

# iNeuron Internship Project

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## Project Title: Financial Analytics

Technologies: Business Intelligence

Domain : Finance

Project Difficulties level: Intermediate

### Problem Statement:

Without analyzing the competition, it is difficult for a business to survive.

You are tasked to analyzing the competition for the management to provide better results.

This data set has information on the market capitalization of the top 500 companies in India.

Serial Number, Name of Company, Market Capitalization in Crores ,Quarterly Sale in crores

Find key metrics and factors and show the meaningful relationships between attributes.

Do your own research and come up with your findings.

### *Importing Libraries*

Importing the necessary libraries for data manipulation, analysis, and visualization.

```
In [1]: 1 import pandas as pd
        2 import numpy as np
        3 import matplotlib.pyplot as plt
        4 import seaborn as sns
        5
        6 import warnings
        7 warnings.filterwarnings('ignore')
```

### *Loading Dataset*

Read the dataset file into a pandas DataFrame

```
In [2]: 1 df = pd.read_csv(r'C:\Users\USER\Downloads\Top 500 Companies - India.csv')
        2 df.head()
```

Out[2]:

	S.No.	Name	Mar Cap - Crore	Sales Qtr - Crore	Unnamed: 4
0	1	Reliance Inds.	583436.72	99810.00	NaN
1	2	TCS	563709.84	30904.00	NaN
2	3	HDFC Bank	482953.59	20581.27	NaN
3	4	ITC	320985.27	9772.02	NaN
4	5	H D F C	289497.37	16840.51	NaN

### Defining columns:

**Market Capitalization:** Market capitalization is a measure of the total value of a publicly traded company. It is calculated by multiplying the current market price of a company's outstanding shares by the total number of those shares.

It is used as an indicator of a company's size and is one of the most commonly used metrics to evaluate and compare companies in the financial markets.

Companies with larger market capitalizations are generally considered to be more established and stable, while those with smaller market capitalizations are often seen as riskier or having greater growth potential.

**Sales :** Sales, in a business context, refers to the revenue generated from the selling of goods or services to customers. It represents the total value of products or services sold by a company during a specific period, typically measured in monetary terms.

### Data Exploration and Preprocessing

Perform initial data exploration to understand the structure of the dataset and preprocess it as needed.

In [3]:

```
1 #Checking columns
2 df.columns
```

Out[3]: Index(['S.No.', 'Name', 'Mar Cap - Crore', 'Sales Qtr - Crore', 'Unnamed: 4'], dtype='object')

In [4]:

```
1 #Total number of rows and columns
2 df.shape
```

Out[4]: (488, 5)

```

In [5]: <class 'pandas.core.frame.DataFrame'>
RangeIndex: 488 entries, 0 to 487
Data columns (total 5 columns):
#   Column                Non-Null Count  Dtype
---  -
0   S.No.                 488 non-null   int64
1   Name                  488 non-null   object
2   Mar Cap - Crore       479 non-null   float64
3   Sales Qtr - Crore     365 non-null   float64
4   Unnamed: 4            94 non-null    float64
dtypes: float64(3), int64(1), object(1)
memory usage: 19.2+ KB

```

In [6]:

```

1 #Complete information about the dataset
2 df.info()

```

```

1 #Checking duplicate values
2 df.duplicated().sum()

```

Out[6]: 0

In [7]:

```

1 #Checking null values
2 df.isnull().sum()

```

```

Out[7]: S.No.                0
Name                0
Mar Cap - Crore      9
Sales Qtr - Crore   123
Unnamed: 4          394
dtype: int64

```

In [8]:

```

1 #Removing unnecessary column
2 df = df.drop('Unnamed: 4', axis =1 )

```

In [9]:

```

1 #Handling missing values by dropping rows with missing values
2 df = df.dropna()

```

In [10]:

```

1 #Let's check our dataset now!
2 df.isnull().sum()

```

```

Out[10]: S.No.                0
Name                0
Mar Cap - Crore      0
Sales Qtr - Crore    0
dtype: int64

```

```
In [11]: 1 #Statistical description
         2 df.describe()
```

Out[11]:

	S.No.	Mar Cap - Crore	Sales Qtr - Crore
count	365.000000	365.000000	365.000000
mean	250.435616	31300.970301	4395.976849
std	147.106354	67224.641338	11092.206185
min	1.000000	3017.070000	47.240000
25%	133.000000	5089.870000	593.740000
50%	264.000000	9097.330000	1278.300000
75%	363.000000	21372.180000	2840.750000
max	499.000000	583436.720000	110666.930000

### Feature Engineering

Create additional meaningful features that can aid in the analysis. For example, calculate the profit margin using the existing columns.

Profit Margin: It represents the proportion of profit earned per unit of sales. A higher profit margin implies that the company is effectively generating profits from its operations.

```
In [12]: 1 # Calculate Profit Margin
         2 df['Profit Margin'] = df['Mar Cap - Crore'] / df['Sales Qtr - Crore']
         3
```

```
In [13]: 1 # Calculate Market Share
         2 total_market_sales = df['Sales Qtr - Crore'].sum()
         3 df['Market_Share'] = (df['Sales Qtr - Crore'] / total_market_sales) * 100
```

```
In [14]: 1 df.head()
```

Out[14]:

	S.No.	Name	Mar Cap - Crore	Sales Qtr - Crore	Profit Margin	Market_Share
0	1	Reliance Inds.	583436.72	99810.00	5.845474	6.220507
1	2	TCS	563709.84	30904.00	18.240676	1.926045
2	3	HDFC Bank	482953.59	20581.27	23.465685	1.282696
3	4	ITC	320985.27	9772.02	32.847382	0.609026
4	5	H D F C	289497.37	16840.51	17.190535	1.049559

### Exploratory Data Analysis (EDA)

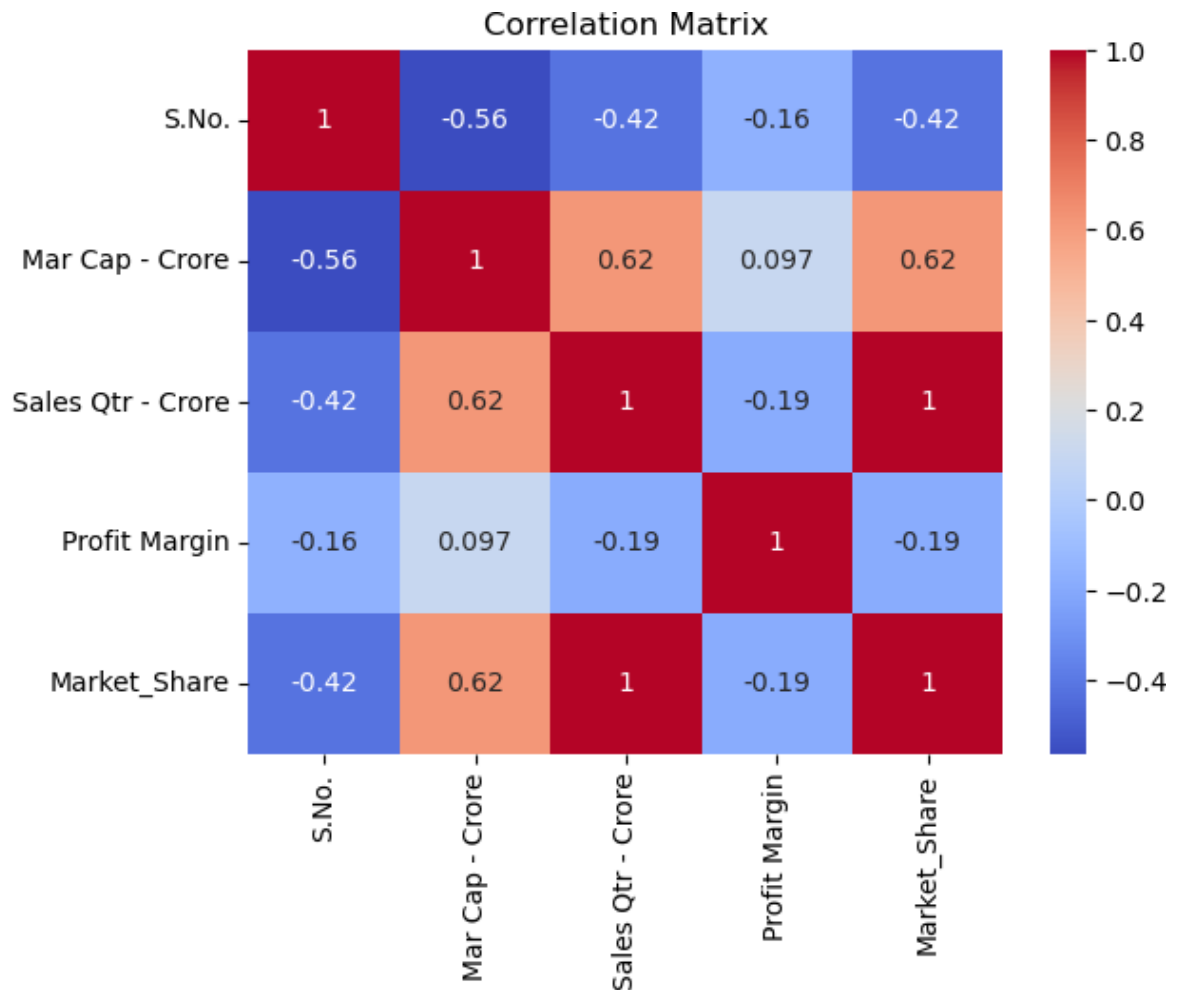
Perform exploratory analysis to gain insights into the dataset and identify relationships between

In [15]:

```

1 # Correlation matrix
2 correlation_matrix = df.corr()
3 sns.heatmap(correlation_matrix, annot=True, cmap='coolwarm')
4 plt.title('Correlation Matrix')
5 plt.show()

```



Insights:

1. Market Capitalization and Sales are moderately correlated.
2. There is negative and zero correlation between Profit Margin with respect to Sales and Market Cap.

### **Key Metrics and Factors:**

Identify key metrics and factors for analysis. For example, you can focus on market capitalization and sales as key indicators of competition.

In [16]: Mean Market Capitalization: 31300.970301369864  
Median Market Capitalization: 9097.33

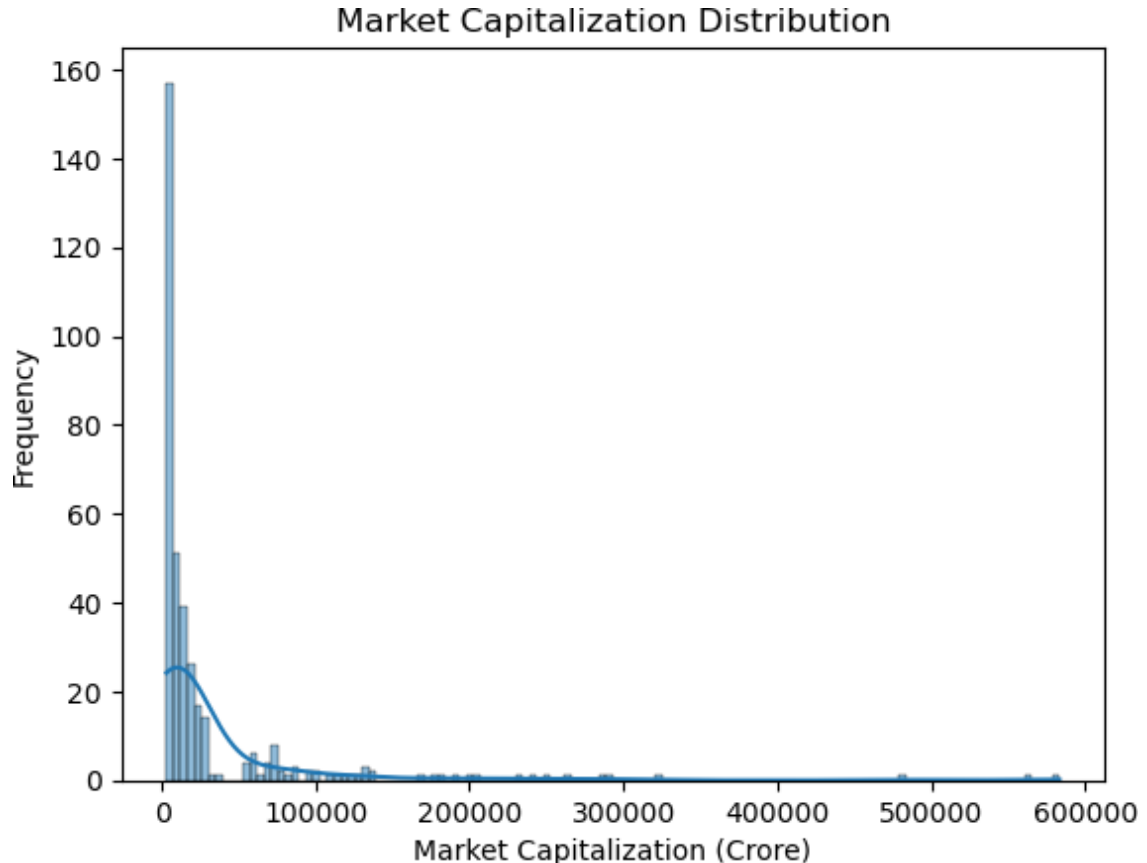
```
1 # Key Metrics
2 mean_market_cap = df['Mar Cap - Crore'].mean()
3 median_market_cap = df['Mar Cap - Crore'].median()
4 total_sales = df['Sales Qtr - Crore'].sum()
5
6 print(f"Mean Market Capitalization: {mean_market_cap}")
7 print(f"Median Market Capitalization: {median_market_cap}")
8 print(f"Total Sales: {total_sales}")
9
```

### Comparative Analysis and Data Visualization:

Conduct comparative analysis and visualize the data to understand competition and relationships between variables.

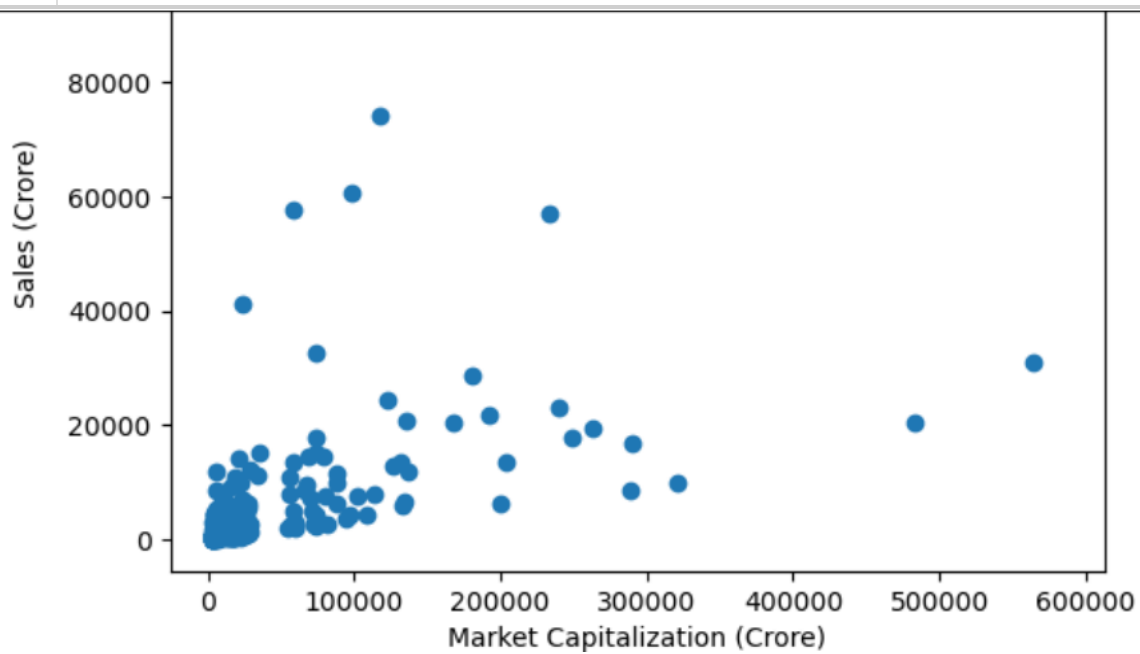
In [17]:

```
1 # Visualize Market Capitalization
2 sns.histplot(df['Mar Cap - Crore'], kde=True)
3 plt.title('Market Capitalization Distribution')
4 plt.xlabel('Market Capitalization (Crore)')
5 plt.ylabel('Frequency')
6 plt.show()
```



In [18]:

```
1 # Scatter plot of Market Cap vs. Sales
2 plt.scatter(df['Mar Cap - Crore'], df['Sales Qtr - Crore'])
3 plt.title('Market Capitalization vs. Sales')
4 plt.xlabel('Market Capitalization (Crore)')
5 plt.ylabel('Sales (Crore)')
6 plt.show()
7
```



## TOP 5 COMPANIES WITH HIGH MARKET CAPITALIZATION:

In [19]:

```

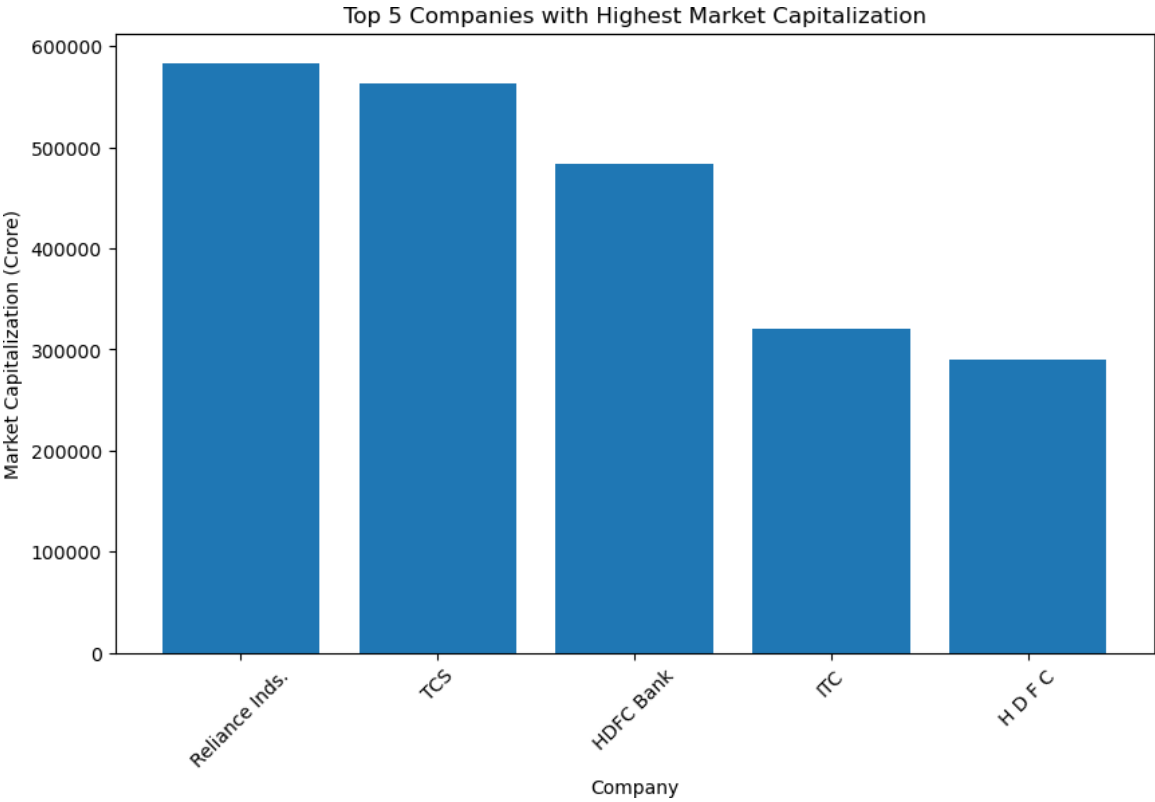
1  # Sort the dataset by market capitalization in descending order
2  sorted_data = df.sort_values('Mar Cap - Crore', ascending=False)
3
4  # Select the top 5 companies with the highest market capitalization
5  top_5_companies = sorted_data.head(5)
6
7
8  print(top_5_companies.value_counts())
9  # Plot the market capitalization of the top 5 companies
10 plt.figure(figsize=(10, 6))
11 plt.bar(top_5_companies['Name'], top_5_companies['Mar Cap - Crore'])
12 plt.title('Top 5 Companies with Highest Market Capitalization')
13 plt.xlabel('Company')
14 plt.ylabel('Market Capitalization (Crore)')
15 plt.xticks(rotation=45)
16 plt.show()

```

S.No.	Name	Mar Cap - Crore	Sales Qtr - Crore	Profit Margin	Mar ket_Share
1	Reliance Inds.	583436.72	99810.00	5.845474	6.2
20507	1				
2	TCS	563709.84	30904.00	18.240676	1.9
26045	1				
3	HDFC Bank	482953.59	20581.27	23.465685	1.2
82696	1				
4	ITC	320985.27	9772.02	32.847382	0.6
09026	1				
5	H D F C	289497.37	16840.51	17.190535	1.0
49559	1				

dtype: int64





**BOTTOM 5 COMPANIES WITH LOW MARKET CAPITALIZATION:**

In [20]:

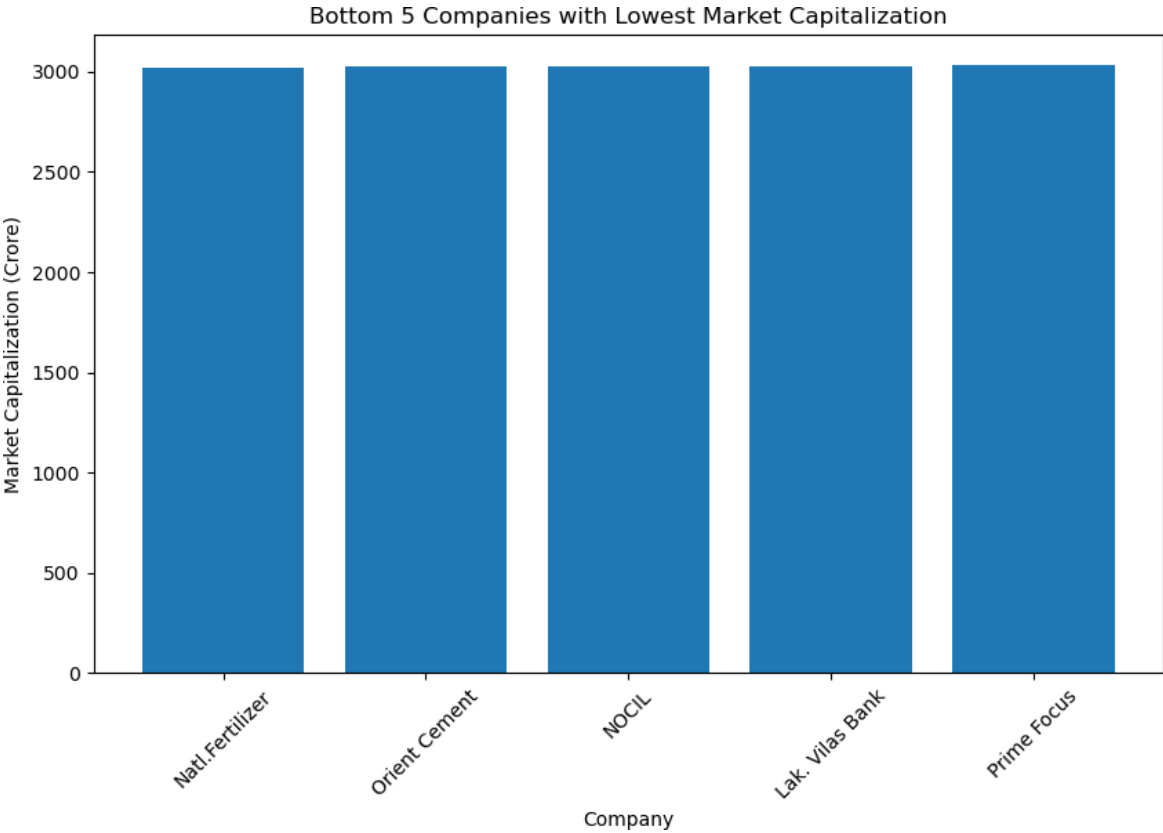
```

1  # Sort the dataset by market capitalization in ascending order
2  sorted_data = df.sort_values('Mar Cap - Crore')
3
4  # Select the bottom 5 companies with the lowest market capitalization
5  bottom_5_companies = sorted_data.head(5)
6
7
8  print(bottom_5_companies.value_counts())
9
10
11
12 # Plot the market capitalization of the bottom 5 companies
13 plt.figure(figsize=(10, 6))
14 plt.bar(bottom_5_companies['Name'], bottom_5_companies['Mar Cap - Crore'])
15 plt.title('Bottom 5 Companies with Lowest Market Capitalization')
16 plt.xlabel('Company')
17 plt.ylabel('Market Capitalization (Crore)')
18 plt.xticks(rotation=45)
19 plt.show()

```

S.No.	Name	Mar Cap - Crore	Sales Qtr - Crore	Profit Margin	Ma
495	Prime Focus	3031.50	609.61	4.972851	0.
037993	1				
496	Lak. Vilas Bank	3029.57	790.17	3.834074	0.
049246	1				
497	NOCIL	3026.26	249.27	12.140490	0.
015535	1				
498	Orient Cement	3024.32	511.53	5.912302	0.
031880	1				
499	Natl.Fertilizer	3017.07	2840.75	1.062068	0.
177045	1				

dtype: int64

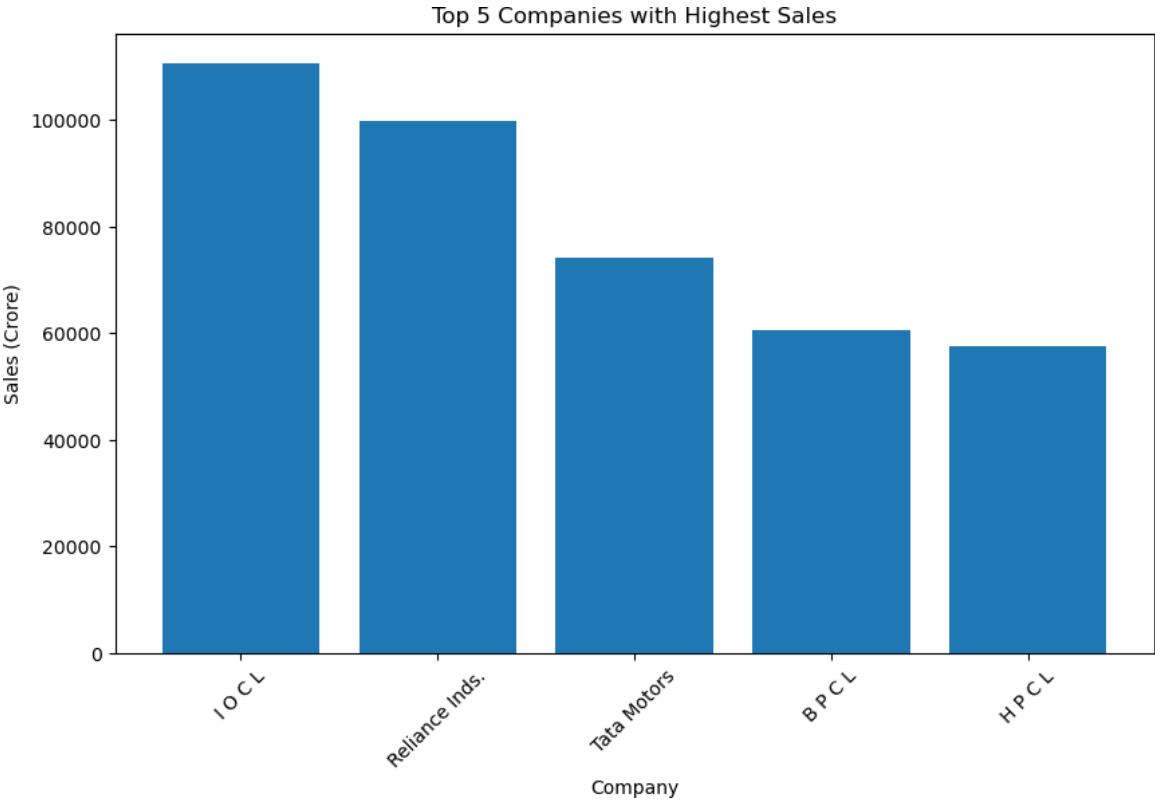


## TOP 5 COMPANIES WITH HIGHEST SALES:

```
In [21]: 1 # Sort the dataset by sales in descending order
2 sorted_data = df.sort_values('Sales Qtr - Crore', ascending=False)
3
4 # Select the top 5 companies with the highest sales
5 top_5_companies_sales = sorted_data.head(5)
6
7 print(top_5_companies_sales.value_counts())
8 # Plot the sales of the top 5 companies
9 plt.figure(figsize=(10, 6))
10 plt.bar(top_5_companies_sales['Name'], top_5_companies_sales['Sales Qtr -
11 plt.title('Top 5 Companies with Highest Sales')
12 plt.xlabel('Company')
13 plt.ylabel('Sales (Crore)')
14 plt.xticks(rotation=45)
15 plt.show()
```

S.No.	Name	Mar Cap - Crore	Sales Qtr - Crore	Profit Margin	Mar
1	Reliance Inds.	583436.72	99810.00	5.845474	6.2
20507	1				
15	I O C L	178017.48	110666.93	1.608588	6.8
97149	1				
24	Tata Motors	117071.87	74156.07	1.578723	4.6
21665	1				
28	B P C L	98278.00	60616.36	1.621311	3.7
77823	1				
55	H P C L	58034.78	57474.25	1.009753	3.5
81996	1				

dtype: int64

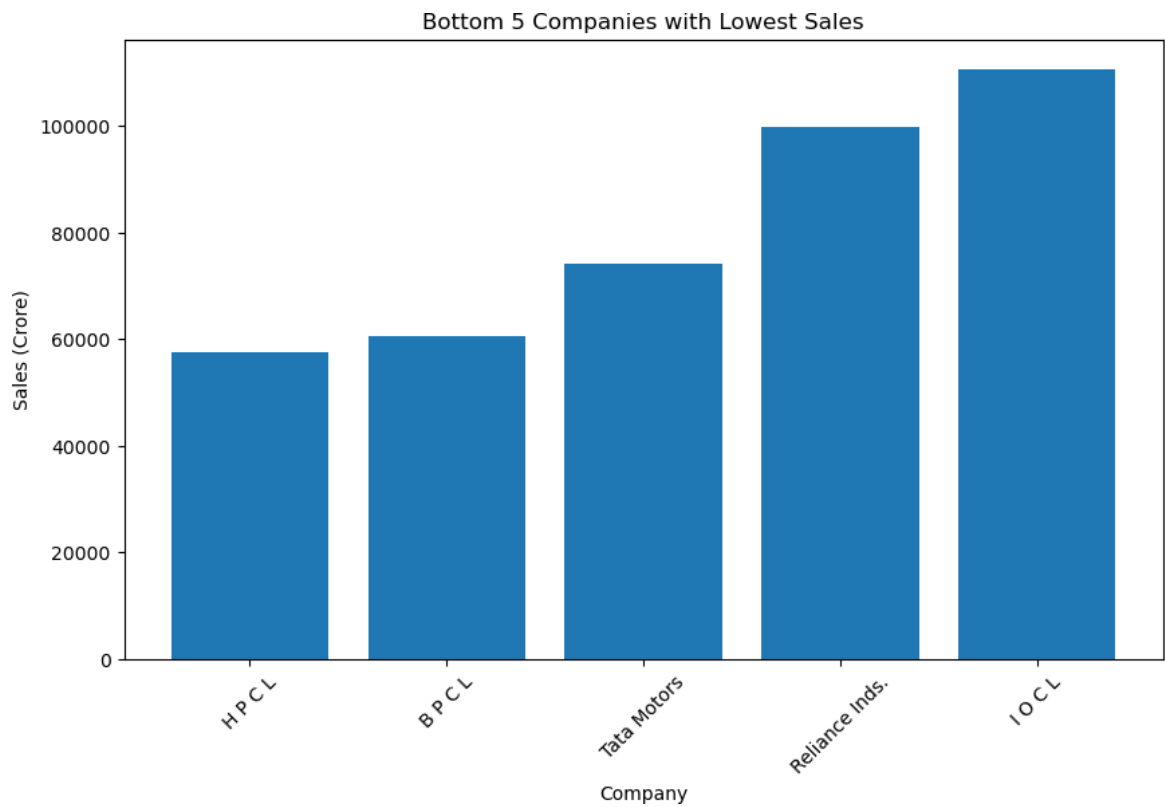


BOTTOM 5 COMPANIES WITH LOWEST SALES:

```
In [22]: 1 # Sort the dataset by sales in ascending order
2 sorted_data = df.sort_values('Sales Qtr - Crore')
3
4 # Select the bottom 5 companies with the highest sales
5 bottom_5_companies_sales = sorted_data.tail(5)
6 print(bottom_5_companies_sales.value_counts())
7 # Plot the sales of the bottom 5 companies
8 plt.figure(figsize=(10, 6))
9 plt.bar(bottom_5_companies_sales['Name'], bottom_5_companies_sales['Sales
10 plt.title('Bottom 5 Companies with Lowest Sales')
11 plt.xlabel('Company')
12 plt.ylabel('Sales (Crore)')
13 plt.xticks(rotation=45)
14 plt.show()
```

S.No.	Name	Mar Cap - Crore	Sales Qtr - Crore	Profit Margin	Mar ket_Share
1	Reliance Inds.	583436.72	99810.00	5.845474	6.2
20507	1				
15	I O C L	178017.48	110666.93	1.608588	6.8
97149	1				
24	Tata Motors	117071.87	74156.07	1.578723	4.6
21665	1				
28	B P C L	98278.00	60616.36	1.621311	3.7
77823	1				
55	H P C L	58034.78	57474.25	1.009753	3.5
81996	1				

dtype: int64

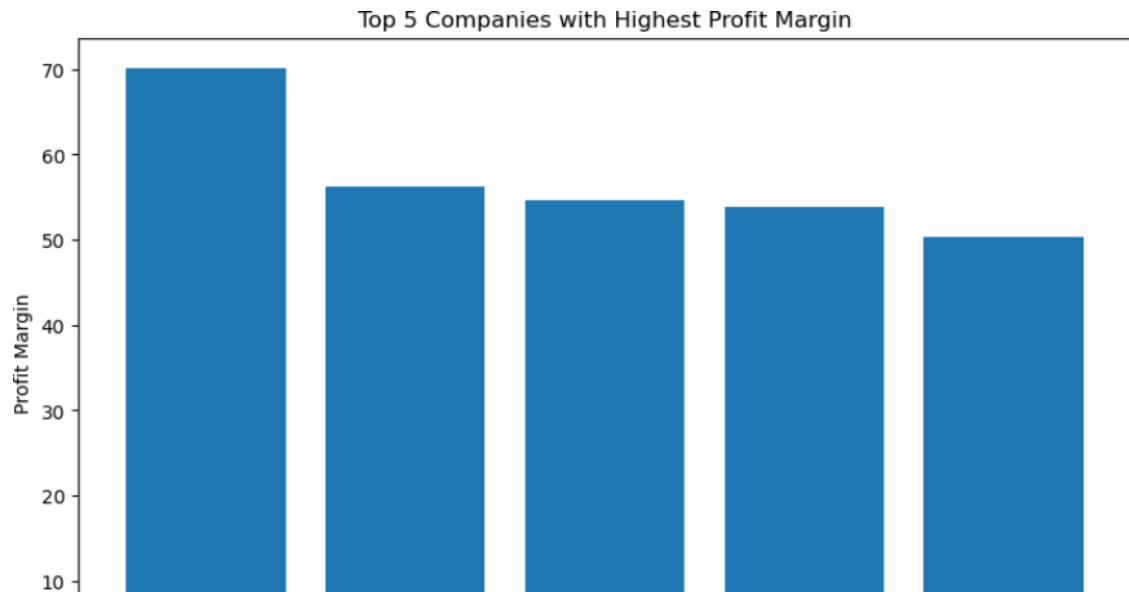


## TOP 5 COMPANIES WITH HIGHEST PROFIT MARGIN:

In [23]:

```
1 # Sort the dataset by profit margin in descending order
2 sorted_data = df.sort_values('Profit Margin', ascending=False)
3
4 # Select the top 5 companies with the highest profit margin
5 top_5_companies_pm = sorted_data.head(5)
6
7
8 print(top_5_companies_pm.value_counts())
9
10 # Plot the profit margin of the top 5 companies
11 plt.figure(figsize=(10, 6))
12 plt.bar(top_5_companies_pm['Name'], top_5_companies_pm['Profit Margin'])
13 plt.title('Top 5 Companies with Highest Profit Margin')
14 plt.xlabel('Company')
15 plt.ylabel('Profit Margin')
16 plt.xticks(rotation=45)
17 plt.show()
18
```

dtype: int64



**BOTTOM 5 COMPANIES WITH LOWEST PROFIT MARGIN:**

In [24]:

```

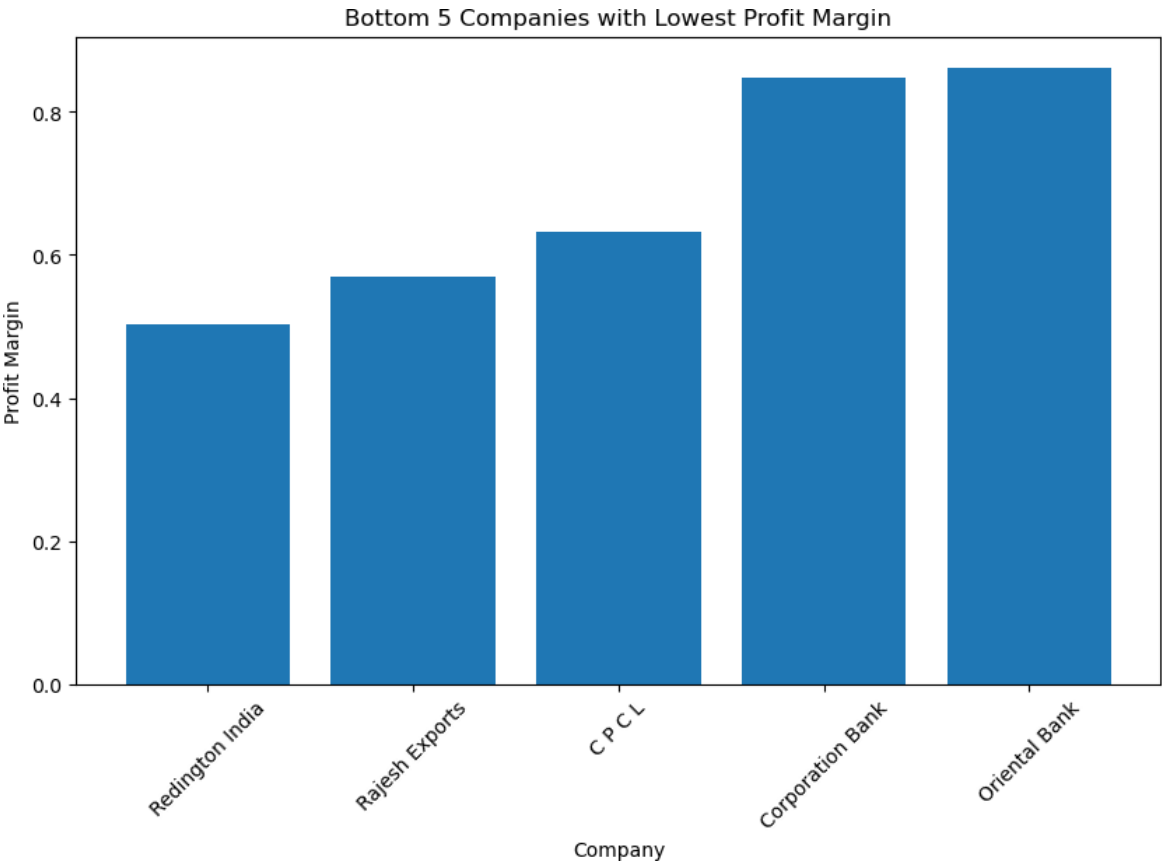
1  # Sort the dataset by profit margin in ascending order
2  sorted_data = df.sort_values('Profit Margin')
3
4  # Select the bottom 5 companies with the lowest profit margin
5  bottom_5_companies_pm = sorted_data.head(5)
6
7
8  print(bottom_5_companies_pm.value_counts())
9
10 # Plot the profit margin of the bottom 5 companies
11
12 plt.figure(figsize=(10, 6))
13 plt.bar(bottom_5_companies_pm['Name'], bottom_5_companies_pm['Profit Margin'])
14 plt.title('Bottom 5 Companies with Lowest Profit Margin')
15 plt.xlabel('Company')
16 plt.ylabel('Profit Margin')
17 plt.xticks(rotation=45)
18 plt.show()

```

S.No.	Name	Mar Cap - Crore	Sales Qtr - Crore	Profit Margin	M
123	Rajesh Exports	23495.54	41304.84	0.568833	
2.574262	1				
333	Redington India	5896.54	11728.40	0.502757	
0.730955	1				
347	C P C L	5427.82	8587.17	0.632085	
0.535182	1				
454	Corporation Bank	3716.46	4387.85	0.846989	
0.273466	1				
457	Oriental Bank	3674.60	4262.08	0.862161	
0.265628	1				

dtype: int64





## TOP 5 COMPANIES WITH HIGHEST MARKET SHARE:

In [25]:

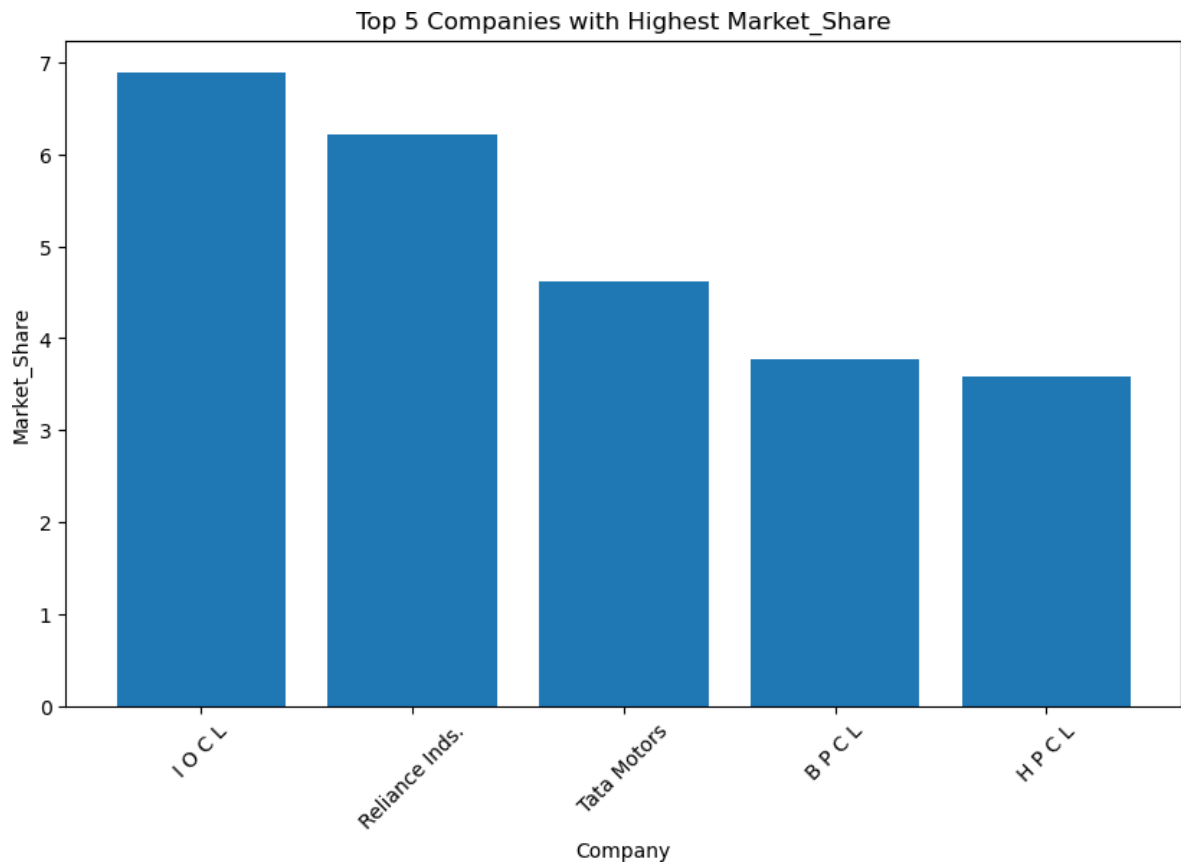
```

1  # Sort the dataset by profit margin in descending order
2  sorted_data_MS = df.sort_values('Market_Share', ascending=False)
3
4  # Select the top 5 companies with the highest profit margin
5  top_5_companies_MS = sorted_data_MS.head(5)
6
7
8  print(top_5_companies_MS.value_counts())
9
10 # Plot the profit margin of the top 5 companies
11 plt.figure(figsize=(10, 6))
12 plt.bar(top_5_companies_MS['Name'], top_5_companies_MS['Market_Share'])
13 plt.title('Top 5 Companies with Highest Market_Share')
14 plt.xlabel('Company')
15 plt.ylabel('Market_Share')
16 plt.xticks(rotation=45)
17 plt.show()
18

```

S.No.	Name	Mar Cap - Crore	Sales Qtr - Crore	Profit Margin	Mar
1	Reliance Inds.	583436.72	99810.00	5.845474	6.2
20507	1				
15	I O C L	178017.48	110666.93	1.608588	6.8
97149	1				
24	Tata Motors	117071.87	74156.07	1.578723	4.6
21665	1				
28	B P C L	98278.00	60616.36	1.621311	3.7
77823	1				
55	H P C L	58034.78	57474.25	1.009753	3.5
81996	1				

dtype: int64



## CONCLUSION

The correlation coefficient of 0.62 indicates a moderate positive relationship between sales and market capitalization. This suggests that as sales increase, market capitalization tends to rise as well, reflecting a general tendency for companies with higher sales to achieve higher market value.

While it is true that increasing sales can positively influence a company's market capitalization, it is important to recognize that market capitalization is also affected by a variety of other factors. These include:

1. **Industry Trends**: The overall health and dynamics of the industry in which the company operates can significantly impact its market capitalization.
2. **Company Profitability**: A company's profitability plays a crucial role; high sales alone may not lead to an increase in market capitalization if the company is not generating profits.
3. **Competitive Landscape**: The presence and strength of competitors can influence market capitalization, even if a company's sales figures are robust.
4. **Management Quality**: Effective leadership and management strategies are essential for sustaining growth and maximizing market value.
5. **Brand Value**: A strong and well-regarded brand can enhance market capitalization, often translating into increased customer loyalty and market value beyond mere sales numbers.
6. **Investor Sentiment**: Market capitalization can be significantly influenced by how investors perceive the company and broader market conditions.
7. **Economic Factors**: Macro-economic elements, such as interest rates and inflation, also play a role in shaping market capitalization.

**In conclusion, while focusing on sales growth is important, it should be part of a comprehensive strategy that considers these additional factors to fully understand and improve a company's market capitalization.**