

Citi_bike_final

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1 ORIE 5530 Midterm Project

Team Member: Dahong Zhang, Zimu Li, Gareth Sun

NetID: zs389, zl898, dz285

1.1 Warm Up Question

Data Load and Pre-process

```
[ ]: import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
import seaborn as sns
from collections import Counter
```

```
[ ]: df = pd.read_csv("202307-citibike-tripdata.csv")
df.head()
```

/var/folders/ck/b09qsxz518x4rlt33mrt531c0000gn/T/ipykernel_1070/2021628630.py:1:
DtypeWarning: Columns (5,7) have mixed types. Specify dtype option on import or
set low_memory=False.

```
df = pd.read_csv("202307-citibike-tripdata.csv")
```

```
[ ]:      ride_id rideable_type      started_at      ended_at \
0  BC78328130CD8A16  classic_bike  2023-07-16 20:42:57  2023-07-16 20:45:13
1  71A4DFA231ADAEA2  classic_bike  2023-07-29 12:22:21  2023-07-29 12:26:54
2  B64451D0D9F52134  classic_bike  2023-07-30 18:30:18  2023-07-30 18:35:42
3  EEECB3E0E05EB0DB  classic_bike  2023-07-04 19:59:20  2023-07-04 20:02:03
4  76B706DF4FC180DC  classic_bike  2023-07-03 19:33:36  2023-07-03 19:43:59
```

```
      start_station_name start_station_id \
0      St. Nicholas Ave & W 126 St      7756.10
1  Adam Clayton Powell Blvd & W 118 St      7670.09
2      St. Nicholas Ave & W 126 St      7756.10
3      St. Nicholas Ave & W 126 St      7756.10
4      St. Nicholas Ave & W 126 St      7756.10
```

```
      end_station_name end_station_id start_lat start_lng \
```

0	Adam Clayton Powell Blvd & W 126 St	7738.04	40.811432	-73.951878
1	Adam Clayton Powell Blvd & W 126 St	7738.04	40.804372	-73.951475
2	Lenox Ave & W 130 St	7753.13	40.811378	-73.951827
3	Adam Clayton Powell Blvd & W 126 St	7738.04	40.811432	-73.951878
4	Pleasant Ave & E 116 St	7450.05	40.811543	-73.951813

	end_lat	end_lng	member_casual
0	40.809495	-73.947765	member
1	40.809495	-73.947765	member
2	40.810792	-73.943068	member
3	40.809495	-73.947765	member
4	40.794988	-73.933335	member

```
[ ]: # Calculate the Time Duration
df['starttime'] = pd.to_datetime(df['started_at'])
df['stoptime'] = pd.to_datetime(df['ended_at'])
df['time_duration'] = df['stoptime'] - df['starttime']

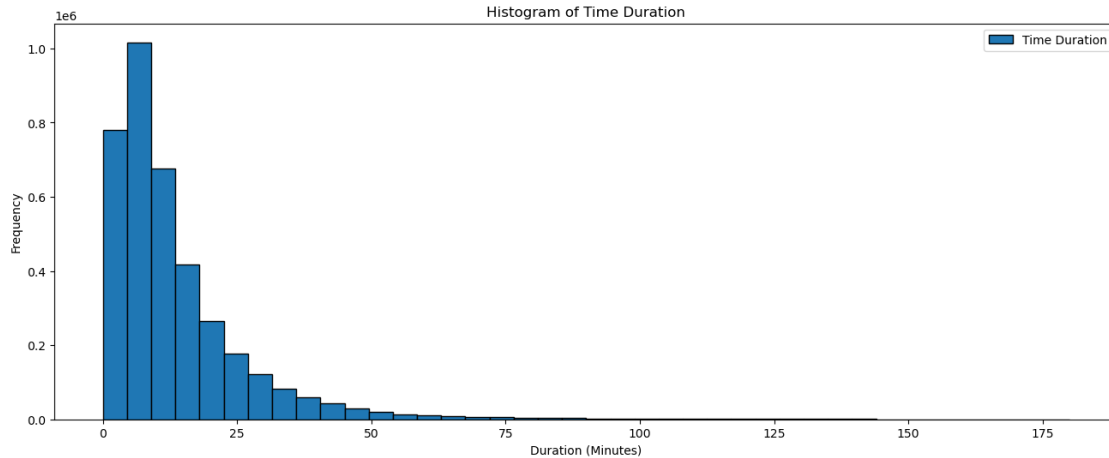
# Data Pre-process
df = df[df['time_duration'] >= pd.Timedelta(hours=0)]
df = df[df['time_duration'] <= pd.Timedelta(hours=3)]
print(df.size)
```

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1.1.1 1. Compute the duration of each ride in minutes and plot the histogram of ride durations.

```
[ ]: time_plot = pd.to_timedelta(df['time_duration'])
time_plot = df['time_duration'].dt.total_seconds() / 60

plt.figure(figsize=(16,6))
plt.hist(time_plot, bins=40, edgecolor='black', label='Time Duration')
plt.xlabel('Duration (Minutes)')
plt.ylabel('Frequency')
plt.title('Histogram of Time Duration')
plt.legend()
plt.show()
```



1.1.2 2. Expected Duration and Variance

Expected ride duration

```
[ ]: avg_time = df['time_duration'].mean()
      avg_time_in_min = avg_time.total_seconds() / 60
      avg_time_in_min = np.round(avg_time_in_min, 2)
      print(f"Expected ride duration is {avg_time_in_min} mins.")
```

Expected ride duration is 13.65 mins.

Empirical variance of ride duration

```
[ ]: df['time_duration_seconds'] = df['time_duration'].dt.total_seconds() / 60
      variance = df['time_duration_seconds'].var()
      variance = np.round(variance, 2)
      print(f"Variance of minutes is {variance}.")
```

Variance of minutes is 203.91.

Rider Duration Longer than 20 Mins

```
[ ]: long_riders = df[ df['time_duration'] >= pd.Timedelta(minutes=20) ]
      probability = len(long_riders) / len(df)
      print(f"The probability that a ride duration is greater than 20 min is_
            ↳ {probability * 100}%")
```

The probability that a ride duration is greater than 20 min is 19.775749812505392%

1.1.3 3. Probability of Ride Duration Greater than 20 Mins

Probability of Member Longer than 20 Mins

```
[ ]: member_riders = df[ df['member_casual'] == 'member' ]
```

```

long_member = member_riders[member_riders['time_duration'] >= pd.
    ↳Timedelta(minutes=20)]
probability = len(long_member)/ len(member_riders)
print(f"The probability that a ride duration is greater than 20 min_
    ↳conditioning on the fact that the user is a CitiBike member is {probability_
    ↳* 100}%")

```

The probability that a ride duration is greater than 20 min conditioning on the fact that the user is a CitiBike member is 15.564672725015486%

1.1.4 4. The probability That Longer Than 25 Mins Ride Belongs to a CitiBike Member

```

[ ]: longer_than_25_min = df[df['time_duration'] >= pd.Timedelta(minutes=25)]
longer_than_25_min_member =_
    ↳longer_than_25_min[longer_than_25_min['member_casual'] == 'member']
probability_2 = len(longer_than_25_min_member) / len(longer_than_25_min)
print(f"The probability that longer than 25 min ride belongs to a CitiBike_
    ↳member is {probability_2 * 100}%")

```

The probability that longer than 25 min ride belongs to a CitiBike member is 57.876337476304165%

1.1.5 5. Expected Duration of Bike

```

[ ]: electric_bike = df[df['rideable_type'] == 'electric_bike']
avg_electric_time = electric_bike['time_duration'].mean()
avg_electric_time_in_min = avg_electric_time.total_seconds() / 60
print(f"Expected ride duration of electric bike is {avg_electric_time_in_min}_
    ↳mins.")

```

Expected ride duration of electric bike is 12.275573333333334 mins.

```

[ ]: classic_bike = df[df['rideable_type'] == 'classic_bike']
avg_classic_time = classic_bike['time_duration'].mean()
avg_classic_time_in_min = avg_classic_time.total_seconds() / 60
print(f"Expected ride duration of classic bike is {avg_classic_time_in_min}_
    ↳mins.")

```

Expected ride duration of classic bike is 13.727032833333334 mins.

1.1.6 6. Suppose that the duration of some ride is less than 10min.

What is the probability that this ride uses an electric bike?

What is the probability that this ride uses a classic bike?

```

[ ]: less_10_min_ride = df[df['time_duration'] <= pd.Timedelta(minutes=10)]
less_10_min_electric_ride = less_10_min_ride[less_10_min_ride['rideable_type']_
    ↳== 'electric_bike']

```

```

less_10_min_classic_ride = less_10_min_ride[less_10_min_ride['rideable_type']_
↳== 'classic_bike']

probability_classic = len(less_10_min_classic_ride) / len(less_10_min_ride)
probability_electric = len(less_10_min_electric_ride) / len(less_10_min_ride)

print(f"The probability that this ride uses a classic bike less than 10min_
↳{probability_classic}.")
print(f"The probability that this ride uses a electric bike less than 10min_
↳{probability_electric}.")

```

The probability that this ride uses a classic bike less than 10min

0.9101874475812783.

The probability that this ride uses a electric bike less than 10min

0.08919753974733986.

The results indicate a strong preference for using classic bikes for shorter rides. For short distance riders, they would prefer to use classic bikes. This may be due to the availability of classic bikes and the fact that the charge for classic bikes is lower than for electric bikes. So, for those who travel short distances, they would not bother to find electric bikes. Electric bikes, on the other hand, are primarily used for long distance travel and take a longer time duration.

However, the expectation that the time duration of classic bike use is shorter than that of electric bikes may be because a small number of classic bikes are left unlocked and used for a long time.

1.2 Project Questions

1.2.1 1. Select the busiest Stations

```

[ ]: # Convert into Datetime
df['started_at'] = pd.to_datetime(df['started_at'])
df['ended_at'] = pd.to_datetime(df['ended_at'])

# Select Weekdays
df['day_of_week'] = df['started_at'].dt.dayofweek
df_weekdays = df[df['day_of_week'].isin(range(5))]

# Calculate the usage
start_station_counts = df_weekdays['start_station_name'].value_counts()
end_station_counts = df_weekdays['end_station_name'].value_counts()
total_station_activity = start_station_counts.add(end_station_counts,
↳fill_value=0)
top_stations = total_station_activity.sort_values(ascending=False).head(10)
top_start_counts = start_station_counts.reindex(top_stations.index,
↳fill_value=0)
top_end_counts = end_station_counts.reindex(top_stations.index, fill_value=0)

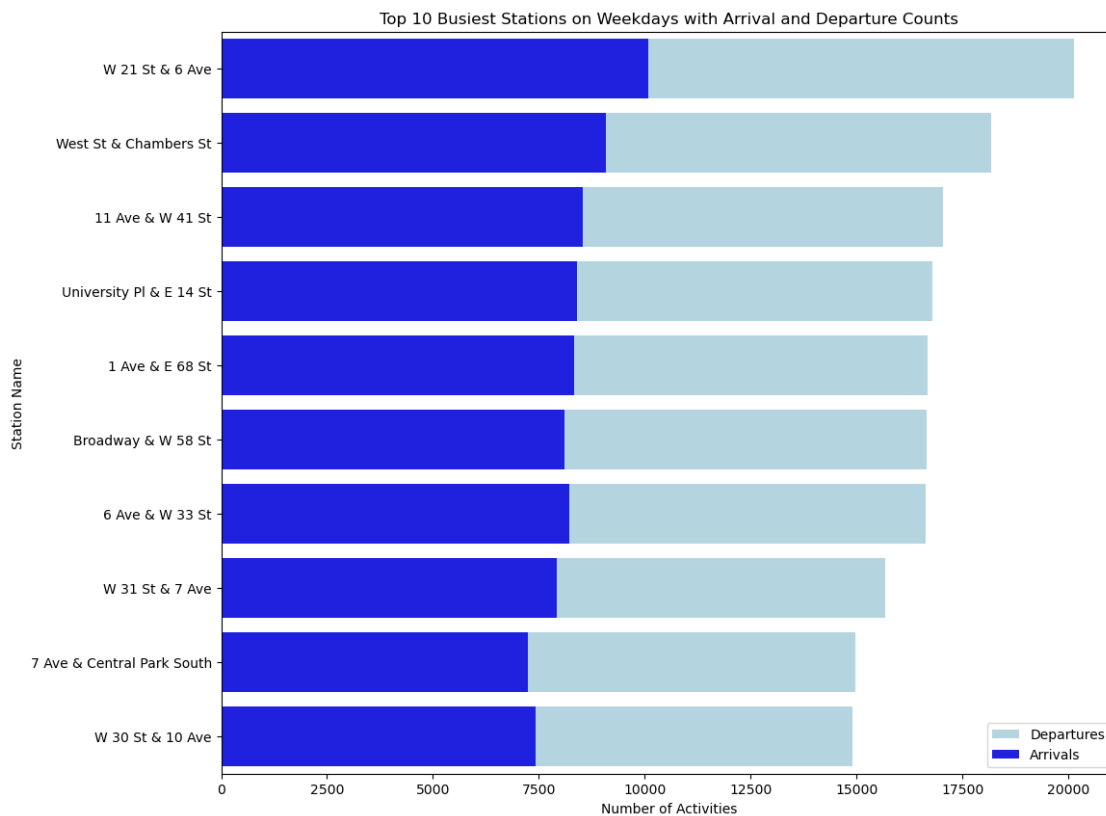
# Plot the figure
plt.figure(figsize=(12,10))

```

```

sns.barplot(x=top_stations.values, y=top_stations.index, color='lightblue',
            label='Departures')
sns.barplot(x=top_end_counts.values, y=top_end_counts.index, color='blue',
            label='Arrivals')
plt.xlabel('Number of Activities')
plt.ylabel('Station Name')
plt.title('Top 10 Busiest Stations on Weekdays with Arrival and Departure
            Counts')
plt.legend()
plt.show()

```



```

[ ]: top_three_stations = total_station_activity.nlargest(3).index
top_stations_data = df_weekdays[df_weekdays['start_station_name'].
    isin(top_three_stations) | df_weekdays['end_station_name'].
    isin(top_three_stations)]
csv_file_path = 'top_three_stations_data.csv'
top_stations_data.to_csv(csv_file_path, index=False)

```

```

[ ]: print(top_three_stations)

```

```

Index(['W 21 St & 6 Ave', 'West St & Chambers St', '11 Ave & W 41 St'],

```

```
dtype='object')
```

1.2.2 2. Calculate the change of bikes

```
[ ]: file_path = 'top_three_stations_data.csv'
df = pd.read_csv(file_path)
target_station = ['W 21 St & 6 Ave', 'West St & Chambers St', '11 Ave & W 41 St']

# Convert str to datetime
df['started_at'] = pd.to_datetime(df['started_at'])
df['ended_at'] = pd.to_datetime(df['ended_at'])

# Create a filter for time
morning_time = (df['started_at'].dt.hour >= 6) & (df['started_at'].dt.hour < 11)
afternoon_time = (df['started_at'].dt.hour >= 16) & (df['started_at'].dt.hour < 22)

# Filter Out required data
df = df[morning_time | afternoon_time]

df['start_time_interval'] = df['started_at'].dt.floor('10T')
df['end_time_interval'] = df['ended_at'].dt.floor('10T')
```

```
[ ]: # Split into morning and afternoon parts
morning_df = df[morning_time]
afternoon_df = df[afternoon_time]

station_w21St6ave_morning = morning_df[(morning_df['start_station_name'] == 'W 21 St & 6 Ave') | (morning_df['end_station_name'] == 'W 21 St & 6 Ave')]
station_w21St6ave_afternoon = afternoon_df[(afternoon_df['start_station_name'] == 'W 21 St & 6 Ave') | (afternoon_df['end_station_name'] == 'W 21 St & 6 Ave')]

# West St & Chambers St - Morning and Afternoon
station_westStChambersSt_morning = morning_df[(morning_df['start_station_name'] == 'West St & Chambers St') | (morning_df['end_station_name'] == 'West St & Chambers St')]
station_westStChambersSt_afternoon = afternoon_df[(afternoon_df['start_station_name'] == 'West St & Chambers St') | (afternoon_df['end_station_name'] == 'West St & Chambers St')]

# 11 Ave & W 41 St - Morning and Afternoon
station_11aveW41St_morning = morning_df[(morning_df['start_station_name'] == '11 Ave & W 41 St') | (morning_df['end_station_name'] == '11 Ave & W 41 St')]
station_11aveW41St_afternoon = afternoon_df[(afternoon_df['start_station_name'] == '11 Ave & W 41 St') | (afternoon_df['end_station_name'] == '11 Ave & W 41 St')]
```

```

/var/folders/ck/b09qsxz518x4rlt33mrt531c0000gn/T/ipykernel_1070/602006012.py:2:
UserWarning: Boolean Series key will be reindexed to match DataFrame index.
    morning_df = df[morning_time]
/var/folders/ck/b09qsxz518x4rlt33mrt531c0000gn/T/ipykernel_1070/602006012.py:3:
UserWarning: Boolean Series key will be reindexed to match DataFrame index.
    afternoon_df = df[afternoon_time]

```

```

[ ]: # West 21 St & 6 Ave - Morning
out_count_per_5_min_morning =
    ↳ station_w21St6ave_morning[station_w21St6ave_morning['start_station_name'] ==
    ↳ 'W 21 St & 6 Ave']['start_time_interval'].value_counts()
in_count_per_5_min_morning =
    ↳ station_w21St6ave_morning[station_w21St6ave_morning['end_station_name'] ==
    ↳ 'W 21 St & 6 Ave']['end_time_interval'].value_counts()

station_w21St6ave_count_morning = pd.concat([in_count_per_5_min_morning,
    ↳ out_count_per_5_min_morning], axis=1).rename(columns={'end_time_interval':
    ↳ 'in_count', 'start_time_interval': 'out_count'})
station_w21St6ave_count_morning.fillna(0, inplace=True)
station_w21St6ave_count_morning['final_count'] =
    ↳ station_w21St6ave_count_morning['in_count'] -
    ↳ station_w21St6ave_count_morning['out_count']
station_w21St6ave_count_morning['final_count'] =
    ↳ station_w21St6ave_count_morning['final_count'].astype(int)

# West 21 St & 6 Ave - Afternoon
out_count_per_5_min_afternoon =
    ↳ station_w21St6ave_afternoon[station_w21St6ave_afternoon['start_station_name']
    ↳ == 'W 21 St & 6 Ave']['start_time_interval'].value_counts()
in_count_per_5_min_afternoon =
    ↳ station_w21St6ave_afternoon[station_w21St6ave_afternoon['end_station_name']
    ↳ == 'W 21 St & 6 Ave']['end_time_interval'].value_counts()

station_w21St6ave_count_afternoon = pd.concat([in_count_per_5_min_afternoon,
    ↳ out_count_per_5_min_afternoon], axis=1).rename(columns={'end_time_interval':
    ↳ 'in_count', 'start_time_interval': 'out_count'})
station_w21St6ave_count_afternoon.fillna(0, inplace=True)
station_w21St6ave_count_afternoon['final_count'] =
    ↳ station_w21St6ave_count_afternoon['in_count'] -
    ↳ station_w21St6ave_count_afternoon['out_count']
station_w21St6ave_count_afternoon['final_count'] =
    ↳ station_w21St6ave_count_afternoon['final_count'].astype(int)

# West St & Chambers St - Morning

```



```

out_count_westStChambersSt_morning =
    ↳station_westStChambersSt_morning[station_westStChambersSt_morning['start_station_name']]
    ↳== 'West St & Chambers St')['start_time_interval'].value_counts()
in_count_westStChambersSt_morning =
    ↳station_westStChambersSt_morning[station_westStChambersSt_morning['end_station_name']]
    ↳== 'West St & Chambers St')['end_time_interval'].value_counts()

station_westStChambersSt_count_morning = pd.
    ↳concat([in_count_westStChambersSt_morning,
    ↳out_count_westStChambersSt_morning], axis=1).
    ↳rename(columns={'end_time_interval': 'in_count', 'start_time_interval':
    ↳'out_count'})
station_westStChambersSt_count_morning.fillna(0, inplace=True)
station_westStChambersSt_count_morning['final_count'] =
    ↳station_westStChambersSt_count_morning['in_count'] -
    ↳station_westStChambersSt_count_morning['out_count']
station_westStChambersSt_count_morning['final_count'] =
    ↳station_westStChambersSt_count_morning['final_count'].astype(int)

# West St & Chambers St - Afternoon
out_count_westStChambersSt_afternoon =
    ↳station_westStChambersSt_afternoon[station_westStChambersSt_afternoon['start_station_name']]
    ↳== 'West St & Chambers St')['start_time_interval'].value_counts()
in_count_westStChambersSt_afternoon =
    ↳station_westStChambersSt_afternoon[station_westStChambersSt_afternoon['end_station_name']]
    ↳== 'West St & Chambers St')['end_time_interval'].value_counts()

station_westStChambersSt_count_afternoon = pd.
    ↳concat([in_count_westStChambersSt_afternoon,
    ↳out_count_westStChambersSt_afternoon], axis=1).
    ↳rename(columns={'end_time_interval': 'in_count', 'start_time_interval':
    ↳'out_count'})
station_westStChambersSt_count_afternoon.fillna(0, inplace=True)
station_westStChambersSt_count_afternoon['final_count'] =
    ↳station_westStChambersSt_count_afternoon['in_count'] -
    ↳station_westStChambersSt_count_afternoon['out_count']
station_westStChambersSt_count_afternoon['final_count'] =
    ↳station_westStChambersSt_count_afternoon['final_count'].astype(int)

# 11 Ave & W 41 St - Morning
out_count_11aveW41St_morning =
    ↳station_11aveW41St_morning[station_11aveW41St_morning['start_station_name']]
    ↳== '11 Ave & W 41 St')['start_time_interval'].value_counts()
in_count_11aveW41St_morning =
    ↳station_11aveW41St_morning[station_11aveW41St_morning['end_station_name']] ==
    ↳'11 Ave & W 41 St')['end_time_interval'].value_counts()

```

```

station_11aveW41St_count_morning = pd.concat([in_count_11aveW41St_morning,
↳ out_count_11aveW41St_morning], axis=1).rename(columns={'end_time_interval':
↳ 'in_count', 'start_time_interval': 'out_count'})
station_11aveW41St_count_morning.fillna(0, inplace=True)
station_11aveW41St_count_morning['final_count'] =
↳ station_11aveW41St_count_morning['in_count'] -
↳ station_11aveW41St_count_morning['out_count']
station_11aveW41St_count_morning['final_count'] =
↳ station_11aveW41St_count_morning['final_count'].astype(int)

# 11 Ave & W 41 St - Afternoon
out_count_11aveW41St_afternoon =
↳ station_11aveW41St_afternoon[station_11aveW41St_afternoon['start_station_name']]
↳ == '11 Ave & W 41 St']['start_time_interval'].value_counts()
in_count_11aveW41St_afternoon =
↳ station_11aveW41St_afternoon[station_11aveW41St_afternoon['end_station_name']]
↳ == '11 Ave & W 41 St']['end_time_interval'].value_counts()

station_11aveW41St_count_afternoon = pd.concat([in_count_11aveW41St_afternoon,
↳ out_count_11aveW41St_afternoon], axis=1).rename(columns={'end_time_interval':
↳ 'in_count', 'start_time_interval': 'out_count'})
station_11aveW41St_count_afternoon.fillna(0, inplace=True)
station_11aveW41St_count_afternoon['final_count'] =
↳ station_11aveW41St_count_afternoon['in_count'] -
↳ station_11aveW41St_count_afternoon['out_count']
station_11aveW41St_count_afternoon['final_count'] =
↳ station_11aveW41St_count_afternoon['final_count'].astype(int)

```

```
[ ]: station_w21St6ave_count_morning
```

```

[ ]:
      in_count  out_count  final_count
2023-07-03 06:20:00      1.0        0.0          1
2023-07-03 06:30:00      0.0        1.0         -1
2023-07-03 06:40:00      2.0        1.0          1
2023-07-03 06:50:00      2.0        0.0          2
2023-07-03 07:30:00      2.0        1.0          1
...
2023-07-31 10:40:00      5.0        5.0          0
2023-07-31 10:50:00      4.0        3.0          1
2023-07-31 11:00:00      6.0        0.0          6
2023-07-31 11:10:00      1.0        0.0          1
2023-07-31 11:30:00      1.0        0.0          1

```

```
[644 rows x 3 columns]
```

```
[ ]: station_w21St6ave_dict_morning =
    dict(sorted(Counter(station_w21St6ave_count_morning['final_count']).items()))
station_w21St6ave_dict_afternoon =
    dict(sorted(Counter(station_w21St6ave_count_afternoon['final_count']).
    items()))

station_westStChambersSt_dict_morning =
    dict(sorted(Counter(station_westStChambersSt_count_morning['final_count']).
    items()))
station_westStChambersSt_dict_afternoon =
    dict(sorted(Counter(station_westStChambersSt_count_afternoon['final_count']).
    items()))

station_11aveW41St_dict_morning =
    dict(sorted(Counter(station_11aveW41St_count_morning['final_count']).
    items()))
station_11aveW41St_dict_afternoon =
    dict(sorted(Counter(station_11aveW41St_count_afternoon['final_count']).
    items()))
```

```
[ ]: station_w21St6ave_dict_morning
```

```
[ ]: {-11: 1,
      -7: 2,
      -6: 3,
      -5: 7,
      -4: 11,
      -3: 21,
      -2: 49,
      -1: 79,
      0: 83,
      1: 118,
      2: 94,
      3: 57,
      4: 43,
      5: 25,
      6: 27,
      7: 11,
      8: 4,
      9: 7,
      11: 2}
```

```
[ ]: def plot_bar(dic, location, time, ax=None):
    if ax is None:
        plt.bar(dic.keys(), dic.values())
        plt.title(f"Bike Counts for {location} - {time}")
        plt.xlabel("Number of Change")
```

```

plt.ylabel("Frequency")
plt.show()
else:
    ax.bar(dic.keys(), dic.values())
    ax.set_title(f"Bike Counts for {location} - {time}")
    ax.set_xlabel("Number of Change")
    ax.set_ylabel("Frequency")

# Create a 2x3 subplot grid
fig, axes = plt.subplots(2, 3, figsize=(15, 10))

# West 21 St & 6 Ave - Morning
plot_bar(station_w21St6ave_dict_morning, 'W 21 St & 6 Ave', 'morning',
    ↪ax=axes[0, 0])

# West 21 St & 6 Ave - Afternoon
plot_bar(station_w21St6ave_dict_afternoon, 'W 21 St & 6 Ave', 'afternoon',
    ↪ax=axes[1, 0])

# West St & Chambers St - Morning
plot_bar(station_westStChambersSt_dict_morning, 'West St & Chambers St',
    ↪'morning', ax=axes[0, 1])

# West St & Chambers St - Afternoon
plot_bar(station_westStChambersSt_dict_afternoon, 'West St & Chambers St',
    ↪'afternoon', ax=axes[1, 1])

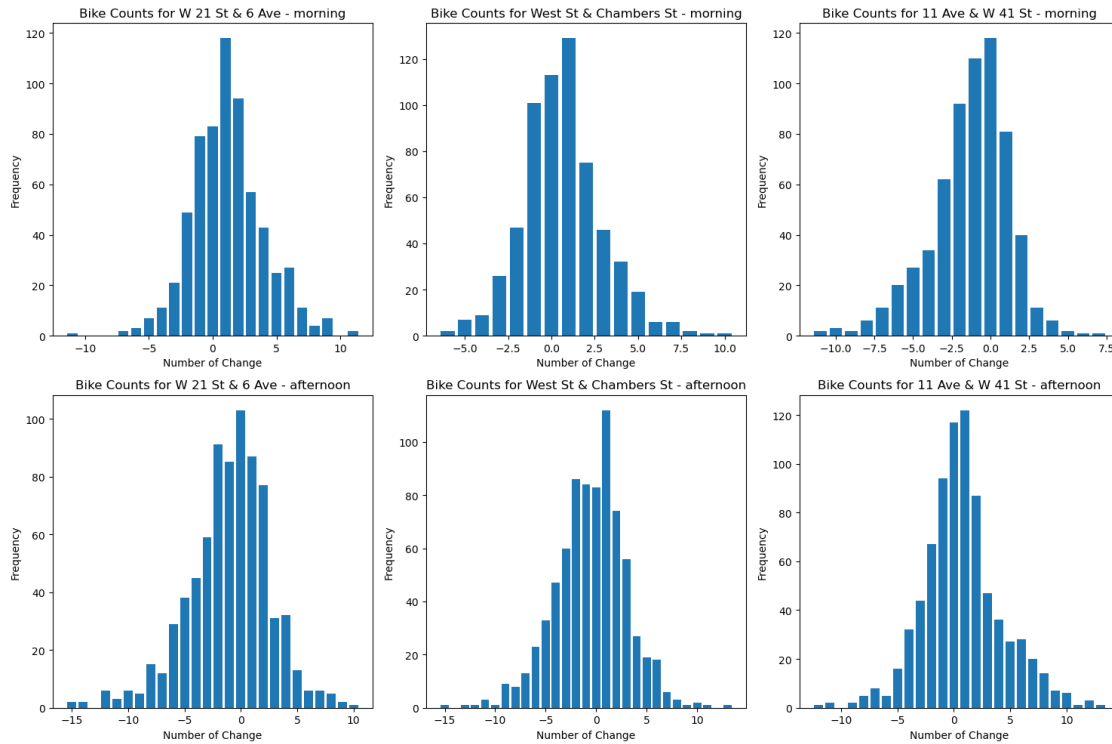
# 11 Ave & W 41 St - Morning
plot_bar(station_11aveW41St_dict_morning, '11 Ave & W 41 St', 'morning',
    ↪ax=axes[0, 2])

# 11 Ave & W 41 St - Afternoon
plot_bar(station_11aveW41St_dict_afternoon, '11 Ave & W 41 St', 'afternoon',
    ↪ax=axes[1, 2])

# Adjust spacing between subplots
plt.tight_layout()

# Show the plot
plt.show()

```



```
[ ]: def create_transition_matrix(docks, bike_change_dict):
    changes = list(bike_change_dict.keys())
    frequency = list(bike_change_dict.values())
    matrix = np.zeros((docks + 1, docks + 1))
    for i in range(matrix.shape[0]):
        for j in range(matrix.shape[1]):
            if (j - i) in changes:
                matrix[i][j] = bike_change_dict[j - i]
            matrix[i] = matrix[i]/sum(matrix[i])
    return matrix

def verify_row_sum_to_one(matrix):
    row_sums = np.sum(matrix, axis=1)
    return np.allclose(row_sums, 1)
```

```
[ ]: w21St6ave_dock = 52
westStChambersSt_dock = 101
station_11aveW41St_dock = 46

w21St6ave_morning = create_transition_matrix(w21St6ave_dock,
↪station_w21St6ave_dict_morning)
w21St6ave_afternoon = create_transition_matrix(w21St6ave_dock,
↪station_w21St6ave_dict_afternoon)
```

```

westStChambersSt_morning = create_transition_matrix(westStChambersSt_dock,
↳station_westStChambersSt_dict_morning)
westStChambersSt_afternoon = create_transition_matrix(westStChambersSt_dock,
↳station_westStChambersSt_dict_afternoon)

station_11aveW41St_morning = create_transition_matrix(station_11aveW41St_dock,
↳station_11aveW41St_dict_morning)
station_11aveW41St_afternoon =
↳create_transition_matrix(station_11aveW41St_dock,
↳station_11aveW41St_dict_afternoon)

```

```
[ ]: print(station_11aveW41St_afternoon)
```

```

[[0.22674419 0.23643411 0.16860465 ... 0.          0.          0.          ]
 [0.15409836 0.19180328 0.2          ... 0.          0.          0.          ]
 [0.09896603 0.13884786 0.17282127 ... 0.          0.          0.          ]
 ...
 [0.          0.          0.          ... 0.19435216 0.20265781 0.14451827]
 [0.          0.          0.          ... 0.18252427 0.22718447 0.2368932 ]
 [0.          0.          0.          ... 0.17048346 0.23918575 0.29770992]]

```

```

[ ]: import numpy as np
np.set_printoptions(threshold=np.inf)

# Define a list of transition matrices for different stations and times
transition_matrices = [w21St6ave_morning, w21St6ave_afternoon,
                        westStChambersSt_morning, westStChambersSt_afternoon,
                        station_11aveW41St_morning, station_11aveW41St_afternoon]

print(w21St6ave_morning)

```

```

[[0.17622081 0.25053079 0.19957537 0.12101911 0.09129512 0.05307856
 0.05732484 0.02335456 0.00849257 0.014862  0.          0.00424628
 0.          0.          0.          0.          0.          0.
 0.          0.          0.          0.          0.          0.
 0.          0.          0.          0.          0.          0.
 0.          0.          0.          0.          0.          0.
 0.          0.          0.          0.          0.          0.
 0.          0.          0.          0.          0.          0.
 0.          0.          0.          0.          0.          ]
 [0.14363636 0.15090909 0.21454545 0.17090909 0.10363636 0.07818182
 0.04545455 0.04909091 0.02          0.00727273 0.01272727 0.
 0.00363636 0.          0.          0.          0.          0.
 0.          0.          0.          0.          0.          0.
 0.          0.          0.          0.          0.          0.
 0.          0.          0.          0.          0.          0.
 0.          0.          0.          0.          0.          ]

```

0.	0.	0.	0.	0.	0.
0.	0.	0.	0.	0.]
[0.08180301	0.13188648	0.13856427	0.19699499	0.15692821	0.0951586
0.07178631	0.04173623	0.04507513	0.01836394	0.0066778	0.01168614
0.	0.0033389	0.	0.	0.	0.
0.	0.	0.	0.	0.	0.
0.	0.	0.	0.	0.	0.
0.	0.	0.	0.	0.	0.
0.	0.	0.	0.	0.	0.
0.	0.	0.	0.	0.	0.
0.	0.	0.	0.	0.]
[0.03387097	0.07903226	0.12741935	0.13387097	0.19032258	0.1516129
0.09193548	0.06935484	0.04032258	0.04354839	0.01774194	0.00645161
0.01129032	0.	0.00322581	0.	0.	0.
0.	0.	0.	0.	0.	0.
0.	0.	0.	0.	0.	0.
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0.	0.	0.	0.	0.	0.
0.	0.	0.	0.	0.	0.
0.	0.	0.	0.	0.]
[0.01743265	0.03328051	0.07765452	0.1251981	0.13153724	0.18700475
0.14896989	0.09033281	0.0681458	0.03961965	0.04278922	0.01743265
0.00633914	0.0110935	0.	0.00316957	0.	0.
0.	0.	0.	0.	0.	0.
0.	0.	0.	0.	0.	0.
0.	0.	0.	0.	0.	0.
0.	0.	0.	0.	0.	0.
0.	0.	0.	0.	0.	0.
0.	0.	0.	0.	0.]
[0.01097179	0.01724138	0.03291536	0.07680251	0.12382445	0.13009404
0.18495298	0.14733542	0.08934169	0.06739812	0.03918495	0.04231975
0.01724138	0.00626959	0.01097179	0.	0.0031348	0.
0.	0.	0.	0.	0.	0.
0.	0.	0.	0.	0.	0.
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0.	0.	0.	0.	0.	0.
0.	0.	0.	0.	0.	0.
0.	0.	0.	0.	0.]
[0.00468019	0.01092044	0.01716069	0.03276131	0.07644306	0.12324493
0.12948518	0.18408736	0.14664587	0.08892356	0.06708268	0.03900156
0.04212168	0.01716069	0.00624025	0.01092044	0.	0.00312012
0.	0.	0.	0.	0.	0.
0.	0.	0.	0.	0.	0.
0.	0.	0.	0.	0.	0.
0.	0.	0.	0.	0.	0.
0.	0.	0.	0.	0.	0.
0.	0.	0.	0.	0.]
[0.00311042	0.00466563	0.01088647	0.01710731	0.03265941	0.07620529

0.12286159	0.12908243	0.18351477	0.14618974	0.08864697	0.06687403
0.03888025	0.04199067	0.01710731	0.00622084	0.01088647	0.
0.00311042	0.	0.	0.	0.	0.
0.	0.	0.	0.	0.	0.
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0.	0.	0.	0.	0.]
[0.	0.00311042	0.00466563	0.01088647	0.01710731	0.03265941
0.07620529	0.12286159	0.12908243	0.18351477	0.14618974	0.08864697
0.06687403	0.03888025	0.04199067	0.01710731	0.00622084	0.01088647
0.	0.00311042	0.	0.	0.	0.
0.	0.	0.	0.	0.	0.
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0.	0.	0.	0.	0.	0.
0.	0.	0.	0.	0.]
[0.	0.	0.00311042	0.00466563	0.01088647	0.01710731
0.03265941	0.07620529	0.12286159	0.12908243	0.18351477	0.14618974
0.08864697	0.06687403	0.03888025	0.04199067	0.01710731	0.00622084
0.01088647	0.	0.00311042	0.	0.	0.
0.	0.	0.	0.	0.	0.
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0.	0.	0.	0.	0.]
[0.	0.	0.	0.00311042	0.00466563	0.01088647
0.01710731	0.03265941	0.07620529	0.12286159	0.12908243	0.18351477
0.14618974	0.08864697	0.06687403	0.03888025	0.04199067	0.01710731
0.00622084	0.01088647	0.	0.00311042	0.	0.
0.	0.	0.	0.	0.	0.
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0.	0.	0.	0.	0.	0.
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0.	0.	0.	0.	0.]
[0.0015528	0.	0.	0.	0.00310559	0.00465839
0.01086957	0.01708075	0.0326087	0.07608696	0.12267081	0.12888199
0.18322981	0.14596273	0.08850932	0.06677019	0.03881988	0.04192547
0.01708075	0.00621118	0.01086957	0.	0.00310559	0.
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0.	0.	0.	0.	0.	0.
0.	0.	0.	0.	0.]
[0.	0.0015528	0.	0.	0.	0.00310559
0.00465839	0.01086957	0.01708075	0.0326087	0.07608696	0.12267081
0.12888199	0.18322981	0.14596273	0.08850932	0.06677019	0.03881988
0.04192547	0.01708075	0.00621118	0.01086957	0.	0.00310559

