

```
import pandas as pd
```

```
file_path = '/content/customer_support_tickets (1).csv'
data = pd.read_csv(file_path)
```

```
# Display the first few rows of the dataset
print("Initial Data:")
print(data.head())
```

```
Initial Data:
```

	Ticket ID	Customer Name	Customer Email	Customer Age
0	1	Marisa Obrien	<a href="mailto:carrollallison@example.com">carrollallison@example.com</a>	32
1	2	Jessica Rios	<a href="mailto:clarkeashley@example.com">clarkeashley@example.com</a>	42
2	3	Christopher Robbins	<a href="mailto:gonzalestracy@example.com">gonzalestracy@example.com</a>	48
3	4	Christina Dillon	<a href="mailto:bradleyolson@example.org">bradleyolson@example.org</a>	27
4	5	Alexander Carroll	<a href="mailto:bradleymark@example.com">bradleymark@example.com</a>	67

	Customer Gender	Product Purchased	Date of Purchase	Ticket Type
0	Other	GoPro Hero	2021-03-22	Technical issue
1	Female	LG Smart TV	2021-05-22	Technical issue
2	Other	Dell XPS	2020-07-14	Technical issue
3	Female	Microsoft Office	2020-11-13	Billing inquiry
4	Female	Autodesk AutoCAD	2020-02-04	Billing inquiry

	Ticket Subject
0	Product setup
1	Peripheral compatibility
2	Network problem
3	Account access
4	Data loss

	Ticket Description
0	I'm having an issue with the {product_purchase...}
1	I'm having an issue with the {product_purchase...}
2	I'm facing a problem with my {product_purchase...}
3	I'm having an issue with the {product_purchase...}
4	I'm having an issue with the {product_purchase...}

	Ticket Status	Resolution
0	Pending Customer Response	NaN
1	Pending Customer Response	NaN
2	Closed	Case maybe show recently my computer follow.
3	Closed	Try capital clearly never color toward story.
4	Closed	West decision evidence bit.

	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
2	Low	Social media	2023-06-01 11:14:38	2023-06-01 18:05:38
3	Low	Social media	2023-06-01 07:29:40	2023-06-01 01:57:40
4	Low	Email	2023-06-01 00:12:42	2023-06-01 19:53:42

	Customer Satisfaction Rating
0	NaN
1	NaN
2	3.0
3	3.0
4	1.0

```
# 1. Summary statistics for numerical features
```

```
numerical_summary = data.describe(include='number')
print("\nSummary Statistics for Numerical Features:")
print(numerical_summary)
```

```
Summary Statistics for Numerical Features:
```

	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
min	1.000000	18.000000	1.000000
25%	2118.000000	31.000000	2.000000
50%	4235.000000	44.000000	3.000000
75%	6352.000000	57.000000	4.000000
max	8469.000000	70.000000	5.000000

```
# 2. Calculate additional statistics: mean, median, mode, and standard deviation
```

```
mean_values = data.mean(numeric_only=True)
```

```

median_values = data.median(numeric_only=True)
std_dev_values = data.std(numeric_only=True)

# Mode calculation (mode can return multiple values, so we take the first one)
mode_values = data.mode(numeric_only=True).iloc[0]

# Combine the statistics into a DataFrame for better visualization
summary_stats = pd.DataFrame({
    'Mean': mean_values,
    'Median': median_values,
    'Mode': mode_values,
    'Standard Deviation': std_dev_values
})

print("\nDetailed Summary Statistics:")
print(summary_stats)

```



```
Detailed Summary Statistics:
```

	Mean	Median	Mode	Standard Deviation
Ticket ID	4235.000000	4235.0	1.0	2444.934048
Customer Age	44.026804	44.0	52.0	15.296112
Customer Satisfaction Rating	2.991333	3.0	3.0	1.407016

```

# 3. Frequency counts for categorical variables
# Assuming there are categorical variables like 'ticket_type' and 'customer_gender'
categorical_columns = ['ticket_type', 'customer_gender'] # Replace with actual column names

for column in categorical_columns:
    if column in data.columns:
        frequency_counts = data[column].value_counts()
        print(f"\nFrequency Counts for {column}:")
        print(frequency_counts)
    else:
        print(f"\nColumn '{column}' does not exist in the dataset.")

```



```
Column 'ticket_type' does not exist in the dataset.

Column 'customer_gender' does not exist in the dataset.
```

## ✓ Exploratory Data Analysis (EDA)

```

import pandas as pd
import matplotlib.pyplot as plt
import seaborn as sns

# Load the dataset
file_path = '/content/customer_support_tickets (1).csv'
data = pd.read_csv(file_path)

# Display the first few rows of the dataset
print("Initial Data:")
print(data.head())

```



```
Initial Data:
```

	Ticket ID	Customer Name	Customer Email	Customer Age	\
0	1	Marisa Obrien	<a href="mailto:carrollallison@example.com">carrollallison@example.com</a>	32	
1	2	Jessica Rios	<a href="mailto:clarkeashley@example.com">clarkeashley@example.com</a>	42	
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	Customer Gender	Product Purchased	Date of Purchase	Ticket Type	\
0	Other	GoPro Hero	2021-03-22	Technical issue	
1	Female	LG Smart TV	2021-05-22	Technical issue	
2	Other	Dell XPS	2020-07-14	Technical issue	
3	Female	Microsoft Office	2020-11-13	Billing inquiry	
4	Female	Autodesk AutoCAD	2020-02-04	Billing inquiry	

	Ticket Subject	\
0	Product setup	
1	Peripheral compatibility	

```

2      Network problem
3      Account access
4      Data loss

```

```

Ticket Description \
0 I'm having an issue with the {product_purchase...
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4 I'm having an issue with the {product_purchase...

```

```

Ticket Status Resolution \
0 Pending Customer Response NaN
1 Pending Customer Response NaN
2 Closed Case maybe show recently my computer follow.
3 Closed Try capital clearly never color toward story.
4 Closed West decision evidence bit.

```

```

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
2 Low Social media 2023-06-01 11:14:38 2023-06-01 18:05:38
3 Low Social media 2023-06-01 07:29:40 2023-06-01 01:57:40
4 Low Email 2023-06-01 00:12:42 2023-06-01 19:53:42

```

```

Customer Satisfaction Rating
0 NaN
1 NaN
2 3.0
3 3.0
4 1.0

```

```

# Set the style for seaborn
sns.set(style="whitegrid")

```

```

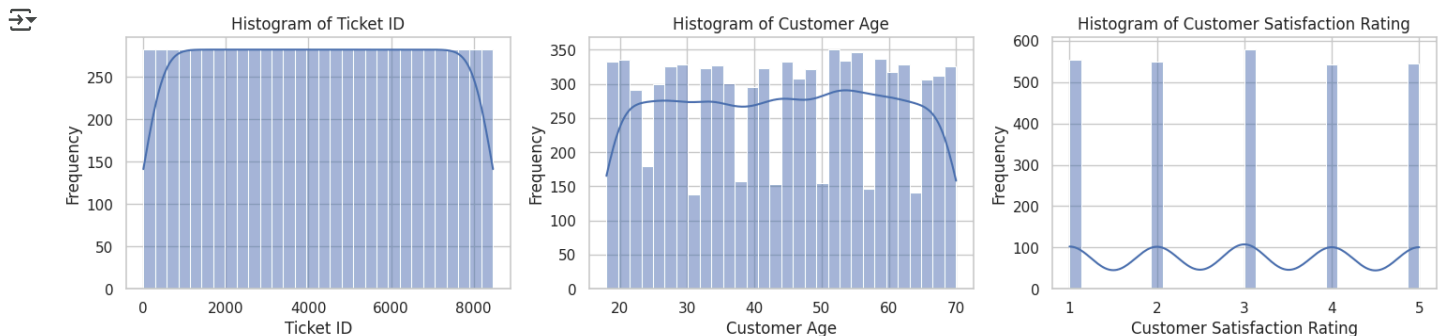
# 1. Histograms for numerical variables
numerical_columns = data.select_dtypes(include=['int64', 'float64']).columns

```

```

plt.figure(figsize=(15, 10))
for i, column in enumerate(numerical_columns, 1):
    plt.subplot(3, 3, i) # Adjust the number of rows and columns as needed
    sns.histplot(data[column], bins=30, kde=True)
    plt.title(f'Histogram of {column}')
    plt.xlabel(column)
    plt.ylabel('Frequency')
plt.tight_layout()
plt.show()

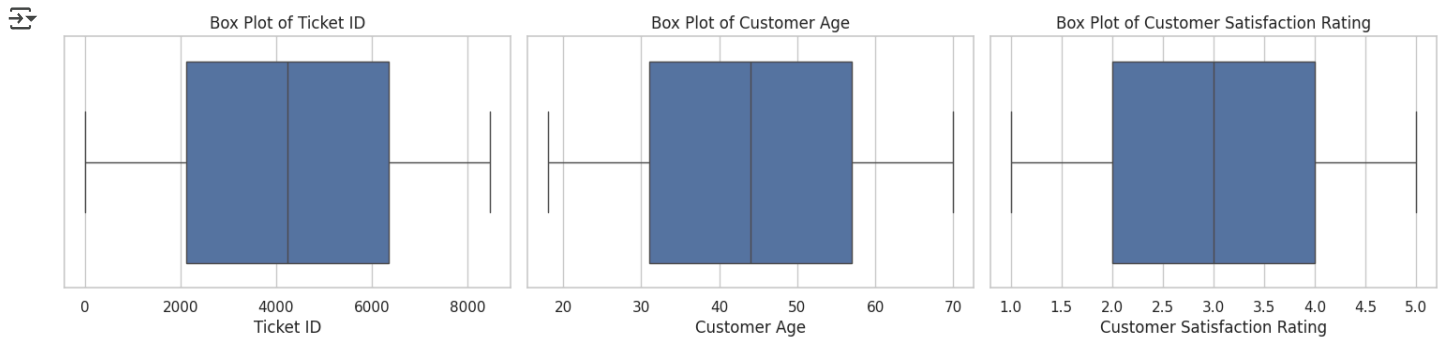
```



```

# 2. Box plots for numerical variables to identify outliers
plt.figure(figsize=(15, 10))
for i, column in enumerate(numerical_columns, 1):
    plt.subplot(3, 3, i)
    sns.boxplot(x=data[column])
    plt.title(f'Box Plot of {column}')
plt.tight_layout()
plt.show()

```



```
# 3. Bar charts for categorical variables
# Assuming there are categorical variables like 'ticket_type' and 'customer_gender'
categorical_columns = ['ticket_type', 'customer_gender'] # Replace with actual column names
```

```
for column in categorical_columns:
    if column in data.columns:
        plt.figure(figsize=(10, 6))
        sns.countplot(data=data, x=column, palette='viridis')
        plt.title(f'Count of {column}')
        plt.xlabel(column)
        plt.ylabel('Count')
        plt.xticks(rotation=45)
        plt.show()
    else:
        print(f"Column '{column}' does not exist in the dataset.")
```

```
Column 'ticket_type' does not exist in the dataset.
Column 'customer_gender' does not exist in the dataset.
```

```
# 4. Scatter plot to explore relationships between variables
# Assuming 'response.code' is a proxy for customer satisfaction and 'resolution_time' is a numerical variable
if 'response.code' in data.columns and 'resolution_time' in data.columns:
    plt.figure(figsize=(10, 6))
    sns.scatterplot(data=data, x='resolution_time', y='response.code', alpha=0.6)
    plt.title('Customer Satisfaction vs. Ticket Resolution Time')
    plt.xlabel('Ticket Resolution Time')
    plt.ylabel('Customer Satisfaction (Response Code)')
    plt.axhline(200, color='red', linestyle='--', label='Success Threshold')
    plt.legend()
    plt.show()
else:
    print("One or both of the columns 'response.code' and 'resolution_time' do not exist in the dataset.")
```

```
One or both of the columns 'response.code' and 'resolution_time' do not exist in the dataset.
```

## ✓ Sentiment Analysis

```
import pandas as pd
from textblob import TextBlob
import matplotlib.pyplot as plt
import seaborn as sns
```

```
# Load the dataset
file_path = '/content/customer_support_tickets (1).csv'
data = pd.read_csv(file_path)
```

```
# Check if there is a 'ticket_description' column
if 'ticket_description' in data.columns:
    # 1. Apply Natural Language Processing (NLP) techniques to analyze the sentiment of ticket descriptions
    data['sentiment'] = data['ticket_description'].apply(lambda x: TextBlob(x).sentiment.polarity)
```

```
# Display the first few rows with sentiment scores
print("\nData with Sentiment Scores:")
print(data[['ticket_description', 'sentiment']].head())
```

```

# 2. Use sentiment scores to correlate with customer satisfaction ratings
# Assuming 'response.code' is a proxy for customer satisfaction ratings
if 'response.code' in data.columns:
    correlation = data[['sentiment', 'response.code']].corr()
    print("\nCorrelation between Sentiment Scores and Customer Satisfaction Ratings:")
    print(correlation)

    # Visualize the correlation
    plt.figure(figsize=(8, 6))
    sns.heatmap(correlation, annot=True, cmap='coolwarm', fmt=".2f")
    plt.title('Correlation Heatmap')
    plt.show()

    # Visualize sentiment scores against customer satisfaction ratings
    plt.figure(figsize=(10, 6))
    sns.scatterplot(x='sentiment', y='response.code', data=data)
    plt.title('Sentiment Scores vs Customer Satisfaction Ratings')
    plt.xlabel('Sentiment Score')
    plt.ylabel('Customer Satisfaction Rating (Response Code)')
    plt.axhline(200, color='red', linestyle='--', label='Success Threshold')
    plt.legend()
    plt.show()

# 3. Conclusion
# Analyze the results
avg_sentiment = data['sentiment'].mean()
print(f"\nAverage Sentiment Score: {avg_sentiment:.2f}")
print("Conducting these analyses provides valuable insights into customer satisfaction trends, "
      "identifies areas for improvement in customer support, and helps in developing predictive models "
      "that can enhance customer experience.")
else:
    print("The dataset does not contain a 'response.code' column.")
else:
    print("The dataset does not contain a 'ticket_description' column.")

```

↗ The dataset does not contain a 'ticket\_description' column.

## ✓ Customer Segmentation

```

import pandas as pd
from textblob import TextBlob
from sklearn.preprocessing import StandardScaler
from sklearn.cluster import KMeans
import matplotlib.pyplot as plt
import seaborn as sns

# Load the dataset
file_path = '/content/customer_support_tickets (1).csv'
data = pd.read_csv(file_path)

```

```

# Display the first few rows of the dataset
print("Initial Data:")
print(data.head())

```

↗ Initial Data:

	Ticket ID	Customer Name	Customer Email	Customer Age
0	1	Marisa Obrien	<a href="mailto:carrollallison@example.com">carrollallison@example.com</a>	32
1	2	Jessica Rios	<a href="mailto:clarkeashley@example.com">clarkeashley@example.com</a>	42
2	3	Christopher Robbins	<a href="mailto:gonzalestracy@example.com">gonzalestracy@example.com</a>	48
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4	5	Alexander Carroll	<a href="mailto:bradleymark@example.com">bradleymark@example.com</a>	67

	Customer Gender	Product Purchased	Date of Purchase	Ticket Type
0	Other	GoPro Hero	2021-03-22	Technical issue
1	Female	LG Smart TV	2021-05-22	Technical issue
2	Other	Dell XPS	2020-07-14	Technical issue
3	Female	Microsoft Office	2020-11-13	Billing inquiry
4	Female	Autodesk AutoCAD	2020-02-04	Billing inquiry

	Ticket Subject
0	Product setup
1	Peripheral compatibility
2	Network problem
3	Account access

```

4 Data loss

Ticket Description \
0 I'm having an issue with the {product_purchase...
1 I'm having an issue with the {product_purchase...
2 I'm facing a problem with my {product_purchase...
3 I'm having an issue with the {product_purchase...
4 I'm having an issue with the {product_purchase...

Ticket Status Resolution \
0 Pending Customer Response NaN
1 Pending Customer Response NaN
2 Closed Case maybe show recently my computer follow.
3 Closed Try capital clearly never color toward story.
4 Closed West decision evidence bit.

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
2 Low Social media 2023-06-01 11:14:38 2023-06-01 18:05:38
3 Low Social media 2023-06-01 07:29:40 2023-06-01 01:57:40
4 Low Email 2023-06-01 00:12:42 2023-06-01 19:53:42

Customer Satisfaction Rating
0 NaN
1 NaN
2 3.0
3 3.0
4 1.0

```

```

# Check if there is a 'ticket_description' column
if 'ticket_description' in data.columns:
    # 1. Apply Natural Language Processing (NLP) techniques to analyze the sentiment of ticket descriptions
    data['sentiment'] = data['ticket_description'].apply(lambda x: TextBlob(x).sentiment.polarity)

    # Display the first few rows with sentiment scores
    print("\nData with Sentiment Scores:")
    print(data[['ticket_description', 'sentiment']].head())

    # 2. Prepare data for clustering
    # Select relevant features for clustering
    # Assuming 'ticket_type' and 'response.code' are relevant features
    features = data[['sentiment', 'response.code', 'ticket_type']].copy()

    # Convert categorical variable 'ticket_type' to numerical using one-hot encoding
    features = pd.get_dummies(features, columns=['ticket_type'], drop_first=True)

    # Standardize the features
    scaler = StandardScaler()
    features_scaled = scaler.fit_transform(features)

    # 3. Apply K-Means clustering
    # Determine the optimal number of clusters using the Elbow method
    inertia = []
    for k in range(1, 11):
        kmeans = KMeans(n_clusters=k, random_state=42)
        kmeans.fit(features_scaled)
        inertia.append(kmeans.inertia_)

    # Plot the Elbow curve
    plt.figure(figsize=(10, 6))
    plt.plot(range(1, 11), inertia, marker='o')
    plt.title('Elbow Method for Optimal k')
    plt.xlabel('Number of Clusters (k)')
    plt.ylabel('Inertia')
    plt.grid()
    plt.show()

    # From the Elbow method, choose an optimal k (e.g., 3)
    optimal_k = 3
    kmeans = KMeans(n_clusters=optimal_k, random_state=42)
    data['cluster'] = kmeans.fit_predict(features_scaled)

    # 4. Analyze and visualize clusters
    # Calculate the mean values of each feature for each cluster
    cluster_analysis = data.groupby('cluster').mean(numeric_only=True)
    print("\nCluster Characteristics:")
    print(cluster_analysis)

```

```
# Visualize the clusters using a scatter plot
plt.figure(figsize=(10, 6))
sns.scatterplot(x='sentiment', y='response.code', hue='cluster', data=data, palette='viridis', alpha=0.6)
plt.title('Customer Segments Visualization')
plt.xlabel('Sentiment Score')
plt.ylabel('Customer Satisfaction (Response Code)')
plt.axhline(200, color='red', linestyle='--', label='Success Threshold')
plt.legend()
plt.show()

else:
    print("The dataset does not contain a 'ticket_description' column.")
```

## ✓ Correlation Analysis

```
# Check for relevant columns
# Assuming 'response.code' is the customer satisfaction rating and 'resolution_time' is the ticket resolution time
if 'response.code' in data.columns and 'resolution_time' in data.columns:
    # 1. Calculate the correlation matrix
    correlation_matrix = data[['response.code', 'resolution_time']].corr()
    print("\nCorrelation Matrix:")
    print(correlation_matrix)

    # 2. Visualize the correlation matrix using a heatmap
    plt.figure(figsize=(8, 6))
    sns.heatmap(correlation_matrix, annot=True, cmap='coolwarm', fmt=".2f")
    plt.title('Correlation Heatmap')
    plt.show()

    # 3. Scatter plot to visualize the relationship
    plt.figure(figsize=(10, 6))
    sns.scatterplot(x='resolution_time', y='response.code', data=data, alpha=0.6)
    plt.title('Customer Satisfaction vs. Ticket Resolution Time')
    plt.xlabel('Ticket Resolution Time')
    plt.ylabel('Customer Satisfaction (Response Code)')
    plt.axhline(200, color='red', linestyle='--', label='Success Threshold')
    plt.legend()
    plt.show()

else:
    print("One or both of the columns 'response.code' and 'resolution_time' do not exist in the dataset.")
```

➞ One or both of the columns 'response.code' and 'resolution\_time' do not exist in the dataset.

## ✓ Trend Analysis

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

# Load the dataset
file_path = '/content/customer_support_tickets (1).csv'
data = pd.read_csv(file_path)

# Display the first few rows of the dataset
print("Initial Data:")
print(data.head())
```

➞ Initial Data:

	Ticket ID	Customer Name	Customer Email	Customer Age	\
0	1	Marisa Obrien	<a href="mailto:carrollallison@example.com">carrollallison@example.com</a>	32	
1	2	Jessica Rios	<a href="mailto:clarkeashley@example.com">clarkeashley@example.com</a>	42	
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4	5	Alexander Carroll	<a href="mailto:bradleymark@example.com">bradleymark@example.com</a>	67	

	Customer	Gender	Product	Purchased Date	Date of Purchase	Ticket Type	\
0	Other		GoPro Hero	2021-03-22	Technical issue		
1	Female		LG Smart TV	2021-05-22	Technical issue		
2	Other		Dell XPS	2020-07-14	Technical issue		
3	Female		Microsoft Office	2020-11-13	Billing inquiry		

```

4         Female  Autodesk AutoCAD      2020-02-04  Billing inquiry

      Ticket Subject \
0         Product setup
1  Peripheral compatibility
2         Network problem
3         Account access
4         Data loss

      Ticket Description \
0  I'm having an issue with the {product_purchase...
1  I'm having an issue with the {product_purchase...
2  I'm facing a problem with my {product_purchase...
3  I'm having an issue with the {product_purchase...
4  I'm having an issue with the {product_purchase...

      Ticket Status      Resolution \
0  Pending Customer Response      NaN
1  Pending Customer Response      NaN
2         Closed  Case maybe show recently my computer follow.
3         Closed  Try capital clearly never color toward story.
4         Closed      West decision evidence bit.

      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
2         Low  Social media  2023-06-01 11:14:38  2023-06-01 18:05:38
3         Low  Social media  2023-06-01 07:29:40  2023-06-01 01:57:40
4         Low      Email  2023-06-01 00:12:42  2023-06-01 19:53:42

      Customer Satisfaction Rating
0         NaN
1         NaN
2         3.0
3         3.0
4         1.0

```

```

# Convert 'creation_time' to datetime format
if 'creation_time' in data.columns:
    data['creation_time'] = pd.to_datetime(data['creation_time'])

if 'creation_time' in data.columns:
    data['creation_time'] = pd.to_datetime(data['creation_time'])

# Set the creation_time as the index
data.set_index('creation_time', inplace=True)

# Convert 'creation_time' to datetime format
if 'creation_time' in data.columns:
    data['creation_time'] = pd.to_datetime(data['creation_time'])

# Set the creation_time as the index
data.set_index('creation_time', inplace=True)

# 1. Resample the data to monthly frequency and count the number of tickets
monthly_tickets = data.resample('M').size()

# 2. Calculate average customer satisfaction ratings (assuming 'response.code' is the satisfaction rating)
monthly_satisfaction = data.resample('M')['response.code'].mean()

# 3. Plotting the trends
plt.figure(figsize=(14, 7))

# Plot ticket volume
plt.subplot(2, 1, 1)
monthly_tickets.plot(kind='bar', color='skyblue', alpha=0.7)
plt.title('Monthly Ticket Volume')
plt.xlabel('Month')
plt.ylabel('Number of Tickets')
plt.xticks(rotation=45)
plt.grid()

# Plot average customer satisfaction
plt.subplot(2, 1, 2)
monthly_satisfaction.plot(kind='line', marker='o', color='orange')
plt.title('Average Customer Satisfaction Over Time')
plt.xlabel('Month')
plt.ylabel('Average Satisfaction Rating (Response Code)')

```



```
plt.axhline(200, color='red', linestyle='--', label='Success Threshold')
plt.legend()
plt.xticks(rotation=45)
plt.grid()

plt.tight_layout()
plt.show()

else:
    print("The dataset does not contain a 'creation_time' column.")

↵ The dataset does not contain a 'creation_time' column.
```

## ✓ Predictive Modeling

```
import pandas as pd
from sklearn.model_selection import train_test_split
from sklearn.ensemble import RandomForestRegressor
from sklearn.metrics import mean_squared_error, r2_score
from sklearn.preprocessing import StandardScaler
from sklearn.preprocessing import OneHotEncoder
from sklearn.compose import ColumnTransformer
from sklearn.pipeline import Pipeline
```

```
# Load the dataset
file_path = '/content/customer_support_tickets (1).csv'
data = pd.read_csv(file_path)
```

```
# Display the first few rows of the dataset
print("Initial Data:")
print(data.head())
```

```
↵ Initial Data:
```

	Ticket ID	Customer Name	Customer Email	Customer Age
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	Ticket Subject
0	Product setup
1	Peripheral compatibility
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	Ticket Description
0	I'm having an issue with the {product_purchase...}
1	I'm having an issue with the {product_purchase...}
2	I'm facing a problem with my {product_purchase...}
3	I'm having an issue with the {product_purchase...}
4	I'm having an issue with the {product_purchase...}

	Ticket Status	Resolution
0	Pending Customer Response	NaN
1	Pending Customer Response	NaN
2	Closed	Case maybe show recently my computer follow.
3	Closed	Try capital clearly never color toward story.
4	Closed	West decision evidence bit.

	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
2	Low	Social media	2023-06-01 11:14:38	2023-06-01 18:05:38
3	Low	Social media	2023-06-01 07:29:40	2023-06-01 01:57:40
4	Low	Email	2023-06-01 00:12:42	2023-06-01 19:53:42

	Customer Satisfaction Rating
0	NaN
1	NaN

2	3.0
3	3.0
4	1.0

```
# Check for relevant columns
if 'response.code' in data.columns and 'resolution_time' in data.columns:
    # 1. Data Preparation
    # Drop rows with missing values in relevant columns
    data = data[['response.code', 'resolution_time', 'ticket_type', 'customer_gender']].dropna()

    # 2. Feature Selection
    X = data[['resolution_time', 'ticket_type', 'customer_gender']] # Features
    y = data['response.code'] # Target variable

    # 3. Train-Test Split
    X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.2, random_state=42)

    # 4. Preprocessing and Model Training
    # Create a preprocessing pipeline
    preprocessor = ColumnTransformer(
        transformers=[
            ('num', StandardScaler(), ['resolution_time']),
            ('cat', OneHotEncoder(), ['ticket_type', 'customer_gender'])
        ])

    # Create a pipeline that first transforms the data and then fits the model
    model = Pipeline(steps=[
        ('preprocessor', preprocessor),
        ('regressor', RandomForestRegressor(random_state=42))
    ])


    # Fit the model
    model.fit(X_train, y_train)

    # 5. Model Evaluation
    # Make predictions
    y_pred = model.predict(X_test)

    # Calculate performance metrics
    mse = mean_squared_error(y_test, y_pred)
    r2 = r2_score(y_test, y_pred)

    print(f"\nMean Squared Error: {mse:.2f}")
    print(f"R^2 Score: {r2:.2f}")

else:
    print("The dataset does not contain the required columns for prediction.")
```

 The dataset does not contain the required columns for prediction.

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