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Preface

What is Python?

Python is an interpreted, object-oriented, high-level programming language with dynamic semantics. Its high-level built in data structures, combined with dynamic typing and dynamic binding, make it very attractive for Rapid Application Development, as well as for use as a scripting or glue language to connect existing components together. Python's simple, easy to learn syntax emphasizes readability and therefore reduces the cost of program maintenance. Python supports modules and packages, which encourages program modularity and code reuse. The Python interpreter and the extensive standard library are available in source or binary form without charge for all major platforms, and can be freely distributed.

Often, programmers fall in love with Python because of the increased productivity it provides. Since there is no compilation step, the edit-test-debug cycle is incredibly fast. Debugging Python programs is easy: a bug or bad input will never cause a segmentation fault. Instead, when the interpreter discovers an error, it raises an exception. When the program doesn't catch the exception, the interpreter prints a stack trace. A source level debugger allows inspection of local and global variables, evaluation of arbitrary expressions, setting breakpoints, stepping through the code a line at a time, and so on. The debugger is written in Python itself, testifying to Python's introspective power. On the other hand, often the quickest way to debug a program is to add a few print statements to the source: the fast edit-test-debug cycle makes this simple approach very effective.

Source: https://www.python.org/doc/essays/blurb/



Top 10 Most Popular Programming Languages

- 1. Python
- 2. JavaScript
- 3. Java
- 4. C#
- 5. C
- 6. C++
- 7. Go
- 8. R
- 9. Swift
- 10. PHP

R Vs Python: What's the Difference?

R and Python are both open-source programming languages with a large community. New libraries or tools are added continuously to their respective catalog. R is mainly used for statistical analysis while Python provides a more general approach to data science.

R and Python are state of the art in terms of programming language oriented towards data science. Learning both of them is, of course, the ideal solution. R and Python requires a time-investment, and such luxury is not available for everyone. Python is a general-purpose language with a readable syntax. R, however, is built by statisticians and encompasses their specific language.



Difference between R and Python

Parameter	R	Python		
Objective	Data analysis and statistics	Deployment and production		
Primary Users	Scholar and R&D	Programmers and developers		
Flexibility	Easy to use available library	Easy to construct new models from scratch. I.e., matrix computation and optimization		
Learning curve	Difficult at the beginning	Linear and smooth		
Popularity of Programming Language. Percentage change	4.23% in 2018	21.69% in 2018		
Average Salary	\$99.000	\$100.000		
Integration	Run locally	Well-integrated with app		
Task	Easy to get primary results	Good to deploy algorithm		
Database size	Handle huge size	Handle huge size		
IDE	Rstudio	Spyder, Ipython Notebook		
Important Packages and library	tidyverse, ggplot2, caret, zoo	pandas, scipy, scikit-learn, TensorFlow, caret		
Disadvantages	Slow High Learning curve Dependencies between library	Not as many libraries as R		
Advantages	 Graphs are made to talk. R makes it beautiful Large catalog for data analysis GitHub interface RMarkdown Shiny 	 Jupyter notebook: Notebooks help to share data with colleagues Mathematical computation Deployment Code Readability Speed Function in Python 		



my data is saved in CSV file.

year	g	b	gg	bb	ggg	bbb	gggg	bbbb
48		10680	1760433	992699	0	0		257649
49	9248	10966	1846935	1069331	0	0	623483	297295
50	8706	10602	1874746	1128112	0	0	702481	341018
51	9509	11728	1734310	1069331	161754	97464	797565	390881
52	10025	11748	1825407	1137189	361084	210381	668600	330668
53	18587	22400	1957533	1180938	576045	330293	547629	269295
54	40574	42280	2173074	1348350	739710	411746	524235	273383
56	79978	95446	2321089	1491431	824242	459419	558411	297107
57	97434	114435	2459239	1619235	875516	493394	608754	332835
58	108298	126838	2581835	1732580	917571	528769	634712	354566
59	119662	139835	2625342	1769630	972137	563126	693809	393750
60	109123	127363	3067778	1941547	1063448	626443	674398	438830
61	81163	90840	2899086	1899506	975359	599705	626861	416536
62	92554	103435	3170008	2113369	1078543	671334	572081	414349
63	85329	92196	3308208	2284600	1055466	661631	576781	399662
64	100960	101939	3468923	2525480	1114107	703546	591049	424139
65	36135	41639	3615988	2720028	1236523	782213	628377	450354
66	50376	56610	3828105	2960218	1355589	854653	700784	488771
67	58451	64986	4058853	3173967	1406118	893392	768422	509599
68	69064	77345	4307333	3450374	1498834	968082	825448	550813
69	85223	92756	4537629	3724812	1638550	1086056	944675	628342
70	103345	114151	4790250	4026895	1782133	1213665	1000835	671812
71	108151	119341	5041319	4328327	1892510	1340172	1066653	752813
72	119959	132554	5224343	4563250	2050707	1490871	1192370	838616
73	81307	87557	5270533	4666836	2274665	1724847	1385197	1003113
74	63534	69119	5210412	4652405	2496712	1943259	1476237	1108108
75	97782	43946	5151547	4594053	2621635	2090393	1615458	1304133
76	71299	75765	4996934	4448983	2729990	2225215	1701525	1477244
77	95339	99842	4885675	4352718	2845092	2343720	1816811	1662824
78	100308	104014	4720322	4217915	2880892	2401989	1903939	1800716
79	109723	110711	4560725	4106422	2889797	2404875	2022833	1896793
80		125700	4348771	3938766	2828770	2343746	2050709	1957867
81	144365	142538	4175849	3792588	2748874	2278350	2065265	1998449
82		160822	3924999	3588016	2698328	2255566		
83		195487	3662507	3366417	2636728	2228875	1940924	1887600
84		212276	3457559	3190358	2514943	2163621	1944256	1873515
85		238387	3215507	2991211	2345168	2025926	1917888	1854697
86		262333	3108860		2214730	1931383	1917183	1845854
87		268349	3009561	2818290	2100894	1837767	1890567	1805380
88		247983	2952075	2773554	1967906			1731272
89		219653	2912264		1833240			1677775
90		221844	2878290	2713462	1733942	1558304		1546074
91	239148	225305	2898799	2734026	1705332	1539569		1437252
92		232630	2932282	2769239	1693270			1389637
93		232028	3522913	3328079	1110458	1002417	1683347	1594689
94		271658	3610900	3407385	1085222	985951	1696561	1591595
95		337615	3701398	3492795	1068474	979158		1555271
96	338704	349748	3819488	3603017	1591299	1463060	1130697	1099526



I will explain about data;

Number of level of education of individuals from 1348 to 1396.

- g: is number of preschool girls
- b:: is number of preschool boys
- gg: is number of primary school girls
- bb: is number of primary school boys
- ggg: is number of middle school girls
- bbb : is number of middle school boys
- gggg:: is number of high school girls
- bbbb : is number of high school boys

I got the data from the website of the Statistics Center of Iran



Top some Python Libraries for Data Science and statistics

1. Pandas

Pandas is an open-source Python package that provides high-performance, easy-to-use data structures and data analysis tools for the labeled data in Python programming language. Pandas stand for *Python Data Analysis Library*. Who ever knew that?

When to use? Pandas is a perfect tool for data wrangling or munging. It is designed for quick and easy data manipulation, reading, aggregation, and visualization.

Pandas take data in a CSV or TSV file or a SQL database and create a Python object with rows and columns called a data frame. The data frame is very similar to a table in statistical software, say Excel or SPSS.

2. NumPy

One of the most fundamental packages in Python, *NumPy* is a general-purpose array-processing package. It provides high-performance multidimensional array objects and tools to work with the arrays. NumPy is an efficient container of generic multidimensional data.



NumPy's main object is the homogeneous multidimensional array. It is a table of elements or numbers of the same datatype, indexed by a tuple of positive integers. In NumPy, dimensions are called *axes* and the number of axes is called *rank*. NumPy's array class is called *ndarray* aka *array*.

3. SciPy

The SciPy library is one of the core packages that make up the SciPy stack. Now, there is a difference between SciPy Stack and SciPy, the library. *SciPy* builds on the NumPy array object and is part of the stack which includes tools like Matplotlib, Pandas, and SymPy with additional tools,

SciPy library contains modules for efficient mathematical routines as linear algebra, interpolation, optimization, integration, and statistics. The main functionality of the SciPy library is built upon NumPy and its arrays. SciPy makes significant use of NumPy.

4. Matplotlib

This is undoubtedly my favorite and a quintessential Python library. You can create stories with the data visualized with Matplotlib. Another library from the SciPy Stack, Matplotlib plots 2D figures.

When to use? Matplotlib is the plotting library for Python that provides an object-oriented API for embedding plots into applications. It is a close resemblance to MATLAB embedded in Python programming language.



5. Seaborn

So when you read the official documentation on Seaborn, it is defined as the data visualization library based on Matplotlib that provides a high-level interface for drawing attractive and informative statistical graphics. Putting it simply, seaborn is an extension of Matplotlib with advanced features.

So, what is the difference between Matplotlib and Seaborn? Matplotlib is used for basic plotting; bars, pies, lines, scatter plots and stuff whereas, seaborn provides a variety of visualization patterns with less complex and fewer syntax.

6. Scikit Learn

Introduced to the world as a Google Summer of Code project, Scikit Learn is a robust machine learning library for Python. It features ML algorithms like SVMs, random forests, k-means clustering, spectral clustering, mean shift, cross-validation and more... Even NumPy, SciPy and related scientific operations are supported by Scikit Learn with Scikit Learn being a part of the SciPy Stack.



Start with python

At the first we should import libraries in python IDLE.

```
mabahes.py - E:\disk D\99\mabahes\end\mabahes.py (3.9.0)
File Edit Format Run Options Window Help
  1 import pandas as pd
  2 import math
  3 import statistics
  4 import numpy as np
  5 import scipy.stats
  6 from scipy import stats
  7 import csv
  8 from plotnine import ggplot,facet_grid,aes, labs, geom_point
  9 import matplotlib.pyplot as plt
10 from sklearn.linear_model import LinearRegression
11 import statsmodels.api as sm
12 import seaborn as sns
13 from statsmodels.api import OLS
14 from matplotlib import pyplot
15 from scipy.optimize import curve fit
16 from numpy import arange
17
18
```

In general, We should not use all libraries.



Next step: we must import CSV file. 19 ## import data 20 df = pd.read csv('D:/99/mabahes/end/dd.csv') 21 print(np.array(df)) Out put: *Python 3.9.0 Shell* × File Edit Shell Debug Options Window Help Python 3.9.0 (tags/v3.9.0:9cf6752, Oct 5 2020, 15:34:40) [MSC v.1927 64 bit (AMD64)] on win32 Type "help", "copyright", "credits" or "license()" for more information. ======== RESTART: E:\disk D\99\mabahes\end\mabahes.py ========== 8782 10680 1760433 992699 0 0 542917 257649] 49 9248 10966 1846935 1069331 0 0 623483 297295] 0 702481 341018] 50 8706 10602 1874746 1128112 0 9509 11728 1734310 1069331 161754 97464 797565 390881] 10025 11748 1825407 1137189 361084 210381 668600 330668] 18587 22400 1957533 1180938 576045 330293 547629 269295] 54 40574 42280 2173074 1348350 739710 411746 524235 273383 79978 95446 2321089 1491431 824242 459419 558411 297107] 57 97434 114435 2459239 1619235 875516 493394 608754 332835] 58 108298 126838 2581835 1732580 917571 528769 634712 354566] 59 119662 139835 2625342 1769630 972137 563126 693809 393750] 60 109123 127363 3067778 1941547 1063448 626443 674398 438830] 61 81163 90840 2899086 1899506 975359 599705 626861 416536] 62 92554 103435 3170008 2113369 1078543 671334 572081 414349] 63 85329 92196 3308208 2284600 1055466 661631 576781 399662] 64 100960 101939 3468923 2525480 1114107 703546 591049 424139] 65 36135 41639 3615988 2720028 1236523 782213 628377 450354] 66 50376 56610 3828105 2960218 1355589 854653 700784 488771] 67 58451 64986 4058853 3173967 1406118 893392 768422 509599] 68 69064 77345 4307333 3450374 1498834 968082 825448 550813] 69 85223 92756 4537629 3724812 1638550 1086056 944675 628342] 70 103345 114151 4790250 4026895 1782133 1213665 1000835 671812] 71 108151 119341 5041319 4328327 1892510 1340172 1066653 752813] 110000 120004 0004242 4002000 0000707 1400071 Showed as a matrix because we used numpy.array. And then 24 ## show as a dataframe



26 print(pddf)

25 pddf = pd.DataFrame(df)

a = np.array(df)

Output:

```
gggg
    8782 10680 1760433 992699
        10966 1846935 1069331
         10602 1874746 1128112
                                 0
   9509 11728 1734310 1069331 161754 97464 797565 390881
         11748 1825407 1137189 361084
                                      210381 668600 330668
         22400 1957533 1180938 576045
   40574 42280 2173074 1348350 739710 411746 524235
   79978 95446 2321089 1491431 824242 459419 558411
   97434 114435 2459239 1619235 875516 493394 608754 332835
58 108298 126838 2581835 1732580 917571 528769 634712 354566
59 119662 139835 2625342 1769630 972137 563126 693809 393750
60 109123 127363 3067778 1941547 1063448 626443 674398 438830
   81163 90840 2899086 1899506 975359 599705 626861 416536
   92554 103435 3170008 2113369 1078543 671334 572081 414349
63 85329 92196 3308208 2284600 1055466 661631 576781 399662
64 100960 101939 3468923 2525480 1114107 703546 591049 424139
   36135 41639 3615988 2720028 1236523 782213 628377 450354
         56610 3828105 2960218 1355589
```

Note: we know use Data Frame so we use pandas library for data to change data frame and call that with pddf name.

Next step:

For example, we get some operations such as R.

```
## mean
d = np.mean(df)
print(d)
## mean
print("var")
a2 = df.var(ddof=1)
print(a2)
## ks test
print('ks test')
print(stats.kstest(df['b'],df['g']))
```



Output: Mean: year 7.235417e+01 1.308601e+05 1.322380e+05 gg 3.487624e+06 2.904961e+06 bb 1.587340e+06 ggg bbb 1.242380e+06 gggg 1.238511e+06 bbbb 1.033069e+06 dtype: float64 Variance: var year 2.022336e+02 8.405851e+09 7.827607e+09 1.100020e+12 gg 1.261263e+12 bb 7.165491e+11 ggg 5.722421e+11 bbb gggg 3.251030e+11 bbbb 4.132904e+11 dtype: float64 Ks test: ks test KstestResult(statistic=0.125, pvalue=0.85283384171513) 41 ## summary 42 print(df.describe()) 43 result = scipy.stats.describe(a, ddof=1, bias=False)



44 print(result)

Output:

Line 42 we used describe code for obtain summary:

```
bbbb
      year
                g ...
                        gggg
count 48.000000 48.000000 ... 4.800000e+01 4.800000e+01
mean 72.354167 130860.083333 ... 1.238511e+06 1.033069e+06
std 14.220886 91683.428190 ... 5.701781e+05 6.428766e+05
min 48.000000 8706.000000 ... 5.242350e+05 2.576490e+05
25% 60.750000 70740.250000 ... 6.601280e+05 4.106772e+05
50% 72.500000 102152.500000 ... 1.161534e+06 9.208645e+05
75% 84.250000 226624.500000 ... 1.820710e+06 1.666562e+06
max 96.000000 338704.000000 ... 2.065265e+06 1.998449e+06
```

We can see mean standard deviation and etc.

In the next print:

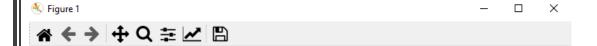
We used scipy library for obtain summary.

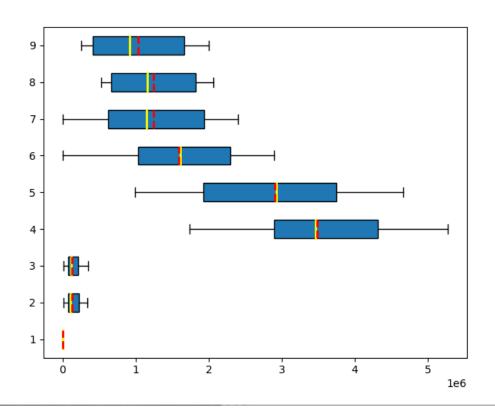
```
Output:
[8 rows x 9 columns]
DescribeResult(nobs=48, minmax=(array([ 48, 8706, 10602, 1734310,
992699.
           0.
    524235, 257649], dtype=int64), array([ 96, 338704, 349748, 52705
33, 4666836, 2889797, 2404875,
   2065265, 1998449], dtype=int64)), mean=array([7.23541667e+01, 1.308
60083e+05, 1.32237979e+05, 3.48762367e+06,
   2.90496108e+06, 1.58733973e+06, 1.24238019e+06, 1.23851138e+0
6,
   1.03306919e+06]), variance=array([2.02233599e+02, 8.40585100e+09,
7.82760667e+09, 1.10002010e+12,
    1.26126255e+12, 7.16549095e+11, 5.72242066e+11, 3.25103019e+1
1,
   4.13290364e+11]), skewness=array([-0.04005161, 0.63203533, 0.639]
80655, 0.07310676, -0.1759978,
   -0.11150676, 0.02351265, 0.08204523, 0.17942308]), kurtosis=array([
-1.16645346, -0.67844912, -0.31807235, -0.94211753, -1.04715452,
   -0.90356133, -1.27509001, -1.74331754, -1.71329171]))
```

We can see the differences.



Boxplot:





For each column we drew a boxplot

- The mean is the red dashed line.
- The median is the yellow dashed line.



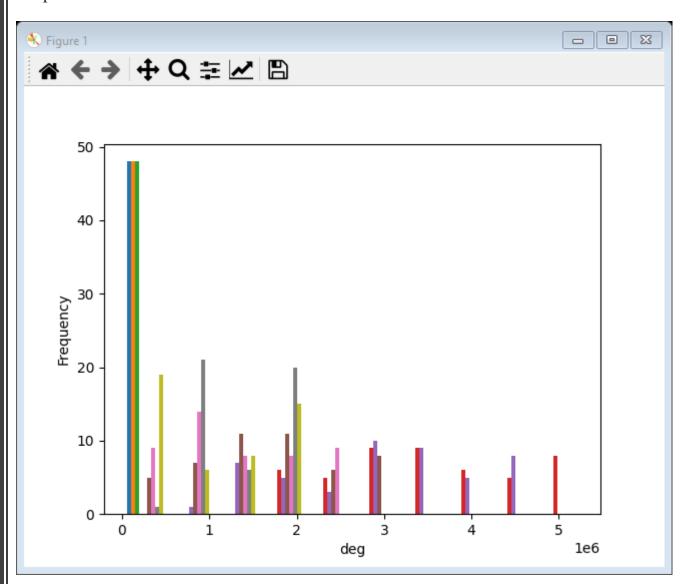
Histogram:

```
#histogram
```

```
fig, ax = plt.subplots()
ax.hist(df, cumulative=False)
ax.set_xlabel('deg')
ax.set_ylabel('Frequency')
plt.show()

ax.set_ylabel('Frequency')
plt.show()
```

Output:





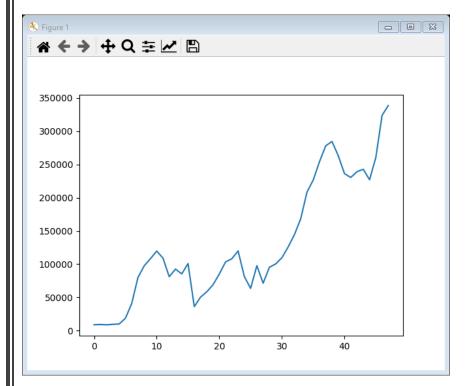
Draw plot for all columns:

```
#####
           plot
plt.plot(df['g'])
plt.show()
plt.plot(df['b'])
plt.show()
plt.plot(df['gg'])
plt.show()
plt.plot(df['bb'])
plt.show()
plt.plot(df['ggg'])
plt.show()
plt.plot(df['bbb'])
plt.show()
plt.plot(df['gggg'])
plt.show()
plt.plot(df['bbbb'])
plt.show()
```

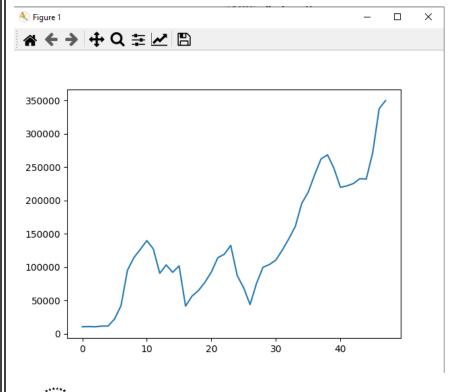


Respectively:

G :



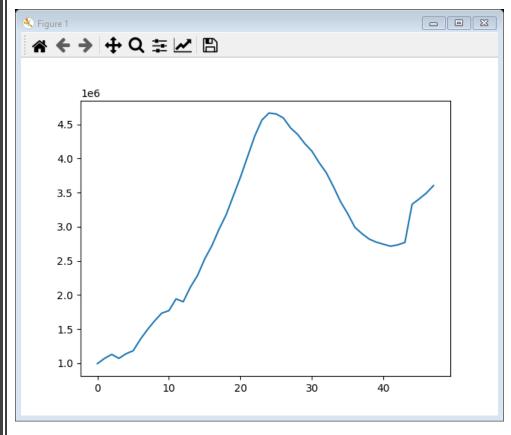
b:





bb:

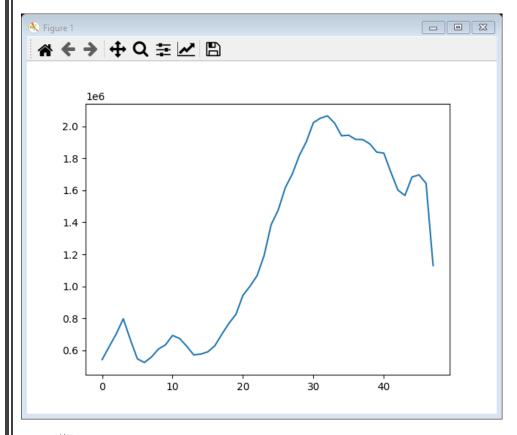
2.0





ggg: 🕙 Figure 1 \square \times **☆←→ +** Q ≠ **∠** 🖺 1e6 3.0 -2.5 2.0 1.5 1.0 0.5 0.0 10 20 30 40

bbbb:



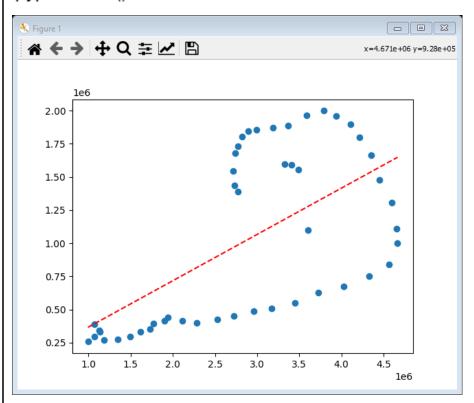


Although you see those charts, In 1357, due to the war, the level of education decreased.

Of course, in some years the level of education is zero, which is not registered.



Creat curve and draw scatter plot: ### Curve def objective(x, a, b): return a * x + b # choose the input and output variables data = df.values # choose the input and output variables x, y = data[:, 4], data[:, -1] # curve fit popt, _ = curve_fit(objective, x, y) # summarize the parameter values a, b = popt print('y = %.5f * x + %.5f' % (a, b)) # plot input vs output pyplot.scatter(x, y) # define a sequence of inputs between the smallest an x line = arange(min(x), max(x), 1) # calculate the output for the range y_line = objective(x_line, a, b) # create a line plot for the mapping function pyplot.plot(x_line, y_line, '--', color='red') pyplot.show()



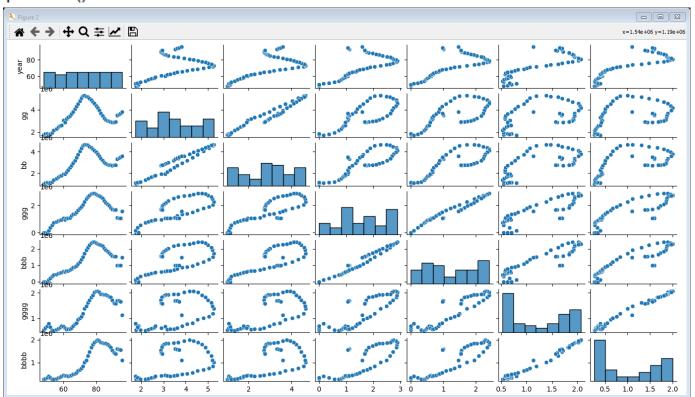


Use simple.eda such as R *

corr() is used to find the pairwise correlation of all columns in the dataframe. Any na values are automatically excluded

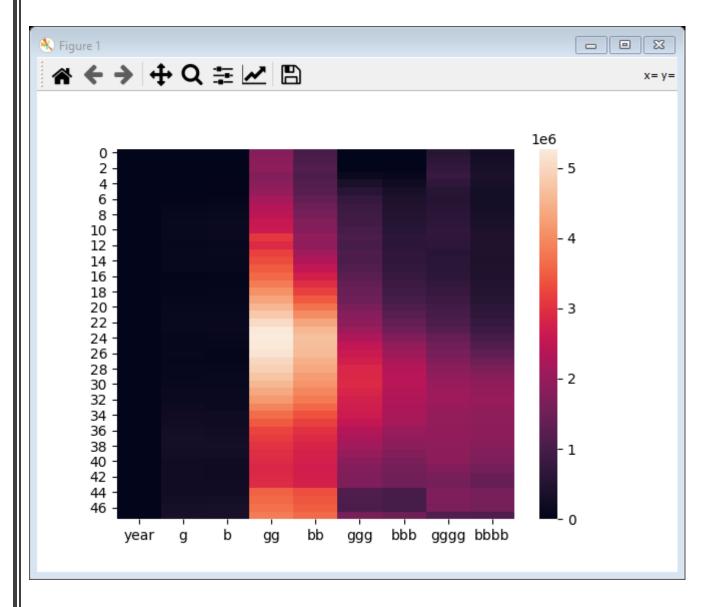
In this part, we had to use (import seaborn as sns import matplotlib.pyplot as plt) library for run code.

plt.show()





Heat map: import csv import pandas as pd import numpy as np import seaborn as sns import matplotlib.pyplot as plt ### import data df = pd.read_csv('E:/disk D/99/mabahes/end/dd.csv') uniform_data = df ax = sns.heatmap(uniform_data) plt.show()





Conclusion:

We can reach these results:

- Python is fast, flexible and accurate.
- Ability to use multiple libraries simultaneously.
- Use a variety of editors for example : anaconda , pycharm or VScode and etc .

(but in this project, I use IDLE)

• Use data with NAN quantities with specific codes.

