

Foreign Direct Investment Analysis

Importing required Libraries

```
In [1]: import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
import seaborn as sns
```

```
In [2]: FDI=pd.read_csv(r"C:\Users\kruna\OneDrive\Desktop\internship\FDI\FDI data (1).csv")
FDI.head()
```

Out[2]:

	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	359.34
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	684.39
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	707.04
3	NON-CONVENTIONAL ENERGY	0.00	0.00	1.70	4.14	1.27	1.35	2.44	58.82	125.88	622.52	214.40	452.17	1106.52	414.25	615.95
4	COAL PRODUCTION	0.00	0.00	0.00	0.04	0.00	9.14	1.30	14.08	0.22	0.00	0.00	0.00	0.00	2.96	0.00

```
In [3]: FDI.columns
```

```
Out[3]: 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')
```

Extracting Detailed Information

```
In [4]: FDI.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
```

Checking the null Values

```
In [5]: FDI.isnull().sum()
```

```
Out[5]: 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 in data

```
In [6]: FDI.duplicated().sum()
```

```
In [20]: FDI.duplicated().sum()

Out[20]: 0

In [13]: FDI.columns.isnull()

Out[13]: array([False, False, False, False, False, False, False, False, False,
        False, False, False, False, False, False, False])

make Sector & Yearwise investment

In [17]: sector = ['Sector']
year = ['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']
```

```
In [18]: FDI.describe().round(2)
```

	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
count	63.00	63.00	63.00	63.00	63.00	63.00	63.00	63.00	63.00	63.00	63.00	63.00	63.00	63.00	63.00	63.00	63.00
mean	37.76	63.93	42.93	34.73	51.09	87.93	198.28	390.09	498.35	410.07	339.41	557.47	355.93	385.70	490.96	634.00	634.00
std	112.23	157.88	86.61	67.65	101.93	206.44	686.78	1026.25	1134.65	926.81	627.14	1031.47	778.09	658.43	837.79	1335.00	1335.00
min	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	0.00	0.00	0.00
25%	0.00	0.00	0.20	0.22	0.72	1.23	4.16	9.95	11.95	7.88	8.43	22.72	15.12	16.61	33.80	30.00	30.00
50%	4.03	5.07	11.01	6.37	9.09	22.62	25.82	58.82	84.88	69.74	58.07	129.36	95.41	113.78	177.22	159.00	159.00
75%	23.51	44.83	36.56	38.66	43.20	63.86	108.32	279.27	383.32	341.60	304.28	593.52	288.02	473.06	595.39	519.00	519.00
max	832.07	873.23	419.96	368.32	527.90	1359.97	4713.78	6986.17	6183.49	5466.13	3296.09	5215.98	4832.98	3982.89	4443.26	6889.00	6889.00

Sort vlaues in Numerial values in this dataset are in US\$ Millions Data is of yearwise investment in all sectors

```
In [23]: FDI_1 = pd.melt(FDI,id_vars = sector, value_vars = year, var_name='year', value_name = 'FDI in US$ Millions')
         FDI_1

Out[23]:
```

	Sector	year	FDI in US\$ Millions
0	METALLURGICAL INDUSTRIES	2000-01	22.69
1	MINING	2000-01	1.32
2	POWER	2000-01	89.42
3	NON-CONVENTIONAL ENERGY	2000-01	0.00
4	COAL PRODUCTION	2000-01	0.00
...
1066	PRINTING OF BOOKS (INCLUDING LITHO PRINTING IN...	2016-17	53.17
1067	COIR	2016-17	0.00
1068	CONSTRUCTION (INFRASTRUCTURE) ACTIVITIES	2016-17	1860.73
1069	CONSTRUCTION DEVELOPMENT: Townships, housing, ...	2016-17	105.14
1070	MISCELLANEOUS INDUSTRIES	2016-17	296.40

1071 rows × 3 columns

```
In [24]: sorted = FDI_1.sort_values(['Sector','year'])
         sorted

Out[24]:
```

	Sector	year	FDI in US\$ Millions
20	AGRICULTURAL MACHINERY	2000-01	3.64
83	AGRICULTURAL MACHINERY	2001-02	1.04
146	AGRICULTURAL MACHINERY	2002-03	13.48
209	AGRICULTURAL MACHINERY	2003-04	47.54
272	AGRICULTURAL MACHINERY	2004-05	0.00
...
794	VEGETABLE OILS AND VANASPATI	2012-13	108.39
857	VEGETABLE OILS AND VANASPATI	2013-14	21.55
920	VEGETABLE OILS AND VANASPATI	2014-15	148.34
983	VEGETABLE OILS AND VANASPATI	2015-16	34.22
1046	VEGETABLE OILS AND VANASPATI	2016-17	108.45

1071 rows × 3 columns

Description of sectors in new format

```
In [26]: FDI_1.describe().round(2)

Out[26]:
```

FDI in US\$ Millions	
count	1071.00
mean	309.98
std	819.04
min	0.00
25%	3.14
50%	37.94
75%	213.74
max	8684.07

Rename sectors

```
In [27]: sorted = sorted.replace(["CONSTRUCTION DEVELOPMENT: Townships, housing, built-up infrastructure and constructio
                                ,"SERVICES SECTOR (Fin.,Banking,Insurance,Non Fin/Business,Outsourcing,R&D,Courie
                                ,'TEA AND COFFEE (PROCESSING & WAREHOUSING COFFEE & RUBBER)']
                                ,["CONSTRUCTION DEVELOPMENT","SERVICES SECTOR",'TEA AND COFFEE'])
```

sector wise total FDI from 2000-17

```
In [28]: total_fdi_sector = sorted.groupby('Sector').sum().sort_values(by = 'FDI in US$ Millions',ascending = False)
total_fdi_sector
```

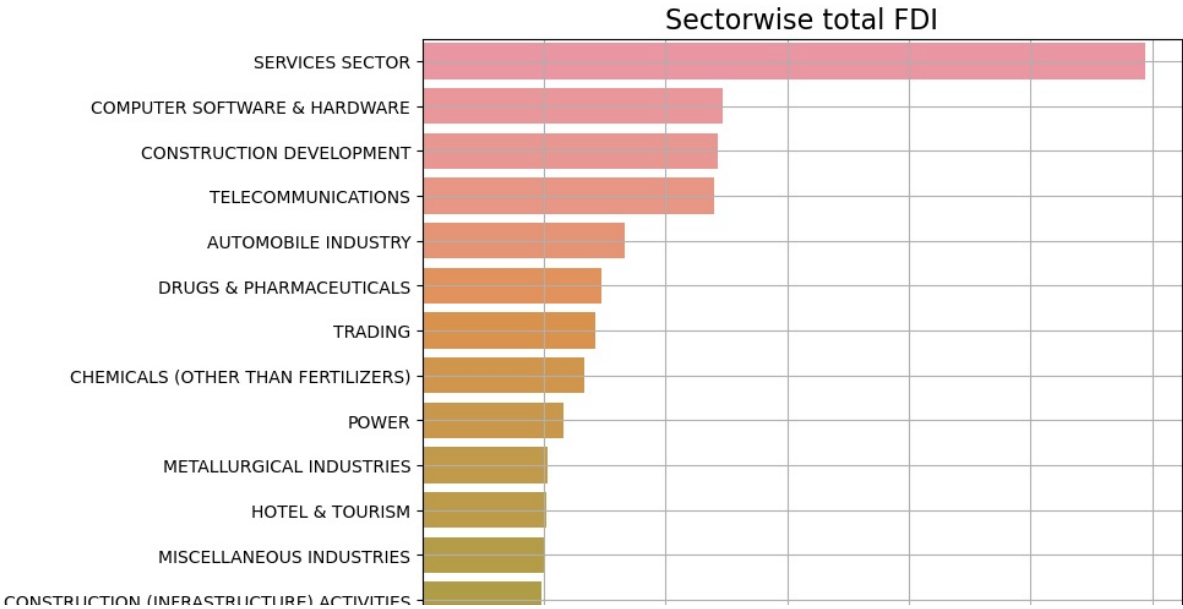
```
Out[28]:
```

	year	FDI in US\$ Millions
Sector		
SERVICES SECTOR	2000-012001-022002-032003-042004-052005-062006...	59476.49
COMPUTER SOFTWARE & HARDWARE	2000-012001-022002-032003-042004-052005-062006...	24669.49
CONSTRUCTION DEVELOPMENT	2000-012001-022002-032003-042004-052005-062006...	24293.09
TELECOMMUNICATIONS	2000-012001-022002-032003-042004-052005-062006...	23946.01
AUTOMOBILE INDUSTRY	2000-012001-022002-032003-042004-052005-062006...	16673.92
...
PHOTOGRAPHIC RAW FILM AND PAPER	2000-012001-022002-032003-042004-052005-062006...	67.28
COAL PRODUCTION	2000-012001-022002-032003-042004-052005-062006...	27.74
MATHEMATICAL,SURVEYING AND DRAWING INSTRUMENTS	2000-012001-022002-032003-042004-052005-062006...	7.98
DEFENCE INDUSTRIES	2000-012001-022002-032003-042004-052005-062006...	5.12
COIR	2000-012001-022002-032003-042004-052005-062006...	4.06

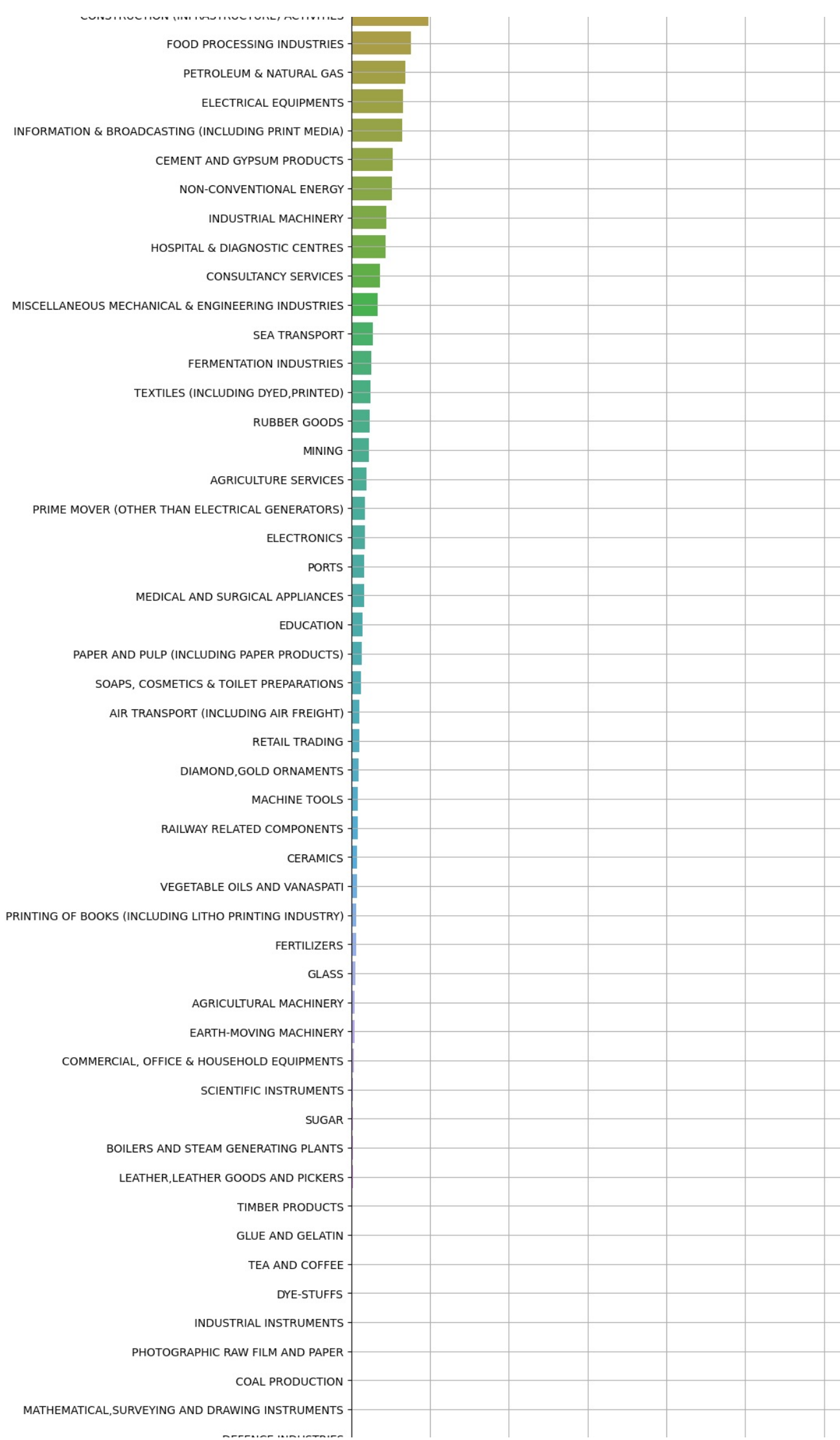
63 rows × 2 columns

Visualisation of sectorwise total FDI

