

```
import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
%matplotlib inline
import seaborn as sns
```

In [ ]:

```
df = pd.read_csv('/content/abalone.csv')
```

In [ ]:

```
df.head()
```

Out[ ]:

	Sex	Length	Diameter	Height	Whole weight	Shucked weight	Viscera weight	Shell weight	Rings
0	M	0.455	0.365	0.095	0.5140	0.2245	0.1010	0.150	15
1	M	0.350	0.265	0.090	0.2255	0.0995	0.0485	0.070	7
2	F	0.530	0.420	0.135	0.6770	0.2565	0.1415	0.210	9
3	M	0.440	0.365	0.125	0.5160	0.2155	0.1140	0.155	10
4	I	0.330	0.255	0.080	0.2050	0.0895	0.0395	0.055	7

In [ ]:

```
df.describe()
```

Out[ ]:

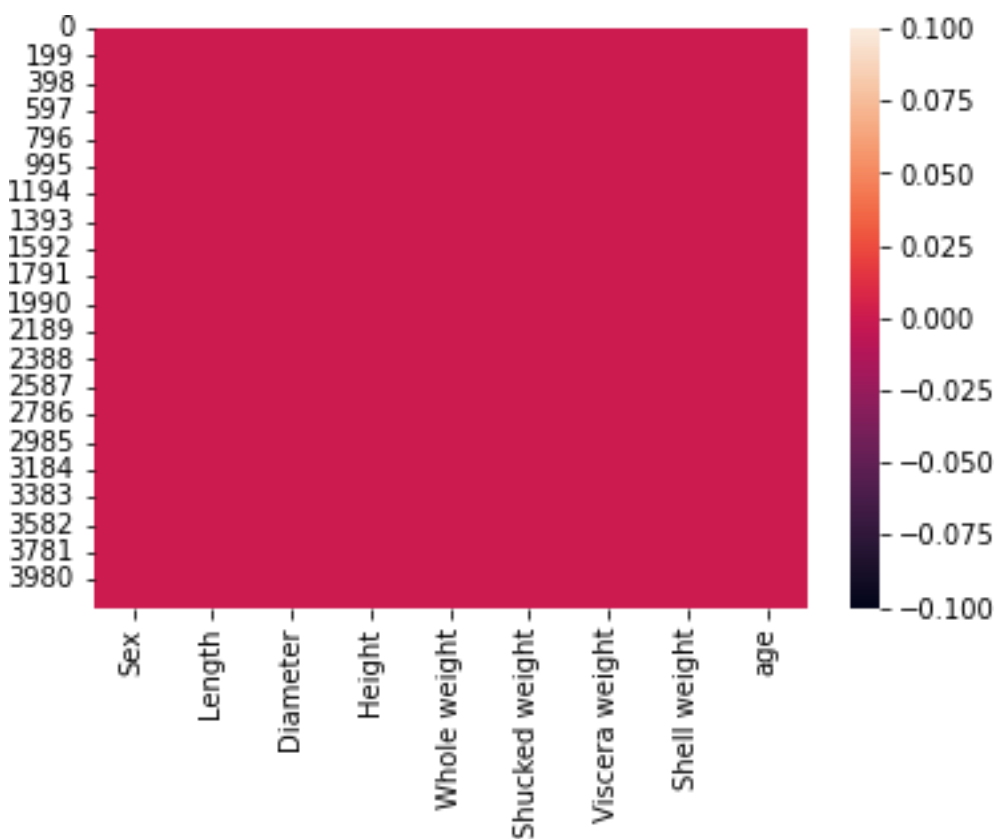
	Length	Diameter	Height	Whole weight	Shucked weight	Viscera weight	Shell weight	Rings
count	4177.000000	4177.000000	4177.000000	4177.000000	4177.000000	4177.000000	4177.000000	4177.000000
mean	0.523992	0.407881	0.139516	0.828742	0.359367	0.180594	0.238831	9.933684
std	0.120093	0.099240	0.041827	0.490389	0.221963	0.109614	0.139203	3.224169
min	0.075000	0.055000	0.000000	0.002000	0.001000	0.000500	0.001500	1.000000
25%	0.450000	0.350000	0.115000	0.441500	0.186000	0.093500	0.130000	8.000000

	Length	Diameter	Height	Whole weight	Shucked weight	Viscera weight	Shell weight	Rings
<b>50%</b>	0.545000	0.425000	0.140000	0.799500	0.336000	0.171000	0.234000	9.000000
<b>75%</b>	0.615000	0.480000	0.165000	1.153000	0.502000	0.253000	0.329000	11.000000
<b>max</b>	0.815000	0.650000	1.130000	2.825500	1.488000	0.760000	1.005000	29.000000

In [ ]:

```
sns.heatmap(df.isnull())
```

Out[ ]:



In [ ]:

```
df.info()
```

```
RangeIndex: 4177 entries, 0 to 4176
```

```
Data columns (total 9 columns):
```

#	Column	Non-Null Count	Dtype
0	Sex	4177 non-null	object
1	Length	4177 non-null	float64
2	Diameter	4177 non-null	float64
3	Height	4177 non-null	float64

```

4   Whole weight    4177 non-null    float64
5   Shucked weight  4177 non-null    float64
6   Viscera weight  4177 non-null    float64
7   Shell weight    4177 non-null    float64
8   age             4177 non-null    float64

```

```
dtypes: float64(8), object(1)
```

```
memory usage: 293.8+ KB
```

In [ ]:

```
numerical_features
```

Out [ ]:

```

Index(['Length', 'Diameter', 'Height', 'Whole weight', 'Shucked weight',
      'Viscera weight', 'Shell weight', 'age'],
      dtype='object')

```

In [ ]:

```
categorical_features
```

Out [ ]:

```
Index(['Sex'], dtype='object')
```

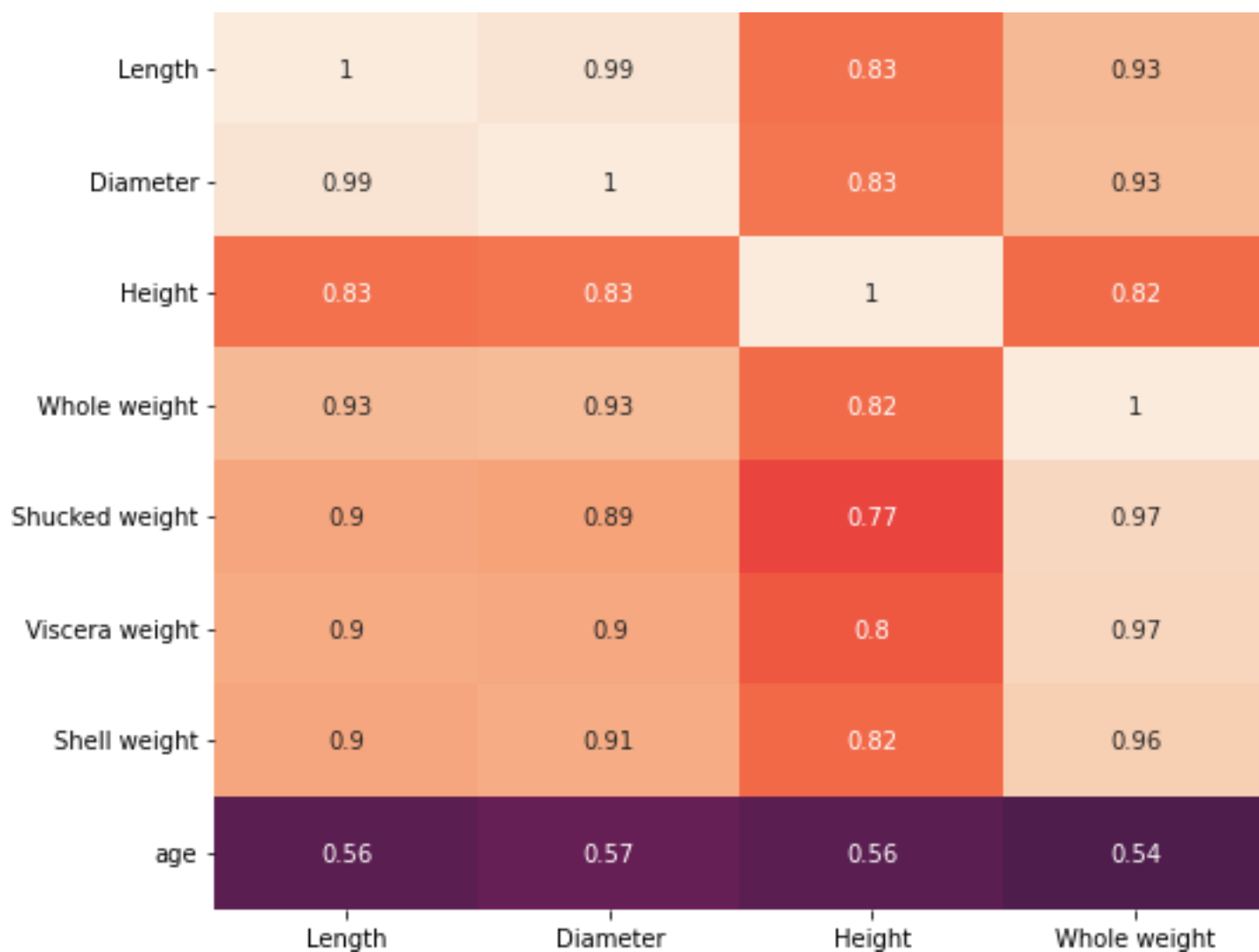
In [ ]:

```

plt.figure(figsize = (20,7))
sns.heatmap(df[numerical_features].corr(),annot = True)

```

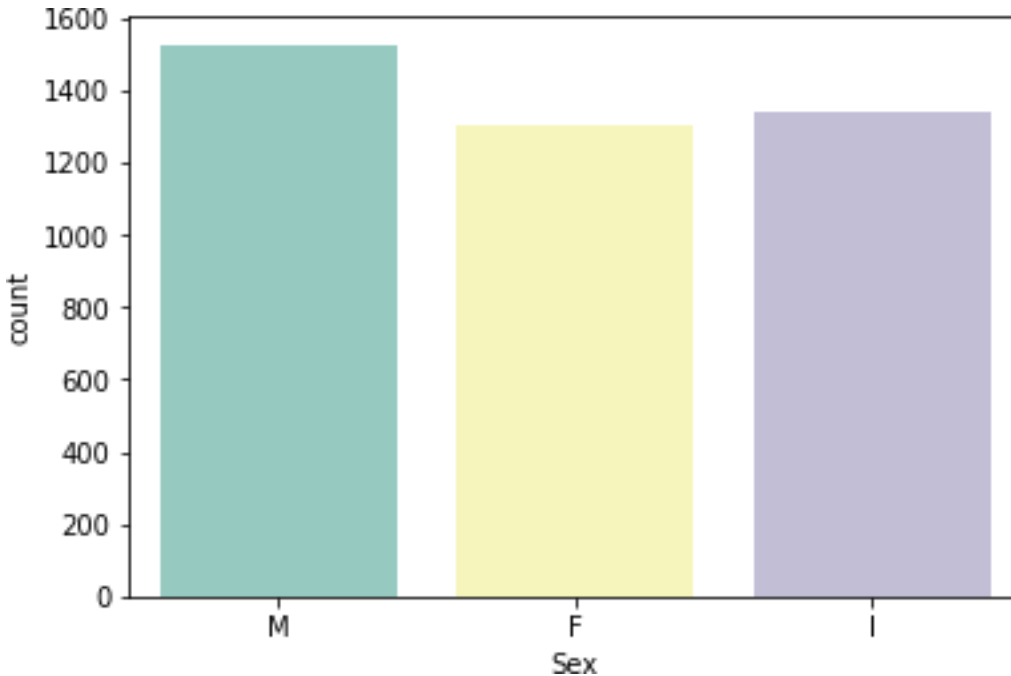
Out [ ]:



```
sns.countplot(x = 'Sex', data = df, palette = 'Set3')
```

In [ ]:

Out[ ]:



In [ ]:

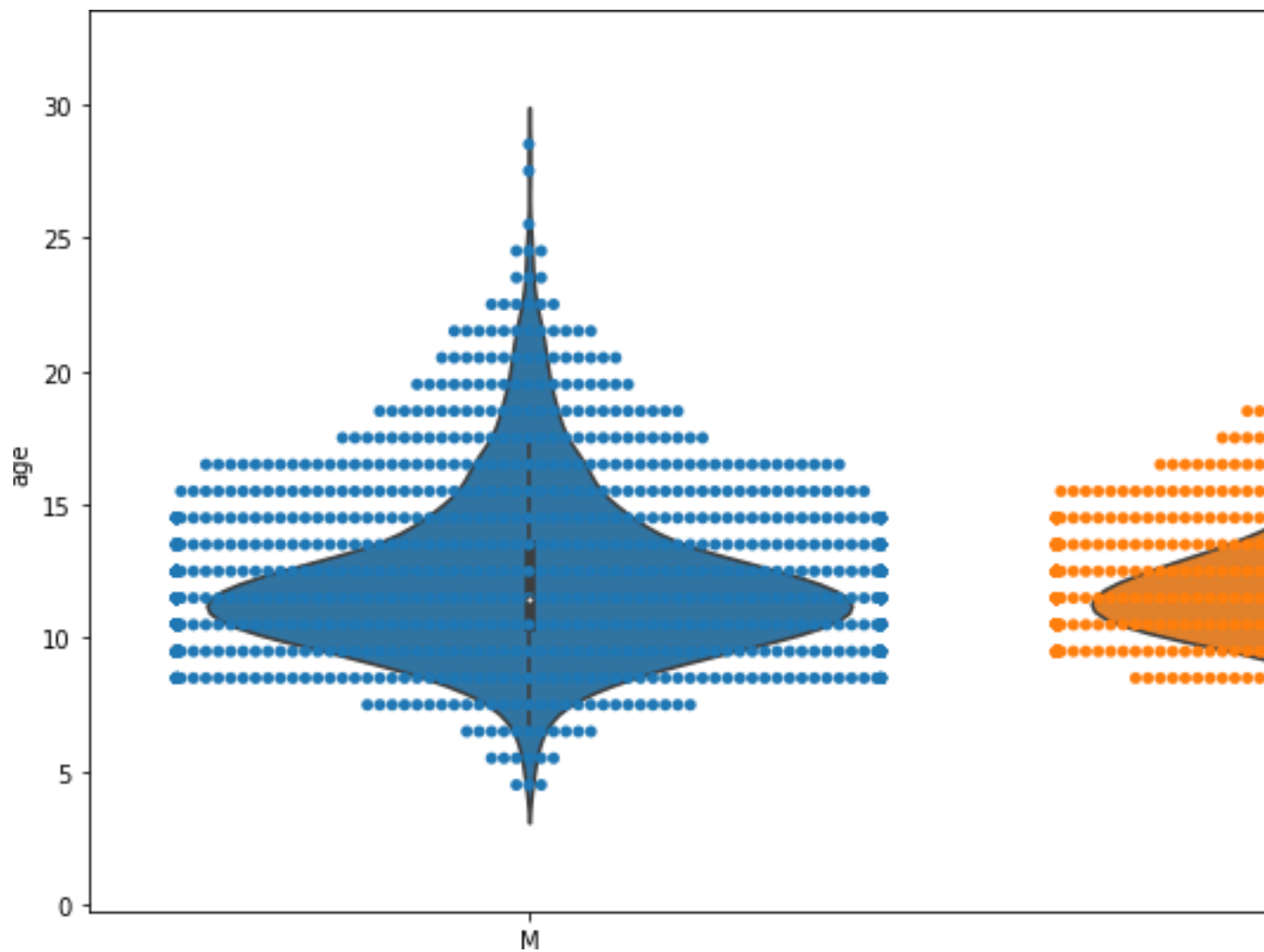
```
plt.figure(figsize = (20,7))
sns.swarmplot(x = 'Sex', y = 'age', data = df, hue = 'Sex')
sns.violinplot(x = 'Sex', y = 'age', data = df)

/usr/local/lib/python3.7/dist-packages/seaborn/categorical.py:1296: UserWarning:
  56.2% of the points cannot be placed; you may want to decrease the size of
  the markers or use stripplot.
  warnings.warn(msg, UserWarning)

/usr/local/lib/python3.7/dist-packages/seaborn/categorical.py:1296: UserWarning:
  52.2% of the points cannot be placed; you may want to decrease the size of
  the markers or use stripplot.
  warnings.warn(msg, UserWarning)

/usr/local/lib/python3.7/dist-packages/seaborn/categorical.py:1296: UserWarning:
  58.5% of the points cannot be placed; you may want to decrease the size of
  the markers or use stripplot.
  warnings.warn(msg, UserWarning)
```

Out[ ]:

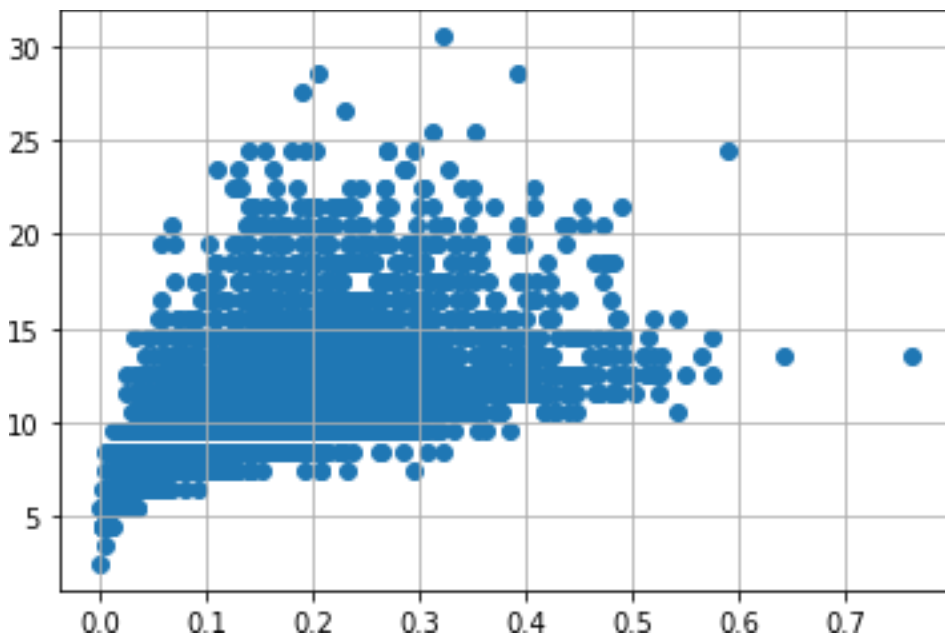


```
df = pd.get_dummies(df)
dummy_df = df
```

In [ ]:

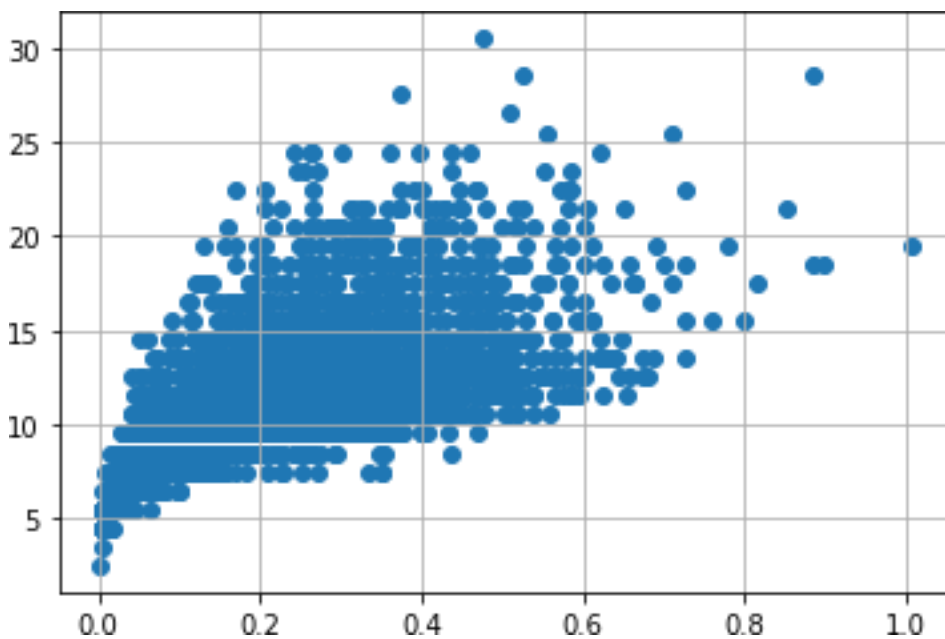
```
#viscera weight
var = 'Viscera weight'
plt.scatter(x = df[var], y = df['age'])
plt.grid(True)
```

In [ ]:



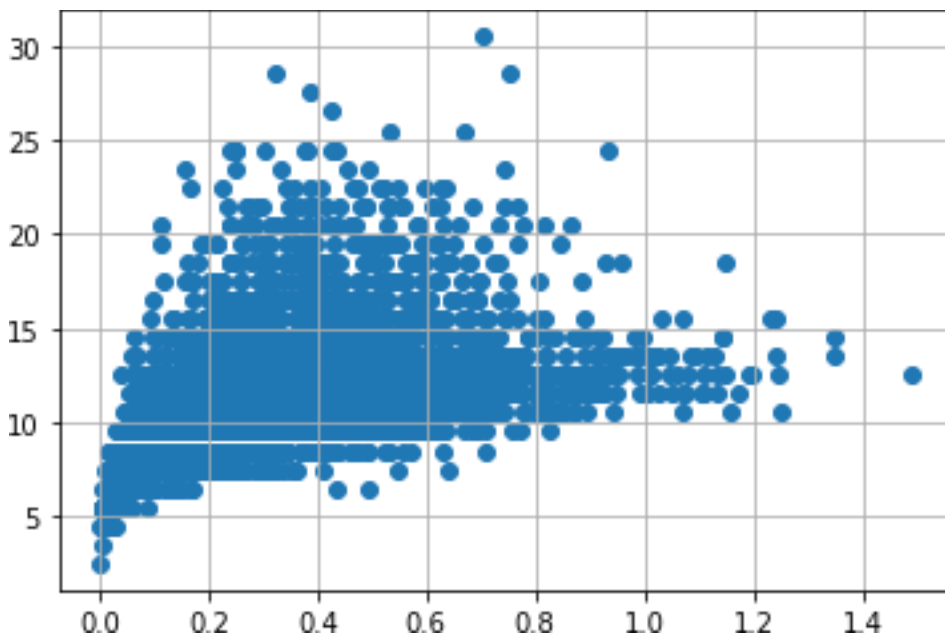
In [ ]:

```
#shell weight
var = 'Shell weight'
plt.scatter(x = df[var], y = df['age'])
plt.grid(True)
```



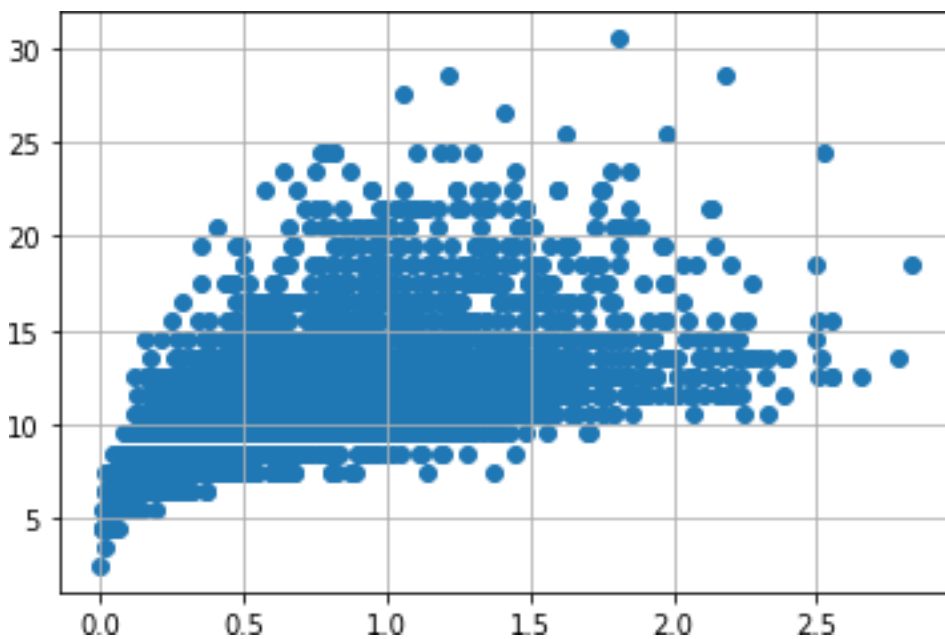
In [ ]:

```
#shucked weight
var = 'Shucked weight'
plt.scatter(x = df[var], y = df['age'])
plt.grid(True)
```



In [ ]:

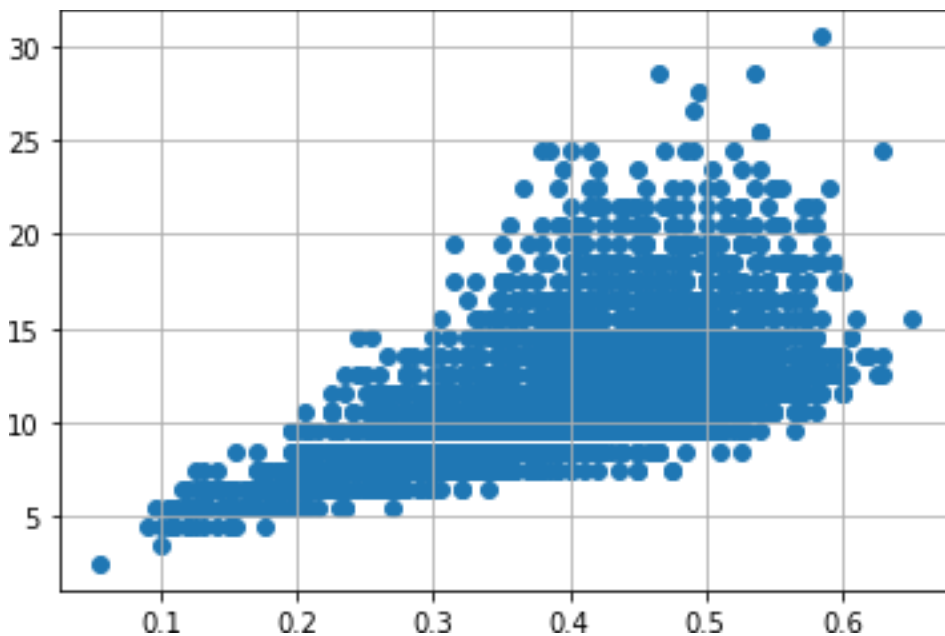
```
#whole weight
var = 'Whole weight'
plt.scatter(x = df[var], y = df['age'])
plt.grid(True)
```



In [ ]:

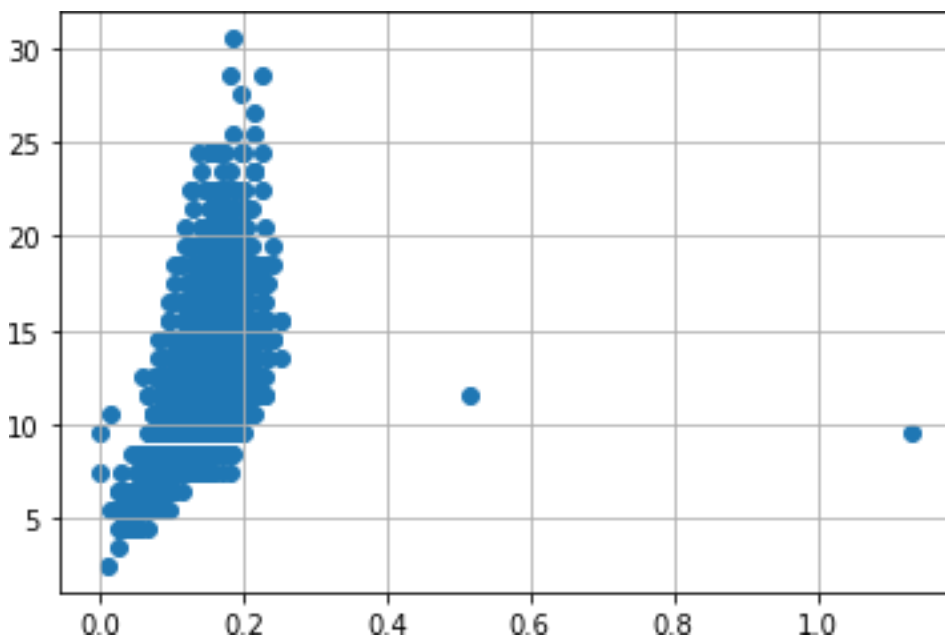
```
#diameter
var = 'Diameter'
plt.scatter(x = df[var], y = df['age'])
plt.grid(True)
```





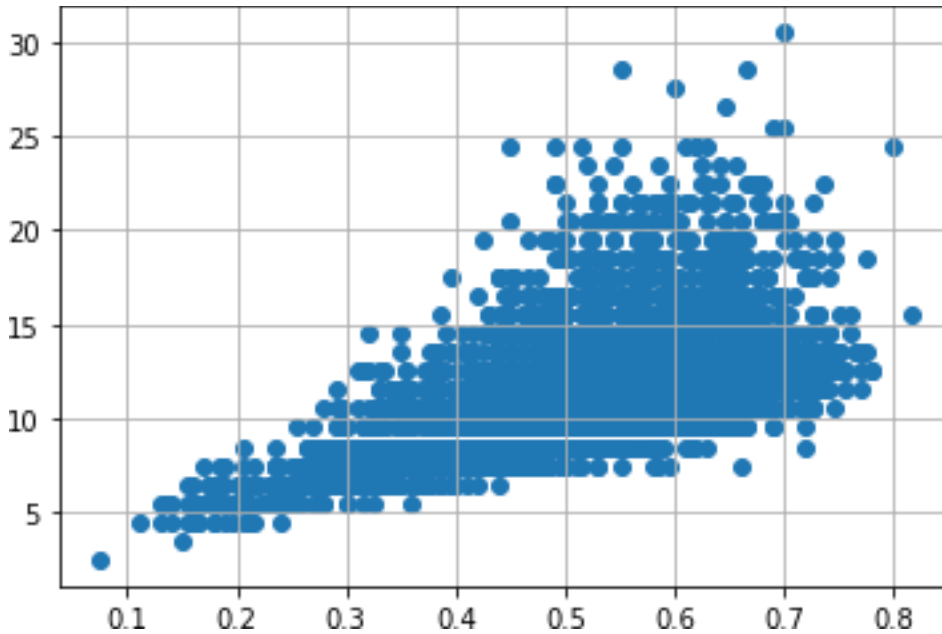
In [ ]:

```
#height
var = 'Height'
plt.scatter(x = df[var], y = df['age'])
plt.grid(True)
```



In [ ]:

```
#length
var = 'Length'
plt.scatter(x = df[var], y = df['age'])
plt.grid(True)
```



*#independent scale*

```
x=df.iloc[:, 0:1]
x
```

In [ ]:

In [ ]:

Out[ ]:

	Length
0	0.455
1	0.350
2	0.530
3	0.440
4	0.330
...	...
4172	0.565
4173	0.590

**Length****4174** 0.600**4175** 0.625**4176** 0.710

4177 rows × 1 columns

In [ ]:

#dependent variable

In [ ]:

y=df.iloc[:,1:]

y

Out[ ]:

	Diameter	Height	Whole weight	Shucked weight	Viscera weight	Shell weight	age	Sex_F	Sex_I	Sex_M
<b>0</b>	0.365	0.095	0.5140	0.2245	0.1010	0.1500	16.5	0	0	1
<b>1</b>	0.265	0.090	0.2255	0.0995	0.0485	0.0700	8.5	0	0	1
<b>2</b>	0.420	0.135	0.6770	0.2565	0.1415	0.2100	10.5	1	0	0
<b>3</b>	0.365	0.125	0.5160	0.2155	0.1140	0.1550	11.5	0	0	1
<b>4</b>	0.255	0.080	0.2050	0.0895	0.0395	0.0550	8.5	0	1	0
...	...	...	...	...	...	...	...	...	...	...
<b>4172</b>	0.450	0.165	0.8870	0.3700	0.2390	0.2490	12.5	1	0	0
<b>4173</b>	0.440	0.135	0.9660	0.4390	0.2145	0.2605	11.5	0	0	1
<b>4174</b>	0.475	0.205	1.1760	0.5255	0.2875	0.3080	10.5	0	0	1
<b>4175</b>	0.485	0.150	1.0945	0.5310	0.2610	0.2960	11.5	1	0	0

	Diameter	Height	Whole weight	Shucked weight	Viscera weight	Shell weight	age	Sex_F	Sex_I	Sex_M
4176	0.555	0.195	1.9485	0.9455	0.3765	0.4950	13.5	0	0	1

4177 rows × 10 columns

In [ ]:

```
#Train the Model, Test the Model, split
from sklearn.model_selection import train_test_split
x_train,x_test,y_train,y_test=train_test_split(x,y,test_size=0.2,random_state=0)
```

In [ ]:

```
x_train.shape
```

Out[ ]:

```
(3341, 1)
```

In [ ]:

```
x_test.shape
```

Out[ ]:

```
(836, 1)
```

In [ ]:

```
x_test
```

Out[ ]:

	Length
668	0.550
1580	0.500
3784	0.620
463	0.220
2615	0.645
...	...
575	0.610
3231	0.410

**Length****1084** 0.445**290** 0.540**2713** 0.250

836 rows × 1 columns

y\_train.shape

(3341, 10)

y\_test.shape

(836, 10)

model of building

# Build the Model

**from** sklearn.linear\_model **import** linearRegression

slr=LinearRegression()

slr.fit(x\_train,y\_train)

x\_test

In [ ]:

Out[ ]:

In [ ]:

Out[ ]:

In [ ]:

In [45]:

Out[45]:

**Length****668** 0.550**1580** 0.500**3784** 0.620**463** 0.220**2615** 0.645

... ...

**Length****575** 0.610**3231** 0.410**1084** 0.445**290** 0.540**2713** 0.250

836 rows × 1 columns

In [46]:

y\_test

Out[46]:

	<b>Diameter</b>	<b>Height</b>	<b>Whole weight</b>	<b>Shucked weight</b>	<b>Viscera weight</b>	<b>Shell weight</b>	<b>age</b>	<b>Sex_F</b>	<b>Sex_I</b>	<b>Sex_M</b>
<b>668</b>	0.425	0.155	0.9175	0.2775	0.2430	0.3350	14.5	0	0	1
<b>1580</b>	0.400	0.120	0.6160	0.2610	0.1430	0.1935	9.5	0	1	0
<b>3784</b>	0.480	0.155	1.2555	0.5270	0.3740	0.3175	12.5	0	0	1
<b>463</b>	0.165	0.055	0.0545	0.0215	0.0120	0.0200	6.5	0	1	0
<b>2615</b>	0.500	0.175	1.5105	0.6735	0.3755	0.3775	13.5	0	0	1
<b>...</b>	...	...	...	...	...	...	...	...	...	...
<b>575</b>	0.475	0.140	1.1330	0.5275	0.2355	0.3500	12.5	1	0	0
<b>3231</b>	0.325	0.120	0.3745	0.1580	0.0810	0.1250	13.5	0	0	1
<b>1084</b>	0.345	0.105	0.4090	0.1675	0.1015	0.1170	8.5	0	1	0

	Diameter	Height	Whole weight	Shucked weight	Viscera weight	Shell weight	age	Sex_F	Sex_I	Sex_M
<b>290</b>	0.435	0.180	0.9960	0.3835	0.2260	0.3250	18.5	0	0	1
<b>2713</b>	0.175	0.060	0.0635	0.0275	0.0080	0.0200	5.5	0	1	0

836 rows × 10 columns

In [49]:

```
#descriptive statistics
df.describe()
```

Out[49]:

	Length	Diameter	Height	Whole weight	Shucked weight	Viscera weight	Shell weight	age	Sex_F	Sex_I	Sex_M
<b>count</b>	4177.00000	4177.00000	4177.00000	4177.00000	4177.00000	4177.00000	4177.00000	4177.00000	4177.00000	4177.00000	4177.00000
<b>mean</b>	0.523992	0.407881	0.139516	0.828742	0.359367	0.180594	0.238831	11.433684	0.312904	0.321283	0.365813
<b>std</b>	0.120093	0.099240	0.041827	0.490389	0.221963	0.109614	0.139203	3.224169	0.463731	0.467025	0.481715
<b>min</b>	0.075000	0.055000	0.000000	0.002000	0.001000	0.000500	0.001500	2.500000	0.000000	0.000000	0.000000
<b>25%</b>	0.450000	0.350000	0.115000	0.441500	0.186000	0.093500	0.130000	9.500000	0.000000	0.000000	0.000000
<b>50%</b>	0.545000	0.425000	0.140000	0.799500	0.336000	0.171000	0.234000	10.500000	0.000000	0.000000	0.000000
<b>75%</b>	0.615000	0.480000	0.165000	1.153000	0.502000	0.253000	0.329000	12.500000	1.000000	1.000000	1.000000
<b>max</b>	0.815000	0.650000	1.130000	2.825500	1.488000	0.760000	1.005000	30.500000	1.000000	1.000000	1.000000

In [52]:

```
#multiple
```

```
df=pd.read_csv('/content/abalone.csv')
df.head()
```

Out[52]:

	Sex	Length	Diameter	Height	Whole weight	Shucked weight	Viscera weight	Shell weight	Rings
0	M	0.455	0.365	0.095	0.5140	0.2245	0.1010	0.150	15
1	M	0.350	0.265	0.090	0.2255	0.0995	0.0485	0.070	7
2	F	0.530	0.420	0.135	0.6770	0.2565	0.1415	0.210	9
3	M	0.440	0.365	0.125	0.5160	0.2155	0.1140	0.155	10
4	I	0.330	0.255	0.080	0.2050	0.0895	0.0395	0.055	7

```
df.tail()
```

In [53]:

Out[53]:

	Sex	Length	Diameter	Height	Whole weight	Shucked weight	Viscera weight	Shell weight	Rings
4172	F	0.565	0.450	0.165	0.8870	0.3700	0.2390	0.2490	11
4173	M	0.590	0.440	0.135	0.9660	0.4390	0.2145	0.2605	10
4174	M	0.600	0.475	0.205	1.1760	0.5255	0.2875	0.3080	9
4175	F	0.625	0.485	0.150	1.0945	0.5310	0.2610	0.2960	10
4176	M	0.710	0.555	0.195	1.9485	0.9455	0.3765	0.4950	12