

```
In [3]: #Innovation In Education By Oybek, Python Project leader

print("Innovation in education across USA, China, and S
```

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litteracy rates across USA, China, and Germany
print(' the following data frames and graphs below show how the literacy rates vary between USA,China and Germany')
import numpy as np
import matplotlib.pyplot as plt
import pandas as pd

dataframe = pd.read_csv('cross-country-literacy-rates two.csv')

dataframe

dataframe.sort_values('rates_of_Literacy', ascending=False)

import re

dataframe.loc[dataframe['COUNTRY_Code'].str.contains('United States USA', flags=re.I, regex=True)]

innovation in education across USA, China, and Germany: world need ideas and innovation to deal progressively with the
major problems the humanity face.Countries should create opportunities and conditions to cultivate such creative and
talented people who contribute to this important work. Let us look at the human resources/quality of those three coun
tries

the following data frames and graphs below show how the literacy rates vary between USA,China and Germany

Out[3]:
COUNTRY_Code  Year  rates_of_Literacy
10  United States USA  1870  80.000000
11  United States USA  1880  83.000000
12  United States USA  1890  86.000000
13  United States USA  1900  89.300003
14  United States USA  1910  92.300003
15  United States USA  1920  94.000000
16  United States USA  1930  95.699997
17  United States USA  1940  97.099998
18  United States USA  1947  97.300003
19  United States USA  1950  96.800003
20  United States USA  1952  97.500000
21  United States USA  1959  97.800003
22  United States USA  1969  99.000000
23  United States USA  1979  99.400002
24  United States USA  2003  99.000000

In [4]:
# Let us Compare Literacy rates across the US, China and Germany
US = dataframe[ dataframe.COUNTRY_Code == 'United States USA' ]
China = dataframe[ dataframe.COUNTRY_Code == 'China CHN' ]
Germany = dataframe[ dataframe.COUNTRY_Code == 'Germany DEU' ]
Germany

Out[4]:
COUNTRY_Code  Year  rates_of_Literacy
0  Germany DEU  1475  9.0
1  Germany DEU  1550  16.0
2  Germany DEU  1650  31.0
3  Germany DEU  1750  38.0
4  Germany DEU  2003  99.0

In [5]:
plt.figure(figsize=(9,7))
plt.plot(US.Year, US.rates_of_Literacy, 'b..')


```

```
plt.plot(china_year, china_rates_of_literacy, 'g.-')
plt.plot(germany_year, germany_rates_of_literacy, 'b.-')
plt.legend(['The US', 'China', 'Germany'])
plt.xlabel('Year')
```

```
plt.savefig('rates_of_Literacy_figure.png', dpi= 300)
plt.show()
```

Country	Percentage of population aged 65 and over in 2010
The US	~12%
China	~10%
Germany	~22%

Year	Percentage of Literacy
1990	65
1991	70
1992	75
1993	80
1994	85
1995	90
1996	95
1997	98
1998	100
1999	100
2000	100



The graph displays a line with circular markers representing literacy rates over time. The x-axis is labeled 'Year' and ranges from 1500 to 2000 in increments of 100. The y-axis ranges from 0 to 20. The data points are approximately as follows:

Year	Literacy Rate (approx.)
1500	10
1550	12
1600	15
1650	18
1700	20

```
In [6]: #as we can see from these graphs, US was leader for gaining higher literacy rates within the past decades

In [7]: import numpy as np
import matplotlib.pyplot as plt
import pandas as pd
researchers = pd.read_csv('researchers-in-rd-per-million-peopleII.csv')
```

```
researchers.sort_values('Researchers_in_R_and_D(per_million_people)', ascending=False)
import re
```

```
Out[7]:
```

	country	Code	Year	Researchers_in_R_and_D_(per_million_people)
40.	United States	USA	1998	2123.57201

41	United States	USA	1997	3224.23913
42	United States	USA	1998	3388.20888

44	United States	USA	2000	3475.69505
45	United States	USA	2001	3545.83215

46	United States	USA	2002	3630.0162
47	United States	USA	2003	3870.5696
48	United States	USA	2004	3765.10261
49	United States	USA	2005	3718.19505
50	United States	USA	2006	3781.80788
51	United States	USA	2007	3757.86698
52	United States	USA	2008	3911.53450
53	United States	USA	2009	4073.17633
54	United States	USA	2010	3868.56644
55	United States	USA	2011	4011.32862
56	United States	USA	2012	4015.88708

```

57 United States USA 2013 4117.67409
58 United States USA 2014 4231.98928

In [8]: plt.style.use('default')

plt.figure(figsize=(8,7))
Germany = researchers.loc[researchers.country == 'Germany']['Researchers_in_R_and_D_(per_million_people)']
United_States = researchers.loc[researchers.country == 'United States']['Researchers_in_R_and_D_(per_million_people)']
China = researchers.loc[researchers.country == 'China']['Researchers_in_R_and_D_(per_million_people)']

labels = ['Germany', 'United States', 'China']
boxes = plt.boxplot([Germany, United_States, China], labels=labels, patch_artist=True, medianprops={'linewidth':3})
for box in boxes['boxes']:
    box.set(color='r', linewidth=3)

plt.title('Researchers in research & development across countries ( USA, China, and Germany)')

```

```
researchers_in_mil = researchers_in_mil / 1000000
plt.xlabel('countries')

plt.savefig('Researchers in research & development across countries ( USA, China, and Germany).png', dpi=300)
plt.show()
```

Researchers in research & development across countries (USA, China, and Germany)

Country	Min	Q1	Median	Mean	Q3	Max
USA	3300	3400	3450	3400	4150	4400
China	3400	3550	3800	3750	4000	4250

Country	Researchers in R&D (per million people)
Israel	~3300
South Korea	~3100
Sweden	~2900
Switzerland	~2800
Finland	~2700
Denmark	~2600
Norway	~2500
Netherlands	~2400
Germany	~2300
Japan	~2200

Country	Min	Q1	Median	Q3	Max
Germany	~350	~380	~400	~420	~450
United States	~350	~380	~400	~420	~450
China	~450	~650	~750	~850	~950

In [9]:

```

#The OECD Programme for International Student Assessment (PISA) examines what students know in reading, mathematics &
#IT provides the most comprehensive and rigorous international assessment of student learning outcomes to date.
#Results from PISA indicate the quality and equity of learning outcomes attained around the world, and allow educators
#NOTE: The Reading, Mathematics and Science scale ranges from 0 to 1000. Some apparent differences between estimates
#SOURCE: Organization for Economic Cooperation and Development (OECD), Program for International Student Assessment (PISA)

```

```
print("women vs men PISA reading performances across countries")
PISA = pd.read_csv('Averages for age 15 years PISA reading scale.csv')
PISA
```

women vs men PISA reading performances across countries

	Year	Country	average	Standard_Error
0	2018	Germany	512	(3.2)
1	2018	United States	517	(3.6)
2	2018	Hong Kong (China)	542	(2.8)
3	2015	Germany	520	(3.1)
4	2015	United States	507	(3.9)

5	2015	Hong Kong (China)	541	(3.6)
6	2012	Germany	530	(3.1)
7	2012	United States	513	(3.8)
8	2012	Hong Kong (China)	558	(3.3)
9	2009	Germany	518	(2.9)
10	2009	United States	513	(3.8)
11	2009	Hong Kong (China)	550	(2.8)
12	2006	Germany	517	(4.4)
13	2006	United States	515	(3.8)
14	2006	Hong Kong (China)	551	(3.0)
15	2003	Germany	542	(2.9)

16	2003	United States	511	(3.5)
17	2003	Hong Kong (China)	525	(3.5)
18	2000	Germany	502	(3.9)
19	2000	United States	518	(6.2)
20	2000	Hong Kong (China)	533	(3.6)


```

In [10]: plt.hist(PISA.average, color='#3e28a8')
plt.yticks([0, 2, 3, 5])
plt.xlabel('Average scores')
plt.ylabel('Frequency')
plt.title('distribution of average scores across countries for women PISA performance')
plt.savefig('distribution of average scores across countries for women PISA performance.png', dpi= 300)

```

distribution of average scores across countries for women PISA performance

average score	Frequency
500	5



```
In [11]: PISAmen = pd.read_csv('Averages for age 15 years PISA reading scale ( male).csv')
PISAmen
PISAmen.sort_values('average', ascending=False)
```

Year	Country	average	Standard Error
8	2012 Hong Kong (China)	533	(3.8)
14	2006 Hong Kong (China)	520	(3.5)
20	2000 Hong Kong (China)	518	(4.8)
11	2009 Hong Kong (China)	518	(3.3)
5	2015 Hong Kong (China)	513	(3.4)
2	2018 Hong Kong (China)	507	(3.5)
3	2015 Germany	499	(3.7)
17	2003 Hong Kong (China)	494	(5.3)
1	2018 United States	494	(4.2)

19	2006	United States	490	(6.4)
13	2006	United States	490	(4.1)
10	2009	United States	488	(4.2)
4	2015	United States	487	(3.7)
6	2012	Germany	486	(2.9)
0	2018	Germany	486	(3.4)
7	2012	United States	482	(4.1)
16	2003	United States	479	(3.7)
9	2009	Germany	478	(3.6)
12	2006	Germany	475	(5.3)
15	2003	Germany	471	(4.2)

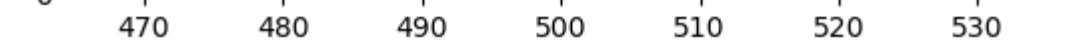
```

18 2000 Germany 468 (3.2)
n [12]: plt.hist(PISAmen.average, color='#3e28a8')
plt.yticks([0, 2, 3, 5])
plt.xlabel('Average scores')
plt.ylabel('Frequency')
plt.title('distribution of average scores across countries for men PISA performance')
plt.savefig('distribution of average scores across countries for men PISA performance.png', dpi=300)
plt.show()

```

distribution of average scores across countries for men PISA performance

Number of children	Frequency
0	2
1	3
2	4
3	3
4	2
5	1



```

In [13]: # mean PISA performance in The science, math and reading across countries
print('mean reading performance across countries')
mean_PISA_reading = pd.read_csv('PISA mean reading performance.csv')
mean_PISA_reading

import re

mean_PISA_reading.loc[mean_PISA_reading['country'].str.contains('United States', flags=re.I, regex=True)]

mean reading performance across countries
country_year_PISA_Mean performance on the reading scale

```

6	United States	2000	504.419691
7	United States	2003	495.182412
8	United States	2009	499.262821
9	United States	2012	497.581718
10	United States	2015	496.935100

```

In [14]: plt.style.use('default')

plt.figure(figsize=(8,7))
Germany = mean_PISA_reading.loc[mean_PISA_reading.country == 'Germany']
United_States = mean_PISA_reading.loc[mean_PISA_reading.country == 'United States']
China = mean_PISA_reading.loc[mean_PISA_reading.country == 'China']

```

```
boxes = plt.boxplot([Germany, United_States, China], labels=labels, patch_artist=True, medianprops={'linewidth': 3})
for box in boxes['boxes']:
    box.set(color='#1557d4', linewidth=3)
plt.title('comparison of mean PISA reading performance across USA, China, Germany')
plt.ylabel('PISA: Mean performance on the reading scale')
plt.savefig('comparison of mean PISA reading performance across USA,China,Germany.png', dpi= 300)

plt.show()
```

A boxplot showing the distribution of reading performance for the 'none' group. The y-axis is labeled 'in performance on the reading scale' and has tick marks at 520 and 540. The box is blue, with the median line at approximately 525. The interquartile range (IQR) spans from approximately 523 to 527. Whiskers extend from the box to approximately 518 (lower) and 532 (upper). There are no outliers.

A box plot comparing PISA Math scores across three countries: Germany, United States, and China. The y-axis is labeled 'PISA Math' and ranges from 480 to 500. Germany has a median around 495, United States around 498, and China around 498. The United States has an outlier above 500.

Country	Min	Q1	Median	Q3	Max	Outliers
Germany	482	492	495	498	502	None
United States	495	497	498	499	500	502
China	492	495	498	500	502	None

```
print('mean math performance across countries')
mean_PISA_math = pd.read_csv('PISA mean math performance.csv')
mean_PISA_math
```

```

# mean math performance across countries
out[15]:

```

	Year	Country	Score
9	2000	Germany	490.000000
10	2003	Germany	502.985532
11	2006	Germany	503.790859
12	2009	Germany	512.777643
13	2012	Germany	513.525056
14	2015	Germany	505.971300

```
[16]: mean_PISA_math.pivot(index='Year', columns='country', values='PISA: Mean performance on the mathematics scale').plot.xlabel('Year')
plt.ylabel('PISA: Mean performance on the math scale')
plt.title('Mean math performance across USA, China, and Germany')
plt.legend(loc='upper right')
```

```
plt.savefig('Mean math performance across USA,China, and Germany.png', dpi=300)
plt.show()
```

Category	China	Germany	United States
1	490	485	495
2	500	480	485
3	500	475	475
4	600	510	490
5	610	500	475

Category	PISA Mean performance on reading literacy
EU average	~500
EU countries with the highest performance	~540
EU countries with the lowest performance	~460

```

n [17]: print('mean science performance across countries')
        mean_PISA_science = pd.read_csv('PISA mean science performance.csv')
        mean_PISA_science

mean science performance across countries
out[17]:
   country  Year  PISA: Mean performance on the science scale
0  United States  2000  489.460132
1  United States  2003  491.263463
2  United States  2006  488.906837

```

4	United States	2012	497.409811
5	United States	2015	496.242400
6	China	2009	574.617355
7	China	2012	580.117831
8	China	2015	517.779300
9	Germany	2000	487.105836
10	Germany	2003	502.336476
11	Germany	2006	515.649130
12	Germany	2009	520.405349
13	Germany	2012	524.120799

```

14 Germany 2015 509.140600
n [18]: mean_PISA_science.pivot(index='Year', columns='country', values='PISA: Mean performance on the science scale').plot()
      plt.xlabel('Year')
      plt.ylabel('PISA: Mean performance on the science scale')
      plt.title('Mean science performance across USA,China, and Germany')
      plt.legend(loc='upper right')


      plt.savefig('Mean science performance across USA,China, and Germany.png', dpi=300)
      plt.show()

```

Country	Mean PISA Science Performance (2015)
China	509.14
Germany	500.00

A bar chart comparing the performance of the United States (green bars) and the United Kingdom (orange bars) on the science scale for different age groups. The y-axis represents the performance score, ranging from 200 to 500. The x-axis lists the age groups: 13-17, 18-24, 25-34, 35-44, 45-54, 55-64, and 65+.

Age Group	United States	United Kingdom
13-17	500	485
18-24	490	505
25-34	490	515
35-44	505	525
45-54	500	515
55-64	500	505
65+	495	505



Year	Digital device usage (orange)	Digital device usage (green)
2000	100	100
2003	100	100
2006	100	100
2009	100	100
2012	100	100
2015	100	100

```
[19]: Digital_device_usage = pd.read_csv('using digital devices.csv')
      Digital_device_usage
```

```
[19]:
```

	Country	Year	15_year_old_students_using_digital_devices_for_practising_and_drilling_(%_of_students)
0	Germany	2006	26
1	Germany	2015	32
2	Hong Kong China	2006	38
3	Hong Kong China	2015	34

```
[20]:
```

```
Germany = Digital_device_usage[ Digital_device_usage.Country == 'Germany' ]
Hong_Kong_China = Digital_device_usage[ Digital_device_usage.Country == 'Hong Kong China' ]

Germany_students = Digital_device_usage.loc[Digital_device_usage.Country == 'Germany']
Hong_Kong_China_students = Digital_device_usage.loc[Digital_device_usage.Country == 'Hong Kong China']

plt.figure(figsize=(10,8))
plt.plot(Germany_year,Germany_students,'b-')
plt.plot(Hong_Kong_China_year,Hong_Kong_China_students,'r-')
```

```
plt.title('Digital device usage by students for practising and drilling across Germany and Hong Kong,China')
plt.legend(['Germany', 'Hong Kong, China', ])
plt.xlabel('Year')
plt.ylabel('15_year_old_students_using_digital_devices_for_practising_and_drilling_(%_of_students)')
plt.savefig('Digital device usage by students for practising and drilling across Germany and Hong Kong,China.png', dpi=100)
plt.show()
```

days	no	yes
1	36.2	31.5
3	34.0	32.0

Year	Percentage of Students Using Digital Devices
2015	26.1
2016	26.5
2017	27.0
2018	27.5
2019	28.0
2020	28.5

Year	Total (%)	Male (%)	Female (%)
2006	18	16	20
2007	19	17	21
2008	20	18	22
2009	21	19	23
2010	22	20	24
2011	23	21	25
2012	24	22	26
2013	25	23	27
2014	26	24	28

In []:

In []:

In []:

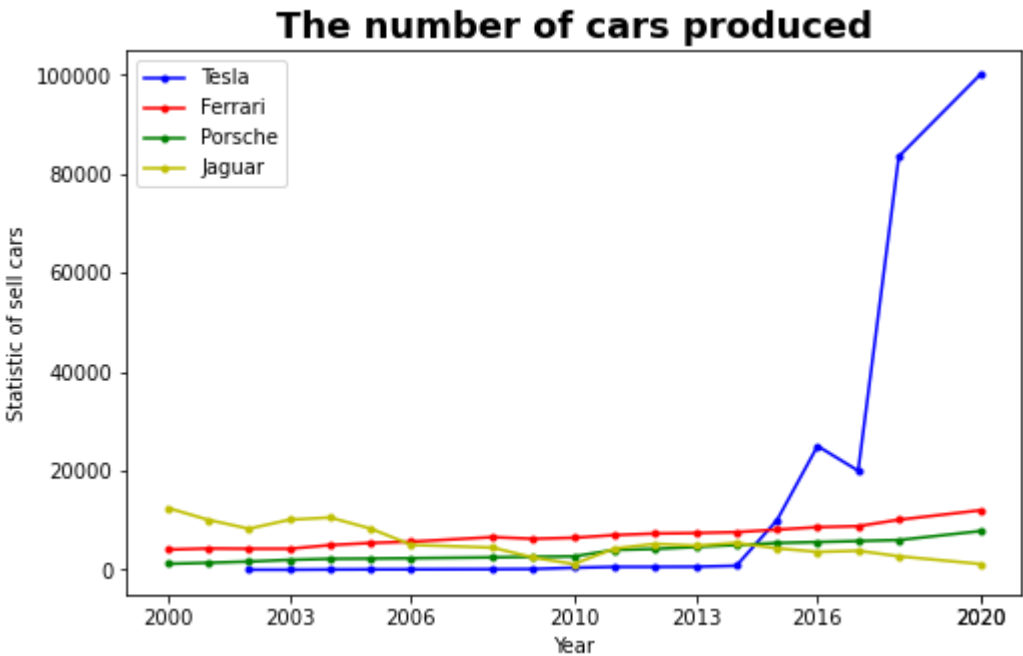
In [1]: `# this part done by Jahongir.U`

```
import matplotlib.pyplot as plt
import numpy as np
import pandas as pd
```

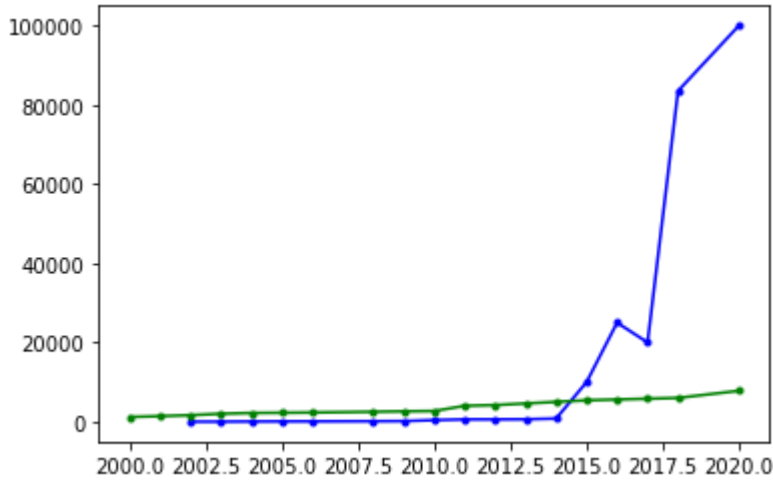
In [10]: `car = pd.read_csv(r'D:\Фотографии\statisti_of_car1.csv')`
`print(car)`

	Year	Tesla	Ferrari	Porsche	Jaguar
0	2000	NaN	4070	1200	12492
1	2001	NaN	4289	1400	10082
2	2002	1.0	4236	1650	8286
3	2003	2.0	4238	2000	10102
4	2004	45.0	4975	2200	10552
5	2005	70.0	5409	2250	8304
6	2006	80.0	5671	2320	5027
7	2008	110.0	6587	2500	4474
8	2009	150.0	6250	2610	2452
9	2010	400.0	6461	2700	1161
10	2011	580.0	7001	4000	4278
11	2012	580.0	7318	4200	5235
12	2013	610.0	7415	4600	4852
13	2014	800.0	7600	5000	5434
14	2015	10000.0	8100	5400	4329
15	2016	25000.0	8600	5600	3611
16	2017	20000.0	8800	5800	3834
17	2018	83500.0	10100	6000	2721
18	2020	100000.0	12000	7800	1167

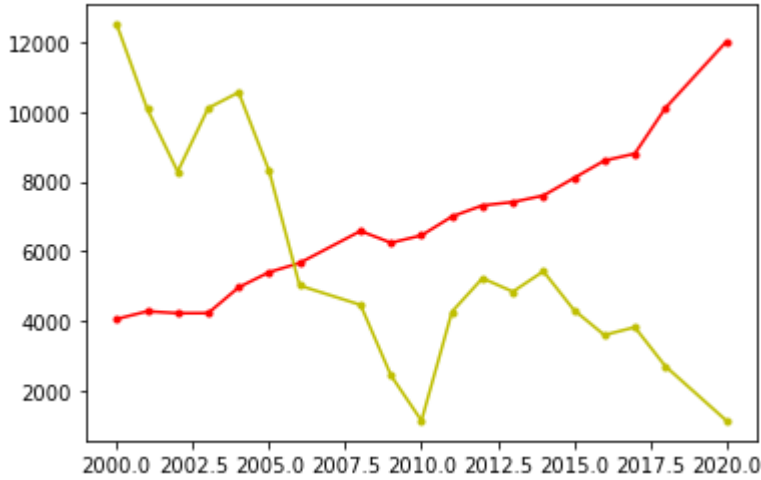
In [3]: `car = pd.read_csv(r'D:\Фотографии\statisti_of_car1.csv')`
`plt.figure(figsize=(8,5))`
`plt.title('The number of cars produced', fontdict={'fontweight':'bold', 'fontsize': 18})`
`plt.plot(car.Year, car.Tesla, 'b.-', label='Tesla')`
`plt.plot(car.Year, car.Ferrari, 'r.-', label = 'Ferrari')`
`plt.plot(car.Year, car['Porsche'],'g.-', label = 'Porsche')`
`plt.plot(car.Year, car.Jaguar,'y.-', label = 'Jaguar')`
`plt.xticks(car.Year[::3].tolist()+[2020])`
`plt.xlabel('Year')`
`plt.ylabel('Statistic of sell cars')`
`plt.legend()`
`plt.savefig('statisti_of_car_figure.png', dpi=300)`
`plt.show()`



In [4]: `car = pd.read_csv(r'D:\Фотографии\statisti_of_car1.csv')`
`plt.plot(car.Year, car.Tesla, 'b.-', label='Tesla')`
`plt.plot(car.Year, car['Porsche'],'g.-', label = 'Porsche')`
`plt.show()`



In [5]: `car = pd.read_csv(r'D:\Фотографии\statisti_of_car1.csv')`
`plt.plot(car.Year, car.Ferrari, 'r.-', label='Tesla')`
`plt.plot(car.Year, car['Jaguar'],'y.-', label = 'Porsche')`
`plt.show()`



In []:

In [4]: `# this part done by Xo'jamurodova Guljahon`

```
import numpy as np
import matplotlib.pyplot as plt
import pandas as pd

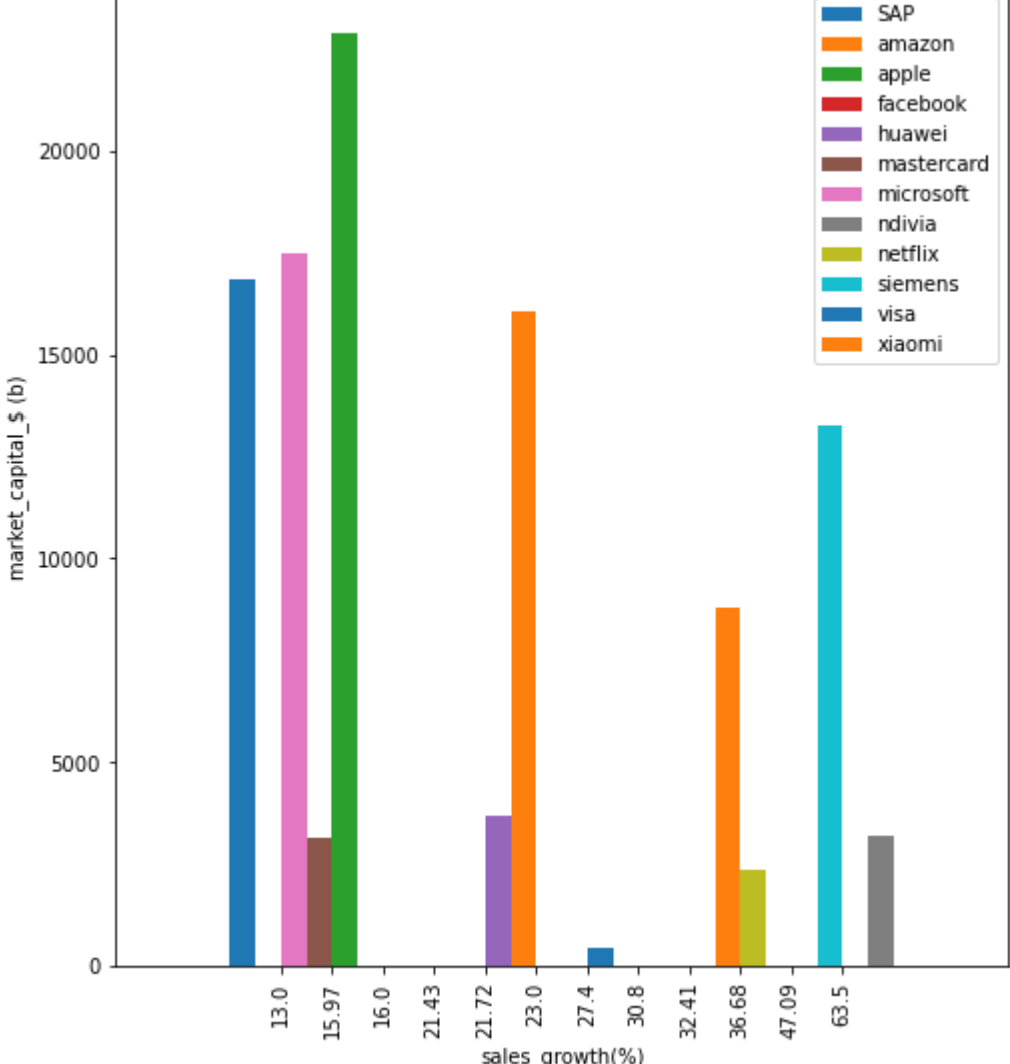
df = pd.read_csv('high-tech companies.csv.')
df.sort_values('market_capital_dollar(b)', ascending=False)
df
```

Out[4]:

	company	sales_growth(%)	market_capital_dollar(b)
0	apple	21.43	22878
1	ndivia	63.50	3216
2	amazon	30.80	16087
3	microsoft	13.00	17495
4	facebook	47.09	7358
5	visa	21.72	426
6	mastercard	15.97	3153
7	netflix	32.41	2352
8	siemens	36.68	13265
9	SAP	16.00	16852
10	huawei	23.00	3698
11	xiaomi	27.40	8793

In [5]: `df.pivot(index='sales_growth%', columns='company', values='market_capital_dollar(b)').plot(kind='bar', figsize=`
`plt.xlabel('sales_growth(%)')`
`plt.ylabel('market_capital_$ (b)')`
`plt.title('high-tech companies growth')`
`plt.legend(loc='upper right')`

`plt.savefig('Mean math performance across USA,China, and Germany.png', dpi= 300)`
`plt.show()`



In [6]: `race = pd.read_csv('race.csv.')`

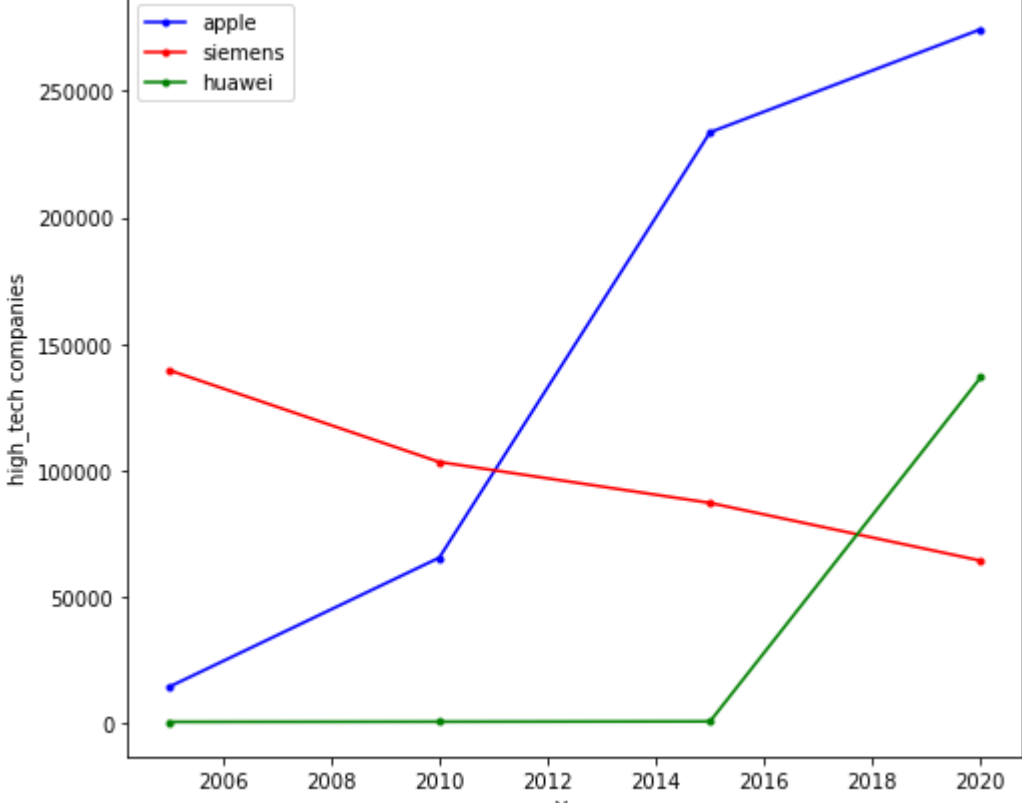
`race['total']= race['apple'] + race['siemens'] + race['huawei']`
`race`

Out[6]:

	year	apple	siemens	huawei	total
0	2005	13931	139573	203	153707
1	2010	65225	103125	276	168626
2	2015	233715	86906	386	321007
3	2020	274515	64030	136717	475262

In [7]: `plt.figure(figsize=(8,7))`
`plt.plot(race.year, race.apple,'b.-')`
`plt.plot(race.year, race.siemens,'r.-')`
`plt.plot(race.year, race.huawei,'g.-')`
`plt.legend(['apple', 'siemens', 'huawei'])`
`plt.xlabel('Year')`
`plt.ylabel('high_tech companies')`

`plt.savefig('high_tech companies.png', dpi= 300)`
`plt.show()`



In [8]: `mobile_cellular_subscriptions = pd.read_csv('mobile-cellular-subscriptions.xls.csv')`
`mobile_cellular_subscriptions`

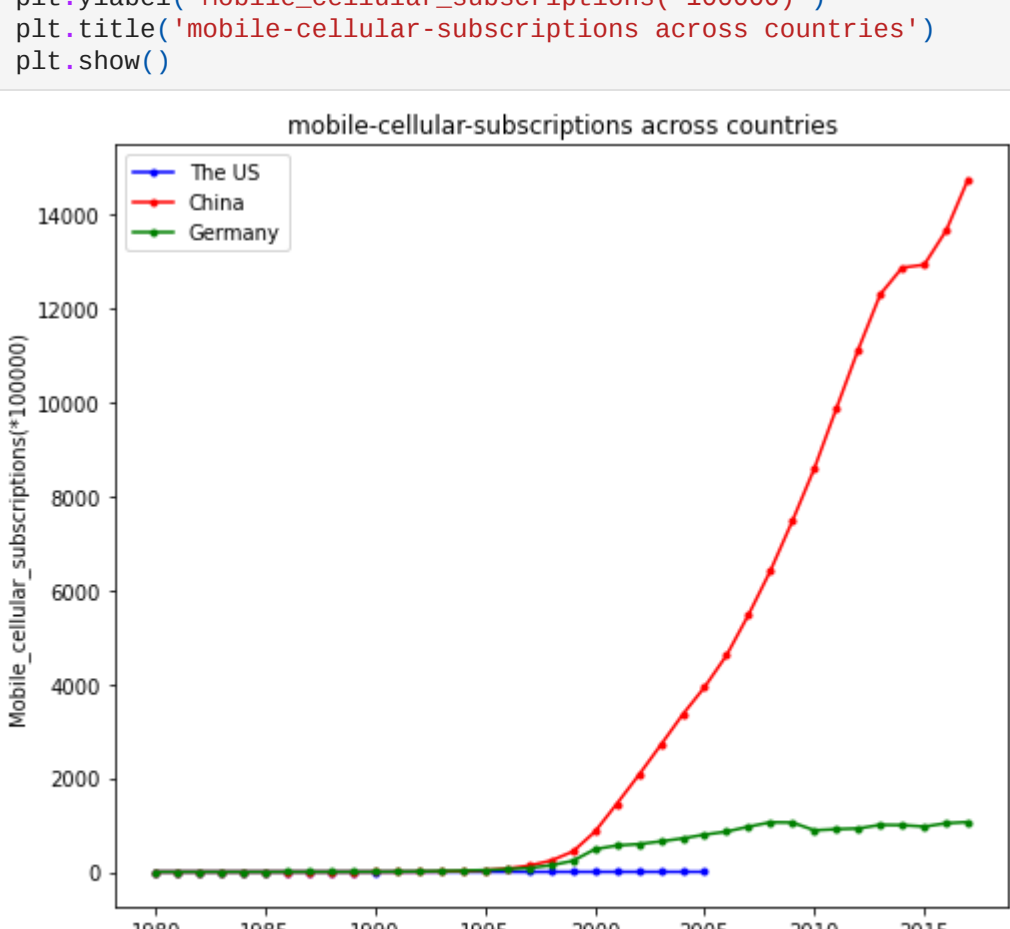
Out[8]:

	Entity	Year	Mobile_cellular_subscriptions
0	China	1980	0
1	China	1981	0
2	China	1982	0
3	China	1983	0
4	China	1984	0
...
126	United States Virgin Islands	2001	41000
127	United States Virgin Islands	2002	45150
128	United States Virgin Islands	2003	49300
129	United States Virgin Islands	2004	64200
130	United States Virgin Islands	2005	80300

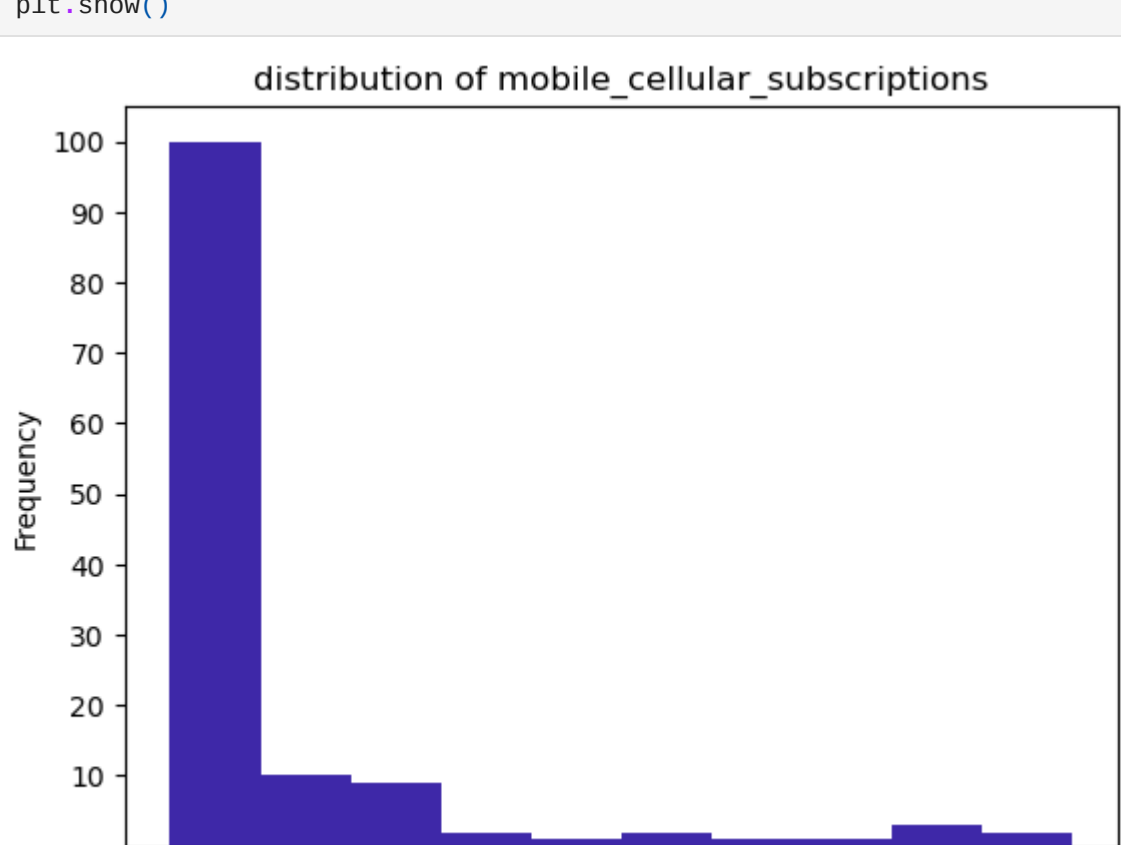
131 rows × 3 columns

In [18]: `US = mobile_cellular_subscriptions[mobile_cellular_subscriptions.Entity == 'United States Virgin Islands']`
`China = mobile_cellular_subscriptions[mobile_cellular_subscriptions.Entity == 'China']`
`Germany = mobile_cellular_subscriptions[mobile_cellular_subscriptions.Entity == 'Germany']`

`plt.figure(figsize=(8,7))`
`plt.plot(US.Year, US.Mobile_cellular_subscriptions/100000,'b.-')`
`plt.plot(China.Year, China.Mobile_cellular_subscriptions/100000,'r.-')`
`plt.plot(Germany.Year, Germany.Mobile_cellular_subscriptions/100000,'g.-')`
`plt.legend(['The US', 'China', 'Germany'])`
`plt.xlabel('Year')`
`plt.ylabel('Mobile_cellular_subscriptions(*100000)')`
`plt.title('mobile-cellular-subscriptions across countries')`
`plt.show()`



In [19]: `plt.style.use('default')`
`plt.hist(mobile_cellular_subscriptions.Mobile_cellular_subscriptions/100000,color='#3e28a8')`
`plt.yticks([10,20,30,40,50,60,70,80,90,100])`
`plt.xlabel('mobile_cellular_subscriptions')`
`plt.ylabel('Frequency')`
`plt.title('distribution of mobile_cellular_subscriptions')`
`plt.savefig('distribution of mobile_cellular_subscriptions.png', dpi= 300)`
`plt.show()`



In []:

Team project GitHub link

[Python-team-project---innovation/complete python project done by team 12.pdf at main · oybek1995-creator/Python-team-project---innovation \(github.com\)](https://github.com/oybek1995-creator/Python-team-project---innovation)