Correlation

- variable within dataset can be related for lots of reasons
- Types
 - Pearson's r
 - spearman,s rho
 - kendall's tau

For Example:

- 1. One variable could cause or depend in the value of another variable.
- 2. one variable could be lightly associated with another variable.
- 3. Two variable could depend on a third unknown variable.

Positive correlation: both variables change in same direction\ **Neutral correlation:** no relationship in the change of variable\ **Negative correlation:** variables change in oppositie direction

Covariance

- Variables can be related by a linear relationship. this is a relationship that is constantly additive across the two data sample.
- This relationship can be summerized 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 change in the different direct (negative).
- The magnitude of the covariance is not easily interpreted. A covariance value of zero indicates that both variables are completely independent.

```
In []: # import Library
import numpy as np
import seaborn as sns
import pandas as pd
import matplotlib.pyplot as plt
# import dataset
kashti = sns.load_dataset('titanic')
phool = sns.load_dataset('iris')
In []: kashti.head()
```

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```
Out[ ]:
           survived pclass
                                 age sibsp parch
                                                    fare embarked class
                                                                          who adult male d
                            sex
        0
                       3
                                22.0
                                                  7.2500
                                                                S Third
                                                                                    True
                           male
                                                                          man
        1
                                38.0
                       1 female
                                                 71.2833
                                                                C
                                                                   First woman
                                                                                    False
        2
                 1
                         female
                                26.0
                                        0
                                                  7.9250
                                                                  Third
                                                                                    False
                                                                        woman
        3
                       1 female
                                35.0
                                                 53.1000
                                                                   First woman
                                                                                    False
        4
                 0
                           male 35.0
                                        0
                                                  8.0500
                                                                 Third
                                                                          man
                                                                                    True
        np.cov(kashti['age'], kashti['fare']) # numpy fucntion tahts why we use np.cov
        # also can find 'google' more libraries of covariance
                                         nan],
        array([[
                          nan,
Out[]:
                          nan, 2469.43684574]])
        # ANTOHER Example
In [ ]:
        import numpy as np
        x= kashti['age']
        y= kashti['fare']
        # find out covariance with respect to columns
        cov_mat = np.stack((x,y), axis = 0)
        cov_mat
                                                   nan, 26.
        array([[22.
                       , 38.
                                , 26.
                                                                         ],
Out[]:
               [ 7.25 , 71.2833, 7.925 , ..., 23.45 , 30.
                                                                , 7.75 ]])
        print(np.cov(cov_mat))
In [ ]:
        # nan because of null values
        [[
                    nan
                                  nan]
                    nan 2469.43684574]]

    Corelation istead of COV

        kashti.columns
In [ ]:
        Out[ ]:
               'alive', 'alone'],
              dtype='object')
        kashti.info()
In [ ]:
        # corelation always lagta ha against int or float dtypes
```

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<class 'pandas.core.frame.DataFrame'>
RangeIndex: 891 entries, 0 to 890
Data columns (total 15 columns):

Column Non-Null Count Dtype --------_____ _ _ _ _ 0 survived 891 non-null int64 891 non-null int64 1 pclass 2 891 non-null object sex 3 714 non-null float64 age int64 4 sibsp 891 non-null 5 parch 891 non-null int64 fare 891 non-null float64 6 7 embarked 889 non-null object 8 class 891 non-null category object 9 who 891 non-null 891 non-null 10 adult_male bool 203 non-null 11 deck category 12 889 non-null embark_town object 13 alive 891 non-null object alone 891 non-null bool

dtypes: bool(2), category(2), float64(2), int64(4), object(5)

memory usage: 80.7+ KB

In []: # corelation at kashti dataset
 kashti.corr()
 # corr function is builtin
 # 1 highly +ve, -1 highly -ve , 0 is no corelation

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

In []: kashti.corr(method='pearson') # for normal data

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

-0.253586

-0.584471

-0.349943

-0.583398

-0.182024

-0.271832

0.280328

0.198270

0.094035

0.135207

-0.557080

-0.203367

adult_male

alone

0.404744

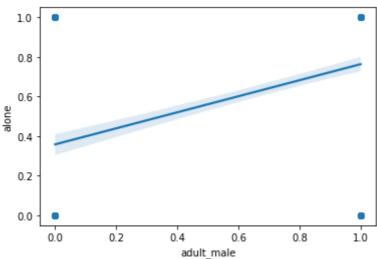
1.000000

1.000000

0.404744

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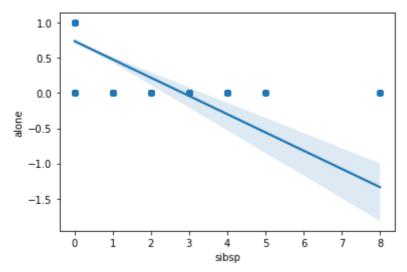
```
08 Correlation
4
             kashti.corr(method='spearman') # for non gaussian distrubution data
   Out[]:
                         survived
                                      pclass
                                                           sibsp
                                                                     parch
                                                                                 fare adult_male
                                                                                                      alone
                                                  age
                         1.000000
                                   -0.339668
                                             -0.052565
                                                        0.088879
                                                                   0.138266
                                                                             0.323736
                                                                                        -0.557080
                                                                                                   -0.203367
               survived
                                                                  -0.022801
                                                                            -0.688032
                         -0.339668
                                    1.000000
                                             -0.361666
                                                       -0.043019
                                                                                         0.099351
                                                                                                   0.135896
                 pclass
                         -0.052565
                                   -0.361666
                                              1.000000
                                                        -0.182061
                                                                  -0.254212
                                                                             0.135051
                                                                                         0.257641
                                                                                                   0.167293
                   age
                                                                                        -0.307948
                                             -0.182061
                                                        1.000000
                                                                   0.450014
                                                                                                  -0.828215
                  sibsp
                         0.088879
                                   -0.043019
                                                                             0.447113
                         0.138266
                                   -0.022801
                                             -0.254212
                                                        0.450014
                                                                   1.000000
                                                                             0.410074
                                                                                        -0.397286
                                                                                                   -0.683719
                  parch
                                              0.135051
                                                                  0.410074
                         0.323736
                                   -0.688032
                                                        0.447113
                                                                             1.000000
                                                                                        -0.308894
                                                                                                   -0.531472
                   fare
             adult male
                         -0.557080
                                   0.099351
                                              0.257641
                                                        -0.307948
                                                                  -0.397286
                                                                            -0.308894
                                                                                         1.000000
                                                                                                   0.404744
                        -0.203367
                                   0.135896
                                              0.167293
                                                       -0.828215
                                                                 -0.683719
                                                                           -0.531472
                                                                                         0.404744
                                                                                                   1.000000
                  alone
             corrp = kashti.corr(method='pearson') # store command as corrp
   In [ ]:
   In [ ]:
             # +ve corealtion
             sns.regplot(kashti['adult_male'], kashti['alone'], data=kashti)
             C:\Users\Junaid\anaconda3\lib\site-packages\seaborn\_decorators.py:36: FutureWarni
             ng: Pass the following variables as keyword args: x, y. From version 0.12, the onl
             y valid positional argument will be `data`, and passing other arguments without an
             explicit keyword will result in an error or misinterpretation.
               warnings.warn(
             <AxesSubplot:xlabel='adult_male', ylabel='alone'>
   Out[ ]:
               1.0
```



```
# -ve corealtion
In [ ]:
        sns.regplot(kashti['sibsp'], kashti['alone'], data=kashti)
```

C:\Users\Junaid\anaconda3\lib\site-packages\seaborn\ decorators.py:36: FutureWarni ng: Pass the following variables as keyword args: x, y. From version 0.12, the onl y valid positional argument will be `data`, and passing other arguments without an explicit keyword will result in an error or misinterpretation. warnings.warn(

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



In []: # at phool dataset
phool.head()

Out[]:		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
	4	5.0	3.6	1.4	0.2	setosa

In []: phool.corr(method='spearman')

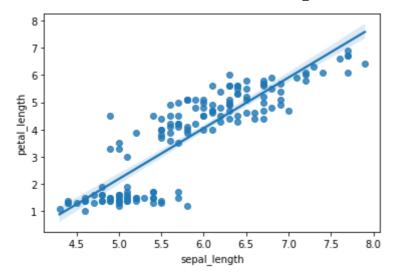
Out[]:

	sepal_length	sepal_width	petal_length	petal_width
sepal_length	1.000000	-0.166778	0.881898	0.834289
sepal_width	-0.166778	1.000000	-0.309635	-0.289032
petal_length	0.881898	-0.309635	1.000000	0.937667
petal_width	0.834289	-0.289032	0.937667	1.000000

```
In [ ]: # +ve corealtion
sns.regplot(phool['sepal_length'], phool['petal_length'], data=phool)
```

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

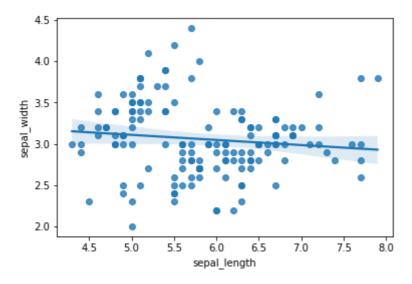
Out[]: <AxesSubplot:xlabel='sepal_length', ylabel='petal_length'>



```
In [ ]: sns.regplot(phool['sepal_length'], phool['sepal_width'], data=phool)
# almost 0 corraltion
```

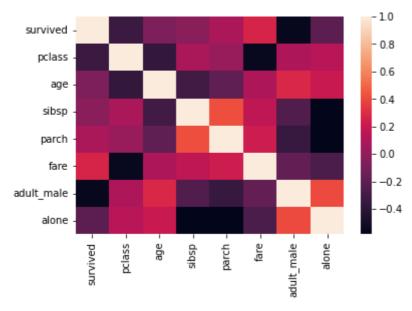
C:\Users\Junaid\anaconda3\lib\site-packages\seaborn_decorators.py:36: FutureWarni
ng: Pass the following variables as keyword args: x, y. From version 0.12, the onl
y valid positional argument will be `data`, and passing other arguments without an
explicit keyword will result in an error or misinterpretation.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

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



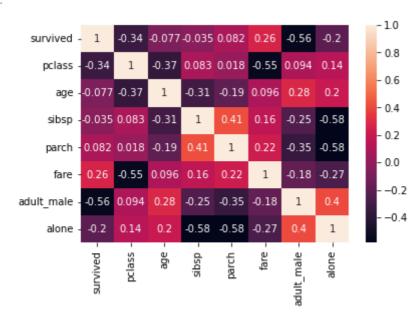
```
In [ ]: # apply corr function
    corr = kashti.corr(method='pearson')
    sns.heatmap(corr)
```

Out[]: <AxesSubplot:>



In []: sns.heatmap(corr, annot=True)

Out[]: <AxesSubplot:>



In []: # ANOTHER STYLE
 corr.style.background_gradient(cmap='coolwarm')

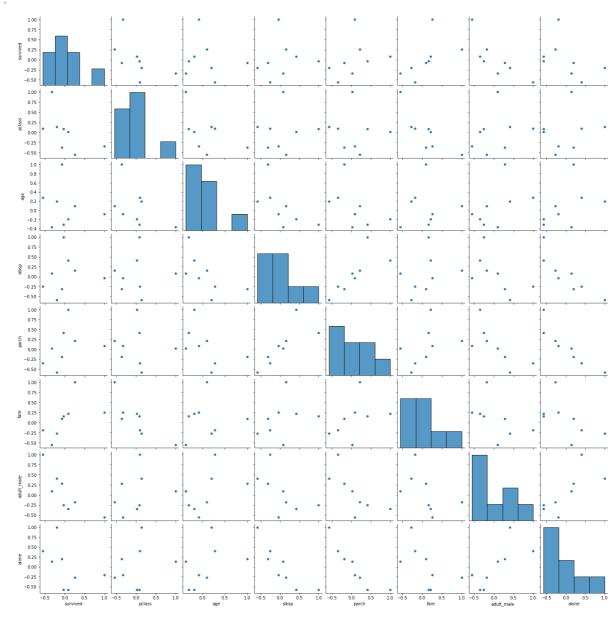
]:		survived	pclass	age	sibsp	parch	fare	adult_male	alone
	survived	1.000000	-0.338481	-0.077221	-0.035322	0.081629	0.257307	-0.557080	-0.203367
	pclass	-0.338481	1.000000	-0.369226	0.083081	0.018443	-0.549500	0.094035	0.135207
	age	-0.077221	-0.369226	1.000000	-0.308247	-0.189119	0.096067	0.280328	0.198270
	sibsp	-0.035322	0.083081	-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
	alone	-0.203367	0.135207	0.198270	-0.584471	-0.583398	-0.271832	0.404744	1.000000
									—

In []: # pairplot

Out[

sns.pairplot(corr)

Out[]: <seaborn.axisgrid.PairGrid at 0x1c145e0c310>



```
In [ ]: # new dataset penguins
    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
		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

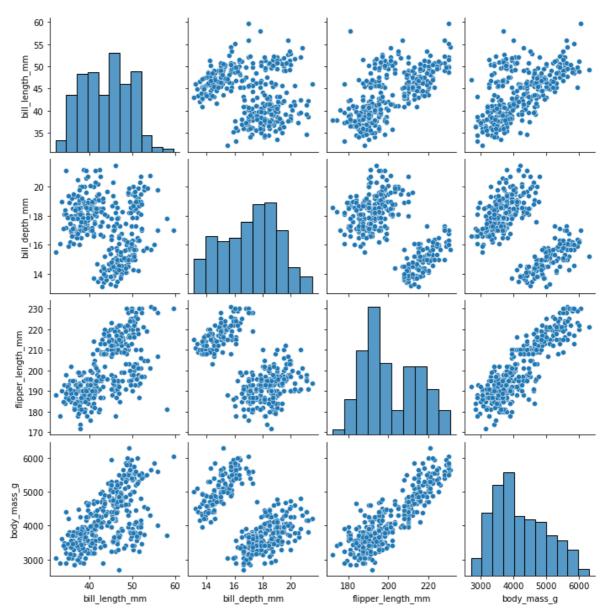
```
In [ ]: penguins.corr(method='spearman')
```

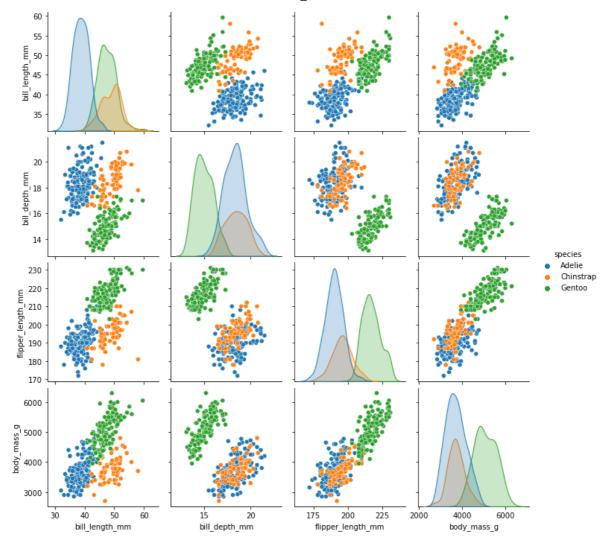
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Out[]: bill_length_mm bill_depth_mm flipper_length_mm body_mass_g bill_length_mm 1.000000 -0.221749 0.672772 0.583800 bill_depth_mm -0.221749 1.000000 -0.523267 -0.432372 flipper_length_mm -0.523267 1.000000 0.839974 0.672772 body_mass_g 0.583800 -0.432372 0.839974 1.000000

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

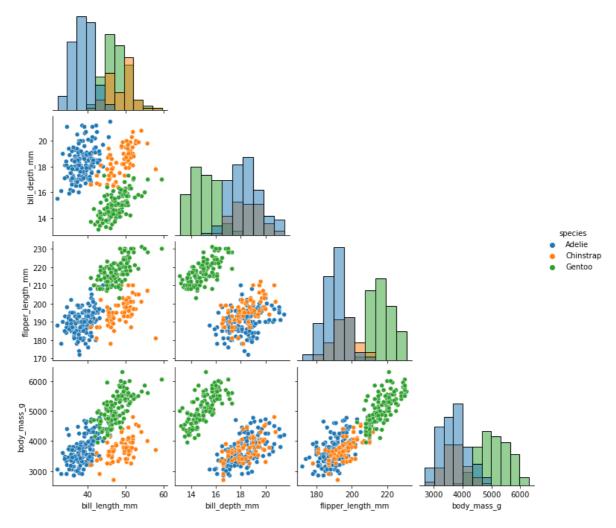
Out[]: <seaborn.axisgrid.PairGrid at 0x1c149f0a280>





In []: # make one side
sns.pairplot(penguins, hue='species', diag_kind='hist', corner=True)

Out[]: <seaborn.axisgrid.PairGrid at 0x1c14ba9ef70>



In []: # calculate pearson's corraltion
phool.head()

Out[]:		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
	4	5.0	3.6	1 4	0.2	setosa

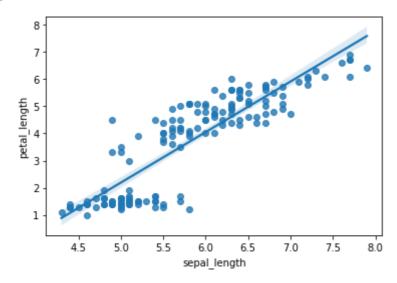
4 types of plots with all types of corelation

• +ve correlation in phool dataset

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

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

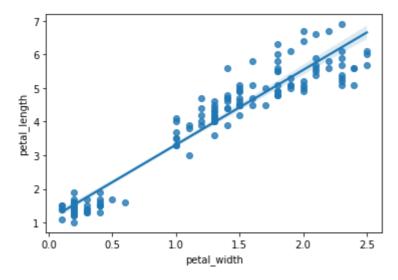
Out[]: <AxesSubplot:xlabel='sepal_length', ylabel='petal_length'>



