Time Series Modeling

CCTS 40500

- AR: Autoregression. A model that uses the dependent relationship between an observation and some number of lagged observations
- I: Integrated. The use of differencing of raw observations (e.g. subtracting an observation from an observation at the previous time step) in order to make the time series stationary.
- MA: Moving Average. A model that uses the dependency between an observation and a residual error from a moving average model applied to lagged observations.

- p: The number of lag observations included in the model, also called the lag order.
- **d**: The number of times that the raw observations are differenced, also called the degree of differencing.
- **q**: The size of the moving average window, also called the order of moving average.

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Autocorrelation

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Partial Autocorrelation

$$R(au) = rac{1}{\sigma^2} \mathbf{E}[(X_t - \mu)(X_{t+ au} - \mu)]$$

Moving Average Model

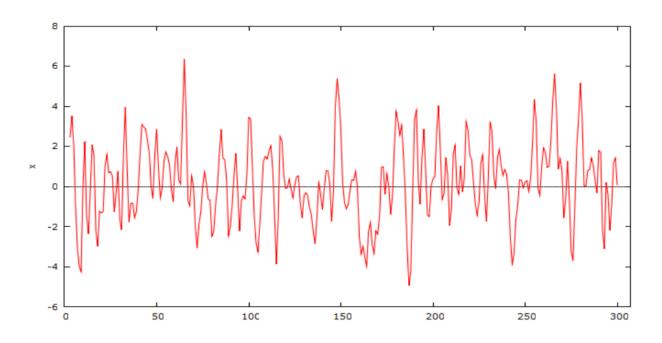


Figure: A simulated time series

Our time series seems the realization of a stationary process with zero mean, thus we can look at sample autocorrelation and partial autocorrelation function to establish the orders p and q of the ARMA model.

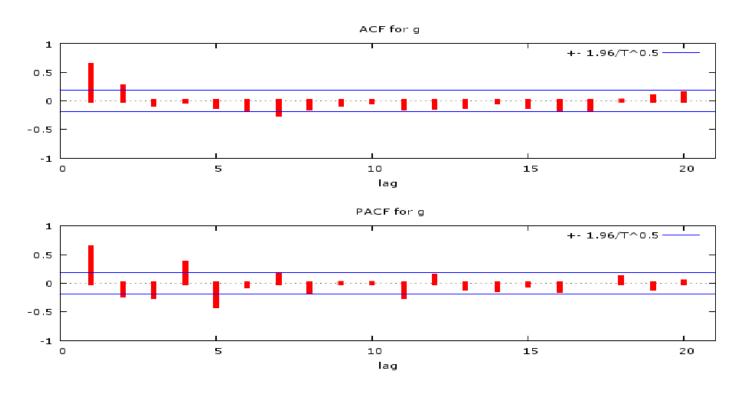


Figure: Sample autocorrelation and sample partial autocorrelation

ACF cuts off after lag 2

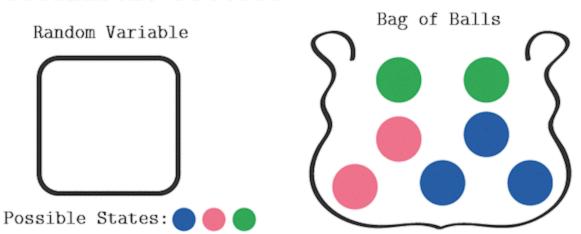
$$X_t = u_t + heta_1 u_{t-1} + heta_2 u_{t-2}, \ u_t \sim WN(0, \sigma^2)$$

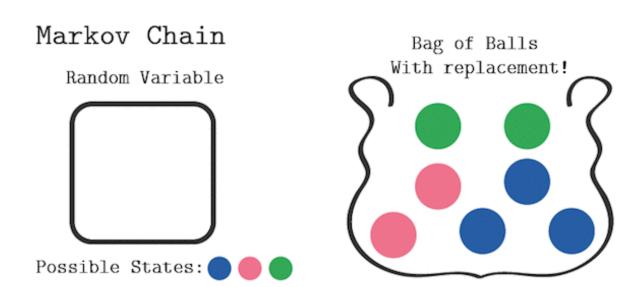
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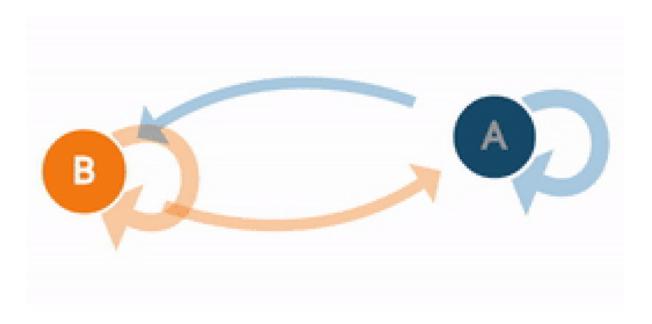
$$egin{aligned} ext{If } d = 0: y_t = Y_t \ ext{If } d = 1: y_t = Y_t - Y_{t-1} \ ext{If } d = 2: y_t = (Y_t - Y_{t-1}) - (Y_{t-1} - Y_{t-2}) = Y_t - 2Y_{t-1} + Y_{t-2} \end{aligned}$$

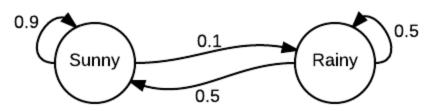
$$\hat{y}_t = \mu + \sum_i \phi_i^d y_{t-i} - \sum_j \theta_j^q e_{t-j}$$

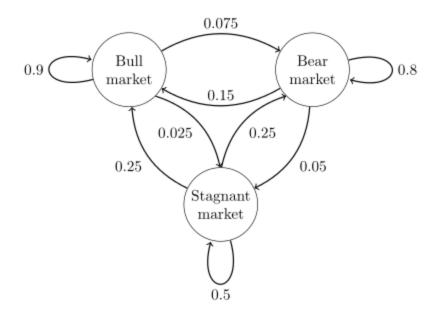
Stochastic Process











$$P = \begin{bmatrix} 0.9 & 0.075 & 0.025 \\ 0.15 & 0.8 & 0.05 \\ 0.25 & 0.25 & 0.5 \end{bmatrix}.$$