

Short introduction to ARMA processes without nonsense. The goal is to state all the theorems rigorously.

**Definition 1.** The process  $(u_t)$  is called white noise if

$$E(u_t) = 0, \quad \text{Var}(u_t) = \sigma^2, \quad \text{Cov}(u_s, u_t) = 0 \text{ for } s \neq t.$$

This definition does not assume that  $u_t$  and  $u_s$  are independent. They may be dependent but uncorrelated.

This definition does not assume normality of  $u_t$  but normality of white noise is often assumed in maximum likelihood estimation.

**Definition 2.** Lag operator  $L$  transforms a stochastic process  $(y_t)$  with  $t \in \mathbb{Z}$  into a new stochastic process by shifting the index back in time,

$$Ly_t = y_{t-1}.$$

**Definition 3.** Forward operator  $F$  transforms a stochastic process  $(y_t)$  with  $t \in \mathbb{Z}$  into a new stochastic process by shifting the index forward in time,

$$Ly_t = y_{t+1}.$$

Simple arithmetic examples are:

$$(1 + 2L + 3L^2)y_t = y_t + 2y_{t-1} + 3y_{t-2},$$

$$(3 + 2F + 5F^2)y_t = 3y_t + 2y_{t+1} + 5y_{t+2},$$

**Teopema 4.** The operators  $L$  and  $F$  are linear and  $L^{-1} = F$ .

*Proof.* The action  $LF$  or  $FL$  does nothing with any process  $(y_t)$ . So operators  $L$  and  $F$  are mutually inverse.  $\square$

**Definition 5.** The process  $(y_t)$  is called stationary in weak sense if

$$E(y_t) = \mu, \quad \text{Cov}(u_s, u_t) = \gamma(t - s).$$

In particular all variances of stationary process are equal,  $\text{Var}(y_t) = \text{Cov}(y_t, y_t) = \gamma_0$ .

When infinite sums do exist?

We *define* division by monomials.

**Definition 6.** For  $|\alpha| < 1$  we define

$$\frac{1}{1 - \alpha L} y_t = (1 + \alpha L + \alpha^2 L^2 + \alpha^3 L^3 + \dots) y_t,$$

and

$$\frac{1}{1 - \alpha F} y_t = (1 + \alpha F + \alpha^2 F^2 + \alpha^3 F^3 + \dots) y_t.$$

**Теорема 7.** If  $(u_t)$  is a white noise and  $|\alpha| < 1$  then  $\frac{1-\alpha L}{1-\alpha F}u_t$  and  $\frac{1-\alpha F}{1-\alpha L}u_t$  are white noises.

**Теорема 8.** The equation

$$P(L)y_t = Q(L)u_t + c,$$

where  $(u_t)$  is a white noise has infinitely many non-stationary solutions  $(y_t)$  if degree of  $P$  is higher than one.

**Теорема 9.** Consider the equation

$$P(L)y_t = Q(L)u_t + c.$$

If polynomials  $P$  and  $Q$  are coprime then

1. There are no stationary solutions  $(y_t)$  at all if  $P$  has at least one root  $\ell$  with  $|\ell| = 1$ .
2. There is exactly one stationary solution  $(y_t)$  if all roots  $\ell$  of  $P$  have  $|\ell| \neq 1$ .

There are two subcases when all roots  $\ell$  of  $P$  have  $|\ell| \neq 1$ :

1. All roots  $\ell$  of  $P$  have  $|\ell| > 1$ . In this case the unique stationary solution has the form

$$y_t = \mu + u_t + c_1 u_{t-1} + c_2 u_{t-2} + c_3 u_{t-3} + \dots,$$

where  $(u_t)$  is the white noise from original equation.

2. At least one root of  $P$  has  $|\ell| < 1$ . In this case the unique stationary solution has the form

$$y_t = \mu + \nu_t + c_1 \nu_{t-1} + c_2 \nu_{t-2} + c_3 \nu_{t-3} + \dots,$$

where  $(\nu_t)$  is a white noise different from  $(u_t)$ .

**Definition 10.** The process  $(y_t)$  is called  $ARMA(p, q)$  process with equation

$$P(L)y_t = Q(L)u_t + c,$$

if

1. the process  $(y_t)$  satisfies this equation;
2. polynomial  $P(L)$  has degree  $p$  and polynomial  $Q(L)$  has degree  $q$ ;
3.  $P(0) = Q(0) = 1$ ;
4.  $P$  and  $Q$  are coprime, in other words they have no common roots.
5. the process  $(y_t)$  can be represented in  $MA(\infty)$  form with respect to  $(u_t)$ :

$$y_t = \mu + u_t + c_1 u_{t-1} + c_2 u_{t-2} + c_3 u_{t-3} + \dots$$

From the last requirement in this definition it follows that all  $ARMA(p, q)$  processes are stationary. By definition. Point.

Not all solutions of equation  $P(L)y_t = Q(L)u_t + c$  are called  $ARMA$  processes.

**Definition 11.** The equation

$$P(L)y_t = Q(L)u_t + c,$$

is called *invertible* if all roots  $\ell$  of  $Q$  have  $|\ell| > 1$ .

Stationarity is the property of a process, invertibility is the property of equation. One cannot check whether a given sequence of random variables is invertible.