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0.	0.	0.	0.	0.]
[0.	0.	0.0015528	0.	0.	0.
0.00310559	0.00465839	0.01086957	0.01708075	0.0326087	0.07608696
0.12267081	0.12888199	0.18322981	0.14596273	0.08850932	0.06677019
0.03881988	0.04192547	0.01708075	0.00621118	0.01086957	0.
0.00310559	0.	0.	0.	0.	0.
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[0.	0.	0.	0.0015528	0.	0.
0.	0.00310559	0.00465839	0.01086957	0.01708075	0.0326087
0.07608696	0.12267081	0.12888199	0.18322981	0.14596273	0.08850932
0.06677019	0.03881988	0.04192547	0.01708075	0.00621118	0.01086957
0.	0.00310559	0.	0.	0.	0.
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[0.	0.	0.	0.	0.0015528	0.
0.	0.	0.00310559	0.00465839	0.01086957	0.01708075
0.0326087	0.07608696	0.12267081	0.12888199	0.18322981	0.14596273
0.08850932	0.06677019	0.03881988	0.04192547	0.01708075	0.00621118
0.01086957	0.	0.00310559	0.	0.	0.
0.	0.	0.	0.	0.	0.
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0.	0.	0.	0.	0.]
[0.	0.	0.	0.	0.	0.0015528
0.	0.	0.	0.00310559	0.00465839	0.01086957
0.01708075	0.0326087	0.07608696	0.12267081	0.12888199	0.18322981
0.14596273	0.08850932	0.06677019	0.03881988	0.04192547	0.01708075
0.00621118	0.01086957	0.	0.00310559	0.	0.
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[0.	0.	0.	0.	0.	0.
0.0015528	0.	0.	0.	0.00310559	0.00465839
0.01086957	0.01708075	0.0326087	0.07608696	0.12267081	0.12888199
0.18322981	0.14596273	0.08850932	0.06677019	0.03881988	0.04192547
0.01708075	0.00621118	0.01086957	0.	0.00310559	0.
0.	0.	0.	0.	0.	0.
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[0.	0.	0.	0.	0.	0.
0.	0.0015528	0.	0.	0.	0.00310559
0.00465839	0.01086957	0.01708075	0.0326087	0.07608696	0.12267081
0.12888199	0.18322981	0.14596273	0.08850932	0.06677019	0.03881988
0.04192547	0.01708075	0.00621118	0.01086957	0.	0.00310559
0.	0.	0.	0.	0.	0.
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0.	0.	0.	0.	0.]
[0.	0.	0.	0.	0.	0.
0.	0.	0.0015528	0.	0.	0.
0.00310559	0.00465839	0.01086957	0.01708075	0.0326087	0.07608696
0.12267081	0.12888199	0.18322981	0.14596273	0.08850932	0.06677019
0.03881988	0.04192547	0.01708075	0.00621118	0.01086957	0.
0.00310559	0.	0.	0.	0.	0.
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[0.	0.	0.	0.	0.	0.
0.	0.	0.	0.0015528	0.	0.
0.	0.00310559	0.00465839	0.01086957	0.01708075	0.0326087
0.07608696	0.12267081	0.12888199	0.18322981	0.14596273	0.08850932
0.06677019	0.03881988	0.04192547	0.01708075	0.00621118	0.01086957
0.	0.00310559	0.	0.	0.	0.
0.	0.	0.	0.	0.	0.
0.	0.	0.	0.	0.	0.
0.	0.	0.	0.	0.]
[0.	0.	0.	0.	0.	0.
0.	0.	0.	0.	0.0015528	0.
0.	0.	0.00310559	0.00465839	0.01086957	0.01708075
0.0326087	0.07608696	0.12267081	0.12888199	0.18322981	0.14596273
0.08850932	0.06677019	0.03881988	0.04192547	0.01708075	0.00621118
0.01086957	0.	0.00310559	0.	0.	0.
0.	0.	0.	0.	0.	0.
0.	0.	0.	0.	0.	0.
0.	0.	0.	0.	0.]
[0.	0.	0.	0.	0.	0.
0.	0.	0.	0.	0.	0.0015528
0.	0.	0.	0.00310559	0.00465839	0.01086957
0.01708075	0.0326087	0.07608696	0.12267081	0.12888199	0.18322981
0.14596273	0.08850932	0.06677019	0.03881988	0.04192547	0.01708075
0.00621118	0.01086957	0.	0.00310559	0.	0.
0.	0.	0.	0.	0.	0.
0.	0.	0.	0.	0.	0.
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[0.	0.	0.	0.	0.	0.

0.	0.	0.	0.	0.	0.
0.0015528	0.	0.	0.	0.00310559	0.00465839
0.01086957	0.01708075	0.0326087	0.07608696	0.12267081	0.12888199
0.18322981	0.14596273	0.08850932	0.06677019	0.03881988	0.04192547
0.01708075	0.00621118	0.01086957	0.	0.00310559	0.
0.	0.	0.	0.	0.	0.
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[0.	0.	0.	0.	0.	0.
0.	0.	0.	0.	0.	0.
0.	0.0015528	0.	0.	0.	0.00310559
0.00465839	0.01086957	0.01708075	0.0326087	0.07608696	0.12267081
0.12888199	0.18322981	0.14596273	0.08850932	0.06677019	0.03881988
0.04192547	0.01708075	0.00621118	0.01086957	0.	0.00310559
0.	0.	0.	0.	0.	0.
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[0.	0.	0.	0.	0.	0.
0.	0.	0.	0.	0.	0.
0.	0.	0.0015528	0.	0.	0.
0.00310559	0.00465839	0.01086957	0.01708075	0.0326087	0.07608696
0.12267081	0.12888199	0.18322981	0.14596273	0.08850932	0.06677019
0.03881988	0.04192547	0.01708075	0.00621118	0.01086957	0.
0.00310559	0.	0.	0.	0.	0.
0.	0.	0.	0.	0.	0.
0.	0.	0.	0.	0.]
[0.	0.	0.	0.	0.	0.
0.	0.	0.	0.	0.	0.
0.	0.	0.	0.0015528	0.	0.
0.	0.00310559	0.00465839	0.01086957	0.01708075	0.0326087
0.07608696	0.12267081	0.12888199	0.18322981	0.14596273	0.08850932
0.06677019	0.03881988	0.04192547	0.01708075	0.00621118	0.01086957
0.	0.00310559	0.	0.	0.	0.
0.	0.	0.	0.	0.	0.
0.	0.	0.	0.	0.]
[0.	0.	0.	0.	0.	0.
0.	0.	0.	0.	0.	0.
0.	0.	0.	0.	0.0015528	0.
0.	0.	0.00310559	0.00465839	0.01086957	0.01708075
0.0326087	0.07608696	0.12267081	0.12888199	0.18322981	0.14596273
0.08850932	0.06677019	0.03881988	0.04192547	0.01708075	0.00621118
0.01086957	0.	0.00310559	0.	0.	0.
0.	0.	0.	0.	0.	0.
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0.	0.	0.	0.	0.	0.
0.	0.	0.	0.	0.	0.0015528
0.	0.	0.	0.00310559	0.00465839	0.01086957

0.01708075	0.0326087	0.07608696	0.12267081	0.12888199	0.18322981
0.14596273	0.08850932	0.06677019	0.03881988	0.04192547	0.01708075
0.00621118	0.01086957	0.	0.00310559	0.	0.
0.	0.	0.	0.	0.	0.
0.	0.	0.	0.	0.]
[0.	0.	0.	0.	0.	0.
0.	0.	0.	0.	0.	0.
0.	0.	0.	0.	0.	0.
0.0015528	0.	0.	0.	0.00310559	0.00465839
0.01086957	0.01708075	0.0326087	0.07608696	0.12267081	0.12888199
0.18322981	0.14596273	0.08850932	0.06677019	0.03881988	0.04192547
0.01708075	0.00621118	0.01086957	0.	0.00310559	0.
0.	0.	0.	0.	0.	0.
0.	0.	0.	0.	0.]
[0.	0.	0.	0.	0.	0.
0.	0.	0.	0.	0.	0.
0.	0.	0.	0.	0.	0.
0.	0.0015528	0.	0.	0.	0.00310559
0.00465839	0.01086957	0.01708075	0.0326087	0.07608696	0.12267081
0.12888199	0.18322981	0.14596273	0.08850932	0.06677019	0.03881988
0.04192547	0.01708075	0.00621118	0.01086957	0.	0.00310559
0.	0.	0.	0.	0.	0.
0.	0.	0.	0.	0.]
[0.	0.	0.	0.	0.	0.
0.	0.	0.	0.	0.	0.
0.	0.	0.	0.	0.	0.
0.	0.	0.0015528	0.	0.	0.
0.00310559	0.00465839	0.01086957	0.01708075	0.0326087	0.07608696
0.12267081	0.12888199	0.18322981	0.14596273	0.08850932	0.06677019
0.03881988	0.04192547	0.01708075	0.00621118	0.01086957	0.
0.00310559	0.	0.	0.	0.	0.
0.	0.	0.	0.	0.]
[0.	0.	0.	0.	0.	0.
0.	0.	0.	0.	0.	0.
0.	0.	0.	0.	0.	0.
0.	0.	0.	0.0015528	0.	0.
0.	0.00310559	0.00465839	0.01086957	0.01708075	0.0326087
0.07608696	0.12267081	0.12888199	0.18322981	0.14596273	0.08850932
0.06677019	0.03881988	0.04192547	0.01708075	0.00621118	0.01086957
0.	0.00310559	0.	0.	0.	0.
0.	0.	0.	0.	0.]
[0.	0.	0.	0.	0.	0.
0.	0.	0.	0.	0.	0.
0.	0.	0.	0.	0.	0.
0.	0.	0.	0.	0.0015528	0.
0.	0.	0.00310559	0.00465839	0.01086957	0.01708075
0.0326087	0.07608696	0.12267081	0.12888199	0.18322981	0.14596273
0.08850932	0.06677019	0.03881988	0.04192547	0.01708075	0.00621118

0.01086957	0.	0.00310559	0.	0.	0.
0.	0.	0.	0.	0.]
[0.	0.	0.	0.	0.	0.
0.	0.	0.	0.	0.	0.
0.	0.	0.	0.	0.	0.
0.	0.	0.	0.	0.	0.0015528
0.	0.	0.	0.00310559	0.00465839	0.01086957
0.01708075	0.0326087	0.07608696	0.12267081	0.12888199	0.18322981
0.14596273	0.08850932	0.06677019	0.03881988	0.04192547	0.01708075
0.00621118	0.01086957	0.	0.00310559	0.	0.
0.	0.	0.	0.	0.]
[0.	0.	0.	0.	0.	0.
0.	0.	0.	0.	0.	0.
0.	0.	0.	0.	0.	0.
0.	0.	0.	0.	0.	0.
0.0015528	0.	0.	0.	0.00310559	0.00465839
0.01086957	0.01708075	0.0326087	0.07608696	0.12267081	0.12888199
0.18322981	0.14596273	0.08850932	0.06677019	0.03881988	0.04192547
0.01708075	0.00621118	0.01086957	0.	0.00310559	0.
0.	0.	0.	0.	0.]
[0.	0.	0.	0.	0.	0.
0.	0.	0.	0.	0.	0.
0.	0.	0.	0.	0.	0.
0.	0.	0.	0.	0.	0.
0.	0.0015528	0.	0.	0.	0.00310559
0.00465839	0.01086957	0.01708075	0.0326087	0.07608696	0.12267081
0.12888199	0.18322981	0.14596273	0.08850932	0.06677019	0.03881988
0.04192547	0.01708075	0.00621118	0.01086957	0.	0.00310559
0.	0.	0.	0.	0.]
[0.	0.	0.	0.	0.	0.
0.	0.	0.	0.	0.	0.
0.	0.	0.	0.	0.	0.
0.	0.	0.	0.	0.	0.
0.	0.	0.0015528	0.	0.	0.
0.00310559	0.00465839	0.01086957	0.01708075	0.0326087	0.07608696
0.12267081	0.12888199	0.18322981	0.14596273	0.08850932	0.06677019
0.03881988	0.04192547	0.01708075	0.00621118	0.01086957	0.
0.00310559	0.	0.	0.	0.]
[0.	0.	0.	0.	0.	0.
0.	0.	0.	0.	0.	0.
0.	0.	0.	0.	0.	0.
0.	0.	0.	0.	0.	0.
0.	0.	0.	0.0015528	0.	0.
0.	0.00310559	0.00465839	0.01086957	0.01708075	0.0326087
0.07608696	0.12267081	0.12888199	0.18322981	0.14596273	0.08850932
0.06677019	0.03881988	0.04192547	0.01708075	0.00621118	0.01086957
0.	0.00310559	0.	0.	0.]
[0.	0.	0.	0.	0.	0.

0.	0.	0.	0.	0.	0.
0.	0.	0.	0.	0.	0.
0.	0.	0.	0.	0.	0.
0.	0.	0.	0.	0.0015528	0.
0.	0.	0.00310559	0.00465839	0.01086957	0.01708075
0.0326087	0.07608696	0.12267081	0.12888199	0.18322981	0.14596273
0.08850932	0.06677019	0.03881988	0.04192547	0.01708075	0.00621118
0.01086957	0.	0.00310559	0.	0.]
[0.	0.	0.	0.	0.	0.
0.	0.	0.	0.	0.	0.
0.	0.	0.	0.	0.	0.
0.	0.	0.	0.	0.	0.
0.	0.	0.	0.	0.	0.0015528
0.	0.	0.	0.00310559	0.00465839	0.01086957
0.01708075	0.0326087	0.07608696	0.12267081	0.12888199	0.18322981
0.14596273	0.08850932	0.06677019	0.03881988	0.04192547	0.01708075
0.00621118	0.01086957	0.	0.00310559	0.]
[0.	0.	0.	0.	0.	0.
0.	0.	0.	0.	0.	0.
0.	0.	0.	0.	0.	0.
0.	0.	0.	0.	0.	0.
0.	0.	0.	0.	0.	0.
0.0015528	0.	0.	0.	0.00310559	0.00465839
0.01086957	0.01708075	0.0326087	0.07608696	0.12267081	0.12888199
0.18322981	0.14596273	0.08850932	0.06677019	0.03881988	0.04192547
0.01708075	0.00621118	0.01086957	0.	0.00310559]	
[0.	0.	0.	0.	0.	0.
0.	0.	0.	0.	0.	0.
0.	0.	0.	0.	0.	0.
0.	0.	0.	0.	0.	0.
0.	0.	0.	0.	0.	0.
0.	0.00155763	0.	0.	0.	0.00311526
0.0046729	0.01090343	0.01713396	0.03271028	0.07632399	0.12305296
0.12928349	0.18380062	0.14641745	0.08878505	0.06697819	0.03894081
0.04205607	0.01713396	0.00623053	0.01090343	0.]
[0.	0.	0.	0.	0.	0.
0.	0.	0.	0.	0.	0.
0.	0.	0.	0.	0.	0.
0.	0.	0.	0.	0.	0.
0.	0.	0.	0.	0.	0.
0.	0.	0.00155763	0.	0.	0.
0.00311526	0.0046729	0.01090343	0.01713396	0.03271028	0.07632399
0.12305296	0.12928349	0.18380062	0.14641745	0.08878505	0.06697819
0.03894081	0.04205607	0.01713396	0.00623053	0.01090343]	
[0.	0.	0.	0.	0.	0.
0.	0.	0.	0.	0.	0.
0.	0.	0.	0.	0.	0.
0.	0.	0.	0.	0.	0.

0.	0.	0.	0.	0.	0.
0.	0.	0.	0.0015748	0.	0.
0.	0.00314961	0.00472441	0.01102362	0.01732283	0.03307087
0.07716535	0.12440945	0.13070866	0.18582677	0.1480315	0.08976378
0.06771654	0.03937008	0.04251969	0.01732283	0.00629921]	
[0.	0.	0.	0.	0.	0.
0.	0.	0.	0.	0.	0.
0.	0.	0.	0.	0.	0.
0.	0.	0.	0.	0.	0.
0.	0.	0.	0.	0.	0.
0.	0.	0.	0.	0.00158479	0.
0.	0.	0.00316957	0.00475436	0.0110935	0.01743265
0.03328051	0.07765452	0.1251981	0.13153724	0.18700475	0.14896989
0.09033281	0.0681458	0.03961965	0.04278922	0.01743265]	
[0.	0.	0.	0.	0.	0.
0.	0.	0.	0.	0.	0.
0.	0.	0.	0.	0.	0.
0.	0.	0.	0.	0.	0.
0.	0.	0.	0.	0.	0.
0.	0.	0.	0.	0.	0.0016129
0.	0.	0.	0.00322581	0.00483871	0.01129032
0.01774194	0.03387097	0.07903226	0.12741935	0.13387097	0.19032258
0.1516129	0.09193548	0.06935484	0.04032258	0.04354839]	
[0.	0.	0.	0.	0.	0.
0.	0.	0.	0.	0.	0.
0.	0.	0.	0.	0.	0.
0.	0.	0.	0.	0.	0.
0.	0.	0.	0.	0.	0.
0.	0.	0.	0.	0.	0.
0.00168634	0.	0.	0.	0.00337268	0.00505902
0.01180438	0.01854975	0.03541315	0.08263069	0.13322091	0.13996627
0.1989882	0.15851602	0.09612142	0.07251265	0.04215852]	
[0.	0.	0.	0.	0.	0.
0.	0.	0.	0.	0.	0.
0.	0.	0.	0.	0.	0.
0.	0.	0.	0.	0.	0.
0.	0.	0.	0.	0.	0.
0.	0.	0.	0.	0.	0.
0.	0.00176056	0.	0.	0.	0.00352113
0.00528169	0.01232394	0.0193662	0.03697183	0.08626761	0.13908451
0.14612676	0.20774648	0.16549296	0.10035211	0.07570423]	
[0.	0.	0.	0.	0.	0.
0.	0.	0.	0.	0.	0.
0.	0.	0.	0.	0.	0.
0.	0.	0.	0.	0.	0.
0.	0.	0.	0.	0.	0.
0.	0.	0.	0.	0.	0.
0.	0.	0.00190476	0.	0.	0.

```

0.00380952 0.00571429 0.01333333 0.02095238 0.04      0.09333333
0.15047619 0.15809524 0.2247619  0.17904762 0.10857143]
[0.      0.      0.      0.      0.      0.
0.      0.      0.      0.      0.      0.
0.      0.      0.      0.      0.      0.
0.      0.      0.      0.      0.      0.
0.      0.      0.      0.      0.      0.
0.      0.      0.      0.      0.      0.
0.      0.      0.      0.00213675 0.      0.
0.      0.0042735 0.00641026 0.01495726 0.02350427 0.04487179
0.10470085 0.16880342 0.17735043 0.25213675 0.2008547 ]
[0.      0.      0.      0.      0.      0.
0.      0.      0.      0.      0.      0.
0.      0.      0.      0.      0.      0.
0.      0.      0.      0.      0.      0.
0.      0.      0.      0.      0.      0.
0.      0.      0.      0.      0.      0.
0.      0.      0.      0.      0.0026738 0.
0.      0.      0.00534759 0.00802139 0.01871658 0.02941176
0.05614973 0.13101604 0.21122995 0.22192513 0.31550802]
[0.      0.      0.      0.      0.      0.
0.      0.      0.      0.      0.      0.
0.      0.      0.      0.      0.      0.
0.      0.      0.      0.      0.      0.
0.      0.      0.      0.      0.      0.
0.      0.      0.      0.      0.      0.
0.      0.      0.      0.      0.      0.00390625
0.      0.      0.      0.0078125 0.01171875 0.02734375
0.04296875 0.08203125 0.19140625 0.30859375 0.32421875]]

```

```

[ ]: # Verify if the row sums equal to one for each matrix
print(verify_row_sum_to_one(w21St6ave_morning))
print(verify_row_sum_to_one(w21St6ave_afternoon))

print(verify_row_sum_to_one(westStChambersSt_morning))
print(verify_row_sum_to_one(westStChambersSt_afternoon))

print(verify_row_sum_to_one(station_11aveW41St_morning))
print(verify_row_sum_to_one(station_11aveW41St_afternoon))

```

```

True
True
True
True
True
True

```



```
[ ]: def calculate_stationary_distribution(transition_matrix):
    # Transpose the transition matrix to find left eigenvectors
    transposed_matrix = np.transpose(transition_matrix)

    # Calculate the eigenvalues and eigenvectors
    eigenvalues, eigenvectors = np.linalg.eig(transposed_matrix)

    # Find the index corresponding to the eigenvalue 1
    index = np.where(np.isclose(eigenvalues, 1))[0][0]

    # Extract the corresponding eigenvector
    stationary_distribution = np.real(eigenvectors[:, index])

    # Normalize the stationary distribution to sum to 1
    stationary_distribution /= np.sum(stationary_distribution)

    return stationary_distribution

w21St6ave_morning_distribution = □
    ↳ calculate_stationary_distribution(w21St6ave_morning)
w21St6ave_afternoon_distribution = □
    ↳ calculate_stationary_distribution(w21St6ave_afternoon)
westStChambersSt_morning_distribution = □
    ↳ calculate_stationary_distribution(westStChambersSt_morning)
westStChambersSt_afternoon_distribution = □
    ↳ calculate_stationary_distribution(westStChambersSt_afternoon)
station_11aveW41St_morning_distribution = □
    ↳ calculate_stationary_distribution(station_11aveW41St_morning)
station_11aveW41St_afternoon_distribution = □
    ↳ calculate_stationary_distribution(station_11aveW41St_afternoon)

[ ]: print(w21St6ave_morning)
```

```
[[0.17622081 0.25053079 0.19957537 0.12101911 0.09129512 0.05307856
 0.05732484 0.02335456 0.00849257 0.014862 0. 0.00424628
 0. 0. 0. 0. 0. 0.
 0. 0. 0. 0. 0. 0.
 0. 0. 0. 0. 0. 0.
 0. 0. 0. 0. 0. 0.
 0. 0. 0. 0. 0. 0.
 0. 0. 0. 0. 0. 0.
 0. 0. 0. 0. 0. ]
[0.14363636 0.15090909 0.21454545 0.17090909 0.10363636 0.07818182
 0.04545455 0.04909091 0.02 0.00727273 0.01272727 0.
 0.00363636 0. 0. 0. 0. 0.
 0. 0. 0. 0. 0. 0.
 0. 0. 0. 0. 0. 0.]
```

0.	0.	0.	0.	0.	0.
0.	0.	0.	0.	0.	0.
0.	0.	0.	0.	0.	0.
0.	0.	0.	0.	0.]
[0.08180301	0.13188648	0.13856427	0.19699499	0.15692821	0.0951586
0.07178631	0.04173623	0.04507513	0.01836394	0.0066778	0.01168614
0.	0.0033389	0.	0.	0.	0.
0.	0.	0.	0.	0.	0.
0.	0.	0.	0.	0.	0.
0.	0.	0.	0.	0.	0.
0.	0.	0.	0.	0.	0.
0.	0.	0.	0.	0.	0.
0.	0.	0.	0.	0.]
[0.03387097	0.07903226	0.12741935	0.13387097	0.19032258	0.1516129
0.09193548	0.06935484	0.04032258	0.04354839	0.01774194	0.00645161
0.01129032	0.	0.00322581	0.	0.	0.
0.	0.	0.	0.	0.	0.
0.	0.	0.	0.	0.	0.
0.	0.	0.	0.	0.	0.
0.	0.	0.	0.	0.	0.
0.	0.	0.	0.	0.	0.
0.	0.	0.	0.	0.]
[0.01743265	0.03328051	0.07765452	0.1251981	0.13153724	0.18700475
0.14896989	0.09033281	0.0681458	0.03961965	0.04278922	0.01743265
0.00633914	0.0110935	0.	0.00316957	0.	0.
0.	0.	0.	0.	0.	0.
0.	0.	0.	0.	0.	0.
0.	0.	0.	0.	0.	0.
0.	0.	0.	0.	0.	0.
0.	0.	0.	0.	0.	0.
0.	0.	0.	0.	0.]
[0.01097179	0.01724138	0.03291536	0.07680251	0.12382445	0.13009404
0.18495298	0.14733542	0.08934169	0.06739812	0.03918495	0.04231975
0.01724138	0.00626959	0.01097179	0.	0.0031348	0.
0.	0.	0.	0.	0.	0.
0.	0.	0.	0.	0.	0.
0.	0.	0.	0.	0.	0.
0.	0.	0.	0.	0.	0.
0.	0.	0.	0.	0.	0.
0.	0.	0.	0.	0.]
[0.00468019	0.01092044	0.01716069	0.03276131	0.07644306	0.12324493
0.12948518	0.18408736	0.14664587	0.08892356	0.06708268	0.03900156
0.04212168	0.01716069	0.00624025	0.01092044	0.	0.00312012
0.	0.	0.	0.	0.	0.
0.	0.	0.	0.	0.	0.
0.	0.	0.	0.	0.	0.
0.	0.	0.	0.	0.	0.
0.	0.	0.	0.	0.	0.

0.	0.	0.	0.	0.]
[0.00311042	0.00466563	0.01088647	0.01710731	0.03265941	0.07620529
0.12286159	0.12908243	0.18351477	0.14618974	0.08864697	0.06687403
0.03888025	0.04199067	0.01710731	0.00622084	0.01088647	0.
0.00311042	0.	0.	0.	0.	0.
0.	0.	0.	0.	0.	0.
0.	0.	0.	0.	0.	0.
0.	0.	0.	0.	0.	0.
0.	0.	0.	0.	0.	0.
0.	0.	0.	0.	0.]
[0.	0.00311042	0.00466563	0.01088647	0.01710731	0.03265941
0.07620529	0.12286159	0.12908243	0.18351477	0.14618974	0.08864697
0.06687403	0.03888025	0.04199067	0.01710731	0.00622084	0.01088647
0.	0.00311042	0.	0.	0.	0.
0.	0.	0.	0.	0.	0.
0.	0.	0.	0.	0.	0.
0.	0.	0.	0.	0.	0.
0.	0.	0.	0.	0.	0.
0.	0.	0.	0.	0.]
[0.	0.	0.00311042	0.00466563	0.01088647	0.01710731
0.03265941	0.07620529	0.12286159	0.12908243	0.18351477	0.14618974
0.08864697	0.06687403	0.03888025	0.04199067	0.01710731	0.00622084
0.01088647	0.	0.00311042	0.	0.	0.
0.	0.	0.	0.	0.	0.
0.	0.	0.	0.	0.	0.
0.	0.	0.	0.	0.	0.
0.	0.	0.	0.	0.	0.
0.	0.	0.	0.	0.]
[0.	0.	0.	0.00311042	0.00466563	0.01088647
0.01710731	0.03265941	0.07620529	0.12286159	0.12908243	0.18351477
0.14618974	0.08864697	0.06687403	0.03888025	0.04199067	0.01710731
0.00622084	0.01088647	0.	0.00311042	0.	0.
0.	0.	0.	0.	0.	0.
0.	0.	0.	0.	0.	0.
0.	0.	0.	0.	0.	0.
0.	0.	0.	0.	0.	0.
0.	0.	0.	0.	0.]
[0.0015528	0.	0.	0.	0.00310559	0.00465839
0.01086957	0.01708075	0.0326087	0.07608696	0.12267081	0.12888199
0.18322981	0.14596273	0.08850932	0.06677019	0.03881988	0.04192547
0.01708075	0.00621118	0.01086957	0.	0.00310559	0.
0.	0.	0.	0.	0.	0.
0.	0.	0.	0.	0.	0.
0.	0.	0.	0.	0.	0.
0.	0.	0.	0.	0.	0.
0.	0.	0.	0.	0.]
[0.	0.0015528	0.	0.	0.	0.00310559
0.00465839	0.01086957	0.01708075	0.0326087	0.07608696	0.12267081

0.12888199	0.18322981	0.14596273	0.08850932	0.06677019	0.03881988
0.04192547	0.01708075	0.00621118	0.01086957	0.	0.00310559
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0.03881988	0.04192547	0.01708075	0.00621118	0.01086957	0.
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0.06677019	0.03881988	0.04192547	0.01708075	0.00621118	0.01086957
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0.08850932	0.06677019	0.03881988	0.04192547	0.01708075	0.00621118
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0.14596273	0.08850932	0.06677019	0.03881988	0.04192547	0.01708075
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0.07608696	0.12267081	0.12888199	0.18322981	0.14596273	0.08850932
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0.18322981	0.14596273	0.08850932	0.06677019	0.03881988	0.04192547
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0.	0.	0.	0.	0.	0.
0.00168634	0.	0.	0.	0.00337268	0.00505902
0.01180438	0.01854975	0.03541315	0.08263069	0.13322091	0.13996627
0.1989882	0.15851602	0.09612142	0.07251265	0.04215852]	
[0.	0.	0.	0.	0.	0.
0.	0.	0.	0.	0.	0.
0.	0.	0.	0.	0.	0.
0.	0.	0.	0.	0.	0.
0.	0.	0.	0.	0.	0.
0.	0.	0.	0.	0.	0.
0.	0.00176056	0.	0.	0.	0.00352113
0.00528169	0.01232394	0.0193662	0.03697183	0.08626761	0.13908451
0.14612676	0.20774648	0.16549296	0.10035211	0.07570423]	
[0.	0.	0.	0.	0.	0.
0.	0.	0.	0.	0.	0.
0.	0.	0.	0.	0.	0.
0.	0.	0.	0.	0.	0.
0.	0.	0.	0.	0.	0.

```

0.          0.          0.          0.          0.          0.
0.          0.          0.00190476 0.          0.          0.
0.00380952 0.00571429 0.01333333 0.02095238 0.04          0.09333333
0.15047619 0.15809524 0.2247619  0.17904762 0.10857143]
[0.          0.          0.          0.          0.          0.
0.          0.          0.          0.          0.          0.
0.          0.          0.          0.          0.          0.
0.          0.          0.          0.          0.          0.
0.          0.          0.          0.          0.          0.
0.          0.          0.          0.          0.          0.
0.          0.          0.          0.00213675 0.          0.
0.          0.0042735  0.00641026 0.01495726 0.02350427 0.04487179
0.10470085 0.16880342 0.17735043 0.25213675 0.2008547 ]
[0.          0.          0.          0.          0.          0.
0.          0.          0.          0.          0.          0.
0.          0.          0.          0.          0.          0.
0.          0.          0.          0.          0.          0.
0.          0.          0.          0.          0.          0.
0.          0.          0.          0.          0.          0.
0.          0.          0.          0.          0.0026738  0.
0.          0.          0.00534759 0.00802139 0.01871658 0.02941176
0.05614973 0.13101604 0.21122995 0.22192513 0.31550802]
[0.          0.          0.          0.          0.          0.
0.          0.          0.          0.          0.          0.
0.          0.          0.          0.          0.          0.
0.          0.          0.          0.          0.          0.
0.          0.          0.          0.          0.          0.
0.          0.          0.          0.          0.          0.
0.          0.          0.          0.          0.          0.00390625
0.          0.          0.          0.0078125  0.01171875 0.02734375
0.04296875 0.08203125 0.19140625 0.30859375 0.32421875]]

```

```

[ ]: def plot_dis(num_of_dock, distribution, site, time, ax):
    num = range(num_of_dock + 1)
    ax.bar(num, distribution, label='Frequency')
    ax.set_title(f'Distribution of {site} in {time}')
    ax.set_xlabel("States")
    ax.set_ylabel("Probability")

# Create a 2x3 grid of subplots
fig, axes = plt.subplots(nrows=2, ncols=3, figsize=(15, 12))

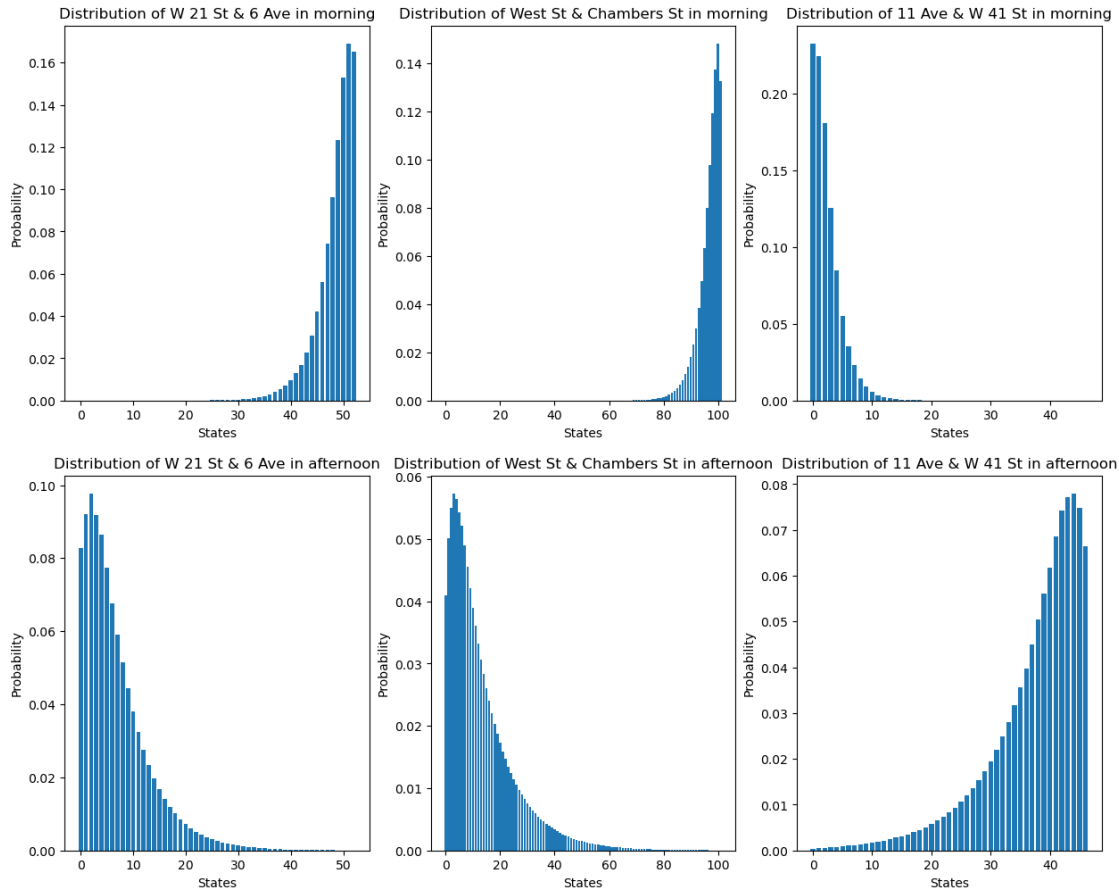
plot_dis(w21St6ave_dock, w21St6ave_morning_distribution, "W 21 St & 6 Ave",
↪ "morning", ax=axes[0, 0])
plot_dis(westStChambersSt_dock, westStChambersSt_morning_distribution, 'West St
↪ & Chambers St', "morning", ax=axes[0, 1])

```

```

plot_dis(station_11aveW41St_dock, station_11aveW41St_morning_distribution, '11 Ave & W 41 St', "morning", ax=axes[0, 2])
plot_dis(w21St6ave_dock, w21St6ave_afternoon_distribution, "W 21 St & 6 Ave", "afternoon", ax=axes[1, 0])
plot_dis(westStChambersSt_dock, westStChambersSt_afternoon_distribution, 'West St & Chambers St', "afternoon", ax=axes[1, 1])
plot_dis(station_11aveW41St_dock, station_11aveW41St_afternoon_distribution, "11 Ave & W 41 St", "afternoon", ax=axes[1, 2])

```



```

[ ]: def exp_cal(stationary_distribution, dock):
    expectation = 0
    for num in range(dock + 1):
        expectation += num * stationary_distribution[num]
    return expectation

[ ]: print(f"W 21 St & 6 Ave (Morning): {round(exp_cal(w21St6ave_morning_distribution, w21St6ave_dock))}")
print(f"W 21 St & 6 Ave (Afternoon): {round(exp_cal(w21St6ave_afternoon_distribution, w21St6ave_dock))}")

```

```

print(f"West St & Chambers St (Morning):␣
↳{round(exp_cal(westStChambersSt_morning_distribution,␣
↳westStChambersSt_dock))}")
print(f"West St & Chambers St (Afternoon):␣
↳{round(exp_cal(westStChambersSt_afternoon_distribution,␣
↳westStChambersSt_dock))}")

print(f"11 Ave & W 41 St (Morning):␣
↳{round(exp_cal(station_11aveW41St_morning_distribution,␣
↳station_11aveW41St_dock))}")
print(f"11 Ave & W 41 St (Afternoon):␣
↳{round(exp_cal(station_11aveW41St_afternoon_distribution,␣
↳station_11aveW41St_dock))}")

```

```

W 21 St & 6 Ave (Morning): 48
W 21 St & 6 Ave (Afternoon): 7
West St & Chambers St (Morning): 97
West St & Chambers St (Afternoon): 13
11 Ave & W 41 St (Morning): 2
11 Ave & W 41 St (Afternoon): 37

```

```

[ ]: w21St6ave_dock = 52
westStChambersSt_dock = 101
station_11aveW41St_dock = 46

EM_w21St6ave_dock = round(exp_cal(w21St6ave_morning_distribution,␣
↳w21St6ave_dock))
EA_w21St6ave_dock = round(exp_cal(w21St6ave_afternoon_distribution,␣
↳w21St6ave_dock))
EM_westStChambersSt_dock = round(exp_cal(westStChambersSt_morning_distribution,␣
↳westStChambersSt_dock))
EA_westStChambersSt_dock =␣
↳round(exp_cal(westStChambersSt_afternoon_distribution,␣
↳westStChambersSt_dock))
EM_station_11aveW41St_dock =␣
↳round(exp_cal(station_11aveW41St_morning_distribution,␣
↳station_11aveW41St_dock))
EA_station_11aveW41St_dock =␣
↳round(exp_cal(station_11aveW41St_afternoon_distribution,␣
↳station_11aveW41St_dock))

```

```

[ ]: dock_labels = ['W 21 St & 6 Ave', 'West St & Chambers St', '11 Ave & W 41 St']
dock_counts = [w21St6ave_dock, westStChambersSt_dock, station_11aveW41St_dock]
morning_exp = [EM_w21St6ave_dock, EM_westStChambersSt_dock,␣
↳EM_station_11aveW41St_dock]

```

```

afternoon_exp = [EA_w21St6ave_dock, EA_westStChambersSt_dock,
    ↪EA_station_11aveW41St_dock]
time_labels = ['Morning', 'Afternoon']

fig, axs = plt.subplots(2, 3, figsize=(12, 8))
fig.suptitle('Percentage of Dock Capacity')

for i in range(3):
    axs[0, i].pie([morning_exp[i], dock_counts[i] - morning_exp[i]],
    ↪labels=[f'Bike Number: {morning_exp[i]}', 'Empty'],
        autopct='%1.1f%%', startangle=90)
    axs[0, i].set_title(f'{dock_labels[i]} ({time_labels[0]})')

    axs[1, i].pie([afternoon_exp[i], dock_counts[i] - afternoon_exp[i]],
    ↪labels=[f'Bike Number: {afternoon_exp[i]}', 'Empty'],
        autopct='%1.1f%%', startangle=90)
    axs[1, i].set_title(f'{dock_labels[i]} ({time_labels[1]})')

plt.show()

```

