

# 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 ( $|\phi| < 1$ ). For ( $|\phi| \geq 1$ ), the process becomes non-stationary.

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

When ( $\phi = 1$ ), 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. *Non-Stationarity*:
  - o For ( $|\phi| < 1$ ), the AR(1) process is stationary and has constant variance.

- For ( $\phi_1 = 1$ ), the variance of ( $X_t$ ) grows linearly with time:  $\text{Var}(X_t) = t$ . 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.