

```
Import pandas as pd
```

```
Import numpy as np
```

```
From sklearn.model_selection import train_test_split
```

```
From sklearn.preprocessing import LabelEncoder, StandardScaler
```

```
From sklearn.ensemble import RandomForestClassifier
```

```
From sklearn.metrics import classification_report, confusion_matrix
```

```
Import seaborn as sns
```

```
Import matplotlib.pyplot as plt
```

```
# 1. Create a synthetic dataset
```

```
Data = {
```

```
    "gender": ["Female", "Male", "Female", "Male", "Female", "Male", "Male", "Female",  
    "Female", "Male"],
```

```
    "SeniorCitizen": [0, 1, 0, 1, 0, 0, 1, 1, 0, 0],
```

```
    "Partner": ["Yes", "No", "No", "Yes", "No", "Yes", "Yes", "No", "No", "Yes"],
```

```
    "Dependents": ["No", "No", "Yes", "No", "Yes", "No", "Yes", "No", "Yes", "No"],
```

```
    "tenure": [1, 34, 12, 45, 5, 22, 10, 3, 18, 6],
```

```
    "PhoneService": ["No", "Yes", "Yes", "Yes", "No", "Yes", "No", "Yes", "Yes", "Yes"],
```

```
    "MultipleLines": ["No phone service", "No", "Yes", "Yes", "No phone service", "No",  
    "No phone service", "Yes", "Yes", "No"],
```

```
    "InternetService": ["DSL", "DSL", "Fiber optic", "Fiber optic", "DSL", "DSL", "DSL",  
    "Fiber optic", "DSL", "Fiber optic"],
```

```
    "OnlineSecurity": ["No", "Yes", "No", "Yes", "No", "No", "Yes", "No", "Yes", "No"],
```

```
    "OnlineBackup": ["Yes", "No", "Yes", "No", "Yes", "No", "Yes", "Yes", "No", "Yes"],
```

```
    "DeviceProtection": ["No", "Yes", "Yes", "No", "Yes", "Yes", "No", "No", "Yes", "Yes"],
```

```

    "TechSupport": ["No", "No", "Yes", "Yes", "No", "Yes", "Yes", "No", "No", "Yes"],
    "StreamingTV": ["No", "Yes", "Yes", "No", "No", "Yes", "No", "Yes", "Yes", "No"],
    "StreamingMovies": ["No", "Yes", "Yes", "No", "No", "Yes", "No", "Yes", "Yes", "Yes"],
    "Contract": ["Month-to-month", "One year", "Month-to-month", "Two year",
    "Month-to-month", "One year", "Two year", "Month-to-month", "One year", "Month-
    to-month"],
    "PaperlessBilling": ["Yes", "No", "Yes", "No", "Yes", "Yes", "No", "Yes", "No", "Yes"],
    "PaymentMethod": ["Electronic check", "Mailed check", "Bank transfer (automatic)",
    "Credit card (automatic)", "Electronic check",
        "Bank transfer (automatic)", "Mailed check", "Credit card (automatic)",
    "Mailed check", "Electronic check"],
    "MonthlyCharges": [29.85, 56.95, 53.85, 42.30, 70.70, 49.95, 30.00, 80.00,
    59.90, 65.25],
    "TotalCharges": [29.85, 1889.50, 646.00, 1840.75, 151.65, 1098.45, 300.00,
    240.00, 1078.20, 391.50],
    "Churn": ["No", "No", "Yes", "No", "Yes", "No", "Yes", "Yes", "No", "Yes"]
}

```

```
Df = pd.DataFrame(data)
```

```
# 2. Encode categorical variables
```

```
Le = LabelEncoder()
```

```
For col in df.select_dtypes(include='object').columns:
```

```
    Df[col] = le.fit_transform(df[col])
```

```
# 3. Scale numeric columns
```

```
Scaler = StandardScaler()
```

```
Df[['tenure', 'MonthlyCharges', 'TotalCharges']] = scaler.fit_transform(df[['tenure',  
'MonthlyCharges', 'TotalCharges']])
```

```
# 4. Split data
```

```
X = df.drop('Churn', axis=1)
```

```
Y = df['Churn']
```

```
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.2, random_state=42)
```

```
# 5. Train model
```

```
Model = RandomForestClassifier(n_estimators=100, random_state=42)
```

```
Model.fit(X_train, y_train)
```

```
# 6. Predict and evaluate
```

```
Y_pred = model.predict(X_test)
```

```
Print("Classification Report:\n", classification_report(y_test, y_pred))
```

```
Print("Confusion Matrix:\n", confusion_matrix(y_test, y_pred))
```

```
# 7. Feature Importance
```

```
Importances = model.feature_importances_
```

```
Features = X.columns
```

```
Sns.barplot(x=importances, y=features)
```

```
Plt.title("Feature Importances")
```

```
Plt.tight_layout()
```

```
Plt.show()
```

## Output

Usr/local/lib/python3.11/dist-packages/sklearn/metrics/\_classification.py:1565:  
UndefinedMetricWarning: Recall is ill-defined and being set to 0.0 in labels with no  
true samples. Use `zero\_division` parameter to control this behavior.

```
_warn_prf(average, modifier, f"{metric.capitalize()} is", len(result))
```

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```

### Classification Report:

	Precision	recall	f1-score	support
0	1.00	0.50	0.67	2
1	0.00	0.00	0.00	0
Accuracy			0.50	2
Macro avg	0.50	0.25	0.33	2
Weighted avg	1.00	0.50	0.67	2

Confusion Matrix:

[[1 1]

[0 0]]



