


Foreign Direct Investment Analysis

```
import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
import seaborn as sns
```

```
from google.colab import files
uploaded = files.upload()
```



Choose Files

 No file chosen


Upload widget is only available when the cell has been executed in the current browser session. Please rerun this cell to enable.

Saving FDI data.csv to FDI data.csv

```
df = pd.read_csv('/content/FDI data.csv')
```

Data Exploration

```
df.head()
```



	Sector	2000-01	2001-02	2002-03	2003-04	2004-05	2005-06	2006-07	2007-08	2008-09	2009-10	2010-11	2011-12	2012-13	2013-14	2014-15
0	METALLURGICAL INDUSTRIES	22.69	14.14	36.61	8.11	200.38	149.13	169.94	1175.75	959.94	419.88	1098.14	1786.14	1466.23	567.63	359
1	MINING	1.32	6.52	10.06	23.48	9.92	7.40	6.62	444.36	34.16	174.40	79.51	142.65	57.89	12.73	684
2	POWER	89.42	757.44	59.11	27.09	43.37	72.69	157.15	988.68	907.66	1271.79	1271.77	1652.38	535.68	1066.08	707
3	NON-FUEL MINING AND QUARRYING	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0

```
df.tail()
```



	Sector	2000-01	2001-02	2002-03	2003-04	2004-05	2005-06	2006-07	2007-08	2008-09	2009-10	2010-11	2011-12	2012-13	2013-14	2014-15
58	PRINTING OF BOOKS (INCLUDING LITHO PRINTING IN...	0.00	0.00	6.30	0.00	0.06	9.90	20.04	35.54	31.61	70.51	36.63	47.39	14.34	113	
59	COIR	0.00	0.00	0.00	0.00	0.47	0.59	0.04	0.01	0.00	0.25	0.10	0.55	0.15	0	
60	CONSTRUCTION (INFRASTRUCTURE) ACTIVITIES	0.00	0.00	0.00	0.00	0.00	0.93	64.06	182.92	172.70	324.56	675.07	386.28	283.89	485	
61	CONSTRUCTION OF RESIDENTIAL BUILDINGS	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0

```
df.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 63 entries, 0 to 62
Data columns (total 18 columns):
#   Column      Non-Null Count  Dtype
---  -
0    Sector      63 non-null    object
1    2000-01     63 non-null    float64
2    2001-02     63 non-null    float64
3    2002-03     63 non-null    float64
4    2003-04     63 non-null    float64
5    2004-05     63 non-null    float64
6    2005-06     63 non-null    float64
7    2006-07     63 non-null    float64
8    2007-08     63 non-null    float64
9    2008-09     63 non-null    float64
10   2009-10     63 non-null    float64
11   2010-11     63 non-null    float64
12   2011-12     63 non-null    float64
13   2012-13     63 non-null    float64
```

```

14 2013-14 63 non-null float64
15 2014-15 63 non-null float64
16 2015-16 63 non-null float64
17 2016-17 63 non-null float64
dtypes: float64(17), object(1)
memory usage: 9.0+ KB

```

```
df.describe()
```

	2000-01	2001-02	2002-03	2003-04	2004-05	2005-06	2006-07	2007-08	2008-09	2009-10	2
count	63.000000	63.000000	63.000000	63.000000	63.000000	63.000000	63.000000	63.000000	63.000000	63.000000	63
mean	37.757302	63.931587	42.925714	34.727778	51.090317	87.932540	198.281905	390.085714	498.348571	410.069524	339
std	112.227860	157.878737	86.606439	67.653735	101.934873	206.436967	686.783115	1026.249935	1134.649040	926.814626	627
min	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0
25%	0.000000	0.000000	0.200000	0.215000	0.715000	1.230000	4.160000	9.950000	11.950000	7.880000	8
50%	4.030000	5.070000	11.010000	6.370000	9.090000	22.620000	25.820000	58.820000	84.880000	69.740000	58
75%	23.510000	44.830000	36.555000	38.660000	43.205000	63.855000	108.325000	279.270000	383.320000	341.595000	304
max	832.070000	873.230000	419.960000	368.320000	527.900000	1359.970000	4713.780000	6986.170000	6183.490000	5466.130000	3296

```
df.columns
```

```

Index(['Sector', '2000-01', '2001-02', '2002-03', '2003-04', '2004-05',
      '2005-06', '2006-07', '2007-08', '2008-09', '2009-10', '2010-11',
      '2011-12', '2012-13', '2013-14', '2014-15', '2015-16', '2016-17'],
      dtype='object')

```

```

missing_values = df.isnull().sum()
print("Missing Values\n", missing_values)

```

```

Missing Values
Sector      0
2000-01     0
2001-02     0
2002-03     0
2003-04     0
2004-05     0
2005-06     0
2006-07     0
2007-08     0
2008-09     0
2009-10     0
2010-11     0
2011-12     0
2012-13     0
2013-14     0
2014-15     0
2015-16     0
2016-17     0
dtype: int64

```

```

duplicate_values = df.duplicated().sum()
print('Number of duplicates', duplicate_values)

```

```
Number of duplicates 0
```

✓ KPI's

Total FDI Inflow (Year-Wise)

```

# Transpose the data to get years as rows for easier aggregation
total_fdi_yearwise = df.set_index('Sector').sum()
total_fdi_yearwise

```

«*Математика в школе*» № 1 (2014) ▶

```
# Sector-wise FDI Inflow
total_fdi_sectorwise = df.set_index('Sector').sum(axis=1)
total_fdi_sectorwise
```

63 rows x 1 columns

«*Il y a une différence entre la connaissance et la compréhension. La connaissance est un fait, la compréhension est une expérience.*»

```
def calculate_cagr(fdi_start, fdi_end, years):
    return (fdi_end / fdi_start) ** (1 / years) - 1

# Example for total FDI across all sectors (CAGR from 2000-01 to 2016-17)
fdi_start = total_fdi_yearwise['2000-01']
fdi_end = total_fdi_yearwise['2016-17']
years = 2017 - 2000
```

 0.18640021060790324

Top Contributing Sectors

```
# Top Contributing Sectors (e.g., top 5 sectors)
top_sectors = total_fdi_sectorwise.sort_values(ascending=False).head(5)
top_sectors
```



0

	Sector
SERVICES SECTOR (Fin.,Banking,Insurance,Non Fin/Business,Outsourcing,R&D,Courier,Tech. Testing and Analysis, Other)	59476.49
COMPUTER SOFTWARE & HARDWARE	24669.49
CONSTRUCTION DEVELOPMENT: Townships, housing, built-up infrastructure and construction-development projects	24293.09
TELECOMMUNICATIONS	23946.01
AUTOMOBILE INDUSTRY	16673.92

63 rows × 1 columns	
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FDI Contribution Percentage (Sector-wise)

```
# Total FDI for all sectors over the entire period
total_fdi_all = total_fdi_sectorwise.sum()

# Sector-wise Contribution Percentage
sector_percentage = (total_fdi_sectorwise / total_fdi_all) * 100
sector_percentage
```



0

	Sector
METALLURGICAL INDUSTRIES	3.111693
MINING	0.684305
POWER	3.490797
NON-CONVENTIONAL ENERGY	1.560732
COAL PRODUCTION	0.008356
...	...
PRINTING OF BOOKS (INCLUDING LITHO PRINTING INDUSTRY)	0.191168
COIR	0.001223
CONSTRUCTION (INFRASTRUCTURE) ACTIVITIES	2.957150
CONSTRUCTION DEVELOPMENT: Townships, housing, built-up infrastructure and construction-development projects	7.317394
MISCELLANEOUS INDUSTRIES	3.025218


63 rows × 1 columns

63 rows × 1 columns	
---------------------	--

Year-over-Year (YoY) Growth Rate

```
# Calculate YoY growth for each sector
df.set_index(df.columns[0], inplace=True)
df = df.T # Transpose so years are rows

# Year-over-Year Growth Rate
yoy_growth = df.pct_change() * 100
# Optionally, reset index if you need the years as a column again
yoy_growth.reset_index(inplace=True)
yoy_growth
```



	22.69	2000-01	14.14	36.61	8.11	200.38	149.13	169.94	1175.75	959.94	41
0	1.32		NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	
1	89.42	1.151718e+04	487.574553	1.537479e+01	337.197581	882.297297	2273.867069	1.224953e+02	2.557084e+03	6.292374	
2	0.00	-1.000000e+02	-97.124006	-8.471761e+01	-97.071709	-98.142798	-98.447343	-9.405065e+01	-8.613137e+01	-5.105167	
3	0.00		NaN	-100.000000	-9.903382e+01	-100.000000	577.037037	-46.721311	-7.606256e+01	-9.982523e+01	-1.000000
4	9.35		inf	inf	2.015000e+05	inf	32.275711	6646.923077	9.878977e+03	1.586682e+05	
...	
57	0.00	-1.000000e+02		inf	-1.000000e+02	500.000000	592.307692	223.225806	8.764520e+01	-1.475189e+01	7.651534
58	0.00		NaN	-100.000000		NaN	683.333333	-94.040404	-99.800399	-9.997186e+01	-1.000000e+02
59	0.00		NaN	NaN	NaN	-100.000000	57.627119	160050.000000	1.829100e+06		inf
60	24.33		inf	inf	inf	inf	24492.473118	2074.445832	2.025153e+03	2.596879e+03	1.584166
61	832.07	3.277681e+02	505.983380	4.005952e+02	-19.880310	-27.961174	-78.113357	-8.640661e+01	-6.672686e+01	-7.900595	

62 rows × 17 columns

Sector-wise Average FDI per Year

```
# Sector-wise Average FDI per Year
df.set_index(df.columns[0], inplace=True)
fdi_data = df.apply(pd.to_numeric, errors='coerce')
average_fdi_sectorwise = df.mean(axis=1)
average_fdi_sectorwise
```




	0
36.61	
10.06	160.995000
59.11	763.082857
1.70	369.985000
0.00	1.981429
56.78	469.925714
...	...
6.30	44.882857
0.00	0.290000
0.00	701.247857
36.10	1727.207857
218.76	626.517857

62 rows × 1 columns

Volatility of FDI (Standard Deviation)

```
# Volatility of FDI (Standard Deviation)
sector_volatility = df.std(axis=1)
sector_volatility
```



	0
36.61	
10.06	221.827041
59.11	523.591507
1.70	365.131948
0.00	4.272657
56.78	607.424548
...	...
6.30	38.677425
0.00	0.385507
0.00	1201.777983
36.10	1828.387280
218.76	474.493065

62 rows × 1 columns

« »

✓ Data Analysis and Visualisation

FDI Inflow Distribution both Sector-wise and Year-wise

```
# Set 'Sector' as the index
df.set_index(df.columns[0], inplace=True)

# Convert data to numeric (in case there are any non-numeric values)
df = fdi_data.apply(pd.to_numeric, errors='coerce')

# Calculate Total FDI by Sector
total_fdi_by_sector = fdi_data.sum(axis=1)

# Calculate Total FDI by Year
total_fdi_by_year = fdi_data.sum(axis=0)

# Plot Sector-wise FDI Distribution
plt.figure(figsize=(12, 6))


# Bar plot for Sector-wise FDI
plt.subplot(1, 2, 1)
total_fdi_by_sector.plot(kind='bar', color='skyblue')
plt.title('Total FDI Inflow Distribution by Sector')
plt.xlabel('Sector')
plt.ylabel('Total FDI Inflow (in Crores)')
plt.xticks(rotation=45)
plt.grid(axis='y')

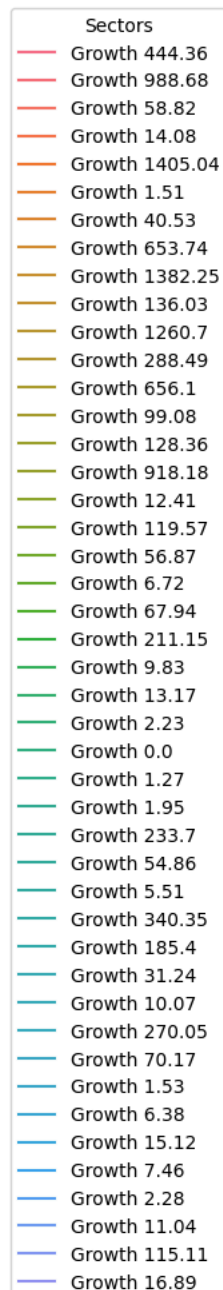
# Plot Year-wise FDI Distribution
plt.subplot(1, 2, 2)
total_fdi_by_year.plot(kind='bar', color='lightgreen')
plt.title('Total FDI Inflow Distribution by Year')
plt.xlabel('Year')
plt.ylabel('Total FDI Inflow (in Crores)')
plt.xticks(rotation=45)
plt.grid(axis='y')

# Show the plots
plt.tight_layout()
plt.show()
```




<https://colab.research.google.com/drive/1GnQvjS3KRXCwhlUP4F4Qk9hKdxo-U96e?authuser=1#printMode=true>

 <ipython-input-54-75ef5fb2a06c>:44: UserWarning: Tight layout not applied. tight_layout cannot make axes height small enough to a
plt.tight_layout()



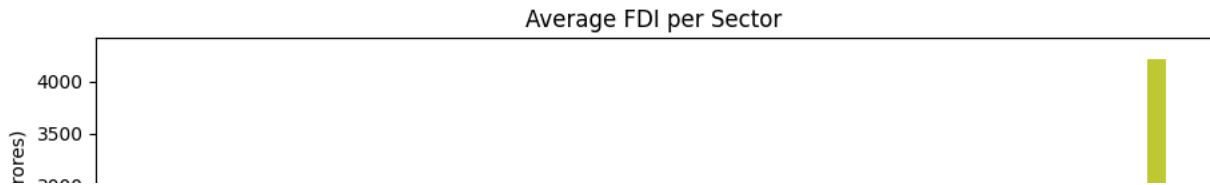
Sector-wise Average FDI

```
plt.figure(figsize=(18, 12))
plt.subplot(3, 2, 2)
sns.barplot(x=average_fdi_sectorwise.index, y=average_fdi_sectorwise.values, palette='viridis')
plt.title('Average FDI per Sector')
plt.xlabel('Sector')
plt.ylabel('Average FDI Inflow (in Crores)')
plt.tight_layout()
plt.show()
```


 <ipython-input-51-ca8fef5be86f>:3: FutureWarning:


Passing `palette` without assigning `hue` is deprecated and will be removed in v0.14.0. Assign the `x` variable to `hue` and set `le`

```
sns.barplot(x=average_fdi_sectorwise.index, y=average_fdi_sectorwise.values, palette='viridis')
```



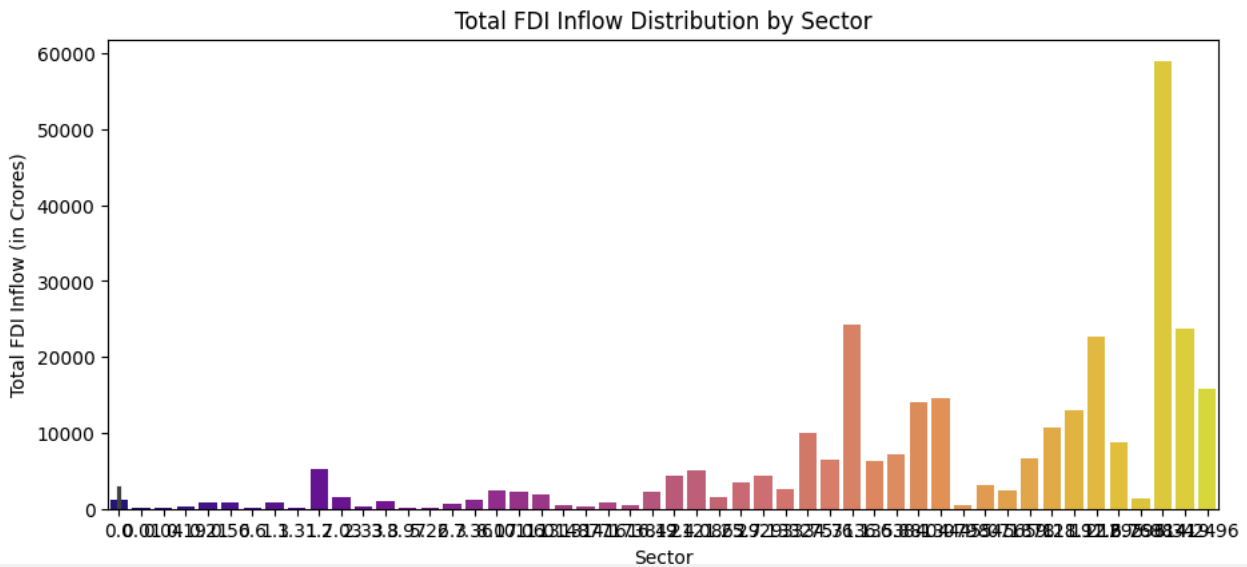
Total FDI by Sector

```
plt.figure(figsize=(18, 12))
plt.subplot(3, 2, 3)
sns.barplot(x=total_fdi_by_sector.index, y=total_fdi_by_sector.values, palette='plasma')
plt.title('Total FDI Inflow Distribution by Sector')
plt.xlabel('Sector')
plt.ylabel('Total FDI Inflow (in Crores)')
plt.tight_layout()
plt.show()
```

 <ipython-input-52-923dd591038a>:3: FutureWarning:


Passing `palette` without assigning `hue` is deprecated and will be removed in v0.14.0. Assign the `x` variable to `hue` and set `le`

```
sns.barplot(x=total_fdi_by_sector.index, y=total_fdi_by_sector.values, palette='plasma')
```



Total FDI by Year

```
plt.figure(figsize=(18, 12))
plt.subplot(3, 2, 4)
sns.barplot(x=total_fdi_by_year.index, y=total_fdi_by_year.values, palette='magma')
plt.title('Total FDI Inflow Distribution by Year')
plt.xlabel('Year')
plt.ylabel('Total FDI Inflow (in Crores)')
plt.xticks(rotation=45)
plt.tight_layout()
plt.show()
```

 <ipython-input-53-b7bf51451a45>:3: FutureWarning:

Passing `palette` without assigning `hue` is deprecated and will be removed in v0.14.0. Assign the `x` variable to `hue` and set `le`