

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

from pathlib import Path
from scipy.stats import trim_mean

anime = pd.read_csv(r"C:\Users\HP\Downloads\Stats project\anime.csv")
print(anime.head(8))
```

	anime_id	name \
0	32281	Kimi no Na wa.
1	5114	Fullmetal Alchemist: Brotherhood
2	28977	Gintama°
3	9253	Steins;Gate
4	9969	Gintama'
5	32935	Haikyuu!!: Karasuno Koukou VS Shiratorizawa Ga...
6	11061	Hunter x Hunter (2011)
7	820	Ginga Eiyuu Densetsu

	rating \	genre	type	episodes
0	9.37	Drama, Romance, School, Supernatural	Movie	1
1	9.26	Action, Adventure, Drama, Fantasy, Magic, Mili...	TV	64
2	9.25	Action, Comedy, Historical, Parody, Samurai, S...	TV	51
3	9.17	Sci-Fi, Thriller	TV	24
4	9.16	Action, Comedy, Historical, Parody, Samurai, S...	TV	51
5	9.15	Comedy, Drama, School, Shounen, Sports	TV	10
6	9.13	Action, Adventure, Shounen, Super Power	TV	148
7	9.11	Drama, Military, Sci-Fi, Space	OVA	110

	members
0	200630
1	793665
2	114262
3	673572
4	151266

```

5      93351
6      425855
7      80679

```

anime

	anime_id		name \
0	32281		Kimi no Na wa.
1	5114		Fullmetal Alchemist: Brotherhood
2	28977		Gintama°
3	9253		Steins;Gate
4	9969		Gintama'
...
12289	9316	Toushindai My Lover: Minami tai Mecha-Minami	
12290	5543		Under World
12291	5621		Violence Gekiga David no Hoshi
12292	6133	Violence Gekiga Shin David no Hoshi: Inma Dens...	
12293	26081		Yasuji no Pornorama: Yacchimaee!!

		genre	type
episodes \			
0		Drama, Romance, School, Supernatural	Movie
1			
1	Action, Adventure, Drama, Fantasy, Magic, Mili...		TV
64			
2	Action, Comedy, Historical, Parody, Samurai, S...		TV
51			
3		Sci-Fi, Thriller	TV
24			
4	Action, Comedy, Historical, Parody, Samurai, S...		TV
51			
...	
..			
12289		Hentai	OVA
1			
12290		Hentai	OVA
1			
12291		Hentai	OVA
4			
12292		Hentai	OVA
1			
12293		Hentai	Movie
1			

	rating	members
0	9.37	200630
1	9.26	793665
2	9.25	114262
3	9.17	673572
4	9.16	151266

```

...
12289    4.15    211
12290    4.28    183
12291    4.88    219
12292    4.98    175
12293    5.46    142

```

```
[12294 rows x 7 columns]
```

Estimates of Location

```

print("Mean rating =")
print(anime['rating'].mean())

```

```

Mean rating =
6.473901690981445

```

```

print("median of rating =")
print(anime['rating'].median())

```

```

median of rating =
6.57

```

```

print("mode of episodes =")
print(anime['episodes'].mode())

```

```

mode of episodes =
0    1
Name: episodes, dtype: object

```

```
print(anime.head(10))
```

	anime_id	name \
0	32281	Kimi no Na wa.
1	5114	Fullmetal Alchemist: Brotherhood
2	28977	Gintama°
3	9253	Steins;Gate
4	9969	Gintama'
5	32935	Haikyuu!!: Karasuno Koukou VS Shiratorizawa Ga...
6	11061	Hunter x Hunter (2011)
7	820	Ginga Eiyuu Densetsu
8	15335	Gintama Movie: Kanketsu-hen - Yorozuya yo Eien...
9	15417	Gintama';: Enchousen

	genre	type	episodes
rating \			
0	Drama, Romance, School, Supernatural	Movie	1
9.37			
1	Action, Adventure, Drama, Fantasy, Magic, Mili...	TV	64

```

9.26
2 Action, Comedy, Historical, Parody, Samurai, S... TV 51
9.25
3 Sci-Fi, Thriller TV 24
9.17
4 Action, Comedy, Historical, Parody, Samurai, S... TV 51
9.16
5 Comedy, Drama, School, Shounen, Sports TV 10
9.15
6 Action, Adventure, Shounen, Super Power TV 148
9.13
7 Drama, Military, Sci-Fi, Space OVA 110
9.11
8 Action, Comedy, Historical, Parody, Samurai, S... Movie 1
9.10
9 Action, Comedy, Historical, Parody, Samurai, S... TV 13
9.11

```

```

members
0 200630
1 793665
2 114262
3 673572
4 151266
5 93351
6 425855
7 80679
8 72534
9 81109

print(trim_mean(anime['members'],0.1))

5589.94286295242

sample_anime = anime.head(100)

print(np.average(sample_anime['rating'],
weights=sample_anime['members']))

8.754444415324095

```

Estimates of Variability

```

print(anime['members'].std())

54820.676924907515

#Interquartile range is calculated as the difference of the 75% and 25% quantile.

```

```

print(anime['members'].quantile(0.75) -
      anime['members'].quantile(0.25))

9212.0

#Median absolute deviation from the median
#method- 1
#print(abs(anime['members'] - anime['members'].median()).median() /
#       0.6744897501960817)
#method-2
print(robust.scale.mad(anime['members']))

2172.0122501107066

```

Estimates on Percentiles

```

print(anime['members'].quantile([0.05, 0.25, 0.5, 0.75, 0.95]))

0.05      58.0
0.25     225.0
0.50    1550.0
0.75    9437.0
0.95   93164.3
Name: members, dtype: float64

percentages = [0.05, 0.25, 0.5, 0.75, 0.95]
df = pd.DataFrame(anime['members'].quantile(percentages))
df.index = [f'{p * 100}%' for p in percentages]
print(df.transpose())

```

	5.0%	25.0%	50.0%	75.0%	95.0%
members	58.0	225.0	1550.0	9437.0	93164.3

Explore data distribution

Frequency table

```

animax = anime.head(2000)
binnedmembers = pd.cut(animax['members'], 20)
print(binnedmembers.value_counts())

```

(-644.548, 51046.4]	1190
(51046.4, 101723.8]	356
(101723.8, 152401.2]	156
(152401.2, 203078.6]	96
(203078.6, 253756.0]	66
(253756.0, 304433.4]	37

(304433.4, 355110.8]	33
(355110.8, 405788.2]	16
(405788.2, 456465.6]	11
(456465.6, 507143.0]	10
(557820.4, 608497.8]	8
(507143.0, 557820.4]	8
(608497.8, 659175.2]	5
(659175.2, 709852.6]	2
(709852.6, 760530.0]	2
(861884.8, 912562.2]	2
(760530.0, 811207.4]	1
(963239.6, 1013917.0]	1
(811207.4, 861884.8]	0
(912562.2, 963239.6]	0

Name: members, dtype: int64

```

binnedmembers.name = 'binnedmembers'
df = pd.concat([animax, binnedmembers], axis=1)
df = df.sort_values(by='members')

```

```

groups = []
for group, subset in df.groupby(by='binnedmembers'):
    groups.append({
        'BinRange': group,
        'Count': len(subset),
        'Genre': ','.join(subset.genre)
    })
print(pd.DataFrame(groups))

```

	BinRange	Count	\
0	(-644.548, 51046.4]	1190	
1	(51046.4, 101723.8]	356	
2	(101723.8, 152401.2]	156	
3	(152401.2, 203078.6]	96	
4	(203078.6, 253756.0]	66	
5	(253756.0, 304433.4]	37	
6	(304433.4, 355110.8]	33	
7	(355110.8, 405788.2]	16	
8	(405788.2, 456465.6]	11	
9	(456465.6, 507143.0]	10	
10	(507143.0, 557820.4]	8	
11	(557820.4, 608497.8]	8	
12	(608497.8, 659175.2]	5	
13	(659175.2, 709852.6]	2	
14	(709852.6, 760530.0]	2	
15	(760530.0, 811207.4]	1	
16	(811207.4, 861884.8]	0	
17	(861884.8, 912562.2]	2	
18	(912562.2, 963239.6]	0	
19	(963239.6, 1013917.0]	1	

```

                                Genre
0  Action, Fantasy, Historical, Martial Arts,Dram...
1  Action, Adventure, Samurai,Romance, School, Sh...
2  Comedy, Drama, Mystery, Romance, Slice of Life...
3  Action, Adventure, Drama, Fantasy, Historical,...
4  Action, Seinen,Action, Fantasy, Magic, Romance...
5  Comedy, School, Slice of Life,Action, Adventur...
6  Action, Comedy, Dementia, Mecha, Parody, Sci-F...
7  Drama, Fantasy, Psychological, Thriller,Fantas...
8  Action, Drama, Horror, Mystery, Psychological,...
9  Drama, Fantasy, Romance, Slice of Life, Supern...
10 Action, Comedy, School, Super Power,Action, Po...
11 Action, Adventure, Comedy, Mecha, Sci-Fi,Comed...
12 Action, Drama, Horror, Mystery, Psychological,...
13 Sci-Fi, Thriller,Action, Comedy, Martial Arts,...
14 Action, Mecha, Military, School, Sci-Fi, Super...
15 Action, Adventure, Drama, Fantasy, Magic, Mili...
16
17 Action, Adventure, Fantasy, Game, Romance,Acti...
18
19 Mystery, Police, Psychological, Supernatural, ...

```

```
print(pd.DataFrame(groups))
```

```

      BinRange  Count  \
0  (-644.548, 51046.4]    1190
1  (51046.4, 101723.8]     356
2  (101723.8, 152401.2]    156
3  (152401.2, 203078.6]     96
4  (203078.6, 253756.0]     66
5  (253756.0, 304433.4]     37
6  (304433.4, 355110.8]     33
7  (355110.8, 405788.2]     16
8  (405788.2, 456465.6]     11
9  (456465.6, 507143.0]     10
10 (507143.0, 557820.4]      8
11 (557820.4, 608497.8]      8
12 (608497.8, 659175.2]      5
13 (659175.2, 709852.6]      2
14 (709852.6, 760530.0]      2
15 (760530.0, 811207.4]      1
16 (811207.4, 861884.8]      0
17 (861884.8, 912562.2]      2
18 (912562.2, 963239.6]      0
19 (963239.6, 1013917.0]     1

```

```

                                Genre
0  Action, Fantasy, Historical, Martial Arts,Dram...
1  Action, Adventure, Samurai,Romance, School, Sh...

```

```

2 Comedy, Drama, Mystery, Romance, Slice of Life...
3 Action, Adventure, Drama, Fantasy, Historical,...
4 Action, Seinen, Action, Fantasy, Magic, Romance...
5 Comedy, School, Slice of Life, Action, Adventur...
6 Action, Comedy, Dementia, Mecha, Parody, Sci-F...
7 Drama, Fantasy, Psychological, Thriller, Fantas...
8 Action, Drama, Horror, Mystery, Psychological,...
9 Drama, Fantasy, Romance, Slice of Life, Supern...
10 Action, Comedy, School, Super Power, Action, Po...
11 Action, Adventure, Comedy, Mecha, Sci-Fi, Comed...
12 Action, Drama, Horror, Mystery, Psychological,...
13 Sci-Fi, Thriller, Action, Comedy, Martial Arts,...
14 Action, Mecha, Military, School, Sci-Fi, Super...
15 Action, Adventure, Drama, Fantasy, Magic, Mili...
16
17 Action, Adventure, Fantasy, Game, Romance, Acti...
18
19 Mystery, Police, Psychological, Supernatural, ...

```

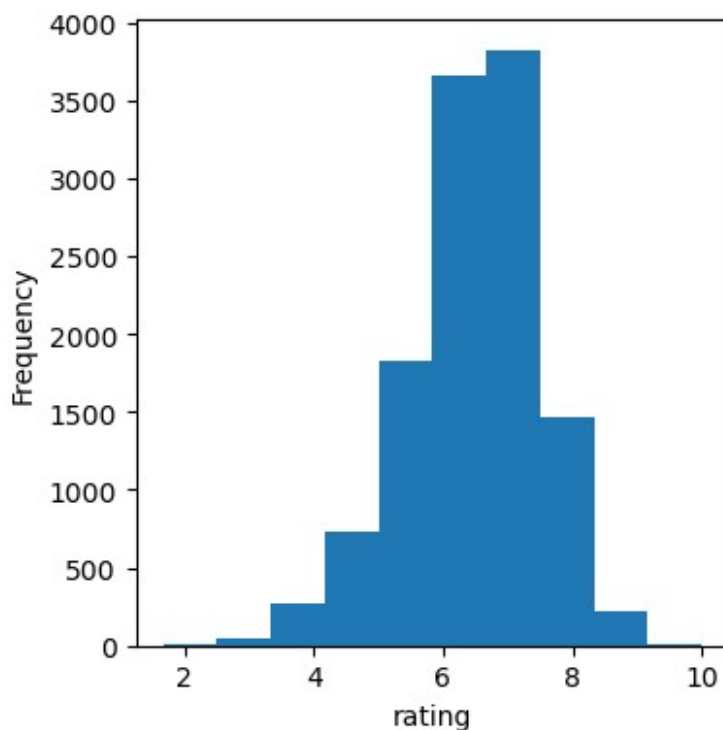
Histograms

```

ax = (anime['rating']).plot.hist(figsize=(4, 4))
ax.set_xlabel('rating')

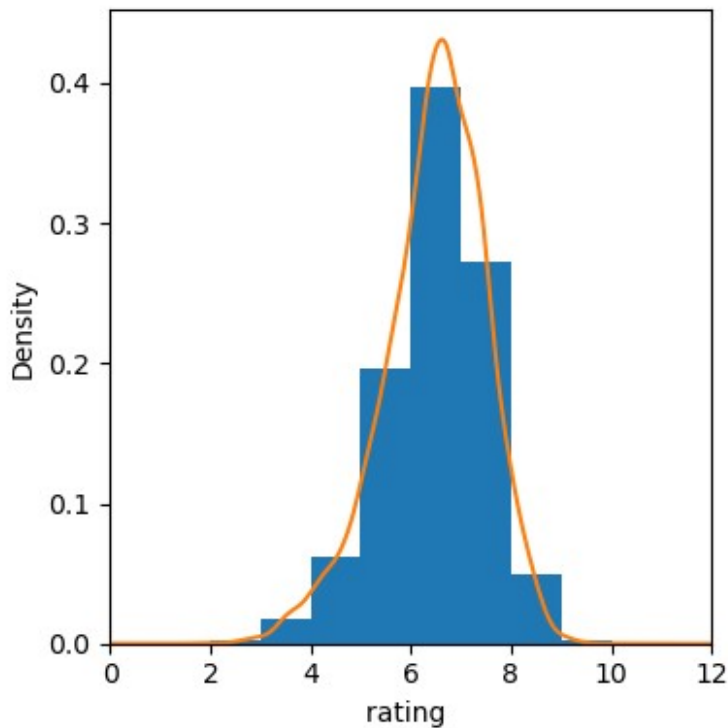
plt.tight_layout()
plt.show()

```



density plot

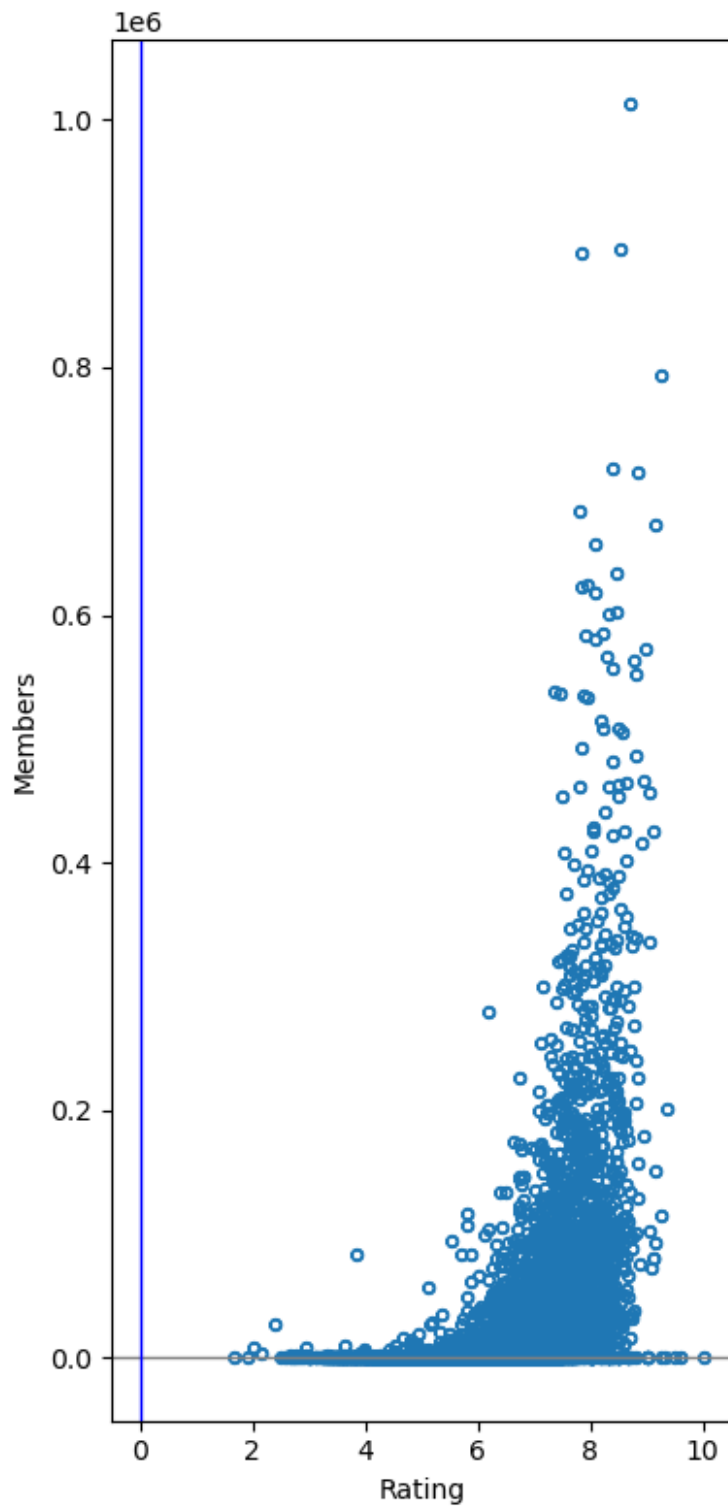
```
ax = anime['rating'].plot.hist(density=True, xlim=[0, 12],  
                               bins=range(1,12), figsize=(4, 4))  
anime['rating'].plot.density(ax=ax)  
ax.set_xlabel('rating ')  
  
plt.tight_layout()  
plt.show()
```



Exploring Binary and Categorical Data

Scatterplots

```
ax = anime.plot.scatter(x='rating', y='members', figsize=(4, 8),  
                        marker='$\u25EF$')  
ax.set_xlabel('Rating')  
ax.set_ylabel('Members')  
ax.axhline(0, color='grey', lw=1)  
ax.axvline(0, color='blue', lw=1)  
  
plt.tight_layout()  
plt.show()
```



Binning

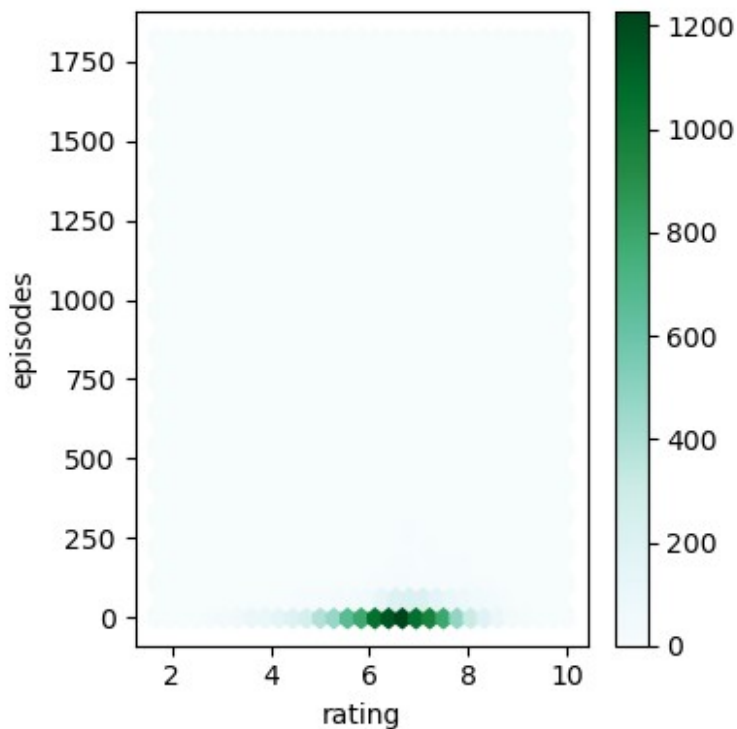
```
animes = anime.drop(anime[anime.episodes == 'Unknown'].index)
print(animes)
```

	anime_id		name \
0	32281		Kimi no Na wa.
1	5114		Fullmetal Alchemist: Brotherhood
2	28977		Gintama°
3	9253		Steins;Gate
4	9969		Gintama'
...
12289	9316	Toushindai My Lover: Minami tai Mecha-Minami	
12290	5543		Under World
12291	5621		Violence Gekiga David no Hoshi
12292	6133	Violence Gekiga Shin David no Hoshi: Inma Dens...	
12293	26081		Yasuji no Pornorama: Yacchimaee!!
		genre	type
episodes \			
0		Drama, Romance, School, Supernatural	Movie
1			
1		Action, Adventure, Drama, Fantasy, Magic, Mili...	TV
64			
2		Action, Comedy, Historical, Parody, Samurai, S...	TV
51			
3		Sci-Fi, Thriller	TV
24			
4		Action, Comedy, Historical, Parody, Samurai, S...	TV
51			
...	
..			
12289		Hentai	OVA
1			
12290		Hentai	OVA
1			
12291		Hentai	OVA
4			
12292		Hentai	OVA
1			
12293		Hentai	Movie
1			
	rating	members	
0	9.37	200630	
1	9.26	793665	
2	9.25	114262	
3	9.17	673572	
4	9.16	151266	
...	
12289	4.15	211	
12290	4.28	183	
12291	4.88	219	
12292	4.98	175	
12293	5.46	142	

```
[11954 rows x 7 columns]

animes = animes.astype({'episodes':'float'})
ax = animes.plot.hexbin(x='rating', y='episodes',
                        gridsize=30, sharex=False, figsize=(4, 4))
ax.set_xlabel('rating')
ax.set_ylabel('episodes')

plt.tight_layout()
plt.show()
```



```
animes['true_weight'] = animes['episodes'] * animes['members'] /
100000
animes.head(10)
```

	anime_id	name \
0	32281	Kimi no Na wa.
1	5114	Fullmetal Alchemist: Brotherhood
2	28977	Gintama°
3	9253	Steins;Gate
4	9969	Gintama'
5	32935	Haikyuu!!: Karasuno Koukou VS Shiratorizawa Ga...
6	11061	Hunter x Hunter (2011)
7	820	Ginga Eiyuu Densetsu
8	15335	Gintama Movie: Kanketsu-hen - Yorozuya yo Eien...

```

9      15417      Gintama&#039;;: Enchousen
                                genre    type  episodes
rating \
0      Drama, Romance, School, Supernatural  Movie      1.0
9.37
1  Action, Adventure, Drama, Fantasy, Magic, Mili...    TV      64.0
9.26
2  Action, Comedy, Historical, Parody, Samurai, S...    TV      51.0
9.25
3      Sci-Fi, Thriller    TV      24.0
9.17
4  Action, Comedy, Historical, Parody, Samurai, S...    TV      51.0
9.16
5      Comedy, Drama, School, Shounen, Sports    TV      10.0
9.15
6      Action, Adventure, Shounen, Super Power    TV      148.0
9.13
7      Drama, Military, Sci-Fi, Space    OVA      110.0
9.11
8  Action, Comedy, Historical, Parody, Samurai, S...  Movie      1.0
9.10
9  Action, Comedy, Historical, Parody, Samurai, S...    TV      13.0
9.11

```

```

    members  true_weight
0    200630      2.00630
1    793665     507.94560
2    114262      58.27362
3    673572     161.65728
4    151266      77.14566
5     93351       9.33510
6    425855     630.26540
7     80679      88.74690
8     72534       0.72534
9     81109     10.54417

```

```

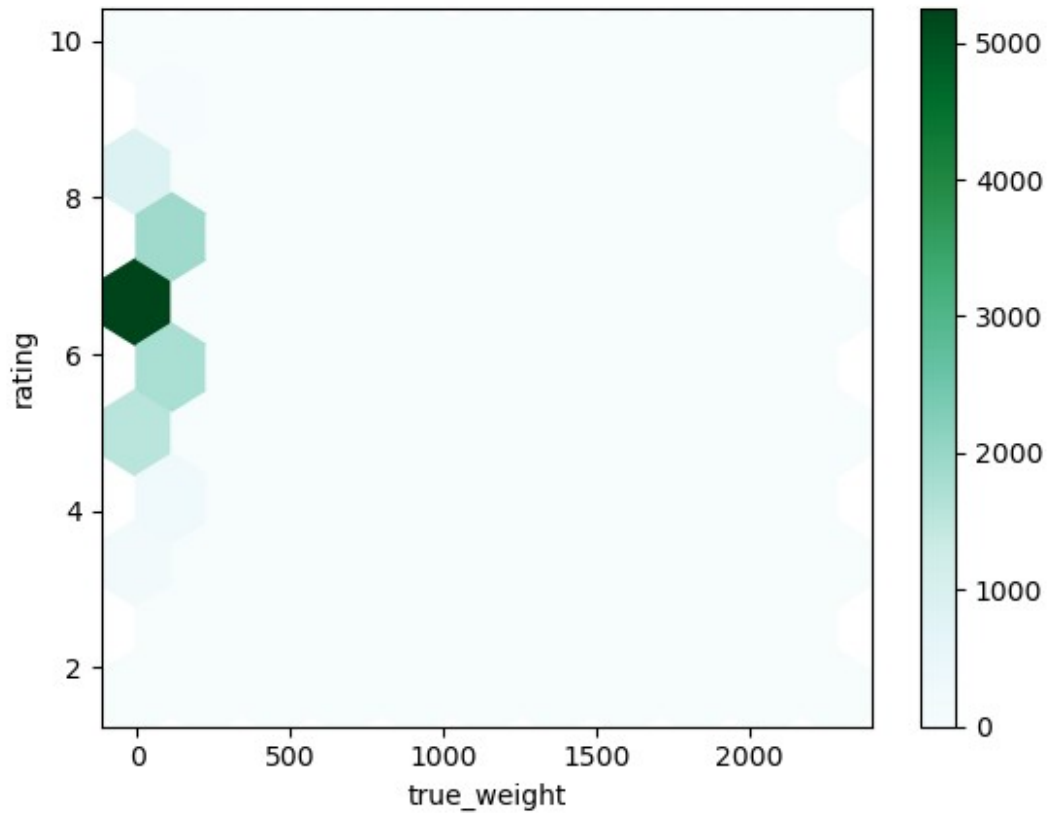
animes.plot(kind='hexbin',x='true_weight', y = 'rating' , gridsize =
10)

```

```

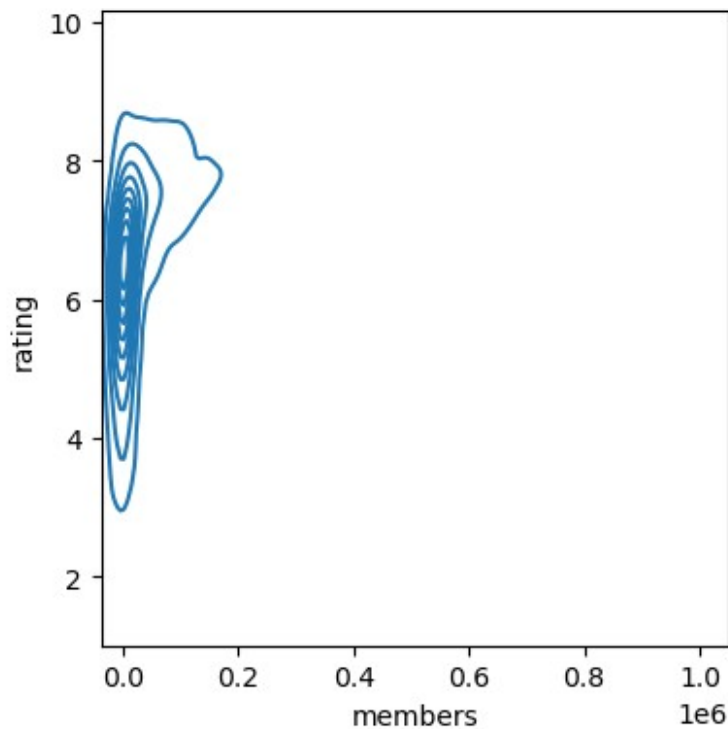
<AxesSubplot:xlabel='true_weight', ylabel='rating'>

```



```
fig, ax = plt.subplots(figsize=(4, 4))
sns.kdeplot(data=animes.sample(10000), x='members', y='rating', ax=ax)
ax.set_xlabel('members')
ax.set_ylabel('rating')

plt.tight_layout()
plt.show()
```



Violin plot

```
fig, ax = plt.subplots(figsize=(4, 4))
sns.violinplot(data=anime, x='episodes', y='members',
               ax=ax, inner='quartile', color='white')
ax.set_xlabel('rating')
ax.set_ylabel('true_weight')

plt.tight_layout()
plt.show()
```

