```
import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
import seaborn as sns
import warnings
warnings.filterwarnings('ignore')
```

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

Choose files FDI data.csv

- **FDI data.csv**(text/csv) - 7992 bytes, last modified: 08/01/2025 - 100% done

```
Saving FDI data.csv to FDI data.csv
```

```
df = pd.read_csv('FDI data.csv')
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	20
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	35
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	68
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	70
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	61
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	

Next steps: Generate code with df | View recommended plots | New interactive sheet

```
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	201
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 DEVELOPMENT: Townships, housing, ...	24.33	51.75	36.10	47.04	152.06	228.71	1392.95	3887.33	4657.51	5466.13	1663.03	3140.78	1332.49	1226
62	MISCELLANEOUS INDUSTRIES	832.07	221.37	218.76	235.48	121.83	164.76	304.87	528.42	1549.70	1147.56	1475.97	813.38	229.49	468

```
df.shape
```

```
(63, 18)
```

```
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
```

All the datatypes are correct

```
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

Here from above we can understand that the data of FDI is largliy skewed specifically right skewed as mean is greater than median(50%)

```
df.isnull().sum()
```

	0
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

There are no null values

```
df.duplicated().sum()
```

```
0
```

From above we can conclude that there are no duplicates involved here!!!

```
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')
```

```
df1 = df.set_index('Sector')  ####making Sector as index column
df1.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
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	359.3
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	684.3
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	707.0
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	615.9
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	0.0

Next steps: Generate code with df1 | View recommended plots | New interactive sheet

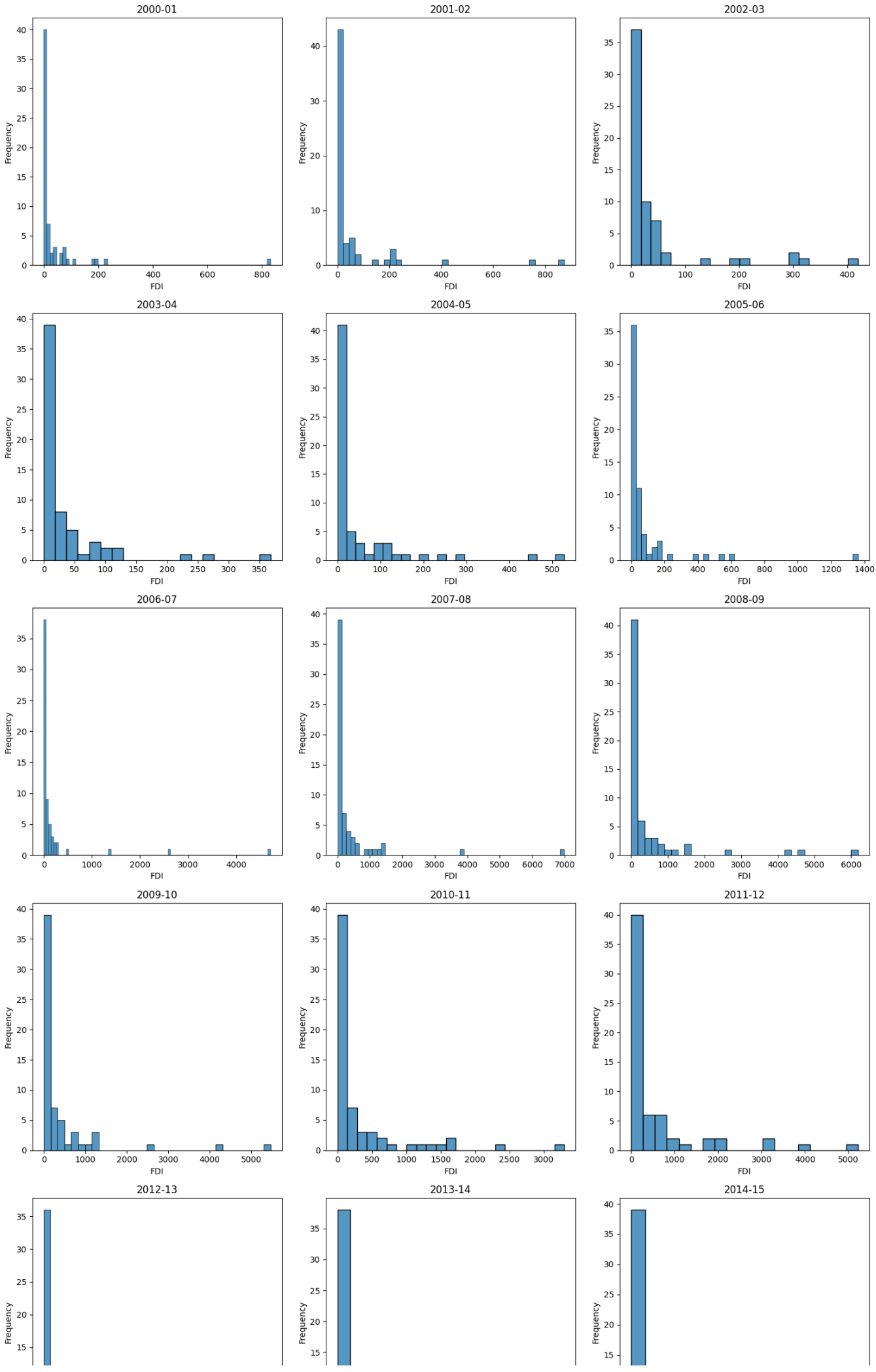
```
df['Sector'].nunique()
```

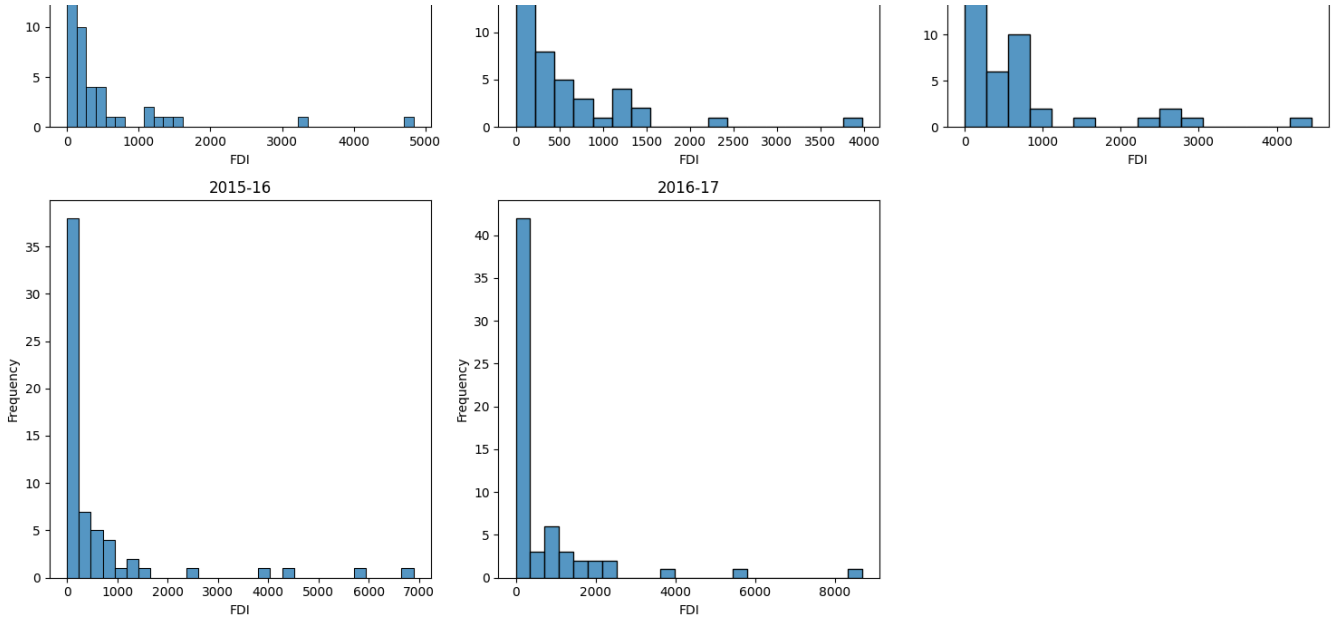
```
63
```

Univariate Plots

Distrubtion of FDI For Each Year

```
def histplots(df1):
    columns = df1.columns
    n_cols = 3
    n_rows = -(-len(columns) // n_cols)

    fig, axes = plt.subplots(n_rows, n_cols, figsize=(15, 5*n_rows))
    axes = axes.flatten()

    for i, col in enumerate(columns):
        sns.histplot(df1[col], ax=axes[i])
        axes[i].set_title(col)
        axes[i].set_xlabel("FDI")
        axes[i].set_ylabel('Frequency')

    # Remove empty subplots
    for j in range(i + 1, n_rows * n_cols):
        fig.delaxes(axes[j])

    plt.tight_layout()
    plt.show()

histplots(df1)
```


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
Frequency
FDI




```
df2 = df1.T
df2.head()
```

Sector	METALLURGICAL INDUSTRIES	MINING	POWER	NON-CONVENTIONAL ENERGY	COAL PRODUCTION	PETROLEUM & NATURAL GAS	BOILERS AND STEAM GENERATING PLANTS	PRIME MOVER (OTHER THAN ELECTRICAL GENERATORS)	ELECTRICAL EQUIPMENTS	COMPUTER SOFTWARE & HARDWARE	.
2000-01	22.69	1.32	89.42	0.00	0.00	9.35	0.00	0.00	79.76	228.39	
2001-02	14.14	6.52	757.44	0.00	0.00	211.07	0.00	0.00	65.76	419.39	
2002-03	36.61	10.06	59.11	1.70	0.00	56.78	0.00	0.00	34.71	314.24	
2003-04	8.11	23.48	27.09	4.14	0.04	80.64	0.04	0.00	73.20	368.32	
2004-05	200.38	9.92	43.37	1.27	0.00	102.78	0.54	2.66	97.40	527.90	

5 rows × 63 columns

```
df2.columns
```

```
Index(['METALLURGICAL INDUSTRIES', 'MINING', 'POWER',
       'NON-CONVENTIONAL ENERGY', 'COAL PRODUCTION', 'PETROLEUM & NATURAL GAS',
       'BOILERS AND STEAM GENERATING PLANTS',
       'PRIME MOVER (OTHER THAN ELECTRICAL GENERATORS)',
       'ELECTRICAL EQUIPMENTS', 'COMPUTER SOFTWARE & HARDWARE', 'ELECTRONICS',
       'TELECOMMUNICATIONS',
       'INFORMATION & BROADCASTING (INCLUDING PRINT MEDIA)',
       'AUTOMOBILE INDUSTRY', 'AIR TRANSPORT (INCLUDING AIR FREIGHT)',
       'SEA TRANSPORT', 'PORTS', 'RAILWAY RELATED COMPONENTS',
       'INDUSTRIAL MACHINERY', 'MACHINE TOOLS', 'AGRICULTURAL MACHINERY',
       'EARTH-MOVING MACHINERY',
       'MISCELLANEOUS MECHANICAL & ENGINEERING INDUSTRIES',
       'COMMERCIAL, OFFICE & HOUSEHOLD EQUIPMENTS',
       'MEDICAL AND SURGICAL APPLIANCES', 'INDUSTRIAL INSTRUMENTS',
       'SCIENTIFIC INSTRUMENTS',
       'MATHEMATICAL,SURVEYING AND DRAWING INSTRUMENTS', 'FERTILIZERS',
       'CHEMICALS (OTHER THAN FERTILIZERS)', 'PHOTOGRAPHIC RAW FILM AND PAPER',
       'DYE-STUFFS', 'DRUGS & PHARMACEUTICALS',
       'TEXTILES (INCLUDING DYED,PRINTED)',
```

```
        'PAPER AND PULP (INCLUDING PAPER PRODUCTS)', 'SUGAR',
        'FERMENTATION INDUSTRIES', 'FOOD PROCESSING INDUSTRIES',
        'VEGETABLE OILS AND VANASPATI',
        'SOAPS, COSMETICS & TOILET PREPARATIONS', 'RUBBER GOODS',
        'LEATHER,LEATHER GOODS AND PICKERS', 'GLUE AND GELATIN', 'GLASS',
        'CERAMICS', 'CEMENT AND GYPSUM PRODUCTS', 'TIMBER PRODUCTS',
        'DEFENCE INDUSTRIES', 'CONSULTANCY SERVICES',
        'SERVICES SECTOR (Fin.,Banking,Insurance,Non Fin/Business,Outsourcing,R&D,Courier,Tech. Testing and Analysis,
    Other)',
        'HOSPITAL & DIAGNOSTIC CENTRES', 'EDUCATION', 'HOTEL & TOURISM',
        'TRADING', 'RETAIL TRADING', 'AGRICULTURE SERVICES',
        'DIAMOND,GOLD ORNAMENTS',
        'TEA AND COFFEE (PROCESSING & WAREHOUSING COFFEE & RUBBER)',
        'PRINTING OF BOOKS (INCLUDING LITHO PRINTING INDUSTRY)', 'COIR',
        'CONSTRUCTION (INFRASTRUCTURE) ACTIVITIES',
        'CONSTRUCTION DEVELOPMENT: Townships, housing, built-up infrastructure and construction-development projects',
        'MISCELLANEOUS INDUSTRIES'],
       dtype='object', name='Sector')
```

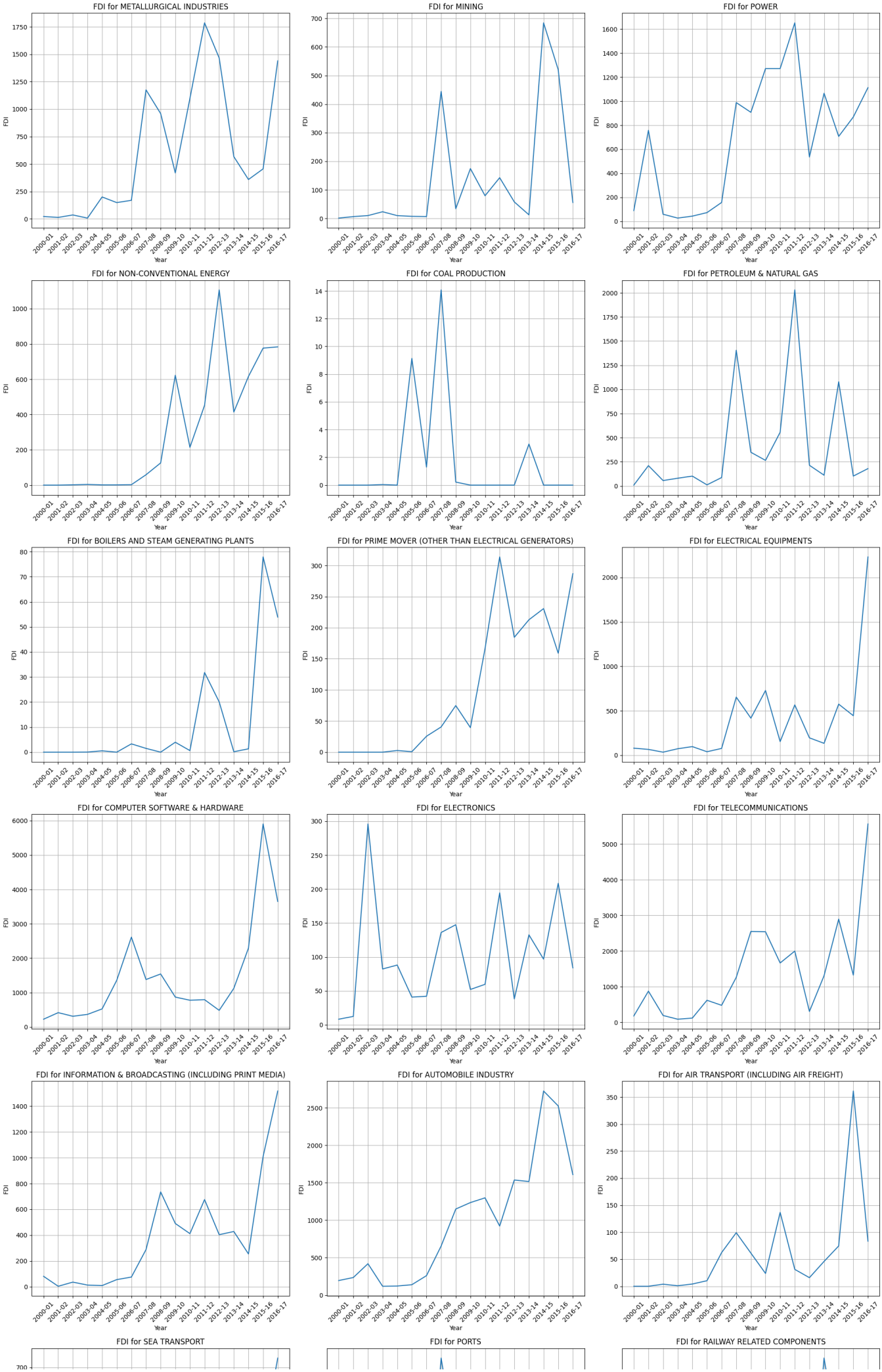
```
df2.rename(columns={'CONSTRUCTION DEVELOPMENT: Townships, housing, built-up infrastructure and construction-development proj
df2.head()
```

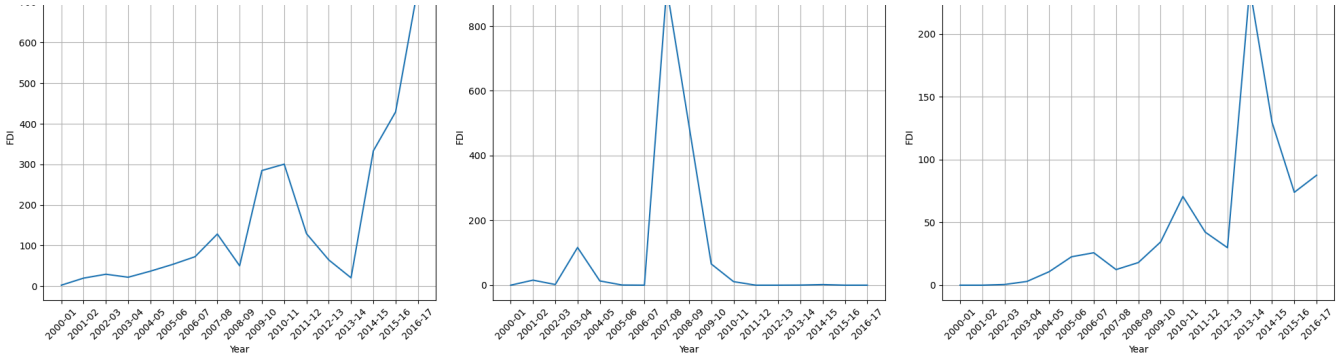
Sector	METALLURGICAL INDUSTRIES	MINING	POWER	NON-CONVENTIONAL ENERGY	COAL PRODUCTION	PETROLEUM & NATURAL GAS	BOILERS AND STEAM GENERATING PLANTS	PRIME MOVER (OTHER THAN ELECTRICAL GENERATORS)	ELECTRICAL EQUIPMENTS	COMPUTER SOFTWARE & HARDWARE
2000-01	22.69	1.32	89.42	0.00	0.00	9.35	0.00	0.00	79.76	228.39
2001-02	14.14	6.52	757.44	0.00	0.00	211.07	0.00	0.00	65.76	419.39
2002-03	36.61	10.06	59.11	1.70	0.00	56.78	0.00	0.00	34.71	314.24
2003-04	8.11	23.48	27.09	4.14	0.04	80.64	0.04	0.00	73.20	368.32
2004-05	200.38	9.92	43.37	1.27	0.00	102.78	0.54	2.66	97.40	527.90

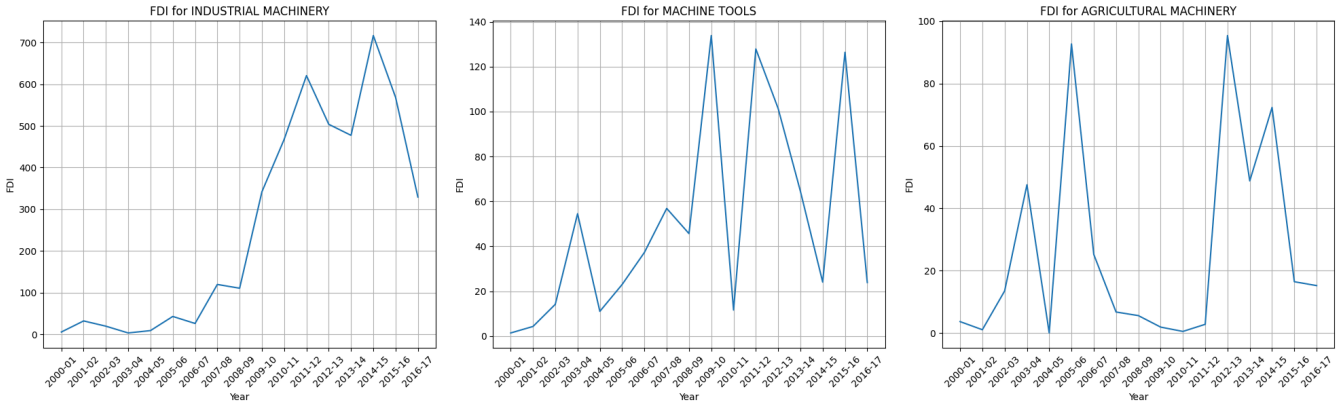
5 rows × 63 columns

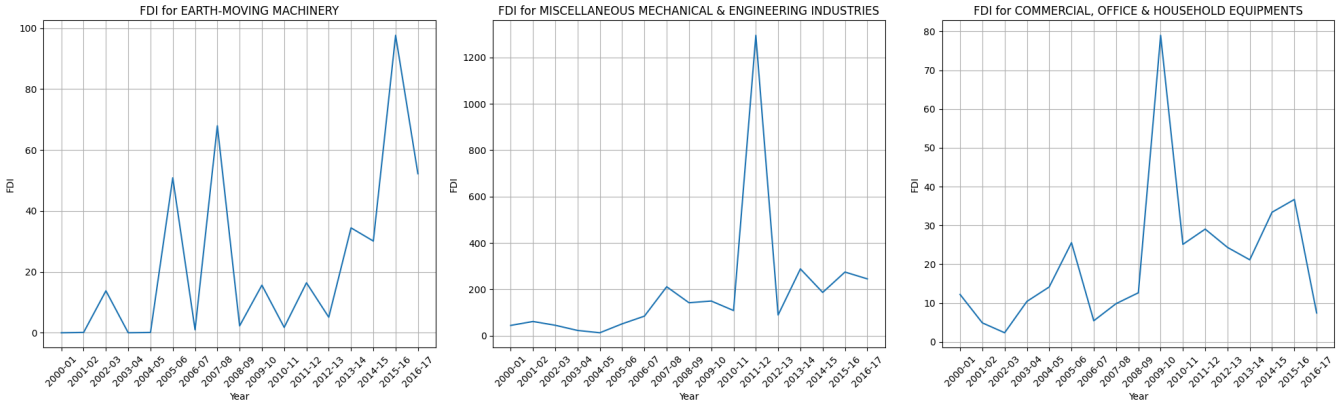
FDI Over The Years For Each Sector

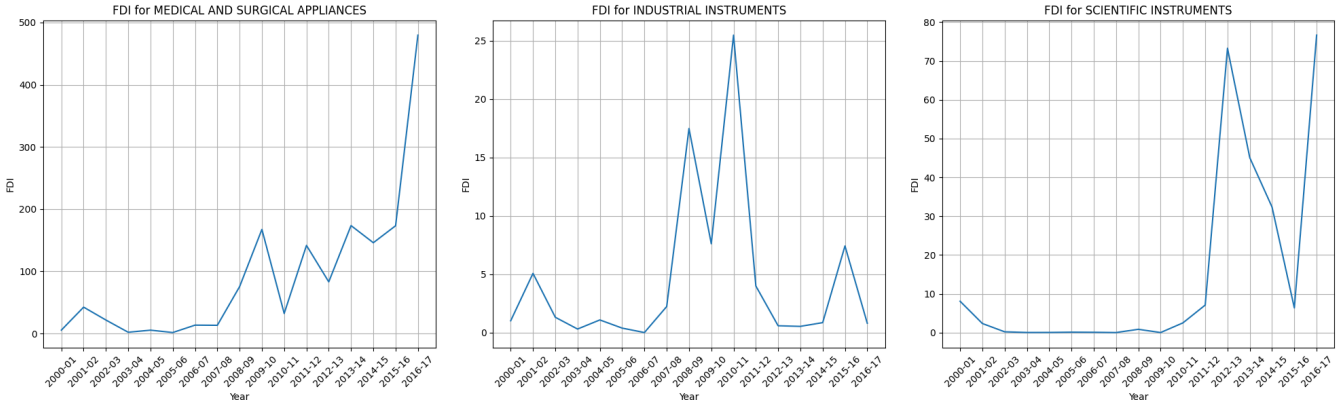
```
def scatterp(df2):
    columns = df2.columns
    n_cols = 3
    n_rows = -(-len(columns) // n_cols)

    fig, axes = plt.subplots(n_rows, n_cols, figsize=(20, 6*n_rows))
    axes = axes.flatten()

    for i, col in enumerate(columns):
        axes[i].plot(df2.index.values, df2[col])
        axes[i].set_title("FDI for " + col)
        axes[i].set_xlabel("Year")
        axes[i].set_xticklabels(df2.index.values, rotation=45)
        axes[i].set_ylabel("FDI")
        axes[i].grid(True)

    # Remove empty subplots
    for j in range(i + 1, n_rows * n_cols):
        fig.delaxes(axes[j])

    plt.tight_layout()
    plt.show()

scatterp(df2)
```

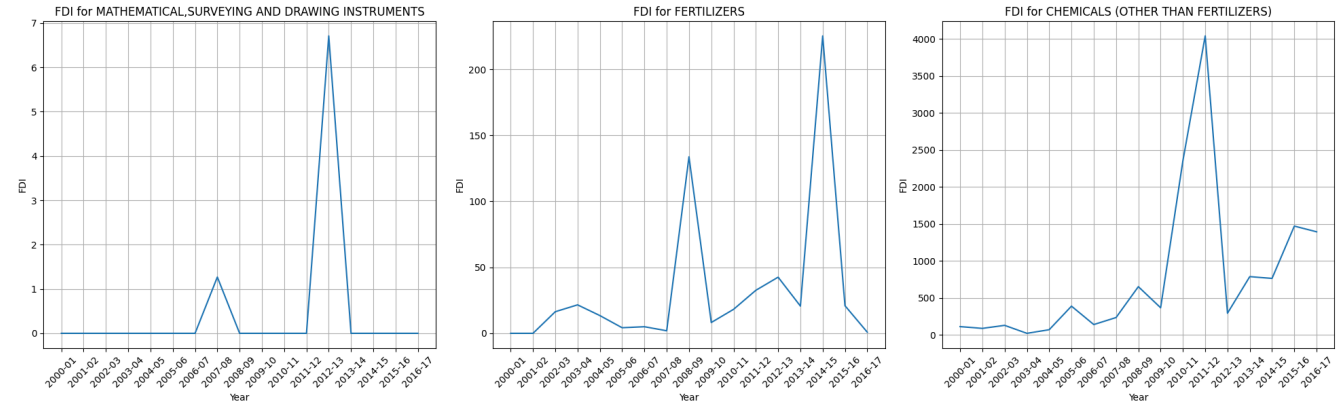

FDI for METALLURGICAL INDUSTRIES
FDI for MINING
FDI for POWER
FDI for NON-CONVENTIONAL ENERGY
FDI for COAL PRODUCTION
FDI for PETROLEUM & NATURAL GAS
FDI for BOILERS AND STEAM GENERATING PLANTS
FDI for PRIME MOVER (OTHER THAN ELECTRICAL GENERATORS)
FDI for ELECTRICAL EQUIPMENTS
FDI for COMPUTER SOFTWARE & HARDWARE
FDI for ELECTRONICS
FDI for TELECOMMUNICATIONS
FDI for INFORMATION & BROADCASTING (INCLUDING PRINT MEDIA)
FDI for AUTOMOBILE INDUSTRY
FDI for AIR TRANSPORT (INCLUDING AIR FREIGHT)
FDI for SEA TRANSPORT
FDI for PORTS
FDI for RAILWAY RELATED COMPONENTS
FDI
Year

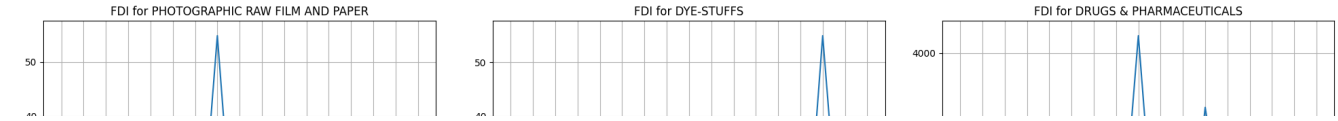

FDI
Year

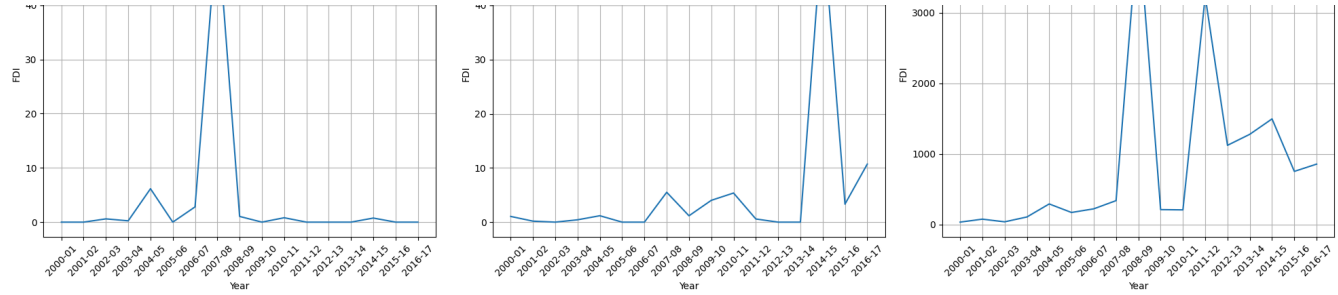

FDI for INDUSTRIAL MACHINERY
FDI for MACHINE TOOLS
FDI for AGRICULTURAL MACHINERY
FDI
Year

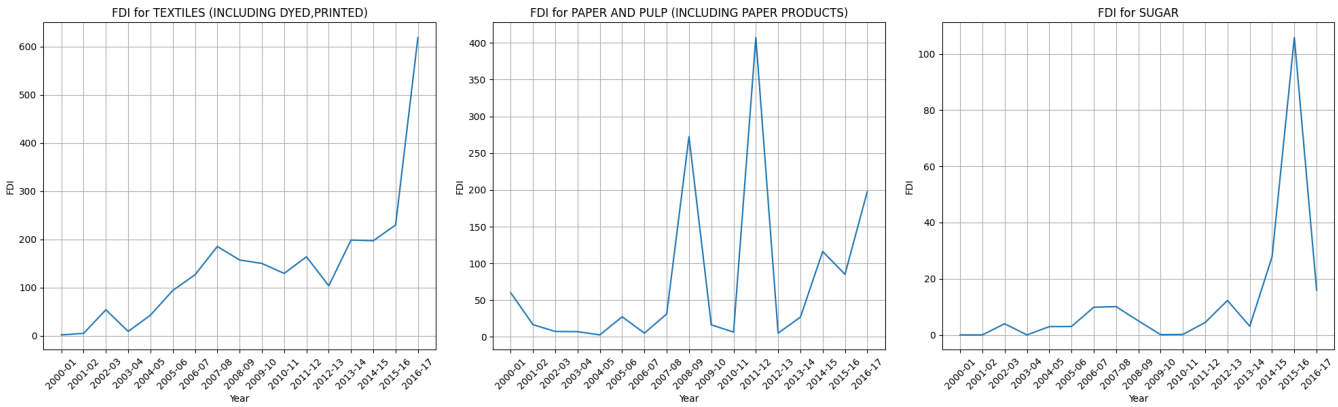

FDI for EARTH-MOVING MACHINERY
FDI for MISCELLANEOUS MECHANICAL & ENGINEERING INDUSTRIES
FDI for COMMERCIAL, OFFICE & HOUSEHOLD EQUIPMENTS
FDI
Year

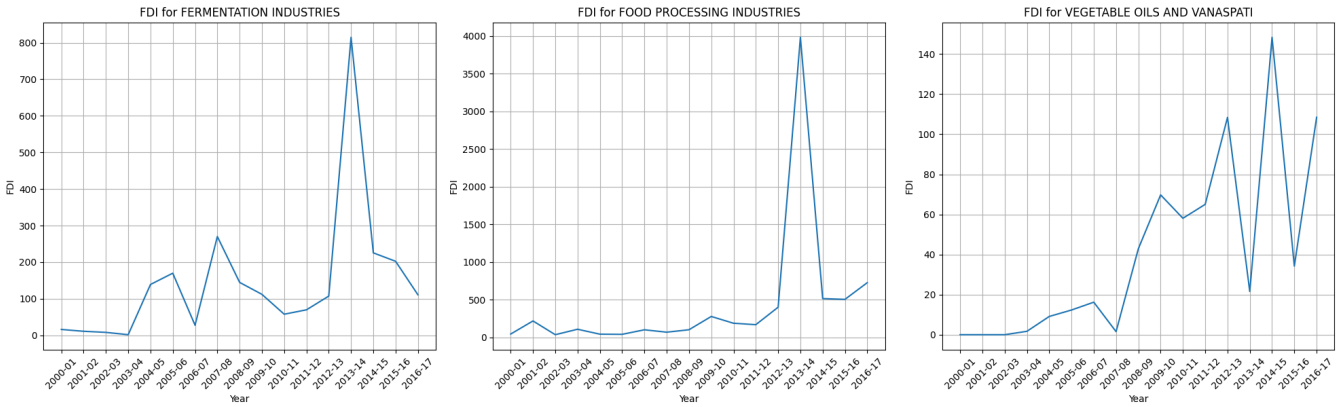

FDI for MEDICAL AND SURGICAL APPLIANCES
FDI for INDUSTRIAL INSTRUMENTS
FDI for SCIENTIFIC INSTRUMENTS
FDI
Year

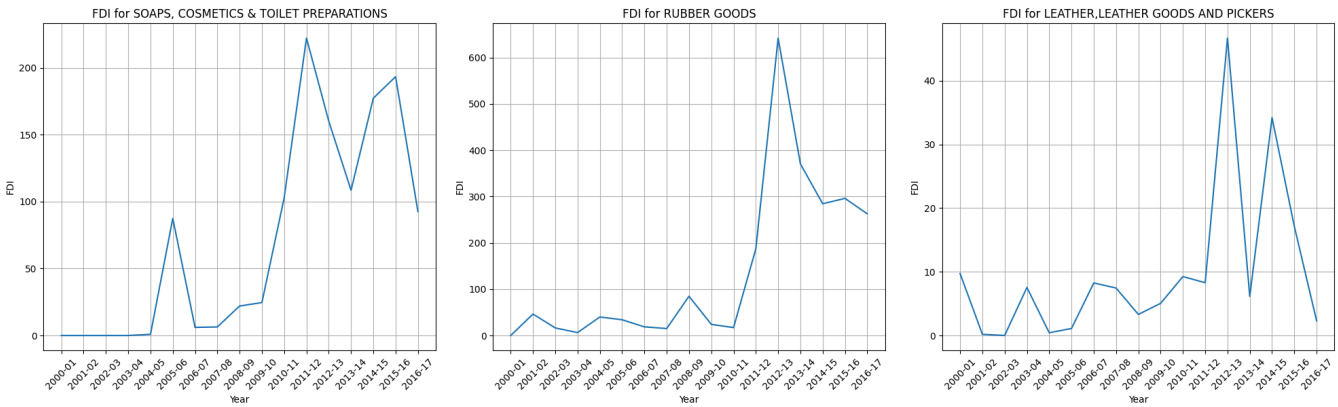

FDI for MATHEMATICAL,SURVEYING AND DRAWING INSTRUMENTS
FDI for FERTILIZERS
FDI for CHEMICALS (OTHER THAN FERTILIZERS)
FDI
Year

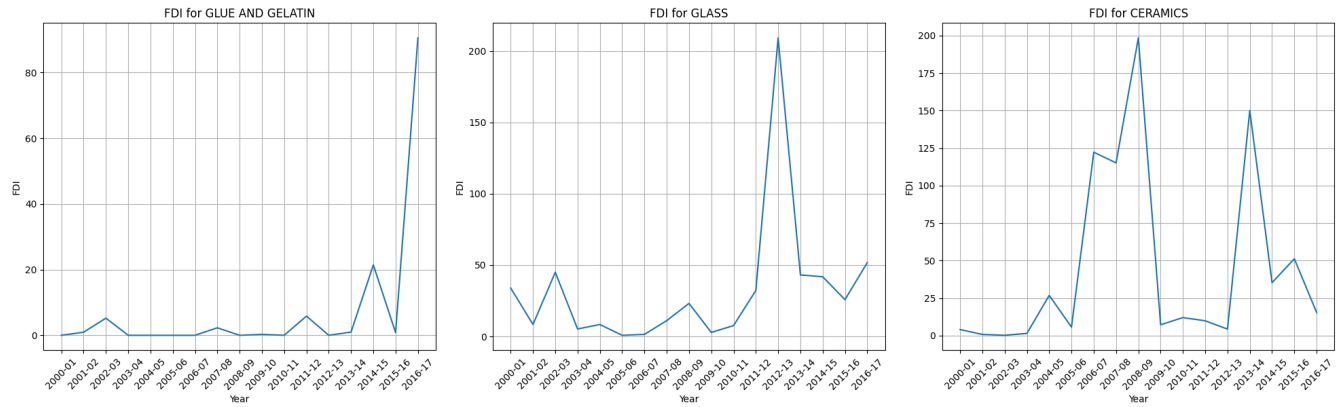

FDI for PHOTOGRAPHIC RAW FILM AND PAPER
FDI for DYE-STUFFS
FDI for DRUGS & PHARMACEUTICALS

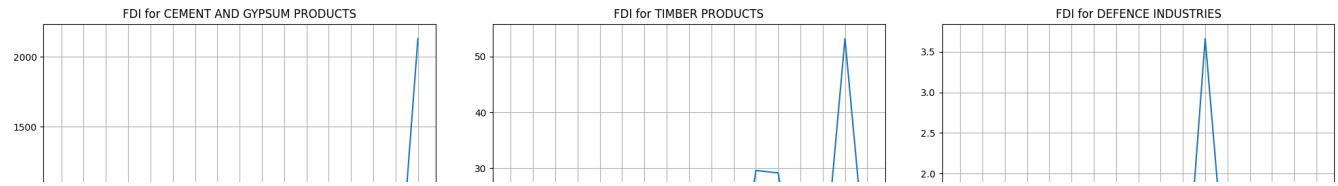

FDI
Year

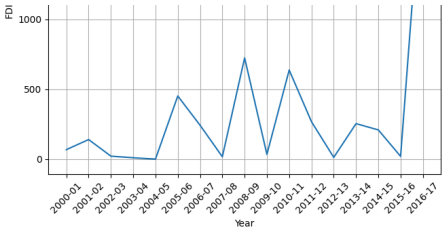

FDI for TEXTILES (INCLUDING DYED,PRINTED)
FDI for PAPER AND PULP (INCLUDING PAPER PRODUCTS)
FDI for SUGAR
FDI
Year

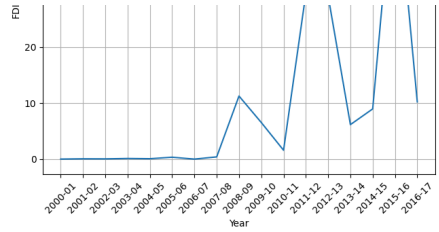

FDI for FERMENTATION INDUSTRIES
FDI for FOOD PROCESSING INDUSTRIES
FDI for VEGETABLE OILS AND VANASPATI
FDI
Year

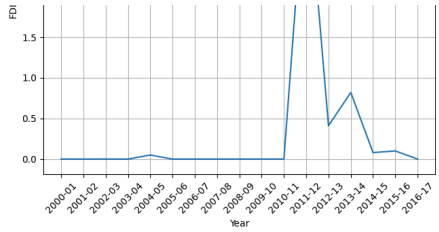

FDI for SOAPS, COSMETICS & TOILET PREPARATIONS
FDI for RUBBER GOODS
FDI for LEATHER,LEATHER GOODS AND PICKERS
FDI
Year

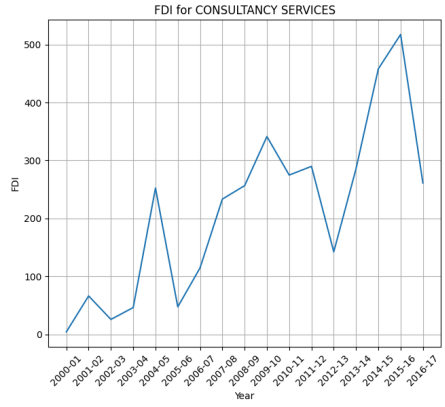

FDI for GLUE AND GELATIN
FDI for GLASS
FDI for CERAMICS
FDI
Year

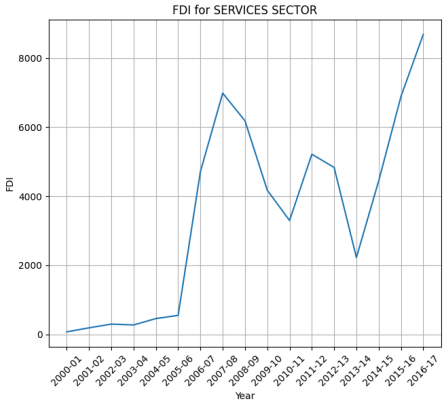

FDI for CEMENT AND GYPSUM PRODUCTS
FDI for TIMBER PRODUCTS
FDI for DEFENCE INDUSTRIES
FDI
Year

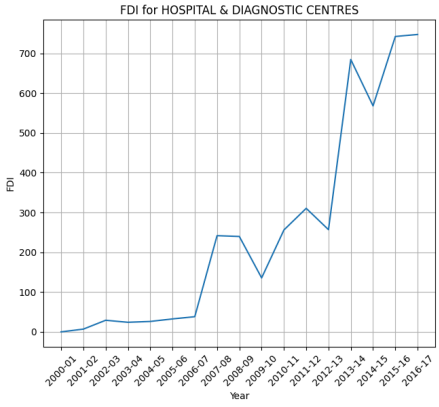

FDI
Year

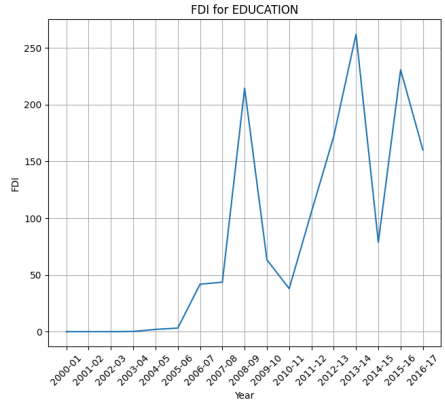

FDI
Year

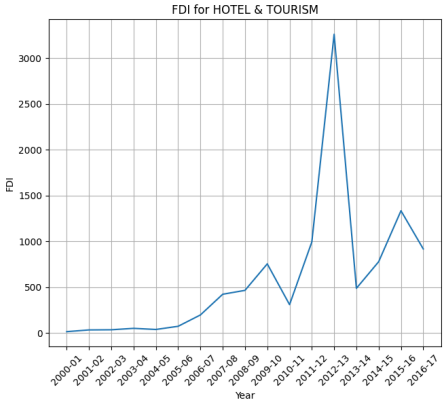

FDI
Year

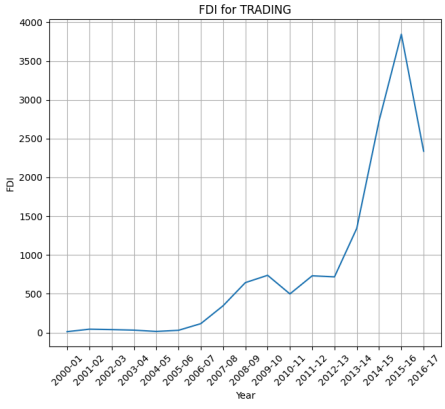

FDI for CONSULTANCY SERVICES
FDI
Year

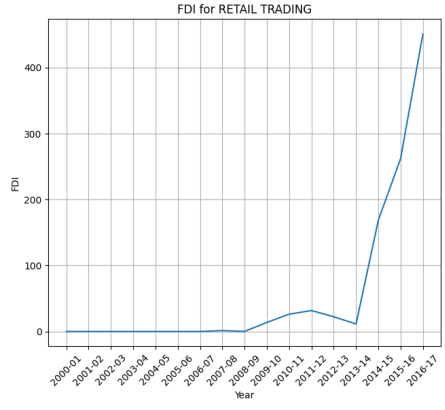

FDI for SERVICES SECTOR
FDI
Year

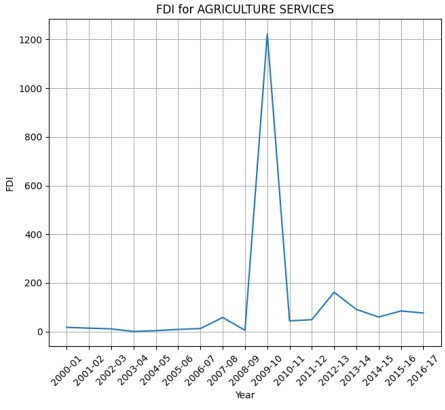

FDI for HOSPITAL & DIAGNOSTIC CENTRES
FDI
Year

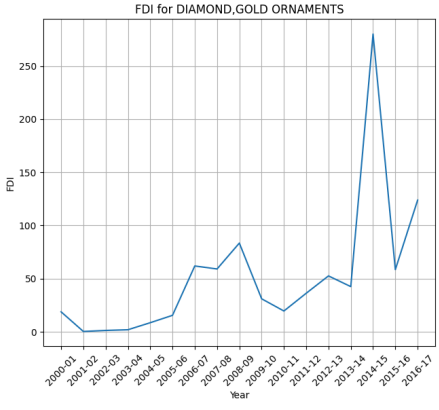

FDI for EDUCATION
FDI
Year

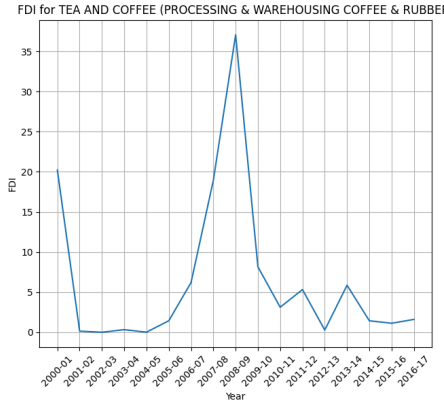

FDI for HOTEL & TOURISM
FDI
Year

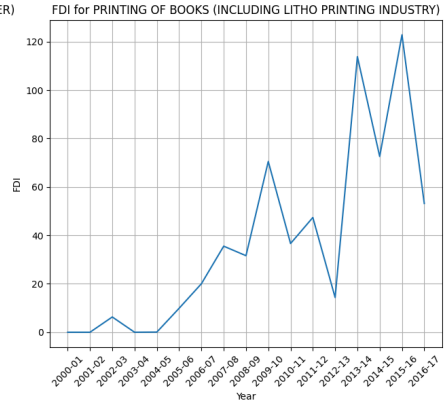

FDI for TRADING
FDI
Year

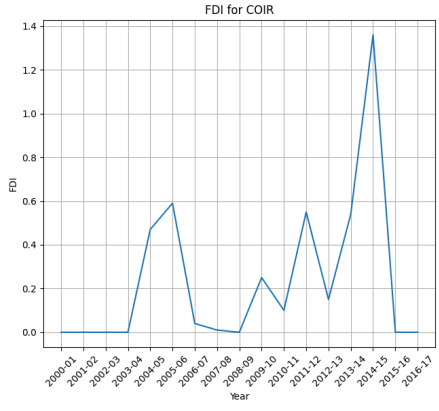

FDI for RETAIL TRADING
FDI
Year

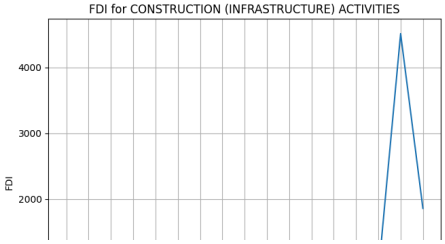

FDI for AGRICULTURE SERVICES
FDI
Year

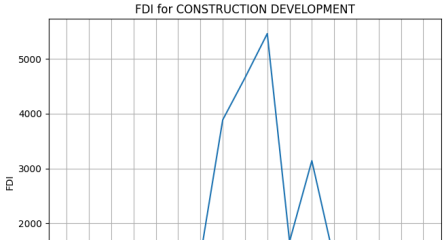

FDI for DIAMOND,GOLD ORNAMENTS
FDI
Year

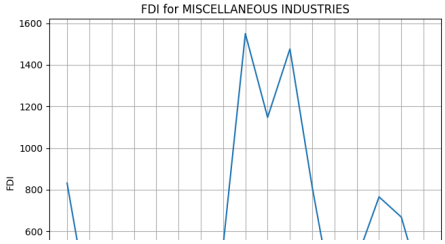

FDI for TEA AND COFFEE (PROCESSING & WAREHOUSING COFFEE & RUBBER)
FDI
Year

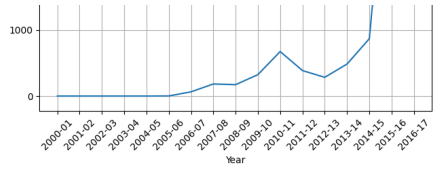

FDI for PRINTING OF BOOKS (INCLUDING LITHO PRINTING INDUSTRY)
FDI
Year

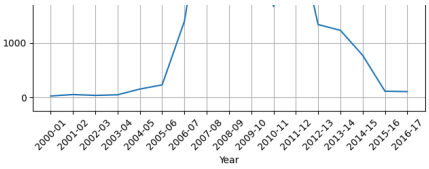

FDI for COIR
FDI
Year


FDI for CONSTRUCTION (INFRASTRUCTURE) ACTIVITIES
FDI


FDI for CONSTRUCTION DEVELOPMENT
FDI


FDI for MISCELLANEOUS INDUSTRIES
FDI


1000
0
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
Year


1000
0
2000-01
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2002-03
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2005-06
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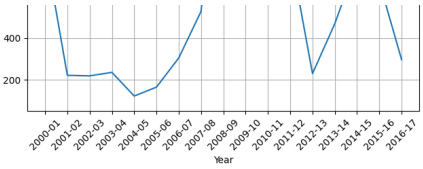
2007-08
2008-09
2009-10
2010-11
2011-12
2012-13
2013-14
2014-15
2015-16
2016-17
Year


400
200
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
Year

The above plot shows us the how FDI changed over the years for each sector

Year wise total FDI

```
year_total = df2.sum(axis=1)
year_total = pd.DataFrame(year_total, columns=['Total_FDI'])
year_total.sort_values(by='Total_FDI', ascending=False)
```

	Total_FDI
2016-17	43478.26
2015-16	40000.99
2011-12	35120.78
2008-09	31395.96
2014-15	30930.47
2009-10	25834.38
2007-08	24575.40
2013-14	24299.32
2012-13	22423.59
2010-11	21383.07
2006-07	12491.76
2005-06	5539.75
2001-02	4027.69
2004-05	3218.69
2002-03	2704.32
2000-01	2378.71
2003-04	2187.85

Year wise Average FDI

```
year_avg = df2.mean(axis=1)
year_avg = pd.DataFrame(year_avg, columns=['Avg_FDI'])
year_avg.sort_values(by='Avg_FDI', ascending=False)
```

	Avg_FDI
2016-17	690.131111
2015-16	634.936349
2011-12	557.472698
2008-09	498.348571
2014-15	490.959841
2009-10	410.069524
2007-08	390.085714
2013-14	385.703492
2012-13	355.930000
2010-11	339.413810
2006-07	198.281905
2005-06	87.932540
2001-02	63.931587
2004-05	51.090317
2002-03	42.925714
2000-01	37.757302
2003-04	34.727778

Top 10 Sector wise total FDI

```
df3 = df2.T
sec_total = df3.sum(axis=1)
sec_total = pd.DataFrame(sec_total, columns=['Total_FDI'])
sec_total_top = sec_total.sort_values(by='Total_FDI', ascending=False).head(10)
sec_total_top
```

Sector	Total_FDI
SERVICES SECTOR	59476.49
COMPUTER SOFTWARE & HARDWARE	24669.49
CONSTRUCTION DEVELOPMENT	24293.09
TELECOMMUNICATIONS	23946.01
AUTOMOBILE INDUSTRY	16673.92
DRUGS & PHARMACEUTICALS	14706.90
TRADING	14210.88
CHEMICALS (OTHER THAN FERTILIZERS)	13293.09
POWER	11589.13
METALLURGICAL INDUSTRIES	10330.54

Next steps: Generate code with sec_total_top | View recommended plots | New interactive sheet

From above data we can see that the service sector was able to draw more FDI than other sectors

Bottom 10 Sector wise total FDI

```
sec_total_tail = sec_total.sort_values(by='Total_FDI', ascending=False).tail(10)
sec_total_tail
```

Sector	Total_FDI
TIMBER PRODUCTS	157.68
GLUE AND GELATIN	128.39
TEA AND COFFEE (PROCESSING & WAREHOUSING COFFEE & RUBBER)	111.22
DYE-STUFFS	88.40
INDUSTRIAL INSTRUMENTS	76.12
PHOTOGRAPHIC RAW FILM AND PAPER	67.28
COAL PRODUCTION	27.74
MATHEMATICAL,SURVEYING AND DRAWING INSTRUMENTS	7.98
DEFENCE INDUSTRIES	5.12
COIR	4.06

Next steps: Generate code with sec_total_tail | View recommended plots | New interactive sheet

From above data we can see that the COIR sector was able to draw least FDI than other sectors

Top 10 Sector wise Average FDI

```
sec_avg = df3.mean(axis=1)
sec_avg = pd.DataFrame(sec_avg, columns=['Avg_FDI'])
sec_avg_top = sec_avg.sort_values(by='Avg_FDI', ascending=False).head(10)
sec_avg_top
```

Sector	Avg_FDI
SERVICES SECTOR	3498.617059
COMPUTER SOFTWARE & HARDWARE	1451.146471
CONSTRUCTION DEVELOPMENT	1429.005294
TELECOMMUNICATIONS	1408.588824
AUTOMOBILE INDUSTRY	980.818824
DRUGS & PHARMACEUTICALS	865.111765
TRADING	835.934118
CHEMICALS (OTHER THAN FERTILIZERS)	781.946471
POWER	681.713529
METALLURGICAL INDUSTRIES	607.678824

Next steps: Generate code with sec_avg_top | View recommended plots | New interactive sheet

Bottom 10 Sector wise Average FDI

```
sec_avg_bottom = sec_avg.sort_values(by='Avg_FDI', ascending=False).tail(10)
sec_avg_bottom
```

Sector	Avg_FDI
TIMBER PRODUCTS	9.275294
GLUE AND GELATIN	7.552353
TEA AND COFFEE (PROCESSING & WAREHOUSING COFFEE & RUBBER)	6.542353
DYE-STUFFS	5.200000
INDUSTRIAL INSTRUMENTS	4.477647
PHOTOGRAPHIC RAW FILM AND PAPER	3.957647
COAL PRODUCTION	1.631765
MATHEMATICAL,SURVEYING AND DRAWING INSTRUMENTS	0.469412
DEFENCE INDUSTRIES	0.301176
COIR	0.238824

Next steps: Generate code with sec_avg_bottom | View recommended plots | New interactive sheet

Yearwise Total_FDI in INDIA

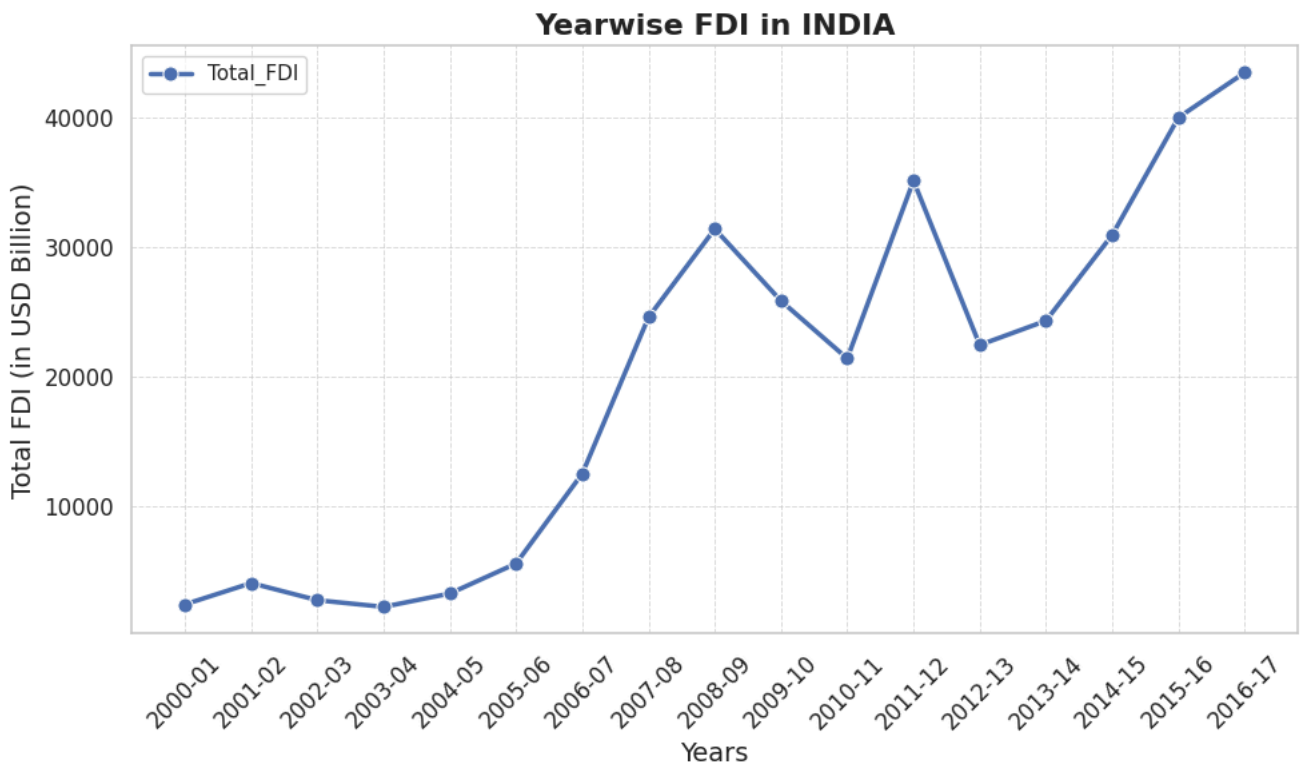
```
sns.set(style="whitegrid")

plt.figure(figsize=(10, 6))

sns.lineplot(data=year_total, marker='o', color='#1f77b4', markersize=8, linewidth=2.5)


plt.title('Yearwise FDI in INDIA', fontsize=16, fontweight='bold')
plt.xlabel('Years', fontsize=14)
plt.ylabel('Total FDI (in USD Billion)', fontsize=14)
plt.xticks(rotation=45, fontsize=12)
plt.yticks(fontsize=12)


plt.grid(True, linestyle='--', linewidth=0.7, alpha=0.7)
plt.tight_layout()
plt.show()
```



Conculsions:

1.The combination of economic reforms, liberalization policies, robust economic growth, and sector-specific developments led to a significant increase in FDI in India on 2006 onwards till 2008-09.

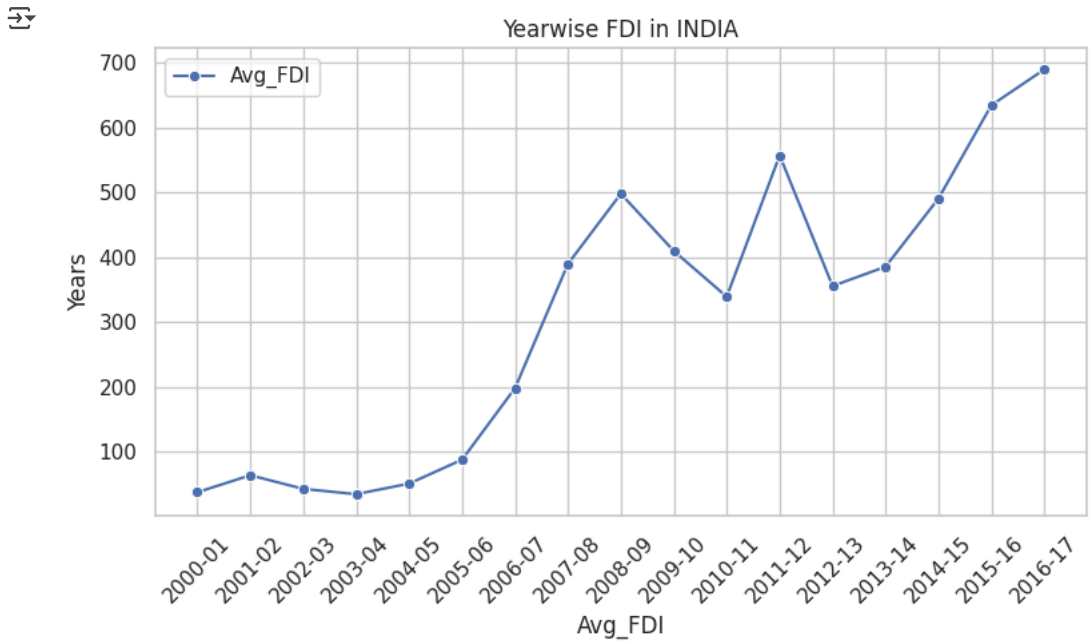
2.Again their was steep decrease in FDI for couple of years due to global financial crisis.

3.Again Following increase and decrease trend, from 2012-13 till today FDI is increasing rapidly.

Yearwise Avg_FDI in INDIA

```
plt.figure(figsize=(8, 5))
sns.lineplot(data=year_avg, marker='o', color='b')
plt.title('Yearwise FDI in INDIA')
plt.xlabel('Avg_FDI')
plt.ylabel('Years')
plt.grid(True)
plt.xticks(rotation=45)

plt.tight_layout()
plt.show()
```



Top10Sectors FDI in INDIA

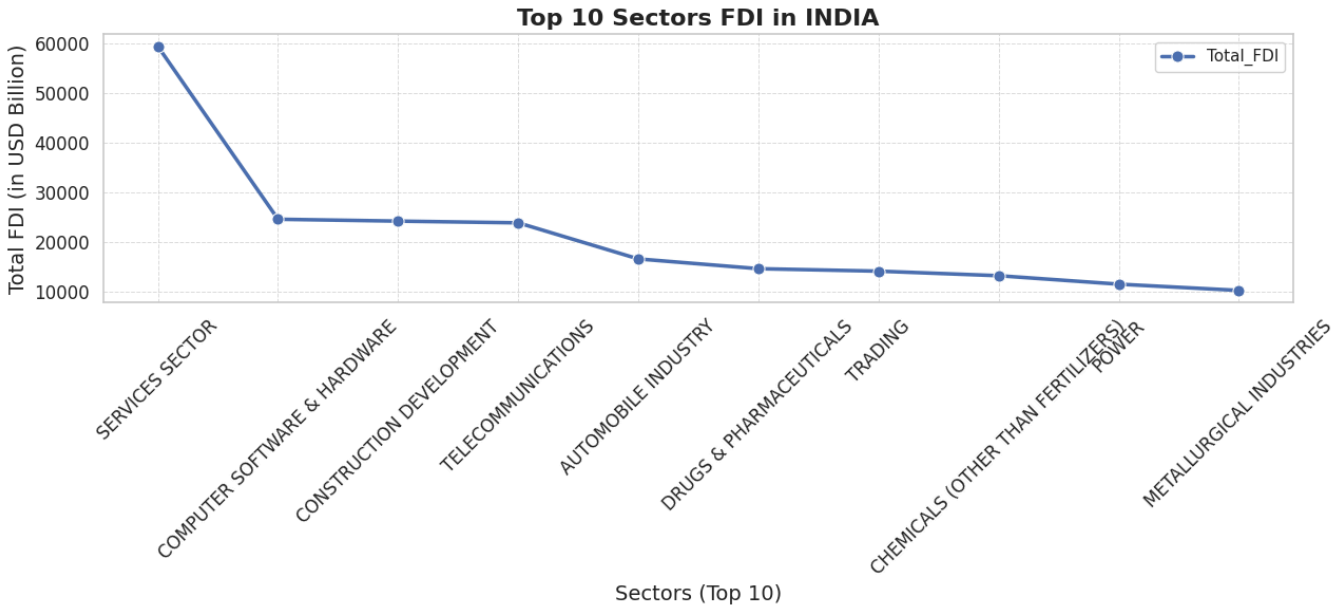
```
sns.set(style="whitegrid")


plt.figure(figsize=(13, 6))

sns.lineplot(data=sec_total_top, marker='o', color='#ff7f0e', markersize=8, linewidth=2.5)


plt.title('Top 10 Sectors FDI in INDIA', fontsize=16, fontweight='bold')
plt.xlabel('Sectors (Top 10)', fontsize=14)
plt.ylabel('Total FDI (in USD Billion)', fontsize=14)
plt.xticks(rotation=45, fontsize=12)
plt.yticks(fontsize=12)

plt.grid(True, linestyle='--', linewidth=0.7, alpha=0.7)
plt.tight_layout()
plt.show()
```



Bottom10 Sectors FDI in INDIA

```
sns.set(style="whitegrid")

plt.figure(figsize=(13, 8))

sns.lineplot(data=sec_total_tail, marker='o', color='#2ca02c', markersize=8, linewidth=2.5)

plt.title('Bottom 10 Sectors FDI in INDIA', fontsize=16, fontweight='bold')
plt.xlabel('Sectors (Bottom 10)', fontsize=14)
plt.ylabel('Total FDI (in USD Billion)', fontsize=14)


plt.xticks(rotation=45, fontsize=12)
plt.yticks(fontsize=12)


plt.grid(True, linestyle='--', linewidth=0.7, alpha=0.7)
plt.tight_layout()
plt.show()
```