Learn_Seaborn

January 18, 2023

0.1 Setup

[2]:
```
import numpy as np
import pandas as pd
import seaborn as sns
import matplotlib.pyplot as plt

%reload_ext autoreload
%autoreload 2
%matplotlib inline
```

0.2 Import Data

[3]:
```
# df = pd.read_csv()
crash_df = sns.load_dataset('car_crashes')
crash_df.head(10)
```

[3]:

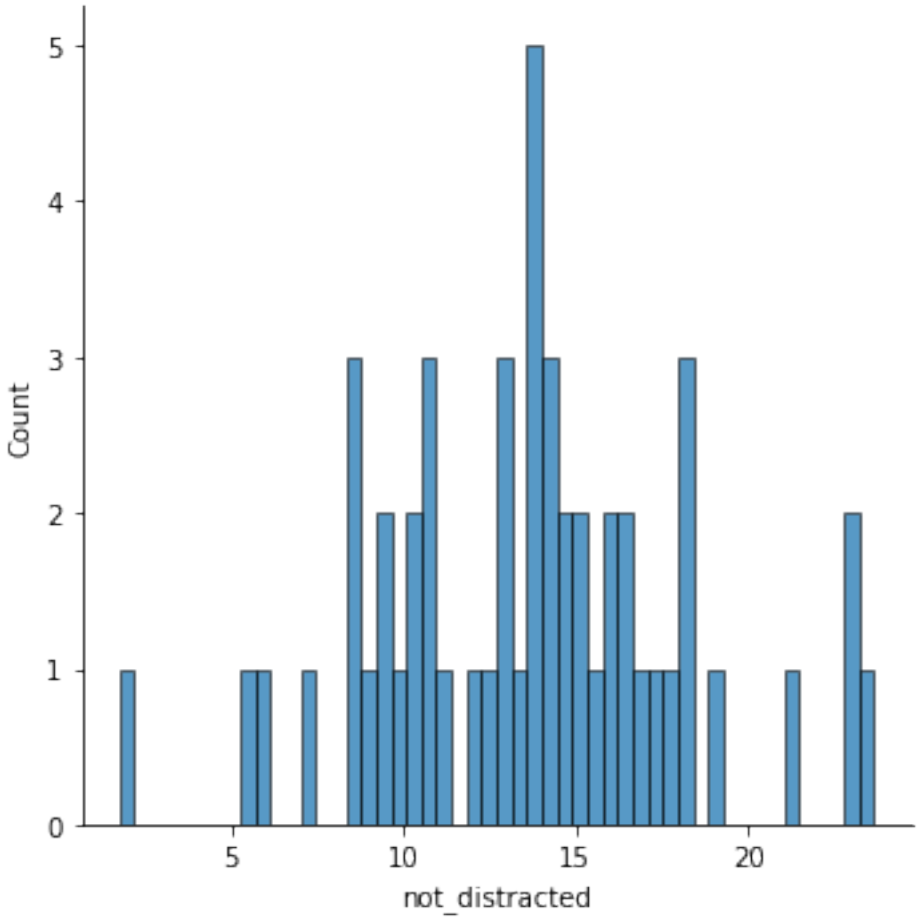
	total	speeding	alcohol	not_distracted	no_previous	ins_premium	ins_losses	abbrev
0	18.8	7.332	5.640	18.048	15.040	784.55	145.08	AL
1	18.1	7.421	4.525	16.290	17.014	1053.48	133.93	AK
2	18.6	6.510	5.208	15.624	17.856	899.47	110.35	AZ
3	22.4	4.032	5.824	21.056	21.280	827.34	142.39	AR
4	12.0	4.200	3.360	10.920	10.680	878.41	165.63	CA
5	13.6	5.032	3.808	10.744	12.920	835.50	139.91	CO
6	10.8	4.968	3.888	9.396	8.856	1068.73	167.02	CT
7	16.2	6.156	4.860	14.094	16.038	1137.87	151.48	DE
8	5.9	2.006	1.593	5.900	5.900	1273.89		
9	17.9	3.759	5.191	16.468	16.826	1160.13		

```
8       136.05      DC
9       144.18      FL
```

1 Distribution Plots

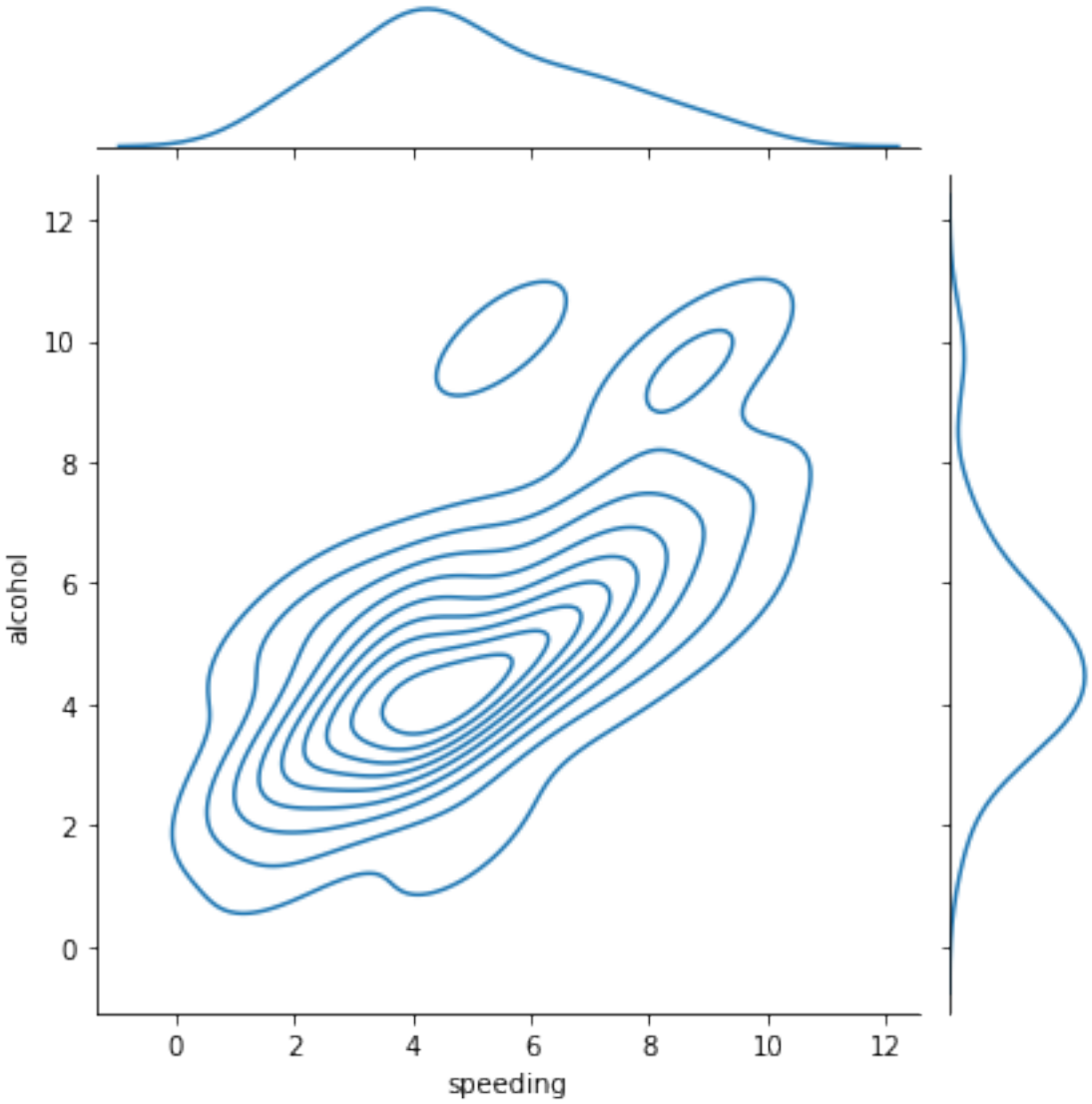
1.1 Distribution Plot

[4]:
```
# sns.distplot(crash_df['not_distracted'])
sns.displot(crash_df['not_distracted'],kde=False,bins=50)
plt.show()
```



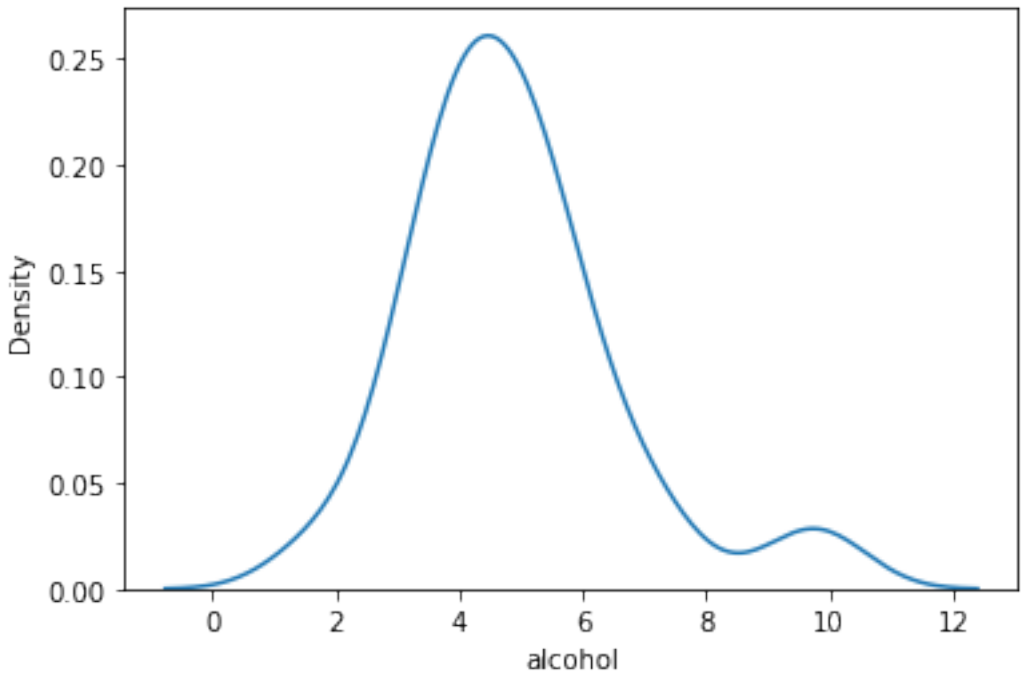
1.2 Joint Plot

[5]:
```
# compare two distribution and plot a scatter plot by default
sns.jointplot(x='speeding',y='alcohol',data=crash_df,kind='kde') # king =␣
↪['kde','hex','reg']
plt.show()
```



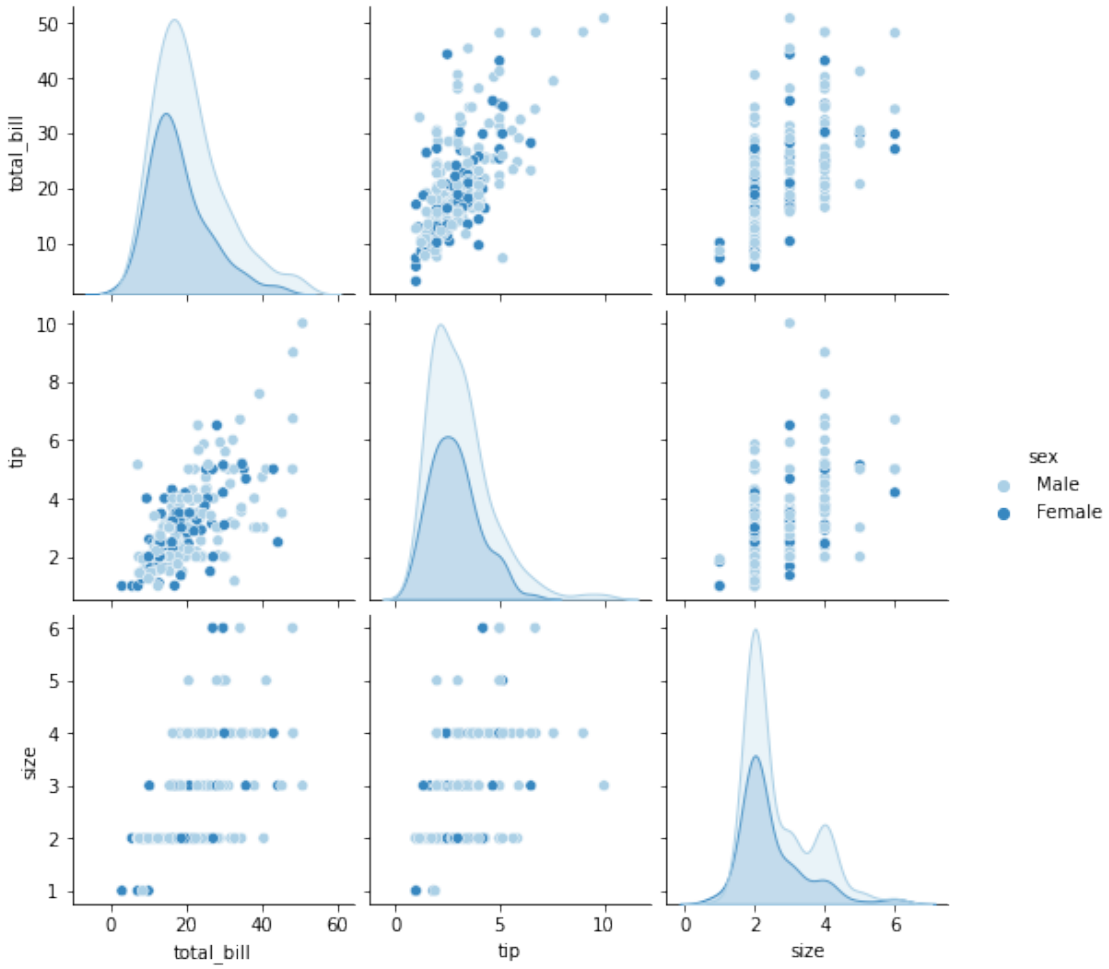
1.3 KDE Plot

[6]:
```
sns.kdeplot(crash_df['alcohol'])
plt.show()
```



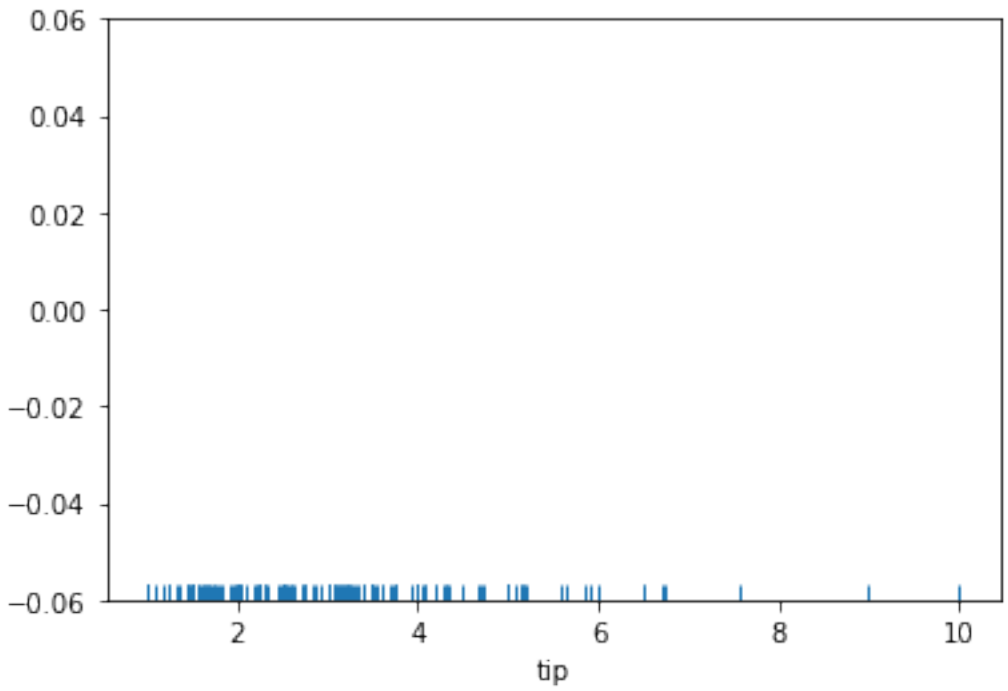
1.4 Pair Plot

[7]:
```
# It plot the the relationship across entire dataframe and  numerical values
# sns.pairplot(crash_df)
tips_df = sns.load_dataset('tips')
sns.pairplot(tips_df,hue='sex',palette='Blues') # hue :-> with hue we can pass␣
 ↪the categorial data and chart will be colorised based on that data
plt.show()
```



1.5 Rug Plot

[8]:
```
# It plots the a sigle column of datapoints in a dataframe as sticks on the x
↪axis, it show more dense number of lines where the amount is more common
sns.rugplot(tips_df['tip'])
plt.show()
```

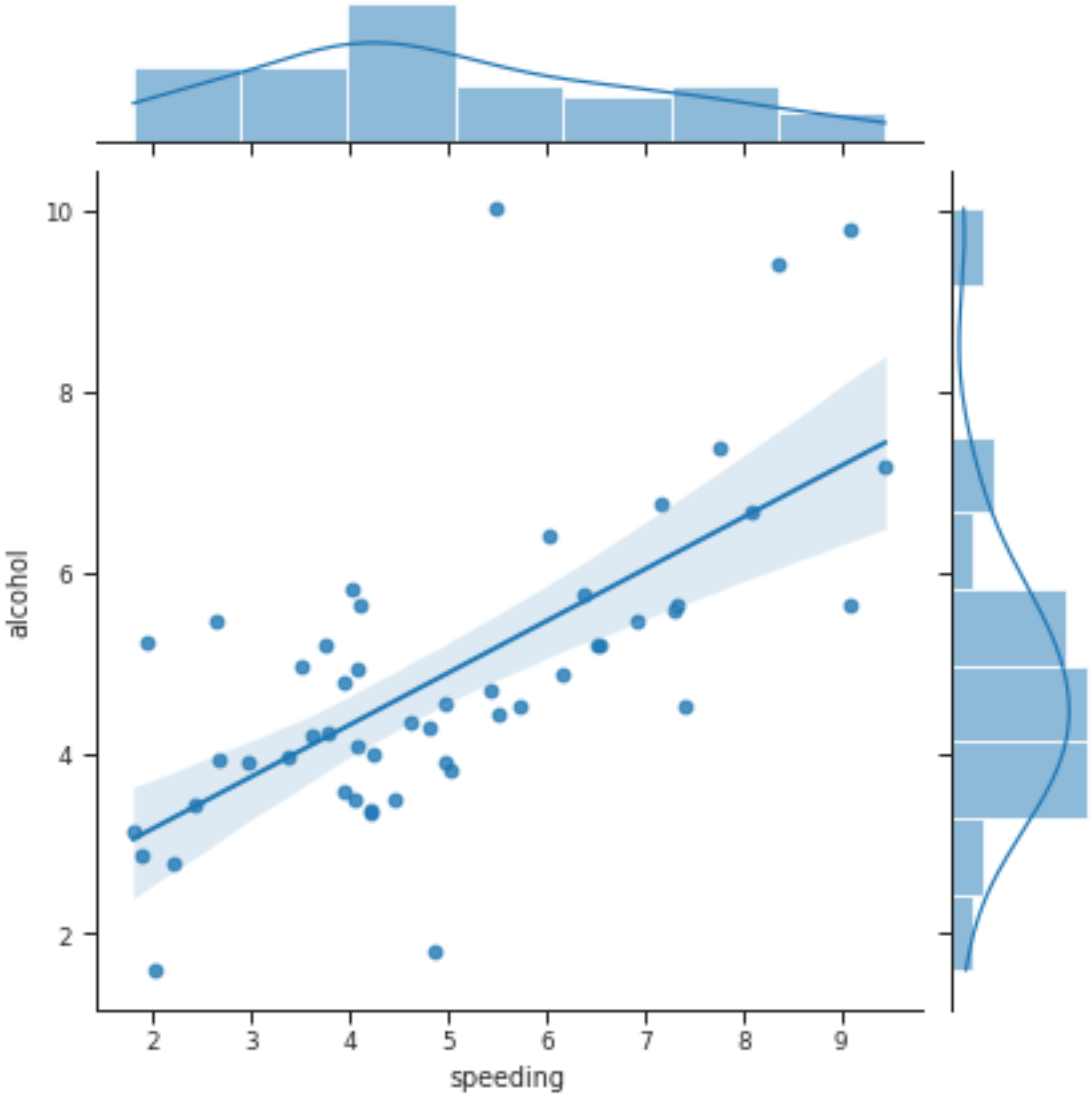


1.6 Styling

[9]:
```
sns.set_style('ticks') # style= ['white','whitegrid',dark,ticks,'darkgrid']
sns.set_context('paper',font_scale=1.0) # [paper,talk,poster]
plt.figure(figsize=(2,2))
sns.jointplot(x='speeding',y='alcohol',data=crash_df,kind='reg')

# sns.despine(left=True,bottom=True) # [right, top]
plt.show()
```

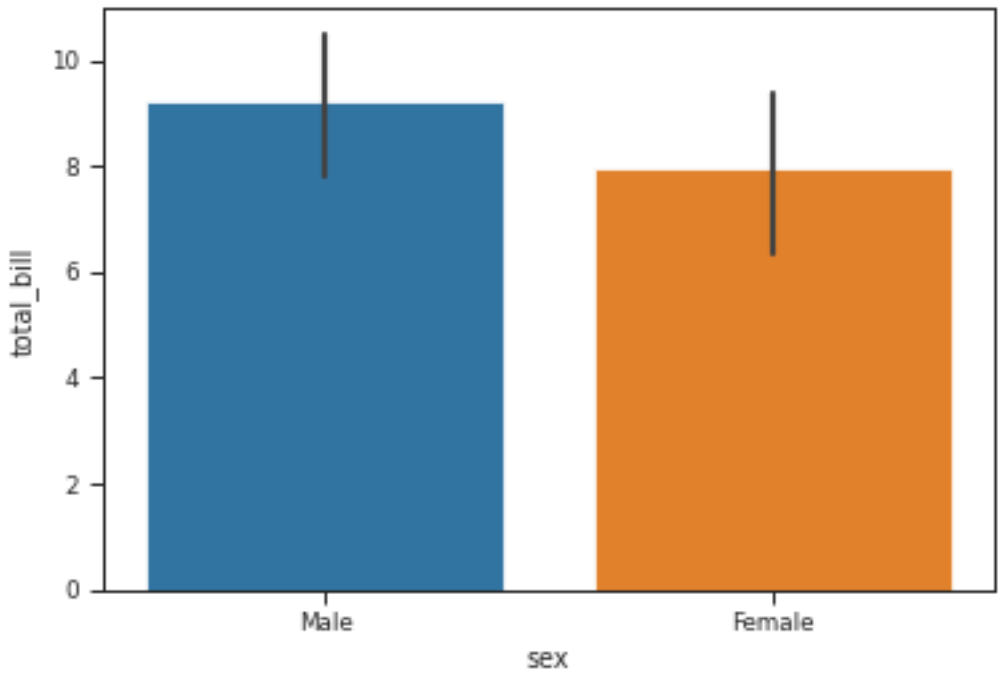
```
<Figure size 144x144 with 0 Axes>
```



2 Categorial Plots

2.1 Bar Plot

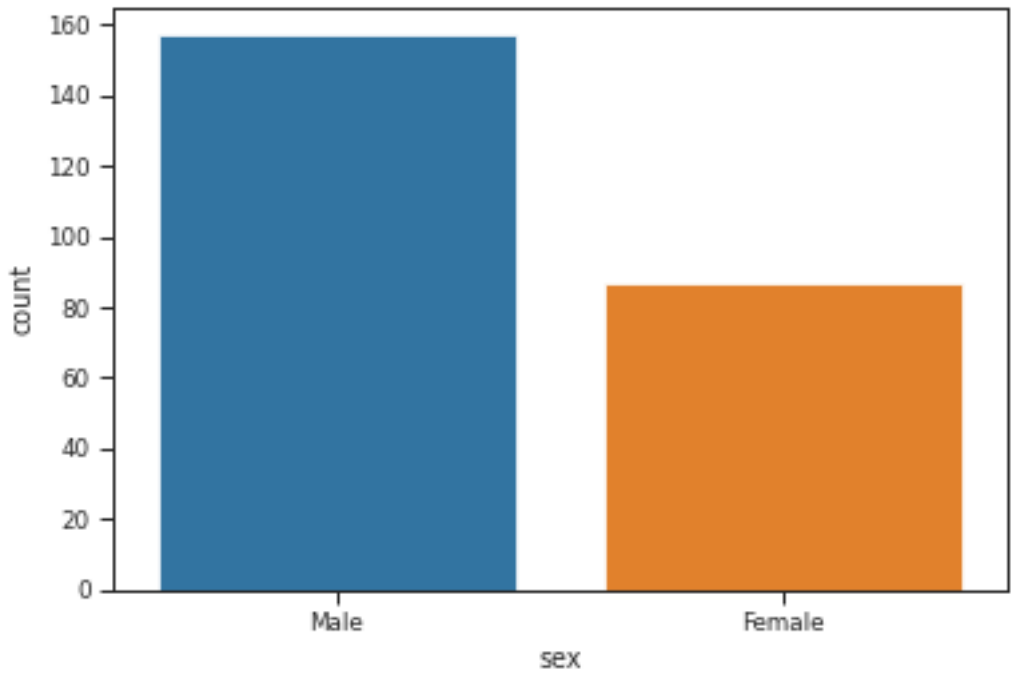
[10]:
```
sns.barplot(x='sex',y='total_bill',data=tips_df,estimator=np.std) #␣
  ↪estimator=[median,mean(default),np.std(StandardDeviation), np.
  ↪var(varience),np.cov(coverience),custom]
plt.show()
```



2.2 Count Plot

[11]:
```
# kind of bar plot but estimator is just count the number of occurances
sns.countplot(x='sex',data=tips_df)
```

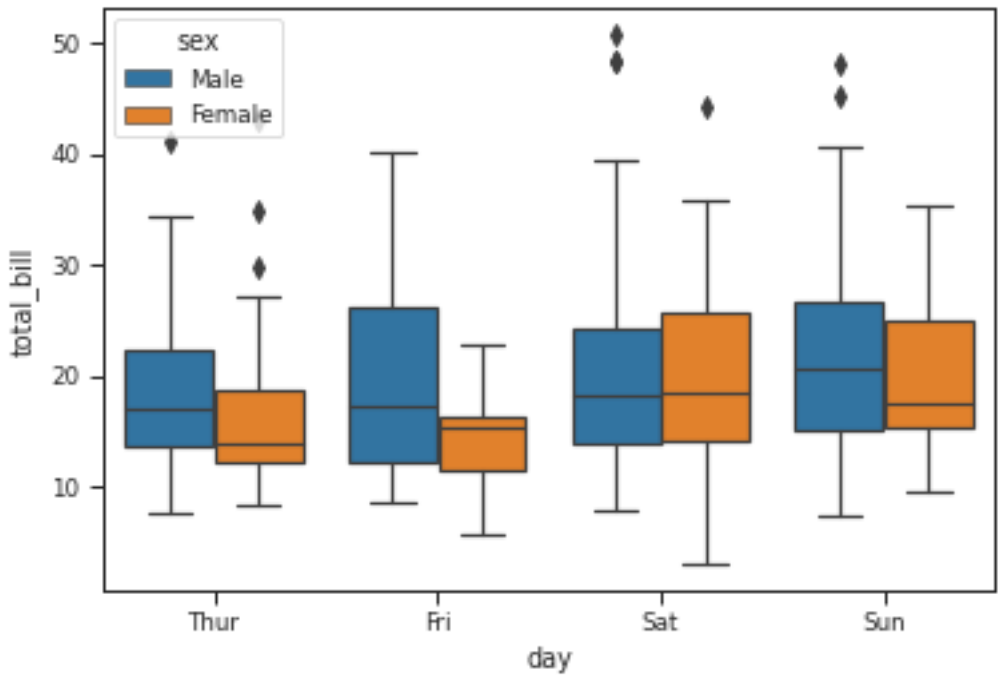
[11]:
```
<matplotlib.axes._subplots.AxesSubplot at 0x7ff4add963a0>
```



2.3 Box Plot

[12]:
```
# A box plot is going to compare the different variables qurtile of the data
sns.boxplot(x='day',y='total_bill',data=tips_df,hue='sex')
```

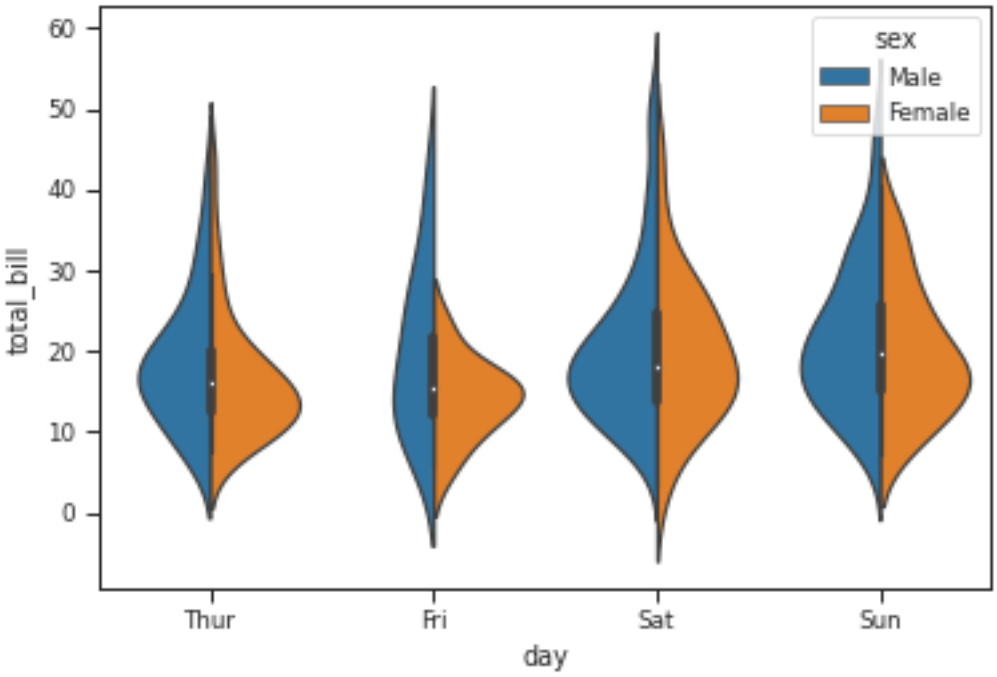
[12]:
```
<matplotlib.axes._subplots.AxesSubplot at 0x7ff4add9f460>
```



2.4 Violin Plot

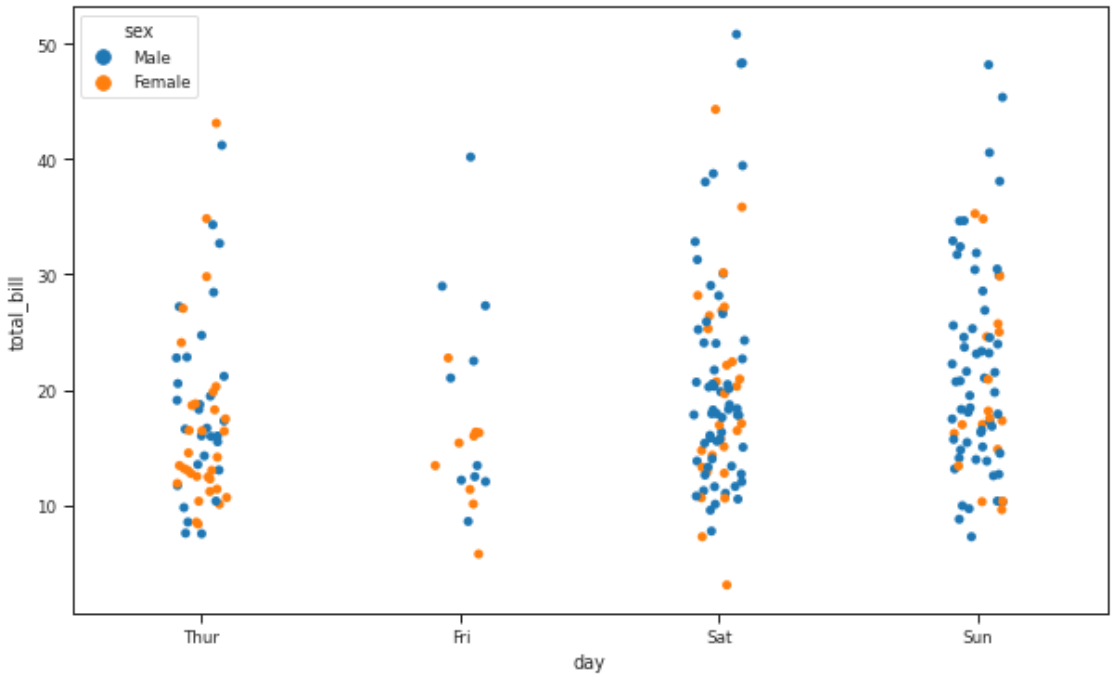
[13]:
```
# it is combination of box plot and a kde plot
# a violin plot uses the kde estimation of data points
sns.violinplot(x='day',y='total_bill',data=tips_df,hue='sex',split=True)
```

[13]: <matplotlib.axes._subplots.AxesSubplot at 0x7ff4b1b82280>



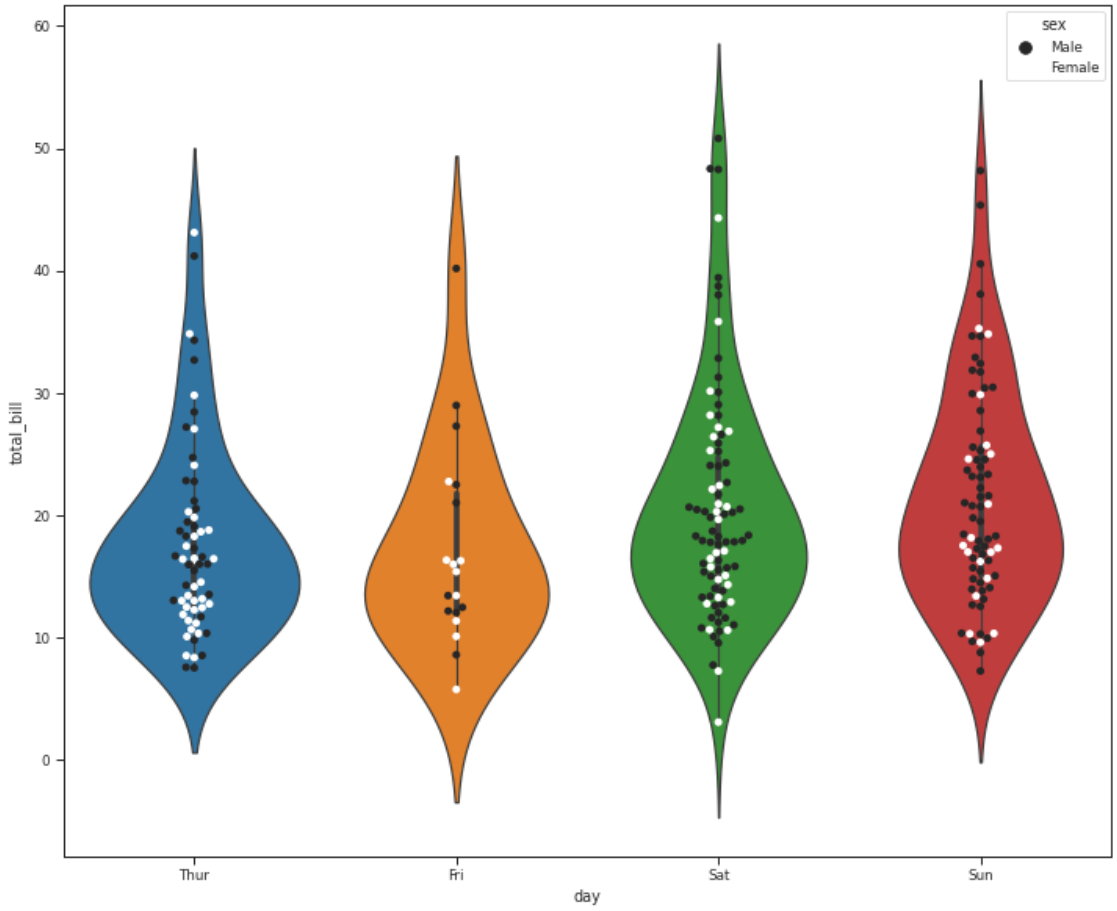
2.5 Strip Plot

[14]:
```
# strip plot draws the scatter plot represents all data points where one␣
 ↪variable is categorical
plt.figure(figsize=(10,6))
sns.stripplot(x='day',y='total_bill',data=tips_df,jitter=True,hue='sex') #␣
 ↪,dodge=True :-> seprates the data
plt.show()
```



2.6 Swarm Plot

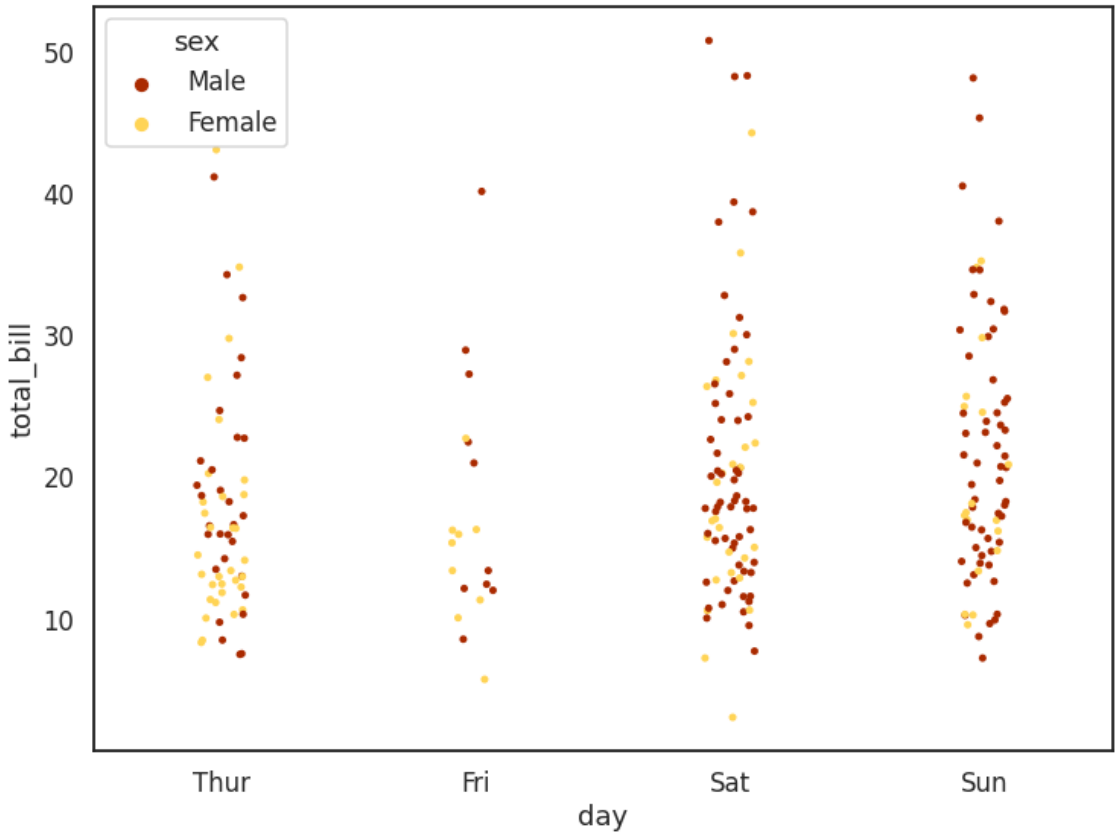
[15]:
```
# It is a combination of violin and a strip plot
plt.figure(figsize=(12,10))
sns.violinplot(x='day',y='total_bill',data=tips_df)
sns.swarmplot(x='day',y='total_bill',data=tips_df,color='white',hue='sex')
plt.show()
```



2.7 Palettes

- by setting the palettes you can change the default coloring of the any plot
- to see more detail about the color palette click here

[16]:
```
# by setting the palettes you can change the default coloring of the any plot
plt.figure(figsize=(12,9))
sns.set_style('white')
sns.set_context('talk')
sns.stripplot(x='day',y='total_bill',data=tips_df,hue='sex',palette='afmhot')
# plt.legend(loc=1)
plt.show()
```



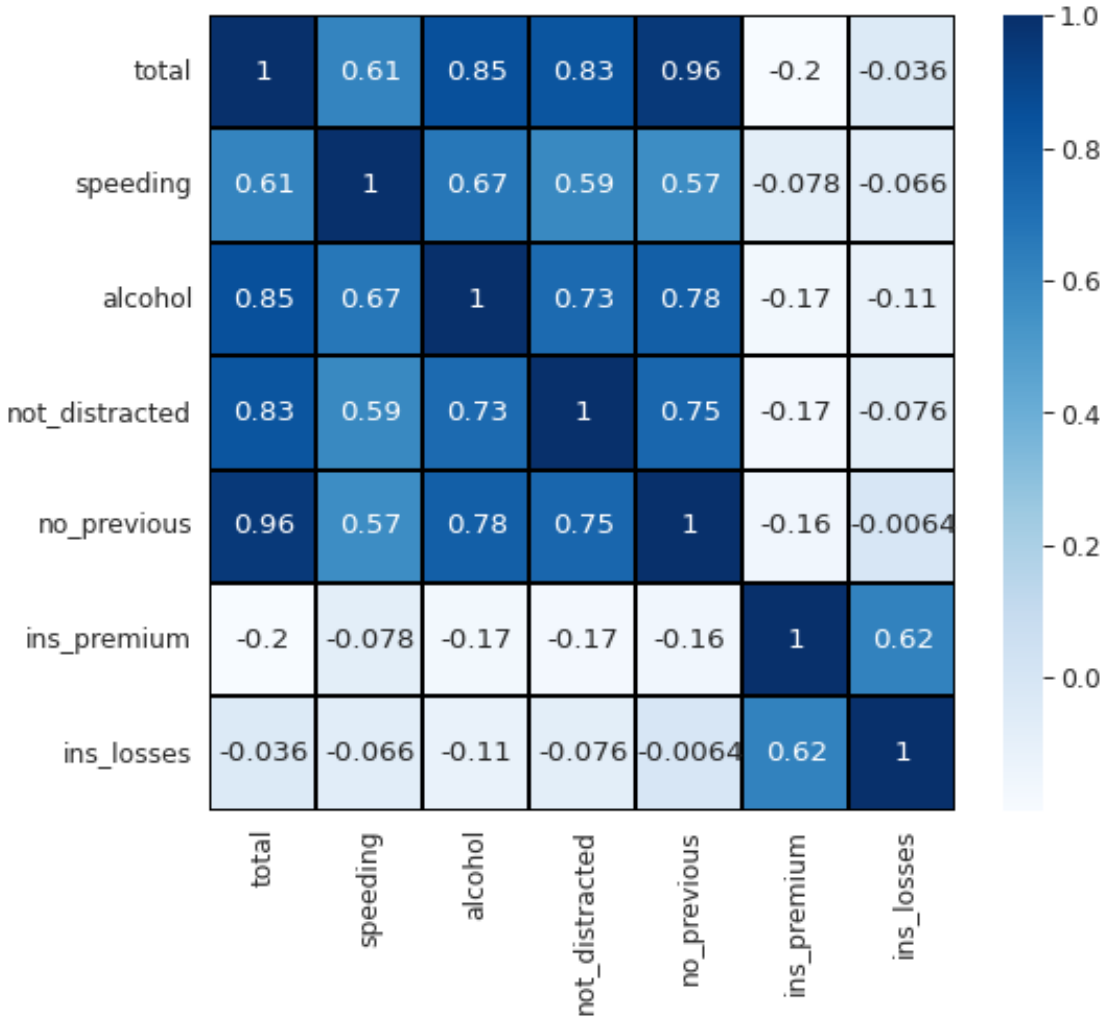
3 Matrix Plot

3.1 Heatmaps

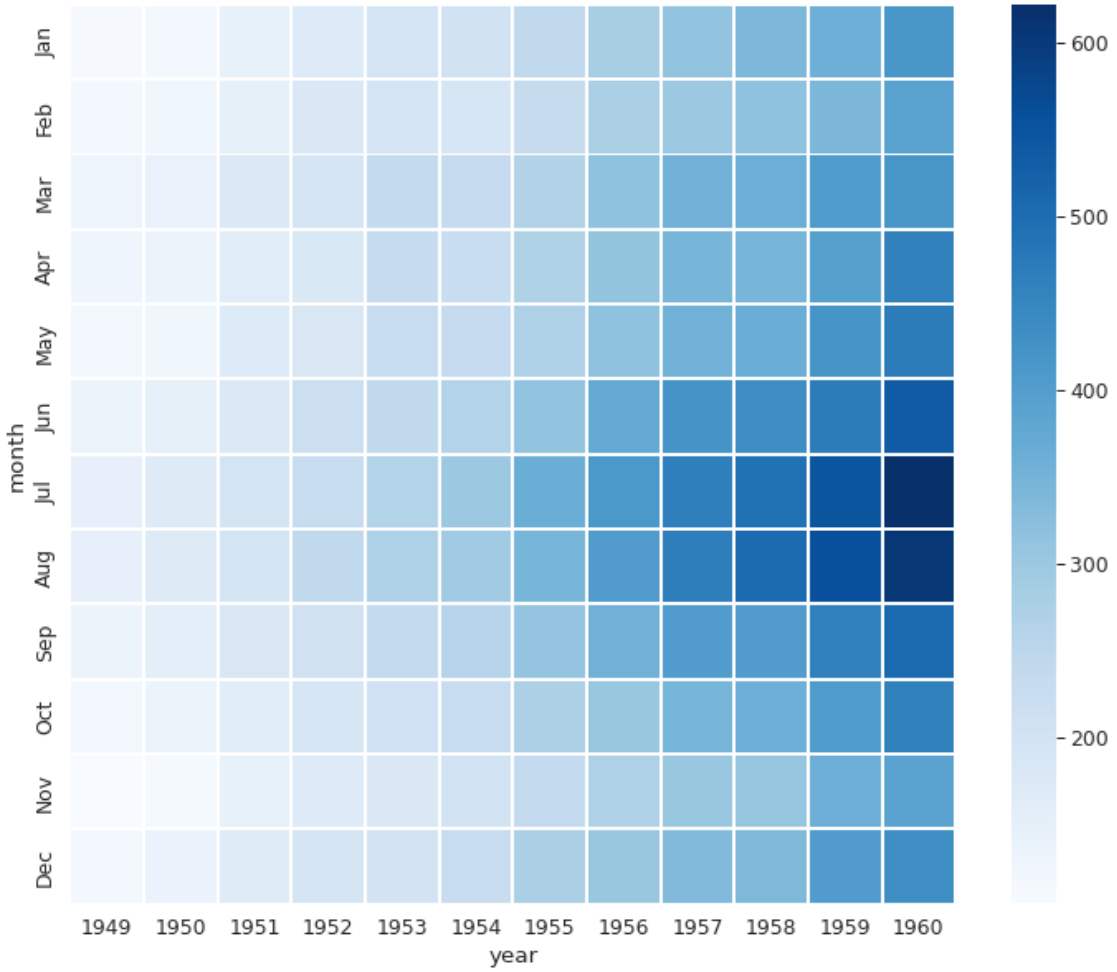
[17]:
```
#
plt.figure(figsize=(8,7))
sns.set_context('paper',font_scale=1.4)

crash_mx = crash_df.corr()

sns.heatmap(crash_mx,annot=True,cmap='Blues',linewidth=1,linecolor='black')
plt.show()
```



[18]:
```
plt.figure(figsize=(12,10))
flights_df = sns.load_dataset('flights')
flights = flights_df.
  ↪pivot_table(index='month',columns='year',values='passengers')
sns.heatmap(flights,cmap='Blues',linecolor='white',linewidth=1)
plt.show()
```

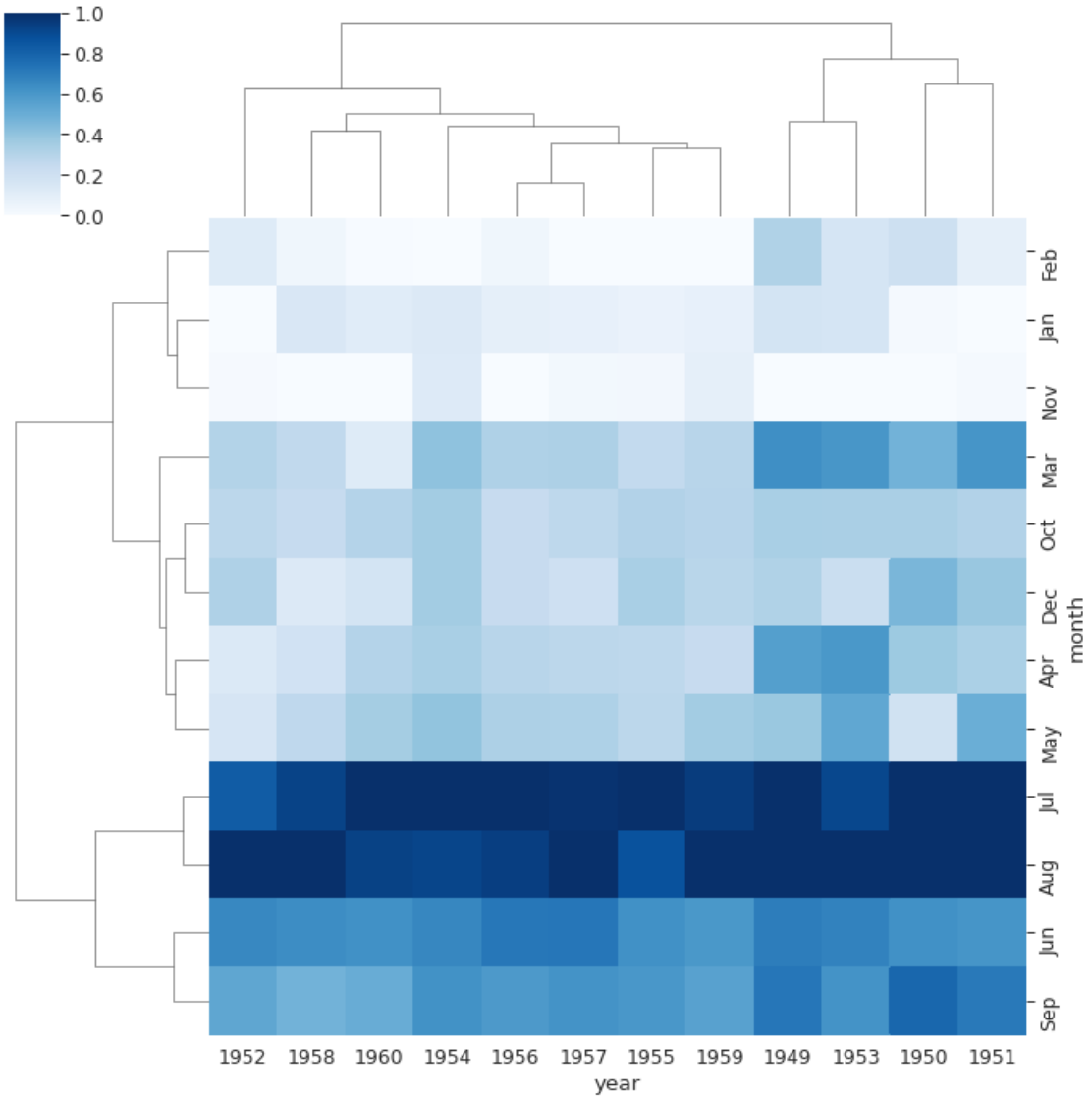


3.2 Cluster Map

[19]:
```
# it gives the cluster data by finding the closest distance between various␣
 ↪points
iris = sns.load_dataset('iris')
# species = iris.pop('species')
# sns.clustermap(iris)
```

[20]:
```
sns.clustermap(flights,cmap='Blues',standard_scale=1)
```

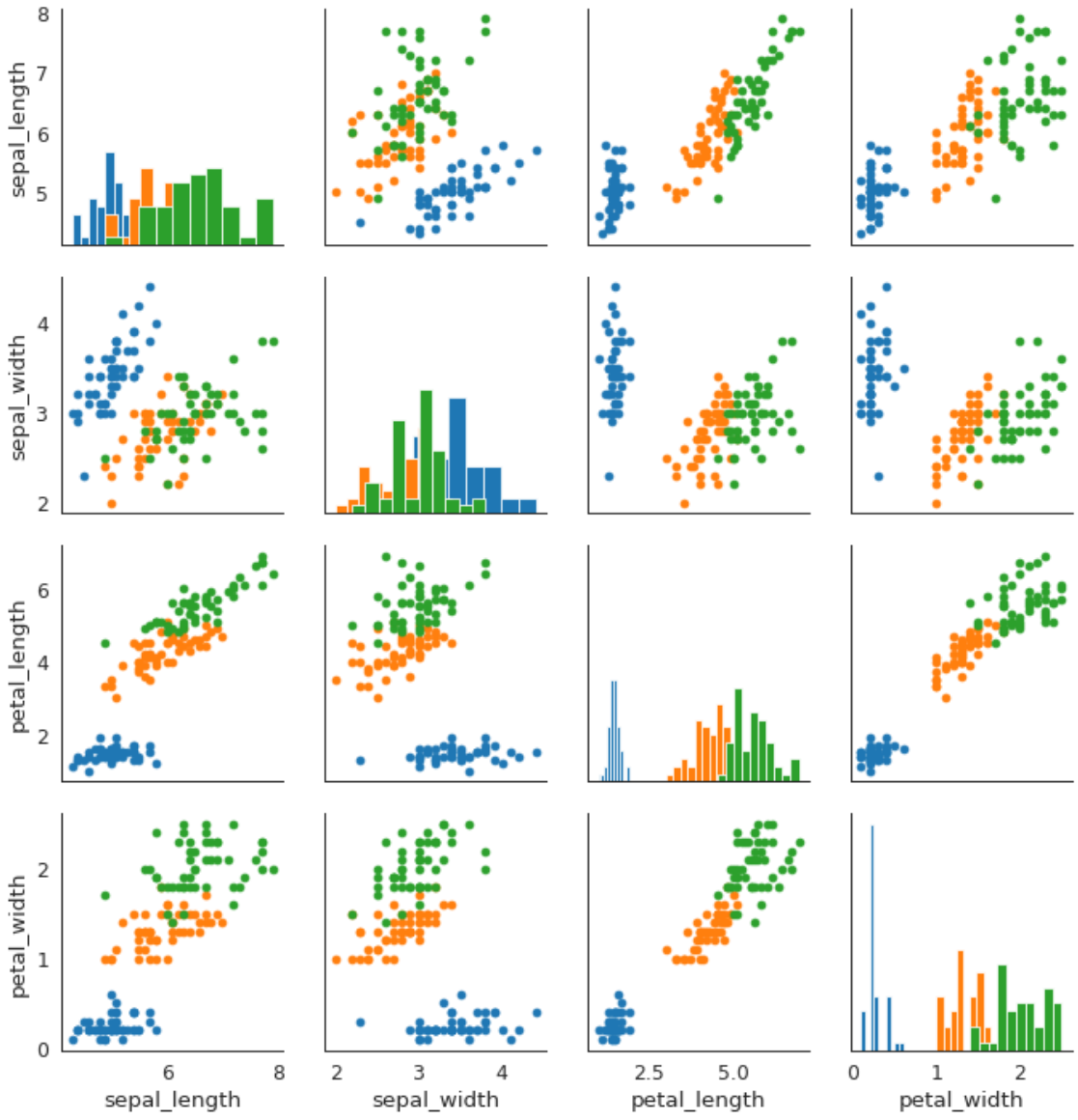
[20]:
```
<seaborn.matrix.ClusterGrid at 0x7ff4adc3b670>
```



3.3 Pair Grid

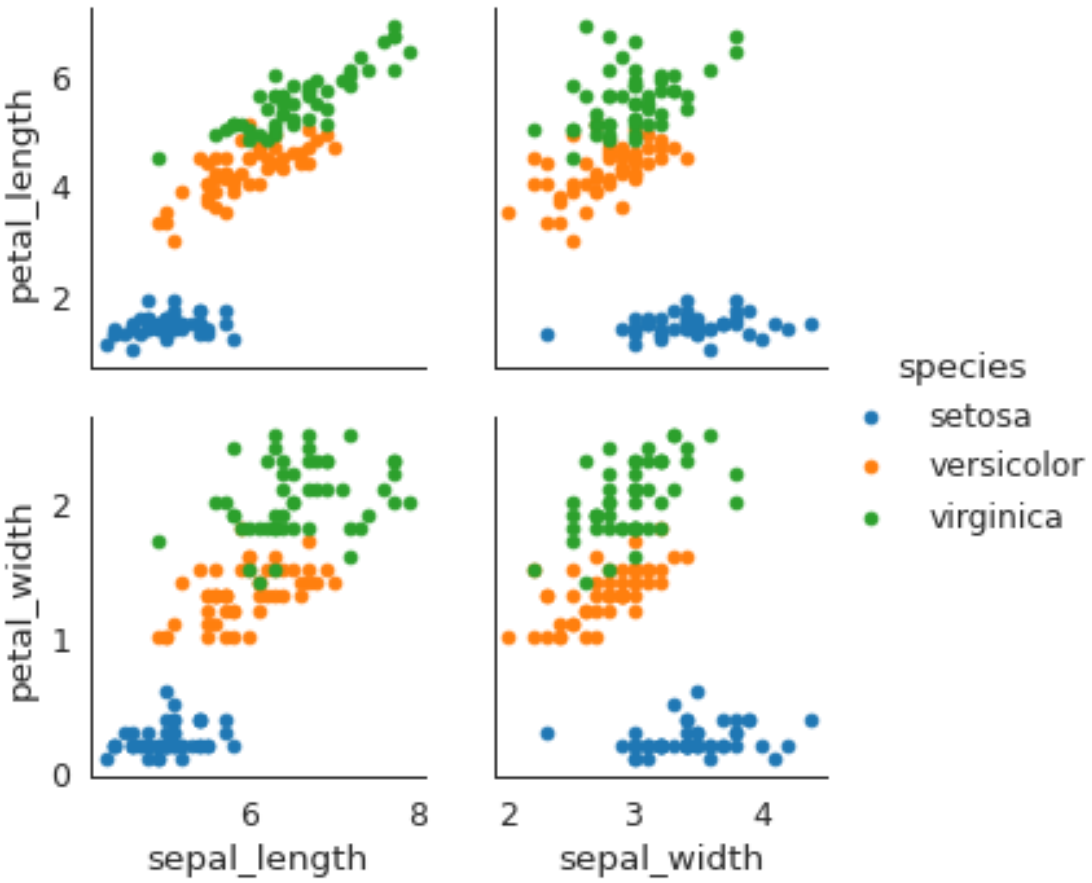
[21]:
```
# to plot or change the default plots in the different position use the
  ↪'shift+tab' [iris_g.map_upper,.map_lower,.map_offdiag,...]
iris_g = sns.PairGrid(iris,hue='species')
# iris_g.map(plt.scatter)
iris_g.map_diag(plt.hist)
iris_g.map_offdiag(plt.scatter)
```

[21]: <seaborn.axisgrid.PairGrid at 0x7ff4b1c02c70>



[22]:
```
# for our custom grid plot we can define the x and y variables with dataset
iris_g = sns.
 ↪PairGrid(iris,hue='species',x_vars=['sepal_length','sepal_width'],y_vars=['petal_length','p
iris_g.map(plt.scatter)
iris_g.add_legend()
```

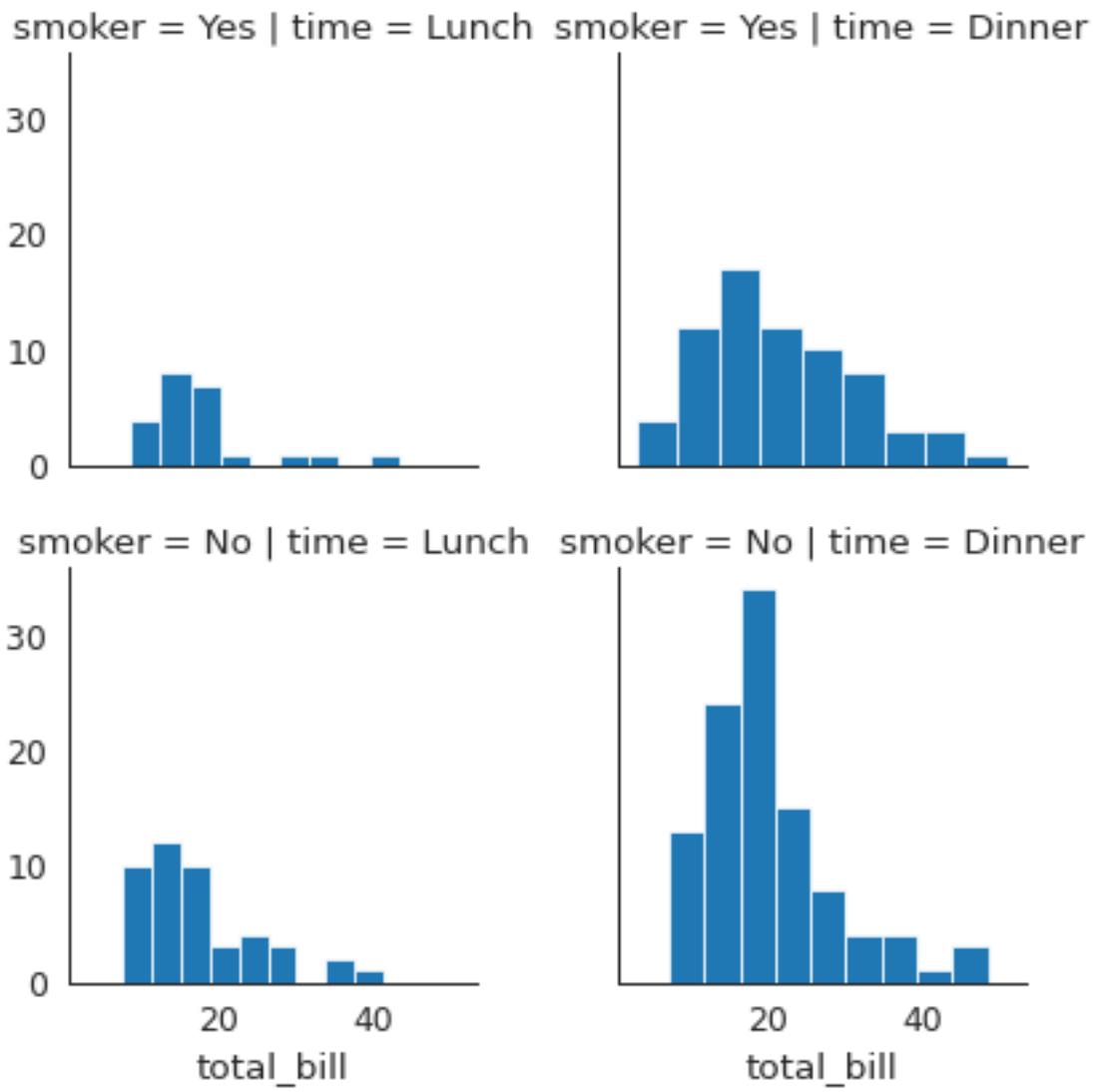
[22]: <seaborn.axisgrid.PairGrid at 0x7ff4addad880>



3.4 Facet Grid

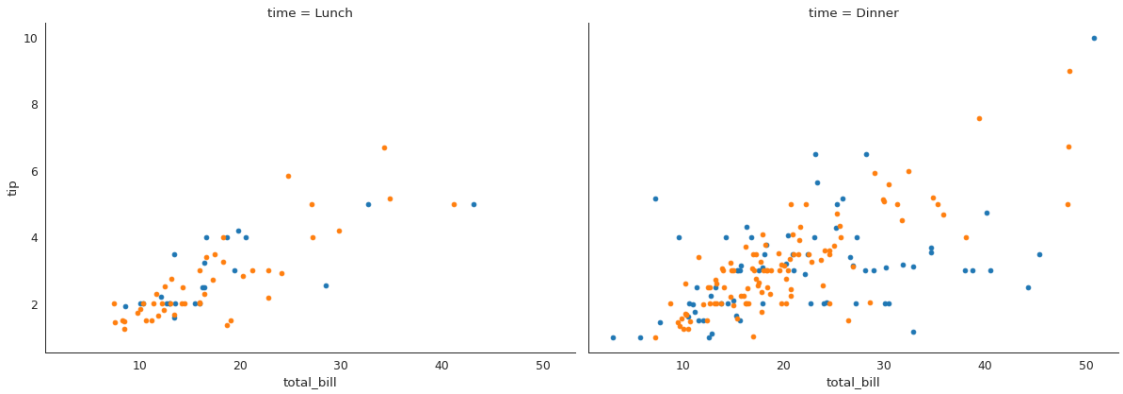
[23]:
```
import seaborn as sns
# this can plot multiple plot in a grid and in this we can define the columns␣
 ↪and rows
tips_fg = sns.FacetGrid(tips_df,col='time',row='smoker') # to draw the plot fig␣
 ↪with only axis no data in it
# sns.set_context('paper',font_scale=1.2)
tips_fg.map(plt.hist, 'total_bill',bins=9) # ['time','smoder',]
# tips_fg.map(plt.scatter, 'total_bill','tip')
```

[23]:
```
<seaborn.axisgrid.FacetGrid at 0x7ff4ad7d3b20>
```



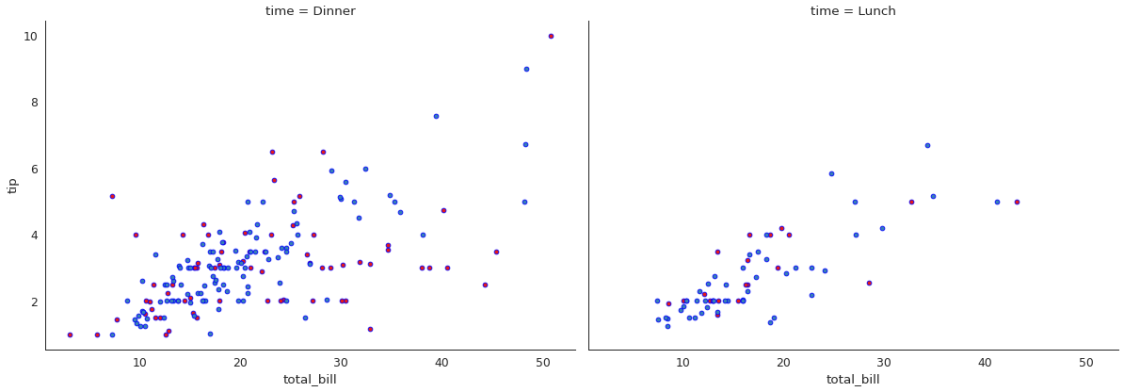
[24]:
```
tips_fg1 = sns.FacetGrid(tips_df,col='time',hue='smoker',height=6,aspect=1.4)
tips_fg1.map(plt.scatter,'total_bill','tip')
```

[24]: <seaborn.axisgrid.FacetGrid at 0x7ff4adf1a7c0>



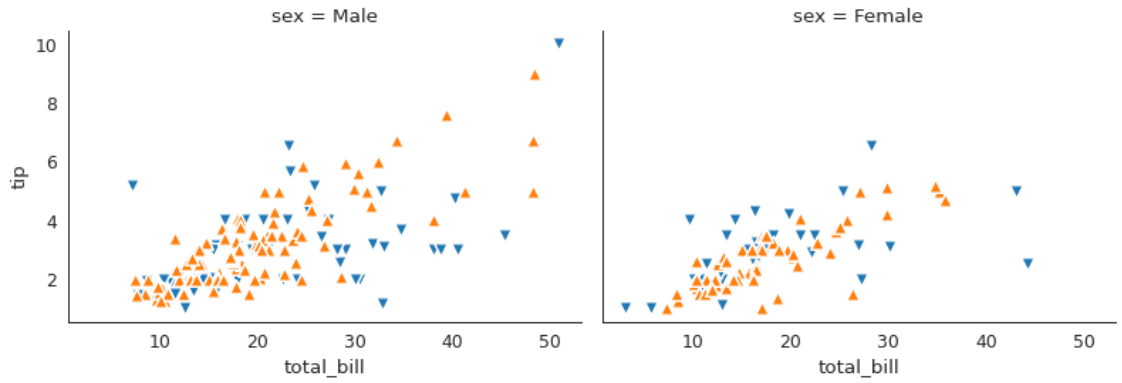
[25]:
```
tips_fg2 = sns.FacetGrid(tips_df,col='time',hue='smoker',height=6,aspect=1.
 ↪4,col_order=['Dinner','Lunch'],palette='Set1')
tips_fg2.map(plt.scatter,'total_bill','tip',edgecolor='b')
```

[25]: <seaborn.axisgrid.FacetGrid at 0x7ff4ad5e6e80>



[26]:
```
kws = dict(s=90,linewidth=1.5,edgecolor='w')
tips_fg2 = sns.FacetGrid(tips_df,col='sex',hue='smoker',height=4,aspect=1.
 ↪4,hue_order=['Yes','No'],hue_kws=dict(marker=['v','^']))
tips_fg2.map(plt.scatter,'total_bill','tip',**kws)
```

[26]: <seaborn.axisgrid.FacetGrid at 0x7ff4ad553df0>

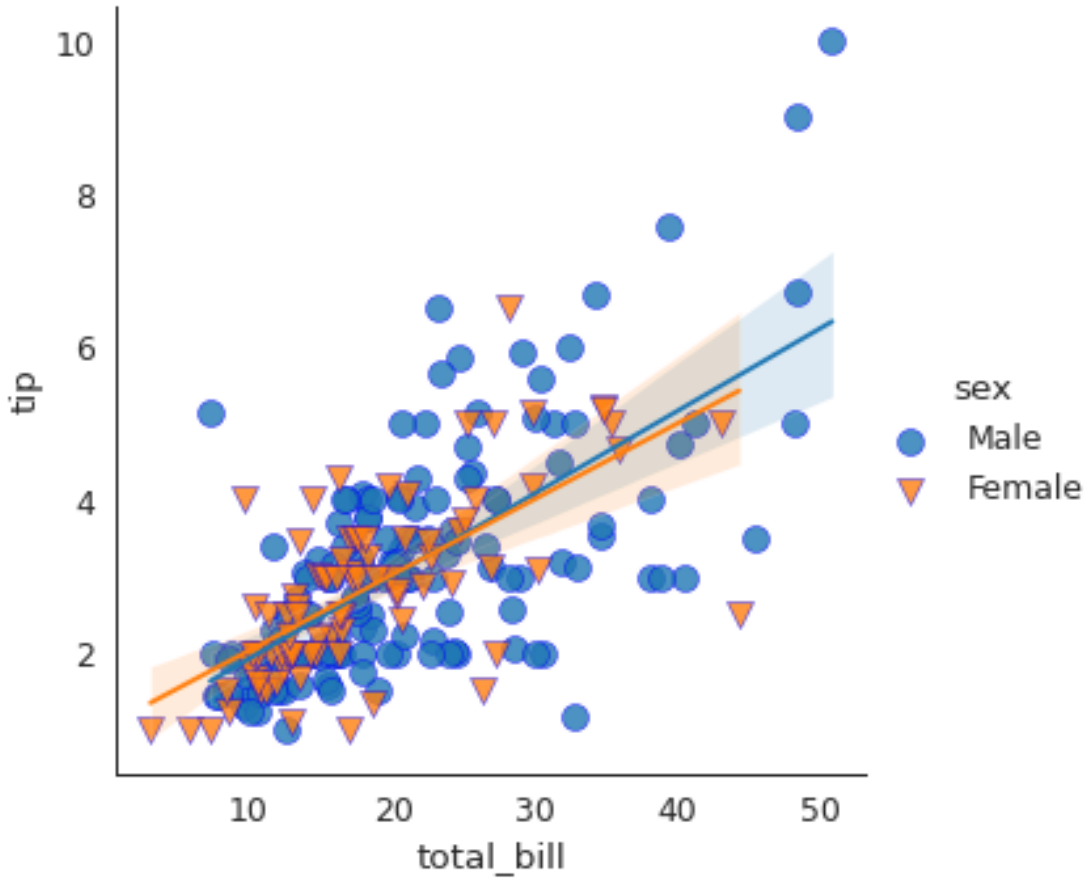


3.5 Regression Plot

[27]:
```
plt.figure(figsize=(8,6))
sns.set_context('paper',font_scale=1.4)
sns.
 ↪lmplot(x='total_bill',y='tip',hue='sex',data=tips_df,markers=['o','v'],scatter_kws={'s':
 ↪100,'linewidth':0.5,'edgecolor':'b'})
```

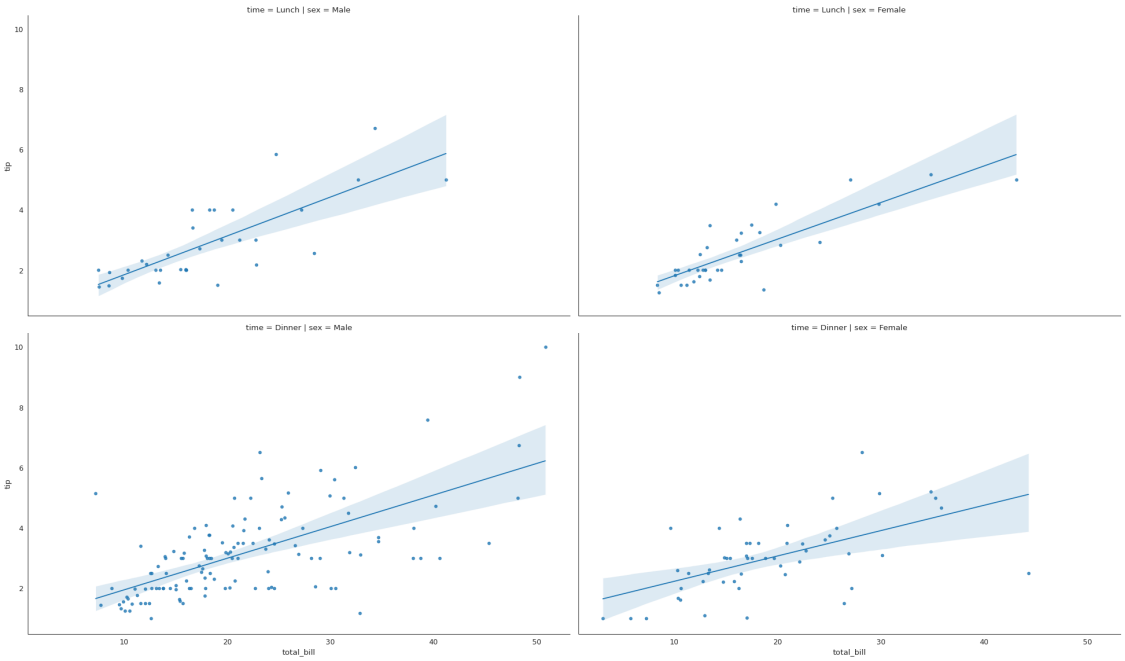
[27]:
```
<seaborn.axisgrid.FacetGrid at 0x7ff4ad5ba460>

<Figure size 576x432 with 0 Axes>
```



[28]:
```
sns.
↪lmplot(x='total_bill',y='tip',col='sex',row='time',data=tips_df,height=8,aspect=1.
↪7)
```

[28]: <seaborn.axisgrid.FacetGrid at 0x7ff4ad4743d0>


time = Lunch | sex = Male
time = Lunch | sex = Female
time = Dinner | sex = Male
time = Dinner | sex = Female
tip
total_bill
2
4
6
8
10
20
30
40
50