import numpy as np import pandas as pd import seaborn as sns import matplotlib.pyplot as plt from scipy.stats import pearsonr, spearmanr from scipy.stats import poisson, binom  Correlation
<pre>In [10]: # Ravi&gt; heights = np.array([68, 62, 64, 61, 70, 66, 61, 65, 71, 72]) # inches weights = np.array([72, 58, 67, 72, 79, 61, 68, 64, 80, 79]) # kgs  sns.scatterplot(x=heights, y=weights) plt.axvline(heights.mean(),color="green",linestyle="") plt.axhline(weights.mean(),color="green",linestyle="")  plt.xlabel("Heights (inches)") plt.ylabel("Weights (Kgs)") plt.show()</pre>
80 - 75 - 80 - 70 - 86 -
In [11]: covariance=(((heights-heights.mean())*(weights-weights.mean())).sum())/len(heights) print("Covariance : ", covariance)  Covariance : 20.4  In [12]: # Rishav> USA
heights1 = np.array([68, 62, 64, 61, 70, 66, 61, 65, 71, 72])*2.54 # cms weights1 = np.array([72, 58, 67, 72, 79, 61, 68, 64, 80, 79])*1000 # gms  sns.scatterplot(x=heights1, y=weights1) plt.axvline(heights1.mean(),color="green",linestyle="") plt.axhline(weights1.mean(),color="green",linestyle="") plt.xlabel("Heights1 (cms)") plt.ylabel("Weights1 (gms)") plt.show()
80000 - 75000 - (SE) 70000
In [13]: covariance1=(((heights1-heights1.mean())*(weights1-weights1.mean())).sum())/len(heights1) print("Covariance: ",covariance1)  Covariance: 51816.00000000001  In [14]: covariance/(heights.std()*weights.std())
Out[14]: 0.7094289771951879  In [15]: covariance1/(heights1.std()*weights1.std())  Out[15]: 0.7094289771951878  In [18]: np.corrcoef(weights, heights)  Out[18]: array([[1. , 0.70942898],
[0.70942898, 1. ]]) In [19]: np.corrcoef(weights, heights)[0,1] Out[19]: 0.7094289771951878  In [20]: pearsonr(weights, heights) Out[20]: (0.7094289771951878, 0.021575164034828447)
<pre>spearmanr(weights, heights) Out[38]: SpearmanrResult(correlation=0.6024493002399199, pvalue=0.06529313205986141) In [23]:</pre>
1 62 58 2 64 67 3 61 72 4 70 79 5 66 61 6 61 68
7 65 64 8 71 80 9 72 79  In [30]: heights_ranks=df[0].rank() weights_ranks=df[1].rank()  In [31]: heights_ranks
Out[31]: 0 7.0  1 3.0  2 4.0  3 1.5  4 8.0  5 6.0  6 1.5  7 5.0  8 9.0  9 10.0
Name: 0, dtype: float64  In [32]: weights_ranks  Out[32]: 0 6.5  1 1.0  2 4.0  3 6.5  4 8.5  5 2.0  6 5.0
7 3.0 8 10.0 9 8.5 Name: 1, dtype: float64  In [35]: sns.scatterplot(x=heights_ranks, y=weights_ranks) plt.axvline(weights_ranks.mean(),color="green",linestyle="") plt.axhline(weights_ranks.mean(),color="green",linestyle="") plt.xlabel("heights_ranks ") plt.ylabel("weights_ranks ")
plt.show()
In [37]: pearsonr(heights_ranks, weights_ranks) Out[37]: (0.6024493002399198, 0.06529313205986152)
Dut[39]:
3 48 0 37 30 9 1 160.00 2 Bangladesh Sharjah 1995-04-05 India True False 4 4 0 13 9 1 0 44.44 2 Pakistan Sharjah 1995-04-07 Pakistan False False
358 114 0 205 147 12 1 77.55 1 Bangladesh Dhaka 2012-03-16 Bangladesh False True 359 52 0 93 48 5 1 108.33 2 Pakistan Dhaka 2012-03-18 India True False  360 rows × 14 columns  In [40]: df.info() <class 'pandas.core.frame.dataframe'=""> RangeIndex: 360 entries, 0 to 359</class>
Data columns (total 14 columns):  # Column Non-Null Count Dtype
8
Notic sixes
Inns -0.076998 0.104015 -0.114803 -0.014255 -0.045657 0.058539 1.00000 0.043287 -0.137966  Won
runs - 1
Won - 019 021 015 023 0064 017 0043 1 011
Truns - 1
Won - 0.19 0.21 0.16 0.21 0.078 0.18 0.043 1 0.11
Out [45]:         mpg         cylinders         displacement         horsepower         weight         acceleration         model_year         origin         name           0         18.0         8         307.0         130.0         350.4         12.0         70         usa         buick skylark 320           2         18.0         8         318.0         150.0         3436         11.0         70         usa         plymouth satellite           3         16.0         8         304.0         150.0         3433         12.0         70         usa         amc rebel sst           4         17.0         8         302.0         140.0         3449         10.5         70         usa         ford torino
393 27.0 4 140.0 86.0 2790 15.6 82 usa ford mustang gl 394 44.0 4 97.0 52.0 2130 24.6 82 europe ww pickup 395 32.0 4 135.0 84.0 2295 11.6 82 usa dodge rampage 396 28.0 4 120.0 79.0 2625 18.6 82 usa ford ranger 397 31.0 4 119.0 82.0 2720 19.4 82 usa chevy s-10 398 rows × 9 columns
In [47]: sns.heatmap(df_mpg.corr(), annot=True, cmap="coolwarm") Out[47]: <axessubplot:>  mpg - 1</axessubplot:>
2000
In [49]: df_mpg.isnull()
dtype: int64  In [50]: pearsonr(df_mpg["displacement"], df_mpg["mpg"])  Out[50]: (-0.8042028248058989, 1.655888910912744e-91)  In [51]: spearmanr(df_mpg["displacement"], df_mpg["mpg"])  Out[51]: SpearmanrResult(correlation=-0.8556920118178749, pvalue=2.3646353779685863e-115)  In []:
In []:  In []:  In []:  In []:
Poisson Distribution  In [60]: import math def calc_poisson(k,mu): return ((mu**k)*((math.exp(-1*mu))))/math.factorial(k)  In [61]: calc_poisson(5,3)
Out[61]:       0.10081881344492448         In [67]:       calc_poisson(2,3)         Out[67]:       0.22404180765538775         In [68]:       calc_poisson(3,3)         Out[68]:       0.22404180765538775
In [80]: calc_poisson(1,3)+calc_poisson(0,3) Out[80]: 0.19914827347145578  In [81]: poisson.pmf(k=1,mu=3)+poisson.pmf(k=0,mu=3) Out[81]: 0.1991482734714558  In [82]: poisson.cdf(k=1,mu=3) Out[82]: 0.100140373474559
Out[82]:       0.1991482734714558         In []:       poisson.pmf(k=5, mu=3)         Out[62]:       0.10081881344492458         In [63]:       poisson.pmf(k=3, mu=3)
Out[63]: 0.22404180765538775  In [65]: poisson.pmf(k=2,mu=3)  Out[65]: 0.22404180765538775  In [64]: mu = 3  x_values = np.arange(0, 10) y_values = poisson.pmf(x_values, mu=mu)
y_values = poisson.pmf(x_values, mu=mu) sns.barplot(x=x_values, y=y_values)  Out[64]: <axessubplot:>  020 015</axessubplot:>
In [69]: mu = 1
<pre>x_values = np.arange(0, 10) y_values = poisson.pmf(x_values, mu=mu) sns.barplot(x=x_values, y=y_values)  Out[69]: <axessubplot:></axessubplot:></pre>
0.25 - 0.20 - 0.15 - 0.10 - 0.05 - 0.00 - 0 1 2 3 4 5 6 7 8 9
<pre>In [70]: mu = 2     x_values = np.arange(0, 10)     y_values = poisson.pmf(x_values, mu=mu)     sns.barplot(x=x_values, y=y_values)  Out[70]: <axessubplot:></axessubplot:></pre>
0.10 - 0.05 - 0.00
In [71]: mu = 3  x_values = np.arange(0, 10) y_values = poisson.pmf(x_values, mu=mu) sns.barplot(x=x_values, y=y_values)  Dut[71]: <axessubplot:></axessubplot:>
0.15 - 0.10 - 0.05 -
In [72]: mu = 4  x_values = np.arange(0, 10) y_values = poisson.pmf(x_values, mu=mu) sns.barplot(x=x_values, y=y_values)
Sils.barplot(x=x_values, y=y_values)  Out[72]: <axessubplot:>  0.200 0.175 0.150 0.125 0.100 0.075</axessubplot:>
In [73]: mu =5 x_values = np.arange(0, 10)
y_values = poisson.pmf(x_values, mu=mu) sns.barplot(x=x_values, y=y_values)  Out[73]: <axessubplot:>  0.175 - 0.150 - 0.125 -</axessubplot:>
0.100 - 0.075 - 0.050 - 0.025 - 0.000 - 0 1 2 3 4 5 6 7 8 9
In [74]: mu = 6  x_values = np.arange(0, 10) y_values = poisson.pmf(x_values, mu=mu) sns.barplot(x=x_values, y=y_values)  Out[74]: <axessubplot:>  0.16 0.14</axessubplot:>
0.14 - 0.12 - 0.10 - 0.08 - 0.06 - 0.04 - 0.02 -
In [75]: mu = 7  x_values = np.arange(0, 10) y_values = poisson.pmf(x_values, mu=mu) sns.barplot(x=x_values, y=y_values)  Out[75]: <axessubplot:></axessubplot:>
Out[75]: <axessubplot:>  014 - 012 - 010 - 008 - 006 - 004 -</axessubplot:>
In [76]: mu = 8  x_values = np.arange(0, 10)
<pre>y_values = np.arange(0, 10) y_values = poisson.pmf(x_values, mu=mu) sns.barplot(x=x_values, y=y_values)</pre>
y_values = poisson.pmf(x_values, mu=mu) sns.barplot(x=x_values, y=y_values)  Out[76]: <axessubplot:>  0.14 0.12 0.10 0.08</axessubplot:>
<pre>y_values = poisson.pmf(x_values, mu=mu) sns.barplot(x=x_values, y=y_values)  Dut[76]: <axessubplot:>  014 012 010 010 010 010 010 010 010 010 010</axessubplot:></pre>

Importing Libraries