

# fdi-project

May 3, 2023

## FDI Analytics

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### Problem Statement:

Investment is a game of understanding historic data of investment objects under different events but it is still a game of chances to minimize the risk we apply analytics to find the equilibrium investment. To understand the Foreign direct investment in India for the last 17 years from 2000-01 to 2016-17. This dataset contains sector and financial year-wise data of FDI in India Sector-wise investment analysis Year-wise investment analysis. Find key metrics and factors and show the meaningful relationships between attributes. Do your own research and come up with your findings

```
[ ]: import pandas as pd
import numpy as np
import seaborn as sns
import matplotlib.pyplot as plt
import warnings
warnings.filterwarnings("ignore")
import re
```

```
[ ]: data = pd.read_csv("/content/FDI data.csv") #,index_col='Sector')
```

```
[ ]: data.head()
```

```
[ ]:
```

	Sector	2000-01	2001-02	2002-03	2003-04	2004-05	\			
0	METALLURGICAL INDUSTRIES	22.69	14.14	36.61	8.11	200.38				
1	MINING	1.32	6.52	10.06	23.48	9.92				
2	POWER	89.42	757.44	59.11	27.09	43.37				
3	NON-CONVENTIONAL ENERGY	0.00	0.00	1.70	4.14	1.27				
4	COAL PRODUCTION	0.00	0.00	0.00	0.04	0.00				
		2005-06	2006-07	2007-08	2008-09	2009-10	2010-11	2011-12	2012-13	\
0		149.13	169.94	1175.75	959.94	419.88	1098.14	1786.14	1466.23	
1		7.40	6.62	444.36	34.16	174.40	79.51	142.65	57.89	
2		72.69	157.15	988.68	907.66	1271.79	1271.77	1652.38	535.68	
3		1.35	2.44	58.82	125.88	622.52	214.40	452.17	1106.52	
4		9.14	1.30	14.08	0.22	0.00	0.00	0.00	0.00	

	2013-14	2014-15	2015-16	2016-17
0	567.63	359.34	456.31	1440.18
1	12.73	684.39	520.67	55.75
2	1066.08	707.04	868.80	1112.98
3	414.25	615.95	776.51	783.57
4	2.96	0.00	0.00	0.00

```
[ ]: data.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')
```

```
[ ]: years = ['2000', '2001', '2002', '2003', '2004', '2005',
              '2006', '2007', '2008', '2009', '2010', '2011',
              '2012', '2013', '2014', '2015', '2016']
```

```
[ ]: data.columns= ['Sector']+years
```

```
[ ]: data.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    63 non-null        float64
2   2001    63 non-null        float64
3   2002    63 non-null        float64
4   2003    63 non-null        float64
5   2004    63 non-null        float64
6   2005    63 non-null        float64
7   2006    63 non-null        float64
8   2007    63 non-null        float64
9   2008    63 non-null        float64
10  2009    63 non-null        float64
11  2010    63 non-null        float64
12  2011    63 non-null        float64
13  2012    63 non-null        float64
14  2013    63 non-null        float64
15  2014    63 non-null        float64
16  2015    63 non-null        float64
17  2016    63 non-null        float64
dtypes: float64(17), object(1)
memory usage: 9.0+ KB
```

```
[ ]: print(data.nunique())
```

```
Sector    63
2000      41
2001      46
2002      51
2003      53
2004      58
2005      57
2006      56
2007      61
2008      59
2009      59
2010      61
2011      60
2012      59
2013      61
2014      62
2015      59
2016      58
dtype: int64
```

```
[ ]: data.describe()
```

```
[ ]:
```

	2000	2001	2002	2003	2004 \
count	63.000000	63.000000	63.000000	63.000000	63.000000
mean	37.757302	63.931587	42.925714	34.727778	51.090317
std	112.227860	157.878737	86.606439	67.653735	101.934873
min	0.000000	0.000000	0.000000	0.000000	0.000000
25%	0.000000	0.000000	0.200000	0.215000	0.715000
50%	4.030000	5.070000	11.010000	6.370000	9.090000
75%	23.510000	44.830000	36.555000	38.660000	43.205000
max	832.070000	873.230000	419.960000	368.320000	527.900000

	2005	2006	2007	2008	2009 \
count	63.000000	63.000000	63.000000	63.000000	63.000000
mean	87.932540	198.281905	390.085714	498.348571	410.069524
std	206.436967	686.783115	1026.249935	1134.649040	926.814626
min	0.000000	0.000000	0.000000	0.000000	0.000000
25%	1.230000	4.160000	9.950000	11.950000	7.880000
50%	22.620000	25.820000	58.820000	84.880000	69.740000
75%	63.855000	108.325000	279.270000	383.320000	341.595000
max	1359.970000	4713.780000	6986.170000	6183.490000	5466.130000

	2010	2011	2012	2013	2014 \
count	63.000000	63.000000	63.000000	63.000000	63.000000
mean	339.413810	557.472698	355.930000	385.703492	490.959841

std	627.141139	1031.474056	778.091368	658.429944	837.787060
min	0.000000	0.000000	0.000000	0.000000	0.000000
25%	8.430000	22.720000	15.115000	16.610000	33.800000
50%	58.070000	129.360000	95.410000	113.780000	177.220000
75%	304.280000	593.525000	288.025000	473.060000	595.390000
max	3296.090000	5215.980000	4832.980000	3982.890000	4443.260000

	2015	2016
count	63.000000	63.000000
mean	634.936349	690.131111
std	1335.307706	1411.965354
min	0.000000	0.000000
25%	30.000000	19.905000
50%	159.130000	110.860000
75%	519.070000	741.220000
max	6889.460000	8684.070000

```
[ ]: data['Sector_total'] = data[years].sum(axis=1)
data.head()
```

```
[ ]:
      Sector  2000  2001  2002  2003  2004  2005 \
0  METALLURGICAL INDUSTRIES  22.69  14.14  36.61  8.11  200.38  149.13
1                MINING      1.32   6.52  10.06  23.48   9.92   7.40
2                POWER    89.42  757.44  59.11  27.09  43.37  72.69
3  NON-CONVENTIONAL ENERGY   0.00   0.00   1.70   4.14   1.27   1.35
4                COAL PRODUCTION  0.00   0.00   0.00   0.04   0.00   9.14
```

	2006	2007	2008	2009	2010	2011	2012	2013 \
0	169.94	1175.75	959.94	419.88	1098.14	1786.14	1466.23	567.63
1	6.62	444.36	34.16	174.40	79.51	142.65	57.89	12.73
2	157.15	988.68	907.66	1271.79	1271.77	1652.38	535.68	1066.08
3	2.44	58.82	125.88	622.52	214.40	452.17	1106.52	414.25
4	1.30	14.08	0.22	0.00	0.00	0.00	0.00	2.96

	2014	2015	2016	Sector_total
0	359.34	456.31	1440.18	10330.54
1	684.39	520.67	55.75	2271.83
2	707.04	868.80	1112.98	11589.13
3	615.95	776.51	783.57	5181.49
4	0.00	0.00	0.00	27.74

```
[ ]: annual_fdi = (data.loc[0 : ].sum(axis = 0))
annual_fdi
```

```
[ ]: Sector      METALLURGICAL INDUSTRIESMININGPOWERNON-CONVENT...
2000                                     2378.71
2001                                     4027.69
```

```

2002                2704.32
2003                2187.85
2004                3218.69
2005                5539.75
2006               12491.76
2007                24575.4
2008               31395.96
2009               25834.38
2010               21383.07
2011               35120.78
2012               22423.59
2013               24299.32
2014               30930.47
2015               40000.99
2016               43478.26
Sector_total       331990.99
dtype: object

```

```
[ ]: data = data.append(annual_fdi, ignore_index=True)
```

```
[ ]: data = data.replace({data.iloc[-1][0]: 'Annual_total'})
```

```
[ ]: data
```

```
[ ]:
```

	Sector	2000	2001 \
0	METALLURGICAL INDUSTRIES	22.69	14.14
1	MINING	1.32	6.52
2	POWER	89.42	757.44
3	NON-CONVENTIONAL ENERGY	0.00	0.00
4	COAL PRODUCTION	0.00	0.00
..	...	...	...
59	COIR	0.00	0.00
60	CONSTRUCTION (INFRASTRUCTURE) ACTIVITIES	0.00	0.00
61	CONSTRUCTION DEVELOPMENT: Townships, housing, ...	24.33	51.75
62	MISCELLANEOUS INDUSTRIES	832.07	221.37
63	Annual_total	2378.71	4027.69

	2002	2003	2004	2005	2006	2007	2008 \
0	36.61	8.11	200.38	149.13	169.94	1175.75	959.94
1	10.06	23.48	9.92	7.40	6.62	444.36	34.16
2	59.11	27.09	43.37	72.69	157.15	988.68	907.66
3	1.70	4.14	1.27	1.35	2.44	58.82	125.88
4	0.00	0.04	0.00	9.14	1.30	14.08	0.22
..	...	...	...	...	...	...	...
59	0.00	0.00	0.47	0.59	0.04	0.01	0.00
60	0.00	0.00	0.00	0.93	64.06	182.92	172.70
61	36.10	47.04	152.06	228.71	1392.95	3887.33	4657.51

```

62  218.76  235.48  121.83  164.76  304.87  528.42  1549.70
63  2704.32  2187.85  3218.69  5539.75  12491.76  24575.40  31395.96

```

```

      2009      2010      2011      2012      2013      2014      2015  \
0      419.88  1098.14  1786.14  1466.23  567.63  359.34  456.31
1      174.40    79.51   142.65    57.89   12.73  684.39  520.67
2     1271.79  1271.77  1652.38   535.68  1066.08  707.04  868.80
3      622.52   214.40   452.17  1106.52   414.25  615.95  776.51
4         0.00     0.00     0.00     0.00     2.96     0.00     0.00
..      ...      ...      ...      ...      ...      ...      ...
59      0.25     0.10     0.55     0.15     0.54     1.36     0.00
60     324.56   675.07   386.28   283.89   485.37   870.25  4510.71
61   5466.13  1663.03  3140.78  1332.49  1226.05   769.14   112.55
62   1147.56  1475.97   813.38   229.49   468.74   765.88   668.77
63  25834.38  21383.07  35120.78  22423.59  24299.32  30930.47  40000.99

```

```

      2016  Sector_total
0     1440.18    10330.54
1       55.75     2271.83
2     1112.98    11589.13
3       783.57     5181.49
4         0.00      27.74
..      ...      ...
59       0.00       4.06
60    1860.73     9817.47
61     105.14    24293.09
62     296.40    10043.45
63   43478.26    331990.99

```

[64 rows x 19 columns]

```

[ ]: def f(x, y, n):
      if x: return ((y/x)**(1/n))-1
      else: return np.nan
y_start = 2012
y_end = 2016
data['Cagr'] = data[['Sector',str(y_start),str(y_end)]].apply(lambda x: f(x[str(y_start)],x[str(y_end)], y_end-y_start), axis=1)

```

```

[ ]: data[['Sector','Cagr']].sort_values(by='Cagr',ascending=False)

```

```

[ ]:
45                CEMENT AND GYPSUM PRODUCTS  2.621761
34  PAPER AND PULP (INCLUDING PAPER PRODUCTS)  1.496162
54                RETAIL TRADING  1.120337
11                TELECOMMUNICATIONS  1.068563
8                ELECTRICAL EQUIPMENTS  0.837037

```

	...	...
4	COAL PRODUCTION	NaN
16	PORTS	NaN
30	PHOTOGRAPHIC RAW FILM AND PAPER	NaN
31	DYE-STUFFS	NaN
42	GLUE AND GELATIN	NaN

[64 rows x 2 columns]

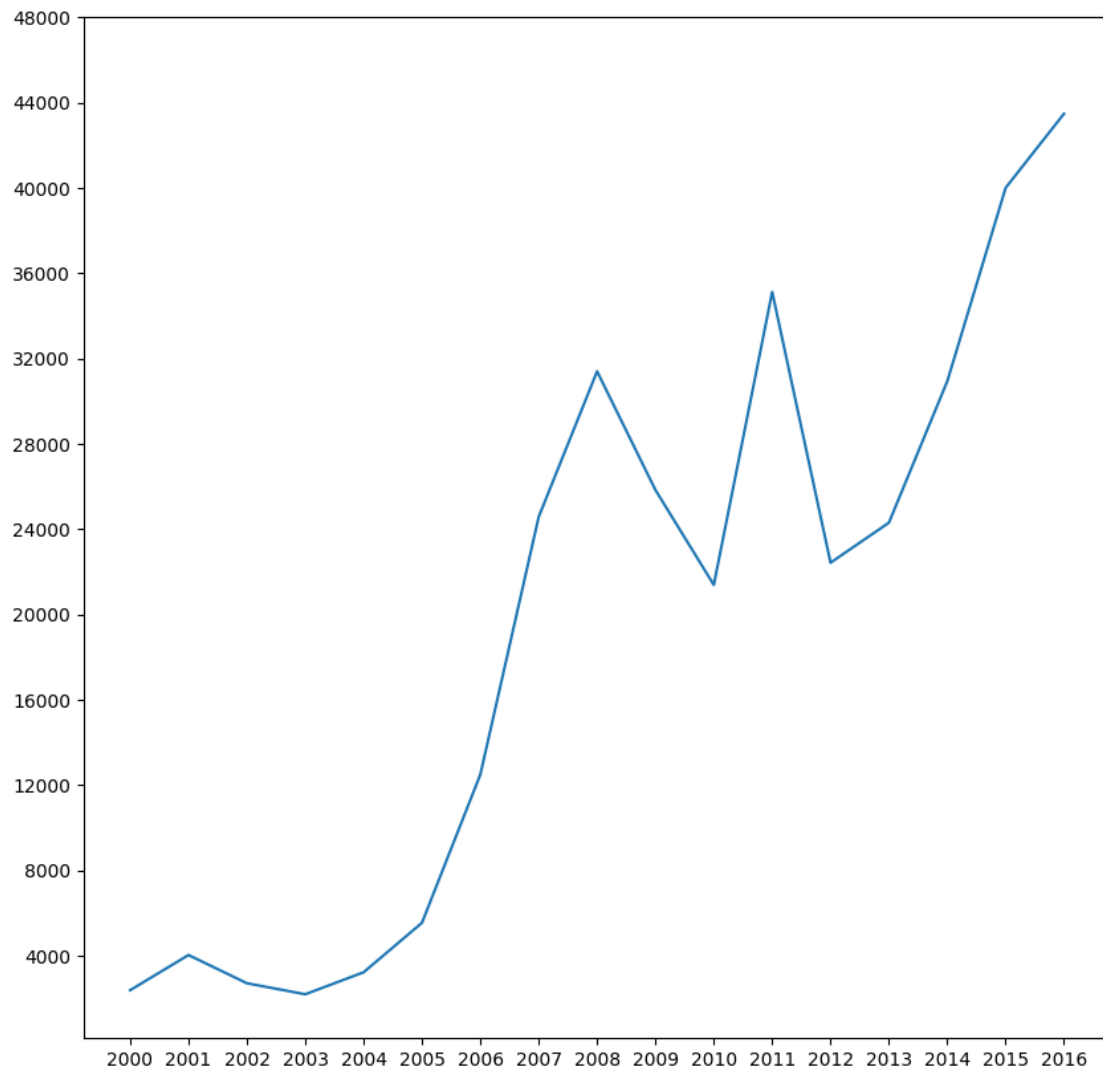
```
[ ]: # data['Sector'] = data['Sector'].replace(" \(.*\):", "")
```

```
[ ]: data1 = data.dropna()
```

```
[ ]: a = data.iloc[-1:,1:-2]
a = a.T
b = a.values.flatten()
c = a.index
```

```
[ ]: plt.figure(figsize=(10, 10))
sns.lineplot(x=c, y=b, markers=True)
plt.
