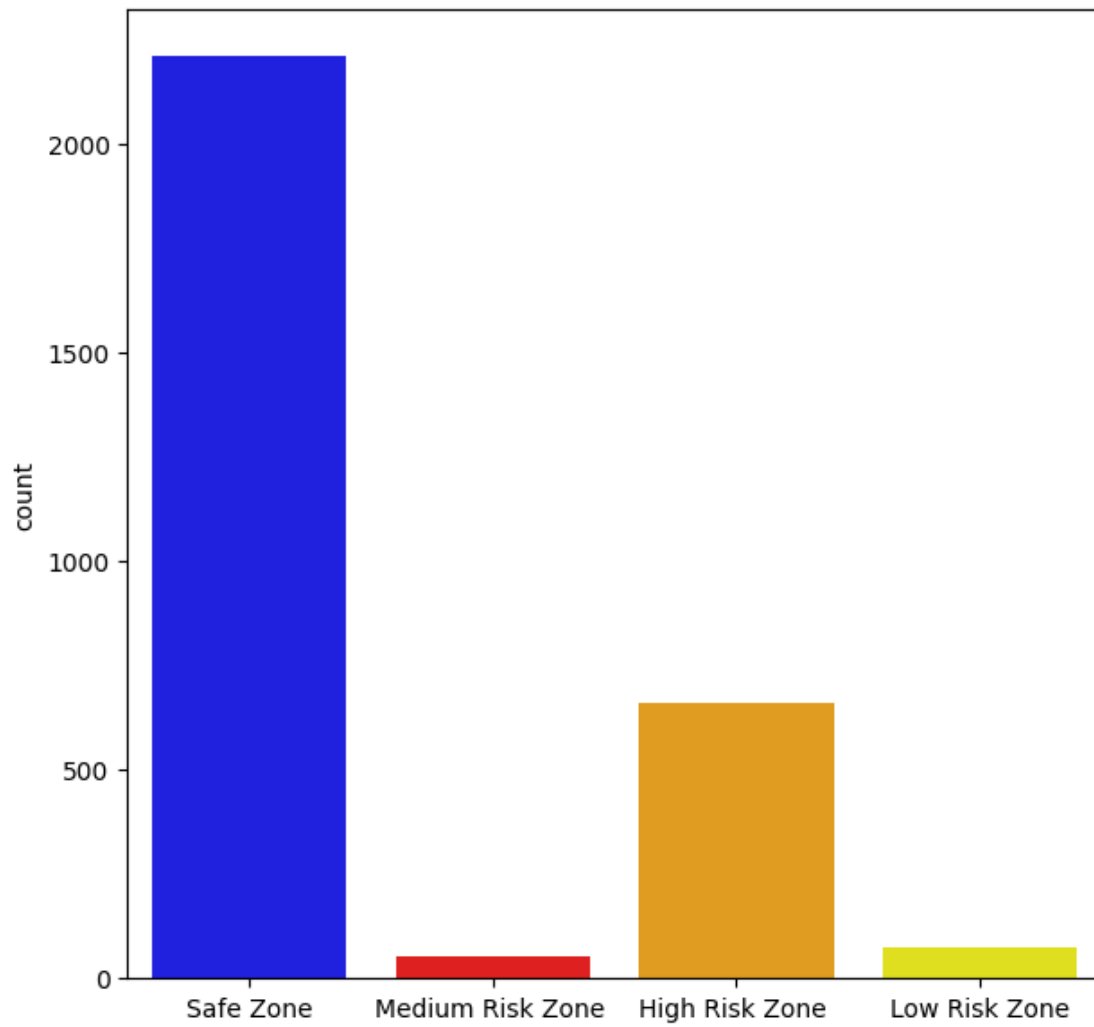
```

C:\Users\priya\.jupyter\New folder\Lib\site-packages\seaborn_oldcore.py:1765:
FutureWarning: unique with argument that is not not a Series, Index,
ExtensionArray, or np.ndarray is deprecated and will raise in a future version.
order = pd.unique(vector)

```

[60]: <Axes: ylabel='count'>

```

```
[61]: data
```

```
[61]:
```

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

	average_monthly_hours	time_spend_company	Work_accident	left	\
0	157	3	0	1	
1	262	6	0	1	
2	272	4	0	1	
3	223	5	0	1	
4	159	3	0	1	
...	
14994	151	3	0	1	
14995	160	3	0	1	
14996	143	3	0	1	
14997	280	4	0	1	
14998	158	3	0	1	

	promotion_last_5years	sales	salary
0	0	sales	low
1	0	sales	medium
2	0	sales	medium
3	0	sales	low
4	0	sales	low
...
14994	0	support	low
14995	0	support	low
14996	0	support	low
14997	0	support	low
14998	0	support	low

[14999 rows x 10 columns]

[]: