<pre>In [29]: # Lalit 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)")</pre>
plt.ylabel("Weights (Kgs)") plt.show()  80  75
60 - 62 64 66 68 70 72 Heights (inches)
<pre>In [30]: covariance_lalit=((heights-heights.mean())*(weights-weights.mean())).sum()/len(heights) covariance_lalit  Out[30]: 20.4 In [38]: correlation_lalit= covariance_lalit/(heights.std()*weights.std()) correlation_lalit</pre>
Out[38]: 0.7094289771951879  In [31]: # Nakul 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])*2.2 # lbs sns.scatterplot(x=heights1, y=weights1)
<pre>plt.axvline(heights1.mean(), color="green", linestyle="") plt.axhline(weights1.mean(), color="green", linestyle="")  plt.xlabel("Heights (cms)") plt.ylabel("Weights (lbs)") plt.show()</pre>
170 - (g) 160 - (g) 150 -
140
covariance_nakul  Out[32]: 113.9952  In [40]: correlation_nakul= covariance_nakul/(heights1.std()*weights1.std())     correlation_nakul  Out[40]: 0.7094289771951878
<pre>In [36]: # Darsh heights2 = np.array([68, 62, 64, 61, 70, 66, 61, 65, 71, 72])*2.54 # cms weights2 = np.array([72, 58, 67, 72, 79, 61, 68, 64, 80, 79])*1000 # gms  sns.scatterplot(x=heights2, y=weights2) plt.axvline(heights2.mean(),color="green",linestyle="") plt.axhline(weights2.mean(),color="green",linestyle="")</pre>
<pre>plt.xlabel("Heights (cms)") plt.ylabel("Weights (gms)") plt.show()</pre>
75000 - Fig. 70000
In [37]: covariance_darsh=((heights2-heights2-weights2-weights2-weights2-weights2-weights2-weights2) covariance_darsh
Out[37]: 51816.00000000001  In [41]: correlation_darsh= covariance_darsh/(heights2.std()*weights2.std()) correlation_darsh  Out[41]: 0.7094289771951878  In [43]: np.corrcoef(weights, heights)[0,1]
Out[43]:       0.7094289771951878         In [47]:       pearsonr(weights, heights)         Out[47]:       (0.7094289771951878, 0.021575164034828447)
<pre>In [52]: spearmanr(weights, heights) Out[52]: SpearmanrResult(correlation=0.6024493002399199, pvalue=0.06529313205986141) In [50]: df=pd.DataFrame([heights, weights]).transpose()     df.rename(columns={0:"Heights",1:"Weights"},inplace=True)     df</pre>
Out[50]:
<ul> <li>4 70 79</li> <li>5 66 61</li> <li>6 61 68</li> <li>7 65 64</li> <li>8 71 80</li> </ul>
<pre>g 72 79  In [56]: df_rank=df.rank()     df_rank.rename(columns={"Heights":"Heights_rank","Weights_rank"},inplace=True)     df_rank</pre>
Out[56]:
4       8.0       8.5         5       6.0       2.0         6       1.5       5.0         7       5.0       3.0         8       9.0       10.0
9 10.0 8.5  In [57]: pearsonr(df_rank["Heights_rank"], df_rank["Weights_rank"])  Out[57]: (0.6024493002399198, 0.06529313205986152)
tin [58]: df.corr()  Dut[58]: Heights Weights  Heights 1.00000 0.709429  Weights 0.709429 1.000000
In [59]:   df.corr(method="spearman")  Dut[59]:   Heights   Weights    Heights   1.000000   0.602449    Weights   0.602449   1.000000
In [60]: !ls  08_Correlation_Tests_Notebook.ipynb problem_solving.csv Sachin_ODI.csv waiting_time.csv aerofit.csv weight-height.csv iq_two_schools.csv  In [62]: df=pd_read_csv("Sachin_ODI.csv")
Out [62]:   runs   NotOut   mins   bf   fours   sixes   sr   lnns   Opp   Ground   Date   Winner   Won   century    1
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
357 6 0 25 19 1 0 31.57 1 Sri Lanka Dhaka 2012-03-13 India True 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 [63]:         df.corr()           Out [63]:         runs         NotOut         bf         fours         sixes         sr         lnns         won         century           runs         1.00000         0.377869         0.95358         0.907861         0.60407         0.216497         0.104015         0.214461         0.336174
