



# ACF and PACF

Theory



# Python for Time Series

- Let's learn about 2 very useful plot types
  - ACF - AutoCorrelation Function Plot
  - PACF - Partial AutoCorrelation Function Plot
- To understand these plots, we first need to understand correlation!



# Python for Time Series

- Correlation is a measure of the strength of the linear relationship between two variables.

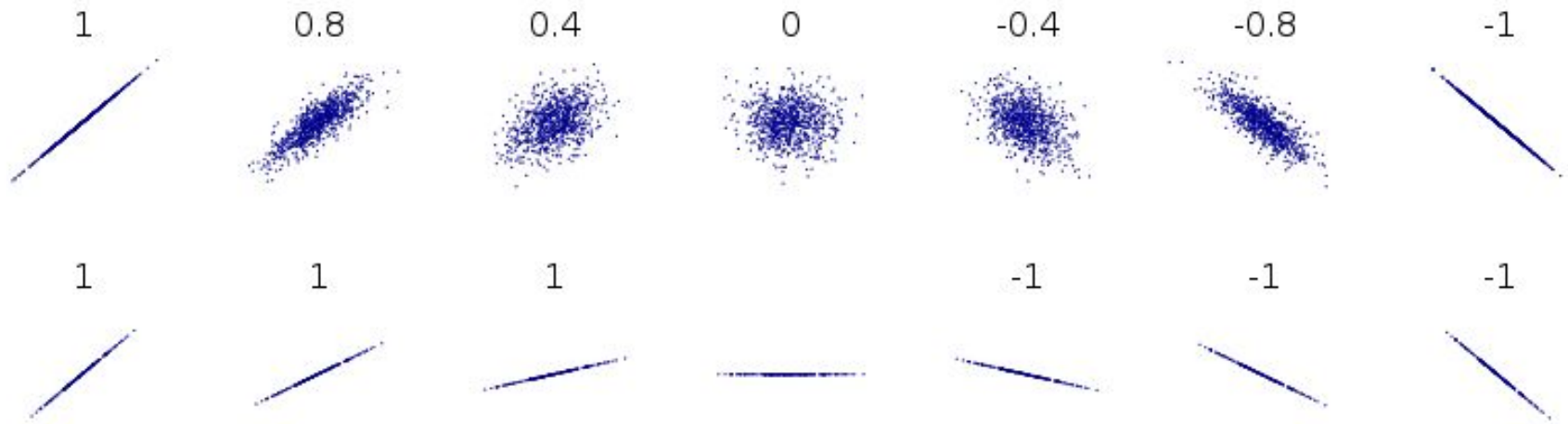


# Python for Time Series

- The closer the correlation is to  $+1$ , the stronger the positive linear relationship
- The closer the correlation is to  $-1$ , the stronger the negative linear relationship.
- And the closer the correlation is to zero, the weaker the linear relationship, or association.



# Python for Time Series





# Python for Time Series

- An autocorrelation plot (also known as a Correlogram ) shows the correlation of the series with itself, lagged by  $x$  time units.
- So the  $y$  axis is the correlation and the  $x$  axis is the number of time units of lag.

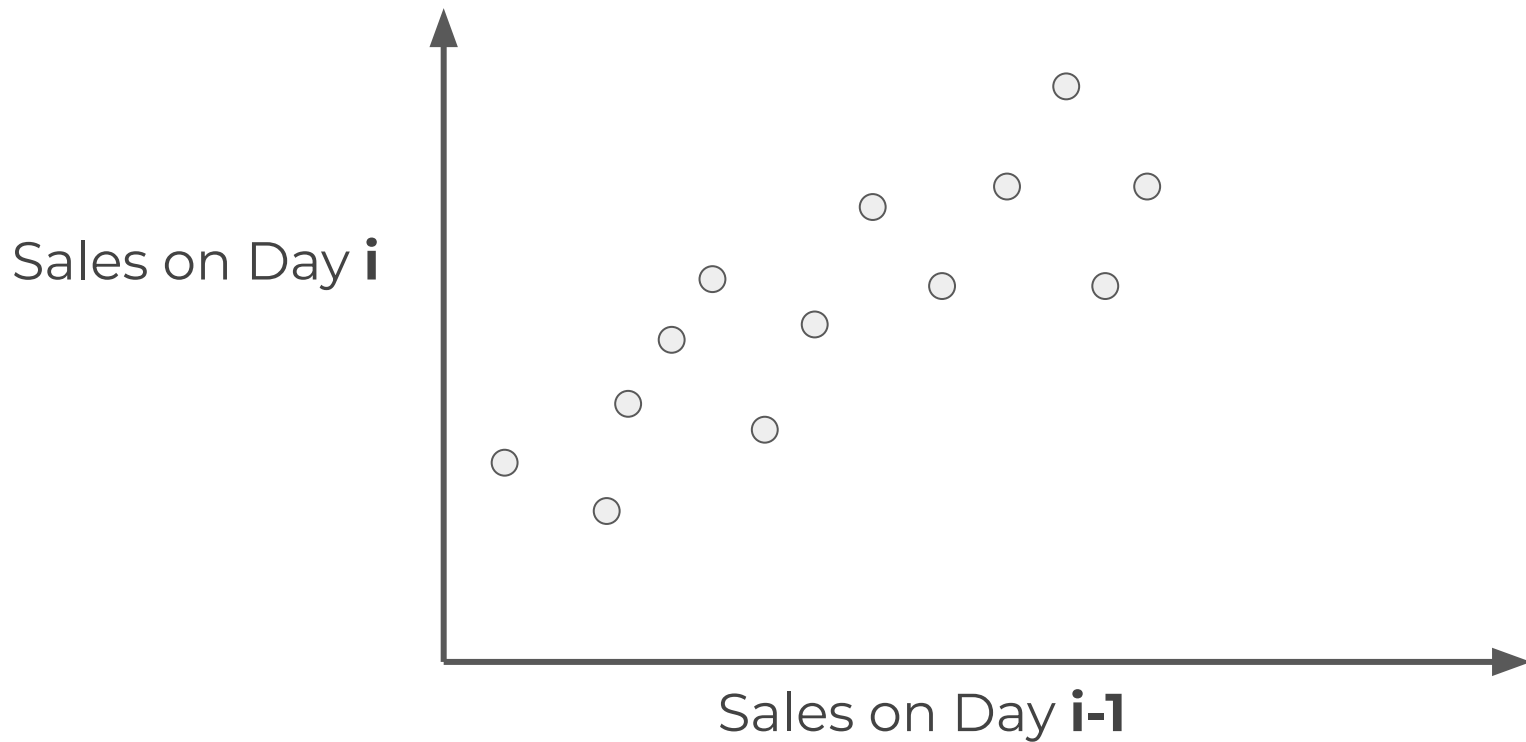


# Python for Time Series

- Imagine we had some sales data.
- We can compare the standard sales data against the sales data shifted by 1 time step.
- This answers the question, “How correlated are today’s sales to yesterday’s sales?”



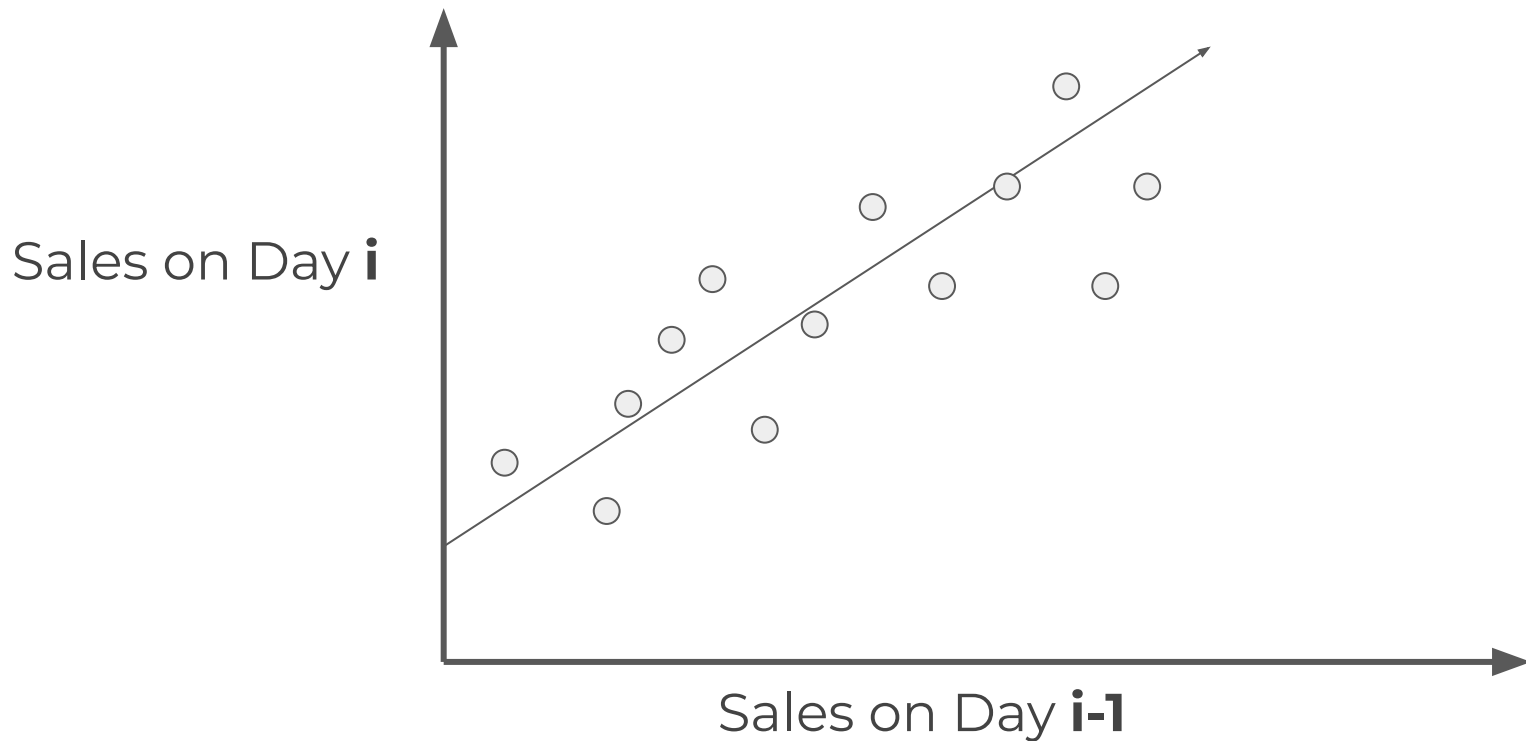
# Python for Time Series





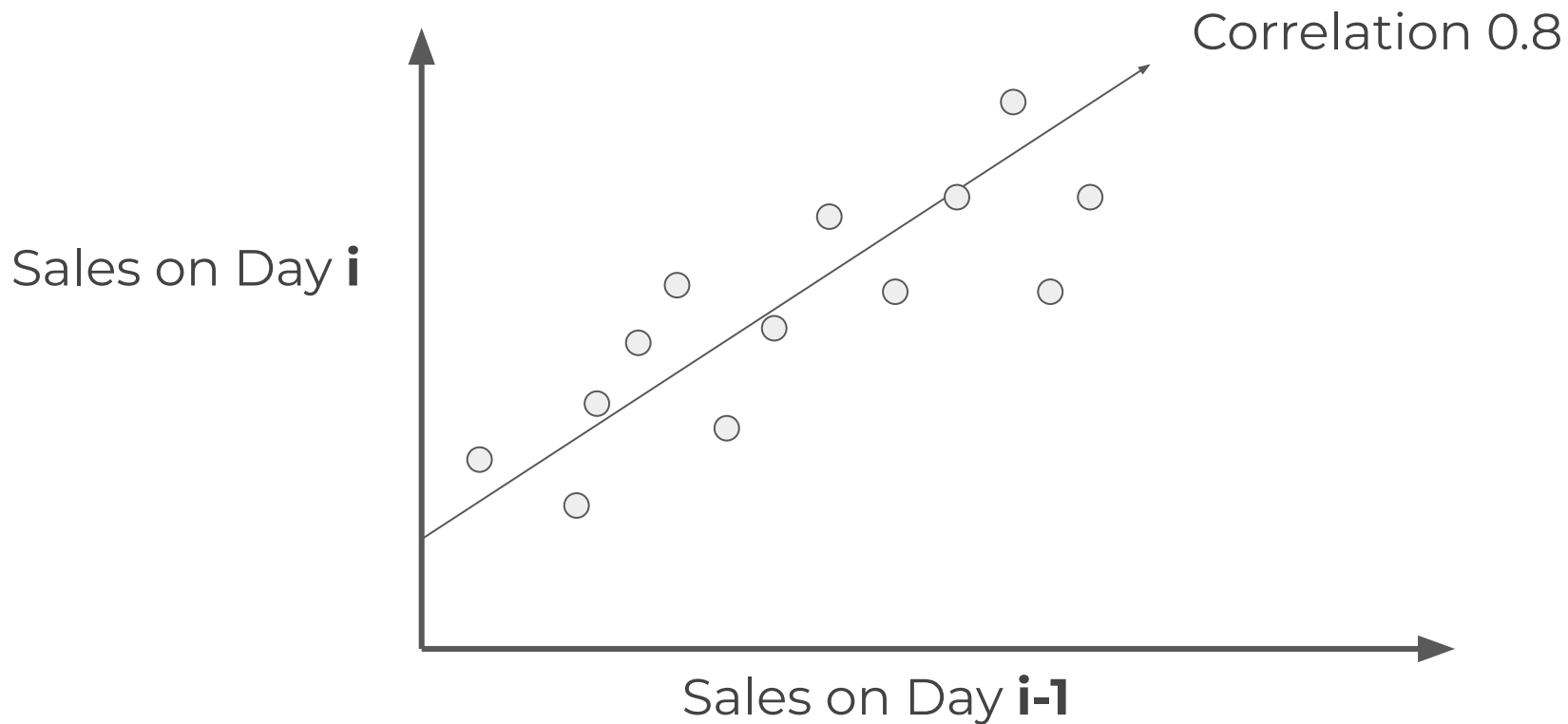


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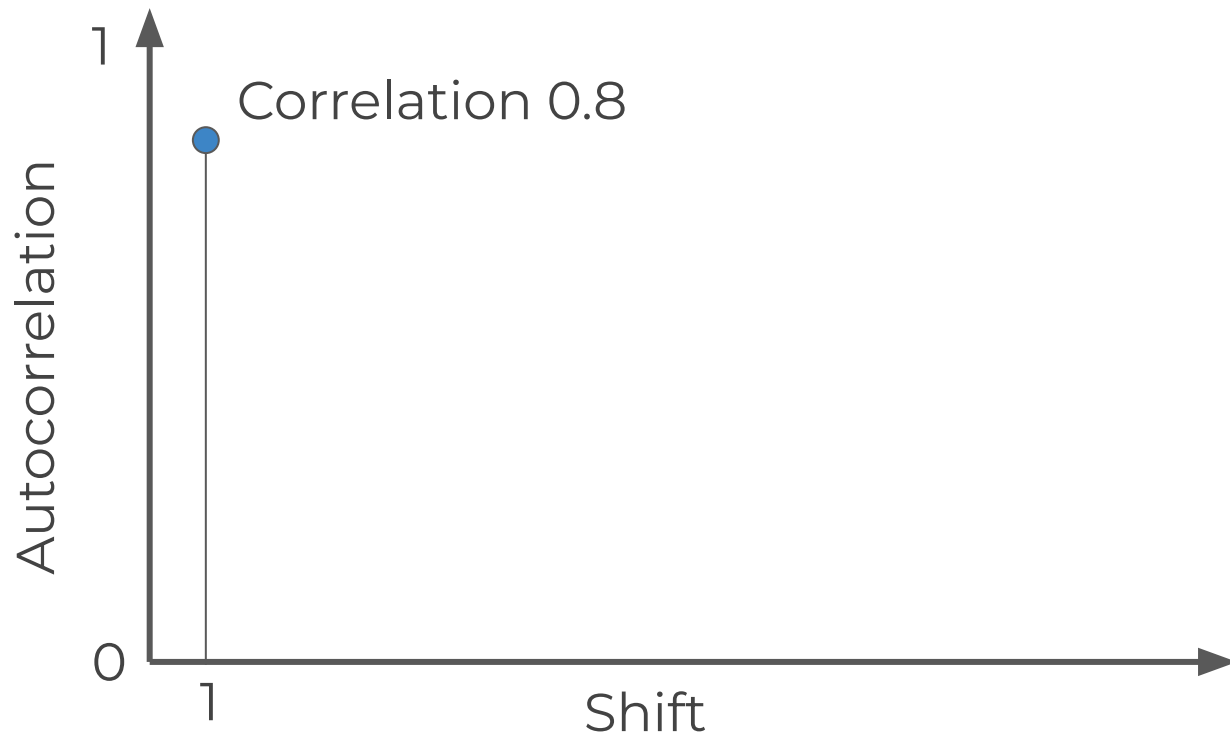


# Python for Time Series



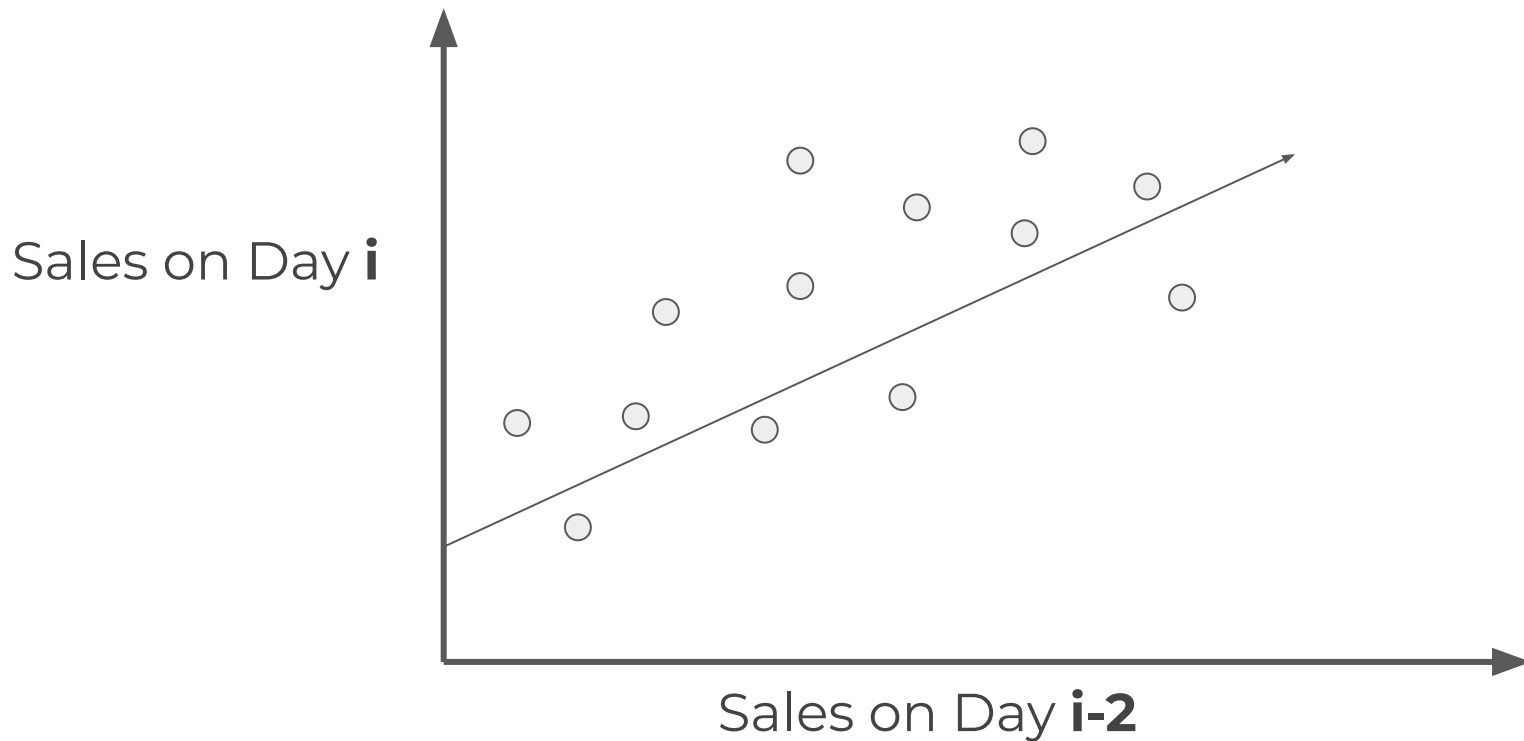


# Python for Time Series



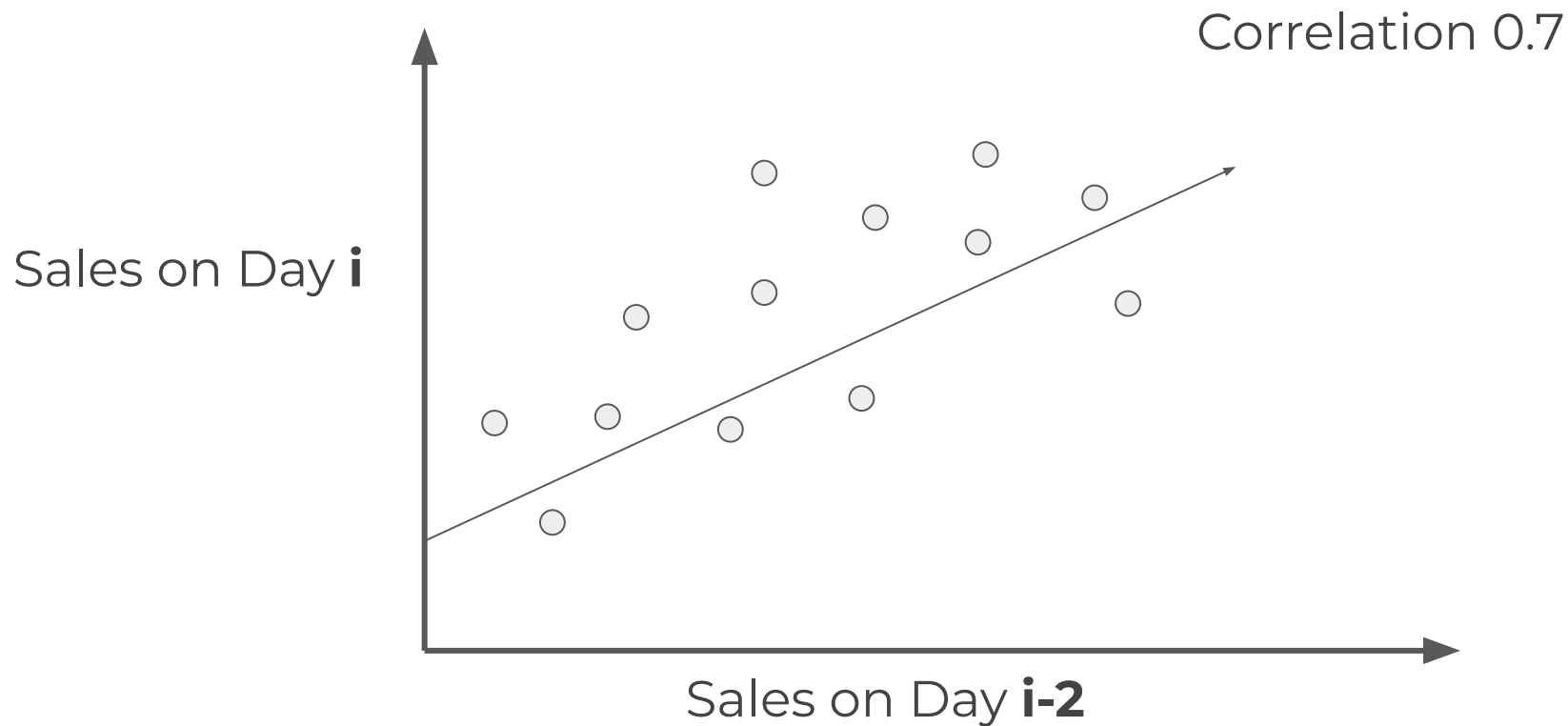


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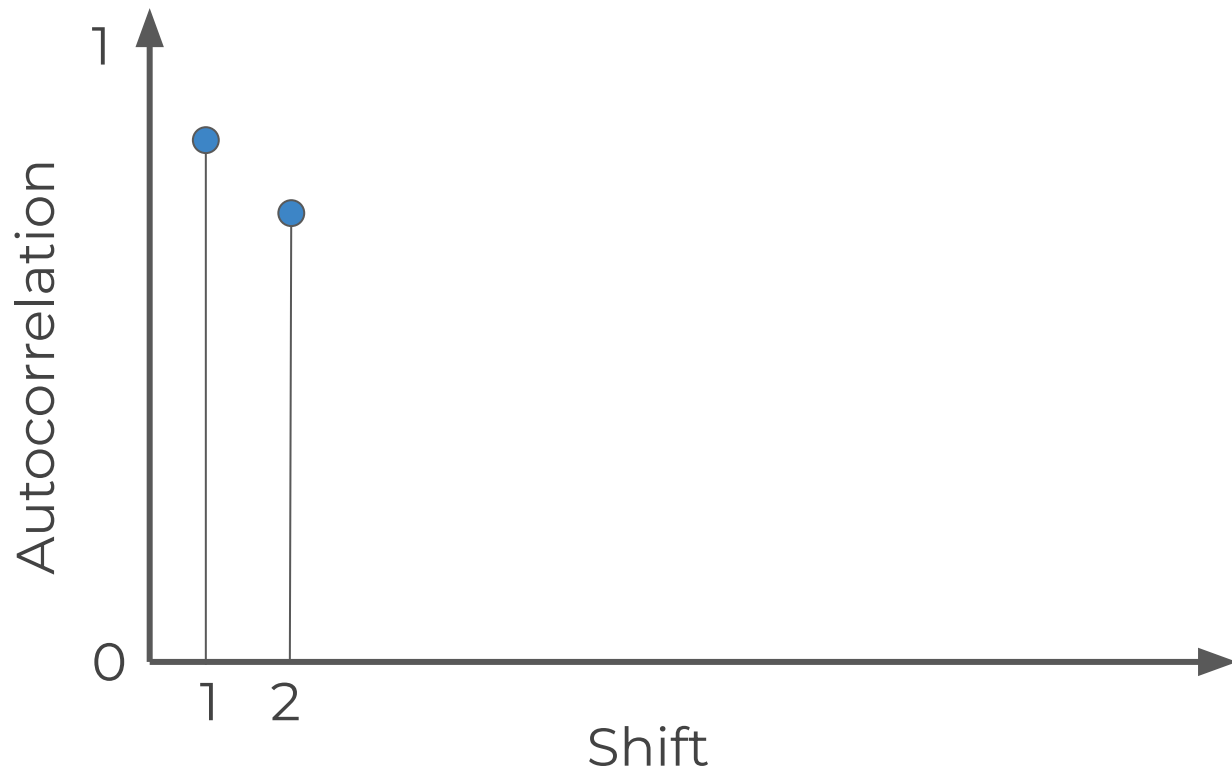


# Python for Time Series





# Python for Time Series





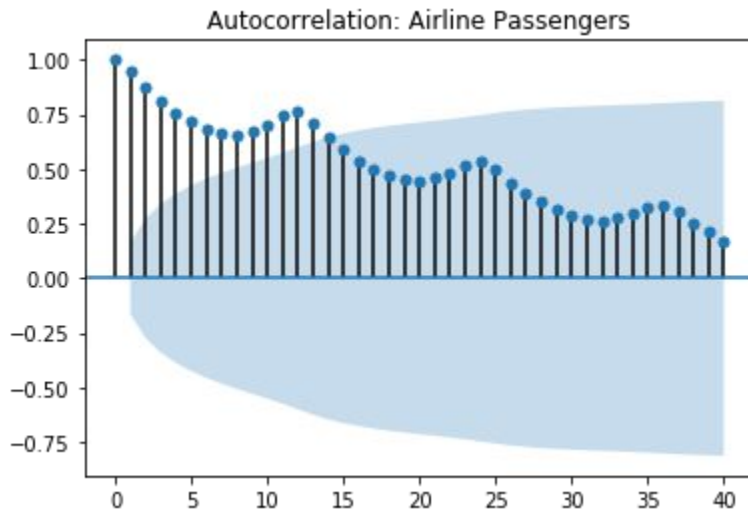
# Python for Time Series

- An autocorrelation plot shows the correlation of the series with itself, lagged by  $x$  time units.
- You go on and do this for all possible time lags  $x$  and this defines the plot.
- Let's see some typical examples!



# Python for Time Series

- Gradual Decline

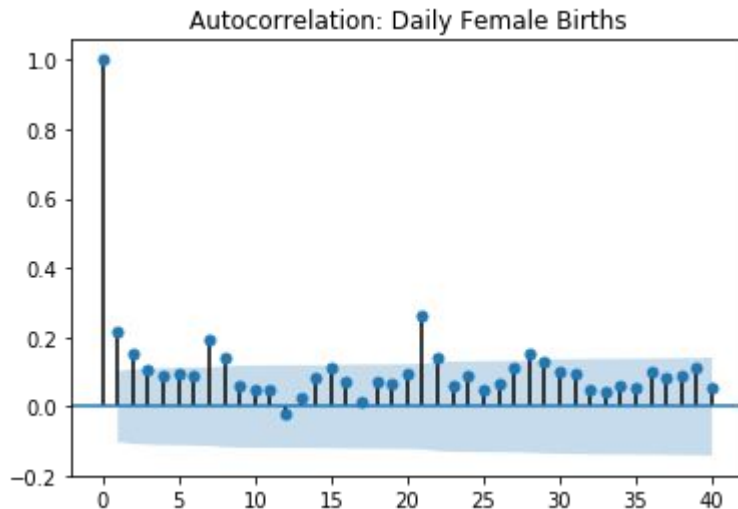






# Python for Time Series

- Sharp Drop-off





# Python for Time Series

- It makes sense that in general there is a decline of some sort, the further away you get with the shift, the less likely the time series would be correlated with itself.



# Python for Time Series

- The actual interpretation and how it relates to ARIMA models can get a bit complicated, but there are some basic common methods we can use for the ARIMA model.



# Python for Time Series

- There are also partial autocorrelation plots!
- These are a little more complicated than autocorrelation plots, but let's show you the basics.

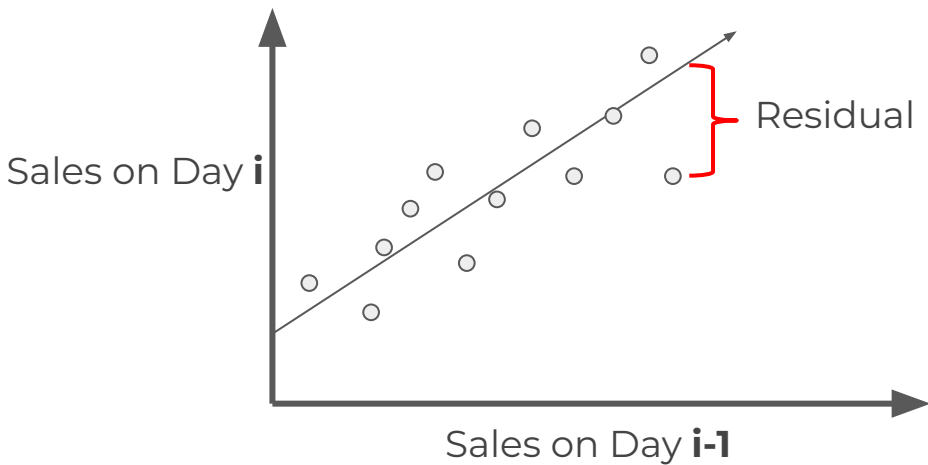


# Python for Time Series



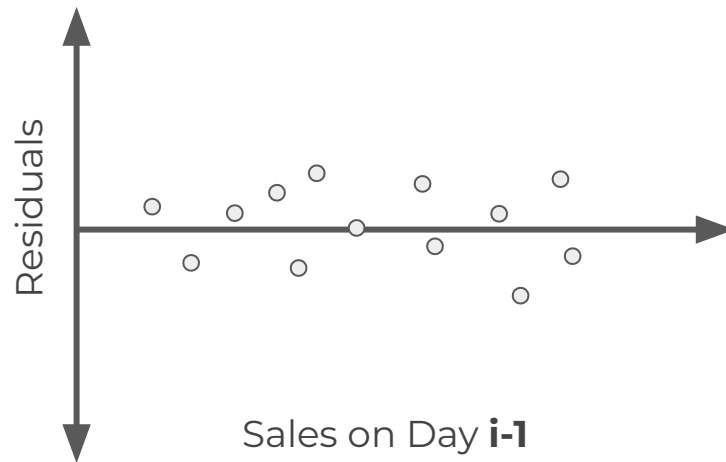
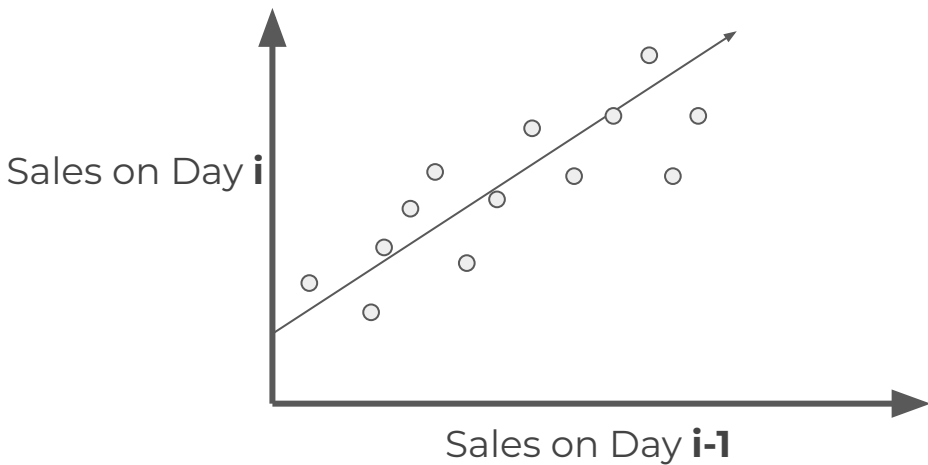


# Python for Time Series





# Python for Time Series





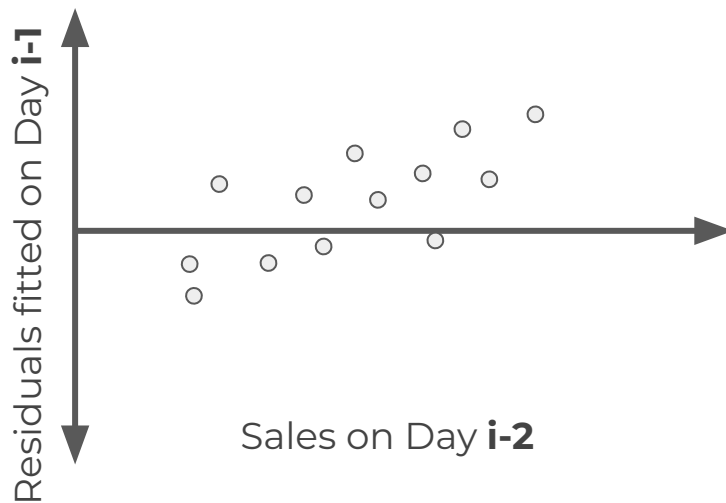
# Python for Time Series





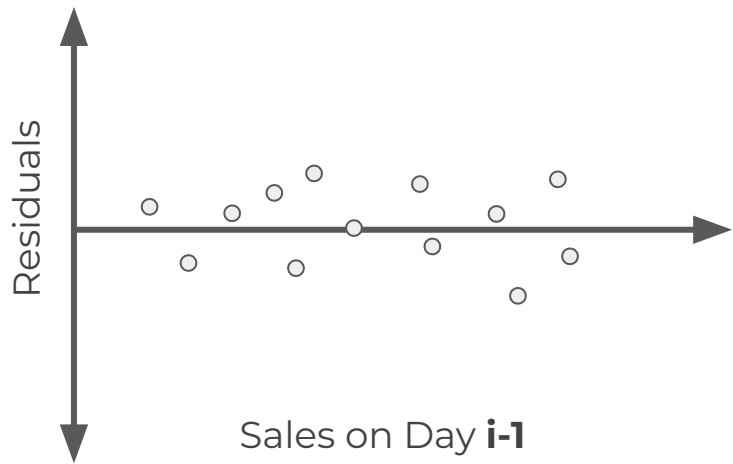


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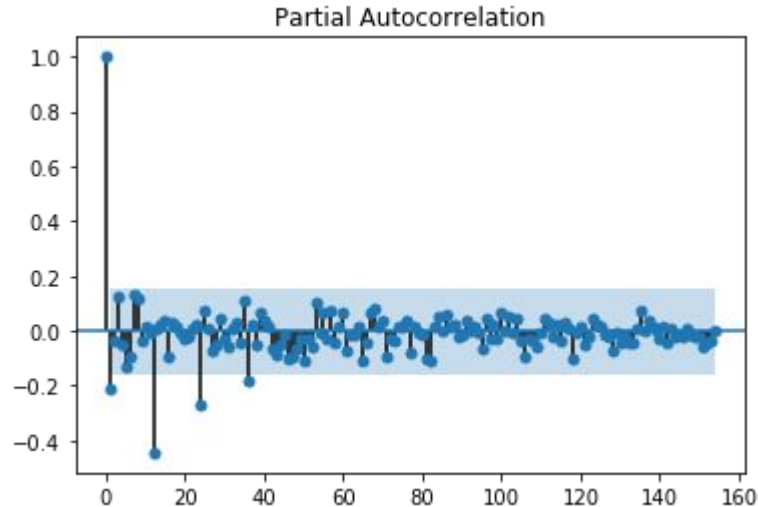
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# Python for Time Series

- Let's see an example of what the plot can look like:





# Python for Time Series

- We essentially plot out the relationship between the previous day's residuals versus the real values of the current day.
- In general we expect the partial autocorrelation to drop off quite quickly.



# Python for Time Series

- The ACF describes the autocorrelation between an observation and another observation at a prior time step that includes direct and indirect dependence information.



# Python for Time Series

- The PACF only describes the direct relationship between an observation and its lag.



# Python for Time Series

- These two plots can help choose order parameters for ARIMA based models.
- Later on, we will see that it is usually much easier to perform a grid search of the parameter values, rather than attempt to read these plots directly.



# Python for Time Series

- Let's explore how to create these plots with statsmodels!