# Call Volume Trend Analysis

Data Analytics Project



Project By:

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# Project Description

This project delves into Customer Experience (CX) analytics, focusing on the inbound calling team of a company. Spanning 23 days, the dataset provides details like agent names, queue times, call timestamps, durations, and call statuses (abandoned, answered, or transferred). The CX team, pivotal in any organization, analyzes customer feedback and data to drive insights for the organization. With Al-powered tools like IVR, RPA, Predictive Analytics, and Intelligent Routing, the project aims to optimize inbound customer support. By handling incoming calls effectively, the goal is to attract, engage, and delight customers, fostering long-term loyalty and advocacy.

# Tech-Stack Used

Jupyter Notebook - Version 7.0.7 is used to implement Python with various other libraries :

- Pandas Data Manipulation and Analysis.
- Numpy Mathematical Computations.
- Matplotlib Data Visualization.
- Seaborn Advanced Data Visualization.

<u>Jupyter Notebook Github Link:</u> &

# Tasks

- A. Average Call Duration
- B. Call Volume Analysis
- C. Manpower Planning
- D. Night Shift Manpower Planning

# Average Call Duration

### Approach

- The 'data' dataframe is grouped by 'Time\_Bucket'.
- Now the average of the 'Call\_Duration' is calculated on this grouped data.
- The result is stored in avg\_time and barplot() is used on it as shown.

```
gd = data.groupby(cols[6])
avg_time = gd[cols[8]].mean()

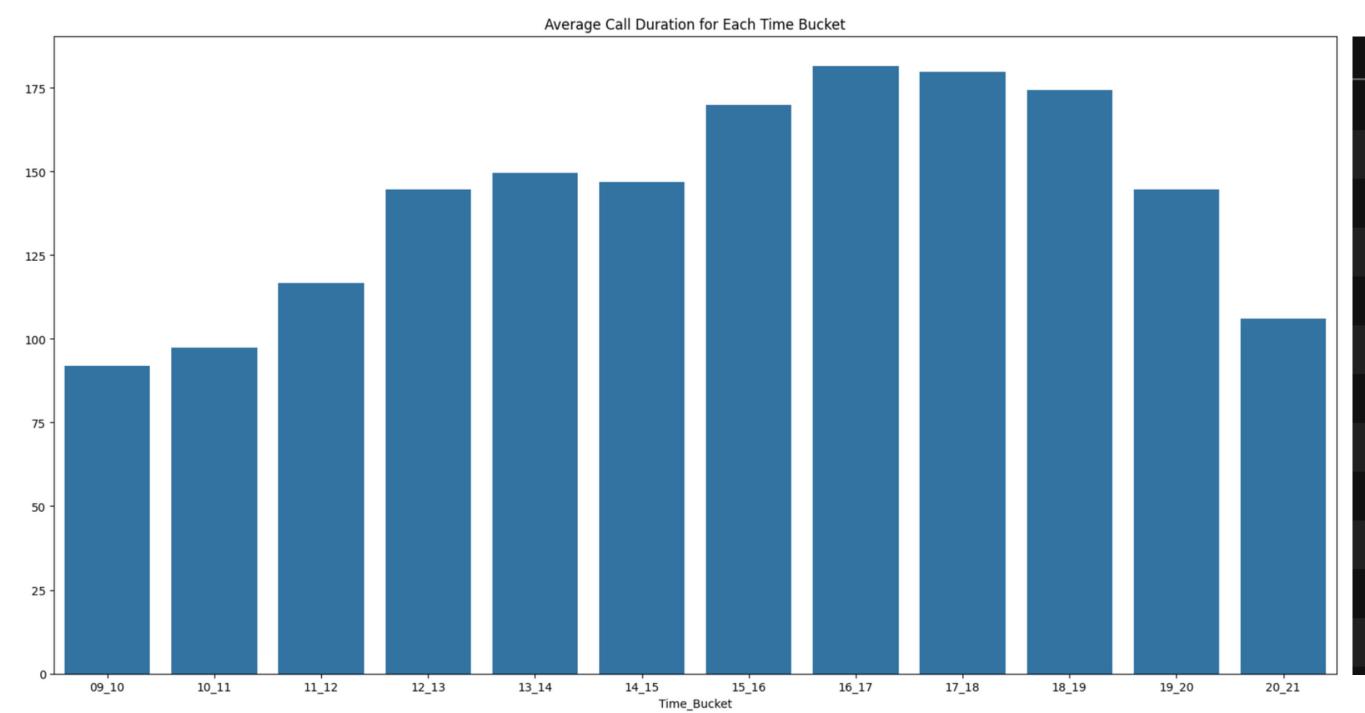
plt.figure( figsize=(20,10) )

sns.barplot(x=avg_time.index, y=avg_time.values)
plt.title("Average Call Duration for Each Time Bucket")

plt.show()
```

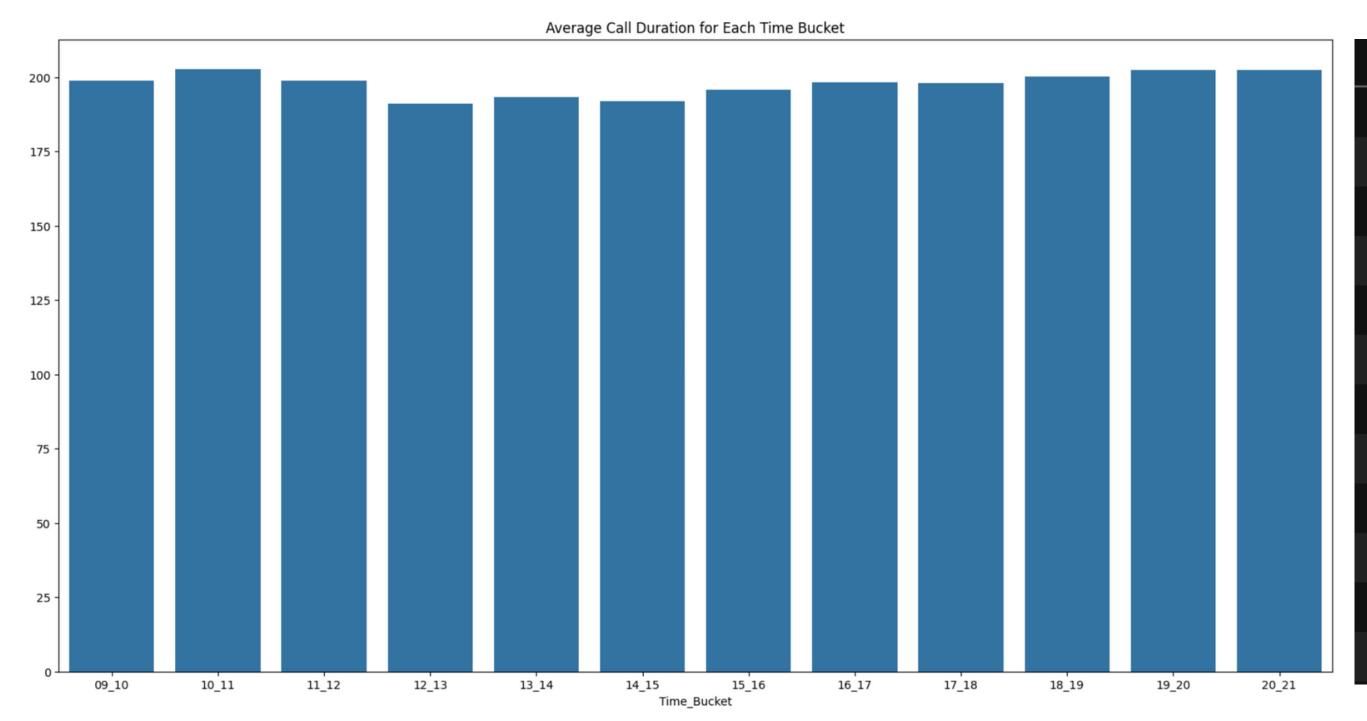
 The same steps are repeated on the data exclunding calls that are abandon with call duration of 0s.

#### Average Call Duration - Including Abandon Calls



Time_Bucket	Call_Seconds (s)
09_10	92.0
10_11	97.0
11_12	117.0
12_13	145.0
13_14	150.0
14_15	147.0
15_16	170.0
16_17	181.0
17_18	180.0
18_19	174.0
19_20	145.0
20_21	106.0

#### Average Call Duration - Excluding Abandon Calls



all_Seconds (s)
199.0
203.0
199.0
191.0
193.0
192.0
196.0
198.0
198.0
200.0
202.0
203.0

- The average call duration exhibited a trend of increase until the 16\_17 time bucket, reaching its peak with 181 s. However, it began to decrease afterwards.
- Upon excluding abandoned calls, the average call duration showed a narrower range, with the highest observed in the 16\_17 time bucket at 203 s and the lowest in the 09\_10 time bucket at 191 s.
- The fluctuations in average call duration could be attributed to abandoned calls skewing the data.
- Excluding abandoned calls provides a clearer picture of actual call handling efficiency, revealing a more uniform distribution of average durations across time buckets.

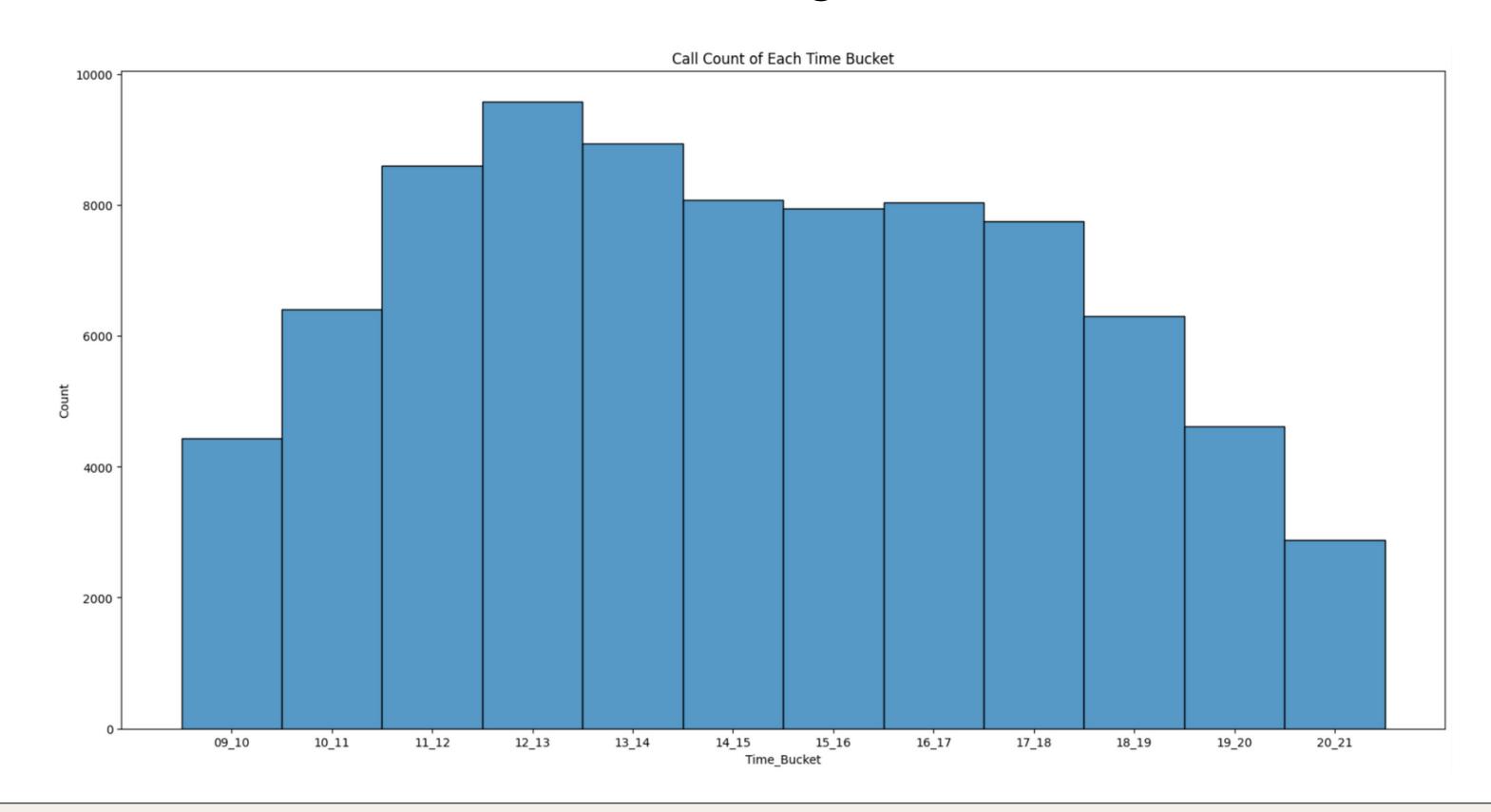
# Call Volume Analysis

### Approach

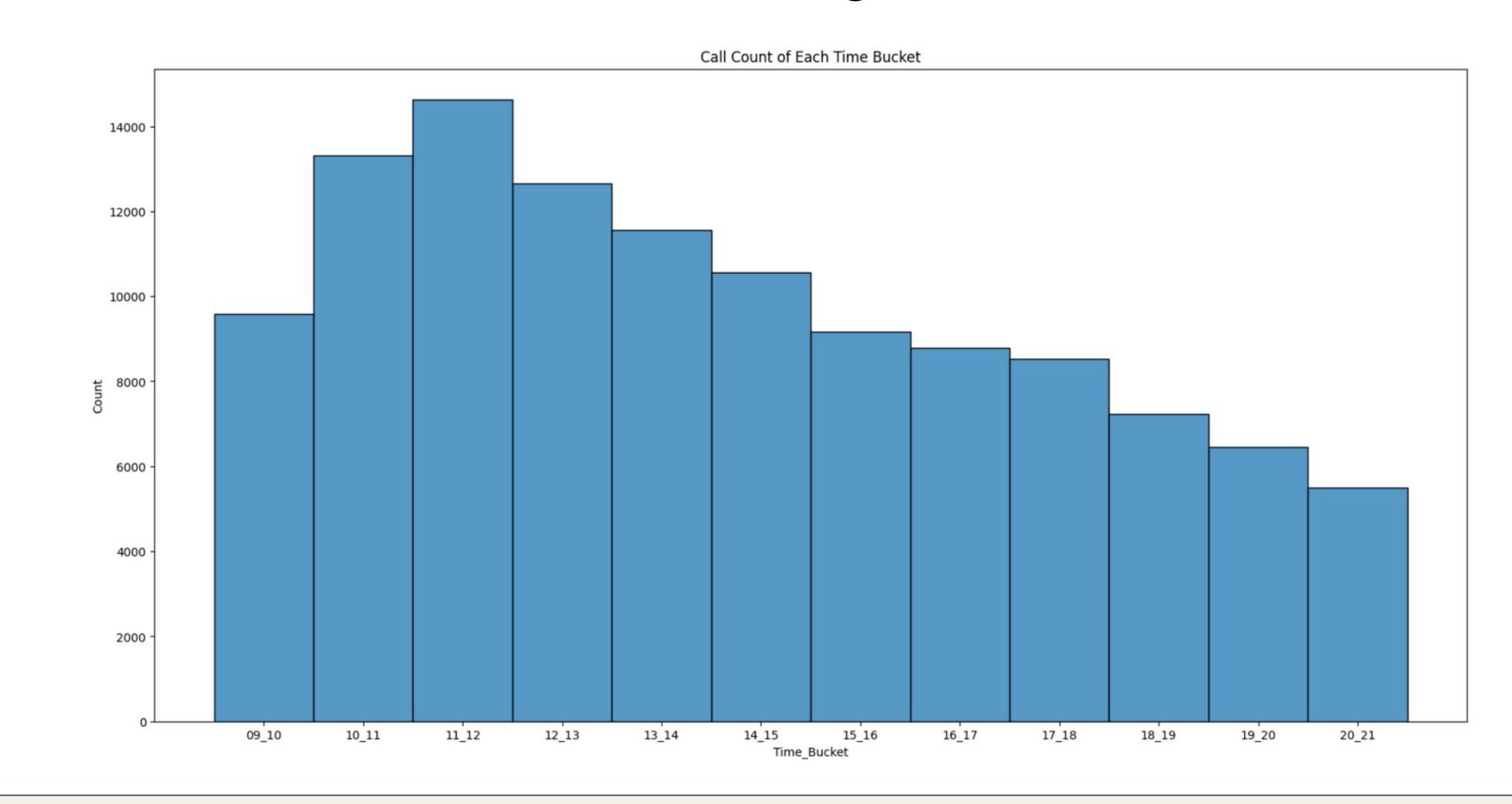
- The data's 'Time\_Bucket' column is passed to the sns.histplot()
  function.
- This generates a histogram based on the values in the 'Time\_Bucket' column.
- data2 is created by filtering out the rows with abandon calls.
- And now, data2's 'Time\_Bucket' column is passed onto sns.histplot().

```
plt.figure(figsize=(20,10))
sns.histplot(data[cols[6]])
plt.title("Call Count of Each Time Bucket")
plt.show()
```

#### Call Volume - Including Abandon Calls



#### Call Volume - Excluding Abandon Calls



- In the case the Call Volume analysis, it observed that the trend was similar before and after removing abandoned calls.
- Including abandoned calls in the analysis revealed the highest call count in the time bucket 12\_13, with a consistent increase followed by a decrease in volume.
- Excluding abandoned calls shifted the peak to an earlier time bucket, 11\_12, and the subsequent decrease in call volume began earlier.
- The significant difference in the pattern occurred in the time bucket 12\_13, which registered the highest call volume when abandoned calls were included, but did not retain its highest position after the removal of abandoned calls..

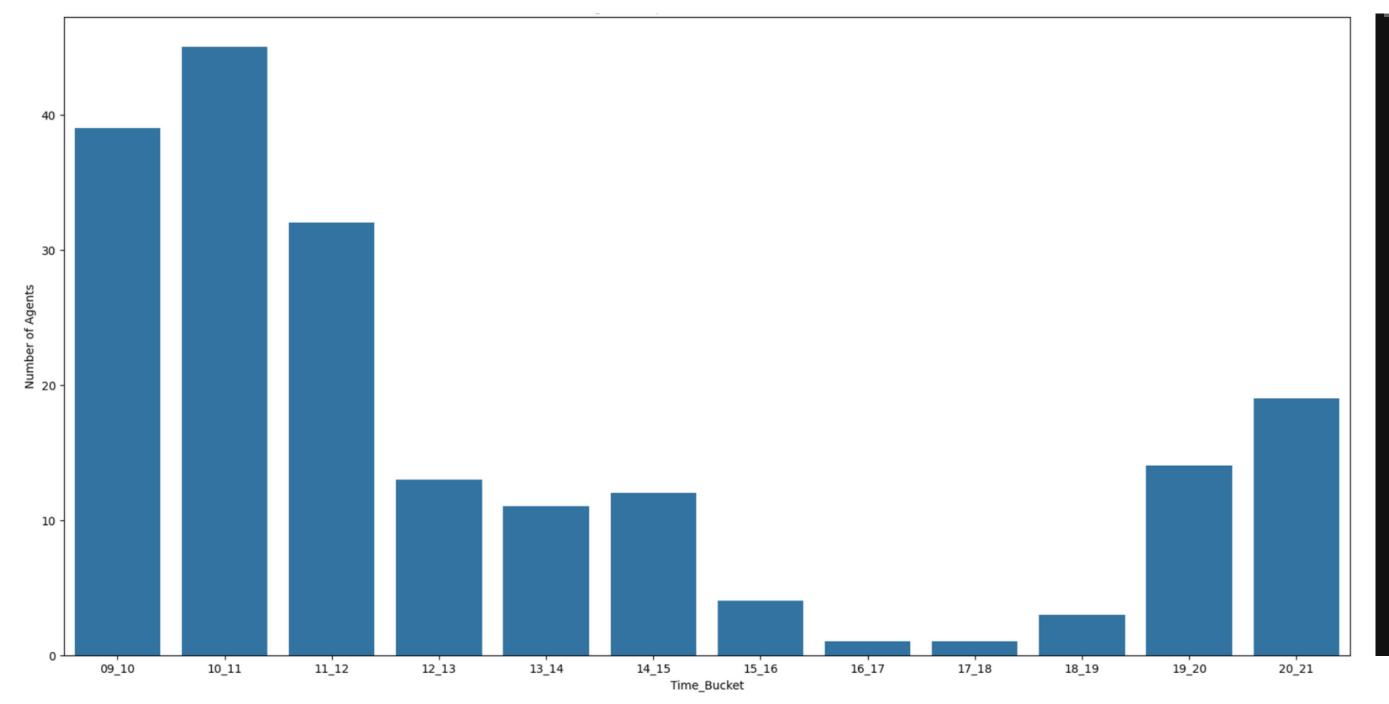
# Manpower Planning

# Approach

- First the percentage of calls answered, abandon and removed are calculated in each Time\_Bucket and stored in 'bs'.
- Then only the percentage of answered calls are filtered out from bs, which is stored in 'ans\_rate'.
- Now, number of agents working in each Time\_Bucket is stored in 'agents'.
- The 'ans\_rate' is divided by 'agents', and multiplied by 90 to get the number of agents required to increase the answer rate to 90.
- This is stored in 'min\_req' and barplot() is used to visualize.

```
tot = gd[cols[9]].count()
bs = (gd[cols[9]].value_counts()/tot)*100
bs = bs.to_frame().reset_index()
bs.rename( columns = {0:'Rate'}, inplace=True)
bs
ans_calls = bs[bs['Call_Status']=='answered']
ans_calls['Rate']
ans_rate = pd.Series( ans_calls['Rate'].values, index=ans_calls['Time_Bucket'] )
ans_rate
agents = gd[cols[1]].nunique()
agents
req = np.ceil((agents/ans_rate)*90)
req
min_req = (req-agents).astype(int)
min_req
plt.figure(figsize=(20,10))
sns.barplot( x=min_req.index, y=min_req.values)
plt.title("Agents required for Night shift")
plt.xlabel('Time_Bucket')
plt.ylabel('Number of Agents')
plt.show()
```

#### Agents Required to Reduce the Abandon Rate to 10%



Time_Bucket		
09_10	39	
10_11	45	
11_12	32	
12_13	13	
13_14	11	
14_15	12	
15_16	4	
16_17	1	
17_18	1	
18_19	3	
19_20	14	
20_21	19	

- A total of 194 agents are required to reduce the percentage of abandoned calls from 30% to 10%.
- The distribution of these 194 agents is inconsistent across different time buckets due to the variation in the percentage of abandoned calls.
- The percentage of abandoned calls was nearly 50% for the first three time buckets, so these 3 time buckets alone require 116 agents out of the total 194 agents, which is almost 60%.
- But during time buckets 15\_16 to 18\_19, the percentage of abandoned calls was already close to 10%, so these 4 time buckets require only 9 agents in total, which is just 4.6%.
- The remaining time buckets require an average of 14 agents.

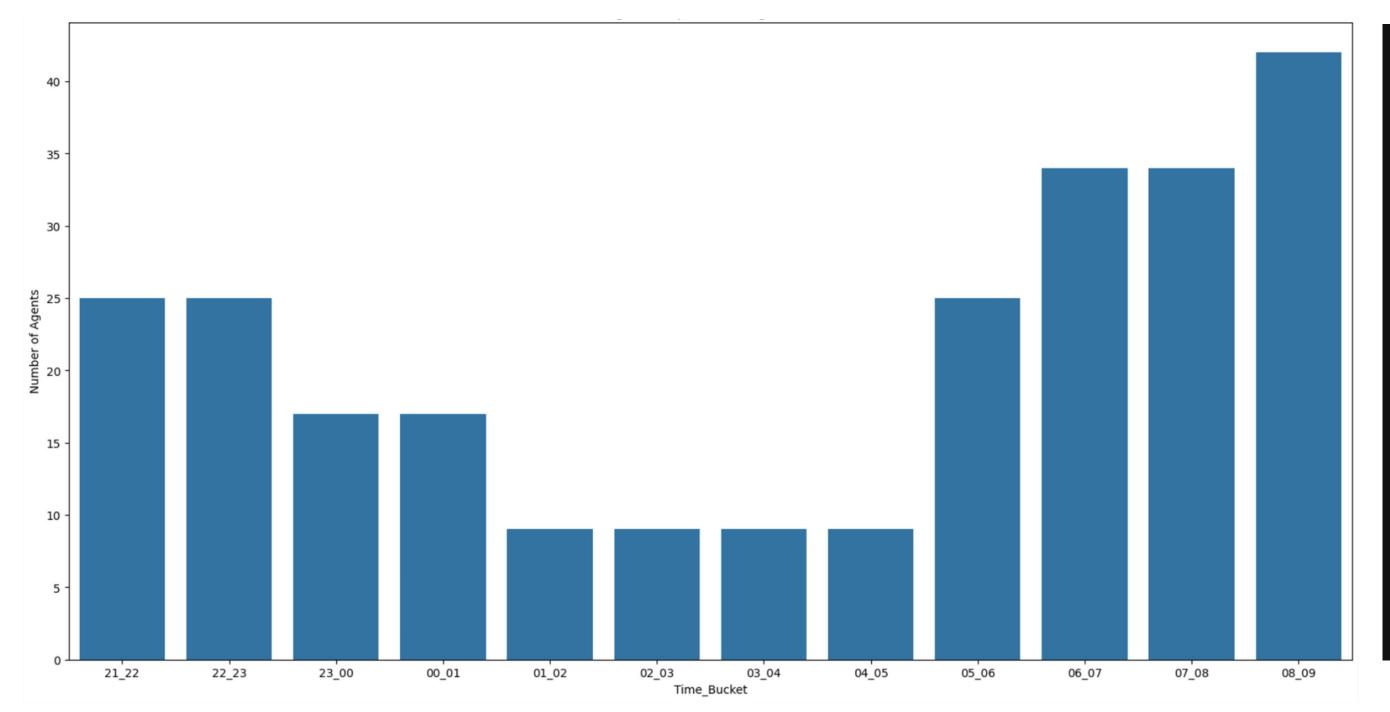
# Night Shift Manpower Planning

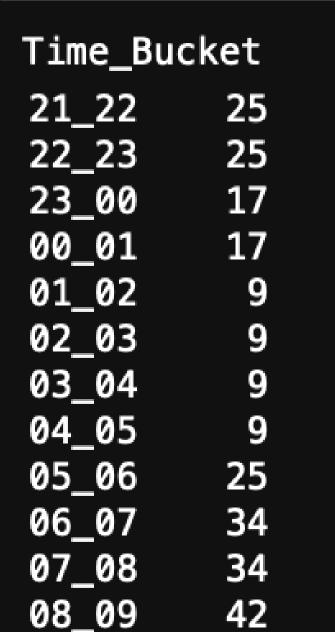
## Approach

- First the given distribution is stored in 'dist' Series.
- This 'dist' values are divided by 100 and multiplied total number of calls to get expected number of calls in each Time\_Bucket, which stored on calls.
- Then the number of agents required to maintain abandon rate at 10% is calculated from 'calls' using the ratio of average agents to average calls which are found from the 'req' and 'gd' reespectively.
- Finally boxplot() function is used for visualization.

```
dist = pd.Series([ 3, 3, 2, 2, 1, 1, 1, 1, 3, 4, 4, 5],
                 index=[ '21_22', '22_23', '23_00', '00_01', '01_02', '02_03', '03_04', '04_05', '05_06', '06_07', '07_08', '08_09'])
calls = np.round((dist/100)*data.shape[0])
calls
avg_agents = np.ceil(req.mean())
avg_agents
avg_calls = int(gd[cols[2]].count().mean())
avg_calls
agents_req = np.ceil((avg_agents/avg_calls)*calls).astype(int)
agents_req
plt.figure(figsize=(20,10))
sns.barplot( x=agents_req.index, y=agents_req.values)
plt.title("Agents required for Night shift")
plt.xlabel('Time_Bucket')
plt.ylabel('Number of Agents')
plt.show()
```

#### Agents Required for Night Shift





- A total of 225 agents are required to maintain the abandoned call rate at 10% during night shifts. The distribution of these agents precisely matches the given distribution as it was derived from this.
- In the original distribution, time buckets from 01\_02 to 04\_05 each had 1 call, now requiring 9 agents each, representing 16%.
- Time buckets 23\_00 and 00\_01, which previously had 2 calls each, now require 17 agents each, accounting for 15.11%. Similarly, time buckets 21\_22, 22\_23, and 05\_06 with 3 calls each, now require 25 agents each, representing 33.33%.
- Subsequently, time buckets 06\_07 and 07\_08, with 4 calls, now requires 34 agents each, making up 30.22%. And finally, time bucket 08\_09, with 5 calls, now requires 42 agents, accounting for 18.66%.