bf         0.953558         0.324253         1.00000         0.841240         0.48939         0.419307         -0.114803         0.153881         0.696394           fours         0.907861         0.360667         0.841240         1.000000         0.414766         0.613975         -0.014255         0.225120         0.591924           sixes         0.604078         0.266437         0.487939         0.414766         1.000000         0.415239         -0.045657         0.063837         0.547777           sr         0.576354         0.210590         0.419307         0.613975         0.415239         1.000000         0.058539         0.174703         0.292648           lnns         -0.076998         0.104015         -0.114803         -0.045657         0.058539         1.000000         0.043287         -0.137966
Won 0.188319 0.214461 0.153881 0.225120 0.063837 0.174703 0.043287 1.000000 0.108010  century 0.751176 0.336174 0.696394 0.591924 0.547777 0.292648 -0.137966 0.108010 1.000000  In [64]: sns.heatmap(df.corr(),annot=True,cmap="coolwarm")  Out[64]: <axessubplot:></axessubplot:>
runs - 1
Sr - 0.58 0.21 0.42 0.61 0.42 1 0.059 0.17 0.29  Inns - 0.077 0.1 0.11 0.014 0.046 0.059 1 0.043 0.14  Won - 0.19 0.21 0.15 0.23 0.064 0.17 0.043 1 0.11  century - 0.75 0.34 0.7 0.59 0.55 0.29 0.14 0.11 1  y
In [70]: sns.heatmap(df.corr(method="spearman"), annot=True, cmap="coolwarm")  Out[70]: <axessubplot:>  runs - 1  0.32  0.96  0.93  0.58  0.71  0.052  0.19  0.58</axessubplot:>
bf - 0.96
Won - 0.19 0.21 0.16 0.21 0.078 0.18 0.043 1 0.11 century - 0.58 0.34 0.56 0.5 0.53 0.34 0.14 0.11 1 century - 0.58 0.34 0.56 0.5 0.53 0.34 0.14 0.11 1 century - 0.58 0.34 0.56 0.5 0.53 0.34 0.14 0.11 1 century - 0.58 0.34 0.56 0.5 0.53 0.34 0.14 0.11 1 century - 0.58 0.34 0.56 0.5 0.53 0.34 0.14 0.11 1 century - 0.58 0.34 0.56 0.5 0.53 0.34 0.14 0.11 1 century - 0.58 0.34 0.56 0.5 0.53 0.34 0.14 0.11 1 century - 0.58 0.34 0.56 0.5 0.53 0.34 0.14 0.11 1 century - 0.58 0.34 0.56 0.5 0.53 0.34 0.14 0.11 1 century - 0.58 0.34 0.56 0.5 0.53 0.34 0.14 0.11 1 century - 0.58 0.34 0.56 0.5 0.53 0.34 0.14 0.11 1 century - 0.58 0.34 0.14 0.11 1 century - 0.58 0.34 0.14 0.14 0.14 0.14 0.14 0.14 0.14 0.1
Pout [66]:         runs         NotOut         mins         bf         fours         sixes         sr         Inns         Opp         Ground         Date         Winner         Woo         century           0         13         0         30         15         3         0         86.66         1         New Zealand         New Zealand         False           1         37         0         75         51         3         1         72.54         2         South Africa         Hamilton         1995-02-18         South Africa         False         False           2         47         0         65         40         7         0         117.50         2         Australia         Dunedin         1995-02-22         India         True         False           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
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
<pre>In [68]: df.columns Out[68]: Index(['runs', 'NotOut', 'mins', 'bf', 'fours', 'sixes', 'sr', 'Inns', 'Opp',</pre>
Out[69]: <seaborn.axisgrid.pairgrid 0x7fe3bbdea490="" at=""></seaborn.axisgrid.pairgrid>
0.0
100 d d d d d d d d d d d d d d d d d d
25   15   16   16   16   16   16   16   1
2
by 100
In [71]: spearmanr(df["sr"],df["fours"]) Out[71]: SpearmanrResult(correlation=0.73256962000253, pvalue=9.019200542754248e-62) In [72]: pearsonr(df["sr"],df["fours"])
Out[72]: (0.6139753782596058, 1.127729171277218e-38)  In []: In []:
In []:  In []:  In []:
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import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
import seaborn as sns
from scipy.stats import pearsonr, spearmanr

In [ ]:

In [ ]: