```
#Importing Libraries and Load Data
In [31]:
         import pandas as pd
         import numpy as np
         import matplotlib.pyplot as plt
         import seaborn as sns
         from sklearn.model selection import train test split
         from sklearn.ensemble import RandomForestClassifier
         from sklearn.metrics import classification_report, confusion_matrix
         from sklearn.preprocessing import LabelEncoder
         import joblib
In [33]:
         # Define the path to the Excel file
         file_path = r"C:\Churn Data Analysis\Data\Prediction Data.xlsx"
         # Define the sheet name to read data from
         sheet name = 'vw ChurnData'
         # Read the data from the specified sheet into a pandas DataFrame
         data = pd.read excel(file path, sheet name=sheet name)
          # Display the first few rows of the fetched data
         data.head()
```

Out[33]:		Customer_ID	Gender	Age	Married	State	Number_of_Referrals	Tenure_in_Months	Value_D
	0	11098-MAD	Female	30	Yes	Madhya Pradesh	0	31	Dea
	1	11114-PUN	Male	51	No	Punjab	5	9	Dea
	2	11167-WES	Female	43	Yes	West Bengal	3	28	De
	3	11179-MAH	Male	35	No	Maharashtra	10	12	Ν
	4	11180-TAM	Male	75	Yes	Tamil Nadu	12	27	Dea

5 rows × 32 columns

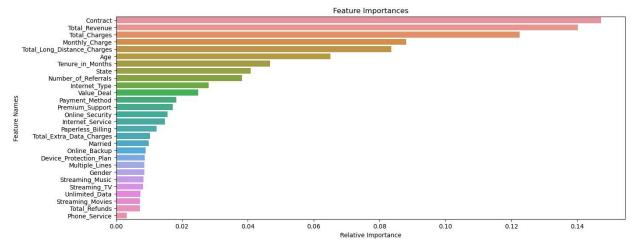
```
In [34]: # Data Preprocessing

# Dropping columns that won't be used for prediction
data = data.drop(['Customer_ID', 'Churn_Category', 'Churn_Reason'], axis=1)

# List of columns to be label encoded
columns_to_encode = [
    'Gender', 'Married', 'State', 'Value_Deal', 'Phone_Service', 'Multiple_Lines',
    'Internet_Service', 'Internet_Type', 'Online_Security', 'Online_Backup',
    'Device_Protection_Plan', 'Premium_Support', 'Streaming_TV', 'Streaming_Movies',
    'Streaming_Music', 'Unlimited_Data', 'Contract', 'Paperless_Billing',
    'Payment_Method'
]

# Encoding categorical variables except the target variable
label_encoders = {}
```

```
for column in columns to encode:
             label encoders[column] = LabelEncoder()
             data[column] = label encoders[column].fit transform(data[column])
          # Encoding the target variable 'Customer_Status'
         data['Customer Status'] = data['Customer Status'].map({'Stayed': 0, 'Churned': 1})
        # Split data into features and target
In [35]:
         X = data.drop('Customer Status', axis=1)
         y = data['Customer_Status']
         # Split data into training and testing sets
         X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.2, random_state=
In [36]: # Initialize the Random Forest Classifier
         rf_model = RandomForestClassifier(n_estimators=100, random_state=42)
         # Train the model
         rf_model.fit(X_train, y_train)
         # Make predictions
         y pred = rf model.predict(X test)
         # Evaluate the model
         print("Confusion Matrix:")
         print(confusion matrix(y test, y pred))
         print("\nClassification Report:")
         print(classification_report(y_test, y_pred))
         # Feature Selection using Feature Importance
         importances = rf_model.feature_importances_
         indices = np.argsort(importances)[::-1]
         # Plot the feature importances
         plt.figure(figsize=(15, 6))
         sns.barplot(x=importances[indices], y=X.columns[indices])
         plt.title('Feature Importances')
         plt.xlabel('Relative Importance')
         plt.ylabel('Feature Names')
         plt.show()
         Confusion Matrix:
         [[783 64]
          [126 229]]
         Classification Report:
                       precision
                                   recall f1-score
                                                        support
                    0
                            0.86
                                      0.92
                                                 0.89
                                                            847
                    1
                            0.78
                                      0.65
                                                 0.71
                                                            355
                                                 0.84
                                                           1202
             accuracy
                                                 0.80
            macro avg
                            0.82
                                      0.78
                                                           1202
                            0.84
                                      0.84
                                                0.84
                                                           1202
         weighted avg
```



## Use Model on New Data

```
# Path to the Joiner Data
In [38]:
         file path = r"C:\Churn Data Analysis\Data\Prediction Data.xlsx"
         sheet name = 'vw JoinedData'
         new_data = pd.read_excel(file_path, sheet_name=sheet_name)
         print(new_data.head())
         # Retain the original Data
         original data = new data.copy()
         customer_ids = new_data['Customer_ID']
         # Data Preprocessing
         new_data = new_data.drop(['Customer_ID', 'Customer_Status', 'Churn_Category', 'Churn_F
         # Encoding
         for column in new_data.select_dtypes(include=['object']).columns:
             new_data[column] = label_encoders[column].transform(new_data[column])
         # Making predictions
         new_predictions = rf_model.predict(new_data)
         # Adding predictions to the original DataFrame
         original_data['Customer_Status_Predicted'] = new_predictions
          # Filter the DataFrame to include only records predicted as "Churned"
         original data = original data[original data['Customer Status Predicted'] == 1]
          # Save the results
         original data.to csv(r"C:\Churn Data Analysis\Data\Predictions.csv", index=False)
```

```
Customer ID
               Gender
                        Age Married
                                             State
                                                    Number of Referrals
    11751-TAM
               Female
                         18
                                       Tamil Nadu
                                                                       5
0
                                  No
                                                                       2
1
    12056-WES
                  Male
                         27
                                  No
                                      West Bengal
2
    12136-RAJ
               Female
                                 Yes
                                        Rajasthan
                                                                       2
3
                                  No
                                                                       9
    12257-ASS
               Female
                         39
                                             Assam
4
    12340-DEL
               Female
                         51
                                 Yes
                                            Delhi
                                                                       0
   Tenure_in_Months Value_Deal Phone_Service Multiple_Lines
0
                   7
                         Deal 5
                                                             No
1
                  20
                            NaN
                                           Yes
                                                             No
2
                  35
                            NaN
                                           Yes
                                                            No
3
                   1
                            NaN
                                           Yes
                                                             No
4
                  10
                            NaN
                                           Yes
                                                             No
    Payment_Method Monthly_Charge Total_Charges Total_Refunds
      Mailed Check
0
                         24.299999
                                        38.450001
                                                              0.0
1
   Bank Withdrawal
                         90.400002
                                       268.450012
                                                              0.0
2
   Bank Withdrawal
                                                              0.0
                         19.900000
                                        19.900000
3
       Credit Card
                         19.549999
                                        19.549999
                                                              0.0
4
       Credit Card
                         62.799999
                                        62.799999
                                                              0.0
  Total_Extra_Data_Charges Total_Long_Distance_Charges Total_Revenue \
0
                                                 0.000000
                                                               38.450001
1
                          0
                                                94.440002
                                                              362.890015
2
                          0
                                                11.830000
                                                               31.730000
3
                          0
                                                10.200000
                                                               29.750000
4
                          0
                                                42.189999
                                                              104.989998
  Customer_Status Churn_Category Churn_Reason
0
           Joined
                           Others
                                         Others
1
           Joined
                           Others
                                         Others
2
           Joined
                           Others
                                         Others
3
           Joined
                           Others
                                         Others
4
           Joined
                           Others
                                         Others
```

[5 rows x 32 columns]

In [ ]: