

## Experiment II [1]

August 17, 2025

```
[32]: import numpy as np # useful for many scientific computing in Python
import pandas as pd # primary data structure library
df_can = pd.read_excel('https://s3-api.us-geo.objectstorage.softlayer.net/
cf-courses-data/CognitiveClass/DV0101EN/labs/Data_Files/Canada.xlsx',
sheet_name='Canada by Citizenship',
skiprows=range(20),
skipfooter=2)

print ('Data read into a pandas dataframe!')
```

Data read into a pandas dataframe!

```
[33]: df_can.head()
```

```
[33]:      Type    Coverage        OdName  AREA AreaName   REG \
0  Immigrants  Foreigners  Afghanistan  935     Asia  5501
1  Immigrants  Foreigners       Albania  908   Europe  925
2  Immigrants  Foreigners       Algeria  903   Africa  912
3  Immigrants  Foreigners  American Samoa  909 Oceania  957
4  Immigrants  Foreigners        Andorra  908   Europe  925

                    RegName   DEV          DevName  1980 ...  2004  2005  2006 \
0    Southern Asia  902  Developing regions   16 ... 2978 3436 3009
1  Southern Europe  901  Developed regions    1 ... 1450 1223  856
2  Northern Africa  902  Developing regions   80 ... 3616 3626 4807
3        Polynesia  902  Developing regions    0 ...     0     0     1
4  Southern Europe  901  Developed regions    0 ...     0     0     1

  2007  2008  2009  2010  2011  2012  2013
0  2652  2111  1746  1758  2203  2635  2004
1   702   560   716   561   539   620   603
2  3623  4005  5393  4752  4325  3774  4331
3    0     0     0     0     0     0     0
4    1     0     0     0     0     1     1

[5 rows x 43 columns]
```

```
[34]: df_can.columns.values
```

```
[34]: array(['Type', 'Coverage', 'OdName', 'AREA', 'AreaName', 'REG', 'RegName',
       'DEV', 'DevName', 1980, 1981, 1982, 1983, 1984, 1985, 1986, 1987,
       1988, 1989, 1990, 1991, 1992, 1993, 1994, 1995, 1996, 1997, 1998,
       1999, 2000, 2001, 2002, 2003, 2004, 2005, 2006, 2007, 2008, 2009,
       2010, 2011, 2012, 2013], dtype=object)
```

```
[35]: print(type(df_can.columns))
print(type(df_can.index))
```

```
<class 'pandas.core.indexes.base.Index'>
<class 'pandas.core.indexes.range.RangeIndex'>
```

```
[36]: df_can.columns.tolist()
df_can.index.tolist()

print (type(df_can.columns.tolist()))
print (type(df_can.index.tolist()))
```

```
<class 'list'>
<class 'list'>
```

```
[37]: df_can.shape
```

```
[37]: (195, 43)
```

```
[38]: df_can.drop(['AREA', 'REG', 'DEV', 'Type', 'Coverage'], axis=1, inplace=True)
df_can.head(2)
```

```
[38]:      OdName AreaName           RegName           DevName  1980  1981 \
0  Afghanistan     Asia  Southern Asia  Developing regions    16    39
1      Albania   Europe  Southern Europe  Developed regions     1     0

      1982  1983  1984  1985 ...  2004  2005  2006  2007  2008  2009  2010 \
0      39     47     71    340 ...   2978   3436   3009   2652   2111   1746   1758
1       0      0      0      0 ...   1450   1223    856    702    560    716    561

      2011  2012  2013
0   2203   2635   2004
1    539    620    603

[2 rows x 38 columns]
```

```
[39]: df_can.rename(columns={'OdName': 'Country', 'AreaName': 'Continent', 'RegName':
                           'Region'}, inplace=True)
df_can.columns
```

```
[39]: Index(['Country', 'Continent', 'Region', 'DevName', 1980,
       1981, 1982, 1983, 1984, 1985,
       1986, 1987, 1988, 1989, 1990,
       1991, 1992, 1993, 1994, 1995,
       1996, 1997, 1998, 1999, 2000,
       2001, 2002, 2003, 2004, 2005,
       2006, 2007, 2008, 2009, 2010,
       2011, 2012, 2013],  
      dtype='object')
```

```
[40]: df_can['Total'] = df_can.apply(pd.to_numeric, errors='coerce').sum(axis=1)
```

```
[41]: df_can.columns
```

```
[41]: Index(['Country', 'Continent', 'Region', 'DevName', 1980,
       1981, 1982, 1983, 1984, 1985,
       1986, 1987, 1988, 1989, 1990,
       1991, 1992, 1993, 1994, 1995,
       1996, 1997, 1998, 1999, 2000,
       2001, 2002, 2003, 2004, 2005,
       2006, 2007, 2008, 2009, 2010,
       2011, 2012, 2013, 'Total'],  
      dtype='object')
```

```
[42]: df_can.isnull().sum()
```

```
[42]: Country      0
Continent     0
Region        0
DevName       0
1980          0
1981          0
1982          0
1983          0
1984          0
1985          0
1986          0
1987          0
1988          0
1989          0
1990          0
1991          0
1992          0
1993          0
1994          0
1995          0
1996          0
```

```
1997      0
1998      0
1999      0
2000      0
2001      0
2002      0
2003      0
2004      0
2005      0
2006      0
2007      0
2008      0
2009      0
2010      0
2011      0
2012      0
2013      0
Total     0
dtype: int64
```

```
[43]: df_can.describe()
```

```
[43]:          1980      1981      1982      1983      1984 \
count  195.000000  195.000000  195.000000  195.000000  195.000000
mean   508.394872  566.989744  534.723077  387.435897  376.497436
std    1949.588546 2152.643752 1866.997511 1204.333597 1198.246371
min    0.000000  0.000000  0.000000  0.000000  0.000000
25%    0.000000  0.000000  0.000000  0.000000  0.000000
50%    13.000000  10.000000  11.000000  12.000000  13.000000
75%    251.500000  295.500000  275.000000  173.000000  181.000000
max   22045.000000 24796.000000 20620.000000 10015.000000 10170.000000

          1985      1986      1987      1988      1989 \
count  195.000000  195.000000  195.000000  195.000000  195.000000
mean   358.861538  441.271795  691.133333  714.389744  843.241026
std    1079.309600 1225.576630 2109.205607 2443.606788 2555.048874
min    0.000000  0.000000  0.000000  0.000000  0.000000
25%    0.000000  0.500000  0.500000  1.000000  1.000000
50%    17.000000  18.000000  26.000000  34.000000  44.000000
75%    197.000000 254.000000 434.000000 409.000000 508.500000
max   9564.000000 9470.000000 21337.000000 27359.000000 23795.000000

...          2005      2006      2007      2008 \
count ...  195.000000  195.000000  195.000000  195.000000
mean ... 1320.292308 1266.958974 1191.820513 1246.394872
std ... 4425.957828 3926.717747 3443.542409 3694.573544
min ... 0.000000 0.000000 0.000000 0.000000
```

25%	...	28.500000	25.000000	31.000000	31.000000	
50%	...	210.000000	218.000000	198.000000	205.000000	
75%	...	832.000000	842.000000	899.000000	934.500000	
max	...	42584.000000	33848.000000	28742.000000	30037.000000	
		2009	2010	2011	2012	2013 \
count		195.000000	195.000000	195.000000	195.000000	195.000000
mean		1275.733333	1420.287179	1262.533333	1313.958974	1320.702564
std		3829.630424	4462.946328	4030.084313	4247.555161	4237.951988
min		0.000000	0.000000	0.000000	0.000000	0.000000
25%		36.000000	40.500000	37.500000	42.500000	45.000000
50%		214.000000	211.000000	179.000000	233.000000	213.000000
75%		888.000000	932.000000	772.000000	783.000000	796.000000
max		29622.000000	38617.000000	36765.000000	34315.000000	34129.000000
		Total				
count		195.000000				
mean		32867.451282				
std		91785.498686				
min		1.000000				
25%		952.000000				
50%		5018.000000				
75%		22239.500000				
max		691904.000000				

[8 rows x 35 columns]

[44]: df\_can.Country

[44]: 0 Afghanistan  
1 Albania  
2 Algeria  
3 American Samoa  
4 Andorra  
...  
190 Viet Nam  
191 Western Sahara  
192 Yemen  
193 Zambia  
194 Zimbabwe  
Name: Country, Length: 195, dtype: object

[45]: df\_can[['Country', 1980, 1981, 1982, 1983, 1984, 1985]]

[45]:

	Country	1980	1981	1982	1983	1984	1985
0	Afghanistan	16	39	39	47	71	340
1	Albania	1	0	0	0	0	0

```

2          Algeria    80    67    71    69    63    44
3      American Samoa    0     1     0     0     0     0
4          Andorra     0     0     0     0     0     0
..
190        Viet Nam  1191  1829  2162  3404  7583  5907
191  Western Sahara    0     0     0     0     0     0
192          Yemen     1     2     1     6     0    18
193          Zambia    11    17    11     7    16     9
194        Zimbabwe    72   114   102    44    32    29

```

[195 rows x 7 columns]

[46]: df\_can.set\_index('Country', inplace=True)

[47]: df\_can.head(3)

	Continent	Region	DevName	1980	1981	1982	\					
Country												
Afghanistan	Asia	Southern Asia	Developing regions	16	39	39						
Albania	Europe	Southern Europe	Developed regions	1	0	0						
Algeria	Africa	Northern Africa	Developing regions	80	67	71						
	1983	1984	1985	1986	...	2005	2006	2007	2008	2009	2010	\
Country					...							
Afghanistan	47	71	340	496	...	3436	3009	2652	2111	1746	1758	
Albania	0	0	0	1	...	1223	856	702	560	716	561	
Algeria	69	63	44	69	...	3626	4807	3623	4005	5393	4752	
	2011	2012	2013	Total								
Country												
Afghanistan	2203	2635	2004	58639.0								
Albania	539	620	603	15699.0								
Algeria	4325	3774	4331	69439.0								

[3 rows x 38 columns]

[48]: df\_can.index.name = None # optional: to remove the name of the index

[49]: #To avoid this ambiguity, let's convert the column names into strings: '1980' ↪ to '2013'.

```

df_can.columns = list(map(str, df_can.columns))
[print(type(x)) for x in df_can.columns.values] #-- uncomment to check type of column headers

```

```

<class 'str'>
<class 'str'>
<class 'str'>

```

```
<class 'str'>
```

[49]: [None,  
 None,  
 None,

```
None,  
None]
```

```
[50]: # useful for plotting later on  
years = list(map(str, range(1980, 2014)))  
years
```

```
[50]: ['1980',  
       '1981',  
       '1982',  
       '1983',  
       '1984',  
       '1985',  
       '1986',  
       '1987',  
       '1988',  
       '1989',  
       '1990',  
       '1991',  
       '1992',  
       '1993',  
       '1994',
```

```
'1995',
'1996',
'1997',
'1998',
'1999',
'2000',
'2001',
'2002',
'2003',
'2004',
'2005',
'2006',
'2007',
'2008',
'2009',
'2010',
'2011',
'2012',
'2013']
```

```
[51]: %matplotlib inline
```

```
import matplotlib as mpl
import matplotlib.pyplot as plt
```

```
[52]: print ('Matplotlib version: ', mpl.__version__ )
```

```
Matplotlib version:  3.9.2
```

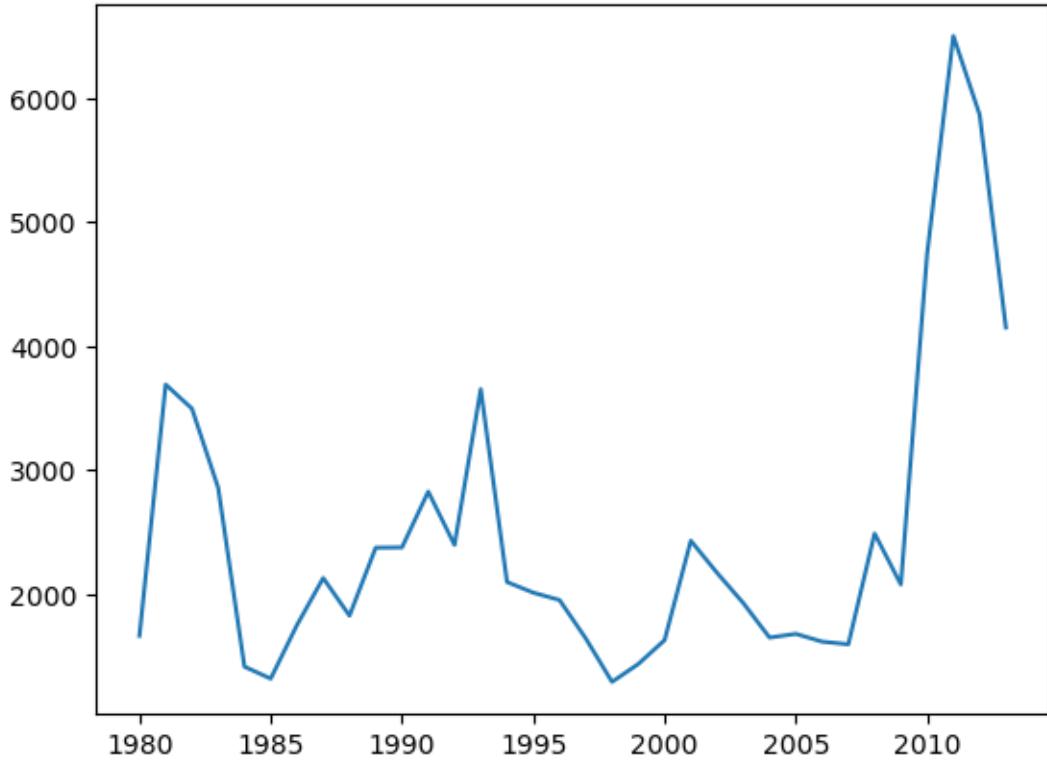
```
[53]: haiti = df_can.loc['Haiti', years] # Passing in years 1980 - 2013 to exclude
       ↴the 'total' column
haiti.head()
```

```
[53]: 1980    1666
      1981    3692
      1982    3498
      1983    2860
      1984    1418
Name: Haiti, dtype: object
```

```
[54]: haiti.plot()
```

```
[54]: <Axes: >
```

```
[55]: plt.show()
```

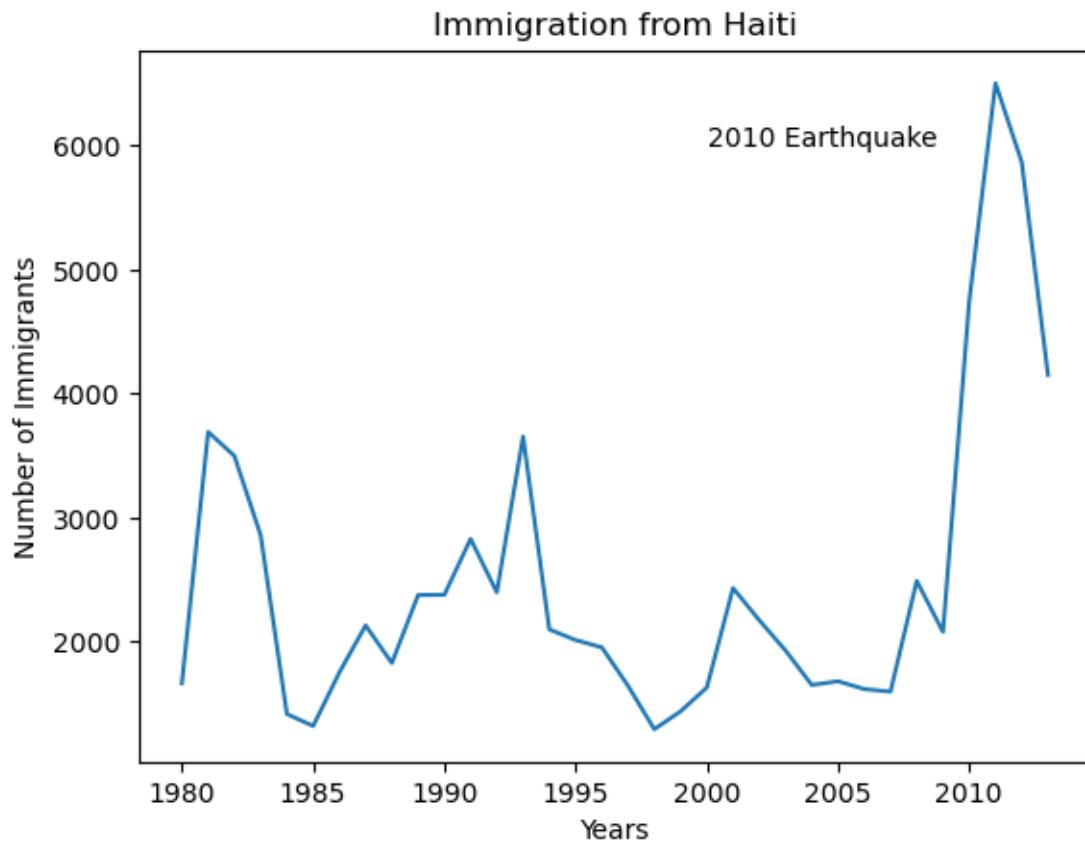


```
[56]: haiti.plot(kind='line')

plt.title('Immigration from Haiti')
plt.ylabel('Number of Immigrants')
plt.xlabel('Years')

# annotate the 2010 Earthquake.
# syntax: plt.text(x, y, label)
plt.text(20, 6000, '2010 Earthquake') # see note below

plt.show()
```



[57]: #QUESTION :: Let us compare the number of immigrants from India and China from 1980 to 2013.

#Step 1: Get the data set for China and India.

```
df_CI = df_can.loc[['India', 'China'], years]
df_CI.head()
```

[57]:

	1980	1981	1982	1983	1984	1985	1986	1987	1988	1989	...	\
India	8880	8670	8147	7338	5704	4211	7150	10189	11522	10343	...	
China	5123	6682	3308	1863	1527	1816	1960	2643	2758	4323	...	

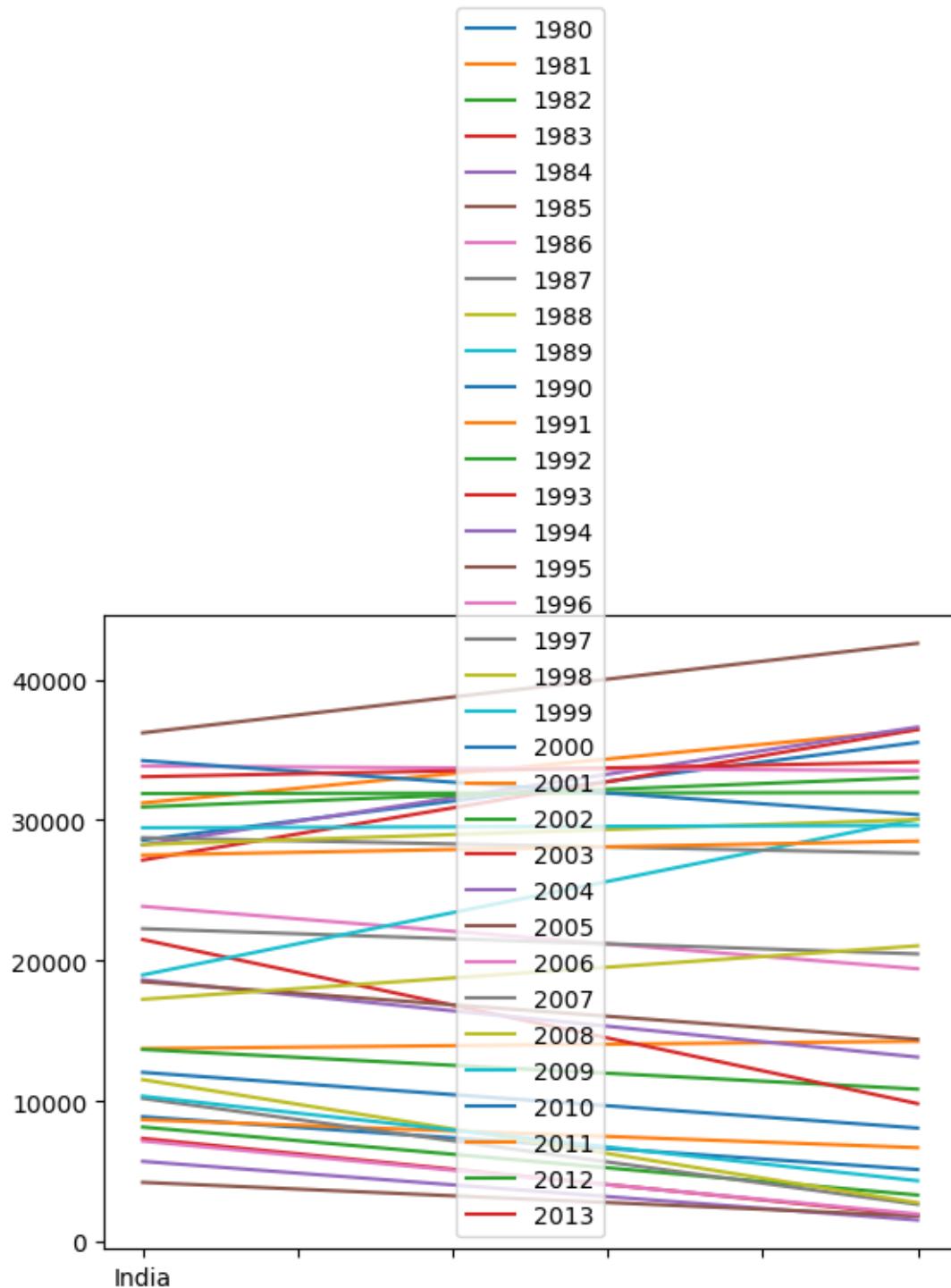
	2004	2005	2006	2007	2008	2009	2010	2011	2012	2013
India	28235	36210	33848	28742	28261	29456	34235	27509	30933	33087
China	36619	42584	33518	27642	30037	29622	30391	28502	33024	34129

[2 rows x 34 columns]

[58]: # Step 2: Plot graph. We will explicitly specify line plot by passing in kind parameter to plot().

```
df_CI.plot(kind='line')
```

```
plt.show()
```



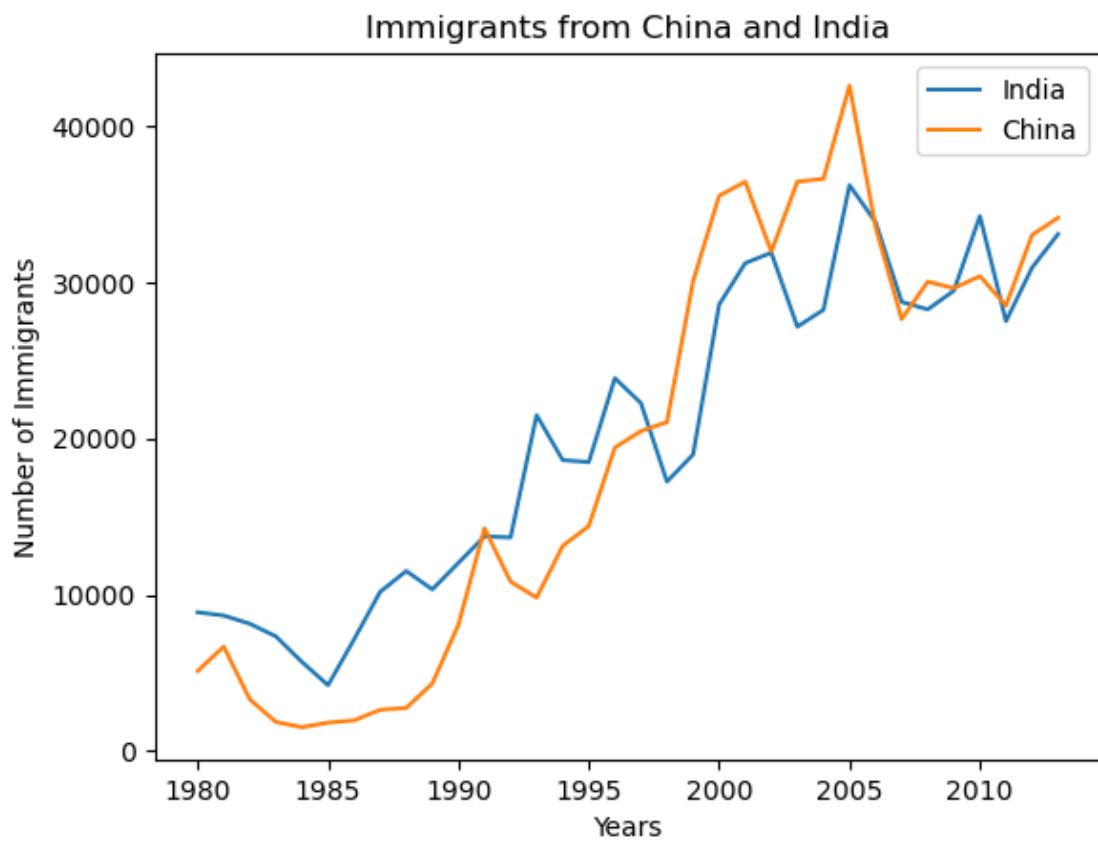
```
[59]: df_CI = df_CI.transpose()
df_CI.head()
```

```
[59]:    India  China
1980    8880   5123
1981    8670   6682
1982    8147   3308
1983    7338   1863
1984    5704   1527
```

```
[60]: df_CI.plot(kind='line')

plt.title('Immigrants from China and India')
plt.ylabel('Number of Immigrants')
plt.xlabel('Years')

plt.show()
```



```
[61]: # Question: Compare the trend of top 5 countries that contributed the most to immigration to Canada.
```

```

# Step 1: Get the dataset. Recall that we created a Total column that
# calculates the cumulative immigration
# by country. We will sort on this column to get our top 5 countries using
# pandas sort_values() method.

# inplace = True parameter saves the changes to the original df_can dataframe
df_can.sort_values(by='Total', ascending=False, axis=0, inplace=True)

# get the top 5 entries
df_top5 = df_can.head(5)

# transpose the dataframe
df_top5 = df_top5[years].transpose()

df_top5

```

	India	China	United Kingdom of Great Britain and Northern Ireland	\
1980	8880	5123		22045
1981	8670	6682		24796
1982	8147	3308		20620
1983	7338	1863		10015
1984	5704	1527		10170
1985	4211	1816		9564
1986	7150	1960		9470
1987	10189	2643		21337
1988	11522	2758		27359
1989	10343	4323		23795
1990	12041	8076		31668
1991	13734	14255		23380
1992	13673	10846		34123
1993	21496	9817		33720
1994	18620	13128		39231
1995	18489	14398		30145
1996	23859	19415		29322
1997	22268	20475		22965
1998	17241	21049		10367
1999	18974	30069		7045
2000	28572	35529		8840
2001	31223	36434		11728
2002	31889	31961		8046
2003	27155	36439		6797
2004	28235	36619		7533
2005	36210	42584		7258
2006	33848	33518		7140
2007	28742	27642		8216
2008	28261	30037		8979

2009	29456	29622	8876
2010	34235	30391	8724
2011	27509	28502	6204
2012	30933	33024	6195
2013	33087	34129	5827

	Philippines	Pakistan
1980	6051	978
1981	5921	972
1982	5249	1201
1983	4562	900
1984	3801	668
1985	3150	514
1986	4166	691
1987	7360	1072
1988	8639	1334
1989	11865	2261
1990	12509	2470
1991	12718	3079
1992	13670	4071
1993	20479	4777
1994	19532	4666
1995	15864	4994
1996	13692	9125
1997	11549	13073
1998	8735	9068
1999	9734	9979
2000	10763	15400
2001	13836	16708
2002	11707	15110
2003	12758	13205
2004	14004	13399
2005	18139	14314
2006	18400	13127
2007	19837	10124
2008	24887	8994
2009	28573	7217
2010	38617	6811
2011	36765	7468
2012	34315	11227
2013	29544	12603

```
[62]: # Step 2: Plot the dataframe. To make the plot more readable, we will change the size using the figsize parameter.
df_top5.plot(kind='line', figsize=(14, 8)) # pass a tuple (x, y) size

plt.title('Immigration Trend of Top 5 Countries')
```

```
plt.ylabel('Number of Immigrants')
plt.xlabel('Years')

plt.show()
```