↳yticks([4000,8000,12000,16000,20000,24000,28000,32000,36000,40000,44000,48000])
plt.xlabel = "Year"
plt.ylabel = "Annual FDI"
plt.title = 'Year Vs Annual FDI'

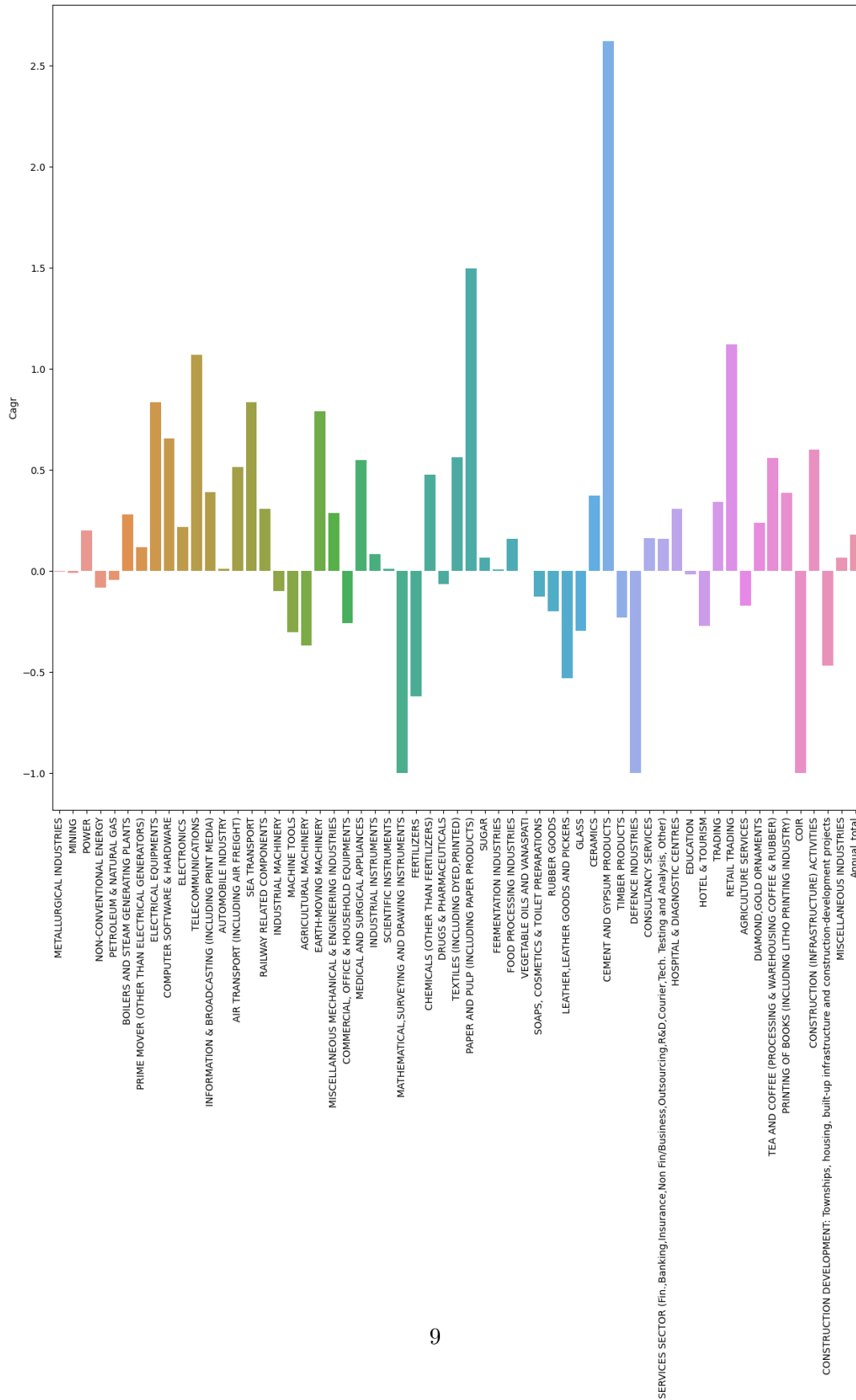
plt.show()
```



```
[ ]: plt.figure(figsize=(15, 15))
sns.barplot(x = data1['Sector'], y = data1['Cagr'])
plt.xticks(rotation=90)
plt.suptitle("Sector vs CAGR")
plt.show()
```

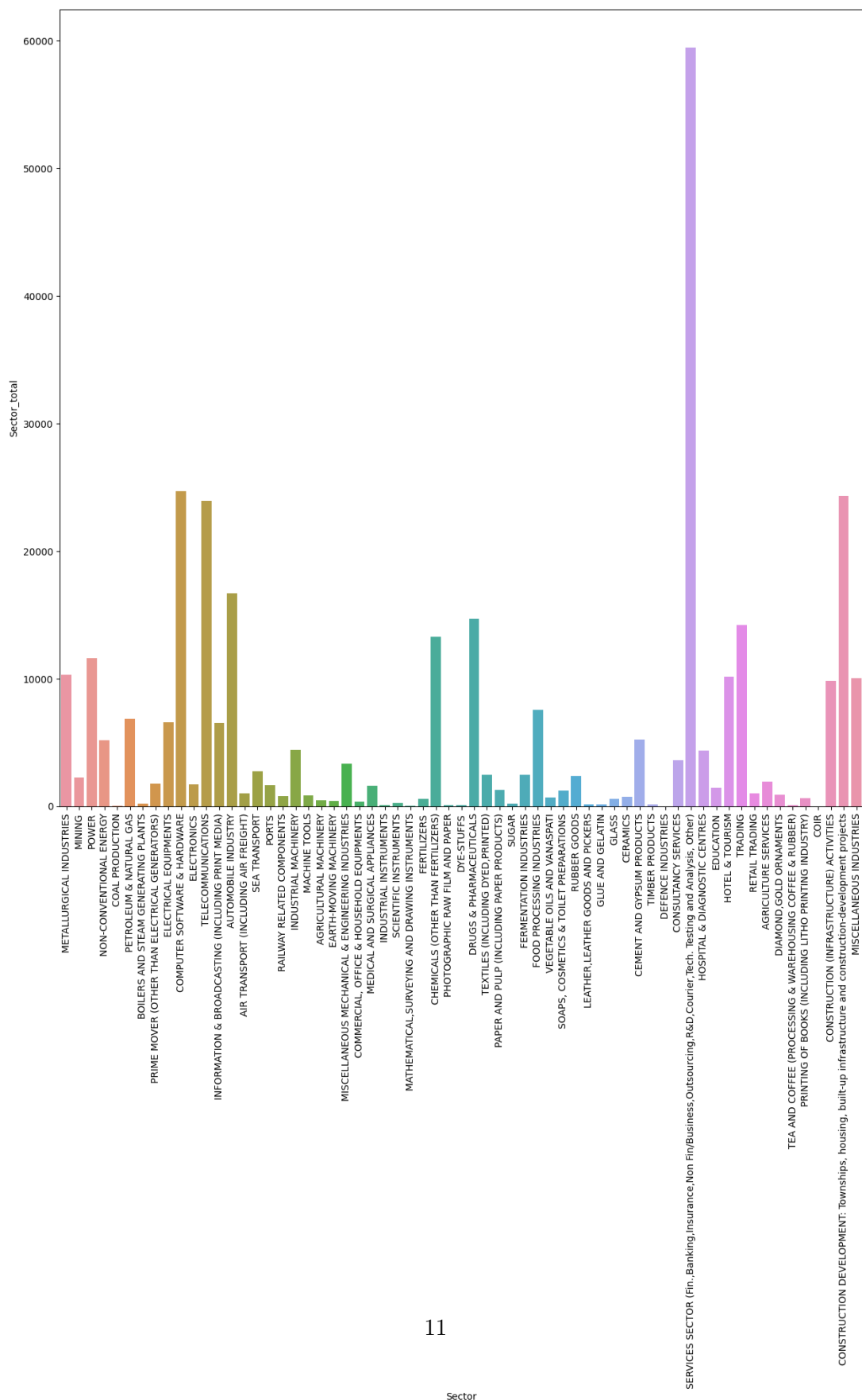


Sector vs CAGR



```
[ ]: plt.figure(figsize=(15, 15))
      sns.barplot(x = data['Sector'][:-1], y = data['Sector_total'][:-1])
      plt.xticks(rotation=90)
      plt.suptitle("Sector vs CAGR")
      plt.show()
```

# Sector vs CAGR



```
[ ]: # data.to_csv("/content/FDI data_1.csv")
```

```
[ ]: # data1.to_csv("/content/FDI data_2.csv")
```

```
[ ]: # final = {'Year': years,  
#             'Annual Total': a.values.flatten()}
```

```
[ ]: # q = pd.DataFrame(final)
```

```
[ ]: # q
```

```
[ ]: # q.to_csv("/content/FDI data_3.csv")
```

```
[ ]:
```