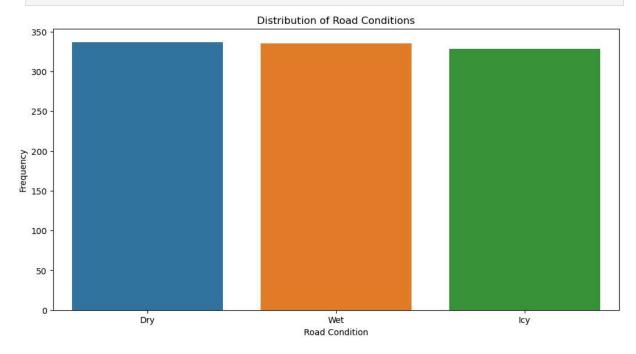
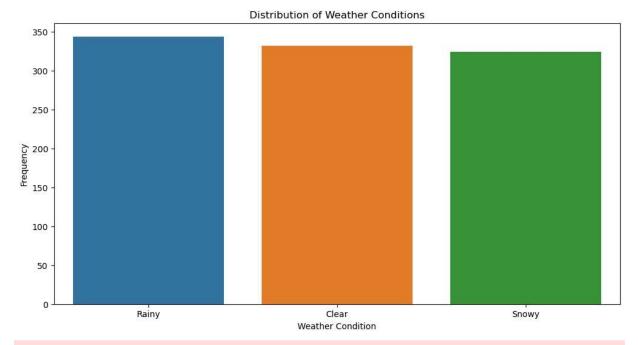
```
In [2]: import pandas as pd
        import numpy as np
        import matplotlib.pyplot as plt
        import seaborn as sns
        from sklearn.cluster import KMeans
        # Generate synthetic data
        np.random.seed(0)
        n \text{ samples} = 1000
        # Simulate road conditions (0: dry, 1: wet, 2: icy)
        road conditions = np.random.choice(['Dry', 'Wet', 'Icy'], size=n samples)
        # Simulate weather conditions (0: clear, 1: rainy, 2: snowy)
        weather_conditions = np.random.choice(['Clear', 'Rainy', 'Snowy'], size=n_samples)
        # Simulate time of day (0-23 hours)
        time of day = np.random.randint(0, 24, size=n samples)
        # Simulate accident locations (latitude and longitude)
        latitude = np.random.uniform(40, 41, size=n samples)
        longitude = np.random.uniform(-74, -73, size=n samples)
        # Simulate accident severity (random integers from 1 to 5)
        severity = np.random.randint(1, 6, size=n samples)
        # Create a synthetic DataFrame
        data = pd.DataFrame({
             'road_condition': road_conditions,
             'weather': weather_conditions,
             'time_of_day': time_of_day,
             'latitude': latitude,
             'longitude': longitude,
             'severity': severity
        })
        # Data preprocessing
        # Exploratory Data Analysis (EDA)
        # Visualize distributions of road conditions, weather, and time of day
        plt.figure(figsize=(12, 6))
        sns.countplot(x='road condition', data=data)
        plt.title('Distribution of Road Conditions')
        plt.xlabel('Road Condition')
        plt.ylabel('Frequency')
        plt.show()
        plt.figure(figsize=(12, 6))
        sns.countplot(x='weather', data=data)
        plt.title('Distribution of Weather Conditions')
        plt.xlabel('Weather Condition')
        plt.ylabel('Frequency')
        plt.show()
        plt.figure(figsize=(12, 6))
```

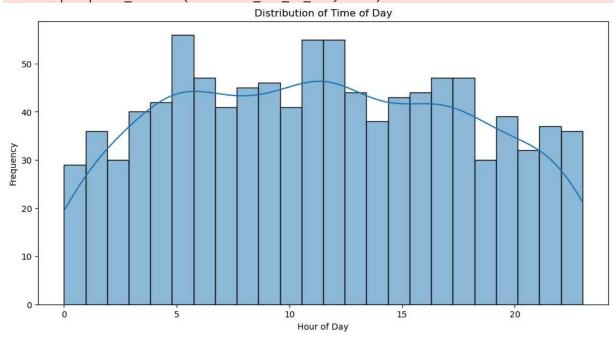
```
sns.histplot(data['time_of_day'], bins=24, kde=True)
plt.title('Distribution of Time of Day')
plt.xlabel('Hour of Day')
plt.ylabel('Frequency')
plt.show()
# Identify accident hotspots using clustering (K-means)
X = data[['latitude', 'longitude']]
kmeans = KMeans(n clusters=5, random state=0).fit(X)
data['cluster'] = kmeans.labels
# Visualize accident hotspots
plt.figure(figsize=(10, 8))
sns.scatterplot(x='longitude', y='latitude', hue='cluster', data=data, palette='vir
plt.title('Accident Hotspots')
plt.xlabel('Longitude')
plt.ylabel('Latitude')
plt.legend(title='Cluster')
plt.show()
# Time of Day Analysis
plt.figure(figsize=(12, 6))
sns.countplot(x='time_of_day', data=data, color='orange')
plt.title('Accidents by Hour of Day')
plt.xlabel('Hour of Day')
plt.ylabel('Number of Accidents')
plt.show()
# Factor Analysis
plt.figure(figsize=(10, 6))
sns.boxplot(x='road_condition', y='severity', data=data)
plt.title('Impact of Road Conditions on Accident Severity')
plt.xlabel('Road Condition')
plt.ylabel('Severity')
plt.show()
```





C:\Users\devat\anaconda3\Lib\site-packages\seaborn_oldcore.py:1119: FutureWarning: use_inf_as_na option is deprecated and will be removed in a future version. Convert inf values to NaN before operating instead.

with pd.option context('mode.use inf as na', True):



C:\Users\devat\anaconda3\Lib\site-packages\sklearn\cluster_kmeans.py:870: FutureWar
ning: The default value of `n_init` will change from 10 to 'auto' in 1.4. Set the va
lue of `n_init` explicitly to suppress the warning
 warnings.warn(

C:\Users\devat\anaconda3\Lib\site-packages\sklearn\cluster_kmeans.py:1382: UserWarn ing: KMeans is known to have a memory leak on Windows with MKL, when there are less chunks than available threads. You can avoid it by setting the environment variable OMP_NUM_THREADS=4.

warnings.warn(

