

Shark_Tank_India_Analysis

May 16, 2024

```
[1]: #importing required libraries
import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
import seaborn as sns
```

```
[2]: #Getting the dataset
data = pd.read_csv('/content/drive/MyDrive/Computers/Python/Datasets for data_
↳analysis/Shark Tank India/Shark Tank India.csv')
```

```
[3]: #Setting max rows and columns
pd.options.display.max_rows = 500
pd.options.display.max_columns = 100
```

```
[4]: #Understanding the data
print(data.head())
```

	Season Number	Startup Name	Episode Number	Pitch Number	Season Start	\
0	1	BluePineFoods	1	1	20-Dec-21	
1	1	BoozScooters	1	2	20-Dec-21	
2	1	HeartUpMySleeves	1	3	20-Dec-21	
3	1	TagzFoods	2	4	20-Dec-21	
4	1	HeadAndHeart	2	5	20-Dec-21	

	Season End	Original Air Date	Episode Title	Anchor	\
0	4-Feb-22	20-Dec-21	Badlegi Business Ki Tasveer	Rannvijay Singh	
1	4-Feb-22	20-Dec-21	Badlegi Business Ki Tasveer	Rannvijay Singh	
2	4-Feb-22	20-Dec-21	Badlegi Business Ki Tasveer	Rannvijay Singh	
3	4-Feb-22	21-Dec-21	Insaan, Ideas Aur Sapne	Rannvijay Singh	
4	4-Feb-22	21-Dec-21	Insaan, Ideas Aur Sapne	Rannvijay Singh	

	Industry	\
0	Food	
1	Vehicles/Electrical Vehicles	
2	Beauty/Fashion	
3	Food	
4	Education	

	Business Description \
0	Frozen Momos
1	Renting e-bike for mobility in private spaces
2	Detachable Sleeves
3	Healthy Potato Chips Snacks
4	Brain Development Course

	Company Website	Started in	Number of Presenters \
0	https://bluepinefoods.com/	2016.0	3
1	https://www.boozup.net/	2017.0	1
2	https://heartupmysleeves.com/	2021.0	1
3	https://tagzfoods.com/	2019.0	2
4	https://thehnh.in/	2015.0	4

	Male Presenters	Female Presenters	Transgender Presenters \
0	2.0	1.0	NaN
1	1.0	NaN	NaN
2	NaN	1.0	NaN
3	2.0	NaN	NaN
4	1.0	3.0	NaN

	Couple Presenters	Pitchers	Average Age	Pitchers City	Pitchers State \
0	0.0		Middle	Delhi	Delhi
1	0.0		Young	Ahmedabad	Gujarat
2	0.0		Young	Delhi	Delhi
3	0.0		Middle	Bangalore	Karnataka
4	1.0		Middle	Patiala	Punjab

	Yearly Revenue	Monthly Sales	Gross Margin	Net Margin	EBITDA	Cash Burn \
0	95.0	8.0	NaN	NaN	NaN	NaN
1	4.0	0.4	NaN	NaN	NaN	NaN
2	NaN	2.0	NaN	NaN	NaN	NaN
3	700.0	NaN	48.0	NaN	NaN	NaN
4	30.0	NaN	NaN	NaN	NaN	NaN

	SKUs	Has Patents	Bootstrapped	Original Ask Amount \
0	NaN	NaN	NaN	50.0
1	NaN	NaN	NaN	40.0
2	NaN	NaN	NaN	25.0
3	NaN	NaN	NaN	70.0
4	NaN	NaN	NaN	50.0

	Original Offered Equity	Valuation Requested	Received Offer \
0	5.0	1000.0	1
1	15.0	267.0	1
2	10.0	250.0	1
3	1.0	7000.0	1
4	5.0	1000.0	0

	Accepted Offer	Total Deal Amount	Total Deal Equity	Total Deal Debt	\
0	1.0	75.0	16.00	NaN	
1	1.0	40.0	50.00	NaN	
2	1.0	25.0	30.00	NaN	
3	1.0	70.0	2.75	NaN	
4	NaN	NaN	NaN	NaN	

	Debt Interest	Deal Valuation	Number of Sharks in Deal	\
0	NaN	469.0	3.0	
1	NaN	80.0	2.0	
2	NaN	83.0	2.0	
3	NaN	2545.0	1.0	
4	NaN	NaN	NaN	

	Deal Has Conditions	Royalty Deal	Advisory Shares Equity	\
0	NaN	NaN	NaN	
1	NaN	NaN	NaN	
2	NaN	NaN	NaN	
3	NaN	NaN	NaN	
4	NaN	NaN	NaN	

	Namita Investment Amount	Namita Investment Equity	Namita Debt Amount	\
0	NaN	NaN	NaN	
1	NaN	NaN	NaN	
2	NaN	NaN	NaN	
3	NaN	NaN	NaN	
4	NaN	NaN	NaN	

	Vineeta Investment Amount	Vineeta Investment Equity	Vineeta Debt Amount	\
0	25.0	5.33	NaN	
1	20.0	25.00	NaN	
2	12.5	15.00	NaN	
3	NaN	NaN	NaN	
4	NaN	NaN	NaN	

	Anupam Investment Amount	Anupam Investment Equity	Anupam Debt Amount	\
0	NaN	NaN	NaN	
1	NaN	NaN	NaN	
2	12.5	15.0	NaN	
3	NaN	NaN	NaN	
4	NaN	NaN	NaN	

	Aman Investment Amount	Aman Investment Equity	Aman Debt Amount	\
0	25.0	5.33	NaN	
1	NaN	NaN	NaN	
2	NaN	NaN	NaN	
3	NaN	NaN	NaN	

4	NaN		NaN		NaN	
	Peyush Investment Amount	Peyush Investment Equity	Peyush Debt Amount	\		
0	NaN	NaN	NaN			
1	NaN	NaN	NaN			
2	NaN	NaN	NaN			
3	NaN	NaN	NaN			
4	NaN	NaN	NaN			
	Amit Investment Amount	Amit Investment Equity	Amit Debt Amount	\		
0	NaN	NaN	NaN			
1	NaN	NaN	NaN			
2	NaN	NaN	NaN			
3	NaN	NaN	NaN			
4	NaN	NaN	NaN			
	Ashneer Investment Amount	Ashneer Investment Equity	Ashneer Debt Amount	\		
0	25.0	5.33	NaN			
1	20.0	25.00	NaN			
2	NaN	NaN	NaN			
3	70.0	2.75	NaN			
4	NaN	NaN	NaN			
	Guest Investment Amount	Guest Investment Equity	Guest Debt Amount	\		
0	NaN	NaN	NaN			
1	NaN	NaN	NaN			
2	NaN	NaN	NaN			
3	NaN	NaN	NaN			
4	NaN	NaN	NaN			
	Invested Guest Name	All Guest Names	Namita Present	Vineeta Present	\	
0	NaN	NaN	1.0	1.0		
1	NaN	NaN	1.0	1.0		
2	NaN	NaN	1.0	1.0		
3	NaN	NaN	1.0	1.0		
4	NaN	NaN	1.0	1.0		
	Anupam Present	Aman Present	Peyush Present	Amit Present	\	
0	1.0	1.0	NaN	NaN		
1	1.0	1.0	NaN	NaN		
2	1.0	1.0	NaN	NaN		
3	1.0	1.0	NaN	NaN		
4	1.0	1.0	NaN	NaN		
	Ashneer Present	Guest Present				
0	1.0	NaN				
1	1.0	NaN				
2	1.0	NaN				

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3          1.0          NaN
4          1.0          NaN
```

```
[5]: #Getting the statistical info about the data
data.describe()
```

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[5]:
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	Season Number	Episode Number	Pitch Number	Started in	\
count	478.000000	478.000000	478.000000	355.000000	
mean	2.010460	23.182008	239.500000	2018.904225	
std	0.804791	15.028357	138.130976	2.770274	
min	1.000000	0.000000	1.000000	1998.000000	
25%	1.000000	10.000000	120.250000	2018.000000	
50%	2.000000	23.000000	239.500000	2019.000000	
75%	3.000000	35.000000	358.750000	2021.000000	
max	3.000000	52.000000	478.000000	2023.000000	

	Number of Presenters	Male Presenters	Female Presenters	\
count	478.000000	412.000000	226.000000	
mean	2.020921	1.679612	1.199115	
std	0.821349	0.819039	0.421847	
min	1.000000	1.000000	1.000000	
25%	1.000000	1.000000	1.000000	
50%	2.000000	1.000000	1.000000	
75%	2.000000	2.000000	1.000000	
max	6.000000	6.000000	3.000000	

	Transgender Presenters	Couple Presenters	Yearly Revenue	\
count	3.0	473.000000	241.000000	
mean	1.0	0.177590	557.033195	
std	0.0	0.382572	1501.919516	
min	1.0	0.000000	0.000000	
25%	1.0	0.000000	72.000000	
50%	1.0	0.000000	170.000000	
75%	1.0	0.000000	460.000000	
max	1.0	1.000000	18700.000000	

	Monthly Sales	Gross Margin	Net Margin	EBITDA	SKUs	\
count	225.000000	129.000000	73.000000	23.000000	28.000000	
mean	68.917253	54.953488	20.575342	10.608696	330.678571	
std	250.631067	21.059017	12.187104	12.822172	1137.602256	
min	0.000000	3.000000	1.000000	-20.000000	1.000000	
25%	6.000000	40.000000	10.000000	5.000000	9.000000	
50%	21.000000	56.000000	20.000000	10.000000	22.500000	
75%	60.000000	70.000000	30.000000	18.500000	95.000000	
max	3500.000000	150.000000	55.000000	35.000000	6000.000000	

	Original Ask Amount	Original Offered Equity	Valuation Requested	\
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count	478.000000	478.000000	478.000000
mean	144.451284	3.811987	5245.346384
std	1370.241566	3.797043	8892.492491
min	0.000000	0.200000	0.000000
25%	50.000000	1.212500	1000.000000
50%	75.000000	2.500000	2857.071428
75%	100.000000	5.000000	6000.000000
max	30000.000000	30.000000	120000.000000

	Received Offer	Accepted Offer	Total Deal Amount	Total Deal Equity \
count	478.000000	321.000000	268.000000	268.000000
mean	0.671548	0.834891	66.177663	8.278694
std	0.470142	0.371859	43.314939	9.366814
min	0.000000	0.000000	0.000000	0.500000
25%	0.000000	1.000000	40.000000	2.500000
50%	1.000000	1.000000	56.300000	5.000000
75%	1.000000	1.000000	90.000000	10.000000
max	1.000000	1.000000	300.000000	75.000000

	Total Deal Debt	Debt Interest	Deal Valuation \
count	62.000000	45.000000	267.000000
mean	46.225806	10.511111	2280.644650
std	26.245636	3.314035	3435.328410
min	20.000000	0.000000	0.000000
25%	25.000000	10.000000	500.000000
50%	40.000000	10.000000	1107.692308
75%	50.000000	12.000000	2500.000000
max	150.000000	18.000000	25000.000000

	Number of Sharks in Deal	Royalty Deal	Advisory Shares Equity \
count	268.000000	17.0	3.000000
mean	2.011194	1.0	1.533333
std	1.149770	0.0	0.950438
min	1.000000	1.0	0.600000
25%	1.000000	1.0	1.050000
50%	2.000000	1.0	1.500000
75%	3.000000	1.0	2.000000
max	5.000000	1.0	2.500000

	Namita Investment Amount	Namita Investment Equity	Namita Debt Amount \
count	89.000000	89.000000	17.000000
mean	32.942025	3.934479	36.332941
std	20.584949	5.116032	22.085984
min	0.000016	0.200000	10.000000
25%	20.000000	1.000000	20.000000
50%	26.660000	2.000000	35.000000
75%	45.000000	5.000000	50.000000

max	100.000000	25.000000	100.000000
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	Vineeta Investment Amount	Vineeta Investment Equity \
count	68.000000	68.000000
mean	31.246507	4.162559
std	20.795928	4.706656
min	0.002500	0.200000
25%	19.415000	1.000000
50%	25.000000	2.500000
75%	40.000000	5.000000
max	100.000000	25.000000

	Vineeta Debt Amount	Anupam Investment Amount \
count	12.000000	75.000000
mean	24.513333	29.593104
std	13.434905	21.221572
min	12.500000	0.000000
25%	14.375000	18.750000
50%	20.000000	25.000000
75%	30.000000	40.000000
max	50.000000	100.000000

	Anupam Investment Equity	Anupam Debt Amount	Aman Investment Amount \
count	75.000000	8.000000	109.000000
mean	4.612107	26.87500	33.868830
std	5.273910	14.92541	24.134884
min	0.166000	12.50000	0.000000
25%	1.000000	16.87500	17.660000
50%	2.250000	22.50000	30.000000
75%	5.500000	31.25000	45.000000
max	25.000000	50.00000	150.000000

	Aman Investment Equity	Aman Debt Amount	Peyush Investment Amount \
count	109.000000	18.000000	89.000000
mean	3.069473	37.656667	34.667222
std	4.384031	21.579428	30.122765
min	0.166000	10.000000	0.000000
25%	1.000000	21.250000	20.000000
50%	2.000000	34.750000	28.000000
75%	4.000000	47.915000	45.000000
max	40.000000	80.000000	250.000000

	Peyush Investment Equity	Peyush Debt Amount	Amit Investment Amount \
count	89.000000	12.000000	35.000000
mean	5.667764	30.583333	35.268571
std	10.834254	14.724490	26.540778
min	0.166000	10.000000	3.500000

25%	1.000000	24.250000	15.830000
50%	2.000000	25.000000	25.000000
75%	5.000000	35.000000	50.000000
max	75.000000	60.000000	100.000000

	Amit Investment Equity	Amit Debt Amount	Ashneer Investment Amount \
count	35.000000	7.000000	21.000000
mean	4.287046	35.000000	25.682381
std	4.713601	18.257419	16.860620
min	0.330000	10.000000	1.000000
25%	1.375000	22.500000	15.000000
50%	2.500000	40.000000	20.000000
75%	5.000000	42.500000	30.000000
max	20.000000	65.000000	70.000000

	Ashneer Investment Equity	Ashneer Debt Amount \
count	21.000000	2.000000
mean	4.440000	57.000000
std	5.065662	59.39697
min	1.000000	15.000000
25%	2.000000	36.000000
50%	3.000000	57.000000
75%	5.000000	78.000000
max	25.000000	99.000000

	Guest Investment Amount	Guest Investment Equity	Guest Debt Amount \
count	48.000000	48.000000	9.000000
mean	39.760057	3.273438	37.257778
std	34.690571	3.551611	25.051059
min	0.000253	0.200000	12.500000
25%	23.750000	1.000000	25.000000
50%	31.665000	2.165000	30.000000
75%	46.250000	4.000000	35.000000
max	200.000000	17.500000	83.320000

	Namita Present	Vineeta Present	Anupam Present	Aman Present \
count	389.0	318.0	424.0	420.0
mean	1.0	1.0	1.0	1.0
std	0.0	0.0	0.0	0.0
min	1.0	1.0	1.0	1.0
25%	1.0	1.0	1.0	1.0
50%	1.0	1.0	1.0	1.0
75%	1.0	1.0	1.0	1.0
max	1.0	1.0	1.0	1.0

	Peyush Present	Amit Present	Ashneer Present	Guest Present
count	307.0	137.0	99.0	144.000000

mean	1.0	1.0	1.0	1.166667
std	0.0	0.0	0.0	0.373979
min	1.0	1.0	1.0	1.000000
25%	1.0	1.0	1.0	1.000000
50%	1.0	1.0	1.0	1.000000
75%	1.0	1.0	1.0	1.000000
max	1.0	1.0	1.0	2.000000

```
[6]: data.info()
#We can see that all the columns have a not null constraint. But still to be
    ↳ sure let's check if there any null values

'''Observations from the data:
    1) Season start, original air date and end are data columns but are in
    ↳ object dtype
    2) Started in is in float but it need to be an int
    3) Male Presenters, Couple Presenters, Trans Presenters and Female
    ↳ Presenters are also in float and need to be changed to int
    4) Number of sharks in the deal is also in float and need to be changed to
    ↳ int '''
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 478 entries, 0 to 477
Data columns (total 78 columns):
#   Column                                Non-Null Count  Dtype
---  -
0   Season Number                        478 non-null    int64
1   Startup Name                        478 non-null    object
2   Episode Number                      478 non-null    int64
3   Pitch Number                       478 non-null    int64
4   Season Start                        478 non-null    object
5   Season End                          478 non-null    object
6   Original Air Date                   447 non-null    object
7   Episode Title                      478 non-null    object
8   Anchor                             478 non-null    object
9   Industry                           478 non-null    object
10  Business Description                478 non-null    object
11  Company Website                    466 non-null    object
12  Started in                         355 non-null    float64
13  Number of Presenters                478 non-null    int64
14  Male Presenters                     412 non-null    float64
15  Female Presenters                   226 non-null    float64
16  Transgender Presenters               3 non-null      float64
17  Couple Presenters                   473 non-null    float64
18  Pitches Average Age                 478 non-null    object
19  Pitches City                       473 non-null    object
20  Pitches State                      474 non-null    object
```

21	Yearly Revenue	241 non-null	float64
22	Monthly Sales	225 non-null	float64
23	Gross Margin	129 non-null	float64
24	Net Margin	73 non-null	float64
25	EBITDA	23 non-null	float64
26	Cash Burn	64 non-null	object
27	SKUs	28 non-null	float64
28	Has Patents	43 non-null	object
29	Bootstrapped	48 non-null	object
30	Original Ask Amount	478 non-null	float64
31	Original Offered Equity	478 non-null	float64
32	Valuation Requested	478 non-null	float64
33	Received Offer	478 non-null	int64
34	Accepted Offer	321 non-null	float64
35	Total Deal Amount	268 non-null	float64
36	Total Deal Equity	268 non-null	float64
37	Total Deal Debt	62 non-null	float64
38	Debt Interest	45 non-null	float64
39	Deal Valuation	267 non-null	float64
40	Number of Sharks in Deal	268 non-null	float64
41	Deal Has Conditions	26 non-null	object
42	Royalty Deal	17 non-null	float64
43	Advisory Shares Equity	3 non-null	float64
44	Namita Investment Amount	89 non-null	float64
45	Namita Investment Equity	89 non-null	float64
46	Namita Debt Amount	17 non-null	float64
47	Vineeta Investment Amount	68 non-null	float64
48	Vineeta Investment Equity	68 non-null	float64
49	Vineeta Debt Amount	12 non-null	float64
50	Anupam Investment Amount	75 non-null	float64
51	Anupam Investment Equity	75 non-null	float64
52	Anupam Debt Amount	8 non-null	float64
53	Aman Investment Amount	109 non-null	float64
54	Aman Investment Equity	109 non-null	float64
55	Aman Debt Amount	18 non-null	float64
56	Peyush Investment Amount	89 non-null	float64
57	Peyush Investment Equity	89 non-null	float64
58	Peyush Debt Amount	12 non-null	float64
59	Amit Investment Amount	35 non-null	float64
60	Amit Investment Equity	35 non-null	float64
61	Amit Debt Amount	7 non-null	float64
62	Ashneer Investment Amount	21 non-null	float64
63	Ashneer Investment Equity	21 non-null	float64
64	Ashneer Debt Amount	2 non-null	float64
65	Guest Investment Amount	48 non-null	float64
66	Guest Investment Equity	48 non-null	float64
67	Guest Debt Amount	9 non-null	float64
68	Invested Guest Name	48 non-null	object

69	All Guest Names	144 non-null	object
70	Namita Present	389 non-null	float64
71	Vineeta Present	318 non-null	float64
72	Anupam Present	424 non-null	float64
73	Aman Present	420 non-null	float64
74	Peyush Present	307 non-null	float64
75	Amit Present	137 non-null	float64
76	Ashneer Present	99 non-null	float64
77	Guest Present	144 non-null	float64

dtypes: float64(55), int64(5), object(18)

memory usage: 291.4+ KB

[6]: 'Observations from the data:\n 1) Season start,original air date and end are data columns but are in object dtype\n 2) Started in is in float but it need to be an int\n 3) Male Presenters, Couple Presenters, Trans Presenters and Female Presenters are also in float and need to be changed to int\n 4) Number of sharks in the deal is also in float and need to be changed to int '

[7]: *#getting the number of nulls in each row*
`print(data.isnull().sum())`

Season Number	0
Startup Name	0
Episode Number	0
Pitch Number	0
Season Start	0
Season End	0
Original Air Date	31
Episode Title	0
Anchor	0
Industry	0
Business Description	0
Company Website	12
Started in	123
Number of Presenters	0
Male Presenters	66
Female Presenters	252
Transgender Presenters	475
Couple Presenters	5
Pitchers Average Age	0
Pitchers City	5
Pitchers State	4
Yearly Revenue	237
Monthly Sales	253
Gross Margin	349
Net Margin	405
EBITDA	455
Cash Burn	414

SKUs	450
Has Patents	435
Bootstrapped	430
Original Ask Amount	0
Original Offered Equity	0
Valuation Requested	0
Received Offer	0
Accepted Offer	157
Total Deal Amount	210
Total Deal Equity	210
Total Deal Debt	416
Debt Interest	433
Deal Valuation	211
Number of Sharks in Deal	210
Deal Has Conditions	452
Royalty Deal	461
Advisory Shares Equity	475
Namita Investment Amount	389
Namita Investment Equity	389
Namita Debt Amount	461
Vineeta Investment Amount	410
Vineeta Investment Equity	410
Vineeta Debt Amount	466
Anupam Investment Amount	403
Anupam Investment Equity	403
Anupam Debt Amount	470
Aman Investment Amount	369
Aman Investment Equity	369
Aman Debt Amount	460
Peyush Investment Amount	389
Peyush Investment Equity	389
Peyush Debt Amount	466
Amit Investment Amount	443
Amit Investment Equity	443
Amit Debt Amount	471
Ashneer Investment Amount	457
Ashneer Investment Equity	457
Ashneer Debt Amount	476
Guest Investment Amount	430
Guest Investment Equity	430
Guest Debt Amount	469
Invested Guest Name	430
All Guest Names	334
Namita Present	89
Vineeta Present	160
Anupam Present	54
Aman Present	58
Peyush Present	171

Amit Present	341
Ashneer Present	379
Guest Present	334
dtype:	int64

```
[8]: #Getting the missing values percentage in each column
print(data.isnull().sum()/len(data)*100)
```

Season Number	0.000000
Startup Name	0.000000
Episode Number	0.000000
Pitch Number	0.000000
Season Start	0.000000
Season End	0.000000
Original Air Date	6.485356
Episode Title	0.000000
Anchor	0.000000
Industry	0.000000
Business Description	0.000000
Company Website	2.510460
Started in	25.732218
Number of Presenters	0.000000
Male Presenters	13.807531
Female Presenters	52.719665
Transgender Presenters	99.372385
Couple Presenters	1.046025
Pitchers Average Age	0.000000
Pitchers City	1.046025
Pitchers State	0.836820
Yearly Revenue	49.581590
Monthly Sales	52.928870
Gross Margin	73.012552
Net Margin	84.728033
EBITDA	95.188285
Cash Burn	86.610879
SKUs	94.142259
Has Patents	91.004184
Bootstrapped	89.958159
Original Ask Amount	0.000000
Original Offered Equity	0.000000
Valuation Requested	0.000000
Received Offer	0.000000
Accepted Offer	32.845188
Total Deal Amount	43.933054
Total Deal Equity	43.933054
Total Deal Debt	87.029289
Debt Interest	90.585774
Deal Valuation	44.142259

Number of Sharks in Deal	43.933054
Deal Has Conditions	94.560669
Royalty Deal	96.443515
Advisory Shares Equity	99.372385
Namita Investment Amount	81.380753
Namita Investment Equity	81.380753
Namita Debt Amount	96.443515
Vineeta Investment Amount	85.774059
Vineeta Investment Equity	85.774059
Vineeta Debt Amount	97.489540
Anupam Investment Amount	84.309623
Anupam Investment Equity	84.309623
Anupam Debt Amount	98.326360
Aman Investment Amount	77.196653
Aman Investment Equity	77.196653
Aman Debt Amount	96.234310
Peyush Investment Amount	81.380753
Peyush Investment Equity	81.380753
Peyush Debt Amount	97.489540
Amit Investment Amount	92.677824
Amit Investment Equity	92.677824
Amit Debt Amount	98.535565
Ashneer Investment Amount	95.606695
Ashneer Investment Equity	95.606695
Ashneer Debt Amount	99.581590
Guest Investment Amount	89.958159
Guest Investment Equity	89.958159
Guest Debt Amount	98.117155
Invested Guest Name	89.958159
All Guest Names	69.874477
Namita Present	18.619247
Vineeta Present	33.472803
Anupam Present	11.297071
Aman Present	12.133891
Peyush Present	35.774059
Amit Present	71.338912
Ashneer Present	79.288703
Guest Present	69.874477

dtype: float64

```
[9]: #Checking the percentage of null values in the columns of type object
obj_cols = data.select_dtypes(include = 'object')
null_val = ((obj_cols.isnull().sum())/len(obj_cols))*100
filtered_null = null_val[null_val > 0] #null values that are greater than 0 and
↳column is object type
print(filtered_null)
```

Original Air Date	6.485356
-------------------	----------

Company Website	2.510460
Pitchers City	1.046025
Pitchers State	0.836820
Cash Burn	86.610879
Has Patents	91.004184
Bootstrapped	89.958159
Deal Has Conditions	94.560669
Invested Guest Name	89.958159
All Guest Names	69.874477

dtype: float64

```
[10]: #Dropping unwanted columns from the above
data.drop(columns = ['Company Website'], inplace = True)
```

```
[11]: #Checking the number of unique values in each column
data.nunique()
```

```
[11]: Season Number          3
Startup Name                478
Episode Number              53
Pitch Number                478
Season Start                3
Season End                  3
Original Air Date           137
Episode Title               138
Anchor                      3
Industry                    17
Business Description         475
Started in                  16
Number of Presenters         6
Male Presenters              6
Female Presenters            3
Transgender Presenters       1
Couple Presenters            2
Pitchers Average Age         3
Pitchers City                113
Pitchers State               47
Yearly Revenue              129
Monthly Sales                104
Gross Margin                 43
Net Margin                   32
EBITDA                       18
Cash Burn                    1
SKUs                         24
Has Patents                  1
Bootstrapped                 2
Original Ask Amount          48
```

Original Offered Equity	32
Valuation Requested	100
Received Offer	2
Accepted Offer	2
Total Deal Amount	39
Total Deal Equity	50
Total Deal Debt	19
Debt Interest	9
Deal Valuation	97
Number of Sharks in Deal	5
Deal Has Conditions	1
Royalty Deal	1
Advisory Shares Equity	3
Namita Investment Amount	31
Namita Investment Equity	35
Namita Debt Amount	11
Vineeta Investment Amount	24
Vineeta Investment Equity	32
Vineeta Debt Amount	7
Anupam Investment Amount	29
Anupam Investment Equity	40
Anupam Debt Amount	6
Aman Investment Amount	35
Aman Investment Equity	44
Aman Debt Amount	14
Peyush Investment Amount	29
Peyush Investment Equity	35
Peyush Debt Amount	7
Amit Investment Amount	19
Amit Investment Equity	18
Amit Debt Amount	6
Ashneer Investment Amount	9
Ashneer Investment Equity	14
Ashneer Debt Amount	2
Guest Investment Amount	24
Guest Investment Equity	25
Guest Debt Amount	8
Invested Guest Name	11
All Guest Names	9
Namita Present	1
Vineeta Present	1
Anupam Present	1
Aman Present	1
Peyush Present	1
Amit Present	1
Ashneer Present	1
Guest Present	2

dtype: int64

```
[12]: #Dealing with null values in deal has conditions  
data['Deal Has Conditions'].unique()
```

```
[12]: array([nan, 'yes'], dtype=object)
```

```
[13]: #Replacing Nan values with No  
data['Deal Has Conditions'] = data['Deal Has Conditions'].fillna('no')  
print(data['Deal Has Conditions'].head(10))
```

```
0    no  
1    no  
2    no  
3    no  
4    no  
5    no  
6    no  
7    no  
8    yes  
9    no  
Name: Deal Has Conditions, dtype: object
```

```
[14]: print(data['Guest Present'].head())  
#we can see that there are NaN values here as well, let us replace them with  
↳Not Present
```

```
0    NaN  
1    NaN  
2    NaN  
3    NaN  
4    NaN  
Name: Guest Present, dtype: float64
```

```
[15]: #Replacing values in Guest Present  
data['Guest Present'] = data['Guest Present'].fillna('Not Present')  
print(data['Guest Present'].head())
```

```
0    Not Present  
1    Not Present  
2    Not Present  
3    Not Present  
4    Not Present  
Name: Guest Present, dtype: object
```

```
[16]: #Getting datatype of the column with float datatype  
print(data.select_dtypes(include='float64').info())
```

<class 'pandas.core.frame.DataFrame'>

RangeIndex: 478 entries, 0 to 477

Data columns (total 54 columns):

#	Column	Non-Null Count	Dtype
0	Started in	355 non-null	float64
1	Male Presenters	412 non-null	float64
2	Female Presenters	226 non-null	float64
3	Transgender Presenters	3 non-null	float64
4	Couple Presenters	473 non-null	float64
5	Yearly Revenue	241 non-null	float64
6	Monthly Sales	225 non-null	float64
7	Gross Margin	129 non-null	float64
8	Net Margin	73 non-null	float64
9	EBITDA	23 non-null	float64
10	SKUs	28 non-null	float64
11	Original Ask Amount	478 non-null	float64
12	Original Offered Equity	478 non-null	float64
13	Valuation Requested	478 non-null	float64
14	Accepted Offer	321 non-null	float64
15	Total Deal Amount	268 non-null	float64
16	Total Deal Equity	268 non-null	float64
17	Total Deal Debt	62 non-null	float64
18	Debt Interest	45 non-null	float64
19	Deal Valuation	267 non-null	float64
20	Number of Sharks in Deal	268 non-null	float64
21	Royalty Deal	17 non-null	float64
22	Advisory Shares Equity	3 non-null	float64
23	Namita Investment Amount	89 non-null	float64
24	Namita Investment Equity	89 non-null	float64
25	Namita Debt Amount	17 non-null	float64
26	Vineeta Investment Amount	68 non-null	float64
27	Vineeta Investment Equity	68 non-null	float64
28	Vineeta Debt Amount	12 non-null	float64
29	Anupam Investment Amount	75 non-null	float64
30	Anupam Investment Equity	75 non-null	float64
31	Anupam Debt Amount	8 non-null	float64
32	Aman Investment Amount	109 non-null	float64
33	Aman Investment Equity	109 non-null	float64
34	Aman Debt Amount	18 non-null	float64
35	Peyush Investment Amount	89 non-null	float64
36	Peyush Investment Equity	89 non-null	float64
37	Peyush Debt Amount	12 non-null	float64
38	Amit Investment Amount	35 non-null	float64
39	Amit Investment Equity	35 non-null	float64
40	Amit Debt Amount	7 non-null	float64
41	Ashneer Investment Amount	21 non-null	float64
42	Ashneer Investment Equity	21 non-null	float64

```

43 Ashneer Debt Amount      2 non-null      float64
44 Guest Investment Amount  48 non-null      float64
45 Guest Investment Equity  48 non-null      float64
46 Guest Debt Amount       9 non-null      float64
47 Namita Present          389 non-null     float64
48 Vineeta Present         318 non-null     float64
49 Anupam Present          424 non-null     float64
50 Aman Present            420 non-null     float64
51 Peyush Present          307 non-null     float64
52 Amit Present            137 non-null     float64
53 Ashneer Present         99 non-null      float64
dtypes: float64(54)
memory usage: 201.8 KB
None

```

```

[17]: #Filling all the null values with 0 in the float columns.
data[['Male Presenters', 'Female Presenters', 'Transgender Presenters',
      'Couple Presenters', 'Number of Sharks in Deal', 'Namita Present',
      'Vineeta Present', 'Anupam Present', 'Aman Present', 'Peyush Present',
      'Amit Present', 'Ashneer Present']] = data[['Male Presenters', 'Female_
↳Presenters', 'Transgender Presenters',
      'Couple Presenters', 'Number of Sharks in Deal', 'Namita Present',
      'Vineeta Present', 'Anupam Present', 'Aman Present', 'Peyush Present',
      'Amit Present', 'Ashneer Present']].fillna(0)

```

```

[18]: print(data[['Male Presenters', 'Female Presenters']].head(10))

```

	Male Presenters	Female Presenters
0	2.0	1.0
1	1.0	0.0
2	0.0	1.0
3	2.0	0.0
4	1.0	3.0
5	1.0	1.0
6	0.0	2.0
7	1.0	0.0
8	2.0	0.0
9	1.0	1.0

```

[19]: #Changing the datatype into int
data[['Male Presenters', 'Female Presenters', 'Transgender Presenters',
      'Couple Presenters', 'Number of Sharks in Deal', 'Namita Present',
      'Vineeta Present', 'Anupam Present', 'Aman Present', 'Peyush Present',
      'Amit Present', 'Ashneer Present']] = data[['Male Presenters', 'Female_
↳Presenters', 'Transgender Presenters',
      'Couple Presenters', 'Number of Sharks in Deal', 'Namita Present',
      'Vineeta Present', 'Anupam Present', 'Aman Present', 'Peyush Present',

```

```
'Amit Present', 'Ashneer Present']]).astype(int)
```

```
[20]: #Checking the erros in cash burn
print(data['Cash Burn'].head())
# as we can see there are NaN values. Let us replace them with No info
```

```
0    NaN
1    NaN
2    NaN
3    NaN
4    NaN
```

Name: Cash Burn, dtype: object

```
[21]: #Replacing nulla values with No
data['Cash Burn'] = data['Cash Burn'].fillna('No')
print(data['Cash Burn'].head())
```

```
0    No
1    No
2    No
3    No
4    No
```

Name: Cash Burn, dtype: object

```
[22]: print(data.columns)
```

```
Index(['Season Number', 'Startup Name', 'Episode Number', 'Pitch Number',
      'Season Start', 'Season End', 'Original Air Date', 'Episode Title',
      'Anchor', 'Industry', 'Business Description', 'Started in',
      'Number of Presenters', 'Male Presenters', 'Female Presenters',
      'Transgender Presenters', 'Couple Presenters', 'Pitchers Average Age',
      'Pitchers City', 'Pitchers State', 'Yearly Revenue', 'Monthly Sales',
      'Gross Margin', 'Net Margin', 'EBITDA', 'Cash Burn', 'SKUs',
      'Has Patents', 'Bootstrapped', 'Original Ask Amount',
      'Original Offered Equity', 'Valuation Requested', 'Received Offer',
      'Accepted Offer', 'Total Deal Amount', 'Total Deal Equity',
      'Total Deal Debt', 'Debt Interest', 'Deal Valuation',
      'Number of Sharks in Deal', 'Deal Has Conditions', 'Royalty Deal',
      'Advisory Shares Equity', 'Namita Investment Amount',
      'Namita Investment Equity', 'Namita Debt Amount',
      'Vineeta Investment Amount', 'Vineeta Investment Equity',
      'Vineeta Debt Amount', 'Anupam Investment Amount',
      'Anupam Investment Equity', 'Anupam Debt Amount',
      'Aman Investment Amount', 'Aman Investment Equity', 'Aman Debt Amount',
      'Peyush Investment Amount', 'Peyush Investment Equity',
      'Peyush Debt Amount', 'Amit Investment Amount',
      'Amit Investment Equity', 'Amit Debt Amount',
      'Ashneer Investment Amount', 'Ashneer Investment Equity',
```

```

'Ashneer Debt Amount', 'Guest Investment Amount',
'Guest Investment Equity', 'Guest Debt Amount', 'Invested Guest Name',
'All Guest Names', 'Namita Present', 'Vineeta Present',
'Anupam Present', 'Aman Present', 'Peyush Present', 'Amit Present',
'Ashneer Present', 'Guest Present'],
dtype='object')

```

```

[23]: #Removing unnecessary columns for our analysis
data.drop(columns=['Has Patents'],inplace = True)

```

```

[24]: #Cleaning the sales data
data[['Yearly Revenue','Monthly Sales','Gross Margin','Net Margin']].head()

```

```

[24]:   Yearly Revenue  Monthly Sales  Gross Margin  Net Margin
0           95.0           8.0           NaN           NaN
1            4.0           0.4           NaN           NaN
2            NaN           2.0           NaN           NaN
3          700.0           NaN          48.0           NaN
4           30.0           NaN           NaN           NaN

```

```

[25]: #Getting the number of nulls in the datas
data[['Yearly Revenue','Monthly Sales','Gross Margin','Net Margin']].isnull().
    ↪sum()

```

```

[25]: Yearly Revenue    237
      Monthly Sales    253
      Gross Margin     349
      Net Margin       405
      dtype: int64

```

```

[26]: #Filling the null values
data[['Yearly Revenue','Monthly Sales',
      'Gross Margin','Net Margin']] = data[['Yearly Revenue','Monthly Sales',
      'Gross Margin', 'Net_
      ↪Margin']].fillna(0)

```

```

[27]: #Checking the errors in Accepted offer
data['Accepted Offer'].isnull().sum()

#Filling null values with 0
data['Accepted Offer'] = data['Accepted Offer'].fillna(0)
print(data['Accepted Offer'].head())

```

```

0    1.0
1    1.0
2    1.0
3    1.0

```

```
4      0.0
Name: Accepted Offer, dtype: float64
```

```
[28]: #Changing datatypes
data['Accepted Offer'] = data[['Accepted Offer']].astype(int)
```

```
[29]: #Fixing the offer data
data[["Total Deal Amount", "Original Offered Equity", "Total Deal Equity", "Total_
↳Deal Debt"
, "Debt Interest", "Deal Valuation"]] = data[["Total Deal Amount", "Original Offered_
↳Equity",
                                                "Total Deal Equity", "Total Deal_
↳Debt", "Debt Interest", "Deal Valuation"]].fillna(0)
```

```
[30]: #Checking what other columns have null vaues
data.isnull().sum()
```

```
[30]: Season Number          0
Startup Name                0
Episode Number             0
Pitch Number               0
Season Start               0
Season End                 0
Original Air Date          31
Episode Title              0
Anchor                    0
Industry                   0
Business Description       0
Started in                 123
Number of Presenters       0
Male Presenters            0
Female Presenters          0
Transgender Presenters     0
Couple Presenters          0
Pitchers Average Age       0
Pitchers City              5
Pitchers State             4
Yearly Revenue             0
Monthly Sales              0
Gross Margin               0
Net Margin                 0
EBITDA                    455
Cash Burn                  0
SKUs                      450
Bootstrapped               430
Original Ask Amount        0
Original Offered Equity    0
```

Valuation Requested	0
Received Offer	0
Accepted Offer	0
Total Deal Amount	0
Total Deal Equity	0
Total Deal Debt	0
Debt Interest	0
Deal Valuation	0
Number of Sharks in Deal	0
Deal Has Conditions	0
Royalty Deal	461
Advisory Shares Equity	475
Namita Investment Amount	389
Namita Investment Equity	389
Namita Debt Amount	461
Vineeta Investment Amount	410
Vineeta Investment Equity	410
Vineeta Debt Amount	466
Anupam Investment Amount	403
Anupam Investment Equity	403
Anupam Debt Amount	470
Aman Investment Amount	369
Aman Investment Equity	369
Aman Debt Amount	460
Peyush Investment Amount	389
Peyush Investment Equity	389
Peyush Debt Amount	466
Amit Investment Amount	443
Amit Investment Equity	443
Amit Debt Amount	471
Ashneer Investment Amount	457
Ashneer Investment Equity	457
Ashneer Debt Amount	476
Guest Investment Amount	430
Guest Investment Equity	430
Guest Debt Amount	469
Invested Guest Name	430
All Guest Names	334
Namita Present	0
Vineeta Present	0
Anupam Present	0
Aman Present	0
Peyush Present	0
Amit Present	0
Ashneer Present	0
Guest Present	0
dtype: int64	

```
[31]: #Dropping more columns
percent_cal = data[['Started in', 'SKUs', 'EBITDA', 'Bootstrapped']]
null_percentage = (percent_cal.isnull().sum()/len(percent_cal))*100
print(null_percentage)
```

```
Started in      25.732218
SKUs            94.142259
EBITDA          95.188285
Bootstrapped    89.958159
dtype: float64
```

```
[32]: #Since the above columns have a lot of null values, we can drop them except for
↳ the started in column, we will replace that with 0
data.drop(columns = ['SKUs', 'EBITDA', 'Bootstrapped'], inplace = True)
```

```
[33]: #Replacing the values in Started in
data['Started in'].fillna(0, inplace = True)
```

```
[33]:
```

1 Analysis and Insights

```
[34]: #Number of seasons and episodes aired per season
episodes = data[data['Episode Number'] != 0]
episodes.groupby(['Season Number'])['Episode Number'].nunique() #number of
↳ episodes per season
```

```
[34]: Season Number
1      36
2      51
3      52
Name: Episode Number, dtype: int64
```

```
[35]: #Sum of all episodes aired till date
episodes.groupby(['Season Number'])['Episode Number'].nunique().sum()
```

```
[35]: 139
```

```
[36]: #Total number of pitches on Shark Tank
data['Pitch Number'].nunique()
```

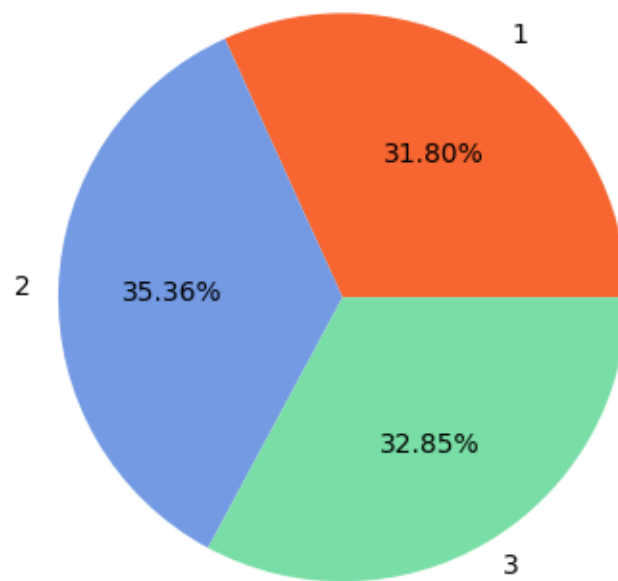
```
[36]: 478
```

```
[37]: #Total number of pitches per season
pitch_no = data[data['Pitch Number'] != 0]
pitch_no.groupby(['Season Number'])['Pitch Number'].count()
```



```
[37]: Season Number
      1    152
      2    169
      3    157
      Name: Pitch Number, dtype: int64
```

```
[38]: plt.pie(data.groupby(["Season Number"])["Pitch Number"].count().
      ↪values,labels=data.groupby(["Season Number"])["Pitch Number"].count().
      ↪index,autopct="%.2f%%",
      colors=['#F76530', '#749AE4', '#79DEA6']);
```



```
[39]: #What are the different industries in the pictes
      industry = data['Industry'].unique()
      print(industry)
```

```
['Food' 'Vehicles/Electrical Vehicles' 'Beauty/Fashion' 'Education'
 'Agriculture' 'Medical/Health' 'Manufacturing' 'Technology/Software'
 'Electronics' 'Animal/Pets' 'Services' 'Hardware' 'Sports'
 'Liquor/Beverages' 'Entertainment' 'Furnishing/Household' 'Others']
```

```
[40]: #Top 5 industries that participated in Shark tank
      data['Industry'].value_counts()[0:5]
```

```
#We can see that food is the most pitched industry followed by Beauty and  
↪Manufacturing
```

```
[40]: Industry  
Food          107  
Beauty/Fashion 95  
Manufacturing 50  
Technology/Software 49  
Services      39  
Name: count, dtype: int64
```

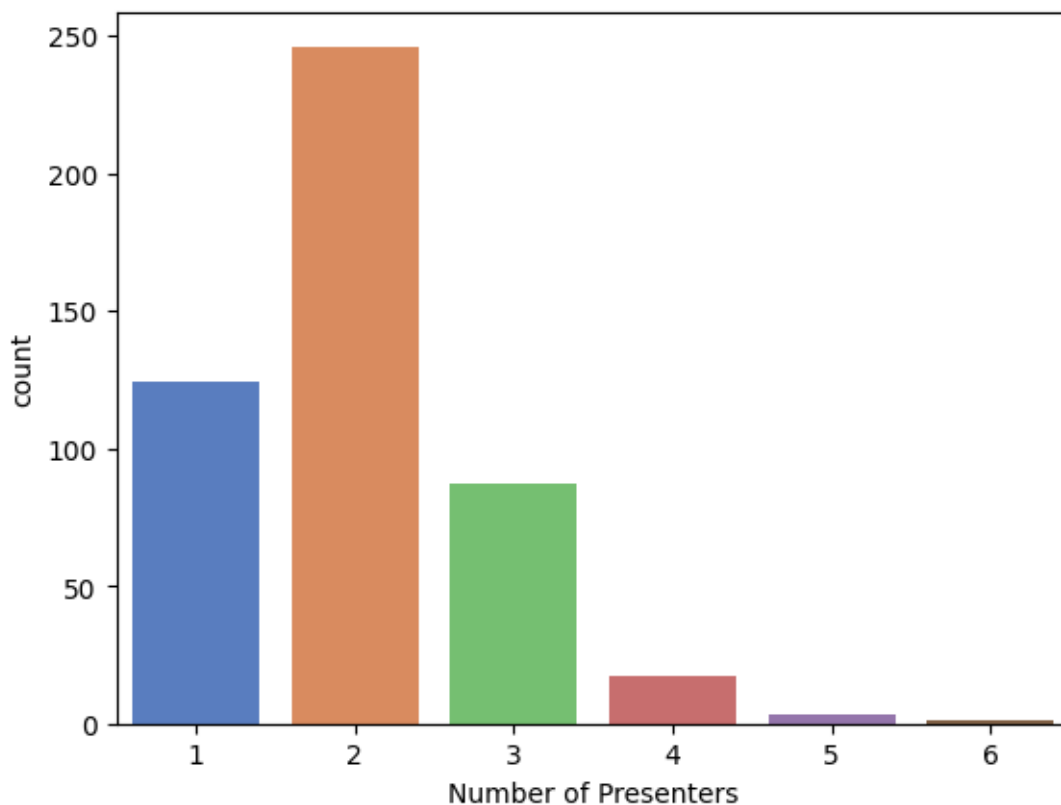
```
[41]: #Number of Presenters in most teams  
sns.countplot(x='Number of Presenters',data=data,palette='muted')
```

```
<ipython-input-41-8e75284ed8e4>:2: FutureWarning:
```

Passing `palette` without assigning `hue` is deprecated and will be removed in v0.14.0. Assign the `x` variable to `hue` and set `legend=False` for the same effect.

```
sns.countplot(x='Number of Presenters',data=data,palette='muted')
```

```
[41]: <Axes: xlabel='Number of Presenters', ylabel='count'>
```



```
[42]: #Total pitchers, Number of male pitchers and female pitchers
print('Total Pitchers:',int(data['Number of Presenters'].sum()))
print('The number of male Pitches:',int(data['Male Presenters'].sum()))
print('The number of Female Pitches:',int(data['Female Presenters'].sum()))
print('The number of Transgender Pitches:',int(data['Transgender Presenters'].
↪sum()))
print('The number of Couple Pitches:',int(data['Couple Presenters'].sum()))
```

Total Pitchers: 966
The number of male Pitches: 692
The number of Female Pitches: 271
The number of Transgender Pitches: 3
The number of Couple Pitches: 84

```
[43]: #Number of startups during COVID (started between 2019 - 2021)
print('The total number of startups started during COVID: ',data[data['Started_
↪in'].between(2019, 2021)]['Startup Name'].count())
```

The total number of startups started during COVID: 182

```
[44]: #Number of startups that rejected the shark offers
number_rejected = data[data['Accepted Offer'] == 0]['Startup Name'].count()
print('The Number of Startups that rejected the shark offers:',number_rejected)
```

The Number of Startups that rejected the shark offers: 210

```
[45]: #Number of all shark deals
all_shark = data[data['Number of Sharks in Deal'] == 5]['Startup Name'].count()
print('Number of all shark deal:',all_shark)

avg_sharks_per_deal = round(data['Number of Sharks in Deal'].mean(),2)
print('The average number of sharks per deal:',avg_sharks_per_deal)
```

Number of all shark deal: 14
The average number of sharks per deal: 1.13

```
[46]: #Number of companies that got same valuation as requested
val_as_req = data[data['Valuation Requested'] == data['Deal_
↪Valuation']][['Startup Name']].count()
print('Number of companies that got same valuation as requested:',val_as_req)

#Percentage of these startups to the total pitches
total_pitches = data['Pitch Number'].nunique()
per_of_val_as_req = round((val_as_req/total_pitches)*100,2)
```

```
print('Percentage of startups that got same valuation as requested:
↳',per_of_val_as_req)
```

Number of companies that got same valuation as requested: 31
 Percentage of startups that got same valuation as requested: 6.49

```
[47]: #Startups with highest Monthly revenue
highest_monthly_rev = int(data['Monthly Sales'].max())
filtered_data = data[data['Monthly Sales'] == highest_monthly_rev]
startup_name = filtered_data['Startup Name']
print('The highest Revenue by a startup in a month:',startup_name.iloc[0] + '
↳earned ' + str(highest_monthly_rev) + ' Lakhs.')

#Startups with highest yearly revenue
highest_yearly_rev = int(data['Yearly Revenue'].max())
filter_data = data[data['Yearly Revenue'] == highest_yearly_rev]
startup_names = filter_data['Startup Name']
print('The highest Revenue by a startup in a year:',startup_names.iloc[0] + '
↳earned ' + str(highest_yearly_rev) + ' Lakhs.')
```

The highest Revenue by a startup in a month: Refit earned 3500 Lakhs.
 The highest Revenue by a startup in a year: Refit earned 18700 Lakhs.

```
[48]: #Number of pre revenue startups
pre_rev = data[data['Yearly Revenue'] == 0]['Startup Name'].count()
print('The number of pre revenue startups:', pre_rev)
print('\n')

#Number of pre revenue startups per season
pre_rev_data = data[data['Yearly Revenue'] == 0]
pre_revenue_per_season = pre_rev_data.groupby('Season Number')['Season Number'].
↳count()
print('Total Number of Pre Revenue startups per season:',pre_revenue_per_season)
```

The number of pre revenue startups: 260

Total Number of Pre Revenue startups per season: Season Number

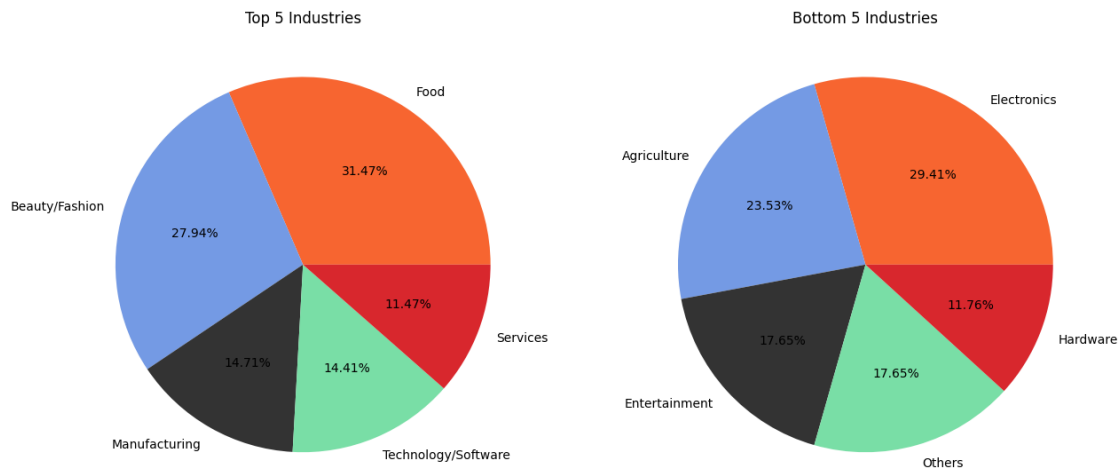
1	85
2	125
3	50

Name: Season Number, dtype: int64

```
[49]: #Top 5 and Bottom 5 industries
color = ['#F76530', '#749AE4', '#333333', '#79DEA6', '#D8272D']
plt.figure(figsize=(15,8))
plt.subplot(1,2,1)
```

```
plt.pie(data["Industry"].value_counts()[0:5].values,labels=data["Industry"].
↪value_counts()[0:5].index,autopct="%.2f%%",colors=color);
plt.title("Top 5 Industries")
plt.subplot(1,2,2)
plt.pie(data["Industry"].value_counts().tail(5).values,labels=data["Industry"].
↪value_counts().tail(5).index,autopct="%.2f%%",colors=color);
plt.title("Bottom 5 Industries")
```

```
[49]: Text(0.5, 1.0, 'Bottom 5 Industries')
```



```
[50]: #Top 5 cities where pitchers were from
top_5 = data['Pitchers City'].value_counts()[0:5]
print('Top 5 cities:',top_5)
```

```
Top 5 cities: Pitchers City
Mumbai      76
Delhi       67
Bangalore   50
Pune        27
Ahmedabad   24
Name: count, dtype: int64
```

```
[51]: #Number of offers that were accepted by season
acc_off = data.groupby(['Season Number'])['Accepted Offer'].value_counts()
print('Total number of offer accepted by season:',acc_off)
#0 = No, 1 = Yes
```

```
Total number of offer accepted by season: Season Number Accepted Offer
1      0      82
      1      70
```

2	1	106
	0	63
3	1	92
	0	65

Name: count, dtype: int64

```
[52]: #Total Amount invested by Anupam per season
anupam_amt = round(data.groupby(['Season Number'])['Anupam Investment Amount'].
    ↪sum(),2)
print('Total Amount invested by Anupam per season in Lakhs:',anupam_amt)
```

Total Amount invested by Anupam per season in Lakhs: Season Number

1	536.33
2	914.83
3	768.32

Name: Anupam Investment Amount, dtype: float64

```
[53]: #Total Amount invested by Ashneer per season
ashneer_amt = round(data.groupby(['Season Number'])['Ashneer Investment_
    ↪Amount'].sum(),2)
print('Total Amount invested by Ashneer per season in Lakhs:',ashneer_amt)
```

Total Amount invested by Ashneer per season in Lakhs: Season Number

1	539.33
2	0.00
3	0.00

Name: Ashneer Investment Amount, dtype: float64

```
[54]: #Total Amount invested by Namita per season
namita_amt = round(data.groupby(['Season Number'])['Namita Investment Amount'].
    ↪sum(),2)
print('Total Amount invested by Namita per season in Lakhs:',namita_amt)
```

Total Amount invested by Namita per season in Lakhs: Season Number

1	708.63
2	1388.88
3	834.33

Name: Namita Investment Amount, dtype: float64

```
[55]: #Total Amount invested by Vineeta per season
vineeta_amt = round(data.groupby(['Season Number'])['Vineeta Investment_
    ↪Amount'].sum(),2)
print('Total Amount invested by Vineeta per season in Lakhs:',vineeta_amt)
```

Total Amount invested by Vineeta per season in Lakhs: Season Number

1	363.62
2	806.98
3	954.16

Name: Vineeta Investment Amount, dtype: float64

```
[56]: #Total Amount invested by Peyush per season
peyush_amt = round(data.groupby(['Season Number'])['Peyush Investment Amount'].
    ↪sum(),2)
print('Total Amount invested by Peyush per season in Lakhs:',peyush_amt)
```

Total Amount invested by Peyush per season in Lakhs: Season Number

1 819.65

2 1390.73

3 875.00

Name: Peyush Investment Amount, dtype: float64

```
[57]: #Total Amount invested by Aman per season
Aman_amt = round(data.groupby(['Season Number'])['Aman Investment Amount'].
    ↪sum(),2)
print('Total Amount invested by Peyush per season in Lakhs:',Aman_amt)
```

Total Amount invested by Peyush per season in Lakhs: Season Number

1 896.99

2 1613.05

3 1181.66

Name: Aman Investment Amount, dtype: float64

```
[58]: #Total Amount invested by sharks

#Total amount invested by Aman
print('Total Amount invest by Peyush in Lakhs:',round(data['Aman Investment_
    ↪Amount'].sum(),2))

#Total amount invested by Peyush
print('Total Amount invest by Peyush in Lakhs:',round(data['Peyush Investment_
    ↪Amount'].sum(),2))

#Total amount invested by Vineeta
print('Total Amount invest by Vineeta in Lakhs:',round(data['Vineeta Investment_
    ↪Amount'].sum(),2))

#Total amount invested by Namita
print('Total Amount invest by Namita in Lakhs:',round(data['Namita Investment_
    ↪Amount'].sum(),2))

#Total amount invested by Ashneer
print('Total Amount invest by Ashneer in Lakhs:',round(data['Ashneer Investment_
    ↪Amount'].sum(),2))

#Total amount invested by Anupam
```

```
print('Total Amount invest by Anupam in Lakhs:',round(data['Anupam Investment_↵↵Amount'].sum(),2))
```

Total Amount invest by Peyush in Lakhs: 3691.7
Total Amount invest by Peyush in Lakhs: 3085.38
Total Amount invest by Vineeta in Lakhs: 2124.76
Total Amount invest by Namita in Lakhs: 2931.84
Total Amount invest by Ashneer in Lakhs: 539.33
Total Amount invest by Anupam in Lakhs: 2219.48