

```

import pandas as pd
import numpy as np
import matplotlib.pyplot as plt

from sklearn.model_selection import train_test_split
from sklearn.preprocessing import LabelEncoder, StandardScaler
from sklearn.linear_model import LinearRegression, LogisticRegression
from sklearn.metrics import mean_absolute_error, accuracy_score, confusion_matrix
from sklearn.neural_network import MLPRegressor, MLPClassifier
from sklearn.neighbors import KNeighborsClassifier, KNeighborsRegressor

```

```

df = pd.read_csv("Food_Delivery_Times.csv")
print(df.head())
print(df.info())

```

	Order_ID	Distance_km	Weather	Traffic_Level	Time_of_Day	Vehicle_Type	\
0	522	7.93	Windy	Low	Afternoon	Scooter	
1	738	16.42	Clear	Medium	Evening	Bike	
2	741	9.52	Foggy	Low	Night	Scooter	
3	661	7.44	Rainy	Medium	Afternoon	Scooter	
4	412	19.03	Clear	Low	Morning	Bike	

	Preparation_Time_min	Courier_Experience_yrs	Delivery_Time_min
0	12	1.0	43
1	20	2.0	84
2	28	1.0	59
3	5	1.0	37
4	16	5.0	68

```

<class 'pandas.core.frame.DataFrame'>
RangeIndex: 1000 entries, 0 to 999
Data columns (total 9 columns):
 #   Column           Non-Null Count  Dtype  
--- 
 0   Order_ID         1000 non-null   int64  
 1   Distance_km     1000 non-null   float64 
 2   Weather          970 non-null   object  
 3   Traffic_Level    970 non-null   object  
 4   Time_of_Day      970 non-null   object  
 5   Vehicle_Type     1000 non-null   object  
 6   Preparation_Time_min  1000 non-null  int64  
 7   Courier_Experience_yrs  970 non-null  float64 
 8   Delivery_Time_min  1000 non-null   int64  
dtypes: float64(2), int64(3), object(4)
memory usage: 70.4+ KB
None

```

```

-----Data Cleaning and preprocessing-----
# Dataset shape
print("\tBEFORE")
print("Initial shape:", df.shape)

# Check duplicates
print("Duplicate rows:", df.duplicated().sum())

# Check missing values
print("Missing values per column:")
print(df.isnull().sum())

#fix missing values
median_exp = df['Courier_Experience_yrs'].median()
df['Courier_Experience_yrs'] = df['Courier_Experience_yrs'].fillna(median_exp)
df.dropna(inplace=True)
print("\n\tAFTER")
print("Shape after handling missing values:", df.shape)
print("Final missing values:")
print(df.isnull().sum())

```

```

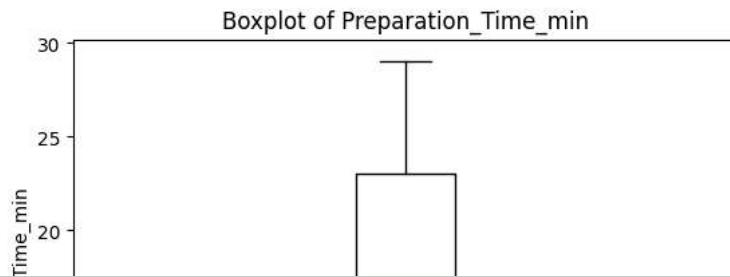
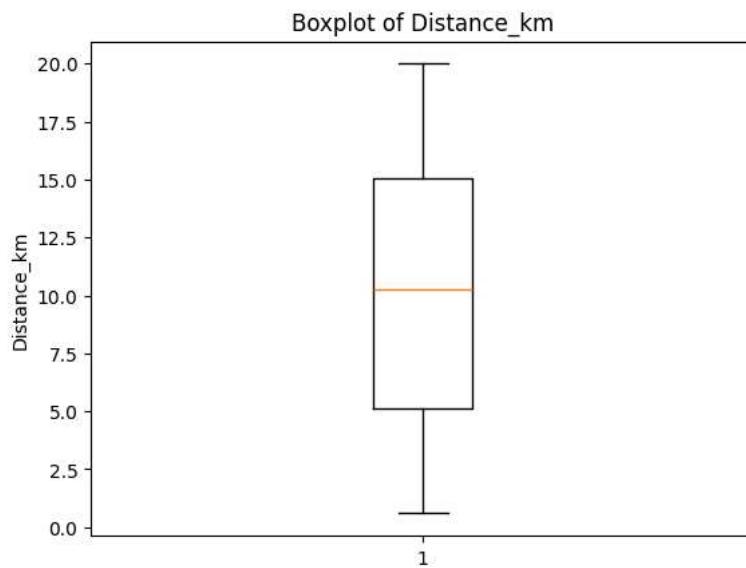
BEFORE
Initial shape: (1000, 9)
Duplicate rows: 0
Missing values per column:
Order_ID          0
Distance_km       0
Weather           30
Traffic_Level     30
Time_of_Day        30
Vehicle_Type       0

```

```
Preparation_Time_min      0
Courier_Experience_yrs    30
Delivery_Time_min         0
dtype: int64
```

```
AFTER
Shape after handling missing values: (912, 9)
Final missing values:
Order_ID                  0
Distance_km                0
Weather                     0
Traffic_Level               0
Time_of_Day                 0
Vehicle_Type                0
Preparation_Time_min        0
Courier_Experience_yrs      0
Delivery_Time_min           0
dtype: int64
```

```
#Check outliers
numerical_cols = ['Distance_km', 'Preparation_Time_min', 'Courier_Experience_yrs', 'Delivery_Time_min']
for col in numerical_cols:
    plt.boxplot(df[col])
    plt.title(f"Boxplot of {col}")
    plt.ylabel(col)
    plt.show()
print("\nSummary statistics:")
print(df.describe())
```

```
#Remove outliers
for col in ['Delivery_Time_min']:
    Q1 = df[col].quantile(0.25)
    Q3 = df[col].quantile(0.75)
    IQR = Q3 - Q1

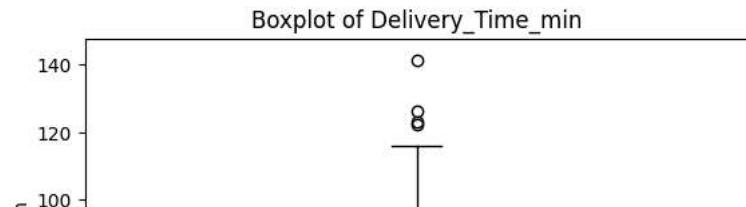
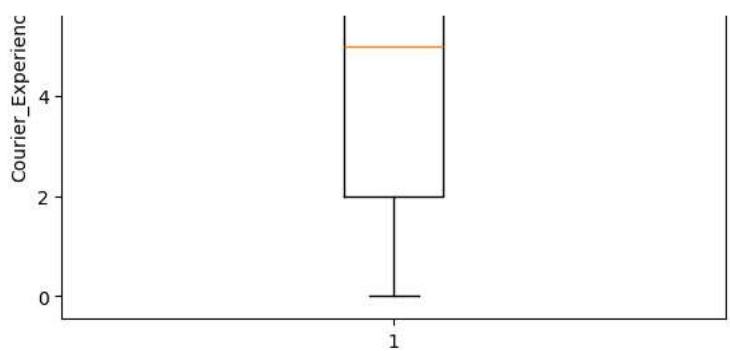
    lower = Q1 - 1.5 * IQR
    upper = Q3 + 1.5 * IQR

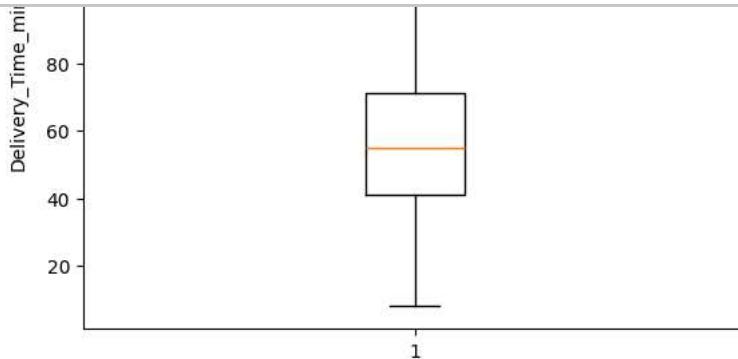
    df = df[(df[col] >= lower) & (df[col] <= upper)]

print("Final cleaned dataset shape:", df.shape)

for col in numerical_cols:
    plt.boxplot(df[col])
    plt.title(f"Boxplot of {col}")
    plt.ylabel(col)
    plt.show()

print("\nSummary statistics:")
print(df.describe())
```



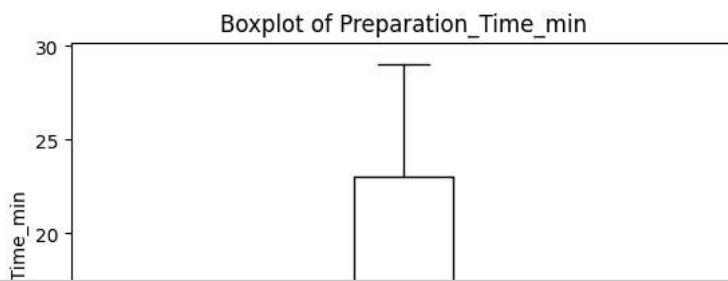
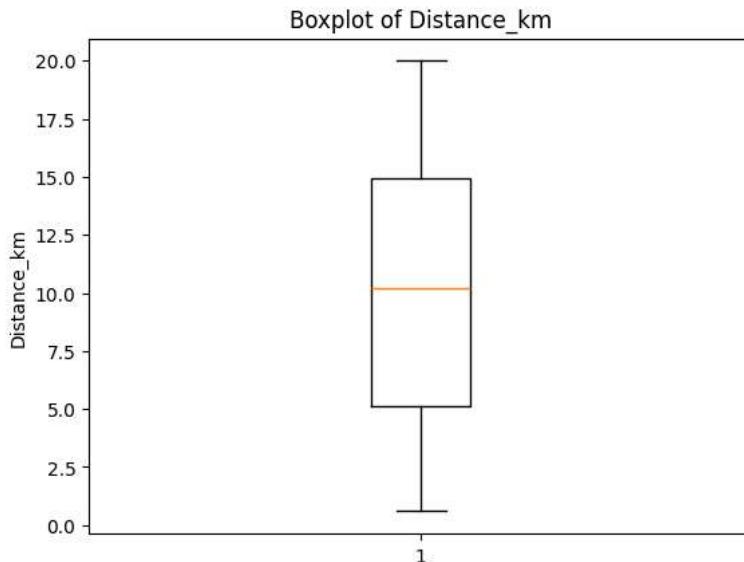


Summary statistics:

	Order_ID	Distance_km	Preparation_Time_min	Courier_Experience_yrs	\
count	912.000000	912.000000	912.000000	912.000000	
mean	507.573465	10.069320	16.993421	4.651316	
std	288.528012	5.692518	7.264636	2.875980	
min	1.000000	0.590000	5.000000	0.000000	
25%	259.750000	5.130000	10.000000	2.000000	
50%	512.500000	10.285000	17.000000	5.000000	
75%	758.250000	15.042500	23.000000	7.000000	
max	1000.000000	19.990000	29.000000	9.000000	

	Delivery_Time_min
count	912.000000
mean	56.458658
std	21.581281
min	8.000000
25%	41.000000
50%	55.000000
75%	71.000000
max	141.000000

```
Final cleaned dataset shape: (908, 9)
```



```
#-----Feature engineering and eda-----
# Distance Category
df['Distance_Category'] = pd.cut(
    df['Distance_km'],
    bins=[0, 5, 10, 20, 50],
    labels=['Short', 'Medium', 'Long', 'Very Long']
)

# Peak Hour Indicator
df['Peak_Hour'] = df['Time_of_Day'].apply(
    lambda x: 1 if x in ['Evening', 'Night'] else 0
)

# Courier Experience Groups
df['Experience_Group'] = pd.cut(
    df['Courier_Experience_yrs'],
    bins=[0, 2, 5, 10, 50],
    labels=['Beginner', 'Intermediate', 'Experienced', 'Expert']
)

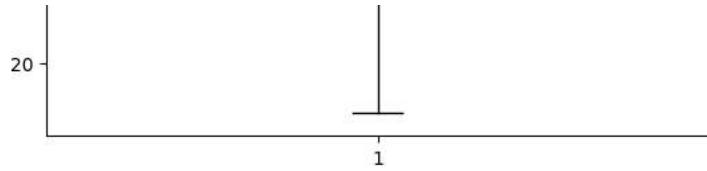
# Delay Status (Target for classification)
df['Delay_Status'] = np.where(
    df['Delivery_Time_min'] >
    df['Preparation_Time_min'] + 15,
    'Delayed', 'On-Time'
)

print(df.head())
# Weather Impact
df.groupby('Weather')['Delivery_Time_min'].mean().plot(kind='bar')
plt.title("Weather Impact on Delivery Time")
plt.ylabel("Average Delivery Time (min)")
plt.show()

# Traffic Impact
df.groupby('Traffic_Level')['Delivery_Time_min'].mean().plot(kind='bar')
plt.title("Traffic Level vs Delivery Time")
plt.ylabel("Average Delivery Time (min)")
plt.show()

# Distance vs Time
```

```
plt.scatter(df['Distance_km'], df['Delivery_Time_min'])
plt.xlabel("Distance (km)")
plt.ylabel("Delivery Time (min)")
plt.title("Distance vs Delivery Time")
plt.show()
```



Summary statistics:

	Order_ID	Distance_km	Preparation_Time_min	Courier_Experience_yrs	\
count	908.000000	908.000000	908.000000	908.000000	
mean	507.378855	10.036960	16.985683	4.650881	
std	288.745428	5.683134	7.270548	2.881016	
min	1.000000	0.590000	5.000000	0.000000	
25%	258.750000	5.105000	10.000000	2.000000	
50%	513.500000	10.220000	17.000000	5.000000	
75%	758.250000	14.965000	23.000000	7.000000	
max	1000.000000	19.990000	29.000000	9.000000	

Delivery_Time_min

	Delivery_Time_min
count	908.000000
mean	56.135463
std	21.091974
min	8.000000
25%	40.750000
50%	55.000000
75%	70.250000
max	116.000000


```

Order_ID Distance_km Weather Traffic_Level Time_of_Day Vehicle_Type \
0      522      7.93  Windy       Low  Afternoon    Scooter
1      738     16.42  Clear    Medium   Evening     Bike
2      741      9.52  Foggy      Low   Night    Scooter
3      661      7.44  Rainy    Medium  Afternoon    Scooter
4      412     19.03  Clear      Low   Morning     Bike

Preparation_Time_min Courier_Experience_yrs Delivery_Time_min \
0                  12             1.0            43
1                  20             2.0            84
2                  28             1.0            59
3                   5             1.0            37
4                 16             5.0            68

```

```

Distance_Category Peak_Hour Experience_Group Delay_Status
0           Medium          0      Beginner    Delayed
1            Long           1      Beginner    Delayed
2           Medium          1      Beginner    Delayed
3           Medium          0      Beginner    Delayed

```

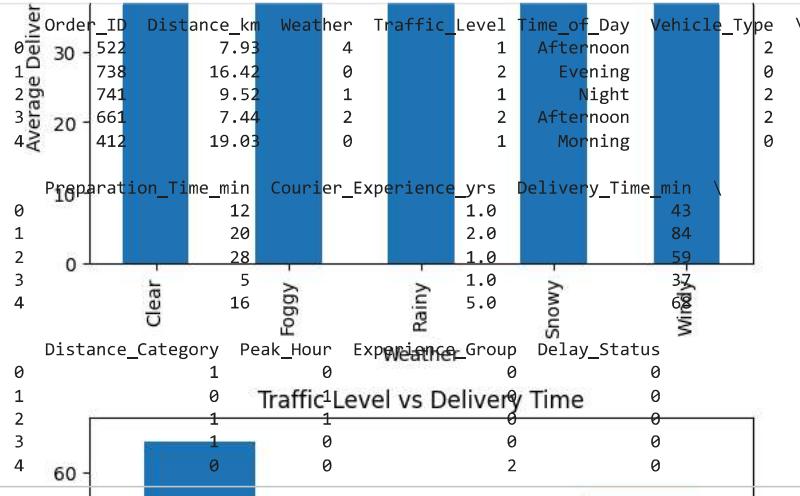
```
#-----Label Encoding for categorical columns-----
categorical_cols = ['Weather', 'Traffic_Level', 'Vehicle_Type', 'Distance_Category', 'Experience_Group', 'Delay_Status']
```

```

le_dict = {}
for col in categorical_cols:
    le = LabelEncoder()
    df[col] = le.fit_transform(df[col])
    le_dict[col] = le # Save encoder for future use

```

```
print(df.head())
```



```
import seaborn as sns
```

```
# ----- Correlation Heatmap -----
plt.figure(figsize=(12, 8))
```

```

# Select only numeric columns
corr = df[['Distance_km',
           'Preparation_Time_min',
           'Courier_Experience_yrs',
           'Weather',
           'Traffic_Level',
           'Vehicle_Type',
           'Distance_Category',
           'Experience_Group',
           'Peak_Hour',
           'Delivery_Time_min',
           'Delay_Status']].corr()

```

```

sns.heatmap(
    corr,
    annot=True,
    cmap='coolwarm',
    fmt=".2f",
    linewidths=0.5
)

```