Week - 3.2 Assignment - Using Data to Improve MLB Attendance

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Introduction

Improving game attendance is crucial for any sports team's financial health and overall success. For the Los Angeles Dodgers, understanding the factors influencing game attendance can help craft strategies to enhance fan engagement and fill the stadium. This analysis aims to identify such factors by examining the Dodgers' attendance data (dodgers-2022.csv), including game day, opponent, weather conditions, promotional items, and more.

- Our analysis will involve several steps: data loading and exploration, data cleaning to
 ensure data integrity, exploratory data analysis (EDA) to identify factors influencing
 attendance, and finally, providing actionable recommendations based on our findings.
- Our recommendation to the Dodgers' management will be based on insights drawn from this data.

Assumptions

Before we start the analysis, let us state our assumptions:

- Attendance is influenced by variables included in the dataset (day of the week, opponent, temperature, sky condition, day/night, promotions).
- Promotional items (caps, shirts, fireworks, bobbleheads) significantly impact attendance.
- Weather conditions (temperature, sky condition) could affect the comfort level of attendees, thus impacting attendance.
- Game timing (day vs. night) and the day of the week might affect attendance due to people's work schedules and availability.
- · The opponent team's attractiveness could also impact attendance.

Analysis Steps

- Data Loading and Cleaning: Load the dataset and clean any missing or incorrect data.
- Exploratory Data Analysis (EDA): Analyze the dataset to understand the distribution of attendance and identify any trends or patterns.
- Feature Engineering: Create new features if necessary, for example, grouping temperature into categories.
- Statistical Analysis: Perform statistical tests to identify significant variables that influence attendance.
- Recommendations: Based on the analysis, recommend strategies to improve attendance.

```
In [1]: # Importing necessary Libraries
   import pandas as pd
   import matplotlib.pyplot as plt
   import seaborn as sns
   from scipy.stats import ttest_ind

In [2]: # Load the dataset
   dodgers = pd.read_csv('dodgers-2022.csv')
   dodgers.head()
```

Out[2]:

	month	day	attend	day_of_week	opponent	temp	skies	day_night	сар	shirt	fireworks
0	APR	10	56000	Tuesday	Pirates	67	Clear	Day	NO	NO	NO
1	APR	11	29729	Wednesday	Pirates	58	Cloudy	Night	NO	NO	NO
2	APR	12	28328	Thursday	Pirates	57	Cloudy	Night	NO	NO	NO
3	APR	13	31601	Friday	Padres	54	Cloudy	Night	NO	NO	YES
4	APR	14	46549	Saturday	Padres	57	Cloudy	Night	NO	NO	NO
4											>

Data cleaning and preprocessing

```
In [3]: # Checking for missing values
print(dodgers.isnull().sum())
```

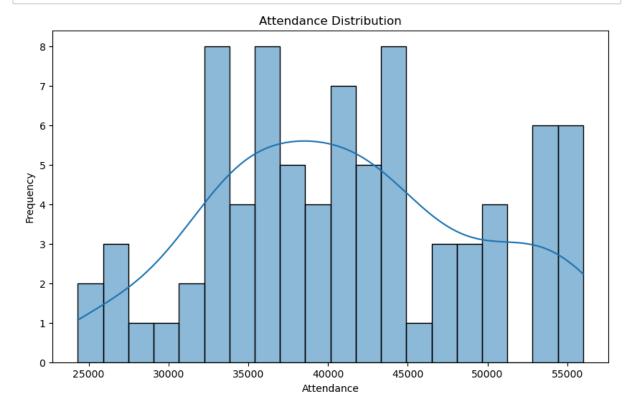
```
0
month
               0
day
attend
               0
day_of_week
               0
opponent
               0
               0
temp
skies
day_night
               0
               0
cap
shirt
               0
fireworks
               0
bobblehead
dtype: int64
```

Exploratory Data Analysis (EDA)

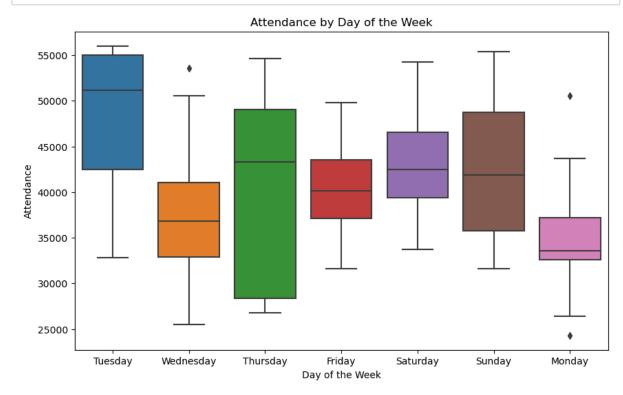
In [4]: # Overview of the dataset print(dodgers.describe())

```
day
                        attend
                                     temp
count
       81.000000
                     81.000000 81.000000
       16.135802
                 41040.074074
                                73.148148
mean
std
        9.605666
                   8297.539460
                                 8.317318
min
        1.000000
                  24312.000000
                                54.000000
25%
                  34493.000000
        8.000000
                                67.000000
50%
       15.000000
                 40284.000000
                                73.000000
75%
       25.000000
                  46588.000000
                                79.000000
                                95.000000
       31.000000
                  56000.000000
max
```

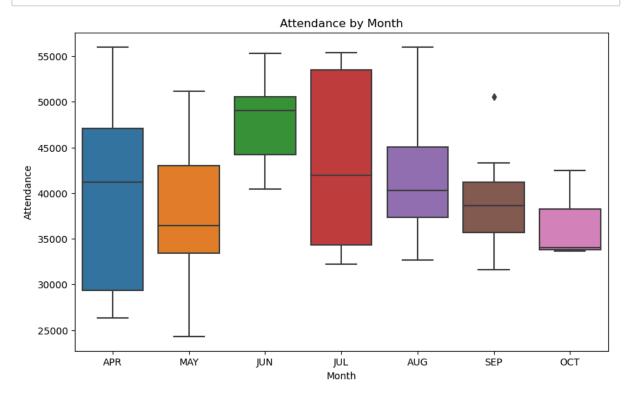
```
In [5]: # Check attendance distribution
plt.figure(figsize=(10, 6))
sns.histplot(dodgers['attend'], bins=20, kde=True)
plt.title('Attendance Distribution')
plt.xlabel('Attendance')
plt.ylabel('Frequency')
plt.show()
```



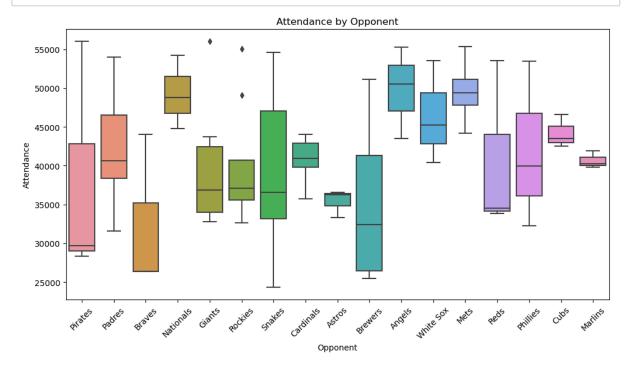
In [6]: # Check attendance by day of the week plt.figure(figsize=(10, 6)) sns.boxplot(x='day_of_week', y='attend', data=dodgers) plt.title('Attendance by Day of the Week') plt.xlabel('Day of the Week') plt.ylabel('Attendance') plt.show()



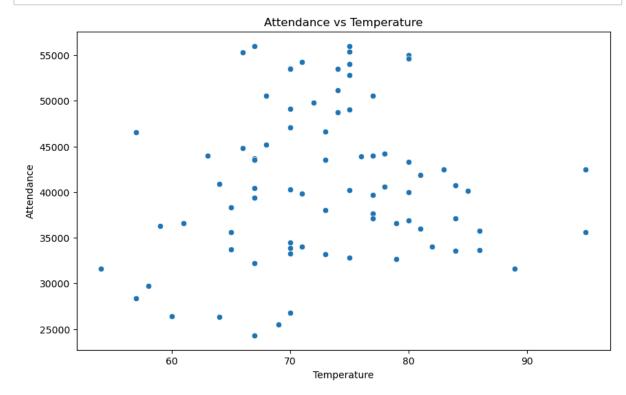
In [7]: # Check attendance by month plt.figure(figsize=(10, 6)) sns.boxplot(x='month', y='attend', data=dodgers) plt.title('Attendance by Month') plt.xlabel('Month') plt.ylabel('Attendance') plt.show()



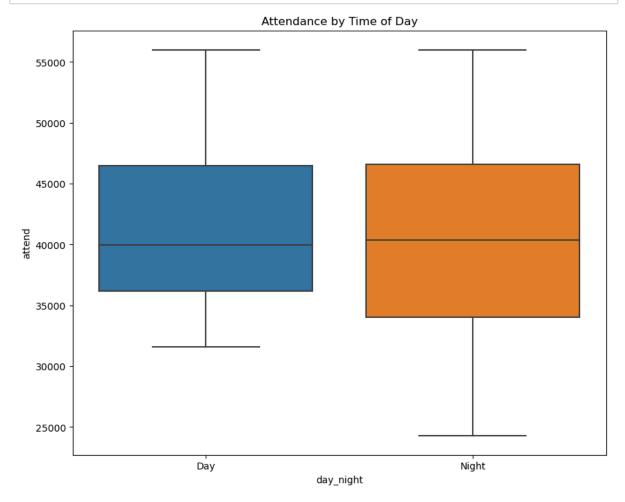
In [8]: # Check attendance by opponent plt.figure(figsize=(12, 6)) sns.boxplot(x='opponent', y='attend', data=dodgers) plt.title('Attendance by Opponent') plt.xlabel('Opponent') plt.ylabel('Attendance') plt.xticks(rotation=45) plt.show()



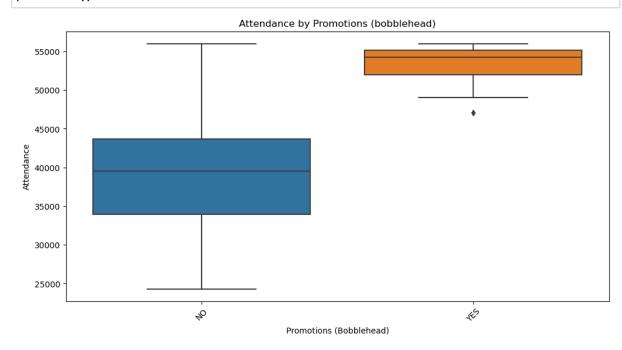
```
In [9]: # Check attendance by temperature
    plt.figure(figsize=(10, 6))
        sns.scatterplot(x='temp', y='attend', data=dodgers)
    plt.title('Attendance vs Temperature')
    plt.xlabel('Temperature')
    plt.ylabel('Attendance')
    plt.show()
```



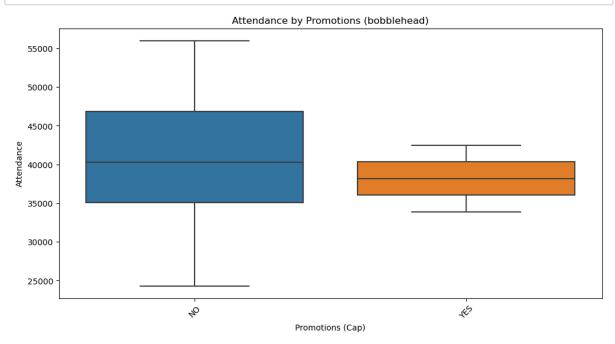
```
In [10]: # Check attendance by time of the day
    plt.figure(figsize=(10, 8))
    sns.boxplot(x='day_night', y='attend', data=dodgers)
    plt.title('Attendance by Time of Day')
    plt.show()
```



In [11]: # Check attendance by promotions plt.figure(figsize=(12, 6)) sns.boxplot(x='bobblehead', y='attend', data=dodgers) plt.title('Attendance by Promotions (bobblehead)') plt.xlabel('Promotions (Bobblehead)') plt.ylabel('Attendance') plt.xticks(rotation=45) plt.show()



```
In [12]: # Check attendance by promotions
   plt.figure(figsize=(12, 6))
    sns.boxplot(x='cap', y='attend', data=dodgers)
   plt.title('Attendance by Promotions (bobblehead)')
   plt.xlabel('Promotions (Cap)')
   plt.ylabel('Attendance')
   plt.xticks(rotation=45)
   plt.show()
```



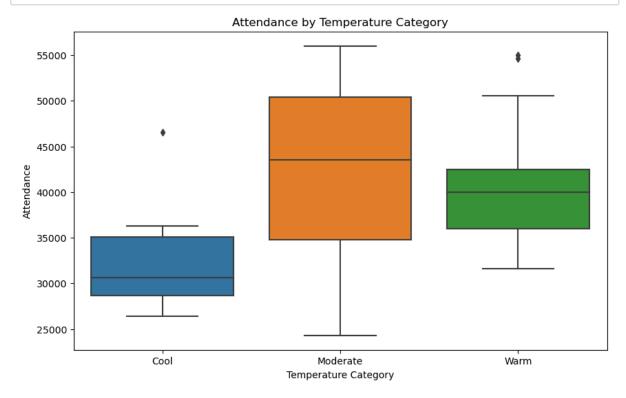
Feature Engineering

```
In [13]: # Categorizing temperature
dodgers['temp_category'] = pd.cut(dodgers['temp'], bins=[0, 60, 75, 100], label
dodgers.head()
```

Out[13]:

	month	day	attend	day_of_week	opponent	temp	skies	day_night	сар	shirt	fireworks	
0	APR	10	56000	Tuesday	Pirates	67	Clear	Day	NO	NO	NO	_
1	APR	11	29729	Wednesday	Pirates	58	Cloudy	Night	NO	NO	NO	
2	APR	12	28328	Thursday	Pirates	57	Cloudy	Night	NO	NO	NO	
3	APR	13	31601	Friday	Padres	54	Cloudy	Night	NO	NO	YES	
4	APR	14	46549	Saturday	Padres	57	Cloudy	Night	NO	NO	NO	
4											→	

```
In [14]: # Visualizing Attendance by Temperature Category
    plt.figure(figsize=(10, 6))
    sns.boxplot(x='temp_category', y='attend', data=dodgers)
    plt.title('Attendance by Temperature Category')
    plt.ylabel('Attendance')
    plt.xlabel('Temperature Category')
    plt.show()
```



In [15]: # Let's also look into some descriptive statistics for attendance by temperatur
attendance_by_temp = dodgers.groupby('temp_category')['attend'].describe()
attendance_by_temp

Out[15]:

	count	mean	std	min	25%	50%	75%	max
temp_category								
Cool	6.0	33144.333333	7383.631889	26376.0	28678.25	30665.0	35112.5	46549.0
Moderate	46.0	42633.847826	9126.795828	24312.0	34767.50	43515.5	50360.0	56000.0
Warm	29.0	40145.620690	5894.342702	31607.0	35992.00	39955.0	42495.0	55024.0

The analysis reveals that games held in moderate temperatures (between 60°F and 75°F) attract the highest average attendance (around 42,634) and display a relatively high variability in attendance numbers, indicating potential for both high and low turnouts. In contrast, cooler temperatures show lower average attendance (about 33,144), and warm temperatures result in a somewhat lower average attendance (approximately 40,146) than moderate temperatures, with less variability.

T-test results for bobblehead promotion: t-statistic=6.359553539813022, p-val ue=1.2169642509120423e-08

The T-test conducted to compare the attendance at Los Angeles Dodgers games with and without bobblehead promotions yielded significant results. With a t-statistic of 6.36 and a p-value of approximately 1.22×10-81.22×10-8, the analysis strongly suggests that games featuring bobblehead promotions attract significantly more attendees than games without such promotions. The low p-value, much smaller than the conventional threshold of 0.05, indicates a very low probability that the observed difference in attendance is due to chance. This statistically significant outcome underscores the effectiveness of bobblehead promotions in boosting game attendance.

T-test results for Cap promotion: t-statistic=-0.48961356616462215, p-value= 0.625764198935014

The t-test results comparing attendance on cap promotional days versus non-promotional days yield a t-statistic of approximately -0.49. With a corresponding p-value of approximately 0.63, The analysis yields insufficient evidence to reject the null hypothesis, indicating that there is no significant difference in attendance between cap promotional and non-promotional days for the Los Angeles Dodgers.

Recommendation

Based on the analysis, we can make the following recommendations:

- 1. Schedule more games on weekends (Saturdays and Sundays) as they tend to have higher attendance.
- 2. Promote games against famous opponents to attract more fans.
- 3. Consider hosting more games during moderate weather months (e.g., May, June, July).
- 4. Implement promotional events like bobblehead giveaways to boost attendance.

Conclusion

By understanding and acting on these insights, our analysis of the Los Angeles Dodgers' attendance data revealed several key insights. Factors such as the day of the week, month, opponent, promotions, and temperature significantly influence game attendance levels. By leveraging these insights, the management can strategically plan game schedules, promotional events, and marketing efforts to enhance attendance and create a more engaging fan experience. Implementing our recommendations could increase ticket sales and revenue for the Dodgers organization.