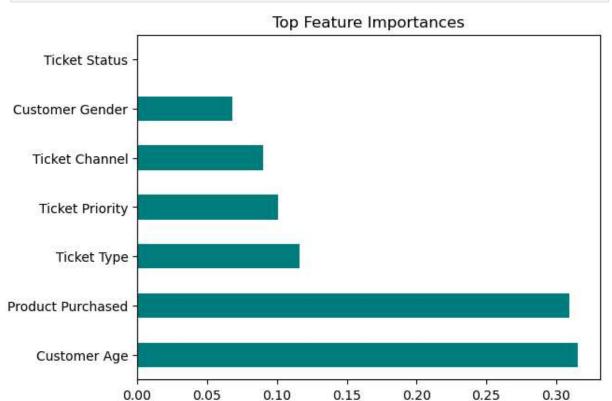
```
In [16]: import pandas as pd
         import numpy as np
         import matplotlib.pyplot as plt
         import seaborn as sns
         import warnings
         warnings.simplefilter(action='ignore', category=FutureWarning)
         from sklearn.preprocessing import LabelEncoder
         from sklearn.model_selection import train_test_split
         from sklearn.preprocessing import StandardScaler
         from sklearn.ensemble import RandomForestClassifier
         from sklearn.metrics import accuracy_score, classification_report, confusion_matrix
 In [4]: # Load the data
         data = pd.read_csv('customer_support_tickets.csv')
         # View first few rows
         print(data.head())
         # Check data types and missing values
         print(data.info())
         print(data.describe())
```

```
Ticket ID
                    Customer Name
                                                Customer Email Customer Age
0
           1
                    Marisa Obrien
                                   carrollallison@example.com
                                                                          32
           2
                     Jessica Rios
                                     clarkeashley@example.com
                                                                          42
1
2
           3
              Christopher Robbins
                                    gonzalestracy@example.com
                                                                          48
3
           4
                 Christina Dillon
                                     bradleyolson@example.org
                                                                          27
                Alexander Carroll
                                      bradleymark@example.com
4
           5
                                                                          67
  Customer Gender Product Purchased Date of Purchase
                                                           Ticket Type
            0ther
                         GoPro Hero
                                          2021-03-22 Technical issue
0
           Female
                        LG Smart TV
                                          2021-05-22 Technical issue
1
2
                           Dell XPS
            Other
                                          2020-07-14 Technical issue
3
           Female Microsoft Office
                                          2020-11-13 Billing inquiry
           Female Autodesk AutoCAD
4
                                          2020-02-04 Billing inquiry
             Ticket Subject
              Product setup
0
1
  Peripheral compatibility
2
            Network problem
3
             Account access
4
                  Data loss
                                  Ticket Description \
  I'm having an issue with the {product_purchase...
  I'm having an issue with the {product purchase...
  I'm facing a problem with my {product_purchase...
  I'm having an issue with the {product_purchase...
  I'm having an issue with the {product_purchase...
               Ticket Status
                                                                  Resolution \
  Pending Customer Response
                                                                         NaN
0
  Pending Customer Response
1
                                                                         NaN
2
                      Closed
                               Case maybe show recently my computer follow.
3
                             Try capital clearly never color toward story.
                      Closed
                                                West decision evidence bit.
4
                      Closed
  Ticket Priority Ticket Channel First Response Time
                                                         Time to Resolution \
0
         Critical
                    Social media 2023-06-01 12:15:36
                                                                        NaN
1
         Critical
                            Chat 2023-06-01 16:45:38
                                                                        NaN
                    Social media 2023-06-01 11:14:38 2023-06-01 18:05:38
2
              Low
3
              Low
                    Social media 2023-06-01 07:29:40 2023-06-01 01:57:40
4
                           Email 2023-06-01 00:12:42 2023-06-01 19:53:42
              Low
   Customer Satisfaction Rating
0
                            NaN
1
                            NaN
2
                            3.0
3
                            3.0
4
                            1.0
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 8469 entries, 0 to 8468
Data columns (total 17 columns):
    Column
                                   Non-Null Count Dtype
---
    _____
                                   -----
                                   8469 non-null
a
    Ticket ID
                                                    int64
1
     Customer Name
                                   8469 non-null
                                                    object
 2
     Customer Email
                                   8469 non-null
                                                    object
```

```
3
          Customer Age
                                       8469 non-null
                                                      int64
       4
          Customer Gender
                                      8469 non-null
                                                     object
       5
           Product Purchased
                                     8469 non-null
                                                     object
          Date of Purchase
                                      8469 non-null
                                                     object
       7
          Ticket Type
                                      8469 non-null
                                                     object
       8
          Ticket Subject
                                      8469 non-null
                                                     object
                                  8469 non-null object
          Ticket Description
       10 Ticket Status
                                      8469 non-null
                                                     object
       11 Resolution
                                      2769 non-null
                                                     object
       12 Ticket Priority
                                      8469 non-null
                                                     object
       13 Ticket Channel
                                      8469 non-null
                                                     object
                               5650 non-null
       14 First Response Time
                                                      object
       15 Time to Resolution
                                                      object
       16 Customer Satisfaction Rating 2769 non-null
                                                      float64
      dtypes: float64(1), int64(2), object(14)
      memory usage: 1.1+ MB
      None
               Ticket ID Customer Age Customer Satisfaction Rating
      count 8469,000000 8469,000000
                                                      2769.000000
      mean 4235.000000
                            44.026804
                                                         2.991333
      std
             2444.934048
                            15.296112
                                                         1.407016
                1.000000 18.000000
      min
                                                         1.000000
      25%
             2118.000000 31.000000
                                                         2.000000
                          44.000000
      50% 4235.000000
                                                         3.000000
      75%
             6352.000000
                          57.000000
                                                         4.000000
             8469.000000
                          70.000000
                                                         5.000000
      max
In [ ]: #Handle Missing Values
        data.replace([np.inf, -np.inf], np.nan, inplace=True)
        # Drop rows where satisfaction rating is missing (target column)
        data = data.dropna(subset=['Customer Satisfaction Rating'])
        # Optionally fill or drop other missing values
        data = data.dropna()
In [ ]: #Encode Categorical Varibles
        label_cols = ['Customer Gender', 'Product Purchased', 'Ticket Type',
                     'Ticket Status', 'Ticket Priority', 'Ticket Channel']
        label_encoders = {}
