# AR1\_Random\_Walk

#### What is a Random Walk?

A random walk is a time series process where each value is the sum of the previous value and a random noise term. It can be described mathematically as:

$$X_t = X_{t-1} + \epsilon_t$$

Here: - (  $X_t$  ): The value at time ( t ), - (  $X_{t-1}$  ): The value at time ( t-1 ), - (  $\epsilon_t$  ): A random noise term, typically assumed to follow a normal distribution with mean 0 and variance (  $\sigma^2$  ).

Key characteristics of a random walk:

- 1. Non-Stationarity: A random walk is not stationary because its mean and variance change over time. The variance grows with time, making it unpredictable in the long run.
- 2. *Memory*: Each step depends directly on the previous value, so the process exhibits strong persistence over time.

### AR(1) Model

The AR(1) model is defined as:

$$X_t = \phi*1*X_{t-1} + \epsilon_t$$

Here: - (  $\phi*1$  ): Autoregressive coefficient, which determines the influence of ( X\*t-1 ) on (  $X_t$  ). - (  $\epsilon_t$ ): White noise.

The AR(1) process is stationary if (  $|\phi_1|<1$  ). For (  $|\phi_1|\geq 1$  ), the process becomes non-stationary.

## How ( $\phi_1 = 1$ ) Makes AR(1) a Random Walk

When (  $\phi_1=1$  ), the AR(1) model becomes:

$$X_t = X_{t-1} + \epsilon_t$$

This is exactly the definition of a random walk. In this case: - Each value ( $X_t$ ) is the previous value ( $X_{t-1}$ ) plus a random noise term. - The time series lacks a tendency to revert to a mean, as there is no dampening factor to bring values back to a central level.

## Key Implications of ( $\phi_1 = 1$ ):

- 1. Non-Stationarity:
  - For  $(|\phi_1| < 1)$ , the AR(1) process is stationary and has constant variance.
  - For ( $\phi_1 = 1$ ), the variance of ( $X_t$ ) grows linearly with time:  $\text{Var}(X_t) = t \cdot \sigma^2$ This makes the process non-stationary.
- 2. Unbounded Growth:
  - As time progresses, the values of (  $X_t$  ) can drift arbitrarily far from their starting point due to the cumulative effect of (  $\epsilon_t$  ).
- 3. Lack of Mean Reversion:
  - In a stationary AR(1) process ((  $|\phi_1| < 1$  )), the series tends to revert to its mean. For (  $\phi_1 = 1$  ), there is no mean reversion, as the process has no equilibrium level.

#### Visualization of Random Walk vs Stationary AR(1):

- 1. Random Walk (  $\phi_1 = 1$  ):
  - The series drifts up or down without a predictable pattern.
  - Variance increases over time, and the path depends heavily on the noise term.
- 2. Stationary AR(1) ( $|\phi_1| < 1$ ):
  - The series oscillates around a fixed mean with a constant variance.
  - Noise contributes to deviations, but the process is "pulled back" to the mean by the coefficient (  $\phi_1$  ).

Understanding this relationship is crucial for time series modeling, as it helps determine whether a process needs differencing or detrending to achieve stationarity before analysis or forecasting.