

Examine the dataset and use the most effective visualization graph to address the following questions. Provide your answers and analyses for each question alongside the visualizations.

- How Does the Funding Ecosystem changes with respect to Time?
- What is the General Amount that Startups get in India?
- Which Kind of Industries are more preferred for Startups?
- Does Location also play a role, In determining the Growth of a Startup?
- Who plays the main role in Indian Startups Ecosystem?
- What are the different Types of Funding for Startups?

```
In [4]: import pandas as pd
import matplotlib.pyplot as plt
import seaborn as sns

# Loading the dataset
df = pd.read_csv("D:\\AIML\\5000 0NB_Data Analytics\\ASSIGN - 1\\startup_funding.csv")
df.head()
```

Out[4]:

	Sr No	Date dd/mm/yyyy	Startup Name	Industry Vertical	SubVertical	City Location	Investors Name	InvestmentnType	Amount in USD	Remarks
0	1	09/01/2020	BYJU'S	E-Tech	E-learning	Bengaluru	Tiger Global Management	Private Equity Round	20,00,00,000	NaN
1	2	13/01/2020	Shuttl	Transportation	App based shuttle service	Gurgaon	Susquehanna Growth Equity	Series C	80,48,394	NaN
2	3	09/01/2020	Mamaearth	E-commerce	Retailer of baby and toddler products	Bengaluru	Sequoia Capital India	Series B	1,83,58,860	NaN
3	4	02/01/2020	https://www.wealthbucket.in/	FinTech	Online Investment	New Delhi	Vinod Khatumal	Pre-series A	30,00,000	NaN
4	5	02/01/2020	Fashor	Fashion and Apparel	Embroided Clothes For Women	Mumbai	Sprout Venture Partners	Seed Round	18,00,000	NaN

\*1. How Does the Funding Ecosystem change with respect to Time?\*

```
In [6]: # Clean the date column (assuming it's in 'dd/mm/yyyy' format)
df['Date'] = pd.to_datetime(df['Date dd/mm/yyyy'], format='%d/%m/%Y', errors='coerce')

# Step to extract the year
df['Year'] = df['Date'].dt.year

# Display the updated dataset with the extracted year
df[['Date', 'Year']].head()
```

Out[6]:

	Date	Year
0	2020-01-09	2020.0
1	2020-01-13	2020.0
2	2020-01-09	2020.0
3	2020-01-02	2020.0
4	2020-01-02	2020.0

```
In [7]: # Step 1: Clean 'Amount in USD' column
# Remove commas and convert to numeric, forcing errors to NaN
df['Amount in USD'] = pd.to_numeric(df['Amount in USD'].str.replace(',', '').str.replace('₹', '').str.strip(), errors='coerce')

# Check the cleaned dataset
df[['Amount in USD']].head()
```

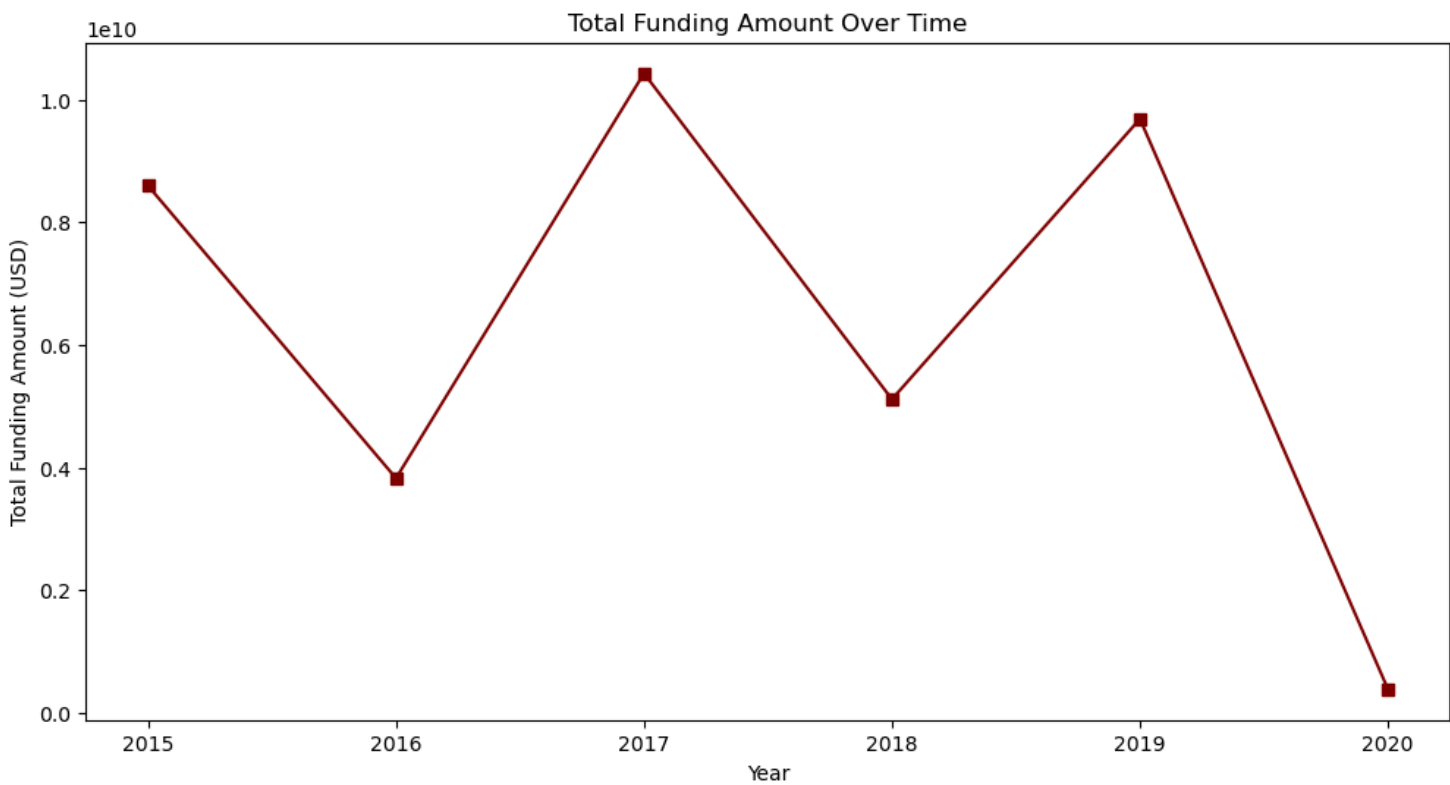
```
Out[7]:
```

	Amount in USD
0	200000000.0
1	8048394.0
2	18358860.0
3	3000000.0
4	1800000.0

```
In [8]: # Aggregate the total funding by year
funding_by_year = df.groupby('Year')['Amount in USD'].sum().reset_index()
print(funding_by_year)
```

```
   Year  Amount in USD
0  2015.0  8.597207e+09
1  2016.0  3.828089e+09
2  2017.0  1.042931e+10
3  2018.0  5.116118e+09
4  2019.0  9.686577e+09
5  2020.0  3.902073e+08
```

```
In [19]: # Create a line plot to visualize the funding trends
plt.figure(figsize=(12, 6))
plt.plot(funding_by_year['Year'], funding_by_year['Amount in USD'], marker='s', color='maroon')
plt.title('Total Funding Amount Over Time')
plt.xlabel('Year')
plt.ylabel('Total Funding Amount (USD)')
plt.xticks(funding_by_year['Year']) # Show all years on x-axis
plt.show()
```



**\*1.A.From the above graph, we can analyse that from 2015 to 2020, there is a similar pattern of rise an fall in each year. Though the highs has a significantly less difference, the lows drastically fall by 2020. From the funding ecosystem, we can expect a rise and expect it to be similar with the highs.\***

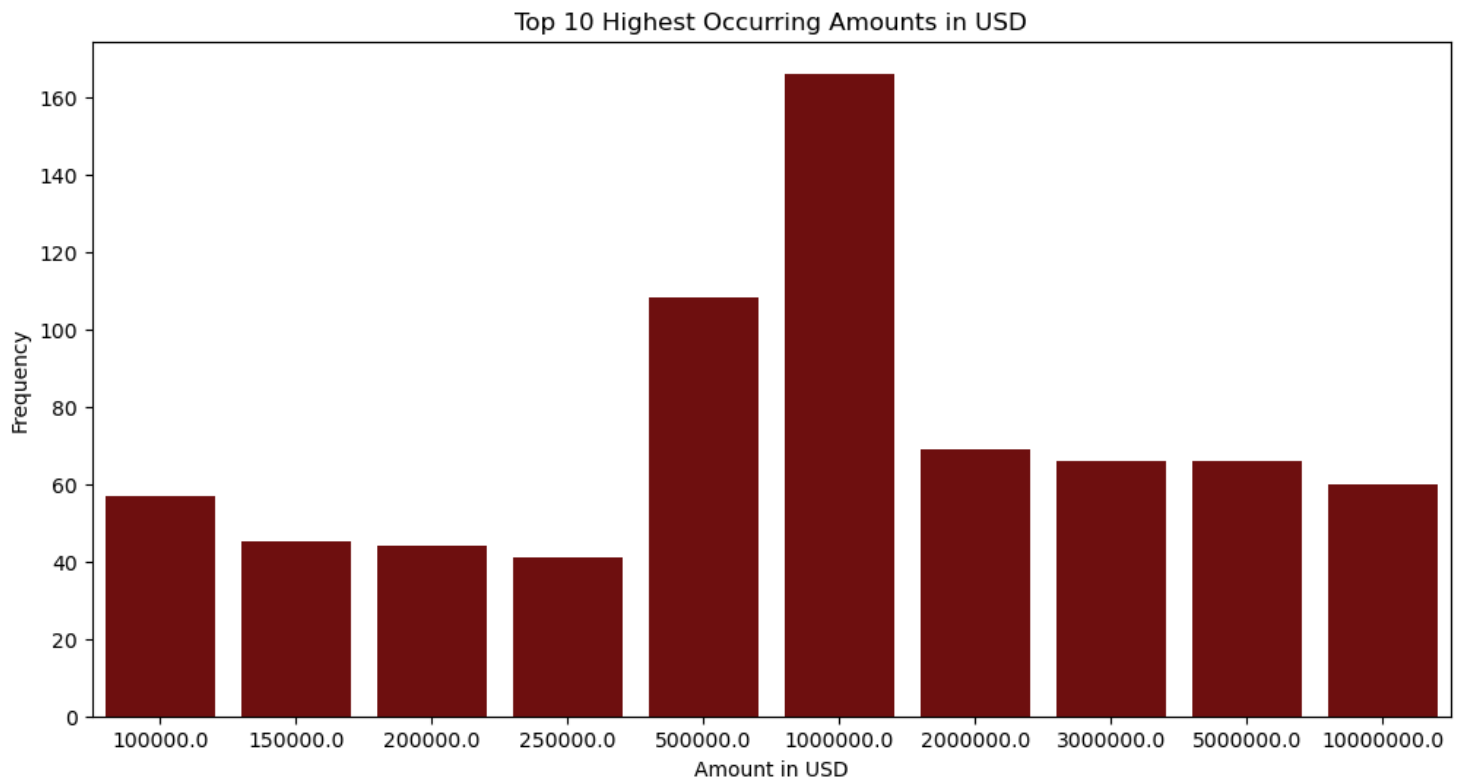
```
In [ ]:
```

**\*2. What is the General Amount that Startups get in India?\***

```
In [34]: # Top 10 frequent values in 'Amount in USD'
top_10_amounts = df['Amount in USD'].value_counts().nlargest(10)

# Top 10 highest occurring amounts using a bar plot
plt.figure(figsize=(12, 6))
sns.barplot(x=top_10_amounts.index, y=top_10_amounts.values, color='maroon')
plt.title('Top 10 Highest Occurring Amounts in USD')
plt.xlabel('Amount in USD')
```

```
plt.ylabel('Frequency')
plt.show()
```



**\*2.A.From the above graph, the general amount startup gets is around 1000000.\***

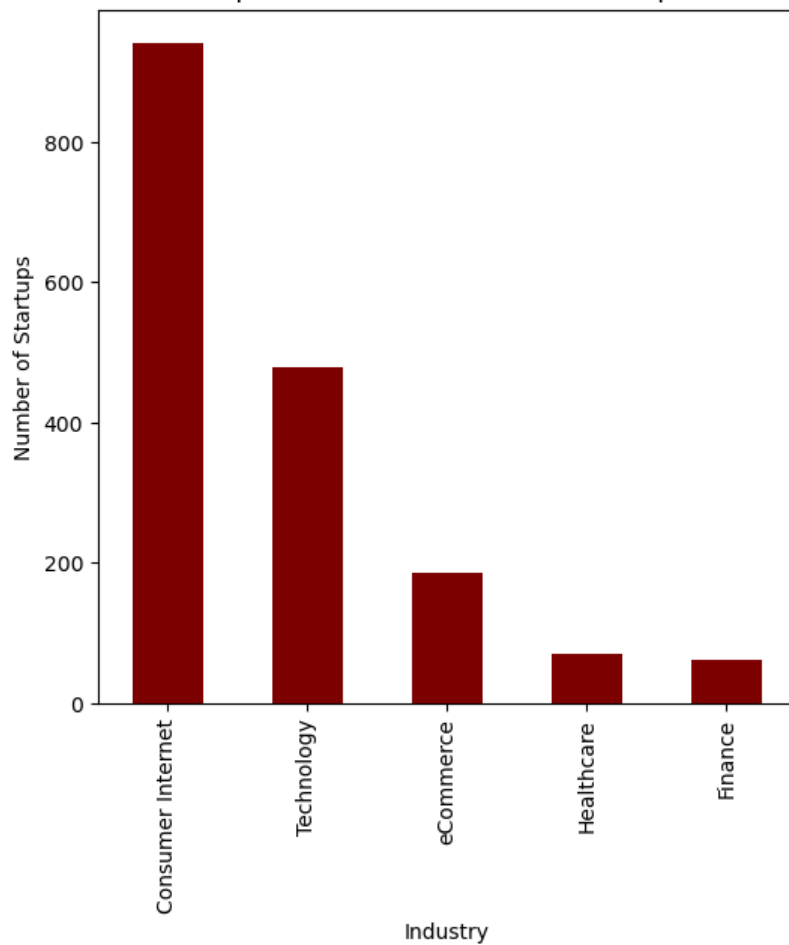
In [ ]:

**\*3. Which Kind of Industries are more preferred for Startups?\***

```
In [44]: industry_counts = df["Industry Vertical"].value_counts().head(5)

# Plot a bar chart
plt.figure(figsize=(6, 6))
industry_counts.plot(kind='bar', color='maroon')
plt.title('Top 5 Preferred Industries for Startups')
plt.xlabel('Industry')
plt.ylabel('Number of Startups')
plt.xticks()
plt.show()
```

Top 5 Preferred Industries for Startups



**\*3.A.From the above graph, Consumer Internet ranks most preferred for Startups.\***

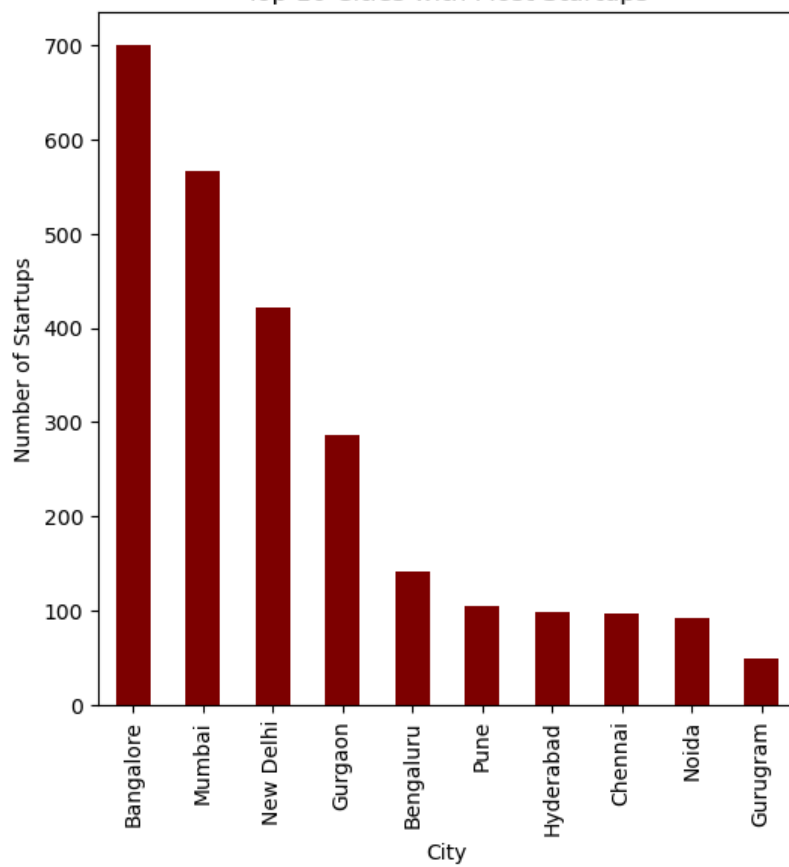
In [ ]:

**\*4.Does Location also play a role, In determining the Growth of a Startup?\***

```
In [55]: # Count the number of startups by city/location
location_counts = df["City Location"].value_counts().head(10)

# Plot a bar chart
plt.figure(figsize=(6, 6))
location_counts.plot(kind='bar', color='maroon')
plt.title('Top 10 Cities with Most Startups')
plt.xlabel('City')
plt.ylabel('Number of Startups')
plt.xticks()
plt.show()
```

Top 10 Cities with Most Startups



\*4.A. From the above graph, the answer is yes, as the number of startups are varying from city to city.\*

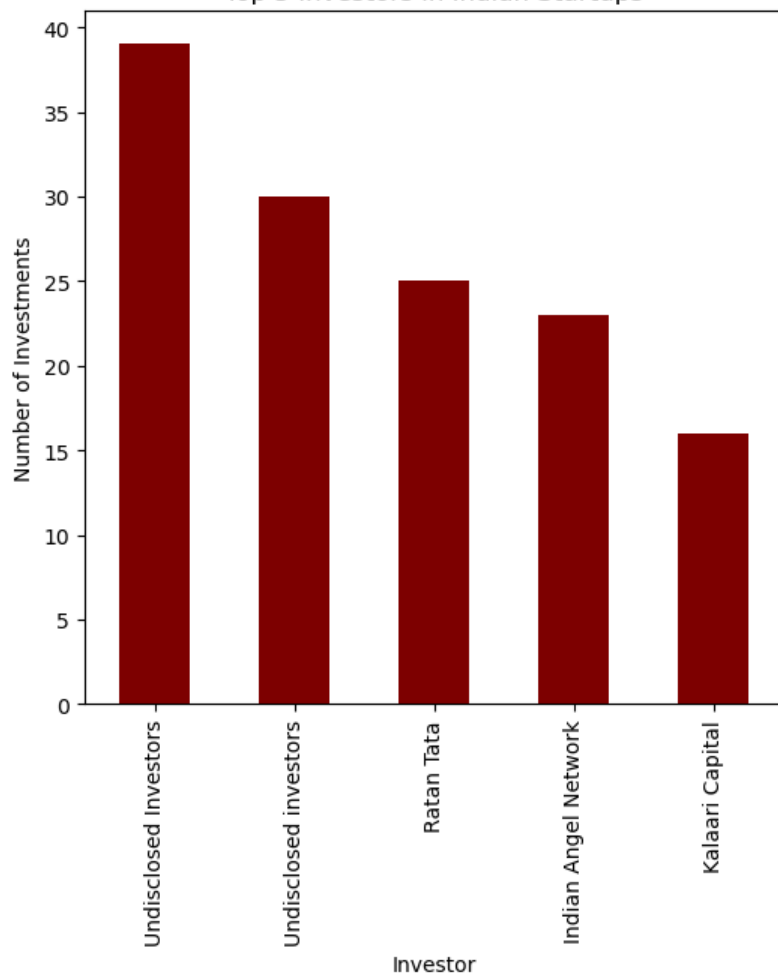
In [ ]:

\*5.Who plays the main role in Indian Startups Ecosystem?\*

```
In [61]: investors= df["Investors Name"].value_counts().head(5)
```

```
# Plot a bar chart
plt.figure(figsize=(6, 6))
investors.plot(kind='bar', color='maroon')
plt.title('Top 5 Investors in Indian Startups')
plt.xlabel('Investor')
plt.ylabel('Number of Investments')
plt.xticks()
plt.show()
```

Top 5 Investors in Indian Startups



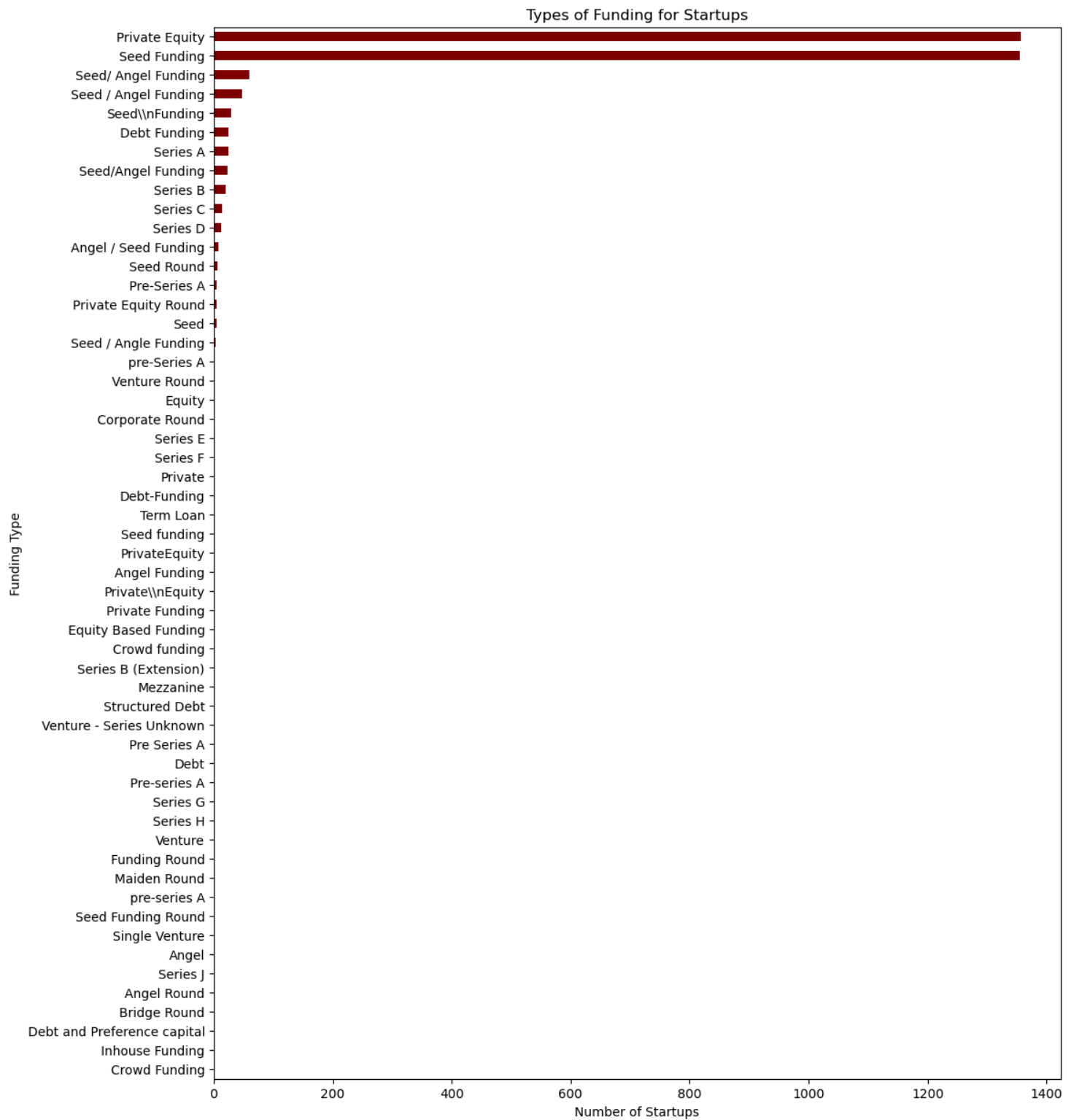
**\*5.A. From the above graph, we can analyse that in Indian Startup ecosystem, many investors who does not disclosed themselves plays major role.\***

In [ ]:

**\*6.What are the different Types of Funding for Startups?\***

```
In [75]: funding_type_counts = df["InvestmentnType"].value_counts()

#Plot a horizontal bar chart for all funding types
plt.figure(figsize=(12, 15))
funding_type_counts.plot(kind='barh', color='maroon')
plt.title('Types of Funding for Startups')
plt.xlabel('Number of Startups')
plt.ylabel('Funding Type')
plt.gca().invert_yaxis()
plt.show()
```



\*6.A.From the graph above we can see different types of funding for startups ranking from the most funded to least.\*