

## Problem Statement:

Uber has received some complaints from their customers facing problems related to ride cancellations by the driver and non-availability of cars for a specific route in the city.

The uneven supply-demand gap for cabs from City to Airport and vice-versa is causing a bad effect on customer relationships as well as Uber is losing out on its revenue.

The aim of analysis is to identify the root cause of the problem (i.e. cancellation and non-availability of cars) and recommend ways to tackle the situation.

```
In [1]: import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
import seaborn as sns
from scipy import stats
from sklearn.linear_model import LinearRegression
from sklearn.model_selection import train_test_split
from sklearn.metrics import mean_squared_error, r2_score
from scipy.stats import chi2_contingency
from sklearn.preprocessing import OneHotEncoder, LabelEncoder
```

```
In [2]: df = pd.read_csv('uber-data.csv', parse_dates=[4,5], dayfirst = True, na_values = "NA")
df
```

```
Out[2]:
```

	Request id	Pickup point	Driver id	Status	Request timestamp	Drop timestamp
0	619	Airport	1.0	Trip Completed	2016-07-11 11:51:00	2016-07-11 13:00:00
1	867	Airport	1.0	Trip Completed	2016-07-11 17:57:00	2016-07-11 18:47:00
2	1807	City	1.0	Trip Completed	2016-07-12 09:17:00	2016-07-12 09:58:00
3	2532	Airport	1.0	Trip Completed	2016-07-12 21:08:00	2016-07-12 22:03:00
4	3112	City	1.0	Trip Completed	2016-07-13 08:33:16	2016-07-13 09:25:47
...	...	...	...	...	...	...
6740	6745	City	NaN	No Cars Available	2016-07-15 23:49:03	NaT
6741	6752	Airport	NaN	No Cars Available	2016-07-15 23:50:05	NaT
6742	6751	City	NaN	No Cars Available	2016-07-15 23:52:06	NaT
6743	6754	City	NaN	No Cars Available	2016-07-15 23:54:39	NaT
6744	6753	Airport	NaN	No Cars Available	2016-07-15 23:55:03	NaT

6745 rows × 6 columns

```
In [3]: # key note here is, we see fee null and nan values in driver id and drop time stamp because:

# when user requests for cab and driver cancels it then the drop time stamp will be automatically null or missing
# And when no cabs available at the point of time, then automatically the driver id and drop time stamp column will be
# or missing

# Because of these reasons, we see null and missing values only in both of these columns

df.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 6745 entries, 0 to 6744
Data columns (total 6 columns):
#   Column                Non-Null Count  Dtype
---  -
0   Request id            6745 non-null  int64
1   Pickup point          6745 non-null  object
2   Driver id             4095 non-null  float64
3   Status                6745 non-null  object
4   Request timestamp     6745 non-null  datetime64[ns]
5   Drop timestamp        2831 non-null  datetime64[ns]
dtypes: datetime64[ns](2), float64(1), int64(1), object(2)
memory usage: 316.3+ KB
```

```
In [76]: # Checking for percentage of null values in each column

df.isnull().sum() / len(df) * 100
```

```
Out[76]: Request id            0.000000
Pickup point          0.000000
Driver id            39.288362
Status              0.000000
Request timestamp     0.000000
Drop timestamp       58.028169
Time_period          0.000000
Cab_Availability      0.000000
Estimated_Demand     0.000000
dtype: float64
```

In [74]: *# Categorizing the time stamp into multiple categories of day*

```
def categorize_timeperiod(hour):  
    if hour <= 4:  
        return 'Dawn'  
    elif hour <= 9 :  
        return 'Early Morning'  
    elif hour <= 16:  
        return 'Afternoon'  
    elif hour <= 21:  
        return 'Late Evening'  
    else:  
        return 'Night'
```

In [75]: `df['Time_period'] = df['Request timestamp'].dt.hour.apply(categorize_timeperiod)`  
`df['Time_period'].value_counts()`

Out[75]:

Late Evening	2342
Early Morning	2103
Afternoon	1224
Dawn	578
Night	498

Name: Time\_period, dtype: int64

In [103]: *# Checking at which hours of the day the cabs are unavailable*

*# we can get to know about this when we put a filter on Trip completed status*  
*# if Trip completed -> then cab is available, else if no cars available or cancelled then it means 'no cab available'*  
*# point of time*

```
def categorize_availability(availability):  
    if availability == 'Trip Completed':  
        return 'Cab Available'  
    else:  
        return 'No Cabs Available'
```

```
In [104]: df['Cab_Availability'] = df['Status'].apply(categorize_availability)
df
```

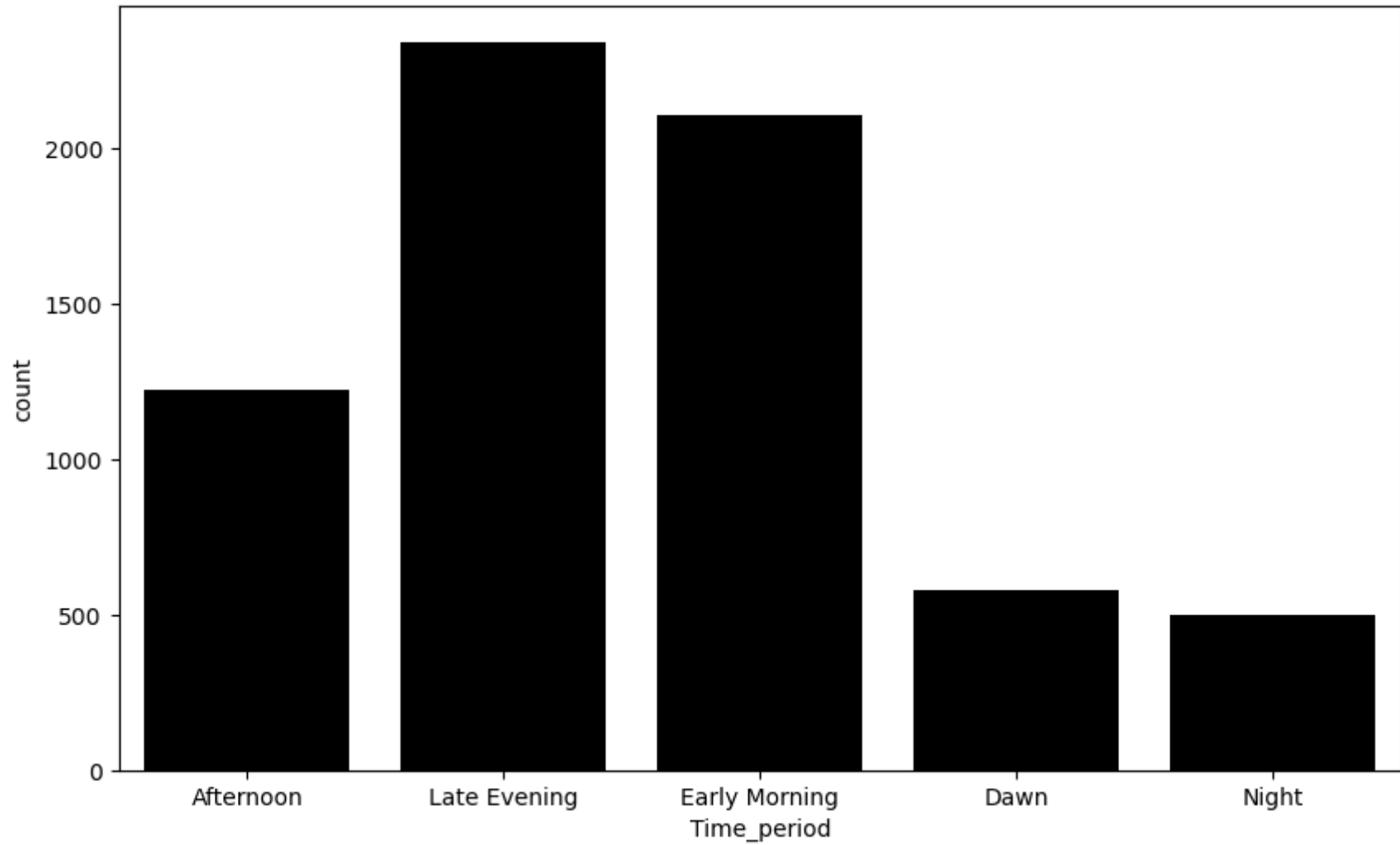
```
Out[104]:
```

	Request id	Pickup point	Driver id	Status	Request timestamp	Drop timestamp	Time_period	Cab_Availability	Estimated_Demand
0	619	Airport	1.0	Trip Completed	2016-07-11 11:51:00	2016-07-11 13:00:00	Afternoon	Cab Available	0
1	867	Airport	1.0	Trip Completed	2016-07-11 17:57:00	2016-07-11 18:47:00	Late Evening	Cab Available	0
2	1807	City	1.0	Trip Completed	2016-07-12 09:17:00	2016-07-12 09:58:00	Early Morning	Cab Available	0
3	2532	Airport	1.0	Trip Completed	2016-07-12 21:08:00	2016-07-12 22:03:00	Late Evening	Cab Available	0
4	3112	City	1.0	Trip Completed	2016-07-13 08:33:16	2016-07-13 09:25:47	Early Morning	Cab Available	0
...	...	...	...	...	...	...	...	...	...
6740	6745	City	NaN	No Cars Available	2016-07-15 23:49:03	NaT	Night	No Cabs Available	0
6741	6752	Airport	NaN	No Cars Available	2016-07-15 23:50:05	NaT	Night	No Cabs Available	0
6742	6751	City	NaN	No Cars Available	2016-07-15 23:52:06	NaT	Night	No Cabs Available	0
6743	6754	City	NaN	No Cars Available	2016-07-15 23:54:39	NaT	Night	No Cabs Available	0
6744	6753	Airport	NaN	No Cars Available	2016-07-15 23:55:03	NaT	Night	No Cabs Available	0

6745 rows × 9 columns

```
In [105]: # Categorizing the requesting time of users in a day per time period
```

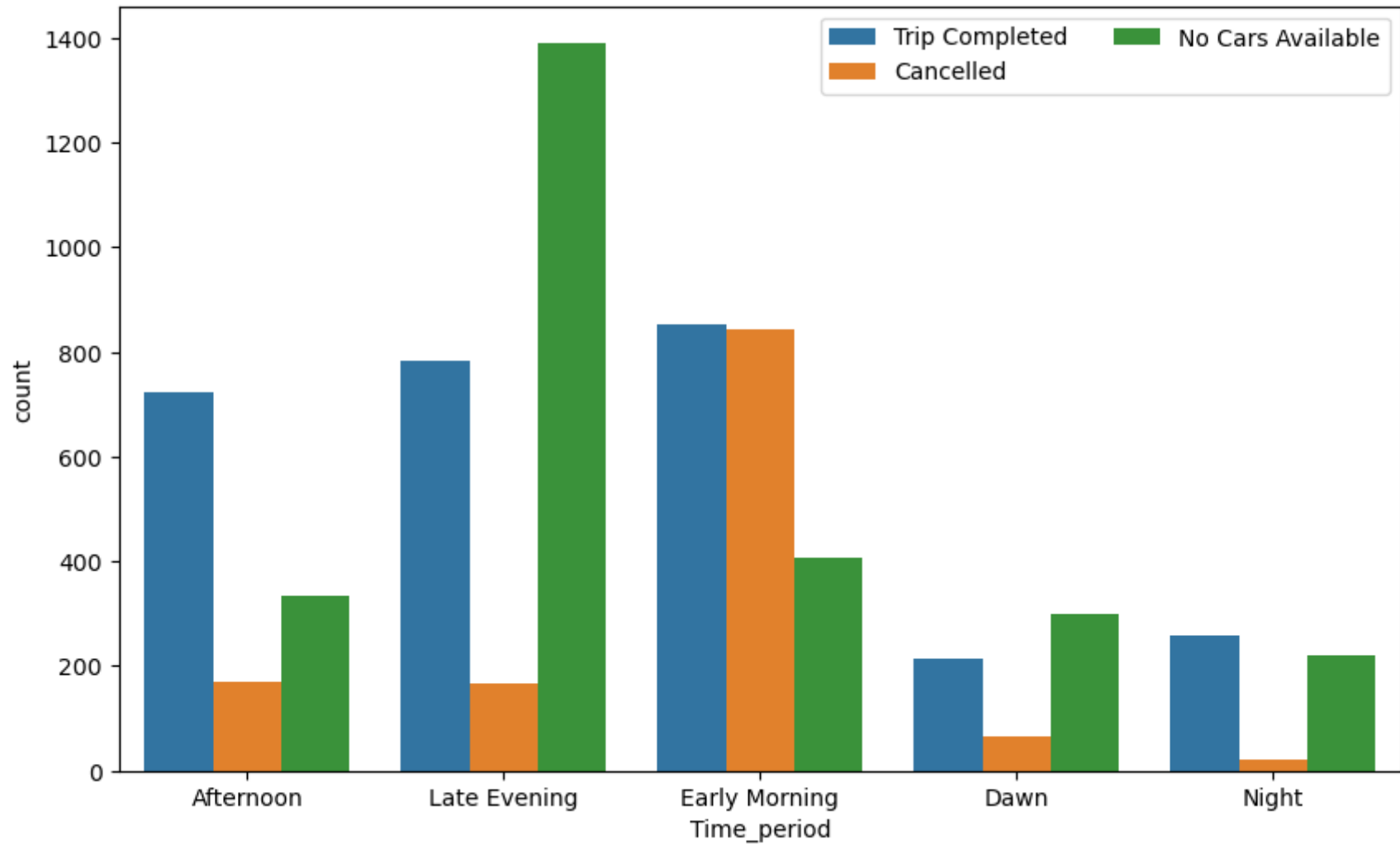
```
plt.figure(figsize = (10,6))  
sns.countplot(data = df, x = 'Time_period', palette=["#000000"])  
plt.show()
```



```
In [106]: # Categorizing the Cabs Availability per time period

# Blank indicates ride is completed
# Red indicates either cancelled or no cabs available at the point of time

plt.figure(figsize = (10,6))
sns.countplot(data = df, x = 'Time_period', hue = 'Status')
plt.legend(loc = 'upper right', frameon = True, ncol = 2)
plt.show()
```



In [107]: *# Morning Time Cab Availability*

```
df.loc[df['Time_period'] == 'Early Morning', 'Status'].value_counts() / len(df) * 100
```

Out[107]:

Trip Completed	12.661231
Cancelled	12.498147
No Cars Available	6.019274

Name: Status, dtype: float64



In [109]: *# Late Evening Cab Availability*

```
df.loc[df['Time_period'] == 'Late Evening', 'Status'].value_counts() / len(df) * 100
```

Out[109]: No Cars Available      20.637509  
Trip Completed            11.623425  
Cancelled                 2.461082  
Name: Status, dtype: float64

In [110]: *# Night time Cab Availability*

```
df.loc[df['Time_period'] == 'Night', 'Status'].value_counts() / len(df) * 100
```

Out[110]: Trip Completed        3.810230  
No Cars Available      3.246850  
Cancelled              0.326168  
Name: Status, dtype: float64

In [111]: *# Dawn time Cab Availability*

```
df.loc[df['Time_period'] == 'Dawn', 'Status'].value_counts() / len(df) * 100
```

Out[111]: No Cars Available      4.432913  
Trip Completed            3.172721  
Cancelled                 0.963677  
Name: Status, dtype: float64

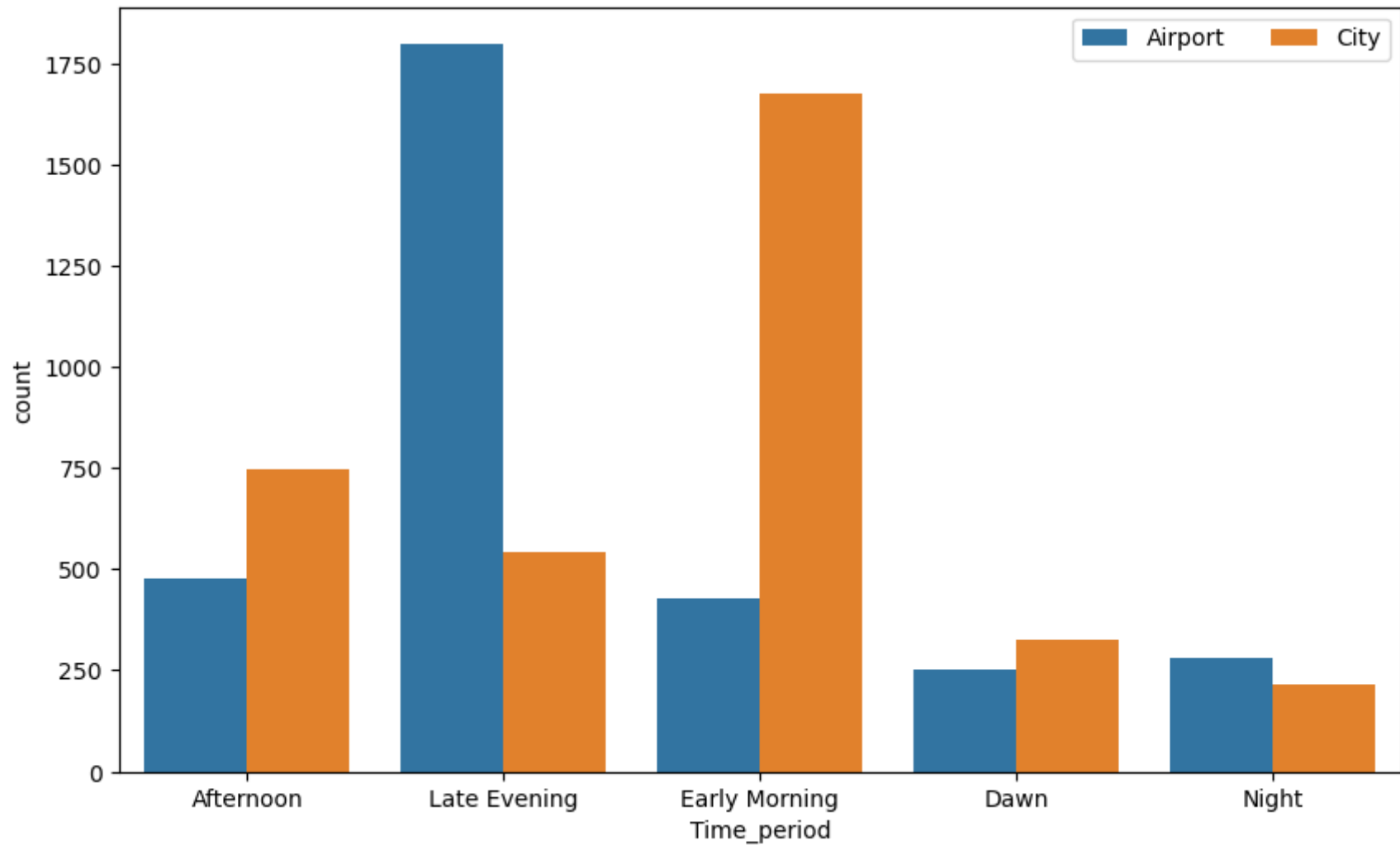
In [112]: *# Afternoon time Cab Availability*

```
df.loc[df['Time_period'] == 'Afternoon', 'Status'].value_counts() / len(df) * 100
```

Out[112]: Trip Completed        10.704225  
No Cars Available      4.951816  
Cancelled              2.490734  
Name: Status, dtype: float64

```
In [113]: # IN late evening, we recieve more ride requests from users travelling from Airport to City, while it is vice versa  
# in the Early Morning
```

```
plt.figure(figsize = (10,6))  
sns.countplot(data = df, x = 'Time_period', hue = 'Pickup point')  
plt.legend(loc = 'upper right', frameon = True, ncol = 2)  
plt.show()
```



In [114]: *# Filtering the commute time period of users going from City to Airport*

```
city_pickups = df.query("`Pickup point` == 'City'")
city_pickups
```

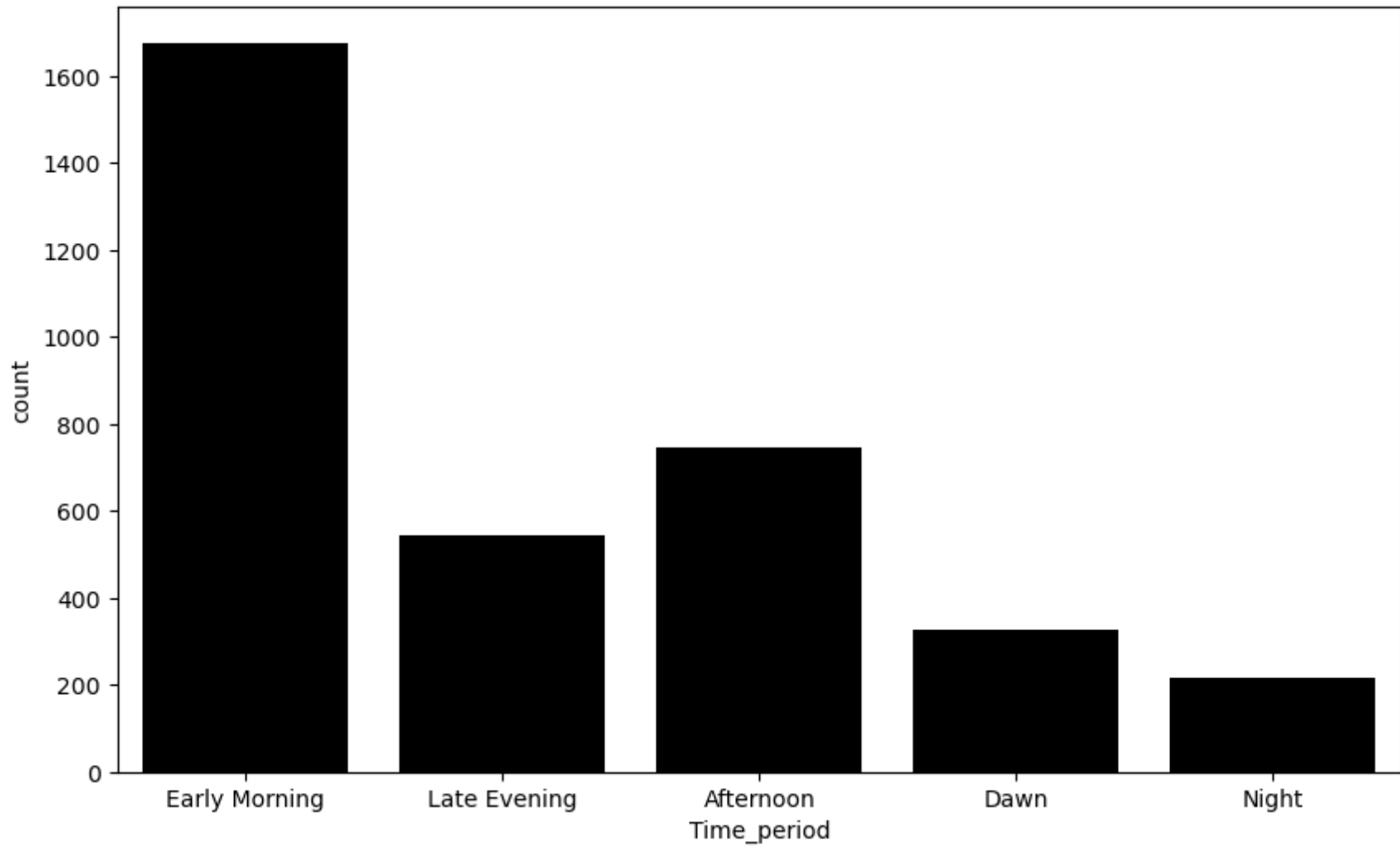
Out[114]:

	Request id	Pickup point	Driver id	Status	Request timestamp	Drop timestamp	Time_period	Cab_Availability	Estimated_Demand
<b>2</b>	1807	City	1.0	Trip Completed	2016-07-12 09:17:00	2016-07-12 09:58:00	Early Morning	Cab Available	0
<b>4</b>	3112	City	1.0	Trip Completed	2016-07-13 08:33:16	2016-07-13 09:25:47	Early Morning	Cab Available	0
<b>8</b>	6248	City	1.0	Trip Completed	2016-07-15 17:57:27	2016-07-15 18:50:51	Late Evening	Cab Available	0
<b>9</b>	267	City	2.0	Trip Completed	2016-07-11 06:46:00	2016-07-11 07:25:00	Early Morning	Cab Available	0
<b>11</b>	1983	City	2.0	Trip Completed	2016-07-12 12:30:00	2016-07-12 12:57:00	Afternoon	Cab Available	0
...	...	...	...	...	...	...	...	...	...
<b>6738</b>	6746	City	NaN	No Cars Available	2016-07-15 23:46:03	NaT	Night	No Cabs Available	0
<b>6739</b>	6739	City	NaN	No Cars Available	2016-07-15 23:46:20	NaT	Night	No Cabs Available	0
<b>6740</b>	6745	City	NaN	No Cars Available	2016-07-15 23:49:03	NaT	Night	No Cabs Available	0
<b>6742</b>	6751	City	NaN	No Cars Available	2016-07-15 23:52:06	NaT	Night	No Cabs Available	0
<b>6743</b>	6754	City	NaN	No Cars Available	2016-07-15 23:54:39	NaT	Night	No Cabs Available	0

3507 rows × 9 columns

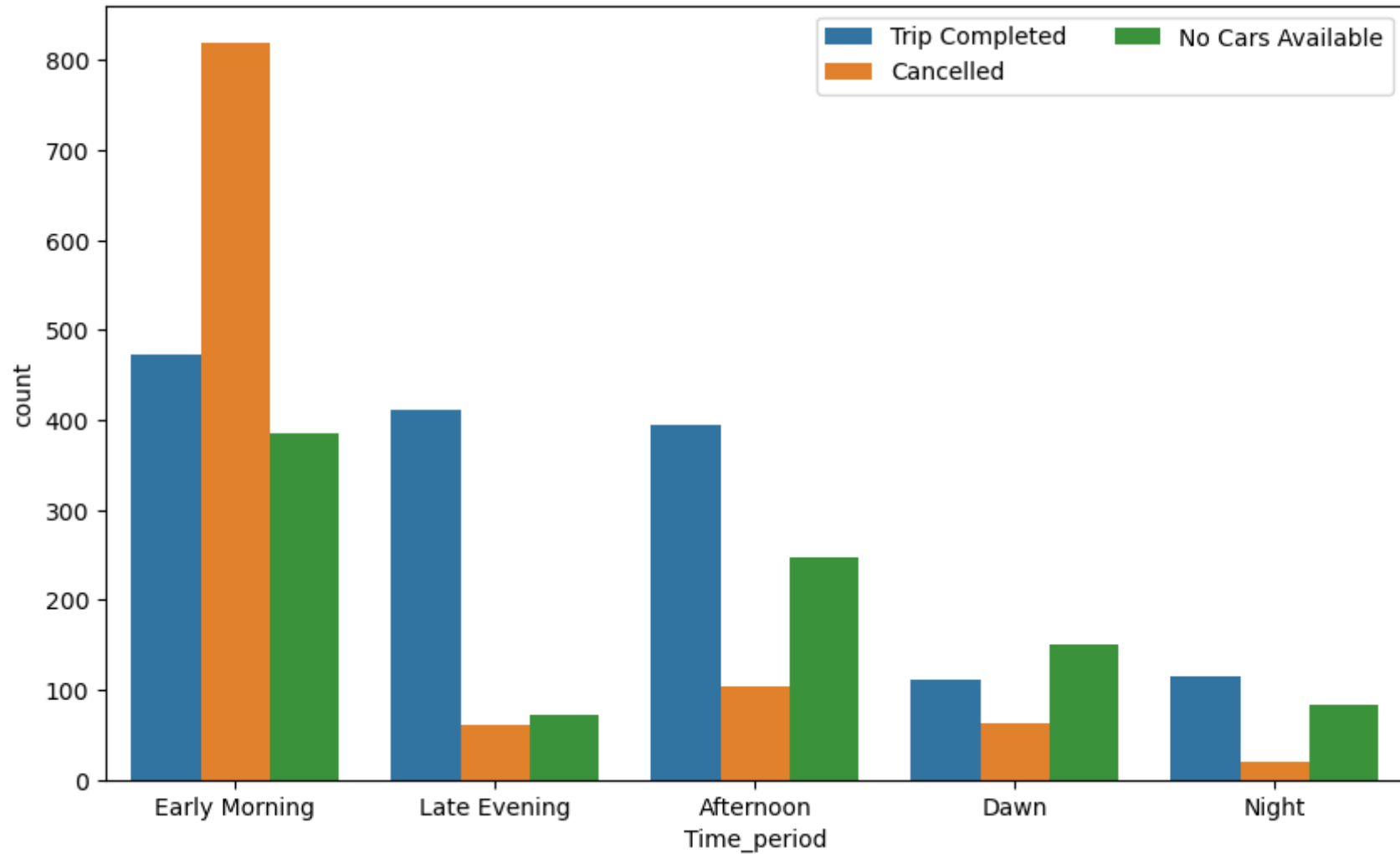
In [115]: *# Count of requests coming from users in a day when they opt to travel to Airport from the City*

```
plt.figure(figsize = (10,6))  
sns.countplot(data = city_pickups, x = 'Time_period', palette="#000000")  
plt.show()
```



In [159]: *# Cabs Availability in a day when users opt to travel to Airport from the City*

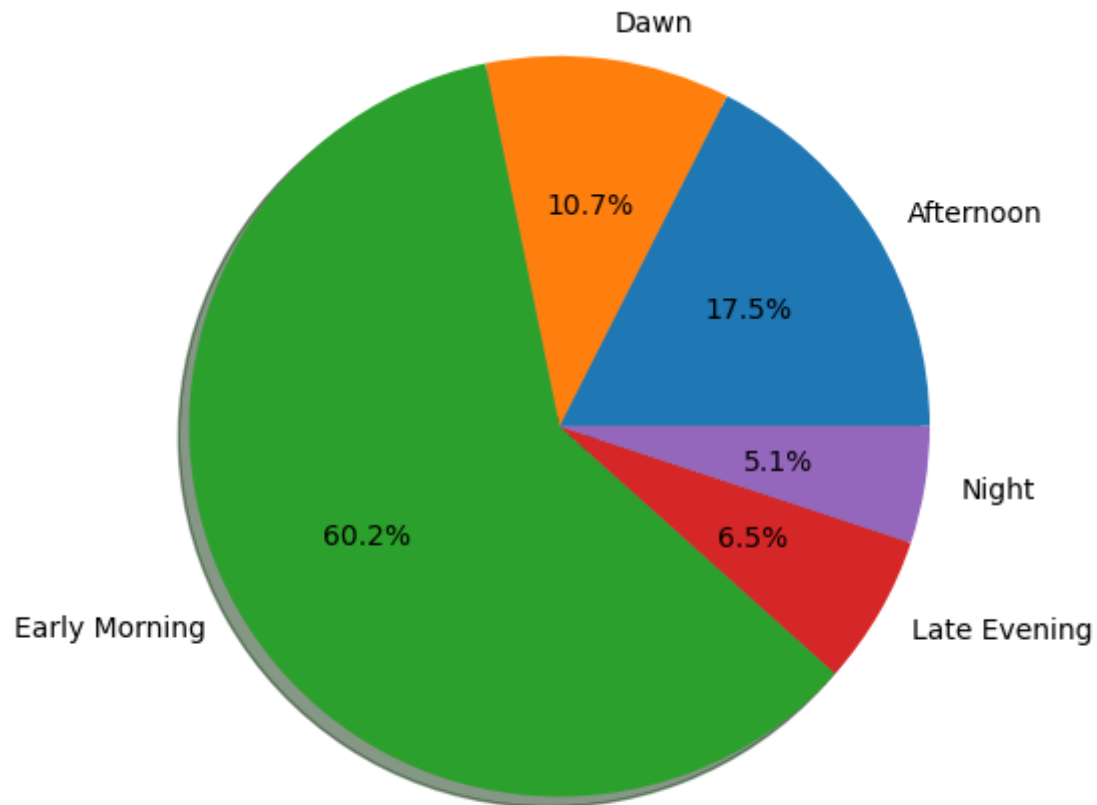
```
plt.figure(figsize = (10,6))  
sns.countplot(data = city_pickups, x = 'Time_period', hue = 'Status')  
plt.legend(loc = 'upper right', frameon = True, ncol = 2)  
plt.show()
```



In [142]: *# Percentage of City pickups where cabs are not available*

```
city_pickups[ (city_pickups["Cab_Availability"]=="No Cabs Available") ].groupby(['Time_period']).size().plot(kind="pie",  
plt.ylabel(""))
```

Out[142]: Text(0, 0.5, '')

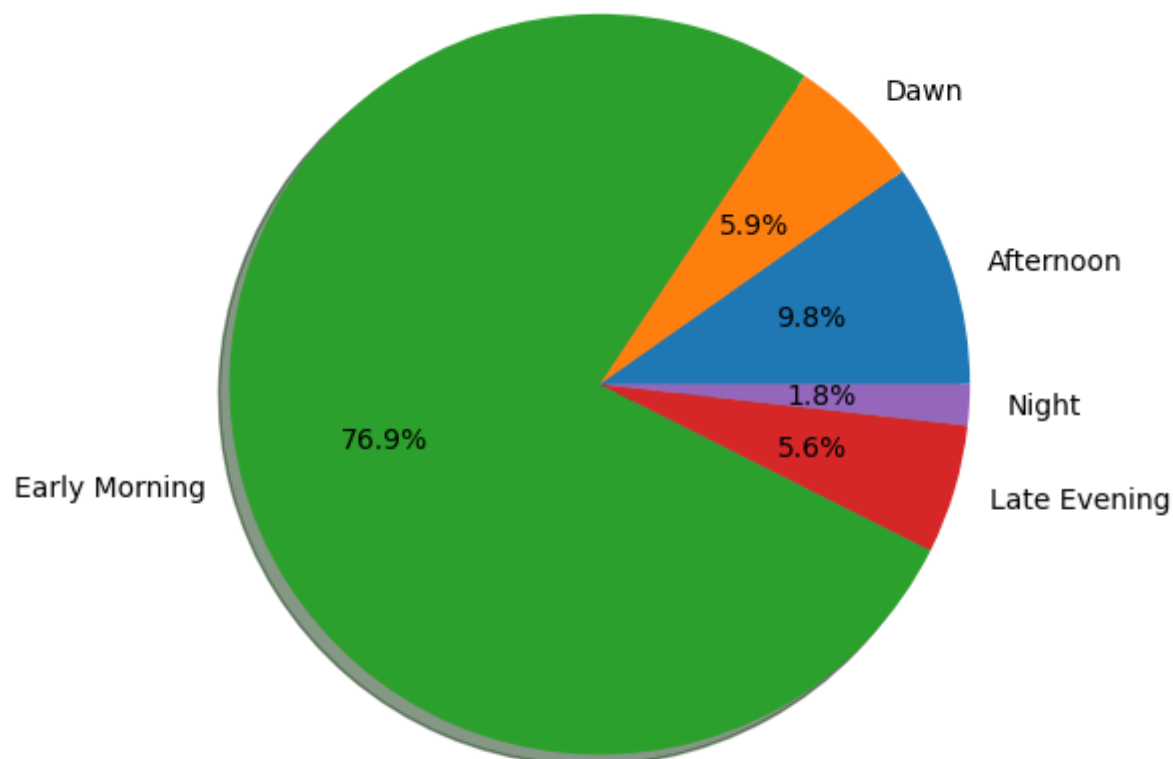


	Afternoon	Dawn	Early Morning	Late Evening	Night
None	351	214	1205	131	102

In [143]: *# Percentage of City pickups where drivers cancelled the rides*

```
city_pickups[ (city_pickups["Status"]=="Cancelled") ].groupby(['Time_period']).size().plot(kind="pie", figsize=(6, 6),  
plt.ylabel(""))
```

Out[143]: Text(0, 0.5, '')



	Afternoon	Dawn	Early Morning	Late Evening	Night
None	104	63	820	60	19

In [117]: *# Percentage of requests in Morning time from City to Airport*

```
city_pickups.loc[city_pickups['Time_period'] == 'Early Morning', 'Status'].value_counts() / len(df) * 100
```

Out[117]: Cancelled 12.157153  
Trip Completed 6.997776  
No Cars Available 5.707932  
Name: Status, dtype: float64

In [118]: *# Percentage of requests in Late Evening time from City to Airport*

```
city_pickups.loc[city_pickups['Time_period'] == 'Late Evening', 'Status'].value_counts() / len(df) * 100
```

Out[118]: Trip Completed 6.093403  
No Cars Available 1.052632  
Cancelled 0.889548  
Name: Status, dtype: float64

In [119]: *# Percentage of requests in Afternoon time from City to Airport*

```
city_pickups.loc[city_pickups['Time_period'] == 'Afternoon', 'Status'].value_counts() / len(df) * 100
```

Out[119]: Trip Completed 5.856190  
No Cars Available 3.661972  
Cancelled 1.541883  
Name: Status, dtype: float64

In [120]: *# Percentage of requests in Night time from City to Airport*

```
city_pickups.loc[city_pickups['Time_period'] == 'Night', 'Status'].value_counts() / len(df) * 100
```

Out[120]: Trip Completed 1.704967  
No Cars Available 1.230541  
Cancelled 0.281690  
Name: Status, dtype: float64



In [121]: *# Percentage of requests in Dawn time from City to Airport*

```
city_pickups.loc[city_pickups['Time_period'] == 'Dawn', 'Status'].value_counts() / len(df) * 100
```

Out[121]: No Cars Available      2.238695  
Trip Completed            1.645663  
Cancelled                0.934025  
Name: Status, dtype: float64

In [122]: *# Filtering the commute time period for users going from Airport to City*

```
airport_pickups = df.query("`Pickup point` == 'Airport' ")
airport_pickups
```

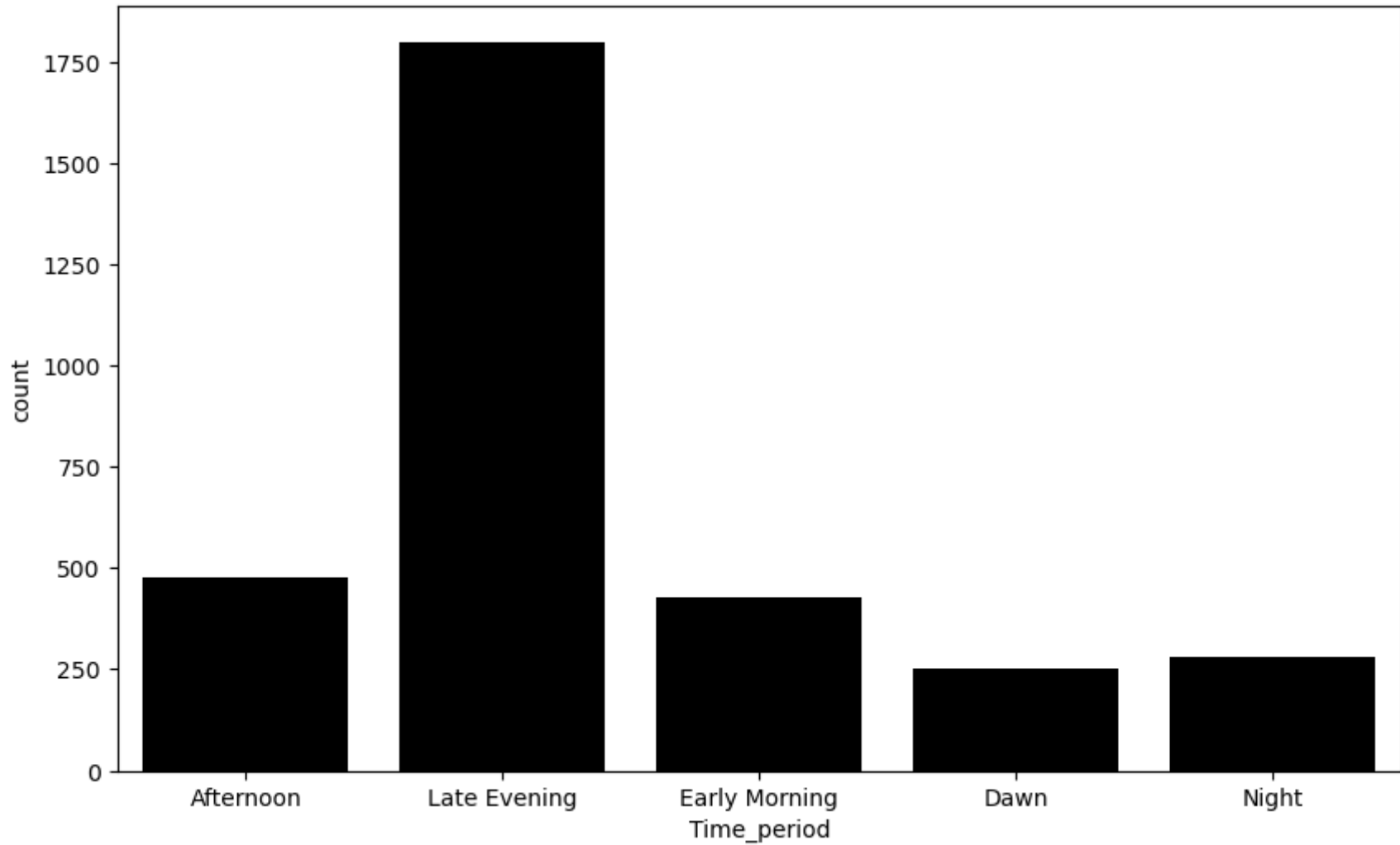
Out[122]:

	Request id	Pickup point	Driver id	Status	Request timestamp	Drop timestamp	Time_period	Cab_Availability	Estimated_Demand
0	619	Airport	1.0	Trip Completed	2016-07-11 11:51:00	2016-07-11 13:00:00	Afternoon	Cab Available	0
1	867	Airport	1.0	Trip Completed	2016-07-11 17:57:00	2016-07-11 18:47:00	Late Evening	Cab Available	0
3	2532	Airport	1.0	Trip Completed	2016-07-12 21:08:00	2016-07-12 22:03:00	Late Evening	Cab Available	0
5	3879	Airport	1.0	Trip Completed	2016-07-13 21:57:28	2016-07-13 22:28:59	Late Evening	Cab Available	0
6	4270	Airport	1.0	Trip Completed	2016-07-14 06:15:32	2016-07-14 07:13:15	Early Morning	Cab Available	0
...	...	...	...	...	...	...	...	...	...
6734	6732	Airport	NaN	No Cars Available	2016-07-15 23:35:50	NaT	Night	No Cabs Available	0
6735	6737	Airport	NaN	No Cars Available	2016-07-15 23:39:15	NaT	Night	No Cabs Available	0
6736	6744	Airport	NaN	No Cars Available	2016-07-15 23:42:51	NaT	Night	No Cabs Available	0
6741	6752	Airport	NaN	No Cars Available	2016-07-15 23:50:05	NaT	Night	No Cabs Available	0
6744	6753	Airport	NaN	No Cars Available	2016-07-15 23:55:03	NaT	Night	No Cabs Available	0

3238 rows × 9 columns

In [123]: *# Count of requests coming from users in a day when they opt to travel to from Airport to City*

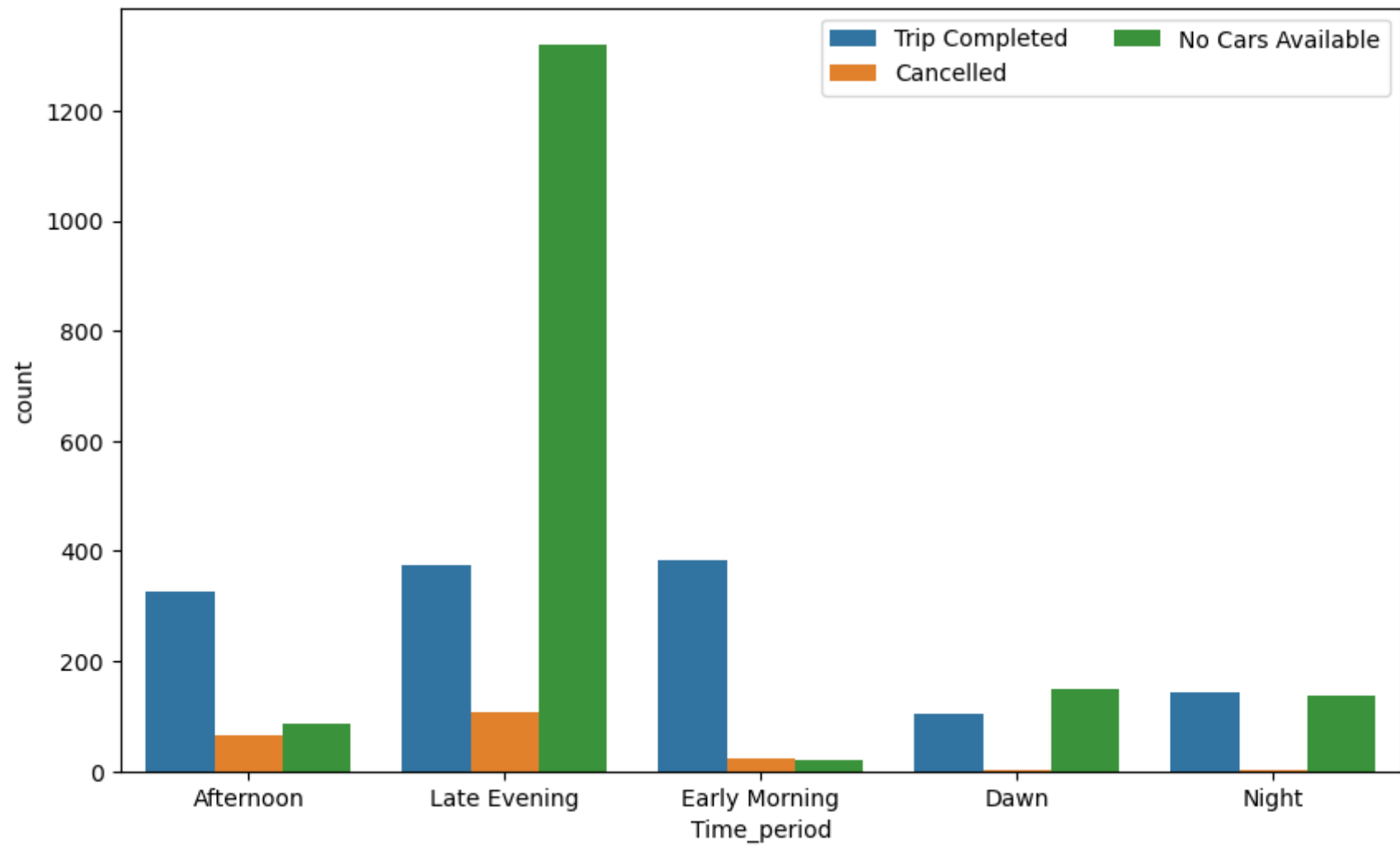
```
plt.figure(figsize = (10,6))  
sns.countplot(data = airport_pickups, x = 'Time_period', palette=["#000000"])  
plt.show()
```



```
In [124]: # Cabs Availability in a day when users opt to travel to City from Airport

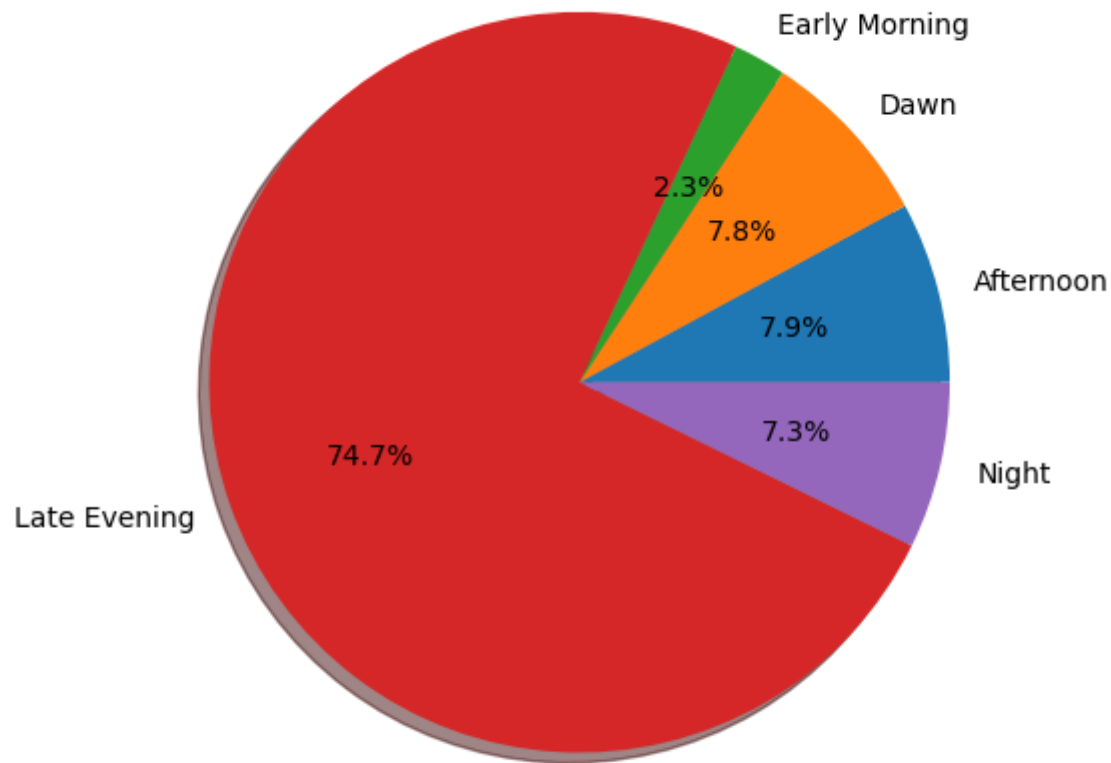
# Blank indicates ride is completed
# Red indicates either cancelled or no cabs available at the point of time

plt.figure(figsize = (10,6))
sns.countplot(data = airport_pickups, x = 'Time_period', hue = 'Status' )
plt.legend(loc = 'upper right', frameon = True, ncol = 2)
plt.show()
```



In [149]: *# Percentage of Airport pickups where cabs are not available*

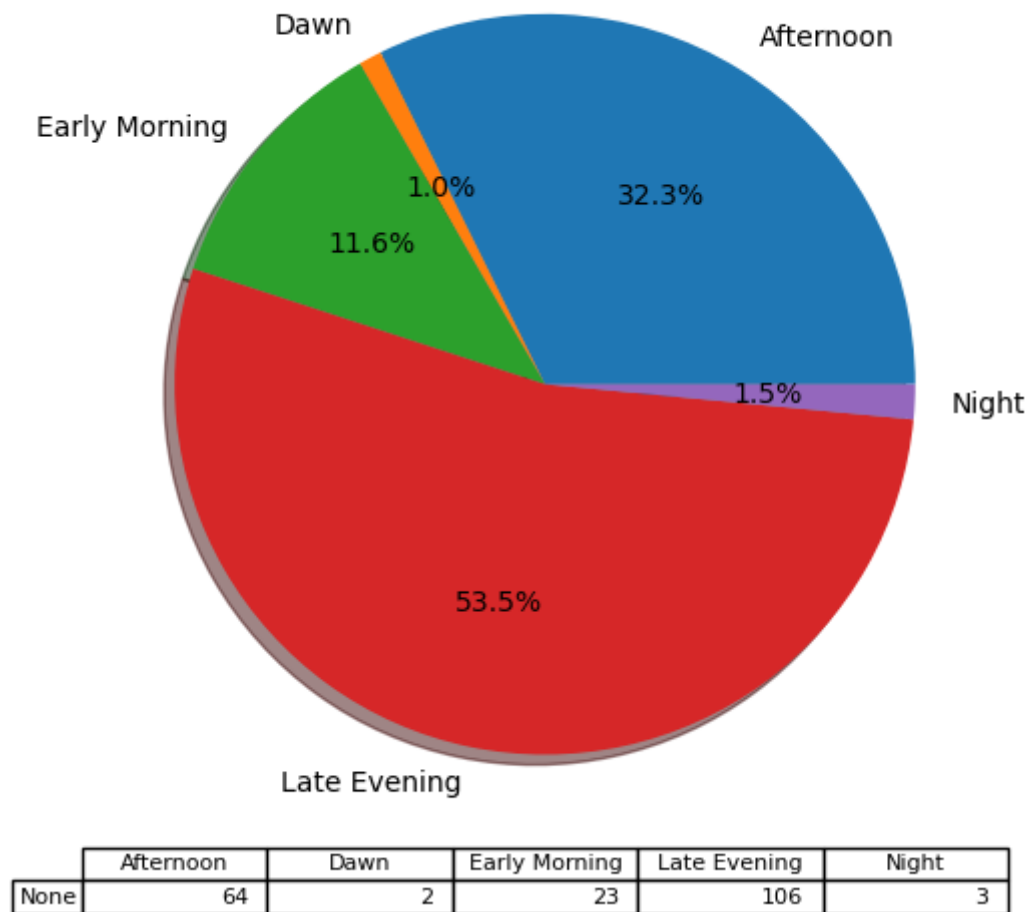
```
airport_pickups[ (airport_pickups["Cab_Availability"]=="No Cabs Available") ].groupby(['Time_period']).size().plot(kind='pie',\nplt.ylabel(''))\nplt.show()
```



	Afternoon	Dawn	Early Morning	Late Evening	Night
None	151	150	44	1427	139

In [158]: *# Percentage of Airport pickups where drivers cancelled the ride*

```
airport_pickups[ (airport_pickups["Status"]=="Cancelled") ].groupby(['Time_period']).size().plot(kind="pie", figsize=(  
plt.ylabel("")  
plt.show()
```



In [125]: *# Percentage of requests in Morning time from Airport to City*

```
airport_pickups.loc[airport_pickups['Time_period'] == 'Early Morning', 'Status'].value_counts() / len(df) * 100
```

Out[125]:

Trip Completed	5.663454
Cancelled	0.340993
No Cars Available	0.311342

Name: Status, dtype: float64

In [126]: *# Percentage of requests in Late Evening time from Airport to City*

```
airport_pickups.loc[airport_pickups['Time_period'] == 'Late Evening', 'Status'].value_counts() / len(df) * 100
```

Out[126]:

No Cars Available	19.584878
Trip Completed	5.530022
Cancelled	1.571534

Name: Status, dtype: float64

In [127]: *# Percentage of requests in Night time from Airport to City*

```
airport_pickups.loc[airport_pickups['Time_period'] == 'Night', 'Status'].value_counts() / len(df) * 100
```

Out[127]:

Trip Completed	2.105263
No Cars Available	2.016308
Cancelled	0.044477

Name: Status, dtype: float64

In [128]: *# Percentage of requests in Dawn time from Airport to City*

```
airport_pickups.loc[airport_pickups['Time_period'] == 'Dawn', 'Status'].value_counts() / len(df) * 100
```

Out[128]:

No Cars Available	2.194218
Trip Completed	1.527057
Cancelled	0.029652

Name: Status, dtype: float64



In [129]: *# Percentage of requests in Afternoon time from Airport to City*

```
airport_pickups.loc[airport_pickups['Time_period'] == 'Afternoon', 'Status'].value_counts() / len(df) * 100
```

Out[129]:

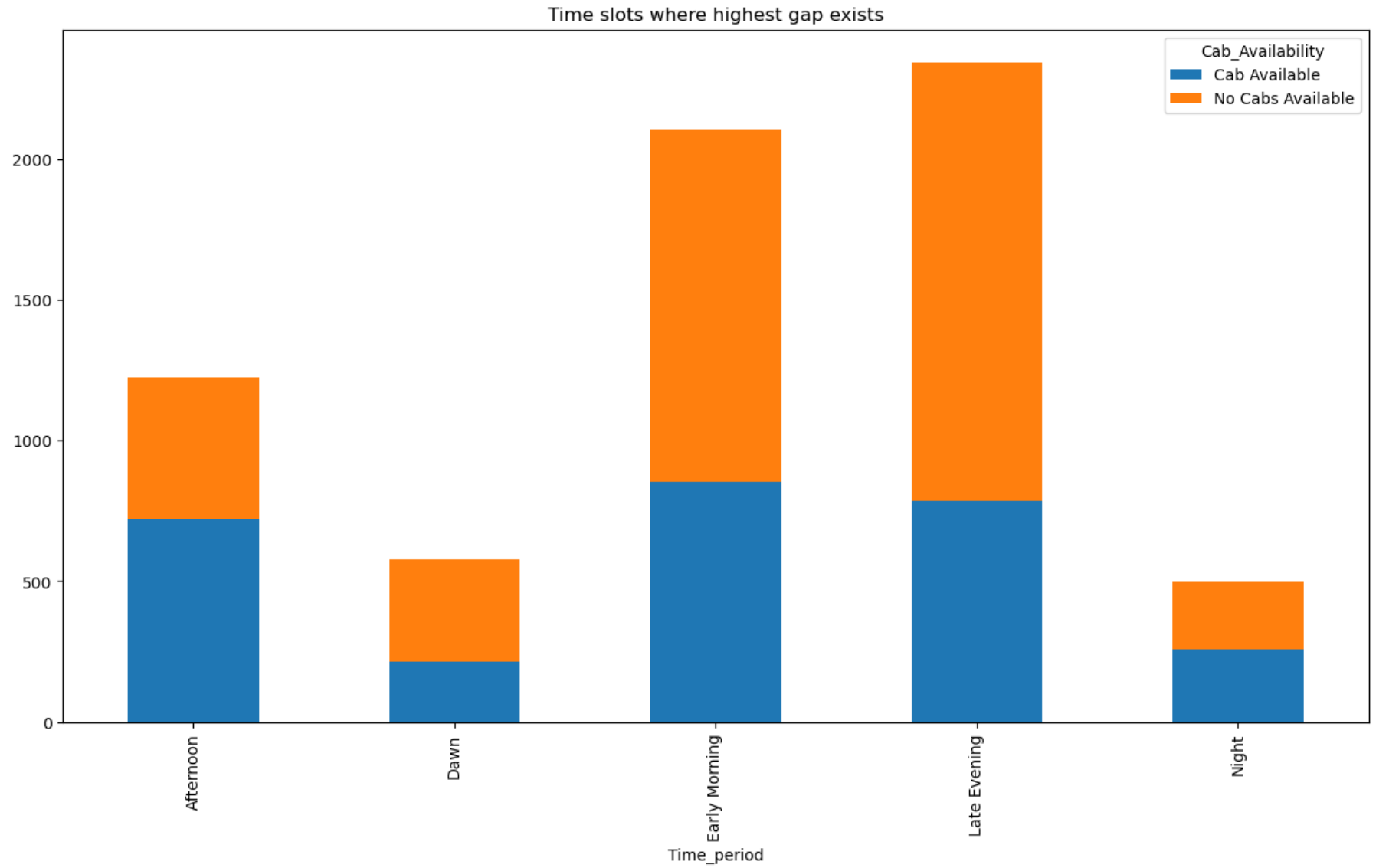
Trip Completed	4.848036
No Cars Available	1.289844
Cancelled	0.948851

Name: Status, dtype: float64

In [167]: *# Highest gap between the rides*

*# Time slots where highest gap exists -*

```
df.groupby(['Time_period', 'Cab_Availability']).size().unstack().plot(kind='bar', stacked=True, figsize=(15, 8))  
plt.title('Time slots where highest gap exists')  
plt.show()
```



```
In [168]: df.groupby(['Time_period', 'Cab_Availability']).size().unstack()
```

```
Out[168]:
```

	Cab_Availability	Cab Available	No Cabs Available
Time_period			
Afternoon		722	502
Dawn		214	364
Early Morning		854	1249
Late Evening		784	1558
Night		257	241

```
In [ ]: # Unstack and Stack

# Unstack function pivots the data, transforming grouped series into a data frame where the cab availability value becomes
# multiple columns and Time period column becomes the data frame index.
# This operation effectively separates the counts of cab availability into separate columns allowing use to plot a stacked
# bar chart with each column representing a different cab availability value

# On the other side stack cannot be used here because
# if we want to use stack we should create a data frame first with the result of group by or that sort and then use stack
# that data frame. The stack method is used to pivot the data frame from a wider format(multi level columns) to long format
# creating a multi-level index

# Using stack in this context would not be appropriate for creating a stacked bar chart since it would create a multi-level
# making the plot more complicated. In most cases, you would want to use unstack when preparing data for a stacked bar chart
# have each category (in this case, 'Cab_Availability') represented by a separate column in the resulting DataFrame, and then
# plot the stacked bars accordingly

df.groupby(['Time_period', 'Cab_Availability']).size().stack().plot(kind='bar', stacked=True, figsize=(15, 8))
```

```
In [130]: # Checking for a relation between Time period and Cab Availability ?  
# Are they linearly dependent on each other ?  
  
# Assuming 'Cab Availability' is the column containing Cab availability in the Uber DataFrame.  
# 'Time period' is the column containing information on desired requests time in a day.  
  
# Since we have two object based columns i.e strings, we can use Chi-square test.  
# For example:  
  
# Create a contingency table  
contingency_table = pd.crosstab(df['Time_period'], df['Cab_Availability'])  
  
# Perform the Chi-Square test  
chi2, p_value, dof, expected = chi2_contingency(contingency_table)  
  
# Define the significance level (alpha)  
alpha = 0.05  
  
# Check the p-value against the significance level to make a decision  
if p_value < alpha:  
    print("Reject the null hypothesis. Time is significantly related to Cab Availability.")  
else:  
    print("Fail to reject the null hypothesis. Time does not significantly improve Cab Availability")
```

Reject the null hypothesis. Time is significantly related to Cab Availability.

In [33]: df

Out[33]:

	Request id	Pickup point	Driver id	Status	Request timestamp	Drop timestamp	Time_period	Cab_Availability
0	619	Airport	1.0	Trip Completed	2016-07-11 11:51:00	2016-07-11 13:00:00	Early Morning	Cab Available
1	867	Airport	1.0	Trip Completed	2016-07-11 17:57:00	2016-07-11 18:47:00	Late Evening	Cab Available
2	1807	City	1.0	Trip Completed	2016-07-12 09:17:00	2016-07-12 09:58:00	Early Morning	Cab Available
3	2532	Airport	1.0	Trip Completed	2016-07-12 21:08:00	2016-07-12 22:03:00	Night	Cab Available
4	3112	City	1.0	Trip Completed	2016-07-13 08:33:16	2016-07-13 09:25:47	Early Morning	Cab Available
...	...	...	...	...	...	...	...	...
6740	6745	City	NaN	No Cars Available	2016-07-15 23:49:03	NaT	Night	No Cabs Available
6741	6752	Airport	NaN	No Cars Available	2016-07-15 23:50:05	NaT	Night	No Cabs Available
6742	6751	City	NaN	No Cars Available	2016-07-15 23:52:06	NaT	Night	No Cabs Available
6743	6754	City	NaN	No Cars Available	2016-07-15 23:54:39	NaT	Night	No Cabs Available
6744	6753	Airport	NaN	No Cars Available	2016-07-15 23:55:03	NaT	Night	No Cabs Available

6745 rows × 8 columns

In [178]: *# Overall Scenario*

*# How many cabs should be ideally available according to the Time period to suffice the customer complaints*

*#Calculate the estimated demand for cabs during each time period*

*# Count the number of rides (demand) during each time period*

```
supply_by_time = df[df['Cab_Availability'] == 'Cab Available'].groupby('Time_period').size().reset_index(name='Supply')
```

*# Count the number of non-availability (supply) during each time period*

```
demand_by_time = df[df['Cab_Availability'] == 'No Cabs Available'].groupby('Time_period').size().reset_index(name='Demand')
```

```
merged_df = pd.merge(demand_by_time, supply_by_time, on='Time_period', how='outer').fillna(0)
```

```
merged_df['Demand_Supply_Gap'] = merged_df['Demand'] - merged_df['Supply']
```

```
print(merged_df)
```

```
plt.figure(figsize=(10, 6))
```

```
plt.bar(merged_df['Time_period'], merged_df['Demand_Supply_Gap'], label='Demand-Supply Gap')
```

```
plt.axhline(0, color='red', linestyle='--', label='Zero Gap (Ideal)')
```

```
plt.xlabel('Time Period')
```

```
plt.ylabel('Demand-Supply Gap')
```

```
plt.title('Demand-Supply Gap for Cabs')
```

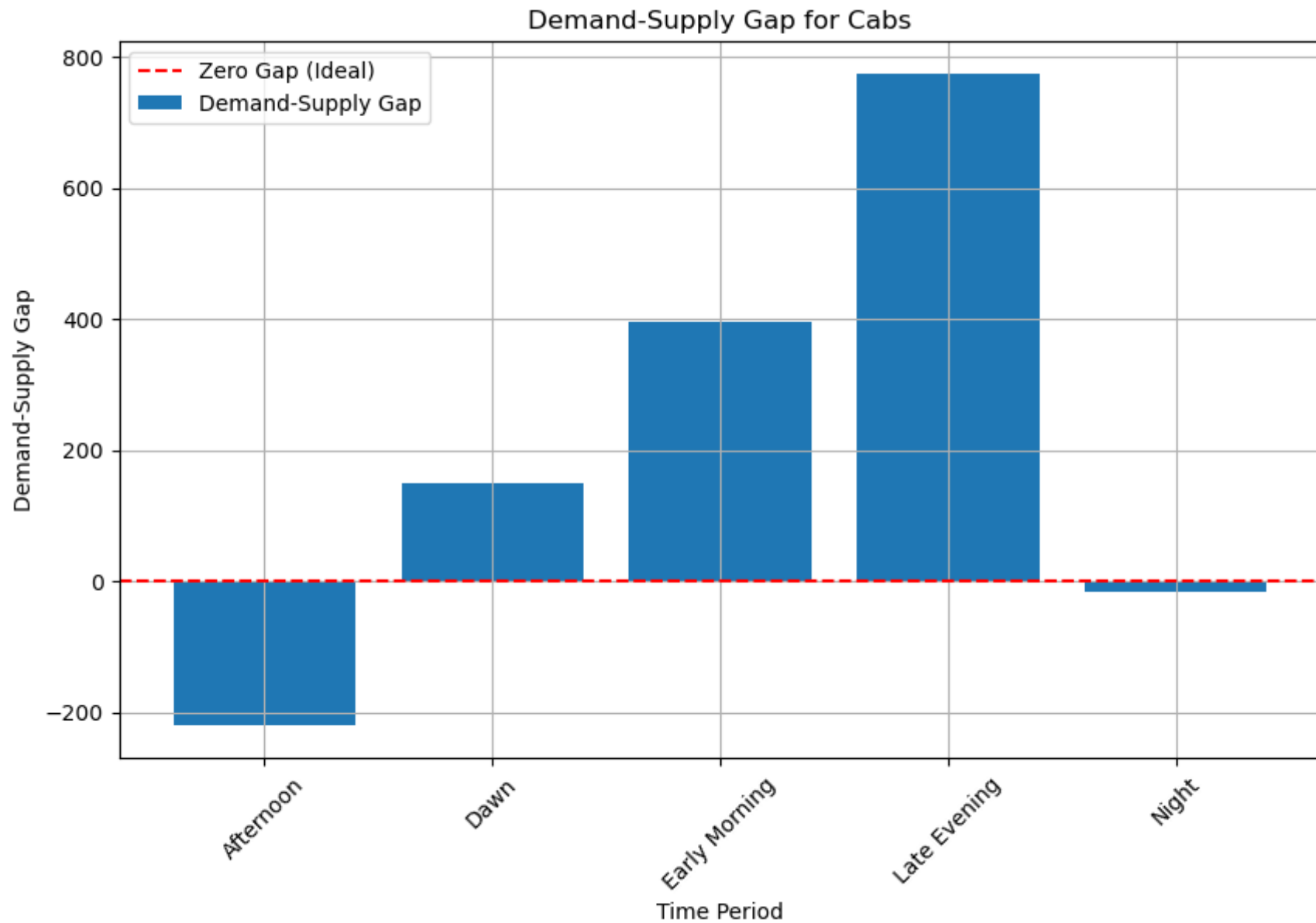
```
plt.legend()
```

```
plt.xticks(rotation=45)
```

```
plt.grid(True)
```

```
plt.show()
```

	Time_period	Demand	Supply	Demand_Supply_Gap
0	Afternoon	502	722	-220
1	Dawn	364	214	150
2	Early Morning	1249	854	395
3	Late Evening	1558	784	774
4	Night	241	257	-16





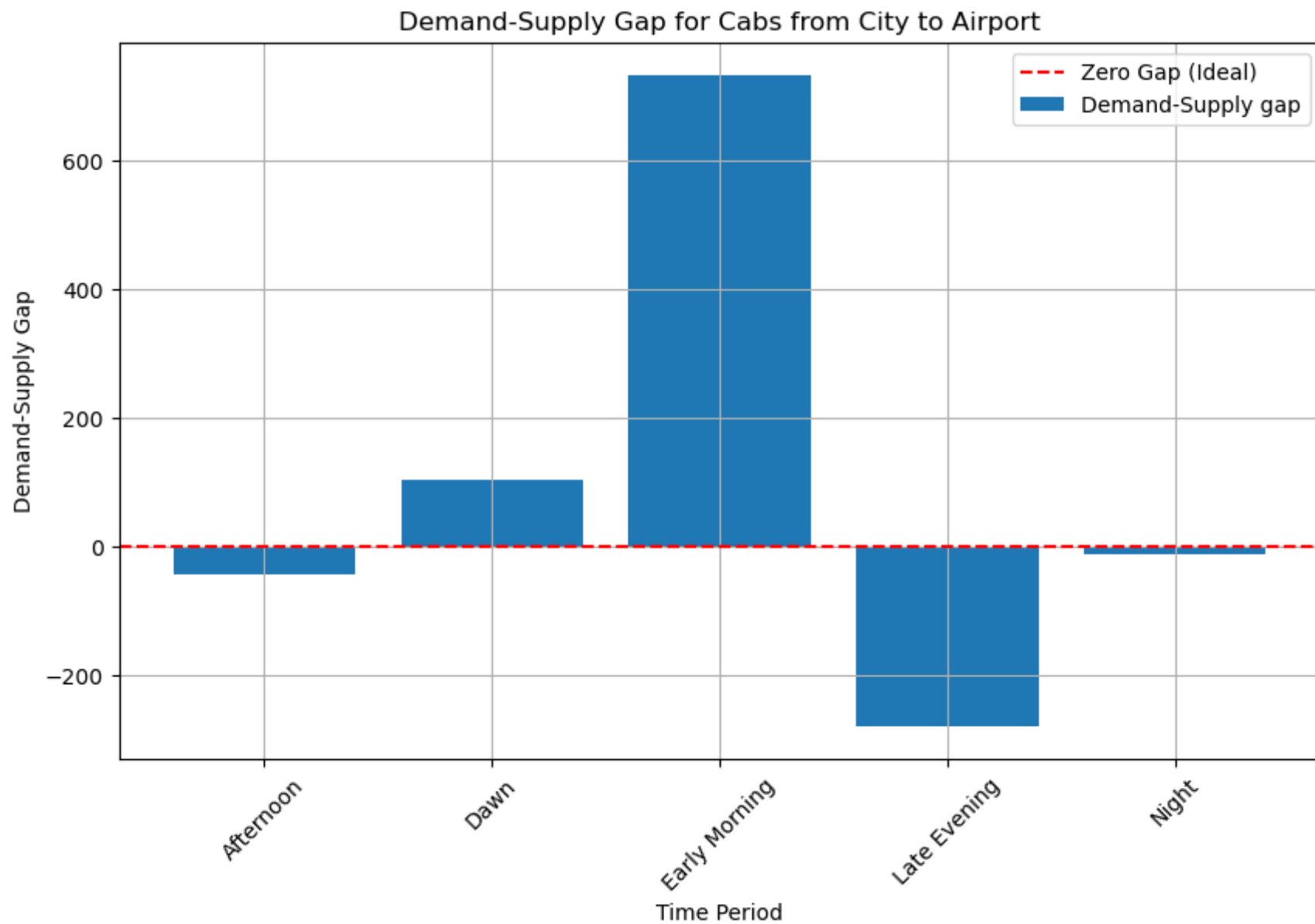
In [189]: *# City Pickup Scenario*

```
demand_surge = city_pickups[city_pickups['Cab_Availability'] == 'No Cabs Available'].groupby('Time_period').size().reset_index()
supply_provided = city_pickups[city_pickups['Cab_Availability'] == 'Cab Available'].groupby('Time_period').size().reset_index()
merged_df = pd.merge(demand_surge, supply_provided, on = 'Time_period', how = 'outer').fillna(0)
merged_df['Demand_Supply_gap'] = merged_df['Demand'] - merged_df['Supply']

print(merged_df)


plt.figure(figsize=(10, 6))
plt.bar(merged_df['Time_period'], merged_df['Demand_Supply_gap'], label='Demand-Supply gap')
plt.axhline(0, color='red', linestyle='--', label='Zero Gap (Ideal)')
plt.xlabel('Time Period')
plt.ylabel('Demand-Supply Gap')
plt.title('Demand-Supply Gap for Cabs from City to Airport')
plt.legend()
plt.xticks(rotation=45)
plt.grid(True)
plt.show()
```

	Time_period	Demand	Supply	Demand_Supply_gap
0	Afternoon	351	395	-44
1	Dawn	214	111	103
2	Early Morning	1205	472	733
3	Late Evening	131	411	-280
4	Night	102	115	-13



In [187]: *# Explore on reset index with name paramter*

```
# demand_surge = city_pickups[city_pickups['Cab_Availability'] == 'No Cabs Available'].groupby('Time_period').size().r  
# demand_surge
```



In [188]: *# Airport Pickup Scenario*

```
demand_surge = airport_pickups[airport_pickups['Cab_Availability'] == 'No Cabs Available'].groupby('Time_period').size
supply_provided = airport_pickups[airport_pickups['Cab_Availability'] == 'Cab Available'].groupby('Time_period').size
merged_df = pd.merge(demand_surge, supply_provided, on = 'Time_period', how = 'outer').fillna(0)
merged_df['Demand_Supply_gap'] = merged_df['Demand'] - merged_df['Supply']

print(merged_df)

plt.figure(figsize=(10, 6))
plt.bar(merged_df['Time_period'], merged_df['Demand_Supply_gap'], label='Demand-Supply gap')
plt.axhline(0, color='red', linestyle='--', label='Zero Gap (Ideal)')
plt.xlabel('Time Period')
plt.ylabel('Demand-Supply Gap')
plt.title('Demand-Supply Gap for Cabs from Airport to City')
plt.legend()
plt.xticks(rotation=45)
plt.grid(True)
plt.show()
```

	Time_period	Demand	Supply	Demand_Supply_gap
0	Afternoon	151	327	-176
1	Dawn	150	103	47
2	Early Morning	44	382	-338
3	Late Evening	1427	373	1054
4	Night	139	142	-3