```
In [29]: plt.figure(figsize=(8,30))
sns.barplot(x='FDI in US$ Millions',y = total_fdi_sector.index ,data = total_fdi_sector)
plt.xlabel('FDI in US$ Millions',fontsize=13)
plt.ylabel('Sectors',fontsize=13)
plt.title('Sectorwise total FDI',fontsize=16)
plt.grid()
plt.show()
```



Sectors





Best and worst performing sectors

```
In [50]: top_10 = total_fdi_sector.head(10)
top_10
```

Out[50]:

	year	FDI in US\$ Millions
Sector		
SERVICES SECTOR	2000-012001-022002-032003-042004-052005-062006...	59476.49
COMPUTER SOFTWARE & HARDWARE	2000-012001-022002-032003-042004-052005-062006...	24669.49
CONSTRUCTION DEVELOPMENT	2000-012001-022002-032003-042004-052005-062006...	24293.09
TELECOMMUNICATIONS	2000-012001-022002-032003-042004-052005-062006...	23946.01
AUTOMOBILE INDUSTRY	2000-012001-022002-032003-042004-052005-062006...	16673.92
DRUGS & PHARMACEUTICALS	2000-012001-022002-032003-042004-052005-062006...	14706.90
TRADING	2000-012001-022002-032003-042004-052005-062006...	14210.88
CHEMICALS (OTHER THAN FERTILIZERS)	2000-012001-022002-032003-042004-052005-062006...	13293.09
POWER	2000-012001-022002-032003-042004-052005-062006...	11589.13
METALLURGICAL INDUSTRIES	2000-012001-022002-032003-042004-052005-062006...	10330.54

check worst 10 performing secotrs

```
In [54]: least_10 = total_fdi_sector.sort_values(by='FDI in US$ Millions', ascending=True).head(10)
least_10
```

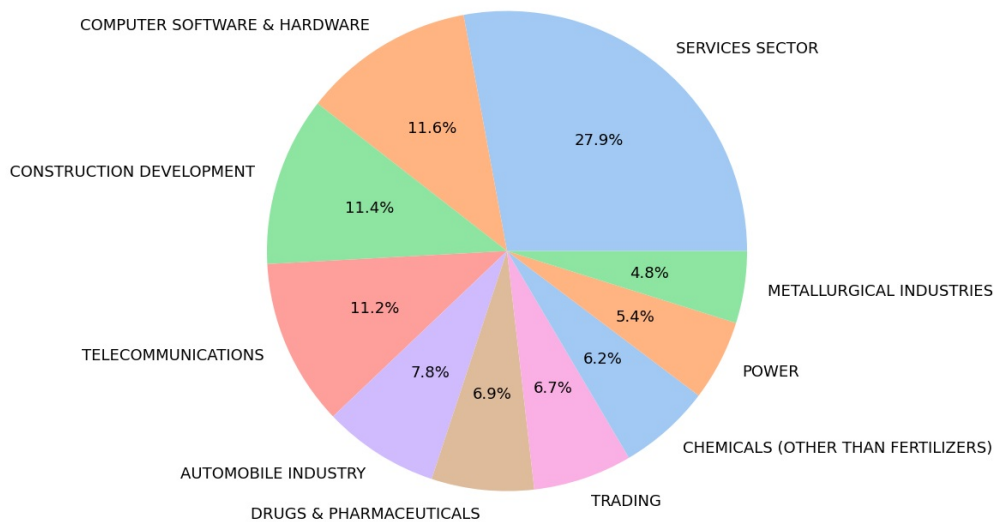
Out[54]:

	year	FDI in US\$ Millions
Sector		
COIR	2000-012001-022002-032003-042004-052005-062006...	4.06
DEFENCE INDUSTRIES	2000-012001-022002-032003-042004-052005-062006...	5.12
MATHEMATICAL,SURVEYING AND DRAWING INSTRUMENTS	2000-012001-022002-032003-042004-052005-062006...	7.98
COAL PRODUCTION	2000-012001-022002-032003-042004-052005-062006...	27.74
PHOTOGRAPHIC RAW FILM AND PAPER	2000-012001-022002-032003-042004-052005-062006...	67.28
INDUSTRIAL INSTRUMENTS	2000-012001-022002-032003-042004-052005-062006...	76.12
DYE-STUFFS	2000-012001-022002-032003-042004-052005-062006...	88.40
TEA AND COFFEE	2000-012001-022002-032003-042004-052005-062006...	111.22
GLUE AND GELATIN	2000-012001-022002-032003-042004-052005-062006...	128.39
TIMBER PRODUCTS	2000-012001-022002-032003-042004-052005-062006...	157.68

pie chart to visualise percentage share of FDI among top 10 sectors

```
In [51]: plt.figure(figsize=(20,8))
colors = sns.color_palette('pastel')[0:7]
textprops = {'fontSize':13}
plt.pie(top_10['FDI in US$ Millions'],labels=top_10.index,colors=colors, autopct="%1.1f%%",textprops = textprop
plt.axis('equal')
plt.title('SHARE AMONG TOP 10 SECTORS', fontsize = 20)
plt.show()
```

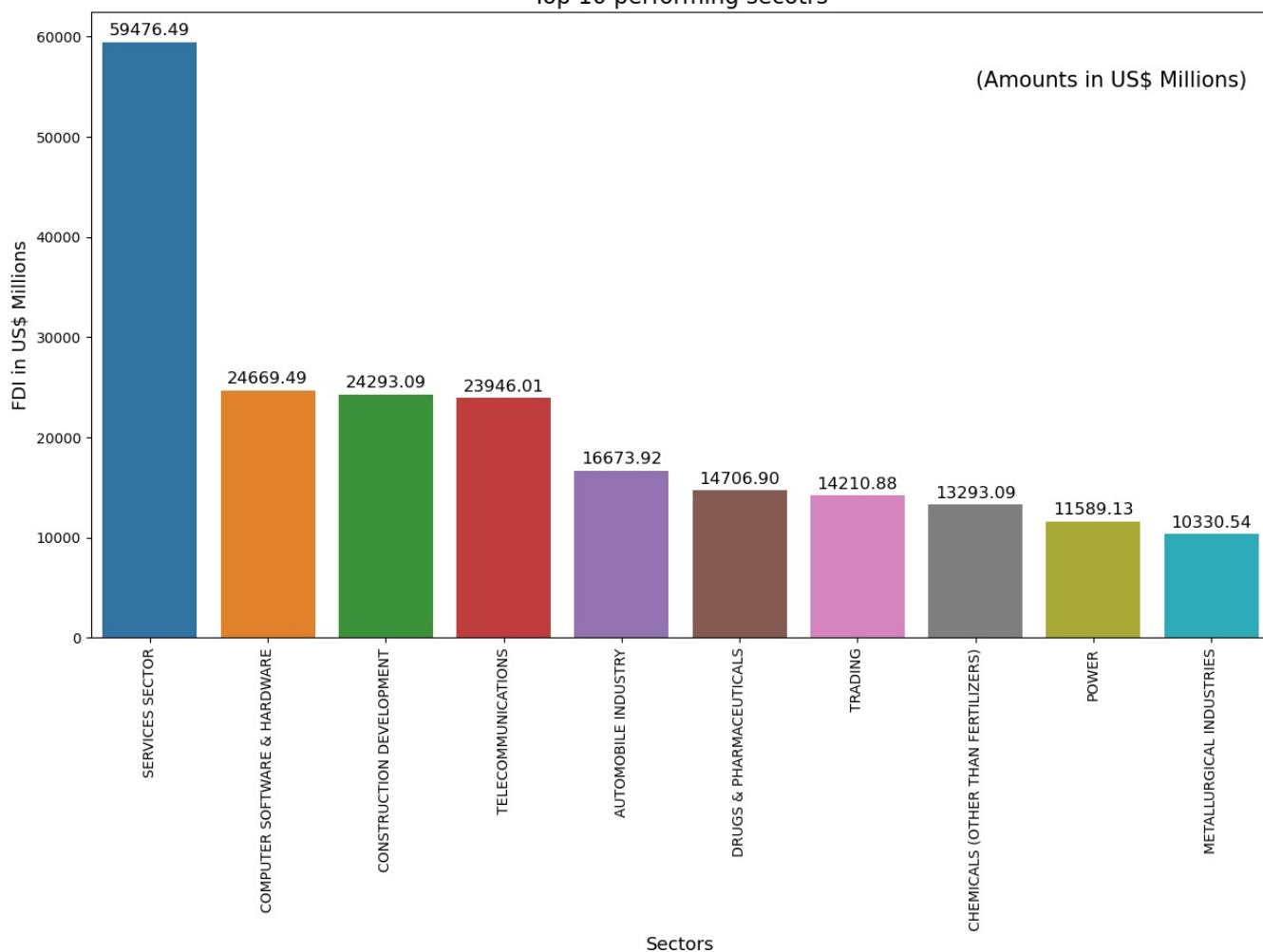
SHARE AMONG TOP 10 SECTORS



visualise Top 10 performing secotrs

```
In [52]: plt.figure(figsize=(15,8))
plots = sns.barplot(x = top_10.index, y='FDI in US$ Millions' ,data = top_10)
for bar in plots.patches:
    plots.annotate(format(bar.get_height(),'.2f'),
                   (bar.get_x() + bar.get_width() / 2,
                    bar.get_height()), ha='center', va='center',
                   size=12, xytext=(0, 8),
                   textcoords='offset points')
plt.xlabel('Sectors',fontsize=13)
plt.ylabel('FDI in US$ Millions',fontsize=13)
plt.xticks(rotation = 90)
plt.text(7,55000, "(Amounts in US$ Millions)", fontsize = 15)
plt.title('Top 10 performing secotrs',fontsize=16)
plt.show()
```

Top 10 performing secotrs



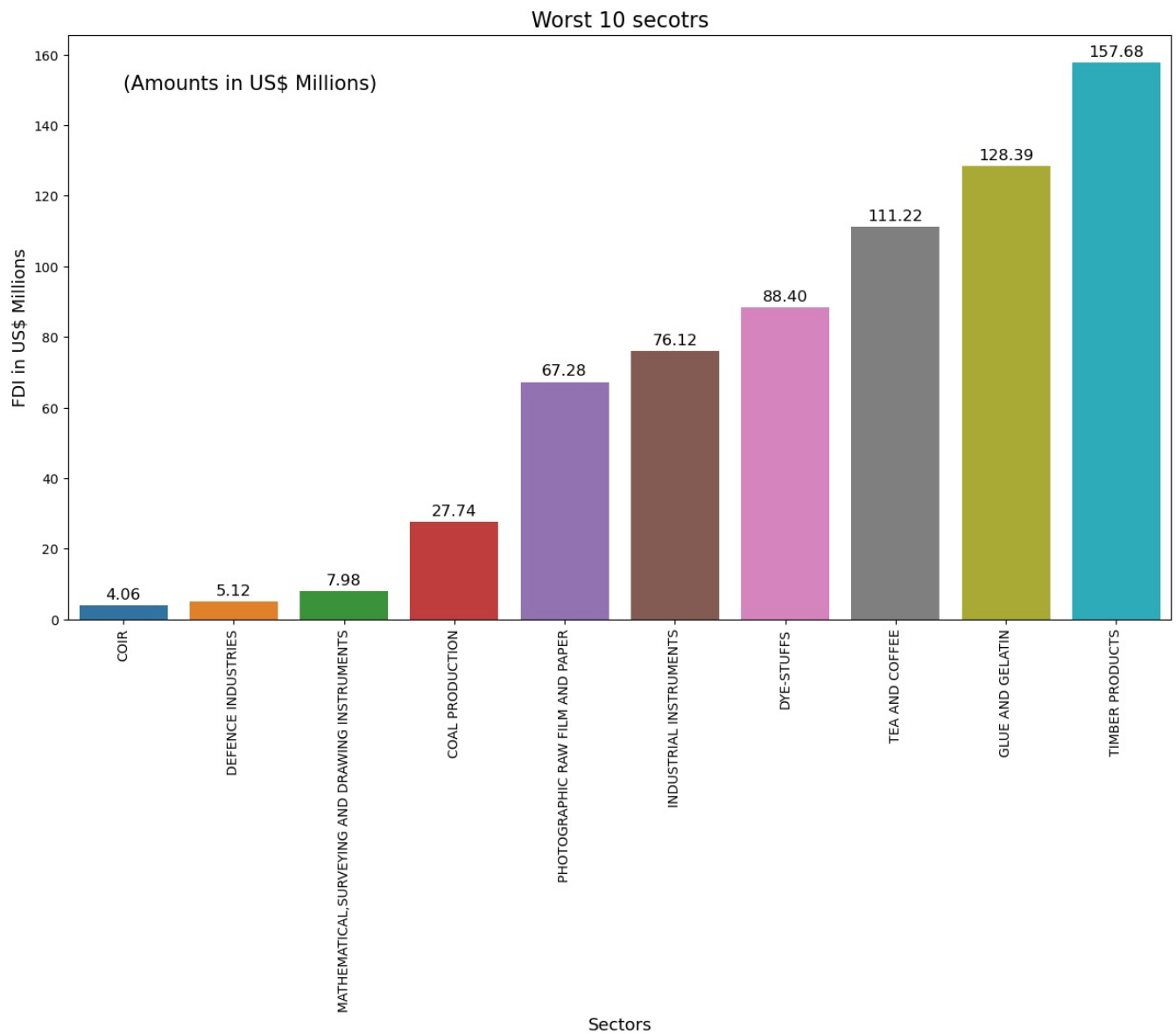
visualise worst 10 performing secotrs

```
In [56]: plt.figure(figsize=(15,8))
plots = sns.barplot(x = least_10.index, y='FDI in US$ Millions' ,data = least_10)
```

```

for bar in plots.patches:
    plots.annotate(format(bar.get_height(),'.2f'),
                   (bar.get_x() + bar.get_width() / 2,
                    bar.get_height()), ha='center', va='center',
                   size=12, xytext=(0, 8),
                   textcoords='offset points')
plt.xlabel('Sectors',fontsize=13)
plt.ylabel('FDI in US$ Millions',fontsize=13)
plt.xticks(rotation = 90)
plt.text(0,150, "(Amounts in US$ Millions)", fontsize = 15)
plt.title('Worst 10 secotrs',fontsize=16)
plt.show()

```



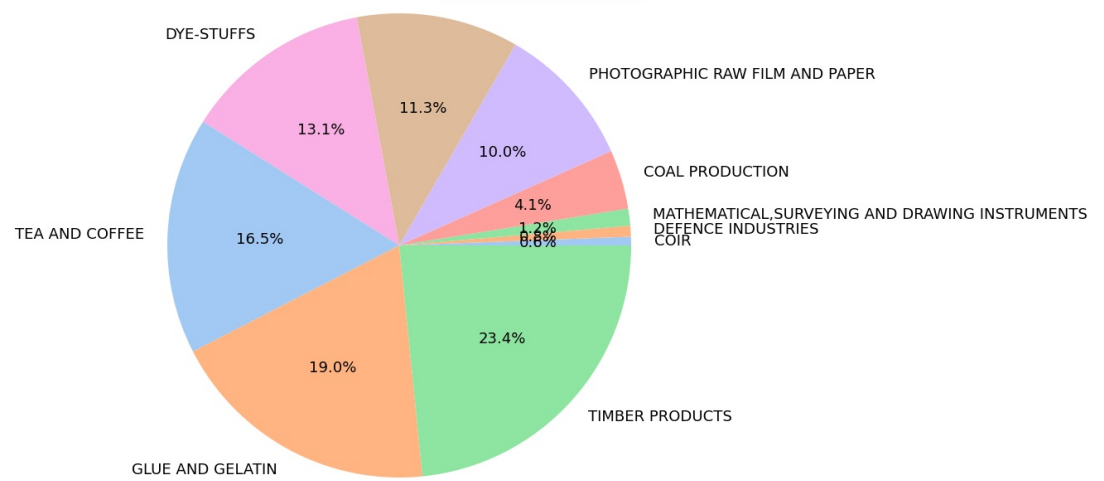
pie chart to visualise percentage share of FDI among worst 10 sectors

```

In [57]: plt.figure(figsize=(20,8))
         colors = sns.color_palette('pastel')[0:7]
         textprops = {'fontsize':13}
         plt.pie(least_10['FDI in US$ Millions'],labels=least_10.index,colors=colors, autopct="%1.1f%%",textprops = text
         plt.axis('equal')
         plt.title('SHARE AMONG WORST 10 SECTORS', fontsize = 20)
         plt.show()

```

SHARE AMONG WORST 10 SECTORS



FDI Inflow Yearwise

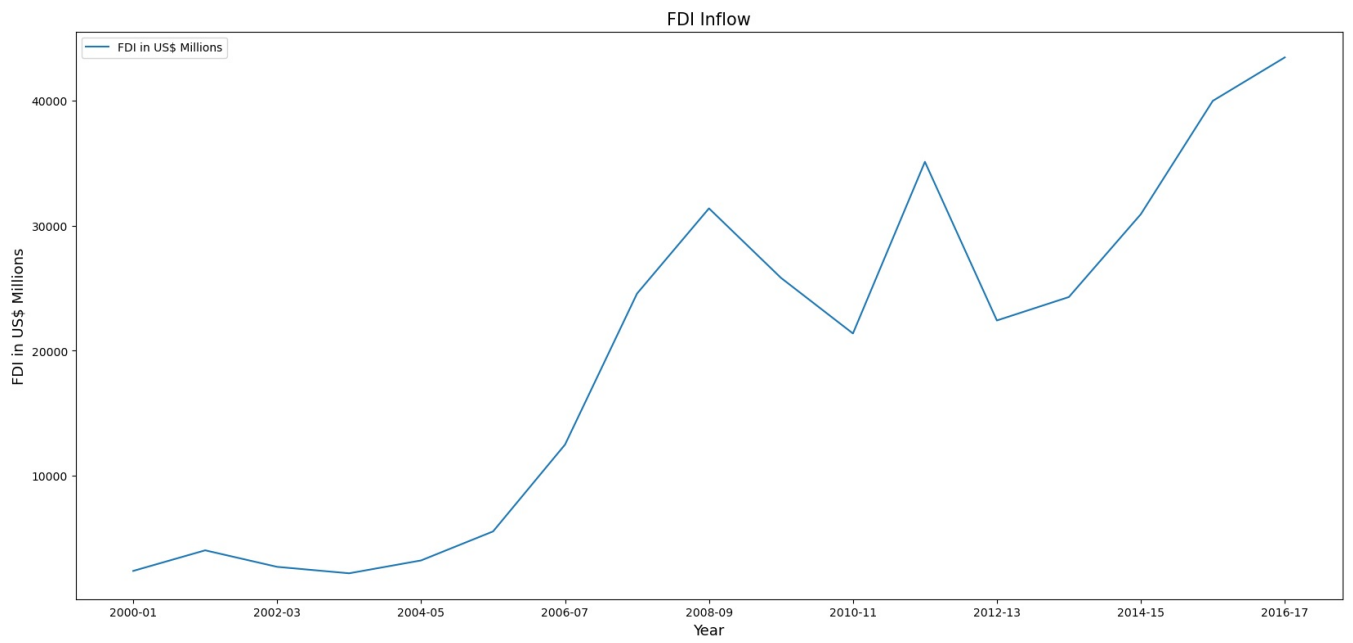
```
In [59]: year_inflow = FDI_1[['year', 'FDI in US$ Millions']]
year_inflow = year_inflow.groupby('year').sum()
year_inflow['% growth over previous year'] = round(year_inflow.pct_change()*100,2)
year_inflow
```

```
Out[59]:
```

	FDI in US\$ Millions	% growth over previous year
year		
2000-01	2378.71	NaN
2001-02	4027.69	69.32
2002-03	2704.32	-32.86
2003-04	2187.85	-19.10
2004-05	3218.69	47.12
2005-06	5539.75	72.11
2006-07	12491.76	125.49
2007-08	24575.40	96.73
2008-09	31395.96	27.75
2009-10	25834.38	-17.71
2010-11	21383.07	-17.23
2011-12	35120.78	64.25
2012-13	22423.59	-36.15
2013-14	24299.32	8.36
2014-15	30930.47	27.29
2015-16	40000.99	29.33
2016-17	43478.26	8.69

plotting to show Year by Year FDI Inflow

```
In [60]: year_inflow.plot.line(y='FDI in US$ Millions',figsize = (20,9))
plt.xlabel('Year', fontsize = 13)
plt.ylabel('FDI in US$ Millions', fontsize = 13)
plt.title('FDI Inflow', fontsize = 15)
plt.show()
```

In []: 2000-01 upto 2005-06 there is not much inflow in India.

In []: 2000-01 upto 2005-06 there is not much inflow in India.

In []: 2008-09 to 2010-11 some downfall

In []: after 2012-13 come back flow in India.

In []: overall there is growth in FDI inflows.

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