

```
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}")
```

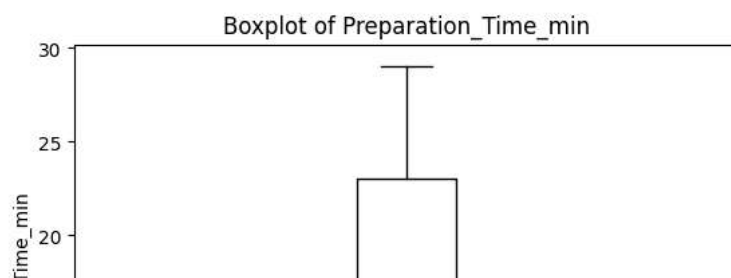
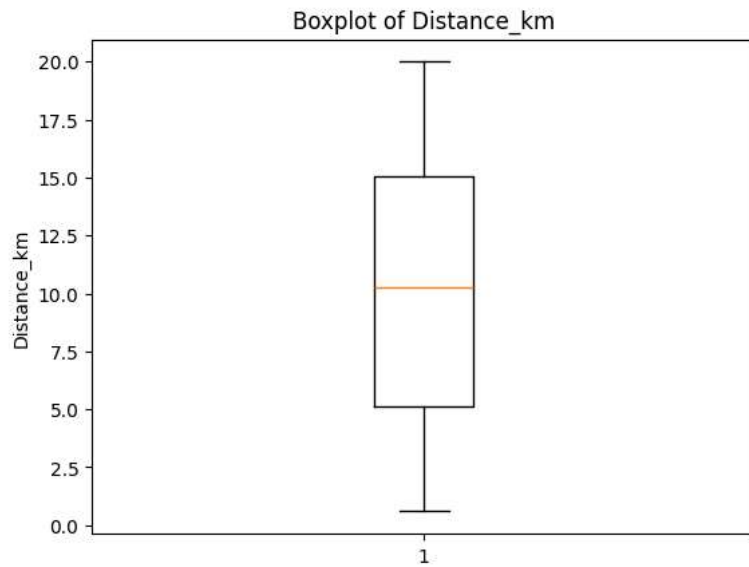
```
    plt.ylabel(col)
```

```
    plt.show()
```

```
print("\nSummary statistics:")
```

```
print(df.describe())
```





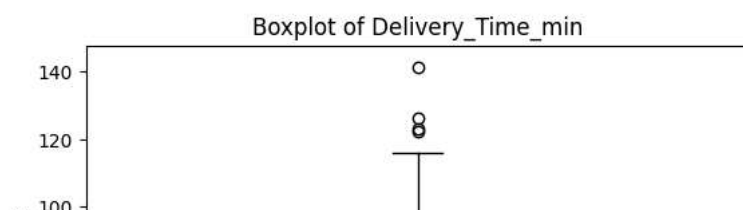
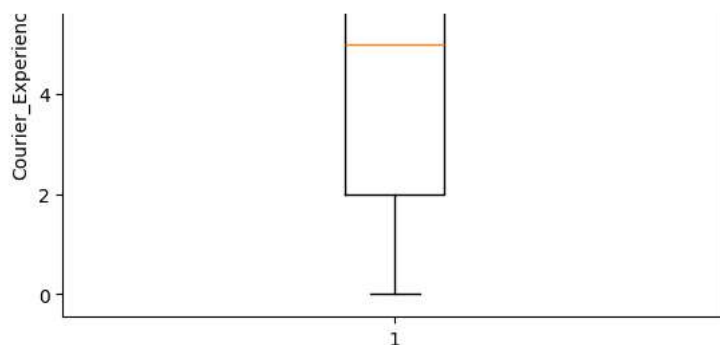
```
#Remove outliers
for col in ['Delivery_Time_min']:
    Q1 = df[col].quantile(0.25)
    Q3 = df[col].quantile(0.75)
    IQR = Q3 - Q1

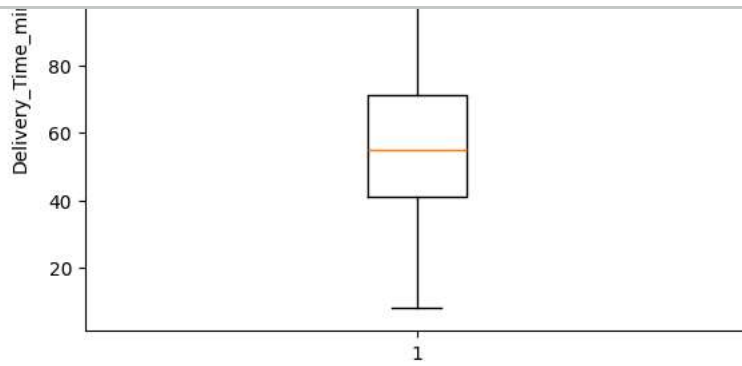
    lower = Q1 - 1.5 * IQR
    upper = Q3 + 1.5 * IQR

    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())
```



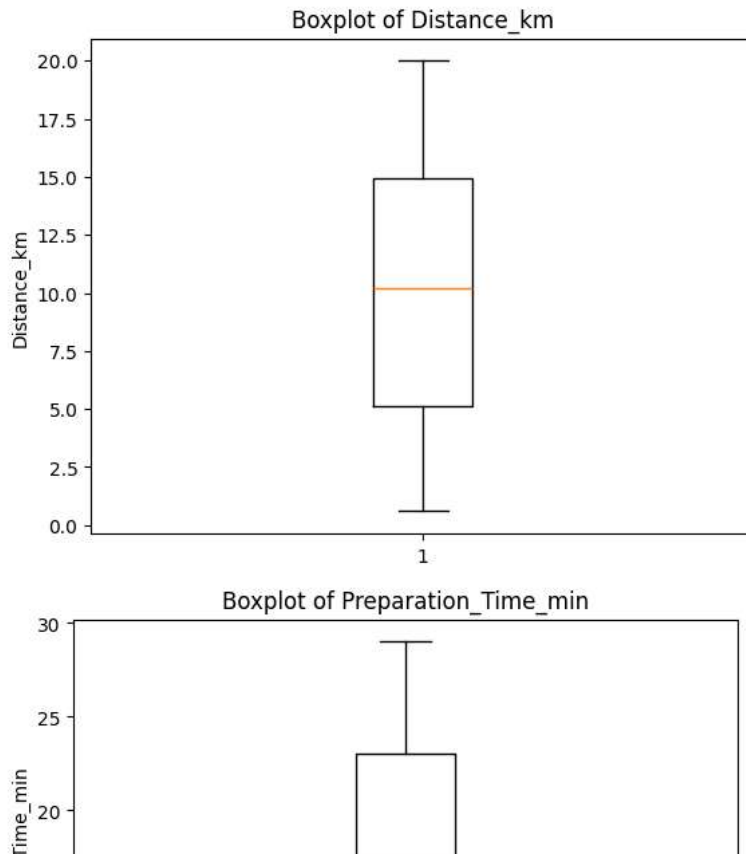


Summary statistics:

	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	56.450658
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']
)

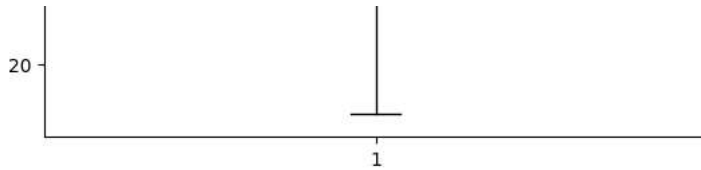
# Delay Status (Target for classification)
df['Delay_Status'] = np.where(
    df['Delivery_Time_min'] >
    df['Preparation_Time_min'] + 15,
    'Delayed', 'On-Time'
)

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
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

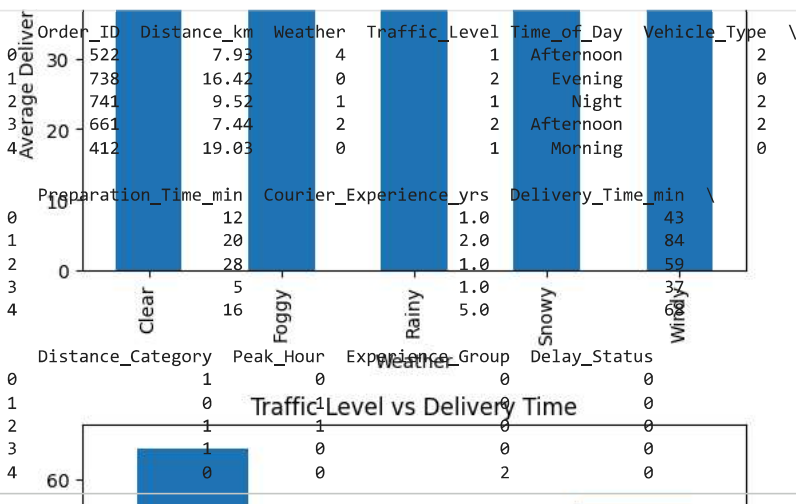
	Preparation_Time_min	Courier_Experience_yrs	Delivery_Time_min
0	12	1.0	43
1	20	2.0	84
2	28	1.0	59
3	5	1.0	37
4	16	5.0	68

	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
)
```