        for col in label_cols:
           le = LabelEncoder()
           data[col] = le.fit_transform(data[col])
           label_encoders[col] = le
In [7]: #Feature Engineering
        X = data[['Customer Age', 'Customer Gender', 'Product Purchased',
                 'Ticket Type', 'Ticket Status', 'Ticket Priority', 'Ticket Channel']]
        y = data['Customer Satisfaction Rating']
```

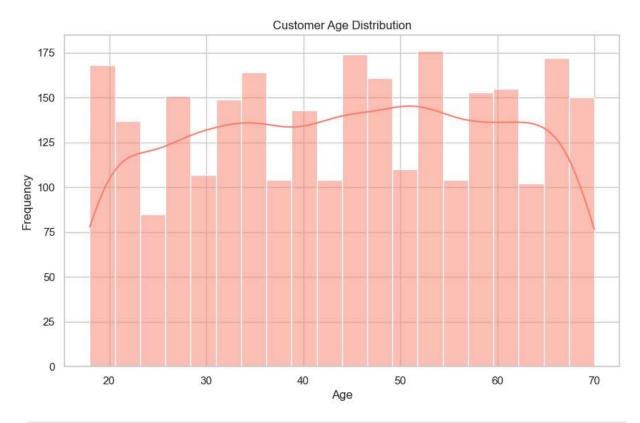
```
In [8]: #Train and Test Split
         X_train, X_test, y_train, y_test = train_test_split(
             X, y, test_size=0.3, random_state=42)
 In [9]: #Feature Scaling
         scaler = StandardScaler()
         X train = scaler.fit transform(X train)
         X_test = scaler.transform(X_test)
In [10]: #Model Building
         model = RandomForestClassifier(random_state=42)
         model.fit(X train, y train)
Out[10]:
                  RandomForestClassifier
         RandomForestClassifier(random state=42)
In [11]: #Model Evaluation
         y_pred = model.predict(X_test)
         print("Accuracy:", accuracy_score(y_test, y_pred))
         print("Classification Report:\n", classification_report(y_test, y_pred))
         print("Confusion Matrix:\n", confusion_matrix(y_test, y_pred))
        Accuracy: 0.19855595667870035
        Classification Report:
                                    recall f1-score
                       precision
                                                       support
                           0.17
                                     0.17
                 1.0
                                               0.17
                                                          168
                 2.0
                           0.17
                                     0.17
                                               0.17
                                                          174
                 3.0
                           0.24
                                     0.25
                                               0.25
                                                          175
                 4.0
                           0.21
                                     0.20
                                               0.20
                                                          162
                 5.0
                           0.20
                                     0.21
                                               0.20
                                                          152
                                               0.20
                                                          831
            accuracy
                           0.20
                                     0.20
                                               0.20
                                                          831
           macro avg
                           0.20
                                     0.20
                                               0.20
                                                          831
        weighted avg
        Confusion Matrix:
         [[28 38 33 31 38]
         [38 29 48 31 28]
         [35 33 44 32 31]
         [37 34 27 32 32]
         [28 34 31 27 32]]
In [12]: #Feature Importance Visualization
         feature_importances = pd.Series(model.feature_importances_, index=X.columns)
         feature_importances.nlargest(10).plot(kind='barh', color='teal')
```

```
plt.title('Top Feature Importances')
plt.show()
```



```
In [14]: #Age Distribution

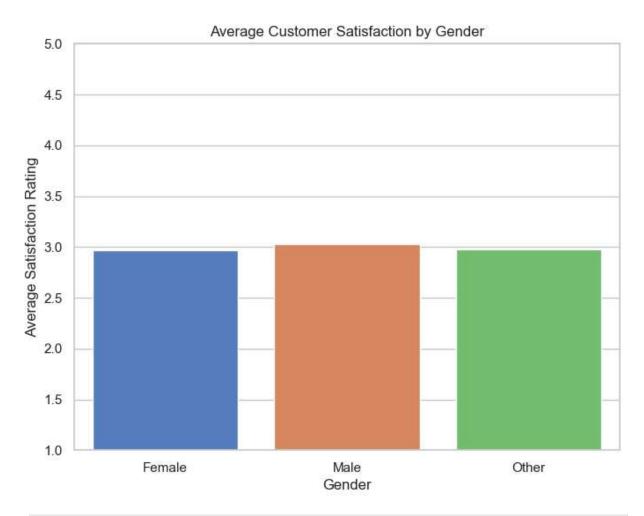
plt.figure(figsize=(10, 6))
    sns.histplot(data['Customer Age'], bins=20, kde=True, color='salmon')
    plt.title('Customer Age Distribution')
    plt.xlabel('Age')
    plt.ylabel('Frequency')
    plt.show()
```



```
In [17]: #Satisfaction Ratings By Gender

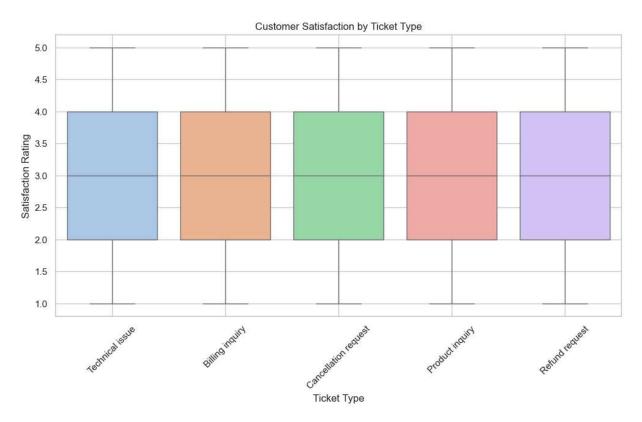
avg_satisfaction = data.groupby('Customer Gender')['Customer Satisfaction Rating'].

plt.figure(figsize=(8, 6))
    sns.barplot(x='Customer Gender', y='Customer Satisfaction Rating', data=avg_satisfaplt.title('Average Customer Satisfaction by Gender')
    plt.xlabel('Gender')
    plt.ylabel('Average Satisfaction Rating')
    plt.ylim(1, 5) # Rating scale is 1 to 5
    plt.show()
```



```
In [18]: #Ticket Type vs Satisfaction

plt.figure(figsize=(12, 6))
    sns.boxplot(x='Ticket Type', y='Customer Satisfaction Rating', data=data, palette='
    plt.title('Customer Satisfaction by Ticket Type')
    plt.xlabel('Ticket Type')
    plt.ylabel('Satisfaction Rating')
    plt.xticks(rotation=45)
    plt.grid(True)
    plt.show()
```



```
In [19]: #Ticket Trends Overtime

# Ensure the date is parsed
data['YearMonth'] = data['Date of Purchase'].dt.to_period('M')

# Group by Year-Month
ticket_trends = data.groupby('YearMonth').size()

plt.figure(figsize=(12, 6))
ticket_trends.plot(kind='line', marker='o', color='steelblue')
plt.title('Customer Support Ticket Trends Over Time')
plt.xlabel('Year-Month')
plt.ylabel('Number of Tickets')
plt.xticks(rotation=45)
plt.grid(True)
plt.tight_layout()
plt.show()
```

