Correlation

- What is correlation?
- Variables within a dataset can be related for lot of reasons.
- Types:
 - Pearson's r (Used for Gaussian distribution)
 - Spearman's rho (Used for non-Gaussian distribution)
 - Kendall's tau (Used for ranking)

For example:

- 1. One variable could cause or depend on the values of another variable.
- 2. One variable could be lightly associated with another variable.
- 3. Two variables could depend on a third known variable.

Possitive Correlation: Both variables change in the same direction (Increase in one results increase in others and vice versa).

Negative Correlation: Variables change in opposite directions (Increase in one results decrease in others and vice versa).

Neutral Correlation: No relationship in the change of the variables (Increase or decrease in one results effectless on other).

Covariance

- Variables can be related by a linear relationship. This is a relationship that is consistently additive across the two data samples.
- This relationship can be summarized between two variables, called the covariance.
- The sign of the covariance can be interpreted as whether the two variables change in the same direction (positive) or in different(negative).
- The magnitude of the covariance is not easily interpreted. A covariance value of zero indicates that both variables are completly independent.

```
In [ ]:
          # import Libraries
          import numpy as np
          import pandas as pd
          import seaborn as sns
          import matplotlib.pyplot as plt
          # import dataset
          kashti = sns.load_dataset('titanic')
          phool = sns.load_dataset('iris')
In [ ]:
          kashti.head()
                                         sibsp
Out[ ]:
            survived pclass
                                                          fare embarked class
                                                                                        adult male deck
                                sex
                                    age
                                                parch
                                                                                  who
         0
                   0
                          3
                              male 22.0
                                                    0
                                                        7.2500
                                                                       S Third
                                                                                              True
                                                                                                    NaN
                                                                                   man
         1
                             female 38.0
                                                    0 71.2833
                                                                           First woman
                                                                                              False
                                                                                                       C
         2
                   1
                                             0
                                                        7.9250
                                                                                                    NaN
                          3
                             female 26.0
                                                    0
                                                                         Third woman
                                                                                              False
         3
                   1
                          1
                             female 35.0
                                             1
                                                    0 53.1000
                                                                           First woman
                                                                                              False
                                                                                                       \mathcal{C}
                              male 35.0
                   0
                          3
                                             0
                                                    0
                                                        8.0500
                                                                       S Third
                                                                                   man
                                                                                              True NaN
```

Out[]: sepal_length sepal_width petal_length petal_width species

In []:

phool.head()

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| | sepal_length | sepal_width | petal_length | petal_width | species |
|---|--------------|-------------|--------------|-------------|---------|
| 0 | 5.1 | 3.5 | 1.4 | 0.2 | setosa |
| 1 | 4.9 | 3.0 | 1.4 | 0.2 | setosa |
| 2 | 4.7 | 3.2 | 1.3 | 0.2 | setosa |
| 3 | 4.6 | 3.1 | 1.5 | 0.2 | setosa |
| - | = - | | | | |

To find Covariance first separate the desired variables then convert them into np.array and then apply function np.cov()

```
In [ ]:
         ks_age = kashti['age']
         ks_age.to_numpy
Out[ ]: <bound method IndexOpsMixin.to_numpy of 0
                                                           22.0
                38.0
         2
                26.0
                35.0
         3
         4
                35.0
                27.0
         886
                19.0
         888
                 NaN
         889
                26.0
         890
                32.0
         Name: age, Length: 891, dtype: float64>
In [ ]:
         ks_fare = kashti['fare']
         ks_fare.to_numpy
\texttt{Out[\ ]:\ } \texttt{<bound\ method\ IndexOpsMixin.to\_numpy\ of\ 0}
                                                            7.2500
                71.2833
         1
         2
                 7.9250
                53.1000
         3
         4
                 8.0500
         886
                13.0000
         887
                30.0000
         888
                23.4500
         889
                30.0000
         890
                 7.7500
         Name: fare, Length: 891, dtype: float64>
         cov_mat = np.stack((ks_age, ks_fare), axis = 0)
         print(np.cov(cov_mat))
         [[
                     nan 2469.43684574]]
In [ ]:
          # Another way
         np.cov(kashti['age'], kashti['fare'])
                                            nan],
Out[]: array([[
                            nan,
                            nan, 2469.43684574]])
```

As it is difficult to interpret **Covariance** so use **Correlation** instead of it.

```
In [ ]:
        kashti.info()
        <class 'pandas.core.frame.DataFrame'>
        RangeIndex: 891 entries, 0 to 890
       Data columns (total 15 columns):
           Column
                        Non-Null Count Dtype
                         -----
                       891 non-null
            survived
                                        int64
            pclass
                        891 non-null
                                       int64
                        891 non-null
            sex
                                       object
                        714 non-null
                                        float64
            age
            sibsp
                        891 non-null
                                       int64
            parch
                        891 non-null
                                       int64
```

```
891 non-null
                                            float64
         6
              fare
         7
              embarked
                           889 non-null
                                            object
         8
                           891 non-null
             class
                                            category
         9
             who
                           891 non-null
                                            object
         10
             adult_male
                           891 non-null
                                            bool
         11
             deck
                           203 non-null
                                            category
         12
             embark_town
                           889 non-null
                                            object
                           891 non-null
         13
             alive
                                            object
                           891 non-null
                                            bool
        dtypes: bool(2), category(2), float64(2), int64(4), object(5)
        memorv usage: 80.7+ KB
In [ ]:
         # Simple correlation
         kashti.corr()
```

<ipython-input-24-d27344ecd7a1>:2: FutureWarning: The default value of numeric only i n DataFrame.corr is deprecated. In a future version, it will default to False. Select only valid columns or specify the value of numeric_only to silence this warning. kashti.corr()

```
survived
Out[]:
                                     pclass
                                                            sibsp
                                                                                   fare adult_male
                                                                       parch
                                                                                                          alone
                                                  age
                                 -0.338481
                                                                               0.257307
            survived
                       1.000000
                                             -0.077221
                                                        -0.035322
                                                                    0.081629
                                                                                           -0.557080
                                                                                                      -0.203367
                                                                                            0.094035
               pclass
                       -0.338481
                                   1.000000
                                             -0.369226
                                                         0.083081
                                                                    0.018443
                                                                              -0.549500
                                                                                                       0.135207
                       -0.077221
                                  -0.369226
                                              1.000000
                                                        -0.308247
                                                                   -0.189119
                                                                               0.096067
                                                                                            0.280328
                                                                                                       0.198270
                 age
                                  0.083081
                sibsp
                       -0.035322
                                             -0.308247
                                                         1.000000
                                                                    0.414838
                                                                               0.159651
                                                                                           -0.253586
                                                                                                      -0.584471
               parch
                       0.081629
                                  0.018443
                                             -0.189119
                                                         0.414838
                                                                    1.000000
                                                                               0.216225
                                                                                           -0.349943
                                                                                                      -0.583398
                 fare
                        0.257307
                                  -0.549500
                                              0.096067
                                                         0.159651
                                                                    0.216225
                                                                               1.000000
                                                                                           -0.182024
                                                                                                      -0.271832
          adult male
                      -0.557080
                                  0.094035
                                              0.280328
                                                        -0.253586
                                                                   -0.349943
                                                                              -0.182024
                                                                                            1.000000
                                                                                                       0.404744
                                                                                            0.404744
               alone
                      -0.203367
                                  0.135207
                                              0.198270 -0.584471 -0.583398 -0.271832
                                                                                                       1.000000
```

```
In [ ]:
         # Pearson'
         corr_pearson = kashti.corr(method='pearson') # for Gaussian
```

<ipython-input-25-0d9f16e57789>:2: FutureWarning: The default value of numeric_only i n DataFrame.corr is deprecated. In a future version, it will default to False. Select only valid columns or specify the value of numeric_only to silence this warning. corr_pearson = kashti.corr(method='pearson') # for Gaussian

```
In [ ]:
         # Spearman'
         corr_spearman = kashti.corr(method='spearman') # for non Gaussian
```

<ipython-input-26-c50da91a56b0>:2: FutureWarning: The default value of numeric_only i n DataFrame.corr is deprecated. In a future version, it will default to False. Select only valid columns or specify the value of numeric_only to silence this warning. corr_spearman = kashti.corr(method='spearman') # for non Gaussian

See positive Correlation in graph

```
In [ ]:
         sns.regplot(kashti['adult_male'], kashti['alone'])
```

 $\verb|c:\Users\kalee\anaconda3\lib\site-packages\seaborn_decorators.py:36: Future \verb|Warning:|$ Pass the following variables as keyword args: x, y. From version 0.12, the only valid positional argument will be `data`, and passing other arguments without an explicit keyword will result in an error or misinterpretation. warnings.warn(

Out[]: <AxesSubplot:xlabel='adult_male', ylabel='alone'>

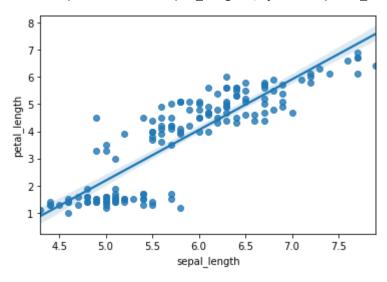
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```
In [ ]: sns.regplot(phool['sepal_length'], phool['petal_length'])
```

c:\Users\kalee\anaconda3\lib\site-packages\seaborn_decorators.py:36: FutureWarning: Pass the following variables as keyword args: x, y. From version 0.12, the only valid positional argument will be `data`, and passing other arguments without an explicit k eyword will result in an error or misinterpretation.

warnings.warn(

Out[]: <AxesSubplot:xlabel='sepal_length', ylabel='petal_length'>

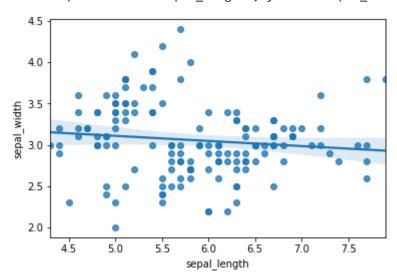


```
In [ ]: sns.regplot(phool['sepal_length'], phool['sepal_width'])
```

c:\Users\kalee\anaconda3\lib\site-packages\seaborn_decorators.py:36: FutureWarning: Pass the following variables as keyword args: x, y. From version 0.12, the only valid positional argument will be `data`, and passing other arguments without an explicit k eyword will result in an error or misinterpretation.

warnings.warn(

Out[]: <AxesSubplot:xlabel='sepal_length', ylabel='sepal_width'>



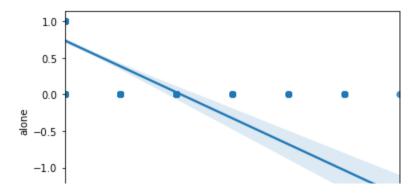
See negative Correlation in graph

```
In [ ]: sns.regplot(kashti['parch'], kashti['alone'])
```

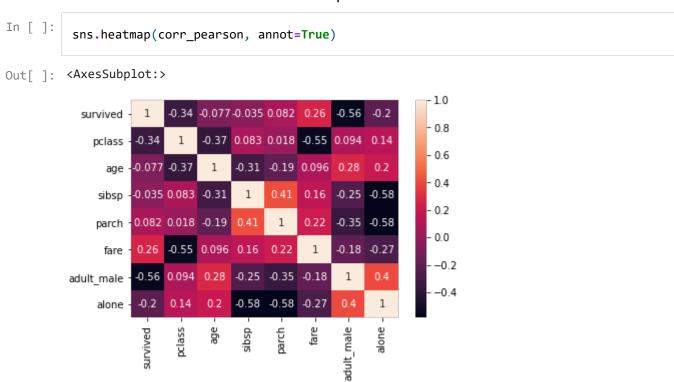
c:\Users\kalee\anaconda3\lib\site-packages\seaborn_decorators.py:36: FutureWarning: Pass the following variables as keyword args: x, y. From version 0.12, the only valid positional argument will be `data`, and passing other arguments without an explicit k eyword will result in an error or misinterpretation.

warnings.warn(

Out[]: <AxesSubplot:xlabel='parch', ylabel='alone'>



Now see correlation in Heatmap



When **Correlation** is more then 0.6 toward 1.0, than it is called highly **+ve Correlated** and below -.06 is called **-ve Correlated** depending on the data

More better representation of Heatmap

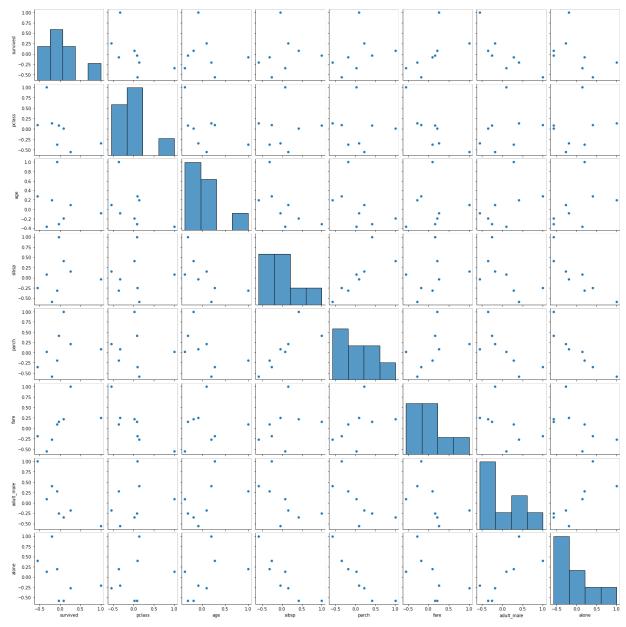
```
In [ ]:
         corr_pearson.style.background_gradient(cmap='coolwarm')
                                                   Traceback (most recent call last)
        <ipython-input-32-1926c9d04fbe> in <module>
        ----> 1 corr_pearson.style.background_gradient(cmap='coolwarm')
        c:\Users\kalee\anaconda3\lib\site-packages\pandas\core\frame.py in style(self)
                        data with HTML and CSS.
           1263
        -> 1264
                        from pandas.io.formats.style import Styler
           1265
           1266
                        return Styler(self)
        c:\Users\kalee\anaconda3\lib\site-packages\pandas\io\formats\style.py in <module>
             54 from pandas.io.formats.format import save_to_buffer
        ---> 56 jinja2 = import_optional_dependency("jinja2", extra="DataFrame.style requires
        jinja2.")
             57
             58 from pandas.io.formats.style_render import (
        c:\Users\kalee\anaconda3\lib\site-packages\pandas\compat\_optional.py in import_optio
        nal_dependency(name, extra, errors, min_version)
            169
                                return None
                            elif errors == "raise":
            170
        --> 171
                                raise ImportError(msg)
            172
            173
                    return module
```

ImportError: Pandas requires version '3.0.0' or newer of 'jinja2' (version '2.11.3' c
urrently installed).

Draw the pairplot now

```
In [ ]: sns.pairplot(corr_pearson)
```

Out[]: <seaborn.axisgrid.PairGrid at 0x20118aae8b0>



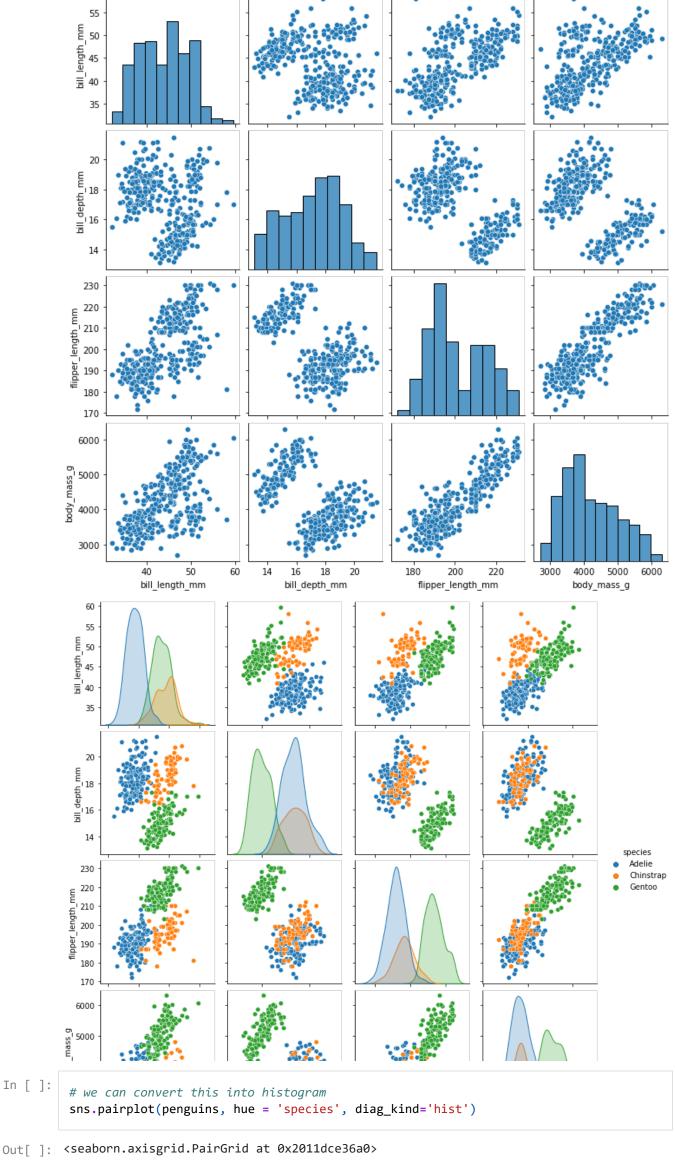
```
In [ ]:
    # We can change the points in pairplot based on category
    # import a new dataset
    penguins = sns.load_dataset('penguins')
    penguins.head()
```

| Out[]: | | species | island | bill_length_mm | bill_depth_mm | flipper_length_mm | body_mass_g | sex |
|---------|---|---------|-----------|----------------|---------------|-------------------|-------------|--------|
| | 0 | Adelie | Torgersen | 39.1 | 18.7 | 181.0 | 3750.0 | Male |
| | 1 | Adelie | Torgersen | 39.5 | 17.4 | 186.0 | 3800.0 | Female |
| | 2 | Adelie | Torgersen | 40.3 | 18.0 | 195.0 | 3250.0 | Female |
| | 3 | Adelie | Torgersen | NaN | NaN | NaN | NaN | NaN |
| | 4 | Adelie | Torgersen | 36.7 | 19.3 | 193.0 | 3450.0 | Female |

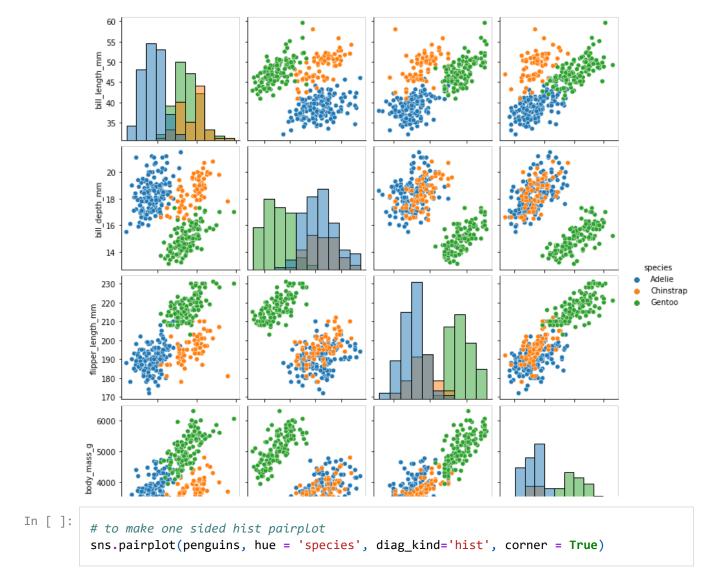
```
In [ ]: sns.pairplot(penguins)
sns.pairplot(penguins, hue = 'species')
```

Out[]: <seaborn.axisgrid.PairGrid at 0x2011b817670>

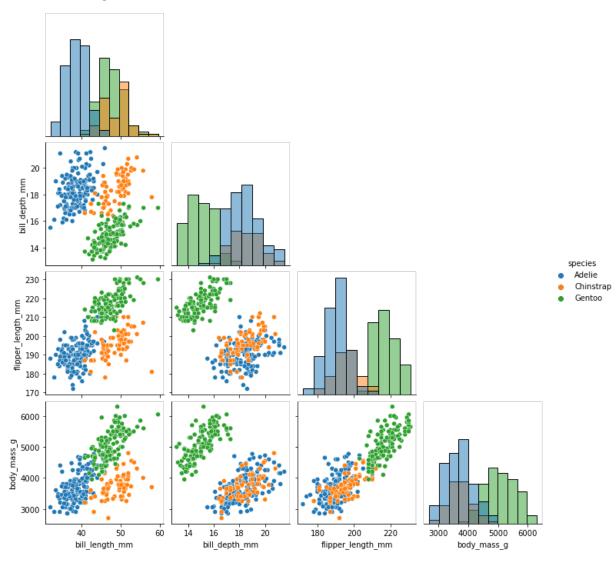
60



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Out[]: <seaborn.axisgrid.PairGrid at 0x2011b684250>



Now use of the scipy stats library for pearson's correlation

```
In [ ]:
          phool.head()
            sepal_length sepal_width petal_length petal_width species
Out[]:
         0
                     5.1
                                                           0.2
                                  3.5
                                               1.4
                                                                setosa
          1
                     4.9
                                  3.0
                                               1.4
                                                           0.2
                                                                setosa
          2
                     4.7
                                  3.2
                                               1.3
                                                           0.2
                                                                setosa
         3
                     4.6
                                  3.1
                                               1.5
                                                           0.2
                                                                setosa
                     5.0
                                  3.6
                                               1.4
                                                           0.2
                                                                setosa
In [ ]:
          # calculate pearson's correlation
          from scipy.stats import pearsonr # and spearsman
          corr, _ = pearsonr(phool['sepal_length'], phool['petal_length'])
          print('Pearsons Correlation: %.3f' % corr)
         Pearsons Correlation: 0.872
```

ASSIGNMENTS

four types of graphs:

- 1. +ve correlation
- 2. -ve correlation
- 3. 0 correlation
- 4. slightly +ve correlation

Sollutions

Load a new dataset

corr_df = df.corr('pearson')

Out[]:

```
In [ ]:
           df = pd.read_csv('D:/Python_Ka_Chilla_Data/Data_Sets/House_Data1.csv')
           df.head()
Out[]:
                            Gas
                     E. Bill
             Month
                                  Grocery
                                           Milk Fruit Meet Wegetables Medicines Eid_guests_pocketmonies
                             Bill
          0
                      1287
                                    26000
                                           7200
                                                 4580
                                                        3800
                                                                     2310
                                                                                6580
                                                                                                         11000
               May
                            250
          1
                Jun
                      7309
                            280
                                     7200
                                           7000
                                                 3400
                                                        4100
                                                                     2500
                                                                                9340
                                                                                                        100500
          2
                     11490
                            270
                                    11500
                                           7350
