### seaborn

#### August 11, 2024

### 1 Seaborn

Seaborn is a Python data visualization library based on matplotlib. It provides a high-level interface for drawing attractive and informative statistical graphics.

- Provides a layer of abstraction hence simpler to use
- better aesthetics
- more graphs included

#### 1.0.1 Main Components of Seaborn

Types of Functions 1. Figure Level 2. Axis Level

Main Classification - Relational Plot - Distribution Plot - Categorical Plot - Regression Plot - Matrix Plot - Multiplots

```
[80]: import seaborn as sns
import numpy as np
import plotly.express as px
```

#### 1.1 1. Relational Plot

Generally used for - see the statistical relation between 2 or more variables. - Bivariate Analysis

```
Figure level - relplot()
```

Axis level - 1. scatterplot() 2. lineplot()

```
[2]: tips = sns.load_dataset('tips')
tips
```

```
[2]:
           total_bill
                         tip
                                  sex smoker
                                                 day
                                                        time
                                                               size
                               Female
     0
                16.99
                        1.01
                                                 Sun
                                                      Dinner
                                                                  2
                                           No
     1
                10.34
                        1.66
                                 Male
                                                      Dinner
                                                                  3
                                           No
                                                 Sun
     2
                21.01
                        3.50
                                                      Dinner
                                                                  3
                                 Male
                                           No
                                                 Sun
     3
                23.68
                        3.31
                                                                  2
                                                      Dinner
                                 Male
                                           No
                                                 Sun
     4
                24.59
                        3.61
                               Female
                                           No
                                                 Sun
                                                      Dinner
                                                                  4
     239
                29.03 5.92
                                 Male
                                           No
                                                 Sat
                                                      Dinner
                                                                  3
                27.18 2.00
                                                                  2
     240
                               Female
                                          Yes
                                                 Sat
                                                      Dinner
     241
                22.67
                        2.00
                                                                  2
                                 Male
                                          Yes
                                                 Sat
                                                      Dinner
```

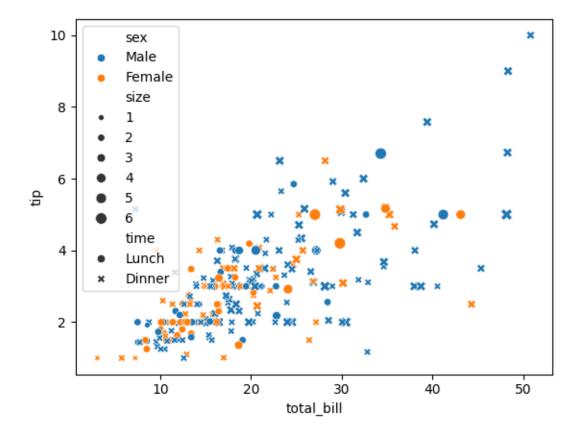
```
242 17.82 1.75 Male No Sat Dinner 2
243 18.78 3.00 Female No Thur Dinner 2
```

[244 rows x 7 columns]

### 1.1.1 sns.scatterplot(data, x, y, hue, style, size)

```
[6]: sns.scatterplot(data=tips, x='total_bill', y='tip', hue='sex', style='time', u size='size')
```

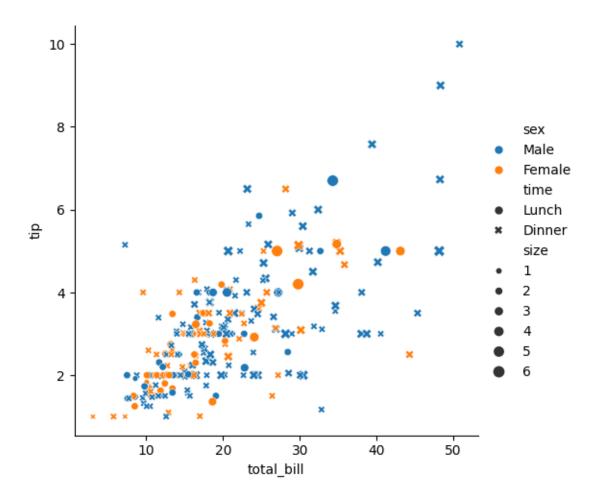
[6]: <Axes: xlabel='total\_bill', ylabel='tip'>



```
sns.relplot(data, x, y, kind, hue, style, size)
```

```
[8]: sns.relplot(data=tips, x='total_bill', y='tip', kind='scatter', hue='sex', u style='time', size='size')
```

[8]: <seaborn.axisgrid.FacetGrid at 0x250fe9cc980>



## 1.1.2 sns.lineplot(data, x, y)

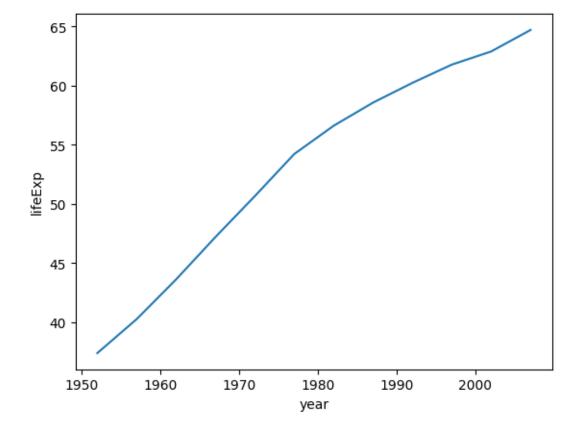
```
[11]: gap = px.data.gapminder()
temp = gap[gap['country'] == 'India']
temp
```

```
[11]:
          country continent
                                      lifeExp
                                                               gdpPercap iso_alpha
                               year
                                                       pop
      696
             India
                         Asia
                               1952
                                       37.373
                                                 372000000
                                                              546.565749
                                                                                IND
      697
             India
                         Asia
                               1957
                                       40.249
                                                 409000000
                                                              590.061996
                                                                                IND
      698
             India
                                       43.605
                                                 454000000
                                                              658.347151
                                                                                IND
                         Asia
                               1962
      699
             India
                         Asia
                               1967
                                       47.193
                                                 506000000
                                                              700.770611
                                                                                IND
      700
             India
                         Asia
                               1972
                                       50.651
                                                 567000000
                                                              724.032527
                                                                                IND
      701
             India
                         Asia
                               1977
                                       54.208
                                                 634000000
                                                              813.337323
                                                                                IND
      702
             India
                         Asia
                               1982
                                       56.596
                                                 708000000
                                                              855.723538
                                                                                IND
      703
             India
                                       58.553
                                                              976.512676
                         Asia
                               1987
                                                 788000000
                                                                                IND
      704
             India
                         Asia
                               1992
                                       60.223
                                                 872000000
                                                             1164.406809
                                                                                IND
      705
                                                 959000000
             India
                         Asia
                               1997
                                       61.765
                                                             1458.817442
                                                                                IND
      706
             India
                         Asia
                               2002
                                       62.879
                                                1034172547
                                                             1746.769454
                                                                                IND
```

```
707
      {\tt India}
                  Asia 2007
                                 64.698 1110396331 2452.210407
                                                                           IND
     iso_num
696
          356
697
          356
698
          356
699
          356
700
          356
701
          356
702
          356
703
          356
704
          356
705
          356
706
          356
707
          356
```

```
[13]: sns.lineplot(data=temp, x='year', y='lifeExp')
```

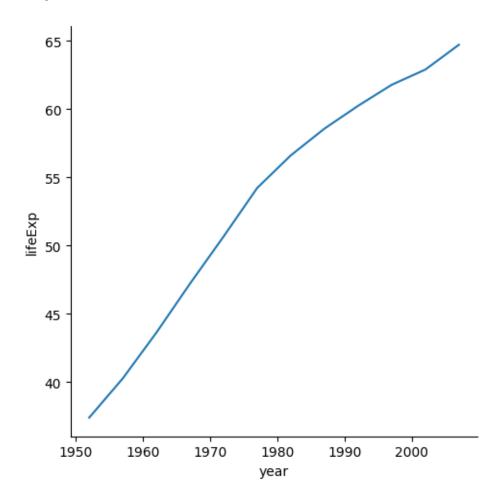
[13]: <Axes: xlabel='year', ylabel='lifeExp'>



sns.relplot(data, x, y, kind, hue, style, size)

```
[14]: sns.relplot(data=temp, x='year', y='lifeExp', kind='line')
```

[14]: <seaborn.axisgrid.FacetGrid at 0x250fe821d90>



```
[17]: temp = gap[gap['country'].isin(['India', 'Brazil', 'Germany'])]
      temp
[17]:
           country continent
                               year
                                     lifeExp
                                                               gdpPercap iso_alpha \
                                                      pop
            Brazil
                               1952
                                      50.917
                                                 56602560
                                                             2108.944355
                                                                               BRA
      168
                    Americas
                                      53.285
                                                             2487.365989
                                                                               BRA
      169
            Brazil
                    Americas
                               1957
                                                 65551171
      170
            Brazil
                    Americas
                               1962
                                      55.665
                                                                               BRA
                                                 76039390
                                                             3336.585802
      171
                                                                               BRA
            Brazil Americas
                               1967
                                      57.632
                                                 88049823
                                                             3429.864357
      172
            Brazil
                    Americas
                               1972
                                      59.504
                                                100840058
                                                             4985.711467
                                                                               BRA
      173
            Brazil
                               1977
                                      61.489
                                                                               BRA
                    Americas
                                                114313951
                                                             6660.118654
      174
            Brazil Americas
                               1982
                                      63.336
                                                128962939
                                                             7030.835878
                                                                               BRA
      175
            Brazil
                               1987
                                      65.205
                                                142938076
                                                             7807.095818
                                                                               BRA
                    Americas
      176
            Brazil
                    Americas
                               1992
                                      67.057
                                                155975974
                                                             6950.283021
                                                                               BRA
      177
            Brazil
                               1997
                                      69.388
                                                168546719
                                                             7957.980824
                                                                               BRA
                    Americas
```

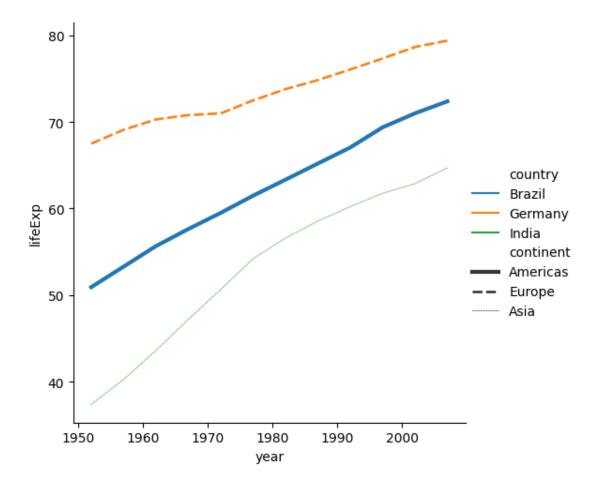
178	Brazil	Americas	2002	71.006	179914212	8131.212843	BRA
179	Brazil	Americas	2007	72.390	190010647	9065.800825	BRA
564	Germany	Europe	1952	67.500	69145952	7144.114393	DEU
565	Germany	Europe	1957	69.100	71019069	10187.826650	DEU
566	Germany	Europe	1962	70.300	73739117	12902.462910	DEU
567	Germany	Europe	1967	70.800	76368453	14745.625610	DEU
568	Germany	Europe	1972	71.000	78717088	18016.180270	DEU
569	Germany	Europe	1977	72.500	78160773	20512.921230	DEU
570	Germany	Europe	1982	73.800	78335266	22031.532740	DEU
571	Germany	Europe	1987	74.847	77718298	24639.185660	DEU
572	Germany	Europe	1992	76.070	80597764	26505.303170	DEU
573	Germany	Europe	1997	77.340	82011073	27788.884160	DEU
574	Germany	Europe	2002	78.670	82350671	30035.801980	DEU
575	Germany	Europe	2007	79.406	82400996	32170.374420	DEU
696	India	Asia	1952	37.373	372000000	546.565749	IND
697	India	Asia	1957	40.249	409000000	590.061996	IND
698	India	Asia	1962	43.605	454000000	658.347151	IND
699	India	Asia	1967	47.193	506000000	700.770611	IND
700	India	Asia	1972	50.651	567000000	724.032527	IND
701	India	Asia	1977	54.208	634000000	813.337323	IND
702	India	Asia	1982	56.596	708000000	855.723538	IND
703	India	Asia	1987	58.553	788000000	976.512676	IND
704	India	Asia	1992	60.223	872000000	1164.406809	IND
705	India	Asia	1997	61.765	959000000	1458.817442	IND
706	India	Asia	2002	62.879	1034172547	1746.769454	IND
707	India	Asia	2007	64.698	1110396331	2452.210407	IND

	iso_num
168	76
169	76
170	76
171	76
172	76
173	76
174	76
175	76
176	76
177	76
178	76
179	76
564	276
565	276
566	276
567	276
568	276
569	276
570	276

```
571
         276
572
         276
573
         276
574
         276
575
         276
696
         356
697
         356
698
         356
699
         356
700
         356
701
         356
702
         356
703
         356
704
         356
705
         356
706
         356
707
         356
```

```
[21]: sns.relplot(data=temp, x='year', y='lifeExp', kind='line', hue='country', u style='continent', size='continent')
```

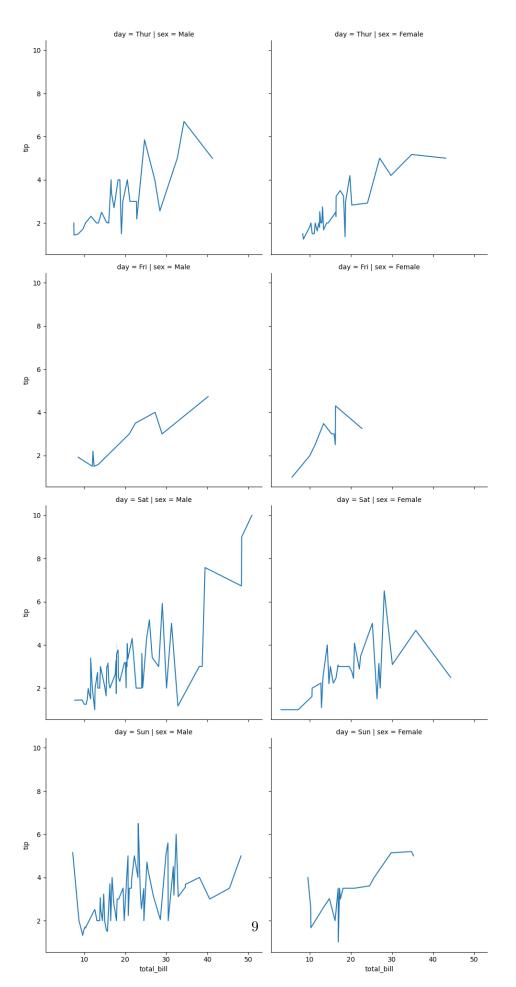
[21]: <seaborn.axisgrid.FacetGrid at 0x250835a25d0>



Facet Plot Figure level function, only work with relplot

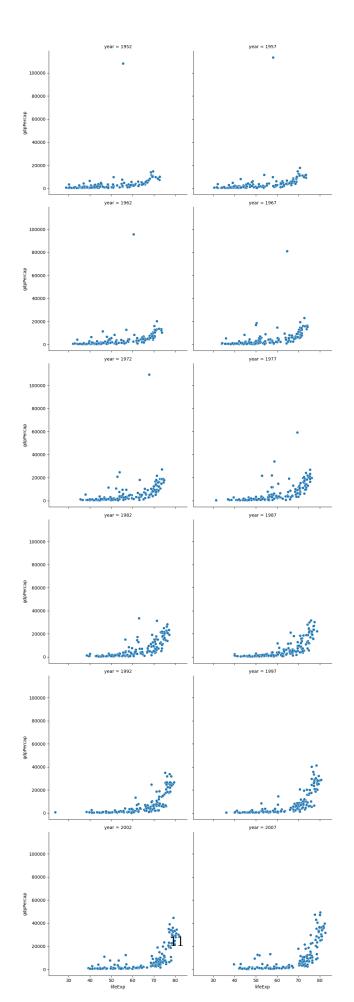
```
[22]: sns.relplot(data=tips, x='total_bill', y='tip', kind='line', col='sex', u orow='day')
```

[22]: <seaborn.axisgrid.FacetGrid at 0x2508358b770>



```
[23]: # Col Wrap
sns.relplot(data=gap, x='lifeExp', y='gdpPercap', kind='scatter', col='year',⊔
→col_wrap=2)
```

[23]: <seaborn.axisgrid.FacetGrid at 0x25082ca3770>



### 1.2 2. Distribution Plots

Generally used for - Univariate Analysis - used to find out the distribution of data - shows the range of the data - central tendency - used to check whether there are any outliers or the data bimodal?

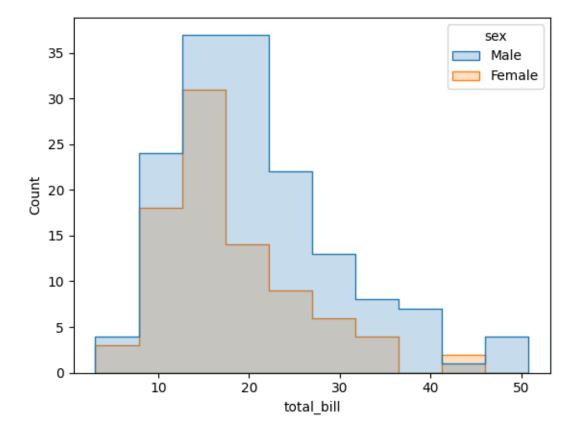
```
Figure level - displot()
Axis level - 1. histplot() 2. kdeplot() 3. rugplot()
```

### 1.2.1 sns.histplot(data, x, bins, hue, element)

element is used to designing

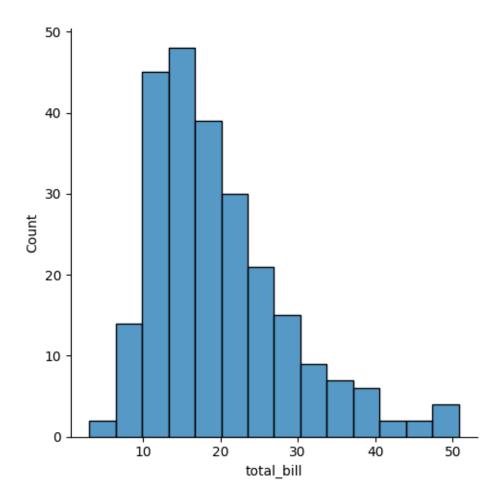
```
[29]: sns.histplot(data=tips, x='total_bill', bins=10, hue='sex', element='step')
```

[29]: <Axes: xlabel='total\_bill', ylabel='Count'>



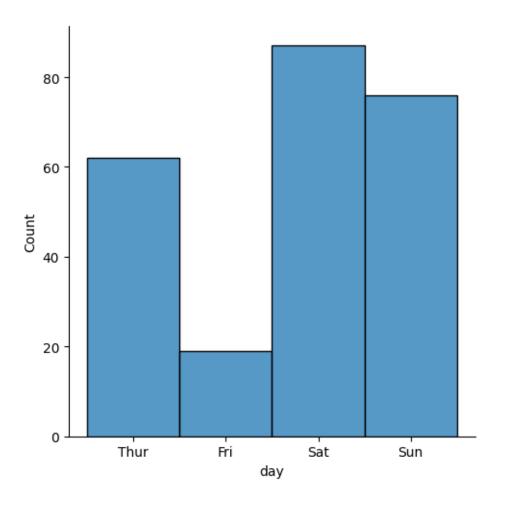
```
sns.displot(data, x, kind)
[25]: sns.displot(data=tips, x='total_bill', kind='hist')
```

[25]: <seaborn.axisgrid.FacetGrid at 0x2508390cf50>



```
[27]: # Categorical variables can also be plotted
sns.displot(data=tips, x='day', kind='hist')
```

[27]: <seaborn.axisgrid.FacetGrid at 0x25085d6e390>



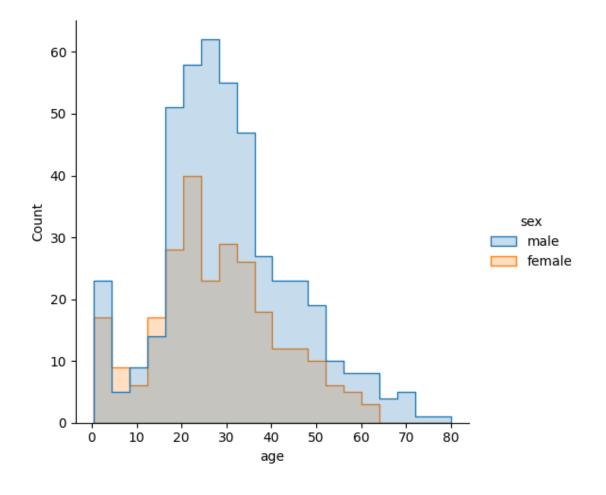
:	survived	pclass	sex	age	sibsp	parch	fare	embarked	class
0	0	3	male	22.0	1	0	7.2500	S	Third
1	1	1	female	38.0	1	0	71.2833	С	First
2	1	3	female	26.0	0	0	7.9250	S	Third
3	1	1	female	35.0	1	0	53.1000	S	First
4	0	3	male	35.0	0	0	8.0500	S	Third
	•••	•••		•••					
886	0	2	male	27.0	0	0	13.0000	S	Second
887	1	1	female	19.0	0	0	30.0000	S	First
888	0	3	female	NaN	1	2	23.4500	S	Third
889	1	1	male	26.0	0	0	30.0000	C	First
890	0	3	${\tt male}$	32.0	0	0	7.7500	Q	Third

```
False
1
     woman
                             С
                                   Cherbourg
                                                yes
                                                      False
2
                   False
                                Southampton
                           NaN
                                                yes
                                                       True
     woman
3
     woman
                   False
                             С
                                Southampton
                                                yes
                                                      False
4
                                Southampton
       man
                    True
                          NaN
                                                 no
                                                       True
886
                                Southampton
                                                       True
       man
                    True
                          {\tt NaN}
                                                 no
887
     woman
                  False
                             В
                                Southampton
                                                yes
                                                       True
888
                  False
                                Southampton
                                                      False
     woman
                          {\tt NaN}
                                                 no
889
                             С
                    True
                                   Cherbourg
                                                       True
                                                yes
       man
890
                    True
                          NaN
                                 Queenstown
                                                       True
       man
                                                 no
```

[891 rows x 15 columns]

```
[31]: sns.displot(data=titanic, x='age', kind='hist', element='step', hue='sex')
```

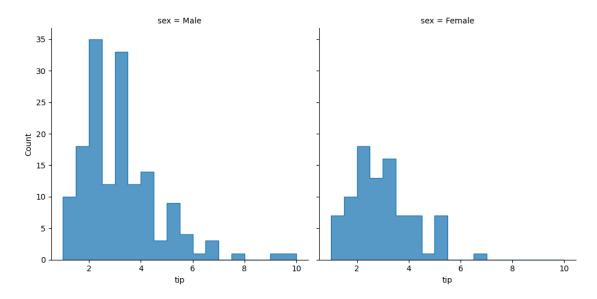
[31]: <seaborn.axisgrid.FacetGrid at 0x25082ca02c0>



Facet Plot

```
[32]: sns.displot(data=tips, x='tip', kind='hist', col='sex', element='step')
```

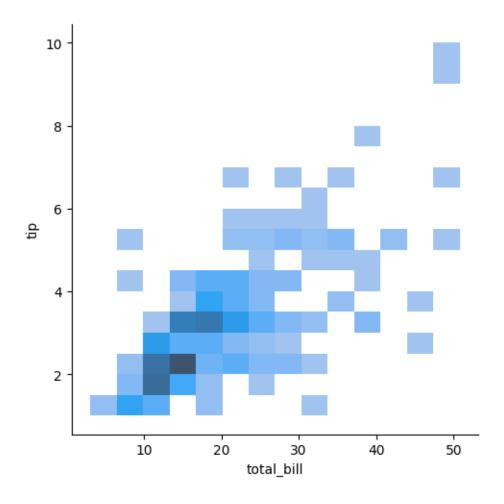
[32]: <seaborn.axisgrid.FacetGrid at 0x25087a84a70>



**Bivariate Histogram** A bivariate histogram bins the data within rectangles that tile the plot and then shows the count of observations within each rectangle with the fill color

```
[42]: # sns.histplot(data=tips, x='total_bill', y='tip')
sns.displot(data=tips, x='total_bill', y='tip', kind='hist')
```

[42]: <seaborn.axisgrid.FacetGrid at 0x25086a75580>

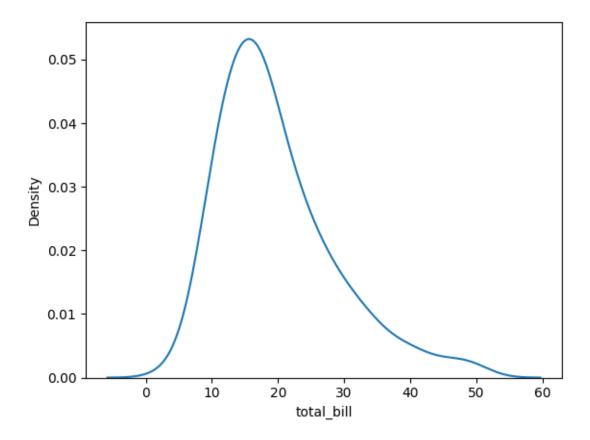


## 1.2.2 sns.kdeplot(data, x, hue, fill)

Rather than using discrete bins, a KDE plot smooths the observations with a Gaussian kernel, producing a continuous density estimate

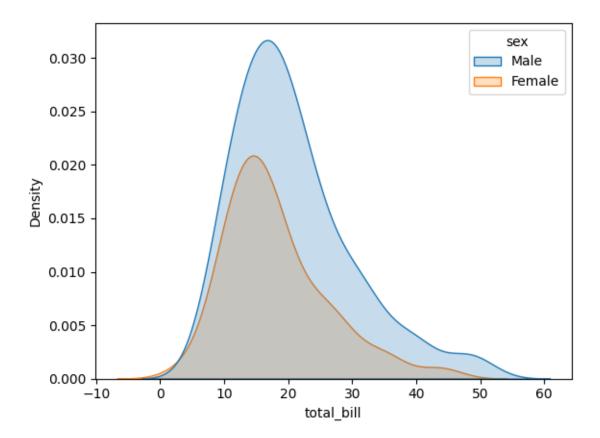
```
[33]: sns.kdeplot(data=tips, x='total_bill')
```

[33]: <Axes: xlabel='total\_bill', ylabel='Density'>



```
[39]: sns.kdeplot(data=tips, x='total_bill', hue='sex', fill=True)
```

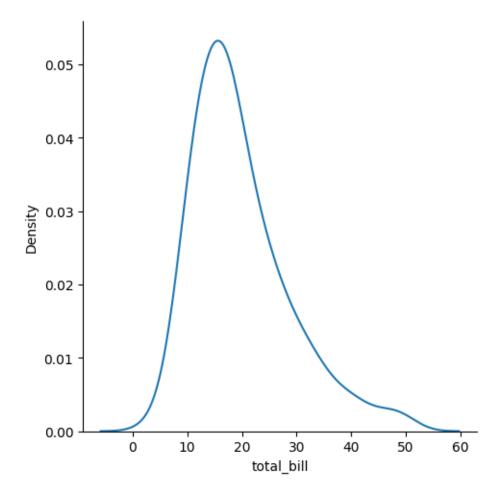
[39]: <Axes: xlabel='total\_bill', ylabel='Density'>



```
sns.displot(data, x, kind)
```

[34]: sns.displot(data=tips, x='total\_bill', kind='kde')

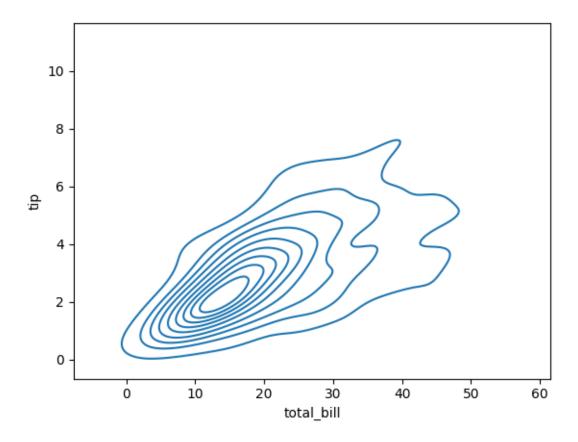
[34]: <seaborn.axisgrid.FacetGrid at 0x25087b63ec0>



Bivariate Kdeplot A bivariate kde plot smoothes the (x,y) observations with a 2D Gaussian

```
[43]: sns.kdeplot(data=tips, x='total_bill', y='tip')
```

[43]: <Axes: xlabel='total\_bill', ylabel='tip'>



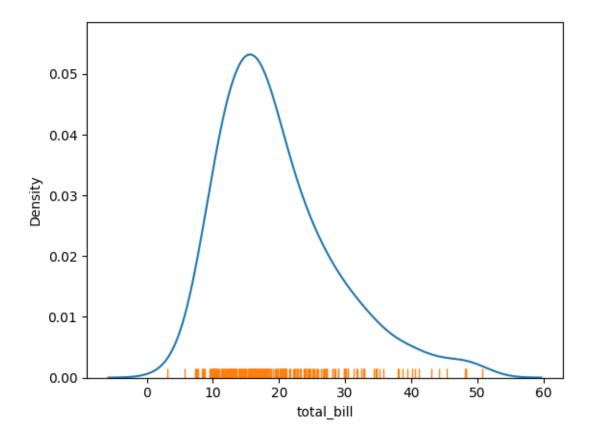
## 1.2.3 sns.rugplot(data, x)

Plot marginal distributions by drawing ticks along the axes.

This function is intended to complement other plots by showing the location of individual observations in an unobtrusive way

```
[40]: sns.kdeplot(data=tips, x='total_bill') sns.rugplot(data=tips, x='total_bill')
```

[40]: <Axes: xlabel='total\_bill', ylabel='Density'>



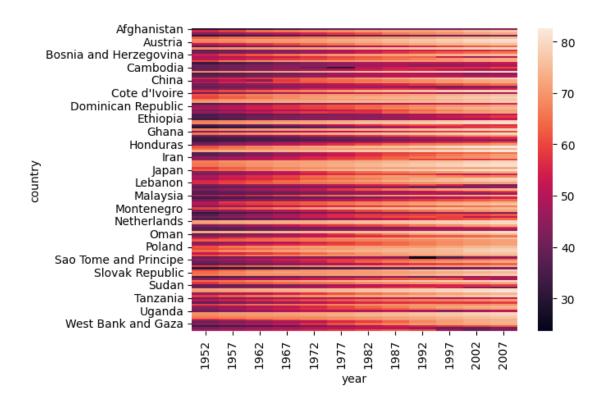
### 1.3 3. Matrix Plots

Figure level - displot() Axis level - 1. heatmap() 2. clustermap()

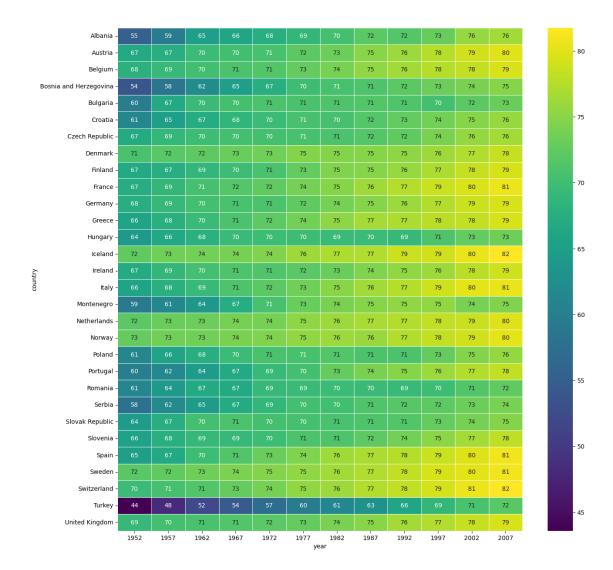
## 1.3.1 sns.heatmap(data, annot=False, linewidth, cmap)

```
[49]: temp = gap.pivot(index='country', columns='year', values='lifeExp') sns.heatmap(temp)
```

[49]: <Axes: xlabel='year', ylabel='country'>



[56]: <Axes: xlabel='year', ylabel='country'>



### 1.3.2 sns.clustermap(data, annot=False, linewidth, cmap)

Plot a matrix dataset as a hierarchically-clustered heatmap. This function requires scipy to be available.

```
[58]: import plotly.express as px
iris = px.data.iris()
iris
```

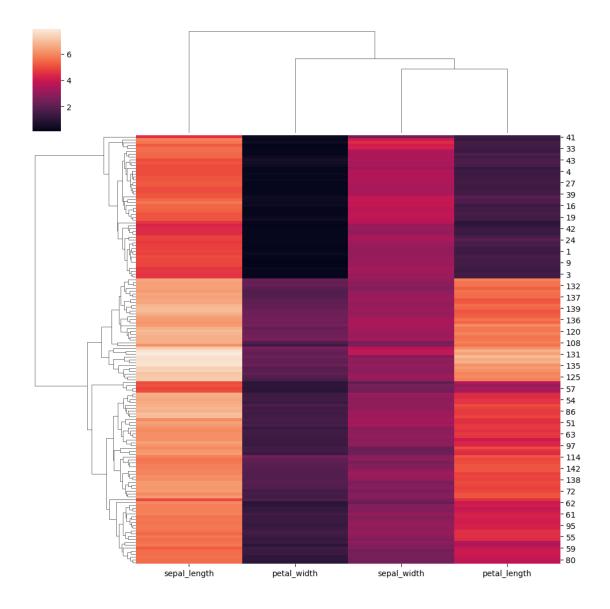
```
[58]:
                           sepal_width petal_length petal_width
            sepal_length
                                                                           species
      0
                      5.1
                                     3.5
                                                    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
                                                                   0.2
                                                    1.4
                                                                            setosa
```

```
5.2
                                                      2.3 virginica
145
              6.7
                           3.0
                                                      1.9 virginica
              6.3
                           2.5
                                         5.0
146
147
              6.5
                           3.0
                                         5.2
                                                      2.0 virginica
                                         5.4
                                                      2.3 virginica
148
              6.2
                           3.4
149
              5.9
                           3.0
                                                      1.8 virginica
                                         5.1
     species_id
0
1
              1
2
              1
3
              1
4
              1
145
              3
146
              3
147
              3
148
              3
149
              3
```

[150 rows x 6 columns]

```
[59]: sns.clustermap(iris.iloc[:, [0,1,2,3]])
```

[59]: <seaborn.matrix.ClusterGrid at 0x2508f22d160>



# 1.4 4. Categorical Plots

## 1.4.1 Categorical Scatter Plot

- Stripplot
- Swarmplot

## 1.4.2 Categorical Distribution Plots

- Boxplot
- Violinplot

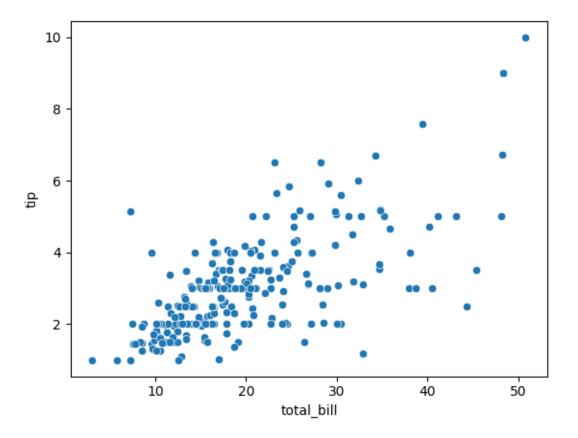
### 1.4.3 Categorical Estimate Plot -> for central tendency

- Barplot
- Pointplot
- Countplot

### 1.4.4 Figure level function -> catplot

```
[60]: sns.scatterplot(data=tips, x='total_bill', y='tip')
```

[60]: <Axes: xlabel='total\_bill', ylabel='tip'>

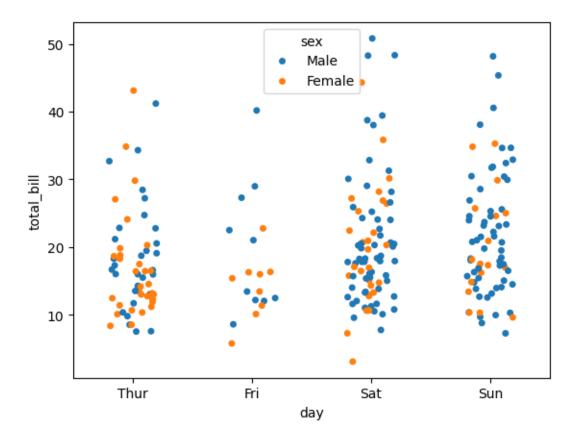


# ${\bf 1.4.5}\quad {\bf Categorical\ Scatter\ Plot}$

stripplot(data, x, y, jitter, hue) Axes level function

```
[63]: sns.stripplot(data=tips, x='day', y='total_bill', jitter=0.2, hue='sex')
```

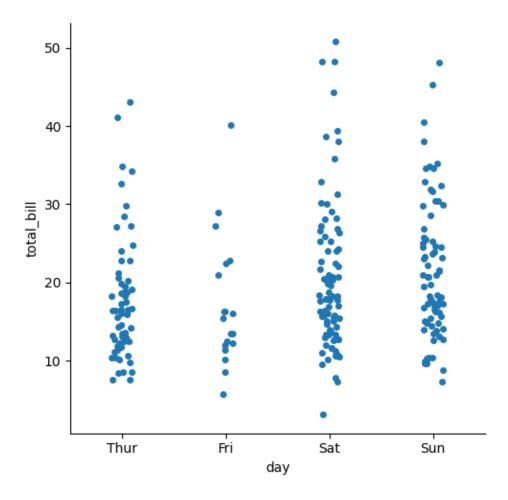
[63]: <Axes: xlabel='day', ylabel='total\_bill'>



 $\mathbf{sns.catplot(data,\,x,\,y,\,kind)} \quad \mathrm{Figure\,\,level\,\,function}$ 

```
[62]: sns.catplot(data=tips, x='day', y='total_bill', kind='strip')
```

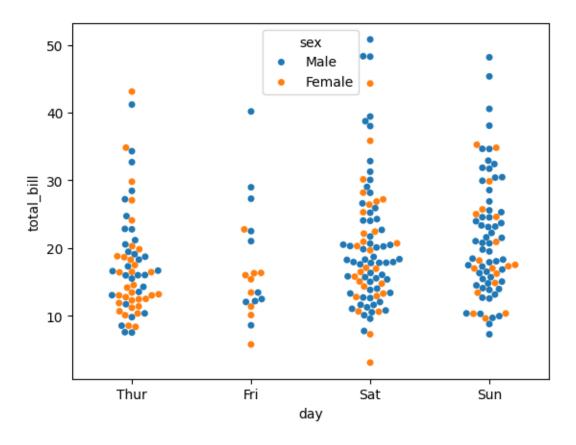
[62]: <seaborn.axisgrid.FacetGrid at 0x2508ea45490>



```
sns.swarmplot(data, x, y, hue)
```

```
[65]: sns.swarmplot(data=tips, x='day', y='total_bill', hue='sex')
```

[65]: <Axes: xlabel='day', ylabel='total\_bill'>



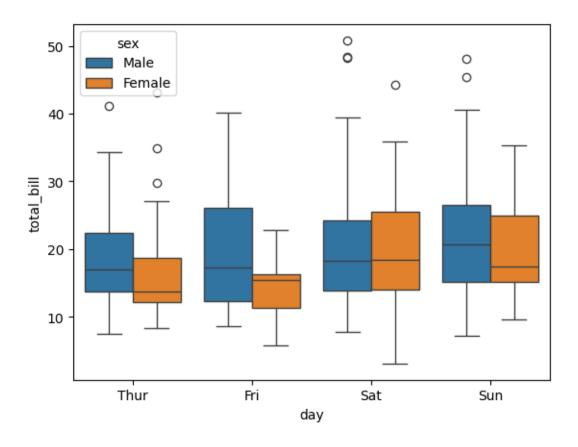
### 1.4.6 Boxplot

A boxplot is a standardized way of displaying the distribution of data based on a five number summary ("minimum", first quartile [Q1], median, third quartile [Q3] and "maximum"). It can tell you about your outliers and what their values are. Boxplots can also tell you if your data is symmetrical, how tightly your data is grouped and if and how your data is skewed.

## 1.4.7 sns.boxplot(data, x, y, hue)

```
[68]: sns.boxplot(data=tips, x='day', y='total_bill', hue='sex')
```

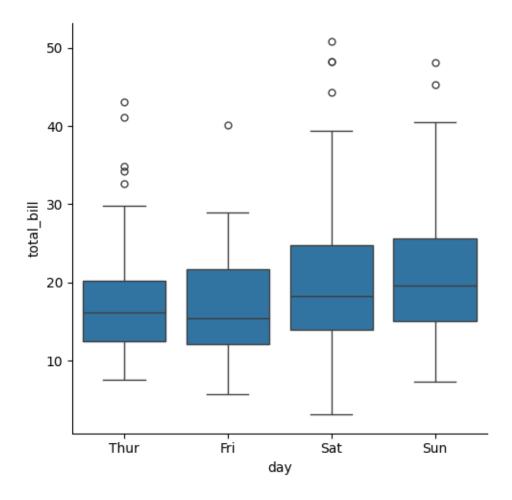
[68]: <Axes: xlabel='day', ylabel='total\_bill'>



# 1.4.8 sns.catplot(data, x, y, kind)

```
[67]: sns.catplot(data=tips, x='day', y='total_bill', kind='box')
```

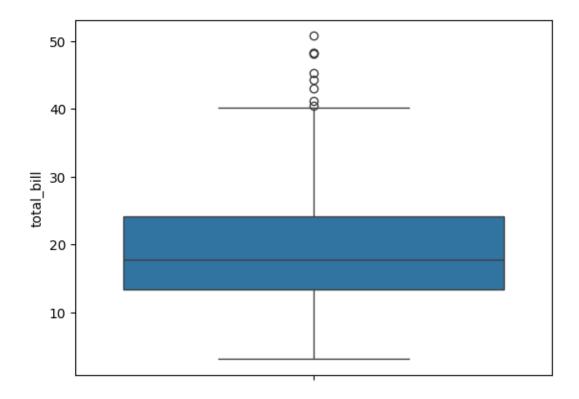
[67]: <seaborn.axisgrid.FacetGrid at 0x2508f8e6870>



Single Boxplot -> numerical column

```
[69]: sns.boxplot(data=tips, y='total_bill')
```

[69]: <Axes: ylabel='total\_bill'>

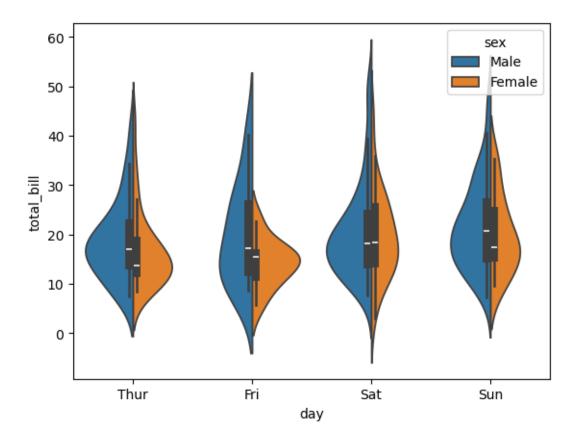


### 1.4.9 sns.violinplot(data, x, y, hue, split=False)

By default split is False which means for every category in hue there is two violin plots, but if split is True then in single violin plot both category is merged.

```
[73]: sns.violinplot(data=tips, x='day', y='total_bill', hue='sex', split=True)
```

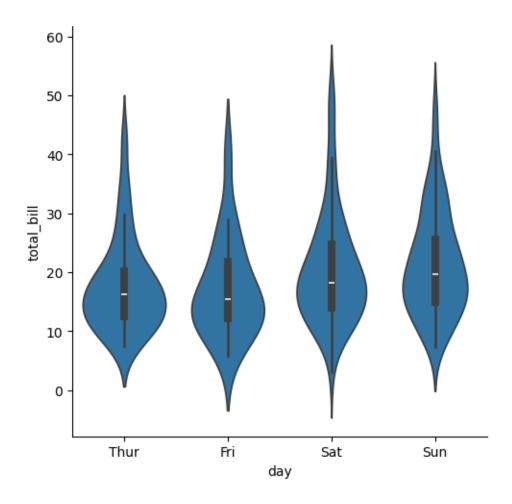
[73]: <Axes: xlabel='day', ylabel='total\_bill'>



```
sns.catplot(data, x, y, kind)
```

```
[71]: sns.catplot(data=tips, x='day', y='total_bill', kind='violin')
```

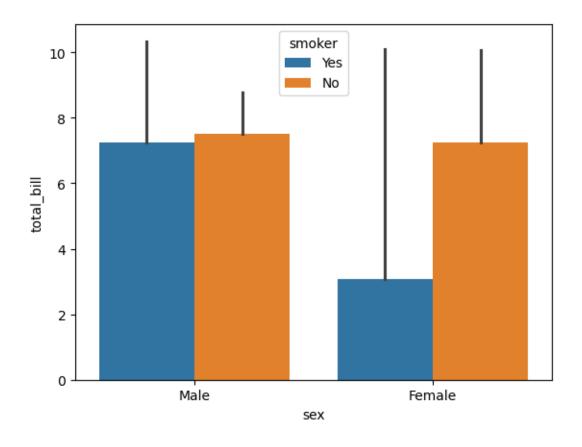
[71]: <seaborn.axisgrid.FacetGrid at 0x25086b73050>



# 1.4.10 sns.barplot(data, x, y, hue, estimator)

```
[79]: sns.barplot(data=tips, x='sex', y='total_bill', hue='smoker', estimator=np.min)
```

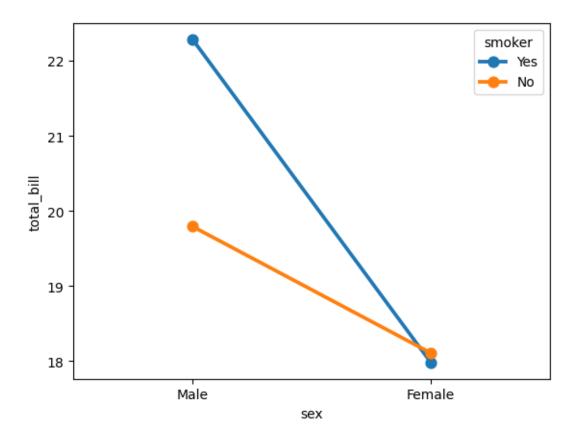
[79]: <Axes: xlabel='sex', ylabel='total\_bill'>



# $1.4.11 \hspace{0.1in} sns.pointplot(data,\,x,\,y,\,hue,\,errorbar)$

```
[83]: sns.pointplot(data=tips, x='sex', y='total_bill', hue='smoker', errorbar=None)
```

[83]: <Axes: xlabel='sex', ylabel='total\_bill'>

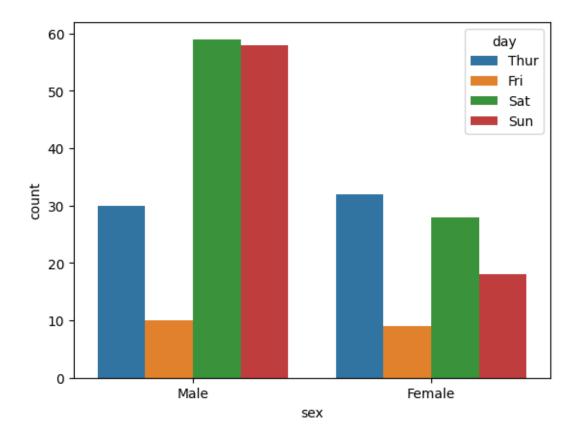


### 1.4.12 sns.countplot(data, x, hue)

When there are multiple observations in each category, it also uses bootstrapping to compute a cofidence interval around the estimate, which is plotted using error bars.

```
[84]: sns.countplot(data=tips, x='sex', hue='day')
```

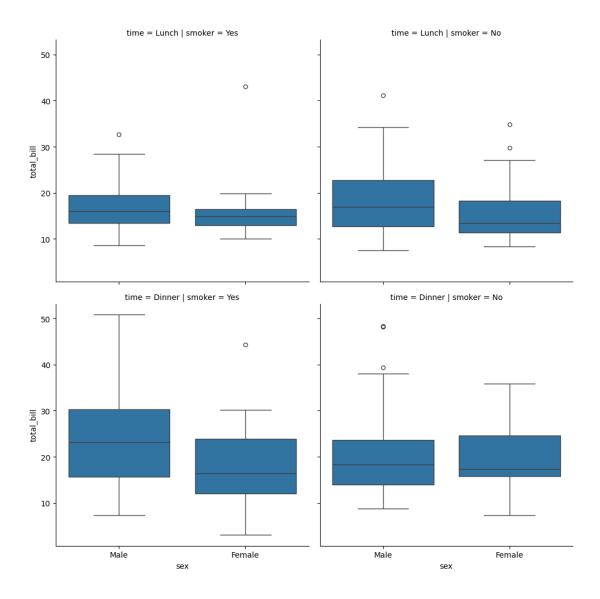
[84]: <Axes: xlabel='sex', ylabel='count'>



# Facet plots

```
[85]: sns.catplot(data=tips, we see state of the state of
```

[85]: <seaborn.axisgrid.FacetGrid at 0x25092bbd520>



#### 1.5 4. Regression Plots

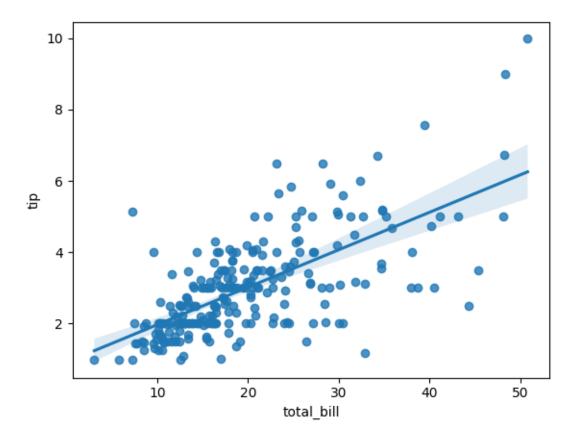
- regplot
- lmplot

In the simplest invocation, both functions draw a scatter plot of two variables, x and y, and then fit the regression model y  $\sim$  x and plot the resulting regression line and a 95% confidence interval for that regression.

Hue parameter is not available.

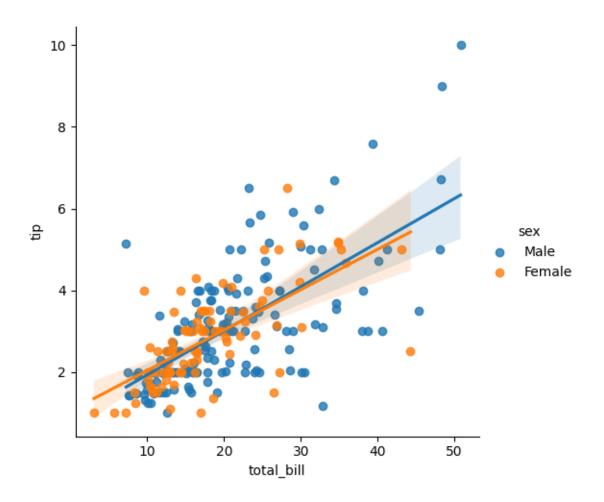
```
[86]: sns.regplot(data=tips, x='total_bill', y='tip')
```

[86]: <Axes: xlabel='total\_bill', ylabel='tip'>



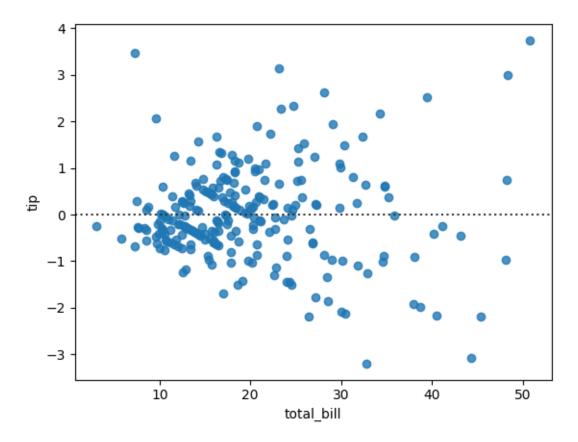
```
[87]: sns.lmplot(data=tips, x='total_bill', y='tip', hue='sex')
```

[87]: <seaborn.axisgrid.FacetGrid at 0x25092e10bc0>



```
[88]: sns.residplot(data=tips, x='total_bill', y='tip')
```

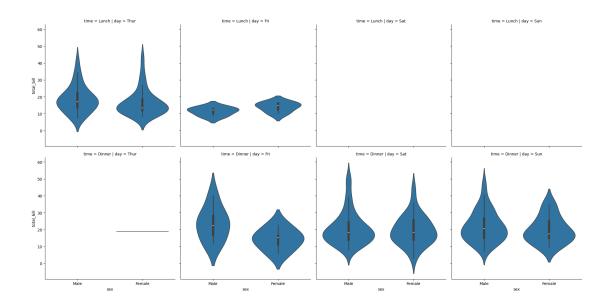
[88]: <Axes: xlabel='total\_bill', ylabel='tip'>



### 1.5.1 Facet Grid

```
[89]: sns.catplot(data=tips, x='sex', y='total_bill', kind='violin', col='day', u orow='time')
```

[89]: <seaborn.axisgrid.FacetGrid at 0x25092e6fd10>



```
[92]: g = sns.FacetGrid(data=tips, col='day', row='time', hue='smoker')
g.map(sns.boxplot, 'sex', 'total_bill')
g.add_legend()
```

c:\Program Files\Python312\Lib\site-packages\seaborn\axisgrid.py:718:
UserWarning:

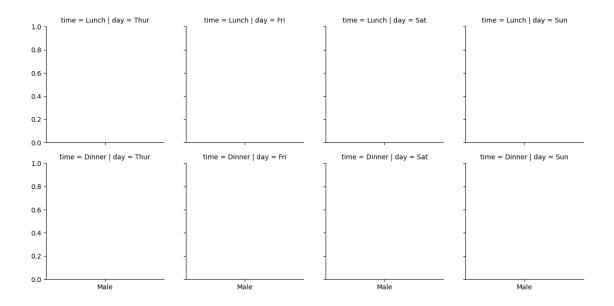
Using the boxplot function without specifying `order` is likely to produce an incorrect plot.

```
TypeError
                                           Traceback (most recent call last)
Cell In[92], line 2
      1 g = sns.FacetGrid(data=tips, col='day', row='time', hue='smoker')
----> 2 g.map(sns.boxplot, 'sex', 'total_bill')
      3 g.add_legend()
File c:\Program Files\Python312\Lib\site-packages\seaborn\axisgrid.py:758, in_

→FacetGrid.map(self, func, *args, **kwargs)
    755
                plot_args = [v.values for v in plot_args]
            # Draw the plot
    757
            self._facet_plot(func, ax, plot_args, kwargs)
--> 758
    760 # Finalize the annotations and layout
    761 self._finalize_grid(args[:2])
File c:\Program Files\Python312\Lib\site-packages\seaborn\axisgrid.py:854, in _{\!\sqcup}
 →FacetGrid._facet_plot(self, func, ax, plot_args, plot_kwargs)
            plot_args = []
    852
```

```
853
             plot_kwargs["ax"] = ax
--> 854 func(*plot_args, **plot_kwargs)
    856 # Sort out the supporting information
    857 self._update_legend_data(ax)
File c:\Program Files\Python312\Lib\site-packages\seaborn\categorical.py:1634,_
 oin boxplot(data, x, y, hue, order, hue_order, orient, color, palette, usaturation, fill, dodge, width, gap, whis, linecolor, linewidth, fliersize, usaturation
 hue_norm, native_scale, log_scale, formatter, legend, ax, **kwargs)
   1627 color = _default_color(
   1628
             ax.fill_between, hue, color,
   1629
             {k: v for k, v in kwargs.items() if k in ["c", "color", "fc",

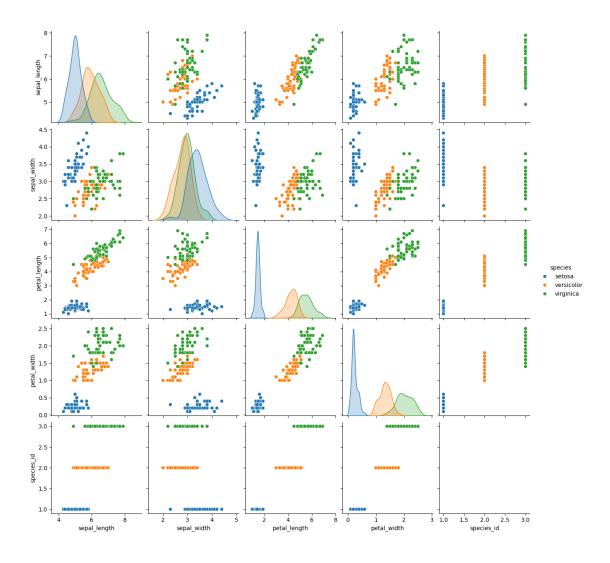
¬"facecolor"]},
   1630
             saturation=saturation,
   1631 )
   1632 linecolor = p._complement_color(linecolor, color, p._hue_map)
-> 1634 p.plot boxes(
   1635
             width=width,
   1636
             dodge=dodge,
   1637
             gap=gap,
             fill=fill,
   1638
   1639
             whis=whis,
   1640
             color=color,
             linecolor=linecolor,
   1641
   1642
             linewidth=linewidth,
   1643
             fliersize=fliersize,
   1644
             plot_kws=kwargs,
   1645 )
   1647 p._add_axis_labels(ax)
   1648 p. adjust cat axis(ax, axis=p.orient)
File c:\Program Files\Python312\Lib\site-packages\seaborn\categorical.py:700, i
 → CategoricalPlotter.plot boxes(self, width, dodge, gap, fill, whis, color, u
 →linecolor, linewidth, fliersize, plot_kws)
    679 default_kws = dict(
             bxpstats=stats.to_dict("records"),
    680
             positions=data[self.orient],
    681
   (...)
    697
             )
    698 )
    699 boxplot_kws = {**default_kws, **plot_kws}
--> 700 artists = ax.bxp(**boxplot kws)
    702 # Reset artist widths after adding so everything stays positive
    703 ori idx = ["x", "y"].index(self.orient)
TypeError: Axes.bxp() got an unexpected keyword argument 'label'
```



# 1.6 Plotting Pairwise Relationship (PairGrid vs Pairplot)

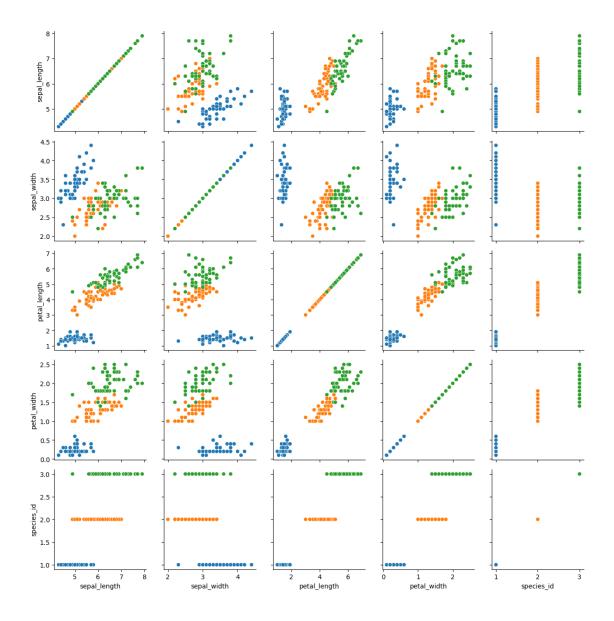
```
[93]: sns.pairplot(iris, hue='species')
```

[93]: <seaborn.axisgrid.PairGrid at 0x25095b760f0>



```
[95]: # pair grid
g = sns.PairGrid(data=iris, hue='species')
# g.map
g.map(sns.scatterplot)
```

[95]: <seaborn.axisgrid.PairGrid at 0x25099c51f10>



```
[96]: # map_diag -> map_offdiag

g = sns.PairGrid(data=iris, hue='species')
g.map_diag(sns.boxplot)
g.map_offdiag(sns.kdeplot)
```

c:\Program Files\Python312\Lib\site-packages\seaborn\axisgrid.py:1615:
UserWarning:

KDE cannot be estimated (0 variance or perfect covariance). Pass `warn\_singular=False` to disable this warning.

c:\Program Files\Python312\Lib\site-packages\seaborn\axisgrid.py:1615:

#### UserWarning:

KDE cannot be estimated (0 variance or perfect covariance). Pass `warn\_singular=False` to disable this warning.

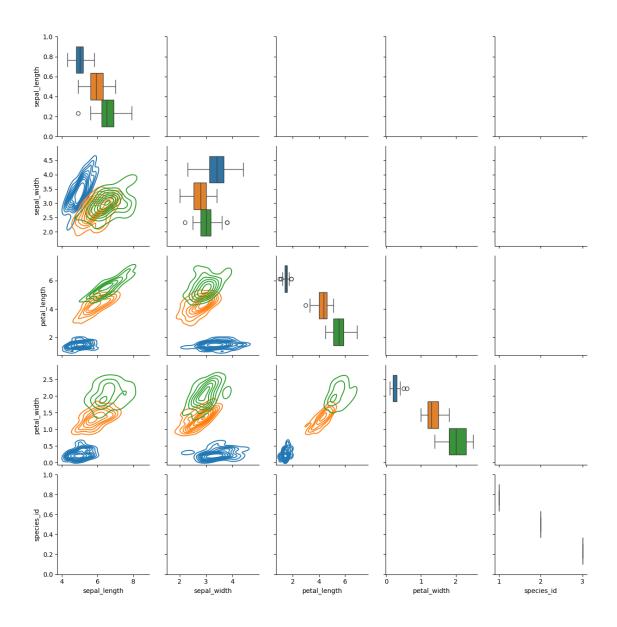
c:\Program Files\Python312\Lib\site-packages\seaborn\axisgrid.py:1615:
UserWarning:

KDE cannot be estimated (0 variance or perfect covariance). Pass `warn\_singular=False` to disable this warning.

```
IndexError
                                          Traceback (most recent call last)
Cell In[96], line 5
      3 g = sns.PairGrid(data=iris, hue='species')
     4 g.map diag(sns.boxplot)
---> 5 g.map_offdiag(sns.kdeplot)
File c:\Program Files\Python312\Lib\site-packages\seaborn\axisgrid.py:1425, in_
 →PairGrid.map_offdiag(self, func, **kwargs)
   1414 """Plot with a bivariate function on the off-diagonal subplots.
   1415
   1416 Parameters
   (...)
   1422
   1423 """
   1424 if self.square_grid:
-> 1425
            self.map_lower(func, **kwargs)
            if not self. corner:
   1426
   1427
                self.map_upper(func, **kwargs)
File c:\Program Files\Python312\Lib\site-packages\seaborn\axisgrid.py:1395, in_
 →PairGrid.map_lower(self, func, **kwargs)
   1384 """Plot with a bivariate function on the lower diagonal subplots.
   1385
   1386 Parameters
   (...)
   1392
   1393 """
   1394 indices = zip(*np.tril_indices_from(self.axes, -1))
-> 1395 self._map_bivariate(func, indices, **kwargs)
   1396 return self
File c:\Program Files\Python312\Lib\site-packages\seaborn\axisgrid.py:1574, in_
 PairGrid._map_bivariate(self, func, indices, **kwargs)
            if ax is None: # i.e. we are in corner mode
```

```
1573
                                       continue
-> 1574
                             self._plot_bivariate(x_var, y_var, ax, func, **kws)
       1575 self._add_axis_labels()
       1577 if "hue" in signature(func).parameters:
File c:\Program Files\Python312\Lib\site-packages\seaborn\axisgrid.py:1615, in_
   →PairGrid._plot_bivariate(self, x_var, y_var, ax, func, **kwargs)
       1611 if "hue" not in kwargs:
       1612
                             kwargs.update({
       1613
                                        "hue": hue, "hue_order": self._hue_order, "palette": self.
  →_orig_palette,
       1614
                             })
-> 1615 func(x=x, y=y, **kwargs)
       1617 self. update legend data(ax)
File c:\Program Files\Python312\Lib\site-packages\seaborn\distributions.py:1715
    in kdeplot(data, x, y, hue, weights, palette, hue_order, hue_norm, color, ا
   ofill, multiple, common_norm, common_grid, cumulative, bw_method, bw_adjust, warn_singular, log_scale, levels, thresh, gridsize, cut, clip, legend, cbar,
   ⇔cbar_ax, cbar_kws, ax, **kwargs)
       1701
                             p.plot_univariate_density(
       1702
                                       multiple=multiple,
       1703
                                       common_norm=common_norm,
       (...)
       1710
                                       **plot_kws,
       1711
                             )
       1713 else:
-> 1715
                             p.plot_bivariate_density(
       1716
                                       common norm=common norm,
       1717
                                       fill=fill.
       1718
                                       levels=levels,
       1719
                                       thresh=thresh,
       1720
                                       legend=legend,
       1721
                                        color=color,
       1722
                                       warn_singular=warn_singular,
       1723
                                        cbar=cbar,
       1724
                                        cbar_ax=cbar_ax,
       1725
                                       cbar_kws=cbar_kws,
       1726
                                       estimate_kws=estimate_kws,
       1727
                                        **kwargs,
       1728
       1730 return ax
File c:\Program Files\Python312\Lib\site-packages\seaborn\distributions.py:1113
   in _DistributionPlotter.plot_bivariate_density(self, common_norm, fill, olevels, thresh, color, legend, cbar, warn_singular, cbar_ax, cbar_kws, olevels, color, legend, cbar, warn_singular, cbar_ax, cbar_kws, olevels, color, legend, cbar, warn_singular, cbar_ax, cbar_kws, olevels, color, legend, cbar_ax, cba
   ⇔estimate_kws, **contour_kws)
       1111 # Transform from iso-proportions to iso-densities
       1112 if common_norm:
```

```
-> 1113
            common_levels = self._quantile_to_level(
                list(densities.values()), levels,
   1114
   1115
            draw_levels = {k: common_levels for k in densities}
   1116
   1117 else:
File c:\Program Files\Python312\Lib\site-packages\seaborn\distributions.py:200,
 →in _DistributionPlotter._quantile_to_level(self, data, quantile)
    198 normalized values = np.cumsum(sorted values) / values.sum()
    199 idx = np.searchsorted(normalized_values, 1 - isoprop)
--> 200 levels = np.take(sorted_values, idx, mode="clip")
    201 return levels
File c:\Program Files\Python312\Lib\site-packages\numpy\core\fromnumeric.py:192
 →in take(a, indices, axis, out, mode)
     95 @array_function_dispatch(_take_dispatcher)
     96 def take(a, indices, axis=None, out=None, mode='raise'):
     97
     98
            Take elements from an array along an axis.
     99
   (...)
    190
                   [5, 7]]
    191
--> 192
            return _wrapfunc(a, 'take', indices, axis=axis, out=out, mode=mode)
File c:\Program Files\Python312\Lib\site-packages\numpy\core\fromnumeric.py:59,
 →in _wrapfunc(obj, method, *args, **kwds)
            return _wrapit(obj, method, *args, **kwds)
     56
     58 try:
---> 59
            return bound(*args, **kwds)
     60 except TypeError:
            # A TypeError occurs if the object does have such a method in its
     62
            # class, but its signature is not identical to that of NumPy's. This
   (...)
            # Call wrapit from within the except clause to ensure a potential
     66
     67
            # exception has a traceback chain.
            return wrapit(obj, method, *args, **kwds)
IndexError: cannot do a non-empty take from an empty axes.
```



```
[97]: # map_diag -> map_upper -> map_lower
g = sns.PairGrid(data=iris, hue='species')
g.map_diag(sns.histplot)
g.map_upper(sns.kdeplot)
g.map_lower(sns.scatterplot)
```

c:\Program Files\Python312\Lib\site-packages\seaborn\axisgrid.py:1615:
UserWarning:

KDE cannot be estimated (0 variance or perfect covariance). Pass `warn\_singular=False` to disable this warning.

c:\Program Files\Python312\Lib\site-packages\seaborn\axisgrid.py:1615:

#### UserWarning:

KDE cannot be estimated (0 variance or perfect covariance). Pass `warn\_singular=False` to disable this warning.

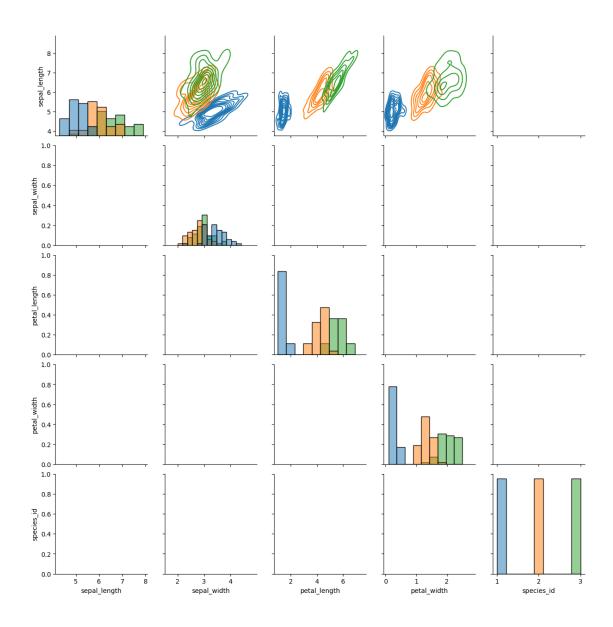
c:\Program Files\Python312\Lib\site-packages\seaborn\axisgrid.py:1615:
UserWarning:

KDE cannot be estimated (0 variance or perfect covariance). Pass `warn\_singular=False` to disable this warning.

```
IndexError
                                          Traceback (most recent call last)
Cell In[97], line 4
      2 g = sns.PairGrid(data=iris, hue='species')
     3 g.map diag(sns.histplot)
---> 4 g.map_upper(sns.kdeplot)
     5 g.map_lower(sns.scatterplot)
File c:\Program Files\Python312\Lib\site-packages\seaborn\axisgrid.py:1410, in_
 →PairGrid.map_upper(self, func, **kwargs)
   1399 """Plot with a bivariate function on the upper diagonal subplots.
   1400
   1401 Parameters
   (...)
   1407
  1408 """
   1409 indices = zip(*np.triu_indices_from(self.axes, 1))
-> 1410 self._map_bivariate(func, indices, **kwargs)
   1411 return self
File c:\Program Files\Python312\Lib\site-packages\seaborn\axisgrid.py:1574, in_
 ←PairGrid._map_bivariate(self, func, indices, **kwargs)
   if ax is None: # i.e. we are in corner mode
   1573
                continue
           self._plot_bivariate(x_var, y_var, ax, func, **kws)
-> 1574
   1575 self._add_axis_labels()
   1577 if "hue" in signature(func).parameters:
File c:\Program Files\Python312\Lib\site-packages\seaborn\axisgrid.py:1615, in_
 PairGrid._plot_bivariate(self, x_var, y_var, ax, func, **kwargs)
   1611 if "hue" not in kwargs:
   1612
           kwargs.update({
                "hue": hue, "hue order": self. hue order, "palette": self.
 →_orig_palette,
  1614
          })
```

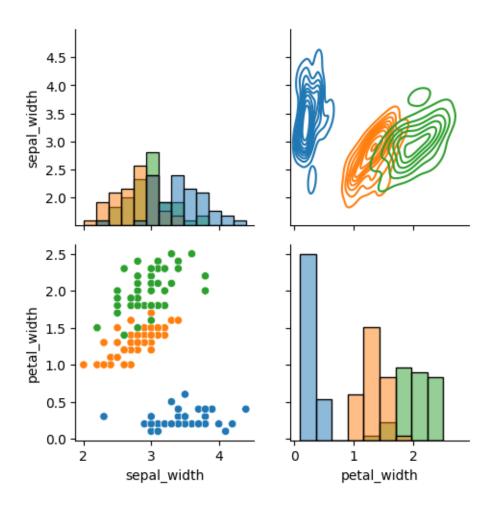
```
-> 1615 func(x=x, y=y, **kwargs)
       1617 self._update_legend_data(ax)
File c:\Program Files\Python312\Lib\site-packages\seaborn\distributions.py:1715
   oin kdeplot(data, x, y, hue, weights, palette, hue_order, hue_norm, color, ofill, multiple, common_norm, common_grid, cumulative, bw_method, bw_adjust, owarn_singular, log_scale, levels, thresh, gridsize, cut, clip, legend, cbar, or color, or co
   ⇔cbar_ax, cbar_kws, ax, **kwargs)
                            p.plot_univariate_density(
       1701
       1702
                                     multiple=multiple,
       1703
                                      common_norm=common_norm,
       (...)
       1710
                                      **plot_kws,
       1711
       1713 else:
-> 1715
                            p.plot bivariate density(
       1716
                                      common norm=common norm,
       1717
                                     fill=fill,
       1718
                                     levels=levels,
       1719
                                     thresh=thresh,
       1720
                                     legend=legend,
       1721
                                     color=color,
       1722
                                     warn_singular=warn_singular,
       1723
                                      cbar=cbar,
       1724
                                      cbar_ax=cbar_ax,
       1725
                                      cbar_kws=cbar_kws,
       1726
                                      estimate kws=estimate kws,
       1727
                                      **kwargs,
       1728
       1730 return ax
File c:\Program Files\Python312\Lib\site-packages\seaborn\distributions.py:1113
   →in DistributionPlotter.plot_bivariate_density(self, common_norm, fill,
   elevels, thresh, color, legend, cbar, warn_singular, cbar_ax, cbar_kws,u
   ⇔estimate kws, **contour kws)
       1111 # Transform from iso-proportions to iso-densities
       1112 if common norm:
-> 1113
                            common_levels = self._quantile_to_level(
       1114
                                     list(densities.values()), levels,
       1115
       1116
                            draw levels = {k: common levels for k in densities}
       1117 else:
File c:\Program Files\Python312\Lib\site-packages\seaborn\distributions.py:200,
   oin DistributionPlotter. quantile to level (self, data, quantile)
         198 normalized_values = np.cumsum(sorted_values) / values.sum()
         199 idx = np.searchsorted(normalized_values, 1 - isoprop)
--> 200 levels = np.take(sorted_values, idx, mode="clip")
         201 return levels
```

```
File c:\Program Files\Python312\Lib\site-packages\numpy\core\fromnumeric.py:192
 →in take(a, indices, axis, out, mode)
     95 @array_function_dispatch(_take_dispatcher)
     96 def take(a, indices, axis=None, out=None, mode='raise'):
     97
            Take elements from an array along an axis.
     98
     99
   (...)
    190
                   [5, 7]]
            11 11 11
    191
            return _wrapfunc(a, 'take', indices, axis=axis, out=out, mode=mode)
--> 192
File c:\Program Files\Python312\Lib\site-packages\numpy\core\fromnumeric.py:59,
 →in _wrapfunc(obj, method, *args, **kwds)
            return _wrapit(obj, method, *args, **kwds)
     58 try:
            return bound(*args, **kwds)
---> 59
     60 except TypeError:
            # A TypeError occurs if the object does have such a method in its
            # class, but its signature is not identical to that of NumPy's. This
     62
   (...)
            # Call _wrapit from within the except clause to ensure a potential
            # exception has a traceback chain.
     68
            return _wrapit(obj, method, *args, **kwds)
IndexError: cannot do a non-empty take from an empty axes.
```



```
[98]: # vars
g = sns.PairGrid(data=iris,hue='species',vars=['sepal_width','petal_width'])
g.map_diag(sns.histplot)
g.map_upper(sns.kdeplot)
g.map_lower(sns.scatterplot)
```

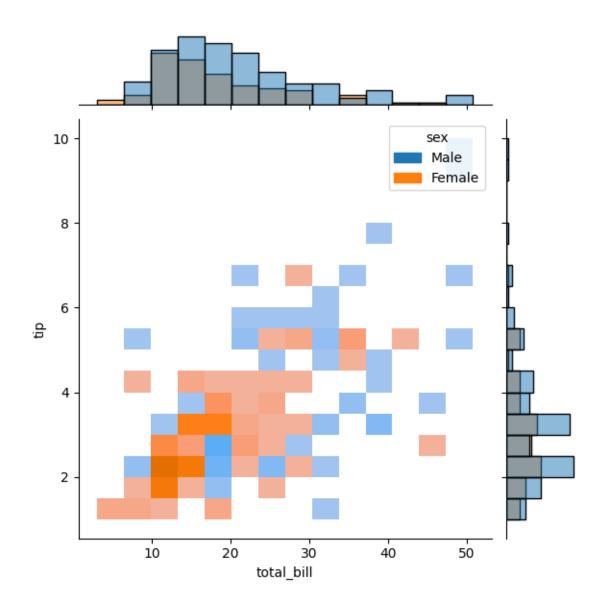
[98]: <seaborn.axisgrid.PairGrid at 0x250a056e690>



# 1.7 JointGrid Vs Jointplot

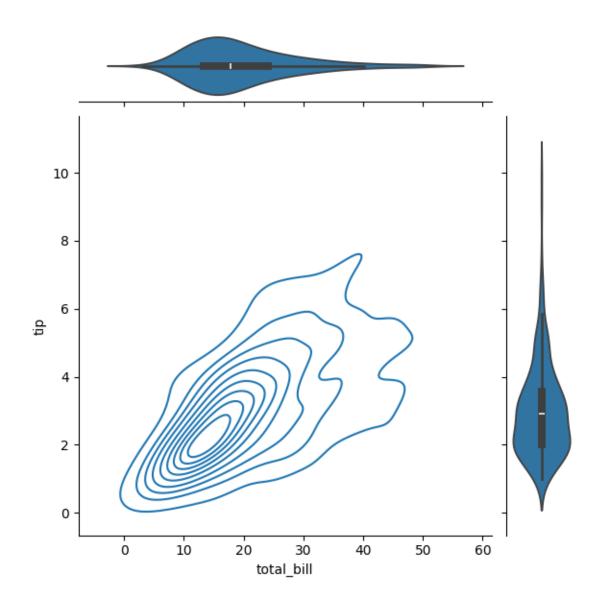
```
[99]: sns.jointplot(data=tips, x='total_bill', y='tip', kind='hist', hue='sex')
```

[99]: <seaborn.axisgrid.JointGrid at 0x250a0feca70>



```
[101]: g = sns.JointGrid(data=tips, x='total_bill', y='tip')
g.plot(sns.kdeplot, sns.violinplot)
```

[101]: <seaborn.axisgrid.JointGrid at 0x250a2aa20f0>



#### 1.7.1 Utility Functions

```
[104]: sns.get_dataset_names()
```

```
File c:\Program Files\Python312\Lib\http\client.py:1336, in HTTPConnection.
 1335 """Send a complete request to the server."""
-> 1336 self._send_request(method, url, body, headers, encode_chunked)
File c:\Program Files\Python312\Lib\http\client.py:1382, in HTTPConnection.
 →_send_request(self, method, url, body, headers, encode_chunked)
           body = _encode(body, 'body')
-> 1382 self.endheaders(body, encode_chunked=encode_chunked)
File c:\Program Files\Python312\Lib\http\client.py:1331, in HTTPConnection.
 ⇔endheaders(self, message_body, encode_chunked)
           raise CannotSendHeader()
-> 1331 self._send_output(message_body, encode_chunked=encode_chunked)
File c:\Program Files\Python312\Lib\http\client.py:1091, in HTTPConnection.
 1090 del self. buffer[:]
-> 1091 self.send(msg)
   1093 if message_body is not None:
   1094
   1095
           # create a consistent interface to message_body
File c:\Program Files\Python312\Lib\http\client.py:1035, in HTTPConnection.
 ⇔send(self, data)
   1034 if self.auto_open:
           self.connect()
-> 1035
   1036 else:
File c:\Program Files\Python312\Lib\http\client.py:1470, in HTTPSConnection.
 ⇔connect(self)
   1468 "Connect to a host on a given (SSL) port."
-> 1470 super().connect()
   1472 if self. tunnel host:
File c:\Program Files\Python312\Lib\http\client.py:1001, in HTTPConnection.
 ⇔connect(self)
   1000 sys.audit("http.client.connect", self, self.host, self.port)
-> 1001 self.sock = self._create_connection(
           (self.host,self.port), self.timeout, self.source_address)
   1003 # Might fail in OSs that don't implement TCP_NODELAY
File c:\Program Files\Python312\Lib\socket.py:829, in create_connection(address
 →timeout, source_address, all_errors)
   828 exceptions = []
--> 829 for res in getaddrinfo(host, port, 0, SOCK_STREAM):
           af, socktype, proto, canonname, sa = res
```

```
File c:\Program Files\Python312\Lib\socket.py:964, in getaddrinfo(host, port, __
 ⇔family, type, proto, flags)
    963 addrlist = []
--> 964 for res in _socket.getaddrinfo(host, port, family, type, proto, flags):
            af, socktype, proto, canonname, sa = res
gaierror: [Errno 11001] getaddrinfo failed
During handling of the above exception, another exception occurred:
URLError
                                          Traceback (most recent call last)
Cell In[104], line 1
---> 1 sns.get_dataset_names()
File c:\Program Files\Python312\Lib\site-packages\seaborn\utils.py:499, in_
 →get_dataset_names()
    493 def get_dataset_names():
    494
            """Report available example datasets, useful for reporting issues.
    495
    496
            Requires an internet connection.
    497
    498
--> 499
            with urlopen(DATASET_NAMES_URL) as resp:
    500
                txt = resp.read()
            dataset_names = [name.strip() for name in txt.decode().split("\n")]
    502
File c:\Program Files\Python312\Lib\urllib\request.py:215, in urlopen(url, data)
 ⇔timeout, cafile, capath, cadefault, context)
    213 else:
    214
            opener = _opener
--> 215 return opener.open(url, data, timeout)
File c:\Program Files\Python312\Lib\urllib\request.py:515, in OpenerDirector.
 ⇔open(self, fullurl, data, timeout)
            req = meth(req)
    512
    514 sys.audit('urllib.Request', req.full_url, req.data, req.headers, req.
 →get method())
--> 515 response = self._open(req, data)
    517 # post-process response
    518 meth_name = protocol+"_response"
File c:\Program Files\Python312\Lib\urllib\request.py:532, in OpenerDirector.
 → open(self, req, data)
    529
            return result
    531 protocol = req.type
--> 532 result = self._call_chain(self.handle_open, protocol, protocol +
    533
                                   '_open', req)
```

```
534 if result:
    535
           return result
File c:\Program Files\Python312\Lib\urllib\request.py:492, in OpenerDirector.

    call chain(self, chain, kind, meth name, *args)

    490 for handler in handlers:
            func = getattr(handler, meth name)
            result = func(*args)
--> 492
    493
            if result is not None:
    494
                return result
File c:\Program Files\Python312\Lib\urllib\request.py:1392, in HTTPSHandler.
 →https_open(self, req)
   1391 def https_open(self, req):
-> 1392
            return self.do_open(http.client.HTTPSConnection, req,
                                context=self._context)
   1393
File c:\Program Files\Python312\Lib\urllib\request.py:1347, in_
 AbstractHTTPHandler.do_open(self, http_class, req, **http_conn_args)
                h.request(req.get method(), req.selector, req.data, headers,
   1344
                          encode_chunked=req.has_header('Transfer-encoding'))
   1345
   1346
            except OSError as err: # timeout error
                raise URLError(err)
-> 1347
   1348
            r = h.getresponse()
   1349 except:
URLError: <urlopen error [Errno 11001] getaddrinfo failed>
```