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PART 1

#code cell 1

```
import pandas as pd
brainFile = './sample_data/brainsize.txt'
brainFrame = pd.read_csv(brainFile, delimiter= '\t')
```

#code cell 2
brainFrame.head()




	Gender\tFSIQ\tVIQ\tPIQ\tWeight\tHeight\tMRI_Cour
0	Female\t133\t132\t124\t118\t64.5\t816932
1	Male\t140\t150\t124\tNA\t72.5\t1001121
2	Male\t139\t123\t150\t143\t73.3\t1038437
3	Male\t133\t129\t128\t172\t68.8\t965353
4	Female\t137\t132\t134\t147\t65.0\t951545

Next steps:

[Generate code with brainFrame](#)

 [View recommended plots](#)

Code cell 3
brainFrame.describe()



	Gender\tFSIQ\tVIQ\tPIQ\tWeight\tHeight\tMRI
count	
unique	
top	Female\t133\t132\t124\t118\t64.5\t
freq	

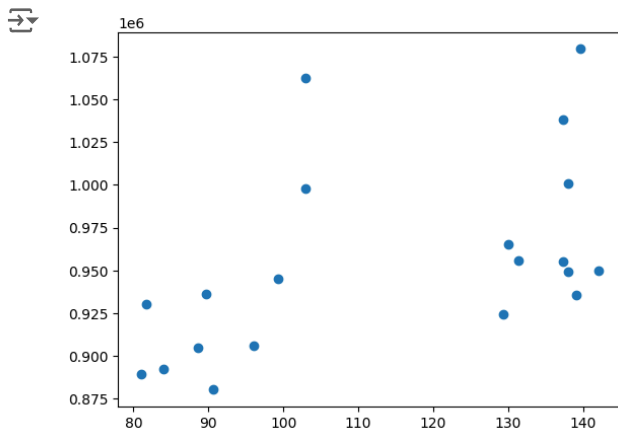
Code cell 4
import numpy as np
import matplotlib.pyplot as plt

Code cell 5
menDf = brainFrame[(brainFrame.Gender == 'Male')]
womenDf = brainFrame[(brainFrame.Gender == 'Female')]

Code cell 6
menMeanSmarts = menDf[["PIQ", "FSIQ", "VIQ"]].mean(axis=1)
plt.scatter(menMeanSmarts, menDf["MRI_Count"])
plt.show()
%matplotlib inline

brainsize.txt X ...

1	Gender	FSIQ	VIQ	PIQ	Weight	Height	MRI_Cou
2	Female	133	132	124	118	64.5	816932
3	Male	140	150	124	NA	72.5	1001121
4	Male	139	123	150	143	73.3	1038437
5	Male	133	129	128	172	68.8	965353
6	Female	137	132	134	147	65.0	951545
7	Female	99	90	110	146	69.0	928799
8	Female	138	136	131	138	64.5	991305
9	Female	92	90	98	175	66.0	854258
10	Male	89	93	84	134	66.3	904858
11	Male	133	114	147	172	68.8	955466
12	Female	132	129	124	118	64.5	833868
13	Male	141	150	128	151	70.0	1079549
14	Male	135	129	124	155	69.0	924059
15	Female	140	120	147	155	70.5	856472
16	Female	96	100	90	146	66.0	878897
17	Female	83	71	96	135	68.0	865363
18	Female	132	132	120	127	68.5	852244
19	Male	100	96	102	178	73.5	945088
20	Female	101	112	84	136	66.3	808020
21	Male	80	77	86	180	70.0	889083
22	Male	83	83	86	NA	NA	892420
23	Male	97	107	84	186	76.5	905940
24	Female	135	129	134	122	62.0	790619
25	Male	139	145	128	132	68.0	955003
26	Female	91	86	102	114	63.0	831772
27	Male	141	145	131	171	72.0	935494
28	Female	85	90	84	140	68.0	798612
29	Male	103	96	110	187	77.0	1062462
30	Female	77	83	72	106	63.0	793549
31	Female	130	126	124	159	66.5	866662
32	Female	133	126	132	127	62.5	857782
33	Male	144	145	137	191	67.0	949589
34	Male	103	96	110	192	75.5	997925
35	Male	90	96	86	181	69.0	879987
36	Female	83	90	81	143	66.5	834344
37	Female	133	129	128	153	66.5	948066
38	Male	140	150	124	144	70.5	949395
39	Female	88	86	94	139	64.5	893983
40	Male	81	90	74	148	74.0	930016
41	Male	89	91	89	179	75.5	935863
42							



```
# Code cell 7
plt.show()
%matplotlib inline
```

```
# Code cell 8
numeric_brainFrame = brainFrame.select_dtypes(include=[np.number])
correlation_matrix = numeric_brainFrame.corr(method='pearson')
print(correlation_matrix)
```

```
FSIQ      FSIQ      VIQ      PIQ      Weight      Height      MRI_Count
FSIQ      1.000000    0.946639    0.934125   -0.051483   -0.086002    0.357641
VIQ      0.946639    1.000000    0.778135   -0.076088   -0.071068    0.337478
PIQ      0.934125    0.778135    1.000000    0.002512    -0.076723    0.386817
Weight   -0.051483   -0.076088    0.002512    1.000000    0.699614    0.513378
Height   -0.086002   -0.071068   -0.076723    0.699614    1.000000    0.601712
MRI_Count 0.357641    0.337478    0.386817    0.513378    0.601712    1.000000
```

```
# Code cell 9
numeric_womenDf = womenDf.select_dtypes(include=[np.number])
correlation_matrix = numeric_womenDf.corr(method='pearson')
print(correlation_matrix)
```

```
FSIQ      FSIQ      VIQ      PIQ      Weight      Height      MRI_Count
FSIQ      1.000000    0.955717    0.939382    0.038192   -0.059011    0.325697
VIQ      0.955717    1.000000    0.802652   -0.021889   -0.146453    0.254933
PIQ      0.939382    0.802652    1.000000    0.113901   -0.001242    0.396157
Weight    0.038192   -0.021889    0.113901    1.000000    0.552357    0.446271
Height   -0.059011   -0.146453   -0.001242    0.552357    1.000000    0.174541
MRI_Count 0.325697    0.254933    0.396157    0.446271    0.174541    1.000000
```

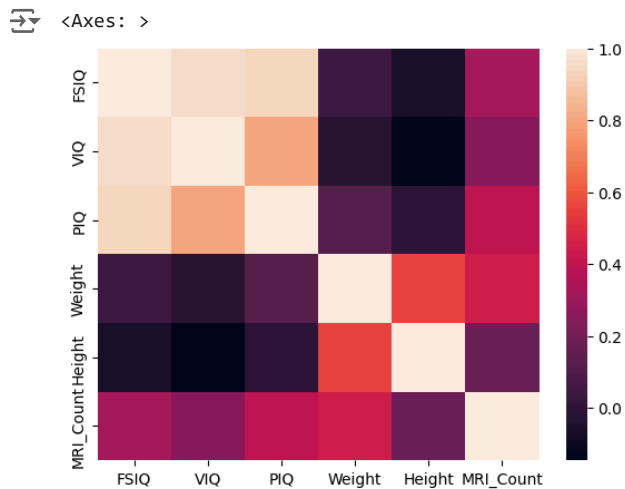
```
# Code cell 10
numeric_df = menDf.select_dtypes(include=[float, int])
correlation_matrix = numeric_df.corr(method='pearson')
print(correlation_matrix)
```

```
FSIQ      FSIQ      VIQ      PIQ      Weight      Height      MRI_Count
FSIQ      1.000000    0.944400    0.930694   -0.278140   -0.356110    0.498369
VIQ      0.944400    1.000000    0.766021   -0.350453   -0.355588    0.413105
PIQ      0.930694    0.766021    1.000000   -0.156863   -0.287676    0.568237
Weight   -0.278140   -0.350453   -0.156863    1.000000    0.406542   -0.076875
Height   -0.356110   -0.355588   -0.287676    0.406542    1.000000    0.301543
MRI_Count 0.498369    0.413105    0.568237   -0.076875    0.301543    1.000000
```

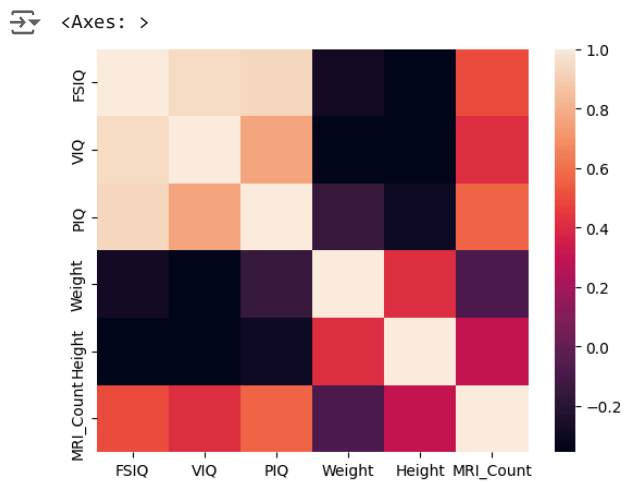
```
# Code cell 11
!pip install seaborn
```

```
Requirement already satisfied: seaborn in /usr/local/lib/python3.10/dist-p
Requirement already satisfied: numpy!=1.24.0,>=1.20 in /usr/local/lib/pyth
Requirement already satisfied: pandas>=1.2 in /usr/local/lib/python3.10/di
Requirement already satisfied: matplotlib!=3.6.1,>=3.4 in /usr/local/lib/p
Requirement already satisfied: contourpy>=1.0.1 in /usr/local/lib/python3.
Requirement already satisfied: cycler>=0.10 in /usr/local/lib/python3.10/d
Requirement already satisfied: fonttools>=4.22.0 in /usr/local/lib/python3
Requirement already satisfied: kiwisolver>=1.0.1 in /usr/local/lib/python3
Requirement already satisfied: packaging>=20.0 in /usr/local/lib/python3.1
Requirement already satisfied: pillow>=6.2.0 in /usr/local/lib/python3.10/
Requirement already satisfied: pyparsing>=2.3.1 in /usr/local/lib/python3.
Requirement already satisfied: python-dateutil>=2.7 in /usr/local/lib/pyth
Requirement already satisfied: pytz>=2020.1 in /usr/local/lib/python3.10/d
Requirement already satisfied: tzdata>=2022.1 in /usr/local/lib/python3.10
Requirement already satisfied: six>=1.5 in /usr/local/lib/python3.10/dist-
```

```
# Code cell 12
numeric_women_df = womenDf.select_dtypes(include=[float, int])
wcorr = numeric_women_df.corr(method='pearson')
sns.heatmap(wcorr)
```



```
# Code cell 14
numeric_men_df = menDf.select_dtypes(include=[float, int])
mcorr = numeric_men_df.corr(method='pearson')
sns.heatmap(mcorr)
```



Many variable pairs present correlation close to zero. What does that mean?

- The reason why there is a lot of the same variable pairs that are presented to correlation that is close to zero because

Why separate the genders?

- The reason why we need to separate the genders is that to be able to find out the difference between those in the graph and they have different results by genders so that we need to separate the genders

What variables have stronger correlation with brain size (MRI_Count)? Is that expected? Explain.

- The Variables that have stronger correlation with brain size is the FSIQ, VIQ, and PIQ. The reason why it has the stronger correlation due it is closer to the value of 1

* SUPPLEMENTARY *

```
import pandas as pd
brainFile = 'nss15.csv'
brainFrame = pd.read_csv(brainFile, delimiter= '\t')
```

```
#code cell 2
brainFrame.head()
```



```
caseNumber, treatmentDate, statWeight, stratum, age, sex, race, diagnosis, body
```

0	1507331
1	15073472
2	15081741
3	15071771
4	15072169

```
# Code cell 3
brainFrame.describe()
```



```
caseNumber, treatmentDate, statWeight, stratum, age, sex, race, diagnosis
```

count	unique	top	freq
			150

```
# Code cell 4
import numpy as np
import matplotlib.pyplot as plt
```

```
print(brainFrame.columns)
```



```
Index(['caseNumber, treatmentDate, statWeight, stratum, age, sex, race, diagnosis
```



```
brainFrame.rename(columns=lambda x: x.strip(), inplace=True)
```

```
# Code cell 5
menDf = brainFrame[(brainFrame.sex == 'Male')]
womenDf = brainFrame[(brainFrame.sex == 'Female')]
```



```
AttributeError                                Traceback (most recent call
last)
```

```
<ipython-input-82-d9e88111f69> in <cell line: 2>()
```

```
1 # Code cell 5
----> 2 menDf = brainFrame[(brainFrame.sex == 'Male')]
      3 womenDf = brainFrame[(brainFrame.sex == 'Female')]
```

```
/usr/local/lib/python3.10/dist-packages/pandas/core/generic.py in
```

```
__getattr__(self, name)
5987     ):
5988         return self[name]
-> 5989     return object.__getattr__(self, name)
5990
5991     def __setattr__(self, name: str, value) -> None:
```

Next steps: [Explain error](#)

```
# Code cell 8
numeric_brainFrame = brainFrame.select_dtypes(include=[np.number])
correlation_matrix = numeric_brainFrame.corr(method='pearson')
print(correlation_matrix)
```