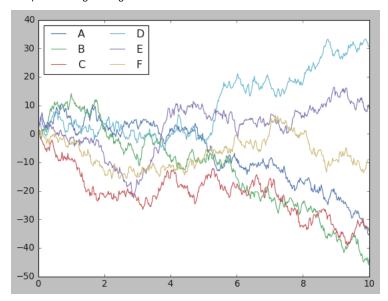
```
In [93]: import matplotlib.pyplot as plt
plt.style.use("classic")
%matplotlib inline
import numpy as np
import pandas as pd
In [94]: rng=np.random.RandomState(0)
x=np.linspace(0,10,500)
y=np.cumsum(rng.randn(500,6),0)
```

Out[94]: <matplotlib.legend.Legend at 0x272162c05e0>

plt.legend("ABCDEF",ncol=2,loc="upper left")

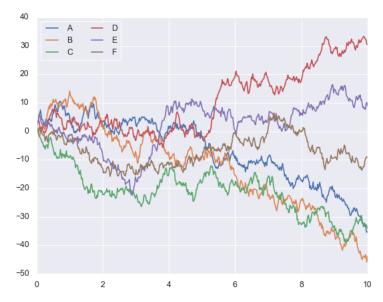
plt.plot(x,y)



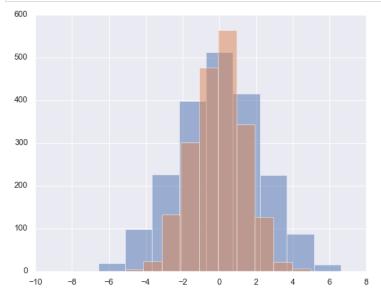
```
In [95]: import seaborn as sns
sns.set()
```

```
In [96]: #plotting the same graph using seaborn
plt.plot(x,y)
plt.legend("ABCDEF",ncol=2,loc="upper left")
```

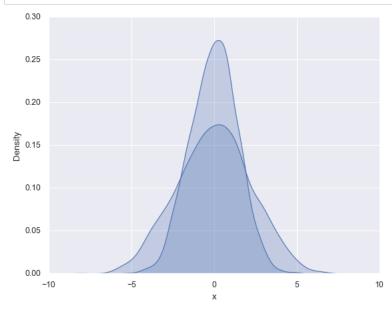
Out[96]: <matplotlib.legend.Legend at 0x27214d15ee0>



```
In [97]: data=np.random.multivariate_normal([0,0],[[5, 2], [2, 2]],size=2000)
data=pd.DataFrame(data,columns=["x","y"])
for col in "xy":
    plt.hist(data[col],alpha=0.5)
```



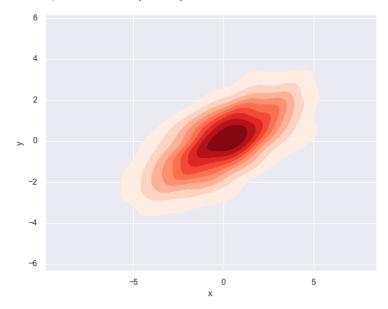
In [98]: #to get a smooth distribution curve we kernal density estimation in seaborn
for col in "xy":
 sns.kdeplot(data[col],shade=True)



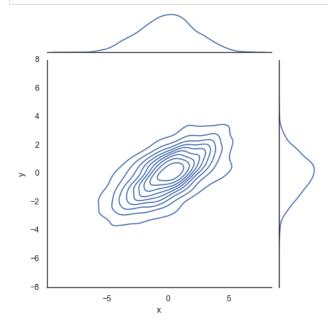
In [99]: #histograms and kde can be combined using distplot however displot is deprecated now

```
In [100]: #passing data[columns] in kdeplot we get 2D data
sns.kdeplot(x=data["x"],y=data["y"],shade=True,cmap="Reds")
```

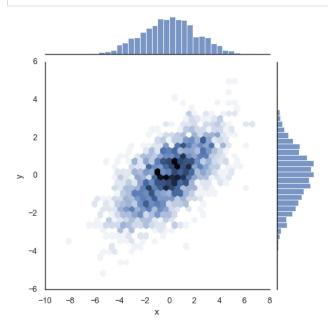
Out[100]: <AxesSubplot:xlabel='x', ylabel='y'>



```
In [101]: with sns.axes_style("white"):#jointplot
    sns.jointplot(x="x",y="y",data=data,kind="kde")
```



In [102]: with sns.axes\_style("white"):#hexagonally based histogram
 sns.jointplot(x="x",y="y",data=data,kind="hex")

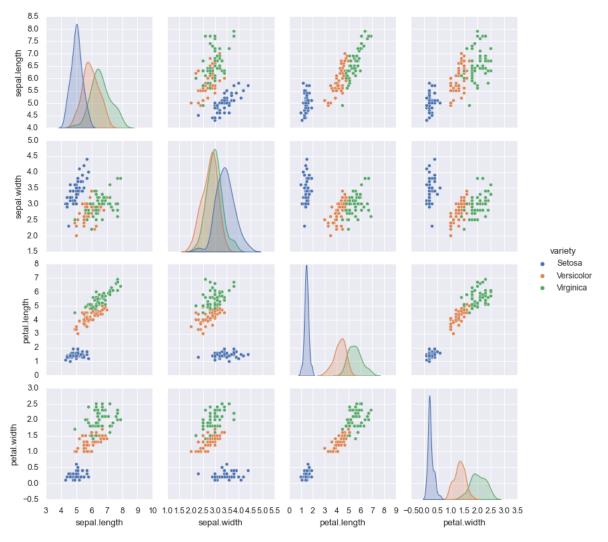


In [103]: iris=pd.read\_csv("Iris.csv")#iris dataset has measurements of petals and sepals from iris species
iris.head()

Out[103]:		sepal.length	sepal.width	petal.length	petal.width	variety
	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
	1	5.0	3.6	1.4	0.2	Setosa

In [104]: sns.pairplot(iris,hue="variety",height=2.5)#hue is given as a clumn name to mark colors

Out[104]: <seaborn.axisgrid.PairGrid at 0x27217aa70a0>



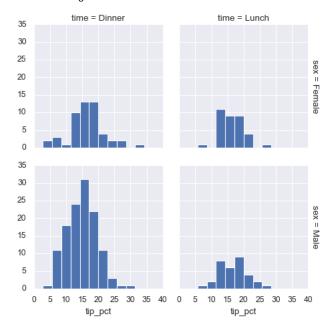
In [105]: tips=pd.read\_csv("tips.csv")
tips.head()

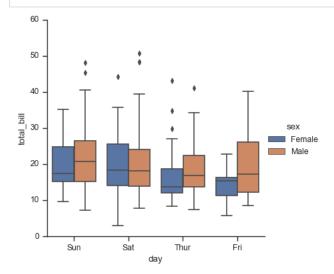
Out[105]:

	total_bill	tip	sex	smoker	day	time	size
0	16.99	1.01	Female	No	Sun	Dinner	2
1	10.34	1.66	Male	No	Sun	Dinner	3
2	21.01	3.50	Male	No	Sun	Dinner	3
3	23.68	3.31	Male	No	Sun	Dinner	2
4	24.50	3 61	Eomala	No	Sun	Dinner	1

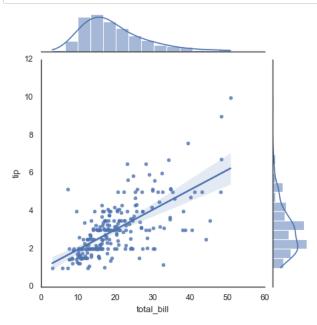
In [106]: #faceted histograms are histograms plotted on subsets of data
 tips['tip\_pct']=100\*(tips["total\_bill"])#adding tips percentage column
 grid=sns.FacetGrid(tips,row="sex",col="time",margin\_titles=True)#assigning row as sex and column and time
 grid.map(plt.hist,"tip\_pct",bins=np.linspace(0, 40, 15))

Out[106]: <seaborn.axisgrid.FacetGrid at 0x272180c0fd0>





```
In [108]: #regression in joint plots
with sns.axes_style("white"):#hexagonally based histogram
sns.jointplot(x="total_bill",y="tip",data=tips,kind="reg")
```

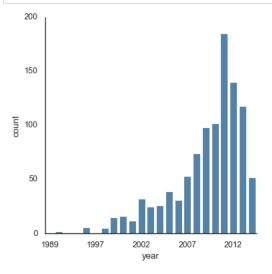


```
In [109]: planets=sns.load_dataset("planets")
    planets.head()
```

## Out[109]:

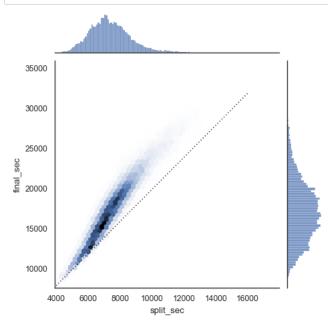
	memou	number	orbitai_periou	IIIass	uistance	year
0	Radial Velocity	1	269.300	7.10	77.40	2006
1	Radial Velocity	1	874.774	2.21	56.95	2008
2	Radial Velocity	1	763.000	2.60	19.84	2011
3	Radial Velocity	1	326.030	19.40	110.62	2007
4	Radial Velocity	1	516.220	10.50	119.47	2009

```
In [110]: #bar plots in seaborn
with sns.axes_style('white'):
    g=sns.catplot(x="year",data=planets,kind="count",color="steelblue")
    g.set_xticklabels(step=5)
```



```
In [111]: with sns.axes_style('white'):
                g=sns.catplot(x="year",data=planets,aspect=4.0,kind="count",color="steelblue",hue="method",order=range(2001, 2015))
                g.set_ylabels("Number of Planets Discovered")
In [128]: data = pd.read_csv("marathon-data.csv")
           data.head()
Out[128]:
               age gender
                              split
                                      final
            0
                                   02:08:51
                33
                          01:05:38
                32
                        M 01:06:26 02:09:28
            2
                31
                        M 01:06:49 02:10:42
                38
                        M 01:06:16 02:13:45
                        M 01:06:32 02:13:59
                31
In [129]: data.dtypes#time is loaded as string hence we have to convert to datetime format
Out[129]: age
                       int64
           gender
                      object
           split
                      object
           final
                      object
           dtype: object
In [130]: def convert_time(s):
               h, m, s = map(int, s.split(':'))
               return pd.Timedelta(hours=h, minutes=m, seconds=s)
           data = pd.read_csv('marathon-data.csv',converters={'split':convert_time, 'final':convert_time})
           data.head()
Out[130]:
               age gender
                                    split
                                                 final
                33
                        M 0 days 01:05:38 0 days 02:08:51
                32
                        M 0 days 01:06:26 0 days 02:09:28
            2
                31
                        M 0 days 01:06:49 0 days 02:10:42
            3
                38
                       M 0 days 01:06:16 0 days 02:13:45
                        M 0 days 01:06:32 0 days 02:13:59
                31
In [131]: data.dtypes
Out[131]: age
                                 int64
           gender
                                object
           split
                      timedelta64[ns]
                      timedelta64[ns]
           final
           dtype: object
In [139]: data["split_sec"]=data["split"]/pd.Timedelta(seconds=1)#creating new column split second
           data["final_sec"]=data["final"]/pd.Timedelta(seconds=1)#creating new column final second
           data.head()
Out[139]:
               age gender
                                    split
                                                 final split_sec final_sec
                33
                        M 0 days 01:05:38 0 days 02:08:51
                                                         3938.0
                                                                  7731.0
                32
                        M 0 days 01:06:26 0 days 02:09:28
                                                        3986.0
                                                                  7768.0
                31
                        M 0 days 01:06:49 0 days 02:10:42
                                                        4009.0
                                                                  7842.0
                38
                        M 0 days 01:06:16 0 days 02:13:45
                                                        3976.0
                                                                  8025.0
                       M 0 days 01:06:32 0 days 02:13:59
                31
                                                        3992.0
                                                                 8039.0
```

```
In [144]: with sns.axes_style("white"):
    g=sns.jointplot(x="split_sec",y="final_sec",data=data,kind="hex")
    plt.plot(np.linspace(4000, 16000),np.linspace(8000, 32000), ':k')
```

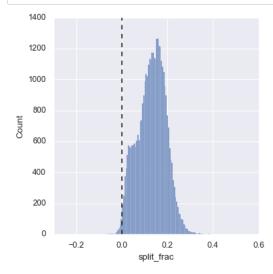


In [146]: data["split\_frac"] =1 -2 \* data["split\_sec"] / data["final\_sec"]#formula for calculating split
#positive split means fast at first half of race and slowing down negative split is the reverse
data.head()

## Out[146]:

	age	gender	split	final	split_sec	final_sec	split_frac
0	33	М	0 days 01:05:38	0 days 02:08:51	3938.0	7731.0	-0.018756
1	32	М	0 days 01:06:26	0 days 02:09:28	3986.0	7768.0	-0.026262
2	31	М	0 days 01:06:49	0 days 02:10:42	4009.0	7842.0	-0.022443
3	38	М	0 days 01:06:16	0 days 02:13:45	3976.0	8025.0	0.009097
4	31	М	0 days 01:06:32	0 days 02:13:59	3992.0	8039.0	0.006842

In [148]: sns.displot(data["split\_frac"], kde=False);#kde false implies that kde plot is invisible plt.axvline(0,color="k",linestyle="--")#axvline is used for plotting a vertical line

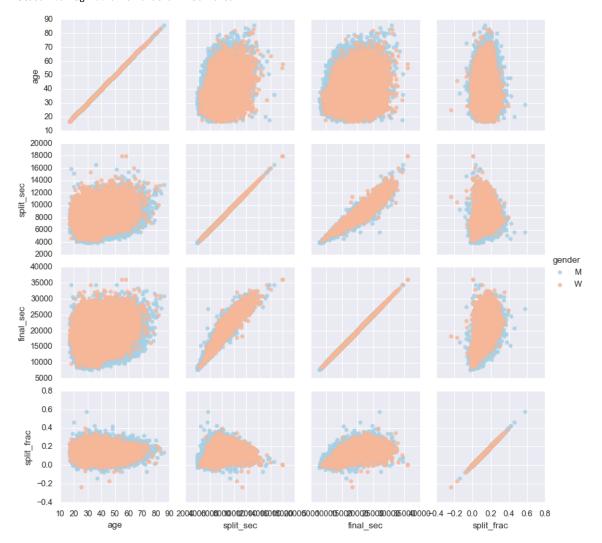


```
In [149]: sum(data.split_frac<0)#number of people with negative split</pre>
```

Out[149]: 251

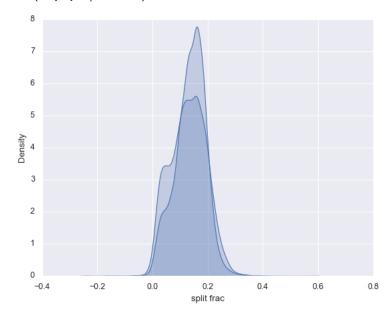
```
In [150]: g=sns.PairGrid(data,vars=["age","split_sec","final_sec","split_frac"],hue="gender",palette="RdBu_r")
    g.map(plt.scatter, alpha=0.8)
    g.add_legend()
```

Out[150]: <seaborn.axisgrid.PairGrid at 0x272567164c0>



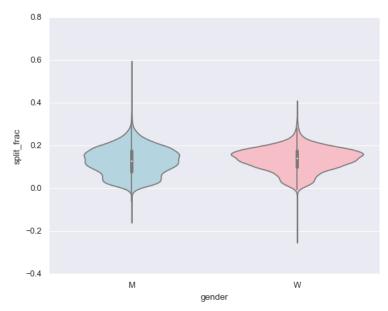
```
In [153]: #comparing split fraction of men and women
sns.kdeplot(data.split_frac[data.gender=="M"],label="men",shade=True)
sns.kdeplot(data.split_frac[data.gender=="W"],label="women",shade=True)
plt.xlabel("split frac")
```

Out[153]: Text(0.5, 0, 'split frac')



```
In [155]: #violin plot comaparison
sns.violinplot(x="gender",y="split_frac",data=data,palette=["lightblue","lightpink"])
```

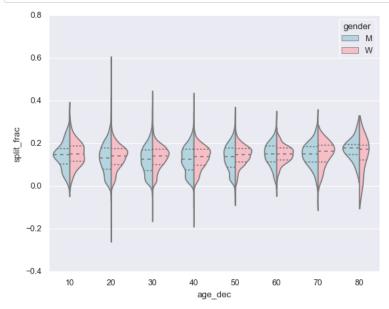
Out[155]: <AxesSubplot:xlabel='gender', ylabel='split\_frac'>



In [156]: data["age\_dec"]=data.age.map(lambda age:10\*(age//10))#adding new column age decade
data.head()

## Out[156]:

	age	gender	split	final	split_sec	final_sec	split_frac	age_dec
0	33	М	0 days 01:05:38	0 days 02:08:51	3938.0	7731.0	-0.018756	30
1	32	М	0 days 01:06:26	0 days 02:09:28	3986.0	7768.0	-0.026262	30
2	31	М	0 days 01:06:49	0 days 02:10:42	4009.0	7842.0	-0.022443	30
3	38	М	0 days 01:06:16	0 days 02:13:45	3976.0	8025.0	0.009097	30
4	31	M	0 days 01:06:32	0 days 02:13:50	3002 N	8030 O	0.006842	30



Out[164]: <seaborn.axisgrid.FacetGrid at 0x272582ce100>

