FDI data analysis

Wireframe

FDI DATA

Required libraries:

```
1. Pandas
```

75%

63.855000

108.325000

```
2.
     Matplotlib
#import libs
import pandas as pd
import matplotlib.pyplot as plt
#import data
df = pd.read_csv('FDI_in India.csv')
!ls
FDI in India.csv sample data
###Generate descriptive statistics.
df.describe()
          2000-01
                       2001-02
                                    2002-03
                                                 2003-04
                                                              2004-05
        63.000000
                     63.000000
                                  63.000000
                                               63.000000
                                                           63.000000
count
mean
        37.757302
                     63.931587
                                  42.925714
                                               34.727778
                                                           51.090317
                    157.878737
std
       112.227860
                                  86.606439
                                               67.653735
                                                          101.934873
         0.000000
                      0.000000
                                   0.000000
                                                0.000000
                                                            0.000000
min
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
       832.070000
                    873.230000
                                 419.960000
                                             368.320000
                                                          527.900000
max
           2005-06
                         2006-07
                                       2007-08
                                                     2008-09
                                                                   2009 - 10
count
         63.000000
                       63,000000
                                     63,000000
                                                   63,000000
                                                                 63,000000
         87.932540
                      198.281905
                                    390.085714
                                                  498.348571
                                                                410.069524
mean
        206.436967
                                   1026.249935
                                                 1134.649040
                                                                926.814626
std
                      686.783115
min
          0.000000
                        0.000000
                                      0.000000
                                                    0.000000
                                                                  0.000000
25%
          1.230000
                        4.160000
                                      9.950000
                                                   11.950000
                                                                  7.880000
         22.620000
                       25.820000
                                     58.820000
                                                   84.880000
                                                                 69.740000
50%
```

383.320000

341.595000

279.270000

max	1359.970000	4713.780000	6986.170000	6183.490000	5466.130000
\	2010-11	2011-12	2012-13	2013-14	2014-15
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
count mean std min 25% 50% 75% max	2015-16 63.000000 634.936349 1335.307706 0.000000 30.000000 159.130000 519.070000 6889.460000	2016-17 63.000000 690.131111 1411.965354 0.000000 19.905000 110.860000 741.220000 8684.070000			

###Output the shape and first 5 rows of the dataframe

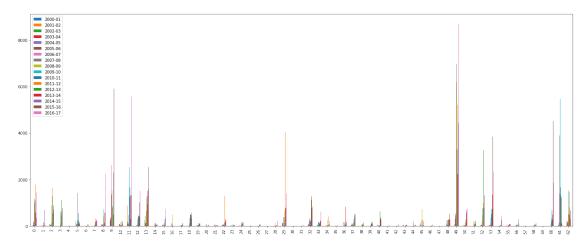
print(df.shape)
df.head()

(63, 18)

Sector	2000-01	2001-02	2002-03	2003-04	2004-
05 \ 0 METALLURGICAL INDUSTRIES 200.38	22.69	14.14	36.61	8.11	
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	

	2005-06	2006 - 07	2007-08	2008-09	2009-10	2010-11	2011-12
	12-13 \ 149.13	169.94	1175.75	050 04	/10 QQ	1098.14	1786.14
-	66.23	109.94	11/3./3	333.34	419.00	1090.14	1700.14
	7.40	6.62	444.36	34.16	174.40	79.51	142.65
_	. 89						
	72.69	157.15	988.68	907.66	1271.79	1271.77	1652.38
53	5.68						
3	1.35	2.44	58.82	125.88	622.52	214.40	452.17
11	06.52						
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			

<matplotlib.axes._subplots.AxesSubplot at 0x7f7dd4d4f610>



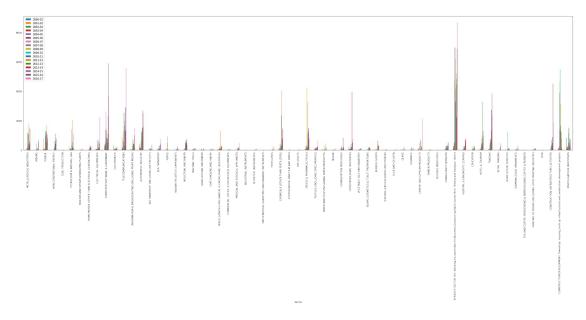
Setting the index

df.set_index('Sector',inplace=True)

df.plot(figsize=(25,10), kind='bar')

Plotting the data after setting the index df.plot(kind='bar', figsize = (40,10))

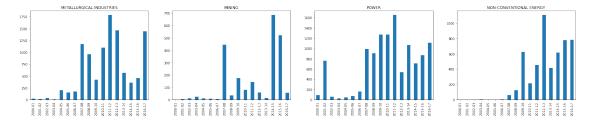
<matplotlib.axes._subplots.AxesSubplot at 0x7f7dd4100b90>



Plotting the first 4 rows

fig, axes = plt.subplots(nrows=1, ncols=4)

for i in range(0,4):
 df.iloc[i].plot(kind='bar',title=df.iloc[i].name, ax = axes[i],
figsize=(30,5))

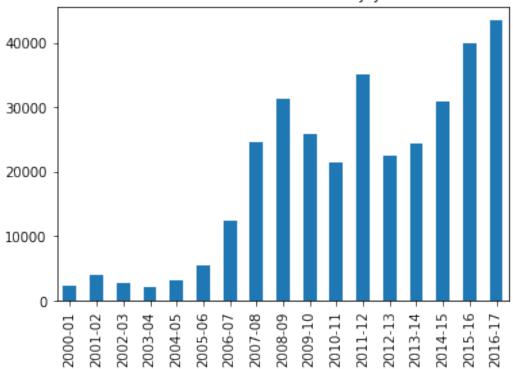


Looking at the sum of investments in all sectors by year

df.sum().plot(kind='bar', title='sum of all investments by year')

<matplotlib.axes._subplots.AxesSubplot at 0x7f7dd126a8d0>

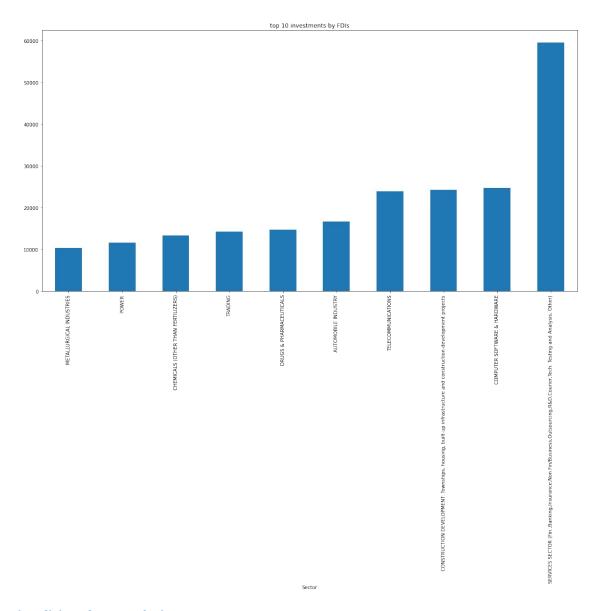
sum of all investments by year



Plotting the sum of invesments of all years by their sector (top 10 sectors)

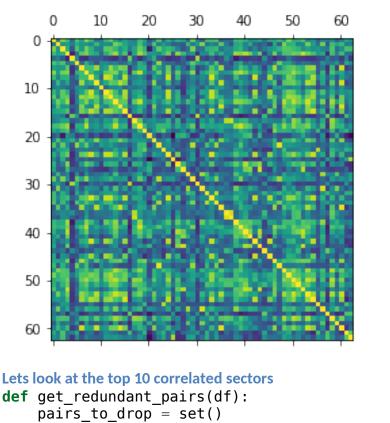
```
df_trans = df.transpose()
df_trans.sum().sort_values()[-10:].plot(figsize=(20,10),kind='bar',
title ='top 10 investments by FDIs')
```

<matplotlib.axes._subplots.AxesSubplot at 0x7f7dd1190c50>



Visualizing the correlation among sectors
corr = df_trans.corr()
plt.matshow(corr)

<matplotlib.image.AxesImage at 0x7f7dd10d4ed0>



Top Absolute Correlations

Sector Sector

MISCELLANEOUS MECHANICAL & ENGINEERING INDUSTRIES
0.958449
SUGAR CONSTRUCTION
(INFRASTRUCTURE) ACTIVITIES 0.937258
ELECTRICAL EQUIPMENTS TEXTILES (INCLUDING DYED, PRINTED) 0.926705

```
MEDICAL AND SURGICAL APPLIANCES
                                                         TEXTILES (INCLUDING
DYED, PRINTED)
                           0.919642
SEA TRANSPORT
                                                         RETAIL TRADING
0.918936
                                                         DIAMOND, GOLD
DYE-STUFFS
ORNAMENTS
                                   0.916723
AIR TRANSPORT (INCLUDING AIR FREIGHT)
                                                         CONSTRUCTION
(INFRASTRUCTURE) ACTIVITIES
                                   0.916622
FERMENTATION INDUSTRIES
                                                          FOOD PROCESSING
INDUSTRIES
                               0.910990
ELECTRICAL EQUIPMENTS
                                                          GLUE AND GELATIN
0.908833
MATHEMATICAL, SURVEYING AND DRAWING INSTRUMENTS
                                                         GLASS
0.908687
dtype: float64
Plotting the Correlated Sectors
abs corr = get top abs correlations(df_trans,10)
plt.rcParams['figure.figsize'] = [25, 5]
for i in range(10):
  plt.plot(df.loc[abs corr.index[i][0]], label = abs corr.index[i][0])
  plt.plot(df.loc[abs corr.index[i][1]], label = abs corr.index[i][1])
  plt.title(f'Correleation between {abs corr.index[i][0]} and
{abs corr.index[i][1]}')
  plt.legend()
  plt.show()

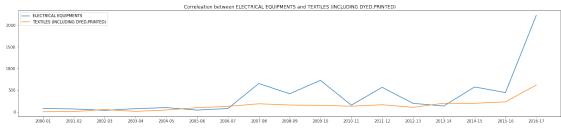
    MISCELLANEOUS MECHANICAL & ENGINEERING INDUSTRIES
    DEFENCE INDUSTRIES

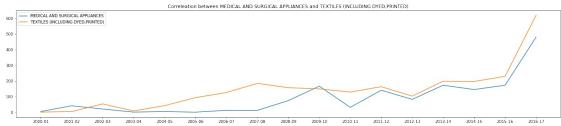
  600
  400
                                                        2013-14
```

2010-11

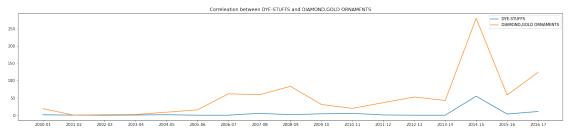
2012-13

SUGAR CONSTRUCTION (INFRASTRUCTURE) ACTIVITIES



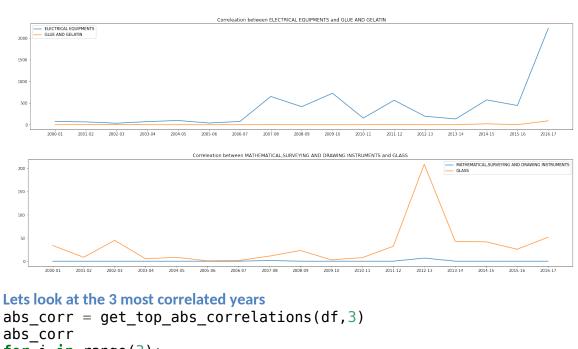












```
abs_corr
for i in range(3):
  plt.plot(df_trans.loc[abs_corr.index[i][0]], label =
abs corr.index[i][0])
  plt.plot(df_trans.loc[abs_corr.index[i][1]], label =
abs corr.index[i][1])
  plt.title(f'Correleation between {abs_corr.index[i][0]} and
{abs_corr.index[i][1]}')
  plt.legend()
  plt.show()
```

