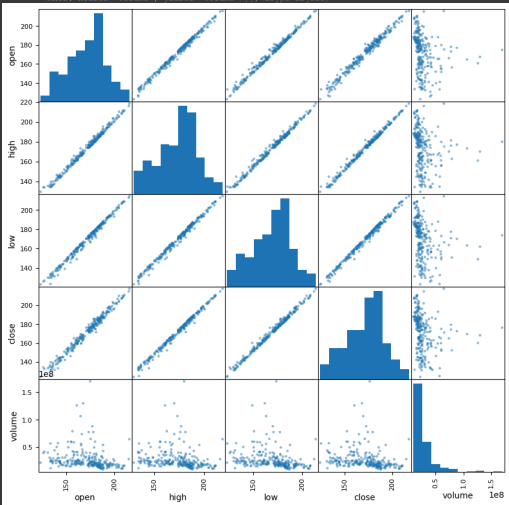


- ✧ pandas.plotting subpackage

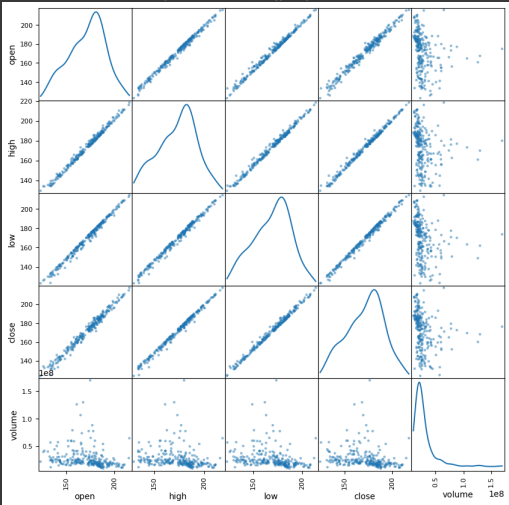
- ▼ About the Data

- Setup

- Scatter matrix

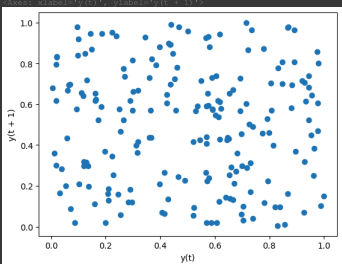


### Changing the diagonal from histograms to KDE



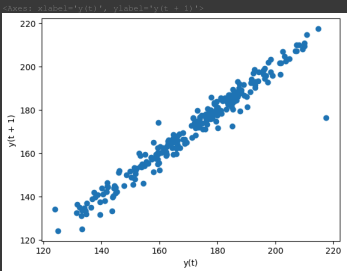
Lag plot

Lag plots let us see how the variable correlations with past observations of itself. Random data has no pattern



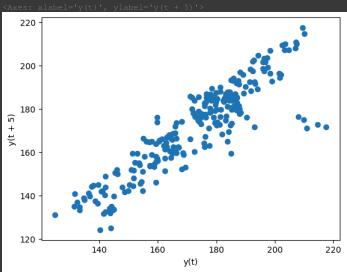
Data with some level of correlation to itself (autocorrelation) may have patterns. Stock prices are highly auto-correlated.

```
1 lag_plot(fb.close)
```



The default lag is 1, but we can alter this with the lag parameter. Let's look at a 5 day lag (a week of trading activity):

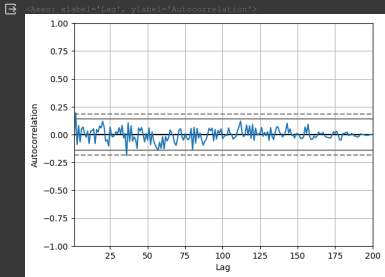
```
1 lag_plot(fb.close, lag=5)
```



#### Autocorrelation plots

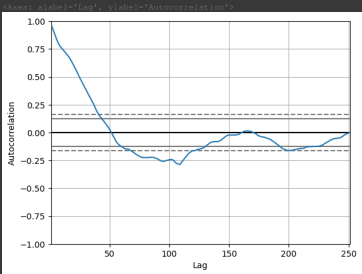
We can use the autocorrelation plot to see if this relationship may be meaningful or just noise. Random data will not have any significant autocorrelation (it stays within the bounds below

```
1 from pandas.plotting import autocorrelation_plot
2
3 np.random.seed(0) # make this repeatable
4 autocorrelation_plot(pd.Series(np.random.random(size=200)))
```



Stock data, on the other hand, does have significant autocorrelation

```
1 autocorrelation_plot(fb.close)
```



#### Bootstrap plot

This plot helps us understand the uncertainty in our summary statistics:

```
1 from pandas.plotting import bootstrap_plot
2
3 fig = bootstrap_plot(fb.volume, fig=plt.figure(figsize=(10, 6)))
```

