In [ ]: Name: Akash Varade Roll No: A-04 In [2]: import seaborn as sns import pandas as pd titanic = sns.load\_dataset("titanic") titanic Out[2]: survived pclass sibsp parch fare embarked class who sex age 0 0 3 male 22.0 1 0 7.2500 S Third man 1 1 female 38.0 0 71.2833 C First woman 2 1 3 female 26.0 0 0 7.9250 S Third woman 3 1 female 35.0 0 53.1000 S First woman 4 0 3 male 35.0 0 0 8.0500 S Third man 886 0 2 male 27.0 0 13.0000 S Second man 887 female 19.0 0 30.0000 S First woman 888 0 3 female NaN 1 2 23.4500 S Third woman 889 1 26.0 0 0 30.0000 C First 1 male man 890 0 3 male 32.0 0 0 7.7500 Q Third man 891 rows × 15 columns In [3]: titanic.info() <class 'pandas.core.frame.DataFrame'> RangeIndex: 891 entries, 0 to 890 Data columns (total 15 columns): # Column Non-Null Count Dtype --int64 0 survived 891 non-null 1 891 non-null int64 pclass 2 sex 891 non-null object float64 714 non-null 3 age 4 891 non-null int64 sibsp 5 891 non-null int64 parch 6 fare 891 non-null float64 7 889 non-null object embarked 8 891 non-null class category 9 891 non-null object who 10 adult\_male 891 non-null bool 11 deck 203 non-null category 12 embark\_town 889 non-null object 13 alive 891 non-null object 14 alone 891 non-null bool

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

memory usage: 80.6+ KB

```
In [4]: x=titanic["fare"]
Out[4]:
         0
                 7.2500
         1
                71.2833
         2
                 7.9250
         3
                53.1000
         4
                 8.0500
                 . . .
         886
                13.0000
         887
                30.0000
         888
                23.4500
         889
                30.0000
         890
                 7.7500
         Name: fare, Length: 891, dtype: float64
In [5]: titanic.describe()
Out[5]:
                  survived
                                pclass
                                             age
                                                        sibsp
                                                                   parch
                                                                                fare
         count 891.000000 891.000000
                                       714.000000
                                                  891.000000
                                                              891.000000
                                                                          891.000000
                  0.383838
                             2.308642
                                        29.699118
                                                     0.523008
                                                                0.381594
                                                                           32.204208
         mean
                  0.486592
                             0.836071
                                        14.526497
                                                     1.102743
                                                                0.806057
                                                                           49.693429
           std
                  0.000000
                             1.000000
                                         0.420000
                                                     0.000000
                                                                0.000000
                                                                            0.000000
          min
          25%
                  0.000000
                             2.000000
                                        20.125000
                                                     0.000000
                                                                0.000000
                                                                            7.910400
          50%
                  0.000000
                                        28.000000
                                                     0.000000
                             3.000000
                                                                0.000000
                                                                           14.454200
          75%
                                        38.000000
                  1.000000
                             3.000000
                                                     1.000000
                                                                0.000000
                                                                           31.000000
                  1.000000
                             3.000000
                                        80.000000
                                                     8.000000
                                                                6.000000
                                                                          512.329200
          max
In [6]: titanic.info()
       <class 'pandas.core.frame.DataFrame'>
       RangeIndex: 891 entries, 0 to 890
       Data columns (total 15 columns):
        #
            Column
                          Non-Null Count Dtype
            ----
                          -----
            survived
                                           int64
        0
                          891 non-null
                                           int64
        1
            pclass
                          891 non-null
        2
                          891 non-null
                                          object
            sex
        3
                          714 non-null
                                          float64
            age
        4
            sibsp
                          891 non-null
                                          int64
        5
            parch
                          891 non-null
                                          int64
        6
            fare
                          891 non-null
                                           float64
        7
            embarked
                          889 non-null
                                          object
        8
            class
                          891 non-null
                                           category
        9
                          891 non-null
            who
                                           object
        10
            adult_male
                          891 non-null
                                           bool
        11
            deck
                          203 non-null
                                           category
        12
            embark_town 889 non-null
                                           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: 80.6+ KB
```

Out[7]:		survived	sex	age	sibsp	parch	fare	class	who	adult_male	alive	а
	0	0	male	22.0	1	0	7.2500	Third	man	True	no	
	1	1	female	38.0	1	0	71.2833	First	woman	False	yes	
	2	1	female	26.0	0	0	7.9250	Third	woman	False	yes	
	3	1	female	35.0	1	0	53.1000	First	woman	False	yes	
	4	0	male	35.0	0	0	8.0500	Third	man	True	no	
	5	0	male	NaN	0	0	8.4583	Third	man	True	no	
	6	0	male	54.0	0	0	51.8625	First	man	True	no	
	7	0	male	2.0	3	1	21.0750	Third	child	False	no	
	8	1	female	27.0	0	2	11.1333	Third	woman	False	yes	
	9	1	female	14.0	1	0	30.0708	Second	child	False	yes	
	10	1	female	4.0	1	1	16.7000	Third	child	False	yes	
	11	1	female	58.0	0	0	26.5500	First	woman	False	yes	
	12	0	male	20.0	0	0	8.0500	Third	man	True	no	
	13	0	male	39.0	1	5	31.2750	Third	man	True	no	
	14	0	female	14.0	0	0	7.8542	Third	child	False	no	
	4											•

In [8]: titanic\_cleaned.info()

<class 'pandas.core.frame.DataFrame'>
RangeIndex: 891 entries, 0 to 890
Data columns (total 11 columns):

#	Column	Non-Null Count	Dtype
0	survived	891 non-null	int64
1	sex	891 non-null	object
2	age	714 non-null	float64
3	sibsp	891 non-null	int64
4	parch	891 non-null	int64
5	fare	891 non-null	float64
6	class	891 non-null	category
7	who	891 non-null	object
8	adult_male	891 non-null	bool
9	alive	891 non-null	object
10	alone	891 non-null	bool

dtypes: bool(2), category(1), float64(2), int64(3), object(3)
memory usage: 58.5+ KB

In [9]: titanic\_cleaned.isnull().sum()

survived	0
sex	0
age	177
sibsp	0
parch	0
fare	0
class	0
who	0
adult_male	0
alive	0
alone	0
dtype: int64	
	sex age sibsp parch fare class who adult_male alive alone

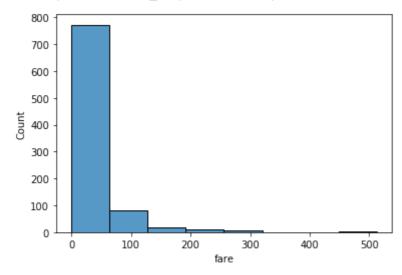
In [10]: titanic\_cleaned.corr(method='pearson')

_		-	_	-
()1	11	11	a	
$\circ$	и С	1 -	$\circ$	

	survived	age	sibsp	parch	fare	adult_male	alone
survived	1.000000	-0.077221	-0.035322	0.081629	0.257307	-0.557080	-0.203367
age	-0.077221	1.000000	-0.308247	-0.189119	0.096067	0.280328	0.198270
sibsp	-0.035322	-0.308247	1.000000	0.414838	0.159651	-0.253586	-0.584471
parch	0.081629	-0.189119	0.414838	1.000000	0.216225	-0.349943	-0.583398
fare	0.257307	0.096067	0.159651	0.216225	1.000000	-0.182024	-0.271832
adult_male	-0.557080	0.280328	-0.253586	-0.349943	-0.182024	1.000000	0.404744
alone	-0.203367	0.198270	-0.584471	-0.583398	-0.271832	0.404744	1.000000

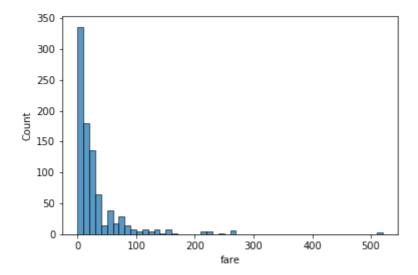
In [12]: sns.histplot(data=titanic,x="fare",bins=8)

Out[12]: <matplotlib.axes.\_subplots.AxesSubplot at 0x7effc7c26990>



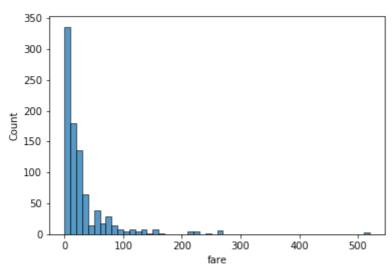
In [13]: sns.histplot(data=titanic,x="fare",binwidth=10)

Out[13]: <matplotlib.axes.\_subplots.AxesSubplot at 0x7effc622aa10>



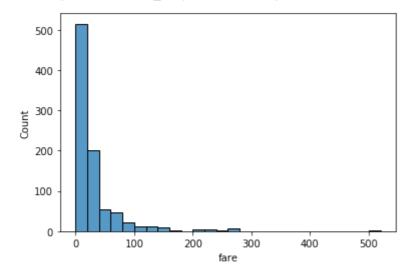
In [14]: sns.histplot(data=titanic,x="fare",bins=20,binwidth=10)

Out[14]: <matplotlib.axes.\_subplots.AxesSubplot at 0x7effc6176b50>



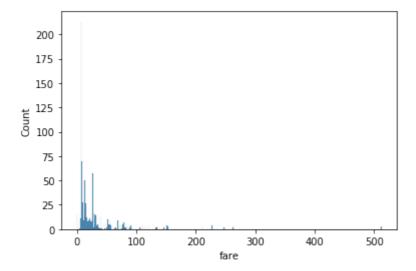
In [15]: sns.histplot(data=titanic,x="fare",binwidth=20)

Out[15]: <matplotlib.axes.\_subplots.AxesSubplot at 0x7effc607a9d0>



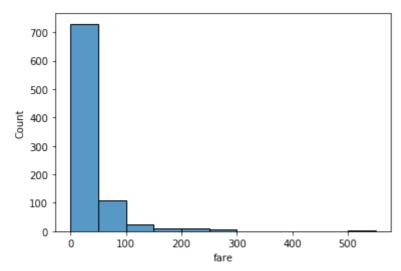
In [16]: sns.histplot(data=titanic,x="fare",binwidth=1)

Out[16]: <matplotlib.axes.\_subplots.AxesSubplot at 0x7effc5f6c490>



In [17]: sns.histplot(data=titanic,x="fare", bins=20,binwidth=50)

Out[17]: <matplotlib.axes.\_subplots.AxesSubplot at 0x7effc596e3d0>



In [ ]: