

EXPLOLATORY DATA ANALYSIS OF INDIAS TRADE TRENDS FROM 2008-2018

November 13, 2019

```
In [46]: #STC PROJECT: WE TRY TO UNDERSTAND HOW INDIA TRADES USING DATA MANIPULATION AND VISUALIZATION  
# BY SHREYAS K 2017120056, VINAYAK JADHAV 2017120020, SOHAM PINGE 2017120047
```

```
In [47]: import numpy as np # linear algebra  
import pandas as pd # data processing, CSV file I/O (e.g. pd.read_csv)  
import matplotlib.pyplot as plt  
import seaborn as sns  
from pandas import DataFrame
```

```
In [48]: import_df=pd.read_csv(r"C:\Users\kailash\Desktop\import.csv")#importing the csv file
```

```
In [49]: export_df=pd.read_csv(r"C:\Users\kailash\Desktop\export.csv")#importing the csv file
```

```
In [50]: from statsmodels.graphics.tsaplots import plot_acf #Auto-Correlation Plots  
from statsmodels.graphics.tsaplots import plot_pacf #Partial-Auto Correlation Plots
```

```
In [51]: import_df.isnull().sum()# calculating the number of nulls
```

```
Out[51]: HSCode          0  
Commodity          0  
value          14027  
country          0  
year          0  
dtype: int64
```

```
In [52]: export_df = export_df.dropna()  
export_df = export_df.reset_index(drop=True)#drop the null values
```

```
In [53]: import_df = import_df.dropna()  
import_df = import_df.reset_index(drop=True)
```

```
In [54]: import_df.head(10)
```

```
Out[54]:
```

	HSCode	Commodity	value	\
0	5	PRODUCTS OF ANIMAL ORIGIN, NOT ELSEWHERE SPECI...	0.00	
1	7	EDIBLE VEGETABLES AND CERTAIN ROOTS AND TUBERS.	12.38	
2	8	EDIBLE FRUIT AND NUTS; PEEL OR CITRUS FRUIT OR...	268.60	
3	9	COFFEE, TEA, MATE AND SPICES.	35.48	

4	12	OIL SEEDS AND OLEA. FRUITS; MISC. GRAINS, SEED...	8.32
5	13	LAC; GUMS, RESINS AND OTHER VEGETABLE SAPS AND...	108.78
6	20	PREPARATIONS OF VEGETABLES, FRUIT, NUTS OR OTH...	0.65
7	25	SALT; SULPHUR; EARTHS AND STONE; PLASTERING MA...	0.05
8	27	MINERAL FUELS, MINERAL OILS AND PRODUCTS OF TH...	0.00
9	41	RAW HIDES AND SKINS (OTHER THAN FURSKINS) AND ...	0.00

	country	year
0	AFGHANISTAN TIS	2018
1	AFGHANISTAN TIS	2018
2	AFGHANISTAN TIS	2018
3	AFGHANISTAN TIS	2018
4	AFGHANISTAN TIS	2018
5	AFGHANISTAN TIS	2018
6	AFGHANISTAN TIS	2018
7	AFGHANISTAN TIS	2018
8	AFGHANISTAN TIS	2018
9	AFGHANISTAN TIS	2018

In [55]: export_df.head(10)

```
Out[55]:
```

	HSCode	Commodity	value \
0	2	MEAT AND EDIBLE MEAT OFFAL.	0.18
1	3	FISH AND CRUSTACEANS, MOLLUSCS AND OTHER AQUAT...	0.00
2	4	DAIRY PRODUCE; BIRDS' EGGS; NATURAL HONEY; EDI...	12.48
3	6	LIVE TREES AND OTHER PLANTS; BULBS; ROOTS AND ...	0.00
4	7	EDIBLE VEGETABLES AND CERTAIN ROOTS AND TUBERS.	1.89
5	8	EDIBLE FRUIT AND NUTS; PEEL OR CITRUS FRUIT OR...	25.01
6	9	COFFEE, TEA, MATE AND SPICES.	13.75
7	10	CEREALS.	0.75
8	11	PRODUCTS OF THE MILLING INDUSTRY; MALT; STARCH...	0.01
9	12	OIL SEEDS AND OLEA. FRUITS; MISC. GRAINS, SEED...	10.02

	country	year
0	AFGHANISTAN TIS	2018
1	AFGHANISTAN TIS	2018
2	AFGHANISTAN TIS	2018
3	AFGHANISTAN TIS	2018
4	AFGHANISTAN TIS	2018
5	AFGHANISTAN TIS	2018
6	AFGHANISTAN TIS	2018
7	AFGHANISTAN TIS	2018
8	AFGHANISTAN TIS	2018
9	AFGHANISTAN TIS	2018

```
In [ ]: importing_countries=import_df[['country']].nunique()
exporting_countries=export_df[['country']].nunique()# CODE TO FIND THE TOTAL NUMBER OF
print("India imports from:",importing_countries,"countries")
print("India exports to:",exporting_countries,"countries")
```

```
In [56]: import_group=import_df.groupby(['country','year']).agg({'value':'sum'})
         export_group=export_df.groupby(['country','year']).agg({'value':'sum'})
```

```
In [57]: import_group
```

```
Out[57]:
```

	country	year	value
	AFGHANISTAN TIS	2010	146.01
		2011	132.49
		2012	159.53
		2013	208.76
		2014	261.92
		2015	615.80
		2016	292.90
		2017	867.54
		2018	870.88
	ALBANIA	2010	0.11
		2011	141.13
		2012	31.05
		2013	198.79
		2014	50.77
		2015	34.36
		2016	6.50
		2017	77.70
		2018	45.24
	ALGERIA	2010	1816.19
		2011	2111.43
		2012	683.54
		2013	860.89
		2014	551.81
		2015	598.86
		2016	605.11
		2017	2521.92
		2018	3393.88
	AMERI SAMOA	2010	0.15
		2011	1.40
		2012	0.78
...			...
	WALLIS F IS	2014	1.01
		2015	0.24
		2017	0.02
	YEMEN REPubLC	2010	1743.89
		2011	970.69
		2012	958.92
		2013	782.18
		2014	540.70
		2015	6.87
		2016	4.81

	2017	306.10
	2018	23.50
ZAMBIA	2010	32.08
	2011	168.83
	2012	324.85
	2013	243.15
	2014	283.34
	2015	475.36
	2016	743.90
	2017	2189.96
	2018	1021.00
ZIMBABWE	2010	11.62
	2011	2.88
	2012	34.54
	2013	12.51
	2014	32.69
	2015	24.44
	2016	60.45
	2017	124.40
	2018	15.62

[2001 rows x 1 columns]

In [58]: export_group

Out [58]:

	country	year	value
	AFGHANISTAN	TIS 2010	422.31
		2011	510.81
		2012	472.55
		2013	474.26
		2014	422.48
		2015	526.51
		2016	506.26
		2017	709.66
		2018	715.35
	ALBANIA	2010	11.49
		2011	12.61
		2012	17.47
		2013	18.71
		2014	19.44
		2015	24.00
		2016	26.43
		2017	29.29
		2018	37.07
	ALGERIA	2010	781.92
		2011	835.59
		2012	1088.67

	2013	1069.44
	2014	1063.65
	2015	787.74
	2016	841.81
	2017	823.18
	2018	940.24
AMERI SAMOA	2010	0.20
	2011	2.05
	2012	0.37
...		...
WALLIS F IS	2014	0.03
	2016	0.00
	2017	0.05
YEMEN REPubLC	2010	513.98
	2011	730.53
	2012	1477.17
	2013	1306.89
	2014	992.02
	2015	399.69
	2016	446.01
	2017	563.37
	2018	741.25
ZAMBIA	2010	118.32
	2011	210.84
	2012	242.95
	2013	377.20
	2014	366.43
	2015	298.02
	2016	237.05
	2017	294.02
	2018	318.90
ZIMBABWE	2010	113.07
	2011	129.22
	2012	153.16
	2013	157.96
	2014	223.84
	2015	205.01
	2016	109.02
	2017	163.46
	2018	181.62

[2099 rows x 1 columns]

```
In [59]: export_group.groupby(['country'])
import_temp=import_group.groupby(['country']).agg({'value':'sum'})
export_temp=export_group.groupby(['country']).agg({'value':'sum'}).loc[import_temp.index]
#GROUPING THE IMPORTS AND EXPORTS COUNTRY WISE AND AGGREGATING THE VALUE AND ARRANGING THEM IN ORDER OF VALUE

In [60]: data_1=import_group.groupby(['country']).agg({'value':'sum'}).sort_values(by='value')
```

```
data_2=export_temp
data_3=data_2-data_1
#GETTING THE TOP 10 VALUES
```

```
In [61]: data_1.columns=['Import']
data_2.columns=['Export']
data_3.columns=['Loss / Profit']
```

```
In [62]: df=pd.DataFrame(index=data_1.index.values)
#df=pd.concat([data_1,data_2,data_3])
df['Import']=data_1
df['Export']=data_2
df['Loss / Profit']=data_3
```

```
df
#IT GIVES US THE TOTAL AMOUNT OF MONEY INDIA IMPORTS AND EXPORTS IN IN INREASING ORDER
```

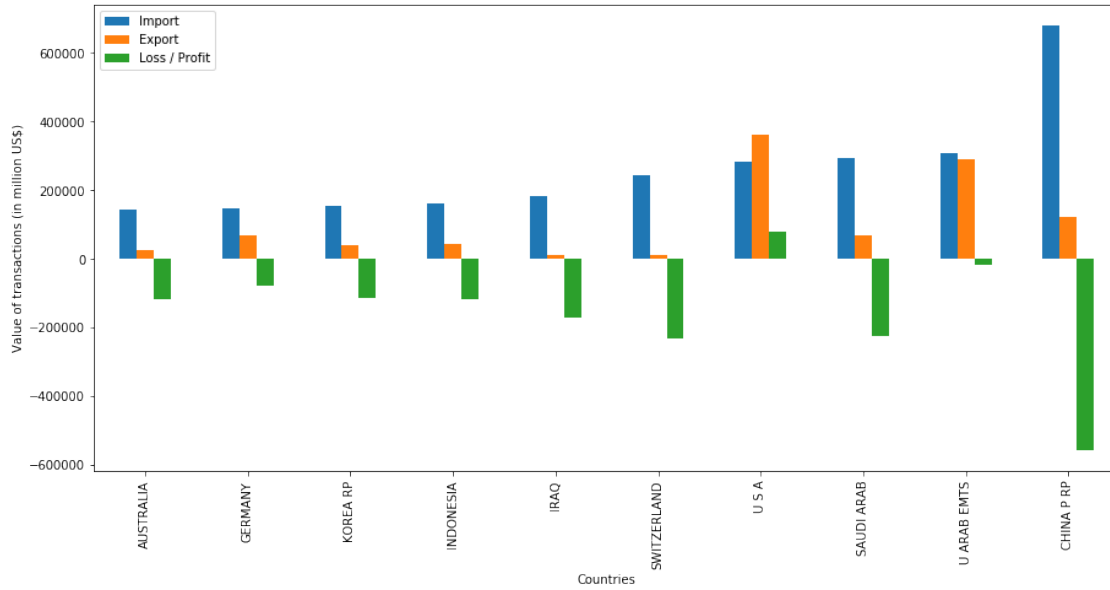
```
Out[62]:
```

	Import	Export	Loss / Profit
AUSTRALIA	142723.67	25372.29	-117351.38
GERMANY	148120.79	68855.45	-79265.34
KOREA RP	154263.94	38023.02	-116240.92
INDONESIA	160452.65	42148.84	-118303.81
IRAQ	182464.57	9833.61	-172630.96
SWITZERLAND	243739.33	9992.56	-233746.77
U S A	281436.58	360609.57	79172.99
SAUDI ARAB	292104.09	65995.99	-226108.10
U ARAB EMTS	307810.15	289374.88	-18435.27
CHINA P RP	678877.15	121806.00	-557071.15

```
In [63]: fig, ax = plt.subplots(figsize=(15,7))
df.plot(kind='bar',ax=ax)
ax.set_xlabel('Countries')
ax.set_ylabel('Value of transactions (in million US$)')
```

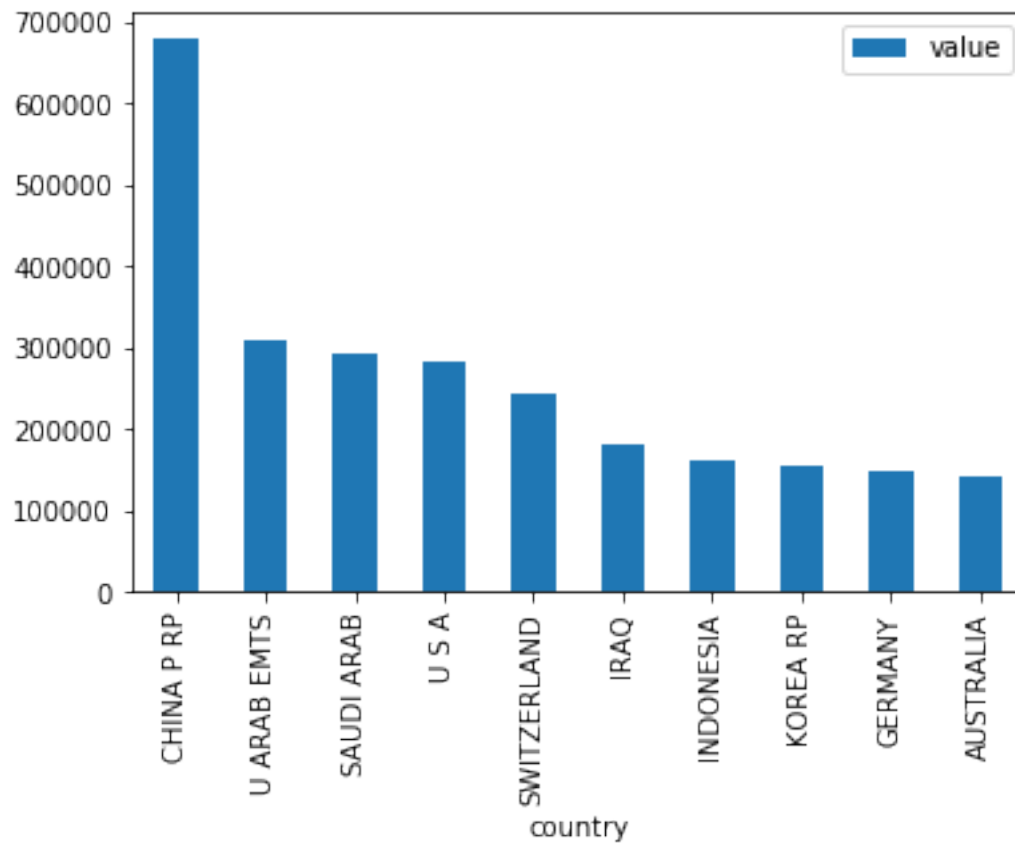
```
# PLOTS THE TOTAL IMPORT EXPORT AND TOTAL INCOME
```

```
Out[63]: Text(0, 0.5, 'Value of transactions (in million US$)')
```



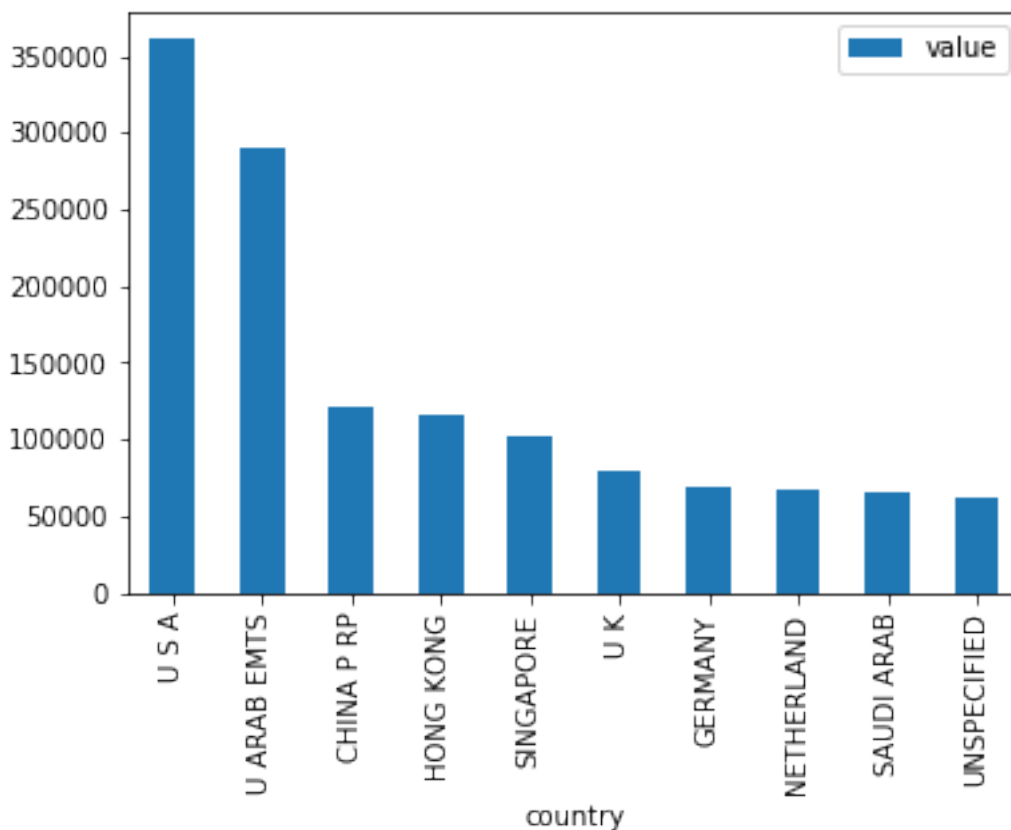
```
In [64]: df_import = import_df.groupby('country').agg({'value': 'sum'}).sort_values(by='value',
df_import.plot(kind='bar')
#PLOTS HOW MUCH INDIA IMPORTS FROM ITS TOP 10 LENDERS
```

```
Out[64]: <matplotlib.axes._subplots.AxesSubplot at 0x19b290caf60>
```



```
In [65]: df_export = export_df.groupby('country').agg({'value': 'sum'}).sort_values(by='value',
df_export.plot(kind='bar')
#PLOTS HOW MUCH INDIA EXPORTS TO ITS TOP 10 BUYERS

Out[65]: <matplotlib.axes._subplots.AxesSubplot at 0x19b28c73978>
```

```
In [66]: Import_ =import_df.groupby(['year']).agg({'value':'sum'})
Export_ =export_df.groupby(['year']).agg({'value':'sum'})
Deficit_=Export_ -Import_
Time_Series=pd.DataFrame(index=Import_.index.values)
Time_Series['Import']=Import_
Time_Series['Export']=Export_
Time_Series['Loss / Profit']=Deficit_
#YEAR WISE ANALYSIS
```

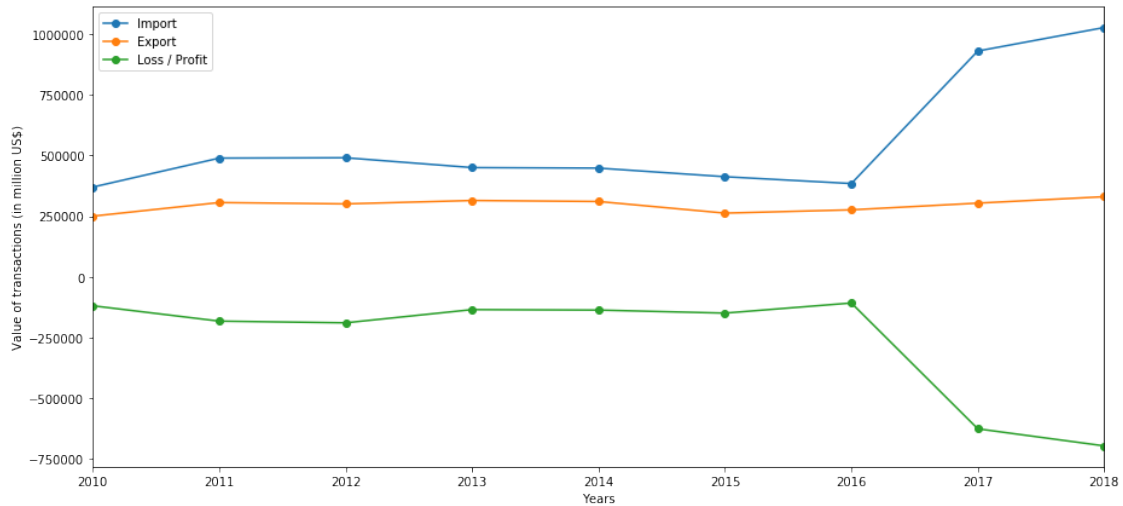
```
In [67]: Time_Series
```

```
Out[67]:
```

	Import	Export	Loss / Profit
2010	369762.25	249801.18	-119961.07
2011	489311.81	305948.28	-183363.53
2012	490730.07	300384.32	-190345.75
2013	450192.99	314388.61	-135804.38
2014	448026.63	310321.02	-137705.61
2015	412537.53	262274.30	-150263.23
2016	384350.29	275835.27	-108515.02
2017	931148.04	303507.85	-627640.19
2018	1028142.66	330058.64	-698084.02

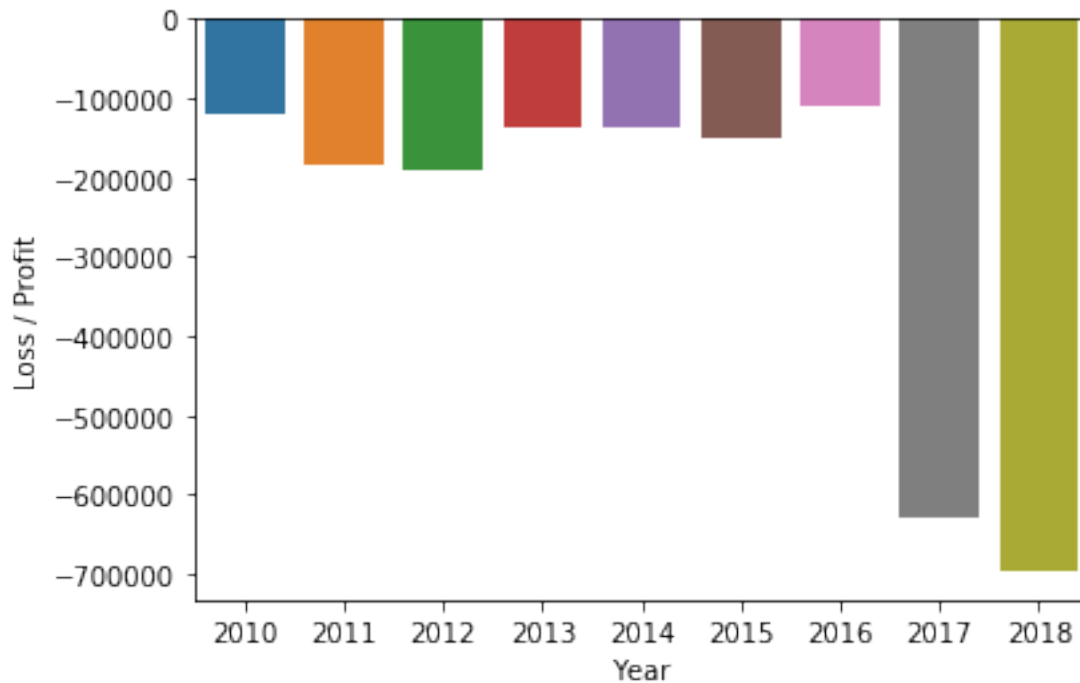
```
In [68]: fig, ax = plt.subplots(figsize=(15,7))
Time_Series.plot(ax=ax,marker='o')
ax.set_xlabel('Years')
ax.set_ylabel('Value of transactions (in million US$)')
# PLOTS HOW MUCH MONEY INDIA HAS MADE OVER THE LAST 10 YEARS FROM TRADE
```

```
Out[68]: Text(0, 0.5, 'Value of transactions (in million US$)')
```



```
In [69]: Time_Series.index.name = 'Year'
Time_Series.reset_index(inplace=True)
#LOSS INDIA HAS MADE FROM TRADE IN THE LAST 10 YEARS
```

```
In [70]: sns.barplot(x = 'Year', y = 'Loss / Profit', data = Time_Series)
plt.show()#LOSS INDIA HAS MADE FROM TRADE IN THE LAST 10 YEARS
```



```
In [71]: import_group=import_df.groupby(['Commodity','year']).agg({'value':'sum'})
export_group=export_df.groupby(['Commodity','year']).agg({'value':'sum'})
#GROUP BY COMMODITY
```

```
In [72]: import_group.sort_values(by = 'value',ascending = False).head(10)
```

```
Out[72]:
```

Commodity	year	value
MINERAL FUELS, MINERAL OILS AND PRODUCTS OF THE...	2018	335743.56
	2017	264589.04
	2013	181382.57
	2012	181344.64
	2011	172753.87
	2014	156399.99
NATURAL OR CULTURED PEARLS,PRECIOUS OR SEMIPREC...	2017	149420.82
	2018	129440.38
MINERAL FUELS, MINERAL OILS AND PRODUCTS OF THE...	2010	115929.10
	2015	107250.79

```
In [73]: import_temp=import_group.groupby(['Commodity']).agg({'value':'sum'})
export_temp=export_group.groupby(['Commodity']).agg({'value':'sum'}).loc[import_temp..]
```

```
In [74]: import_temp
```

```
Out[74]:
```

Commodity	value
-----------	-------

AIRCRAFT, SPACECRAFT, AND PARTS THEREOF.	65212.62
ALBUMINOIDAL SUBSTANCES; MODIFIED STARCHES; GLU...	4097.18
ALUMINIUM AND ARTICLES THEREOF.	42375.88
ANIMAL OR VEGETABLE FATS AND OILS AND THEIR CLE...	115462.40
ARMS AND AMMUNITION; PARTS AND ACCESSORIES THER...	414.10
ARTICLES OF APPAREL AND CLOTHING ACCESSORIES, K...	2884.66
ARTICLES OF APPAREL AND CLOTHING ACCESSORIES, N...	3991.15
ARTICLES OF IRON OR STEEL	45781.95
ARTICLES OF LEATHER,SADDLERY AND HARNESS;TRAVEL...	3750.24
ARTICLES OF STONE, PLASTER, CEMENT, ASBESTOS, M...	7561.06
BEVERAGES, SPIRITS AND VINEGAR.	6191.45
CARPETS AND OTHER TEXTILE FLOOR COVERINGS.	1041.29
CERAMIC PRODUCTS.	7077.66
CEREALS.	2830.67
CLOCKS AND WATCHES AND PARTS THEREOF.	3390.63
COCOA AND COCOA PREPARATIONS.	2378.61
COFFEE, TEA, MATE AND SPICES.	7257.63
COPPER AND ARTICLES THEREOF.	39889.03
CORK AND ARTICLES OF CORK.	65.21
COTTON.	9030.46
DAIRY PRODUCE; BIRDS' EGGS; NATURAL HONEY; EDIB...	807.00
EDIBLE FRUIT AND NUTS; PEEL OR CITRUS FRUIT OR ...	31111.30
EDIBLE VEGETABLES AND CERTAIN ROOTS AND TUBERS.	28442.52
ELECTRICAL MACHINERY AND EQUIPMENT AND PARTS TH...	427570.66
ESSENTIAL OILS AND RESINOIDS; PERFUMERY, COSMET...	8100.83
EXPLOSIVES; PYROTECHNIC PRODUCTS; MATCHES; PYRO...	109.04
FERTILISERS.	68717.35
FISH AND CRUSTACEANS, MOLLUSCS AND OTHER AQUATI...	896.99
FOOTWEAR, GAITERS AND THE LIKE; PARTS OF SUCH A...	5741.58
FURNITURE; BEDDING, MATTRESSES, MATTRESS SUPPOR...	15785.33
...	...
PREPARATIONS OF VEGETABLES, FRUIT, NUTS OR OTHE...	976.57
PREPARED FEATHERS AND DOWN AND ARTICLES MADE OF...	169.11
PRINTED BOOKDS, NEWSPAPERS, PICTURES AND OTHER ...	4782.72
PRODUCTS OF ANIMAL ORIGIN, NOT ELSEWHERE SPECIF...	457.45
PRODUCTS OF THE MILLING INDUSTRY; MALT; STARCHE...	662.15
PROJECT GOODS; SOME SPECIAL USES.	43677.74
PULP OF WOOD OR OF OTHER FIBROUS CELLULOSIC MAT...	19670.78
RAILWAY OR TRAMWAY LOCOMOTIVES, ROLLING-STOCK A...	4808.38
RAW HIDES AND SKINS (OTHER THAN FURSKINS) AND L...	6638.67
RESIDUES AND WASTE FROM THE FOOD INDUSTRIES; PR...	4651.84
RUBBER AND ARTICLES THEREOF.	37362.65
SALT; SULPHUR; EARTHS AND STONE; PLASTERING MAT...	28984.75
SHIPS, BOATS AND FLOATING STRUCTURES.	58294.94
SILK	2791.80
SOAP, ORGANIC SURFACE-ACTIVE AGENTS, WASHING PR...	7500.32
SPECIAL WOVEN FABRICS; TUFTED TEXTILE FABRICS; ...	1943.68
SUGARS AND SUGAR CONFECTIONERY.	8213.27

