## 4.2 Strict Stationarity

A more restrictive definition of stationarity involves all the multivariate distributions of the subsets of TS r.vs.

**Definition 4.4.** A time series  $\{X_t\}$  is called strictly stationary if the random vectors  $(X_{t_1}, \ldots, X_{t_n})^{\mathrm{T}}$  and  $(X_{t_1+\tau}, \ldots, X_{t_n+\tau})^{\mathrm{T}}$  have the same joint distribution for all sets of indices  $\{t_1, \ldots, t_n\}$  and for all integers  $\tau$  and n > 0. It is written as

 $(X_{t_1},\ldots,X_{t_n})^{\mathrm{T}} \stackrel{d}{=} (X_{t_1+\tau},\ldots,X_{t_n+\tau})^{\mathrm{T}},$ 

where  $\stackrel{d}{=}$  means "equal in distribution".

## **Properties of a Strictly Stationary TS**

- 1. The r.vs  $X_t$  are identically distributed for all t. Follows from the definition for n = 1.
- 2. Pairs of r.vs  $(X_t, X_{t+\tau})^T$  are identically distributed for all t and  $\tau$ . That is,

$$(X_t, X_{t+\tau})^{\mathrm{T}} \stackrel{d}{=} (X_1, X_{1+\tau})^{\mathrm{T}}.$$

Follows from the definition for n = 2.

- 3.  $X_t$  is a weakly stationary TS if  $E(X_t^2) < \infty$  for all t. By (1)  $E X_t$  is constant, does not depend on t. If  $E(X_t^2) < \infty$  then all covariances also exist (by Cauchy-Schwarz inequality). By (2)  $\gamma(\tau) = cov(X_t, X_{t+\tau}) = cov(X_1, X_{1+\tau})$ , which also does not depend on t.
- 4. Weak stationarity does not imply strict stationarity. To prove this property we show an example of a weakly stationary TS which is not strictly stationary. Let  $Z_t \sim_{iid} \mathcal{N}(0, 1)$ . Define

$$X_t = \begin{cases} Z_t, & \text{if } t \text{ is even} \\ \frac{1}{\sqrt{2}}(Z_{t-1}^2 - 1) & \text{if } t \text{ is odd.} \end{cases}$$

Then

$$E X_t = \begin{cases} E Z_t = 0, & \text{if } t \text{ is even} \\ E \left[ \frac{1}{\sqrt{2}} (Z_{t-1}^2 - 1) \right] = \frac{1}{\sqrt{2}} E[Z_{t-1}^2 - 1] = 0 & \text{if } t \text{ is odd.} \end{cases}$$

Also,

$$\operatorname{var}(X_t) = \begin{cases} \operatorname{var}(Z_t) = 1, & \text{if } t \text{ is even,} \\ \operatorname{var}(\frac{1}{\sqrt{2}}(Z_{t-1}^2 - 1)) = \frac{1}{2}\operatorname{var}(Z_{t-1}^2) = 1 & \text{if } t \text{ is odd,} \end{cases}$$

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and by part 2 of Theorem 3.2 we obtain

$$\operatorname{cov}(X_t, X_{t+\tau}) = 0.$$

Hence  $X_t$  is a weakly stationary TS, it is in fact WN(0, 1). Is it identically distributed? We will compare  $P(X_t < x_t)$  for t even and odd. We have

$$P(X_t < x_t) = P(Z_t < x_t)$$
 for t even.

For t odd we get

$$P(X_t < x_t) = P\left(\frac{1}{\sqrt{2}}(Z_{t-1}^2 - 1) < x_t\right)$$
  
=  $P\left(Z_{t-1}^2 < \sqrt{2}x_t + 1\right)$   
=  $P\left(-\sqrt{\sqrt{2}x_t + 1} < Z_{t-1} < \sqrt{\sqrt{2}x_t + 1}\right).$ 

Take x = 0. Then

$$P(X_t < 0) = \begin{cases} 0.5, & \text{if } t \text{ is even,} \\ 0.6826, & \text{if } t \text{ is odd.} \end{cases}$$

Hence the c.d.fs are different for t even and t odd, that is series  $X_t$  is not strictly stationary.

5. *An i.i.d. TS is strictly stationary.* For an i.i.d. TS the joint cdf is

$$P(X_{t_1+\tau} < x_1, \dots, X_{t_n+\tau} < x_n) = F(x_1) \cdot \dots \cdot F(x_n).$$

So, it does not depend on the choice of the indices  $\{t_1, \ldots, t_n\}$ .