# TIME SERIES ANALYSIS LECTURE 1

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# Stationarity: Example 2

#### 2. A Stochastic Model With a Deterministic Linear Trend

### A Stochastic Model with a Linear Trend

Consider a model with a deterministic linear trend:

$$x_t = a + bt + w_t$$
where  $x_t$  is a white noise with mean 0 and variance  $\sigma_w^2$ 

The expected value of  $x_t$  is

$$E(x_t) = E(a + bt + w_t)$$

$$= a + btE(w_t)$$

$$= a + bt$$

## 2. A Stochastic Model With a Deterministic Linear Trend (2)

As such, a stochastic model with a linear trend is not mean stationary, as the mean changes with time. If b > 0, then the mean is an increasing function of the time index t. On the other hand, if b < 0, then the mean is an decreasing function of the time index t.

$$Var(x_t) = Var(a + bt + w_t)$$

$$= Var(w_t)$$

$$= \sigma_w^2$$

which is a constant. So, while the model is not mean stationary, it is variance stationary.

## 2. A Stochastic Model With a Deterministic Linear Trend (3)

$$Cov(x_t, x_{t-1}) = Cov (a + bt + w_t, a + b(t-1) + w_{t-1})$$

$$= Cov (w_t, w_{t-1})$$

$$= 0$$

since  $w_t$  is a white noise series.

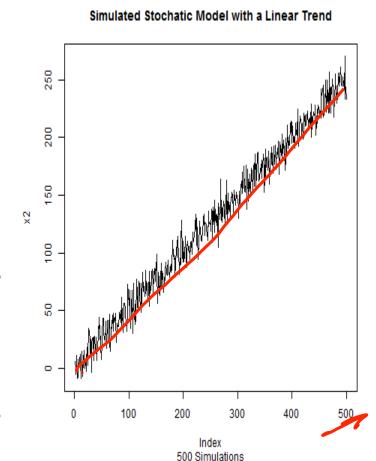
Therefore, the stochastic model with a deterministic linear trend is not (strictly or weakly) stationary, although as we will see in a few lectures, it can be easily transformed into a stationary model.

#### **Trend: Simulations**

 $x_t = 1 + 0.5t + w_t$  where  $w_t$  is a series of independent Gaussian white noise with mean 0 and variance 10

```
set.seed(898)
sigma_w = 10
beta0 = 1
beta1 = 0.5
t = seq(1,500)
w <- rnorm(500,0,sigma_w)
x2 <- beta0 + beta1*t + w
cbind(t, x2, w)|
summary(x2)
mean(x2)
sd(x2)</pre>
```

- The simulated series
   appears as a linear trend,
   as if the white noise does
   not affect the trend.
- The acf shows that the series is very persistence, meaning that is highly correlated with its lags.



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