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# CUSTOMER CHURN PROJECT – OUTPUT REPORT
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```
## 1. Data Preprocessing Output
```

```
Loaded dataset with 7,043 rows and 21 columns.
```

```
After cleaning missing values, dataset is ready.
```

```
Sample rows:
```

```
CustomerID: 7590-VHVEG
```

```
Gender: Female
```

```
Tenure: 1
```

```
MonthlyCharges: 29.85
```

```
Churn: No
```

```
CustomerID: 5575-GNVDE
```

```
Gender: Male
```

```
Tenure: 34
```

```
MonthlyCharges: 56.95
```

```
Churn: No
```

```
## 2. Model Training Output
```

```
RandomForestClassifier trained successfully.
```

```
Model saved as: churn_model.pkl
```

```
Training Accuracy: 0.80
```

```
Validation Accuracy: 0.79
```

```
## 3. Model Evaluation Output
```

```
Accuracy on test set: 0.79
```

```
Classification Report:
```

Precision (Churn = Yes): 0.73

Recall (Churn = Yes): 0.68

F1 Score (Churn = Yes): 0.70

Precision (Churn = No): 0.84

Recall (Churn = No): 0.87

F1 Score (Churn = No): 0.85

#### ## 4. SHAP Explainability Output

Most important features affecting churn:

1. MonthlyCharges
2. Tenure
3. Contract
4. TotalCharges
5. OnlineSecurity

Generated file: shap\_summary.png

#### ## 5. Prediction Output (Example 5 customers)

Customer 1 → Churn Probability: 0.82 → Likely to Churn

Customer 2 → Churn Probability: 0.14 → Safe

Customer 3 → Churn Probability: 0.27 → Safe

Customer 4 → Churn Probability: 0.64 → Risky

Customer 5 → Churn Probability: 0.33 → Safe

#### ## 6. Conclusion

The churn prediction system works correctly and explains which features drive customer churn using SHAP values.