In []: sns.regplot(phool['petal_width'], phool['petal_length'], data=phool)

C:\Users\Junaid\anaconda3\lib\site-packages\seaborn_decorators.py:36: FutureWarni
ng: Pass the following variables as keyword args: x, y. From version 0.12, the onl
y 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='petal_width', ylabel='petal_length'>

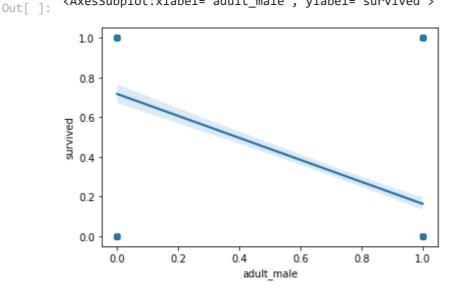


• -ve correlation in kashti dataset

```
In [ ]: sns.regplot(kashti['adult_male'], kashti['survived'], data=kashti)
```

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

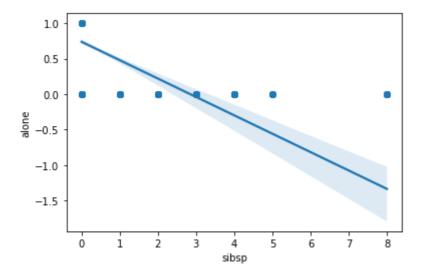
<AxesSubplot:xlabel='adult_male', ylabel='survived'>



```
In [ ]: sns.regplot(kashti['sibsp'], kashti['alone'], data=kashti)
```

C:\Users\Junaid\anaconda3\lib\site-packages\seaborn_decorators.py:36: FutureWarni
ng: Pass the following variables as keyword args: x, y. From version 0.12, the onl
y 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='sibsp', ylabel='alone'>



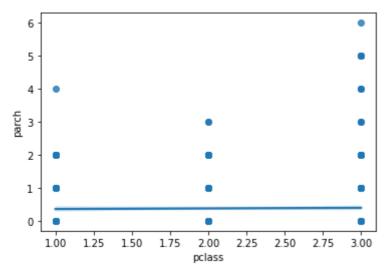
• neutral correlation in kashti dataset

```
In [ ]: sns.regplot(kashti['pclass'], kashti['parch'], data=kashti)
```

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

nutf l. <AxesSubplot:xlabel='pclass', ylabel='parch'>

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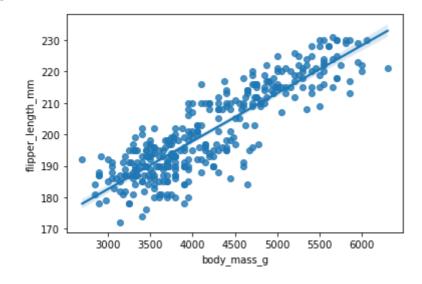


correlations in penguin dataset

```
In [ ]: # +ve corelation
sns.regplot(penguins['body_mass_g'], penguins['flipper_length_mm'], data=penguins)
```

C:\Users\Junaid\anaconda3\lib\site-packages\seaborn_decorators.py:36: FutureWarni
ng: Pass the following variables as keyword args: x, y. From version 0.12, the onl
y 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='body_mass_g', ylabel='flipper_length_mm'>

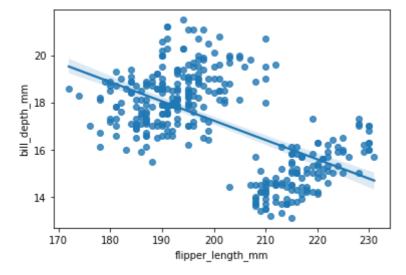


```
In [ ]: # -ve corelation
sns.regplot(penguins['flipper_length_mm'], penguins['bill_depth_mm'], data=penguins
```

C:\Users\Junaid\anaconda3\lib\site-packages\seaborn_decorators.py:36: FutureWarni
ng: Pass the following variables as keyword args: x, y. From version 0.12, the onl
y 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='flipper_length_mm', ylabel='bill_depth_mm'>

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In []: # zero corelation
sns.regplot(penguins['bill_depth_mm'], penguins['bill_length_mm'], data=penguins)

C:\Users\Junaid\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 keyword will result in an error or misinterpretation.

warnings.warn(

Out[]: <AxesSubplot:xlabel='bill_depth_mm', ylabel='bill_length_mm'>