                                                 4800
                                                        3650
                                                                     2700
                                                                                2960
                                                                                                         52710
                 Jul
          3
                     19184
                            250
                                     5400 7200
                                                 4100
                                                        4000
                                                                     2200
                                                                                2000
                                                                                                         70650
                Aug
                                                                     2400
                    20065 260
                                    12310 6900 5200
                                                        4700
                                                                                3450
                                                                                                         34560
                Sep
In [ ]:
           df.columns
Out[]: Index(['Month', 'E. Bill', 'Gas Bill', 'Grocery', 'Milk', 'Fruit', 'Meet',
                  'Wegetables', 'Medicines', 'Eid_guests_pocketmonies', 'Maintenance', 'house_wares', 'Inflation rate', 'Buying Power(PKR)', 'Saving'],
                  'house_wares'
                 dtype='object')
In [ ]:
           corr_df = df.corr('pearson')
           corr_df
          <ipython-input-42-b57e8b96ad48>:1: FutureWarning: The default value of numeric_only i
```

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Gas Bill

E. Bill

n DataFrame.corr is deprecated. In a future version, it will default to False. Select only valid columns or specify the value of numeric_only to silence this warning.

Grocery

Milk

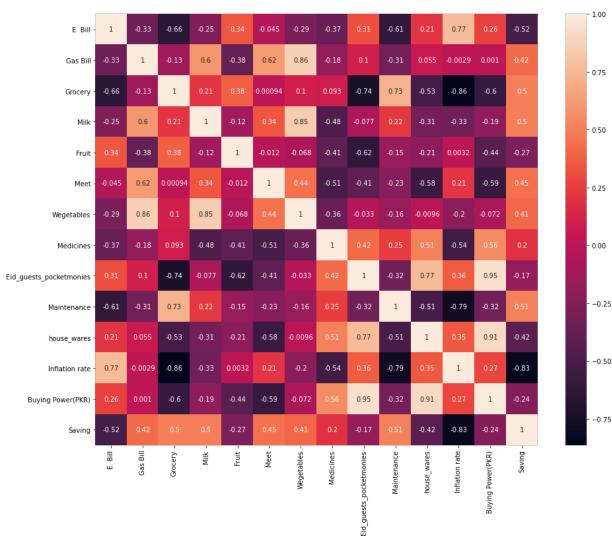
Fruit

Meet Wegetables

| | E. Bill | Gas Bill | Grocery | Milk | Fruit | Meet | Wegetables |
|-------------------------|-----------|-----------|-----------|-----------|-----------|-----------|------------|
| E. Bill | 1.000000 | -0.333596 | -0.658929 | -0.252684 | 0.338906 | -0.044632 | -0.287951 |
| Gas Bill | -0.333596 | 1.000000 | -0.134358 | 0.600796 | -0.381480 | 0.623686 | 0.864663 |
| Grocery | -0.658929 | -0.134358 | 1.000000 | 0.210455 | 0.376687 | 0.000939 | 0.100573 |
| Milk | -0.252684 | 0.600796 | 0.210455 | 1.000000 | -0.117434 | 0.341335 | 0.847616 |
| Fruit | 0.338906 | -0.381480 | 0.376687 | -0.117434 | 1.000000 | -0.012416 | -0.067974 |
| Meet | -0.044632 | 0.623686 | 0.000939 | 0.341335 | -0.012416 | 1.000000 | 0.44374 |
| Wegetables | -0.287951 | 0.864663 | 0.100573 | 0.847616 | -0.067974 | 0.443741 | 1.000000 |
| Medicines | -0.367530 | -0.175674 | 0.093385 | -0.482648 | -0.408359 | -0.507341 | -0.360418 |
| Eid_guests_pocketmonies | 0.310057 | 0.104797 | -0.738716 | -0.077111 | -0.616861 | -0.409191 | -0.033228 |
| Maintenance | -0.611900 | -0.310782 | 0.728834 | 0.224218 | -0.150142 | -0.228264 | -0.156502 |
| house_wares | 0.208273 | 0.055266 | -0.525829 | -0.306631 | -0.205594 | -0.575040 | -0.009608 |
| Inflation rate | 0.767361 | -0.002879 | -0.863095 | -0.330068 | 0.003213 | 0.211376 | -0.201887 |

In

Out[]: <AxesSubplot:>



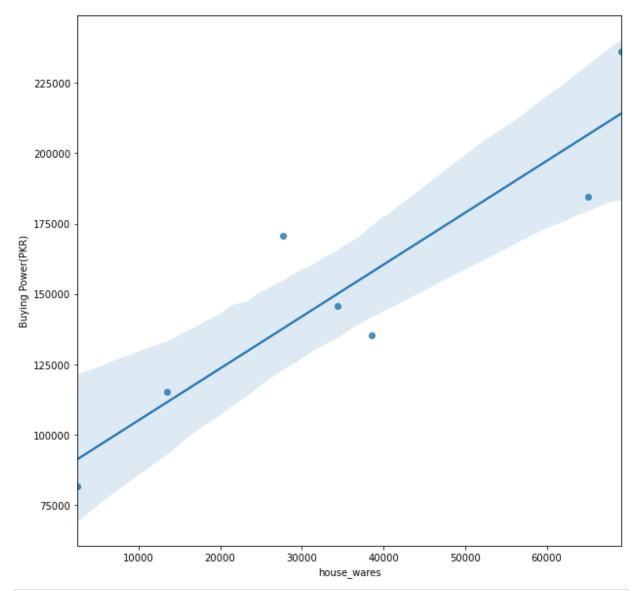
+ve correlation

```
In [ ]:
         plt.figure(figsize=(10,10))
         sns.regplot(df['house_wares'], df['Buying Power(PKR)'])
```

c:\Users\kalee\anaconda3\lib\site-packages\seaborn_decorators.py:36: FutureWarning:
Pass the following variables as keyword args: x, y. From version 0.12, the only valid
positional argument will be `data`, and passing other arguments without an explicit k eyword will result in an error or misinterpretation. warnings.warn(

Out[]: <AxesSubplot:xlabel='house_wares', ylabel='Buying Power(PKR)'>

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```
In [ ]:
    plt.figure(figsize=(10,10))
    sns.regplot(df['Eid_guests_pocketmonies'], df['Buying Power(PKR)'])
```

c:\Users\kalee\anaconda3\lib\site-packages\seaborn_decorators.py:36: FutureWarning:
Pass the following variables as keyword args: x, y. From version 0.12, the only valid
positional argument will be `data`, and passing other arguments without an explicit k
eyword will result in an error or misinterpretation.
 warnings.warn(

Out[]: <AxesSubplot:xlabel='Eid_guests_pocketmonies', ylabel='Buying Power(PKR)'>

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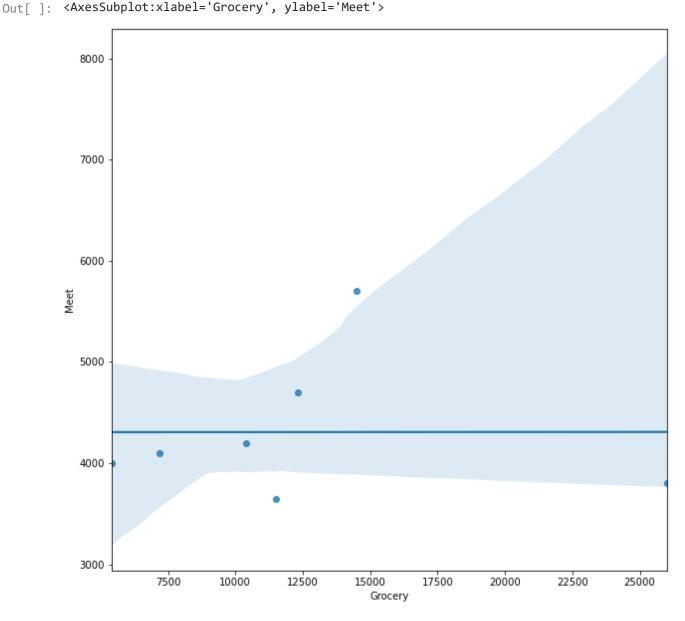
```
300000 -
```

2. 0 correlation

```
In [ ]: plt.figure(figsize=(10,10))
    sns.regplot(df['Grocery'], df['Meet'])
```

c:\Users\kalee\anaconda3\lib\site-packages\seaborn_decorators.py:36: FutureWarning:
Pass the following variables as keyword args: x, y. From version 0.12, the only valid
positional argument will be `data`, and passing other arguments without an explicit k
eyword will result in an error or misinterpretation.
 warnings.warn(

(Augustus and Longraphy) India | Mage

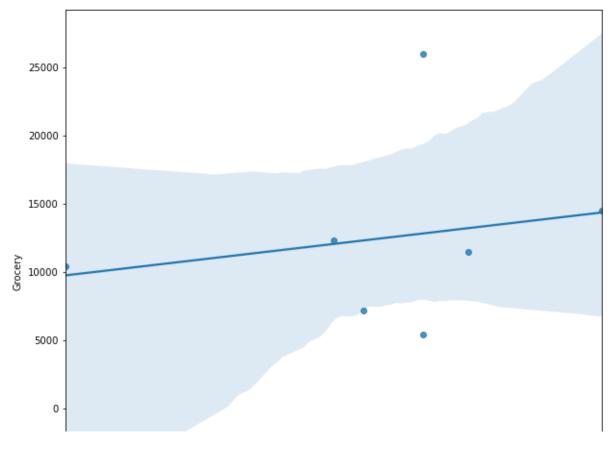


3. Slightly +ve Correlation

```
In [ ]:
    plt.figure(figsize=(10,10))
    sns.regplot(df['Milk'], df['Grocery'])
```

c:\Users\kalee\anaconda3\lib\site-packages\seaborn_decorators.py:36: FutureWarning:
Pass the following variables as keyword args: x, y. From version 0.12, the only valid
positional argument will be `data`, and passing other arguments without an explicit k
eyword will result in an error or misinterpretation.
 warnings.warn(

Out[]: <AxesSubplot:xlabel='Milk', ylabel='Grocery'>

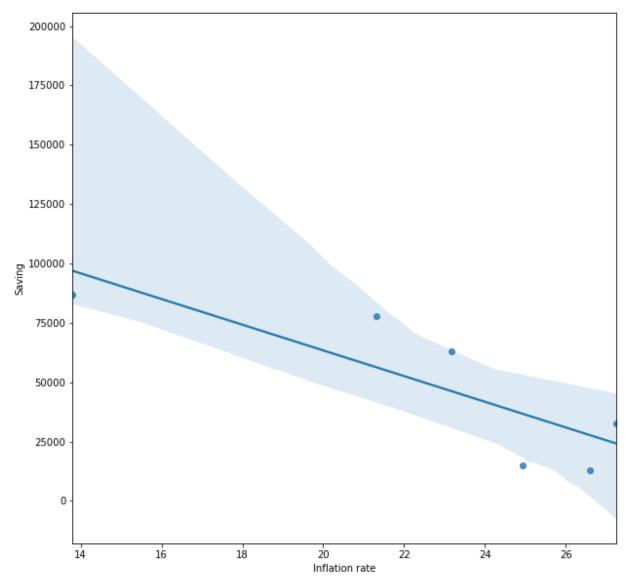


4. -ve Correlation

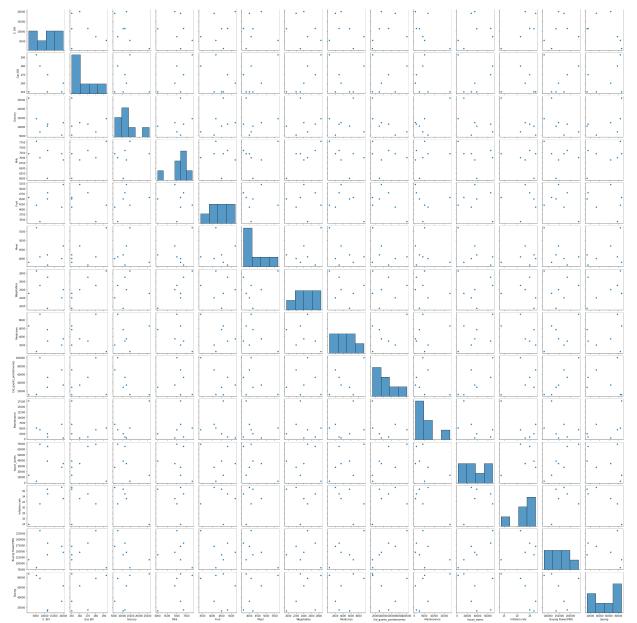
```
In [ ]:
    plt.figure(figsize=(10,10))
    sns.regplot(df['Inflation rate'], df['Saving'])
```

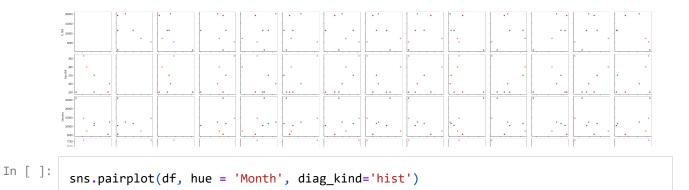
c:\Users\kalee\anaconda3\lib\site-packages\seaborn_decorators.py:36: FutureWarning:
Pass the following variables as keyword args: x, y. From version 0.12, the only valid
positional argument will be `data`, and passing other arguments without an explicit k
eyword will result in an error or misinterpretation.
 warnings.warn(

Out[]: <AxesSubplot:xlabel='Inflation rate', ylabel='Saving'>



Out[]: <seaborn.axisgrid.PairGrid at 0x201218978b0>





Out[]: <seaborn.axisgrid.PairGrid at 0x2012899c040>

