## Stationarity time series: rationale

(weakly) stationary process: Inuitively, we have a stationary process if its mean is finite and constant and the autocovariance function depends only on the lag, and not on the time t.

For a discrete or continuous **stationary** time series  $\{X_t\}$ , the mean and variance functions are respectively defined as:

$$\mu_t = E[X_t] = \mu$$

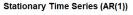
$$\sigma_t^2 = var(X_t) = E[(X_t - \mu)^2] = \sigma^2$$

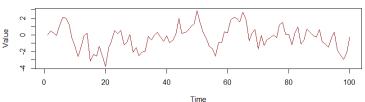
An example of a stationary series is an AR(1) process. An example of a non-stationary series is a Random Walk (RW). We consider those two to illustrate two test for stationary series, namely the augmented Dicky-Fuller test and the Kwiatkowski-Phillips-Schmidt-Shin (KPSS) test.

# R code to generate simple time series

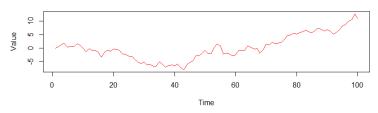
```
1 # simulate two times series (first = stationary, second = non stationary)
 2 set.seed(2024)
 4 # parameters
 5 n = 100; phi = 0.7; sigma = 1 # number of observation, AR(1) parameter value and
         s.d. of WN
7 # stationary time series: AR(1)
8 \pm 1 = numeric(n)
 9 epsilon = rnorm(n, mean = 0, sd = sigma) # White noise
10
11 for (t in 2:n) {
12 t1[t] = phi * t1[t - 1] + epsilon[t]
13 F
14
15 # non-stationary time series: A random walk
16 t2 = numeric(n)
17 epsilon = rnorm(n. mean = 0. sd = sigma) # White noise
18
19 for (t in 2:n) {
20 t2[t] = t2[t - 1] + epsilon[t]
21 }
22
23 # plots
24 \text{ par}(\text{mfrow} = c(2, 1))
25 plot(t1, type = "1", col = "darkred", main = "Stationary Time Series (AR(1))",
       vlab = "Value", xlab = "Time")
26
27 plot(t2, type = "l", col = "red", main = "Non-Stationary Time Series (Random
        Walk)".
28
        vlab = "Value", xlab = "Time")
```

### Plot of the two time series





#### Non-Stationary Time Series (Random Walk)



## Augmented Dickey-Fuller tests in R

```
1 > # augmented Dickey-Fuller test
2 > library(tseries)
3 > adf.test(t1) # p-value: 0.0321; Ho: the series is non-stationary, is rejected
   Augmented Dickey-Fuller Test
7 data: t1
8 Dickey-Fuller = -3.6518. Lag order = 4. p-value = 0.0321
9 alternative hypothesis: stationary
10
11 > adf.test(t2) # p-value: 0.8317; Ho: the series is non-stationary, is not
       rejected
12
13
   Augmented Dickey-Fuller Test
14
15 data: ±2
16 Dickey-Fuller = -1.3844, Lag order = 4, p-value = 0.8317
17 alternative hypothesis: stationary
```

### KPSS tests in R

```
1 > # KPSS test
 2 > kpss.test(t1, null="Trend") # p-value: 0.0764: Ho: the series is trend
        stationary, is not rejected
 4 KPSS Test for Trend Stationarity
6 data: t1
 7 KPSS Trend = 0.13174, Truncation lag parameter = 4, p-value = 0.0764
 9 > kpss.test(t2, null="Trend") # p-value: 0.01; Ho: the series is trend
        stationary, is rejected
10
11 KPSS Test for Trend Stationarity
12
13 data: ±2
14 KPSS Trend = 0.45554, Truncation lag parameter = 4, p-value = 0.01
15
16 Warning message:
17 In kpss.test(t2, null = "Trend") : p-value smaller than printed p-value
```

### References

The R Project for Statistical Computing: https://www.r-project.org/