

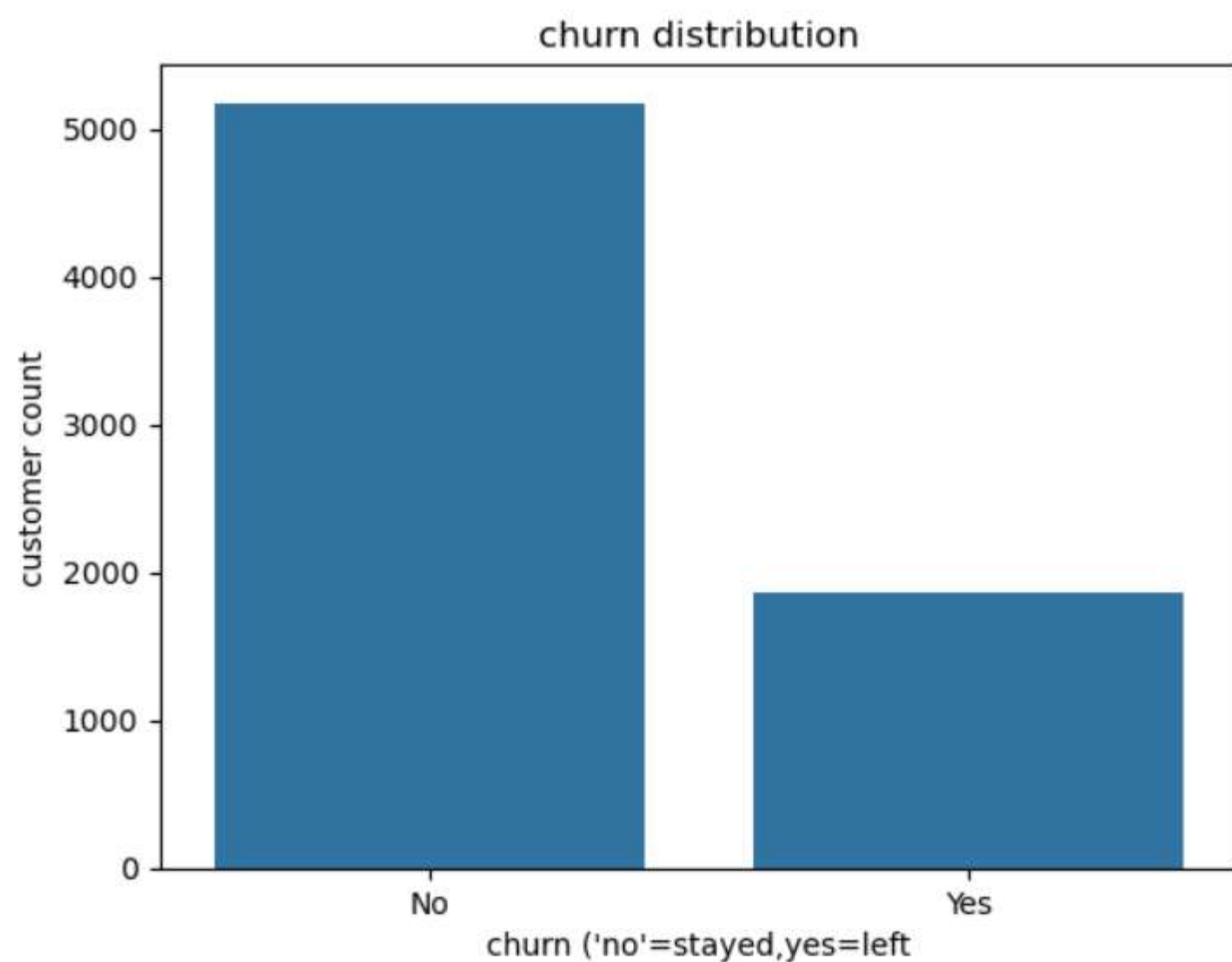
```

sns.countplot(x='Churn',data=df)
plt.title("churn distribution")
plt.xlabel("churn ('no'=stayed,yes=left")
plt.ylabel("customer count")

plt.show()

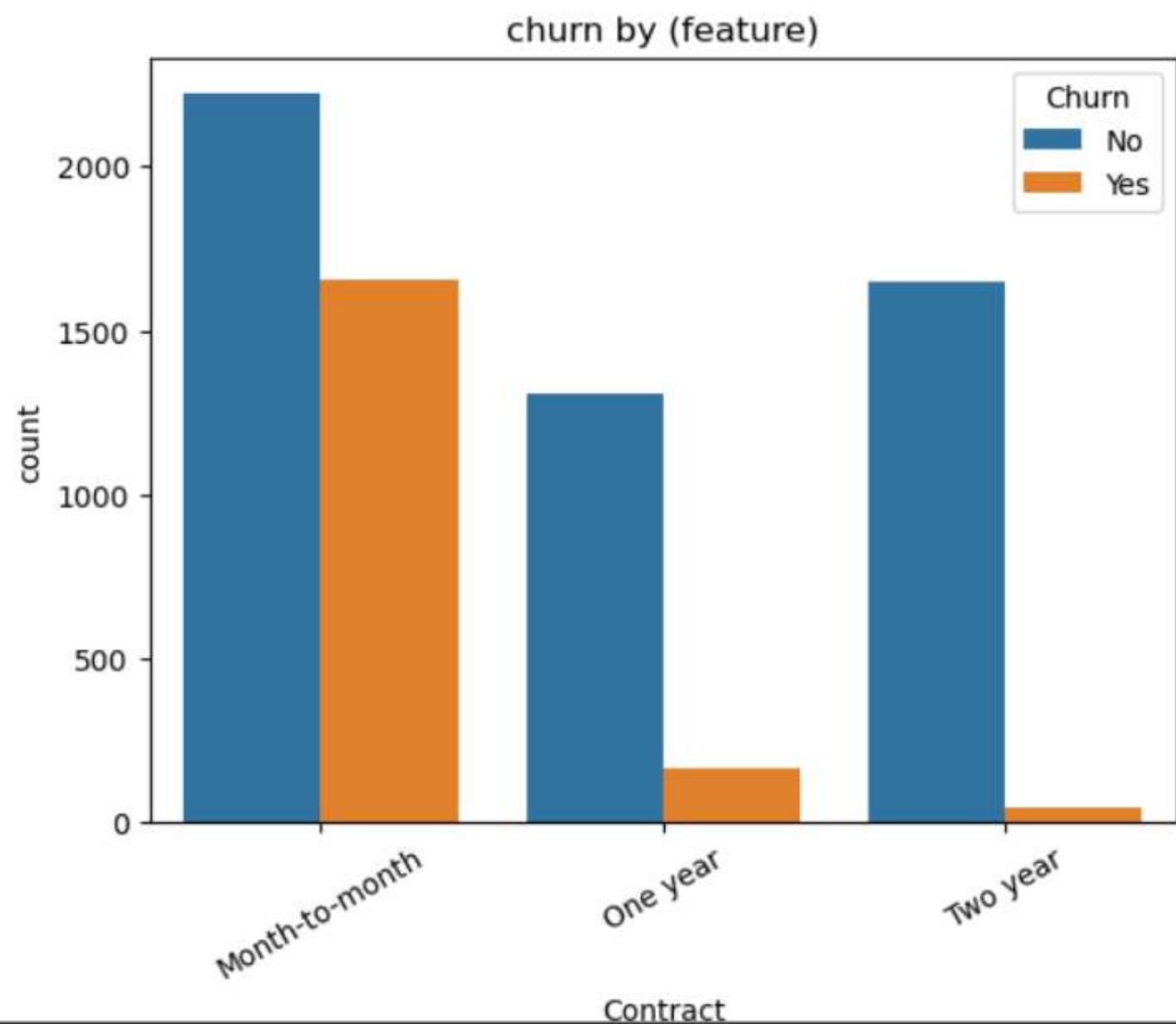
churn_rate=df['Churn'].value_counts(normalize=True)*100
churn_rate

```

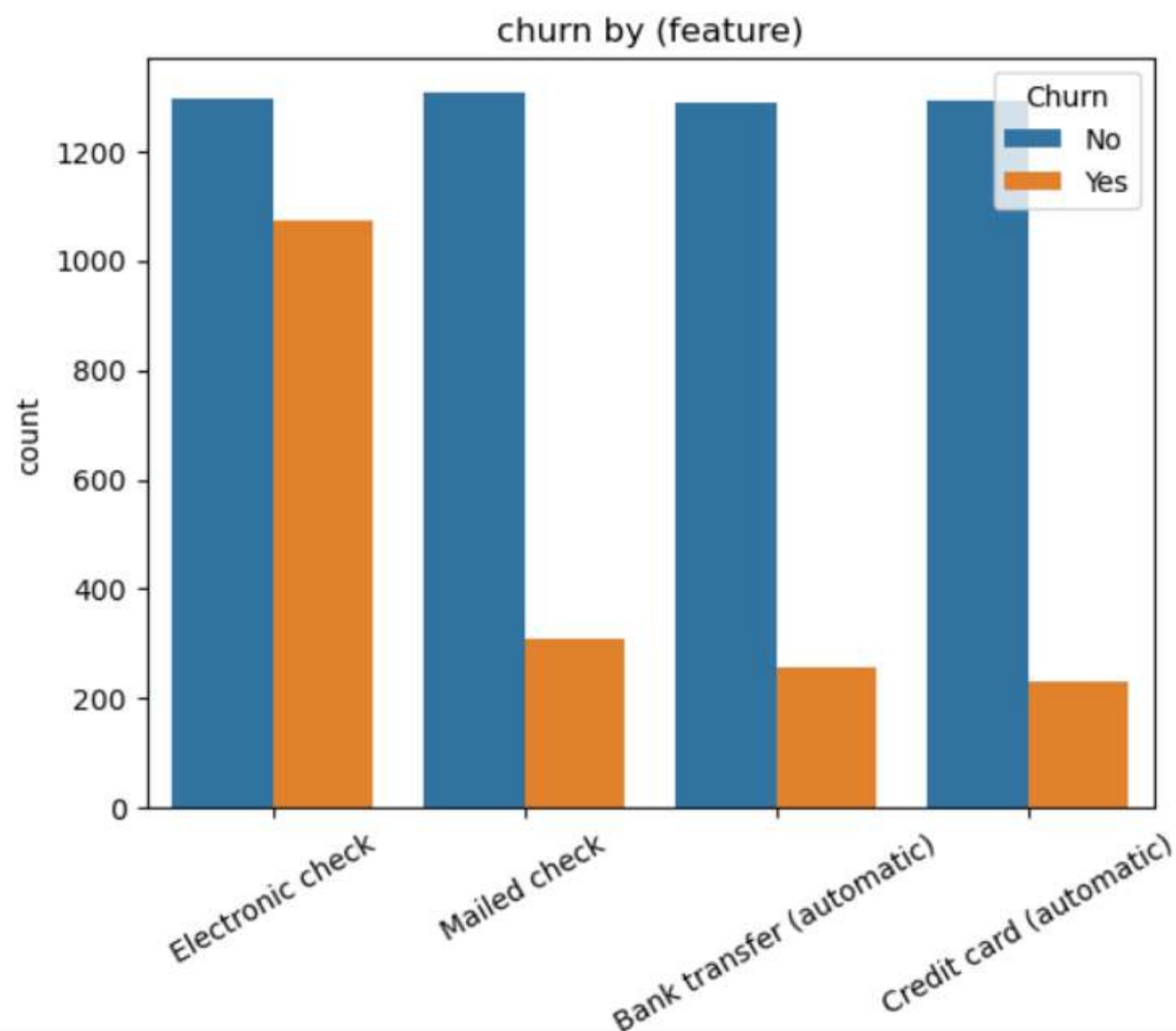


Churn	
No	73.463013
Yes	26.536987

```
feature='Contract'  
sns.countplot(x=feature,hue='Churn',data=df)  
plt.title('churn by (feature)')  
plt.xticks(rotation=30)  
plt.show()
```

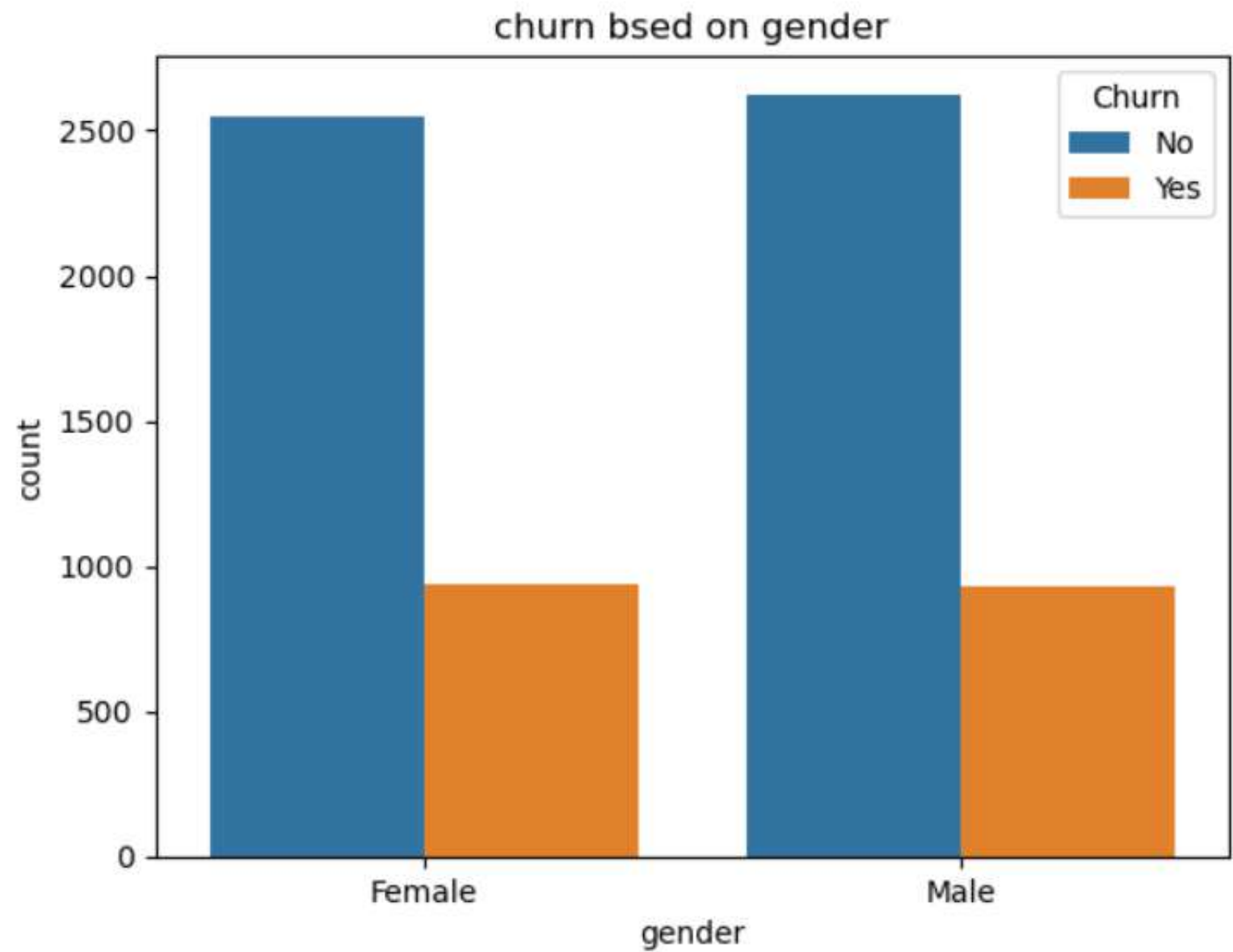


```
sns.countplot(x='PaymentMethod',hue='Churn',data=df)
plt.title('churn by (feature)')
plt.xticks(rotation=30)
plt.show()
pd.crosstab(df['PaymentMethod'],df['Churn'],normalize=True)*100
```



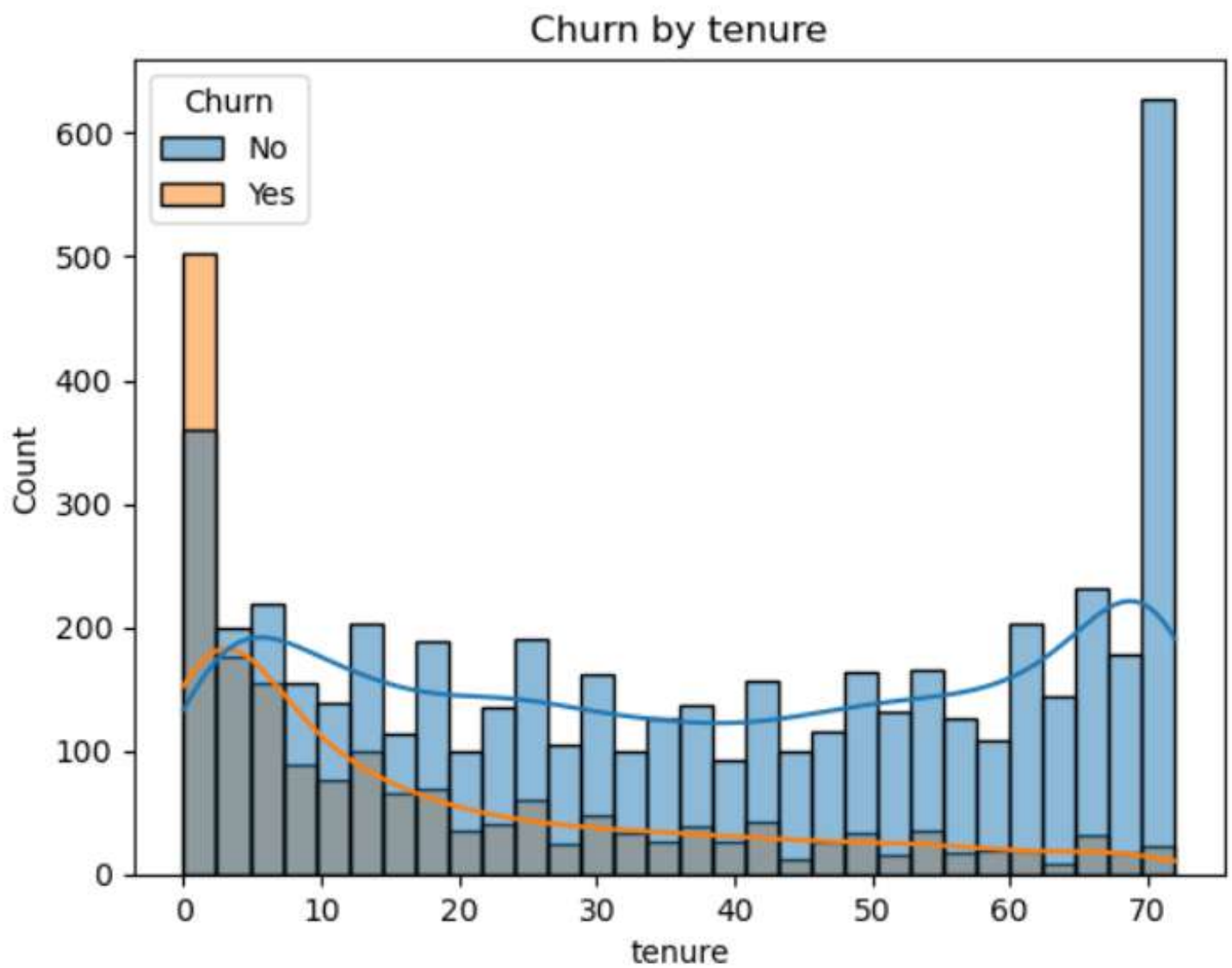
```
sns.countplot(x="gender", hue="Churn",data=df)
plt.title("churn bsd on gender")
plt.show()
```

```
pd.crosstab(df['gender'],df['Churn'])
```

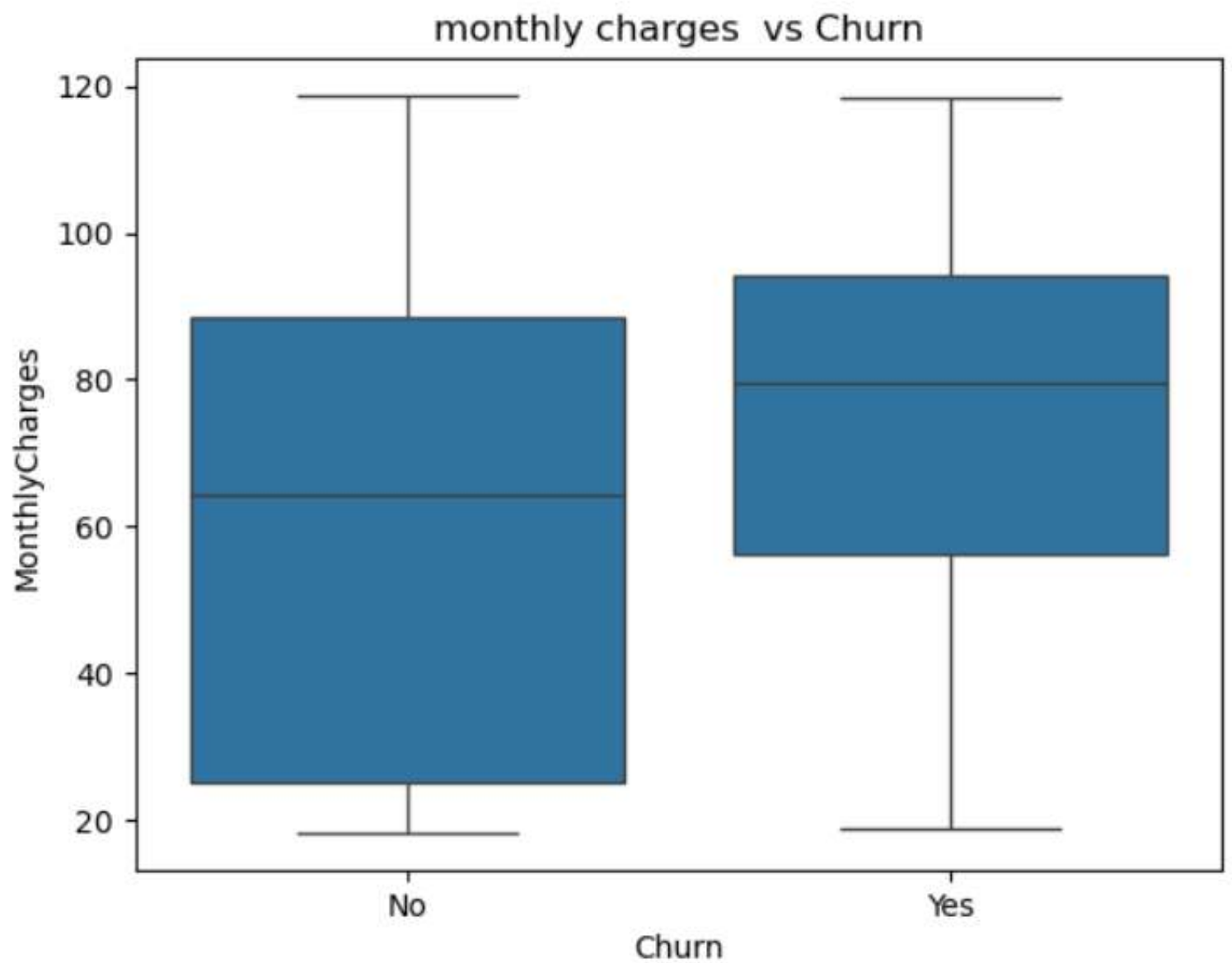


Churn	No	Yes
gender		
Female	2549	939
Male	2625	930

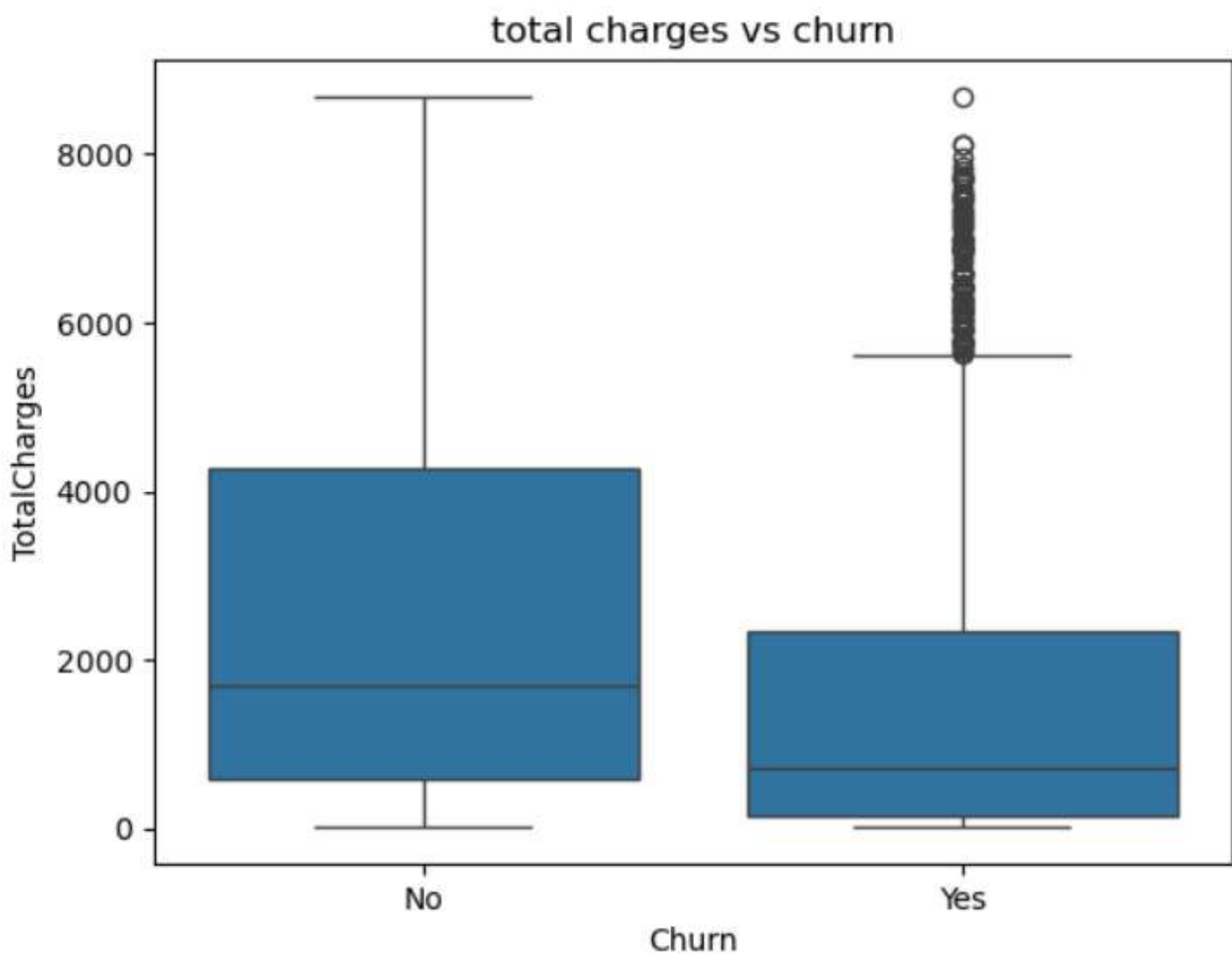
```
sns.histplot(data=df,x='tenure',hue='Churn',bins=30,kde=True)
plt.title("Churn by tenure")
plt.show()
```



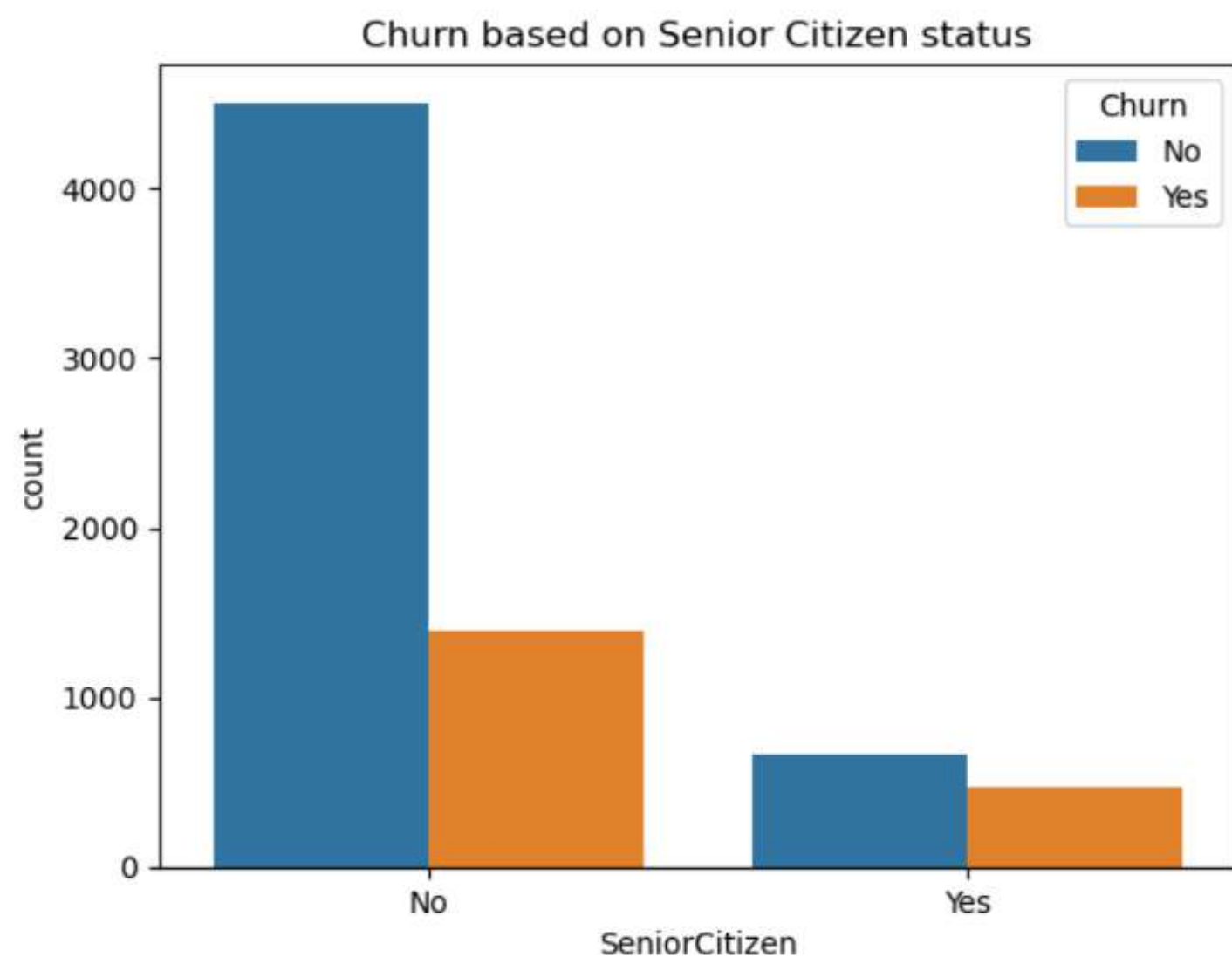
```
sns.boxplot(x='Churn',y='MonthlyCharges',data=df)
plt.title("monthly charges vs Churn")
plt.show()
```



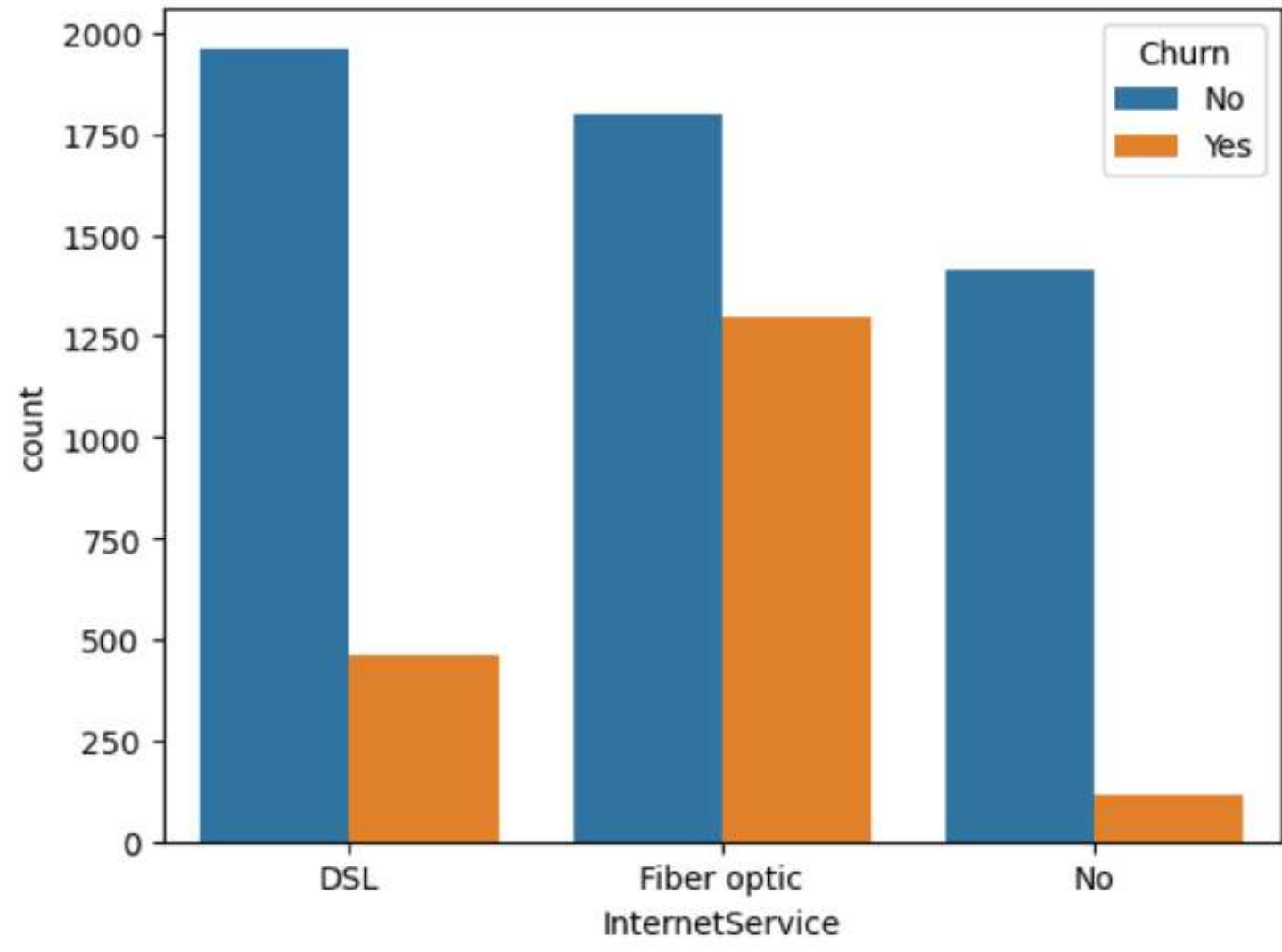
```
sns.boxplot(x='Churn',y='TotalCharges',data=df)
plt.title('total charges vs churn')
plt.show()
```



```
sns.countplot(x="SeniorCitizen", hue='Churn', data=df)
plt.title("Churn based on Senior Citizen status")
plt.show()
```




```
sns.countplot(hue='Churn',x='InternetService',data =df)
plt.show()
```



```

from sklearn.metrics import classification_report, confusion_matrix, roc_auc_score

y_pred=model.predict(x_test)
y_proba=model.predict_proba(x_test)[:,-1]

print("Classification Report:")
print(classification_report(y_test,y_pred))

print("\n confusion matrix:")
print(confusion_matrix(y_test,y_pred))

print("\n ROC AUC Score")
print(roc_auc_score(y_test,y_proba))

```

```

Classification Report:

```

	precision	recall	f1-score	support
0	0.85	0.90	0.87	1035
1	0.66	0.56	0.60	374
accuracy			0.81	1409
macro avg	0.75	0.73	0.74	1409
weighted avg	0.80	0.81	0.80	1409

```

confusion matrix:
[[927 108]
 [166 208]]

```

```

ROC AUC Score
0.8421349040274871

```

```

from sklearn.metrics import classification_report, confusion_matrix, roc_auc_score
rf_pred=rf_model.predict(x_test)
rf_proba=rf_model.predict_proba(x_test)[:,-1]

print("Classification report (Random Forest):")
print(classification_report(y_test, rf_pred))

print("Confusion Matrix:")
print(confusion_matrix(y_test, rf_pred))

print("ROC AUC score:")
print(roc_auc_score(y_test, rf_proba))

```

```

Classification report (Random Forest):

```

	precision	recall	f1-score	support
0	0.84	0.91	0.87	1035
1	0.67	0.52	0.58	374
accuracy			0.80	1409
macro avg	0.75	0.71	0.73	1409
weighted avg	0.79	0.80	0.80	1409

```

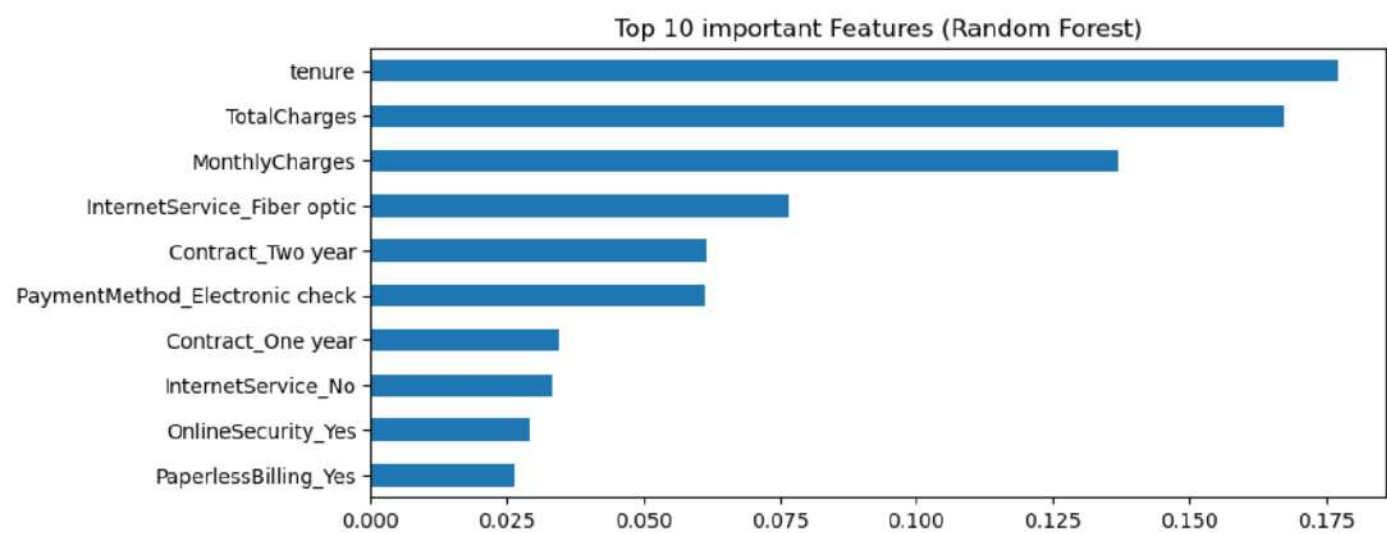
Confusion Matrix:
[[939  96]
 [180 194]]
ROC AUC score:
0.8438528507582216

```

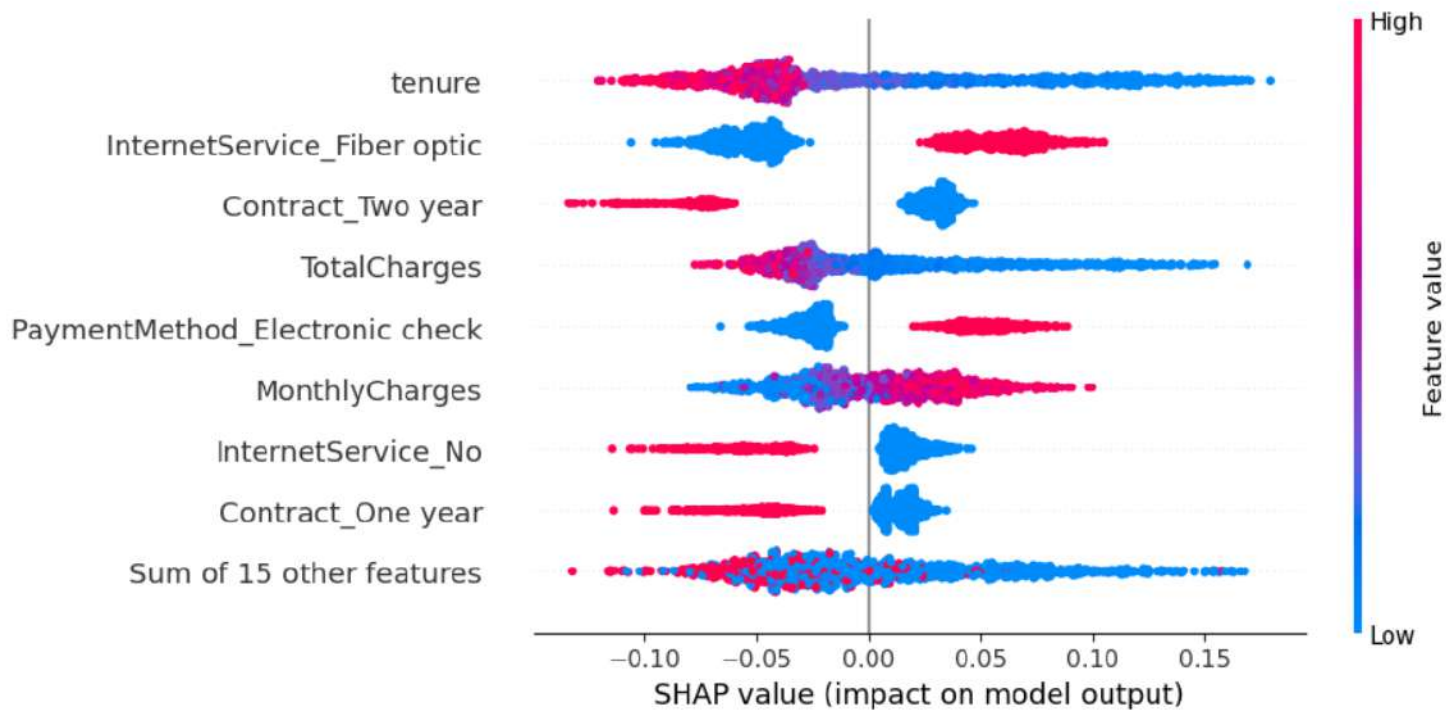
```
import pandas as pd
import matplotlib.pyplot as plt

importances=rf_model.feature_importances_
features=pd.Series(importances,index=x_train.columns)

features.sort_values(ascending=False).head(10).plot(kind='barh',figsize=(9,4))
plt.title("Top 10 important Features (Random Forest)")
plt.gca().invert_yaxis()
plt.show()
```



```
shap_values_class1=shap_values[...,1]  
  
shap.plots.beeswarm(shap_values_class1,max_display=9)
```



Quick Business Metrics

Total Customers

7043

Predicted Loss

4.27M

Payment Method

All



Churn AVG

1.87K

Monthly charges AVG

64.76

Root Cause Analysis:

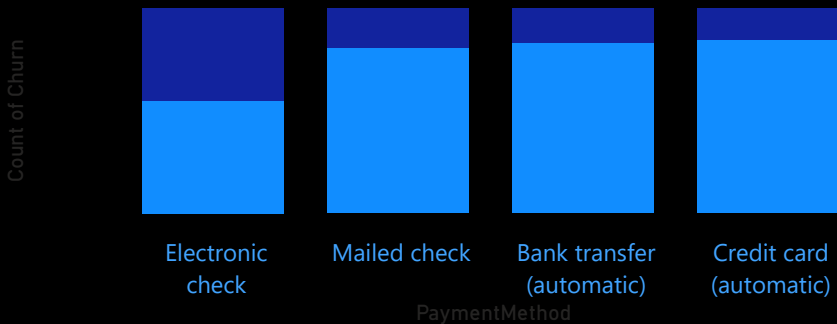
Churn by Contract Type

Churn ● No ● Yes



Count of Churn by PaymentMethod and Churn

Churn ● No ● Yes



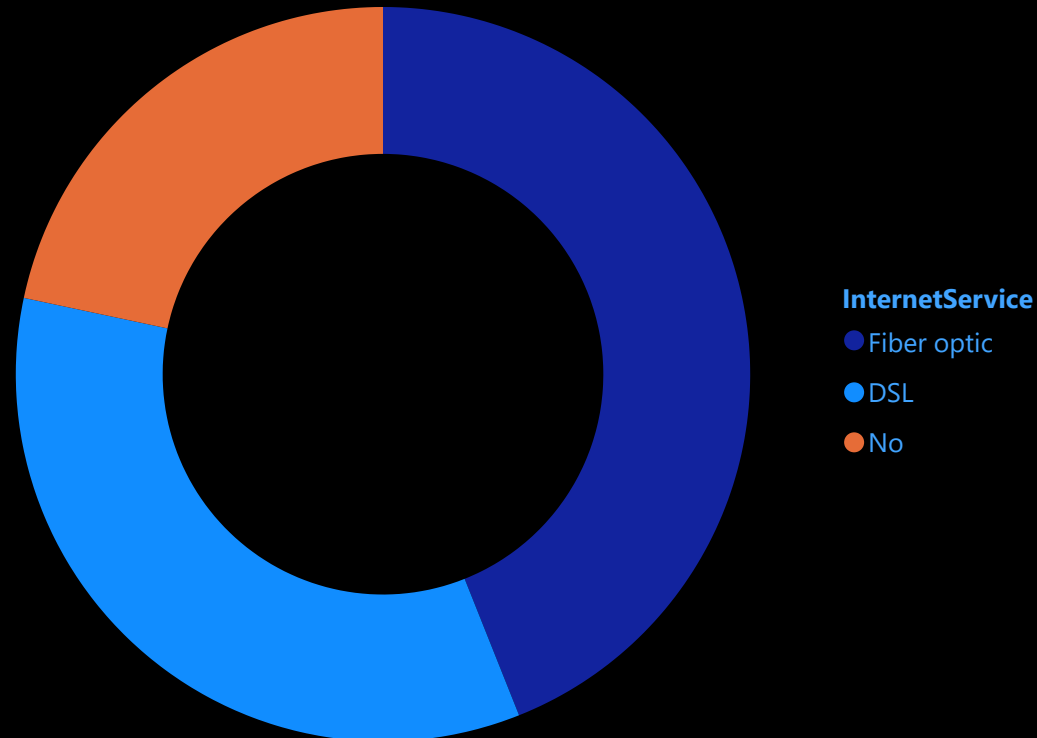
Senior Citizen

Tenure

All

All

contribution of churn by internetservices



SOLUTION SUMMARY

Conversion rate

0.60



Tenure

12



switch to credit card

0.50

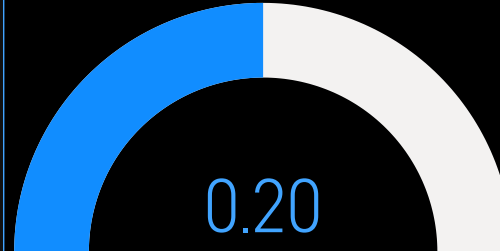


Churn Rate



0.27

Simulated churn rate



0.20