TANNING OR DYEING EXTRACTS; TANNINS AND THEIR D...	19103.44
TIN AND ARTICLES THEREOF.	2277.85
TOBACCO AND MANUFACTURED TOBACCO SUBSTITUTES.	481.77
TOOLS IMPLEMENTS, CUTLERY, SPOONS AND FORKS, OF...	10568.39
TOYS, GAMES AND SPORTS REQUISITES; PARTS AND AC...	5734.05
UMBRELLAS, SUN UMBRELLAS, WALKING-STICKS, SEAT-...	310.79
VEGETABLE PLAITING MATERIALS; VEGETABLE PRODUCT...	308.94
VEHICLES OTHER THAN RAILWAY OR TRAMWAY ROLLING ...	57415.76
WADDING, FELT AND NONWOVENS; SPACIAL YARNS; TWI...	2891.36
WOOD AND ARTICLES OF WOOD; WOOD CHARCOAL.	25978.66
WOOL, FINE OR COARSE ANIMAL HAIR, HORSEHAIR YAR...	4290.27
WORKS OF ART COLLECTORS' PIECES AND ANTIQUES.	1186.38
ZINC AND ARTICLES THEREOF.	6194.61

[98 rows x 1 columns]

```
In [ ]: HSCode=pd.DataFrame()
HSCode['Start']=[1,6,15,16,25,28,39,41,44,47,50,64,68,71,72,84,86,90,93,94,97]
HSCode['End']=[5,14,15,24,27,38,40,43,46,49,63,67,70,71,83,85,89,92,93,96,98]
HSCode['Sections']=['Animals & Animal Products',
'Vegetable Products',
'Animal Or Vegetable Fats',
'Prepared Foodstuffs',
'Mineral Products',
'Chemical Products',
'Plastics & Rubber',
'Hides & Skins',
'Wood & Wood Products',
'Wood Pulp Products',
'Textiles & Textile Articles',
'Footwear, Headgear',
'Articles Of Stone, Plaster, Cement, Asbestos',
'Pearls, Precious Or Semi-Precious Stones, Metals',
'Base Metals & Articles Thereof',
'Machinery & Mechanical Appliances',
'Transportation Equipment',
'Instruments - Measuring, Musical',
'Arms & Ammunition',
'Miscellaneous',
'Works Of Art',]

In [78]: import_df['Sections']=import_df["HSCode"]
export_df['Sections']=export_df["HSCode"]
for i in range(0,len(HSCode)):
    import_df.loc[(import_df["Sections"] >= HSCode['Start'][i]) & (import_df["Sections"] < HSCode['End'][i])] = 'Import'
    export_df.loc[(export_df["Sections"] >= HSCode['Start'][i]) & (export_df["Sections"] < HSCode['End'][i])] = 'Export'

In [79]: import_group=import_df.groupby(['Sections','year']).agg({'value':'sum'})
```

```

export_group=export_df.groupby(['Sections', 'year']).agg({'value': 'sum'})

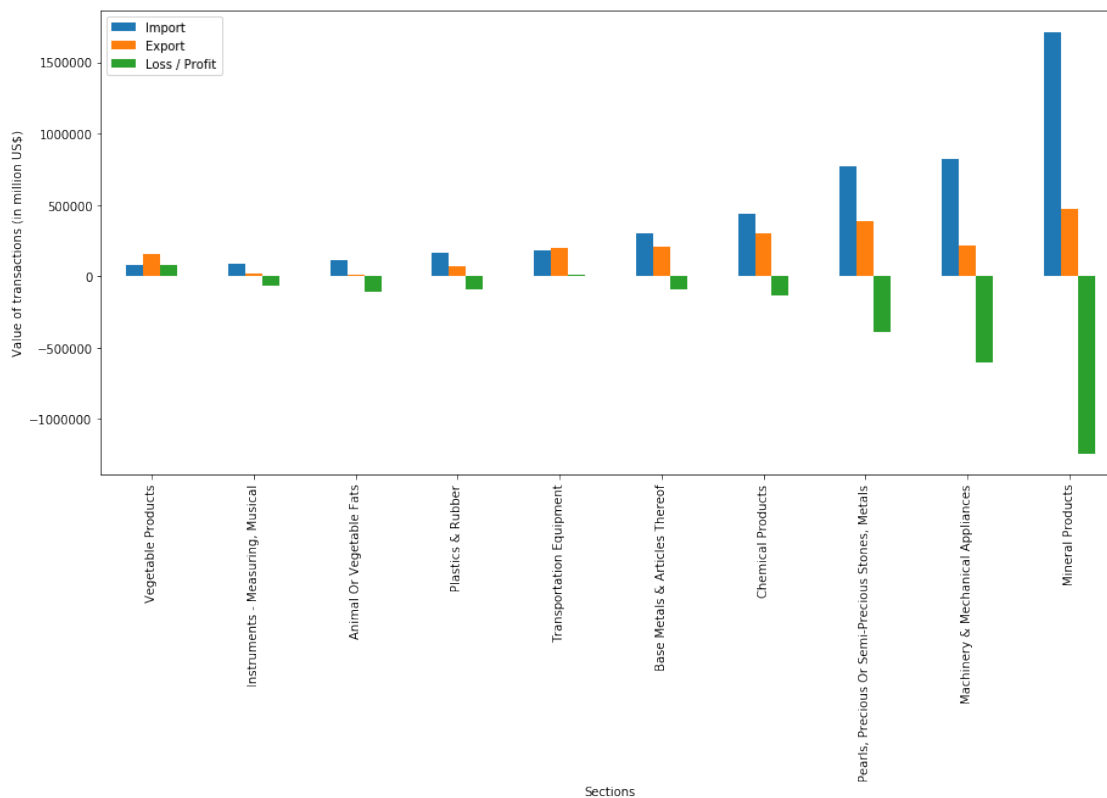
import_temp=import_group.groupby(['Sections']).agg({'value': 'sum'})
export_temp=export_group.groupby(['Sections']).agg({'value': 'sum'}).loc[import_temp.index]

In [80]: data_1=import_group.groupby(['Sections']).agg({'value': 'sum'}).sort_values(by='value')
data_2=export_temp
data_3=data_2-data_1
data_1.columns=['Import']
data_2.columns=['Export']
data_3.columns=['Loss / Profit']
df=pd.DataFrame(index=data_1.index.values)
#df=pd.concat([data_1,data_2,data_3])
df['Import']=data_1
df['Export']=data_2
df['Loss / Profit']=data_3

In [81]: df.index=HSCode['Sections'][data_1.index.values]
fig, ax = plt.subplots(figsize=(15,7))
df.plot(kind='bar',ax=ax)
ax.set_xlabel('Sections')
ax.set_ylabel('Value of transactions (in million US$)')

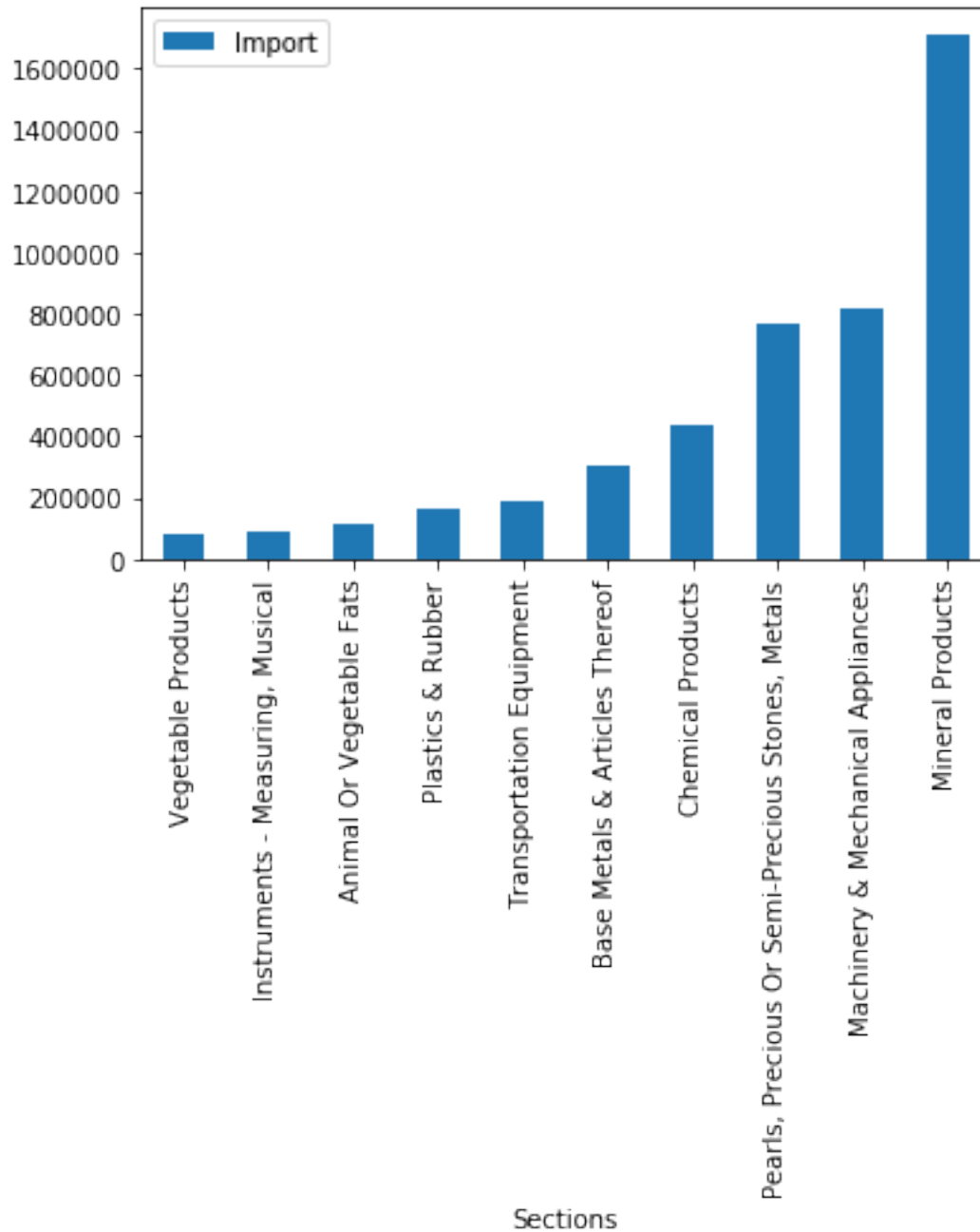
Out[81]: Text(0, 0.5, 'Value of transactions (in million US$)')

```



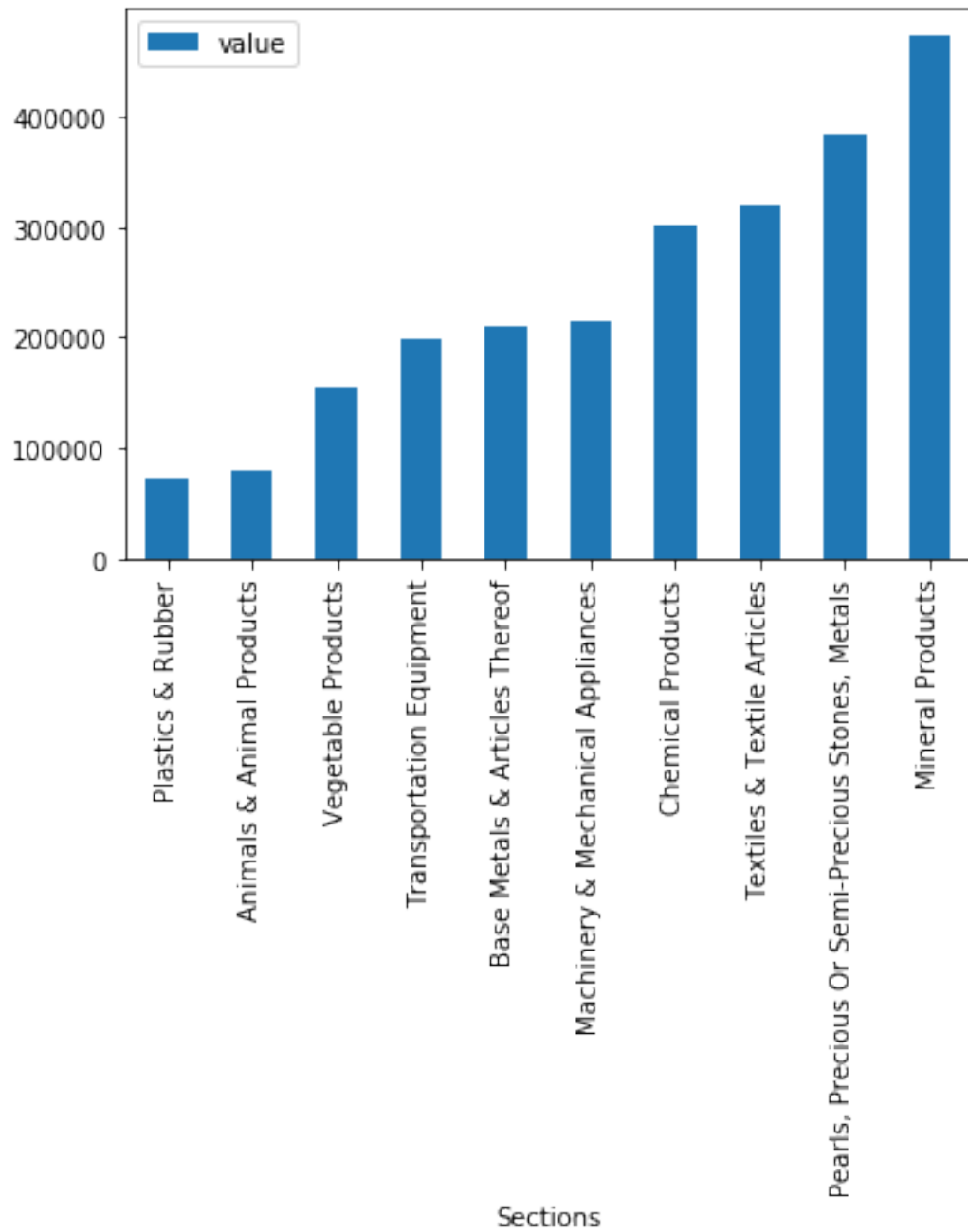
```
In [82]: data_1.index=HSCode['Sections'][data_1.index.values]
data_1.plot(kind='bar')
```

```
Out[82]: <matplotlib.axes._subplots.AxesSubplot at 0x19b2a4b9438>
```



```
In [83]: data_2=export_group.groupby(['Sections']).agg({'value':'sum'}).sort_values(by='value')
data_2.index=HSCode['Sections'][data_2.index.values]
data_2.plot(kind='bar')
```

Out[83]: <matplotlib.axes._subplots.AxesSubplot at 0x19b2ac8e9e8>



In []: