

3. LINEAR TIME-INVARIANT SYSTEMS

(Linear transfer function models)



For simplicity: • dimension = 1 (otherwise $\underline{u}_t, \underline{x}_t$ are vectors)
 $t \in T$ where $T = \mathbb{R}$ or $T = \mathbb{Z}$

$u, X: T \rightarrow \mathbb{R}$ (or \mathbb{C}) ... deterministic } functions, $u(t) := u_t; X(t) := X_t$
 $T \rightarrow$ random var. ... random } over scalar field \mathbb{R} or \mathbb{C} ($=: \mathbb{F}$)

$u \in \mathcal{F}_1$... space of inputs } normed linear spaces (NL-spaces)
 $X \in \mathcal{F}_2$... space of outputs } shift-invariant: $y(\cdot) \in \mathcal{F}_i \Rightarrow y(\cdot - a) \in \mathcal{F}_i \forall a \in T$

$\|\cdot\|_i$: ... norm on \mathcal{F}_i and $\xrightarrow{\mathcal{F}_i}$ convergence in norm: $y_n \xrightarrow{\mathcal{F}_i} y \equiv \|y - y_n\|_i \rightarrow 0$ as $n \rightarrow \infty$.
 $\|\cdot\|_i$: ... norm on $u(T)$ or $X(T)$ (or in another sense) usually abs. value.

3.1. Definition

A mapping $LS: \mathcal{F}_1 \rightarrow \mathcal{F}_2$ is called linear (time-invariant, shift-invariant) system (LTI-system) if the following holds:

! (1) Linearity: $LS(u_1 + u_2) = LS(u_1) + LS(u_2) \quad \forall u_1, u_2 \in \mathcal{F}_1$
 $LS(c \cdot u) = c \cdot LS(u) \quad \forall u \in \mathcal{F}_1, c \in \mathbb{F}$

! (2) Shift-invariance (time-invariance):
 $LS(u) = X \Rightarrow LS(u(\cdot - a)) = X(\cdot - a)$
 $\forall u \in \mathcal{F}_1$ and $a \in T$.

! (3) Continuity: $u_n \xrightarrow{\mathcal{F}_1} u \Rightarrow LS(u_n) \xrightarrow{\mathcal{F}_2} LS(u)$

Moreover LS is called

! (4) Causal if $\forall t \in T: u_1(\tau) = u_2(\tau) \quad \forall \tau \leq t \Rightarrow LS(u_1)(t) = LS(u_2)(t)$
 ($X(t) = LS(u)(t)$ depends only on input values $u(\tau), \tau \leq t$)

! (5) stable if $\|u(t)\| \leq C_1 < \infty \Rightarrow \exists C_2 < \infty: \|LS(u)(t)\| \leq C_2 \quad \forall t$
 (bounded inputs produce bounded outputs).

$T = \mathbb{R}$ } $\Rightarrow LS$ is called { continuous-time linear system
 $T = \mathbb{Z}$ } { discrete-time - " -

3.2. Example of continuous-time LS ($T = \mathbb{R}$)

(3.1) $x(t) \stackrel{a.e.}{=} \int_{-\infty}^{\infty} u(t-\tau) h(\tau) d\tau \equiv x = LS_h(u); h \dots$ fixed
 $\mathcal{F}_1, \mathcal{F}_2 \dots$ suitable function spaces ($\mathcal{F}_i := L_p, 1 \leq p < \infty$)
 integral convolution: $x = u * h$

3.3. Discretization of (3.1) with time step Δt

$u(t) \approx u_m$ for $m\Delta t \leq t < (m+1)\Delta t$... piecewise constant function

$$\begin{aligned} \underline{x_n} := x(n\Delta t) &\stackrel{(3.1)}{=} \int_{-\infty}^{\infty} u(n\Delta t - \tau) h(\tau) d\tau = \sum_{k=-\infty}^{\infty} \int_{(k-1)\Delta t}^{k\Delta t} \underbrace{u(n\Delta t - \tau)}_{\substack{\downarrow \\ (n-k)\Delta t \leq n\Delta t - \tau < (n-k+1)\Delta t}} h(\tau) d\tau \approx \\ &\approx \sum_{k=-\infty}^{\infty} \int_{(k-1)\Delta t}^{k\Delta t} u_{n-k} h(\tau) d\tau \\ &= \sum_{k=-\infty}^{\infty} u_{n-k} \int_{(k-1)\Delta t}^{k\Delta t} h(\tau) d\tau = \sum_{k=-\infty}^{\infty} u_{n-k} \underline{h_k} \quad \text{where } \underline{h_k} := \int_{(k-1)\Delta t}^{k\Delta t} h(\tau) d\tau \end{aligned}$$

\Downarrow usually with $\Delta t = 1$

3.4. Example of discrete-time LS ($T = \mathbb{Z}$)

! (3.2) $x(t) = \sum_{k=-\infty}^{\infty} u(t-k) h(k) \equiv x = LS_h(u)$ where $u, x, h \dots$ sequences provided that the series converges in some sense.

! Discrete Linear Convolution (DLC): $x = u * h$

$\mathcal{F}_1, \mathcal{F}_2 \dots$ suitable spaces of sequences, typically:

$$\mathcal{F}_i = \ell_p = \left\{ \{x(t)\}_{t \in \mathbb{Z}} \mid \sum_{t=-\infty}^{\infty} |x(t)|^p < \infty \right\}, \quad 1 \leq p < \infty$$

$$\|x\|_i := \|x\|_p := \sqrt[p]{\sum_{t=-\infty}^{\infty} |x(t)|^p}; \quad x_n \xrightarrow{\mathcal{F}_i} x \equiv \|x - x_n\|_i \rightarrow 0 \text{ as } n \rightarrow \infty.$$

$p=2 \dots$ Euclidean norm as an analog of vector norm for sequences.

knonecker symbol

! 3.5. Notation

$\delta := \{\delta(t)\}_{t \in \mathbb{Z}}$ where $\delta(t) = \delta_{0,t} = \begin{cases} 1 & \text{for } t=0 \\ 0 & \text{otherwise} \end{cases} \dots$ Dirac sequence

$\{\delta(t-k)\}_{t \in \mathbb{Z}} \dots$ shifted Dirac sequence: $\delta(t-k) = \begin{cases} 1 & \text{for } t=k \\ 0 & \text{otherwise} \end{cases}$

$x^- := \{x(-t)\}_{t \in \mathbb{Z}}$ for $x = \{x(t)\}_{t \in \mathbb{Z}} \dots$ reversed sequence to x

3.6. Theorem

$h \in \ell_1, u \in \ell_p \Rightarrow (3.2)$ converges in usual sense for each $t \in \mathbb{Z}$ and $x \in \ell_p$. Moreover it holds: $\|x\|_p \leq \|u\|_p \cdot \|h\|_1$, which confirms continuity (3). As the validity of (1) and (2) is evident, $LS_h: \ell^p \rightarrow \ell^p$ is a linear system for each $h \in \ell_1$.

3.7. Theorem

If (3.2) converges in usual sense for each $t \in \mathbb{Z}$ and $LS_h: \mathcal{F}_1 \rightarrow \mathcal{F}_2$ defines a linear system, then

(1) $h \in \ell_1$, i.e. $\sum_{k=-\infty}^{\infty} |h(k)| < \infty$ (so-called stability condition),

\Downarrow
 LS_h is stable.

(2) LS_h is causal $\Leftrightarrow h(k) = 0$ for all $k < 0$.

In such a case (3.2) attains the form:

$$(3.2') \quad x(t) = \sum_{k=0}^{\infty} u(t-k) h(k)$$

3.8. Corollary

$h \in \ell_1, \mathcal{F}_1 = \mathcal{F}_2 = \ell_p \Rightarrow LS_h: \ell_p \rightarrow \ell_p$ is a stable linear system.

3.9. Theorem

Let $LS: \mathcal{F}_1 \rightarrow \mathcal{F}_2$ be a discrete linear system, \mathcal{F}_1 and \mathcal{F}_2 spaces of number sequences such that $\delta \in \mathcal{F}_1$ and \mathcal{F}_2 is either convergence in usual sense (point convergence) or convergence in norm $\|\cdot\|_p$ in case $\mathcal{F}_1 = \mathcal{F}_2 = \ell_p$.

Then there exists unique $h \in \mathcal{F}_2: LS = LS_h$ by (3.2) where $h = LS(\delta)$.

3.10. Corollary

$LS: \ell_p \rightarrow \ell_p$ arbitrary $\Rightarrow \exists!$ $h \in \ell_p, h = LS(\delta): LS = LS_h$.

3.11. Definition

Every discrete linear system as of theorem 3.9 is uniquely determined by the sequence $h = LS(\delta)$ which is called its impulse response.

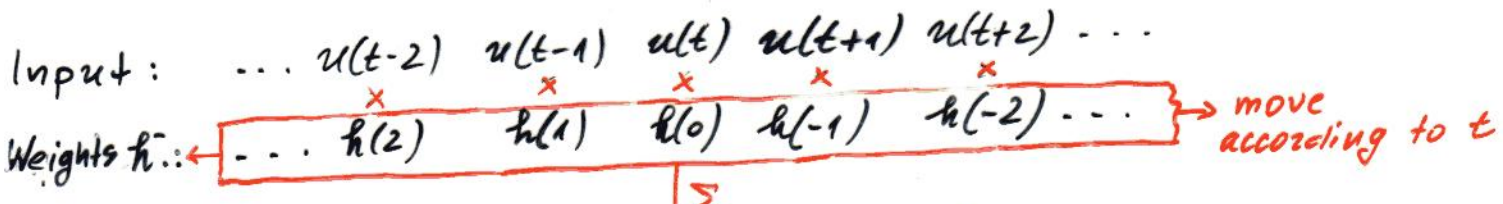
h finite ($h(k) \neq 0$ only for a finitely many $k \in \mathbb{Z}$)... LS is said to be finite impulse response linear system (FIR-LS)

h infinite... LS is said to be an infinite impulse response linear system (IIR-LS)



3.12. Example

Discrete linear filter = Weighted moving average = LSR for suitable h
Geometric visualization of (3.2):



$$x(t) = \sum_{k=-\infty}^{\infty} u(t-k) h(k) = \sum_{k=-\infty}^{\infty} u(t+k) \underbrace{h(-k)}_{h^-(k)}$$

$x(t)$ = weighted mean from values $u(t \pm k)$ using weights $h^-(k)$

Simple Moving average: $h = h^- = \{ \dots, 0, \underbrace{\frac{1}{m}, \dots, \frac{1}{m}}_m \text{-times}, 0, \dots \}$

3.13. Definition (z-transform)

$h \rightsquigarrow H(z) := \sum_{k=-\infty}^{\infty} h(k) z^k, z \in \mathbb{C} \dots$ formal Laurent series
 $h^- \rightsquigarrow H^-(z) = \sum_{k=-\infty}^{\infty} h(-k) z^k = \sum_{k=-\infty}^{\infty} h(k) z^{-k} = H(\frac{1}{z}) \dots$ z-transform of h^-

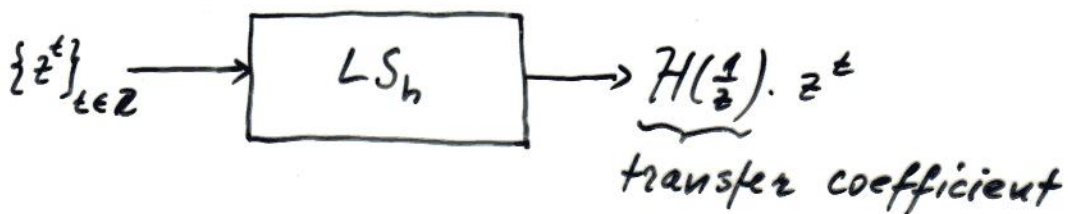
3.14. Theorem

- (1) LSR is causal $\Leftrightarrow H(z) = \sum_{k=0}^{\infty} h(k) z^k$ is formal power series (annulus)
- (1') LSR is causal FIR-LS $\Leftrightarrow H(z)$ is a polynomial.
- (2) LSR stable by $h \in \ell_1 \Leftrightarrow H(z)$ is absolutely convergent on a ring $\{z \mid r_1 < |z| < r_2\} =: C(r_1, r_2)$ for some $0 < r_1 < 1 < r_2 < \infty$ (a ring containing unit circle $\{z \mid |z| = 1\} =: C(1)$).
- (3) LSR both causal and stable (by $h \in \ell_1$) $\Leftrightarrow H(z)$ is a power series absolutely convergent on a disc $\{z \mid |z| < r\} =: D(r)$ for some $r > 1$ (a disc containing unit circle).
- (4) LSR stable by $h \in \ell_1 \Rightarrow$
 - (i) LSR $(z^{-t}) = H(z) z^{-t} \forall z \in C(r_1, r_2)$ as of (2) or $z \in D(r)$ as of (3)
 - (ii) LSR $(z^t) = H(\frac{1}{z}) z^t \forall z \in C(\frac{1}{r_2}, \frac{1}{r_1})$ or $z \in \{z \mid |z| > \frac{1}{r}\}$

where r_1, r_2 are as of (2) or r is as of (3).
- (5) $h(k) \in \mathbb{R} \forall k \in \mathbb{Z} \Rightarrow H(\bar{z}) = \overline{H(z)}$ and $H(\frac{1}{z}) = \overline{H(\frac{1}{\bar{z}})}$.
- (6) $x = u * h \Leftrightarrow X(z) = U(z) \cdot H(z) : U(z)$ abs. conv on $G_1, H(z)$ on $G_2 \Rightarrow X(z)$ on $G_1 \cap G_2$

3.15. Definition

z-transform $H(\frac{1}{z})$ (or alternatively $H(z)$ - it is a matter of convention) of the impulse response h of a linear system LS_h is called its transfer function.



3.16. Theorem (Transfer of sine/cosine waves through a stable LS_h)

Let LS_h be a stable ($h \in \ell_1$) linear system with the transfer function $H(\frac{1}{z})$. Then the following holds ($c \in \mathbb{R}$ or $c \in \mathbb{C}$):

(3.3) $LS_h(c \cdot e^{i\omega t}) = H(e^{-i\omega}) \cdot c \cdot e^{i\omega t}$

(3.3') $LS_h(c \cdot e^{i(\omega t - \varphi)}) = |H(e^{-i\omega})| \cdot c \cdot e^{i(\omega t - (\varphi + \psi))}$

where $H(e^{-i\omega}) = |H(e^{-i\omega})| \cdot e^{-i\psi}$
 $|H(e^{-i\omega})| \dots$ amplitude transfer coefficient
 $\psi \dots$ transfer phase shift

(3.3'') $LS_h(A \cdot \cos(\omega t - \varphi)) = |H(e^{-i\omega})| \cdot A \cdot \cos(\omega t - (\varphi + \psi))$, $A \in \mathbb{R}$, $h(k) \in \mathbb{R}$
↑
transfer phase shift

(3.3''') $LS_h\{c\} = c \cdot \sum_{k=-\infty}^{\infty} h(k)$

$\omega = 2\pi f$ is angle velocity ($f =$ frequency) } of the input wave function
 $\varphi =$ is phase shift

3.17. Definition

A causal and stable LS_h , $h \in \ell_1$, is called a recursive linear system (or ARMA-LS) of order (p, q) , $p > 0, q \geq 0$ integers, if there exist polynomials: $\Theta(z) = \Theta_0 + \Theta_1 z + \dots + \Theta_q z^q$, $\Theta_q \neq 0$ and $\Phi(z) = \Phi_0 - \Phi_1 z - \dots - \Phi_p z^p$, $\Phi_0 \neq 0, \Phi_p \neq 0$ such that $H(z) = \frac{\Theta(z)}{\Phi(z)}$. [We can assume without the loss of generality, that $\Theta(z)$ and $\Phi(z)$ have no common roots and that $\Phi_0 = 1$].

3.18. Theorem

LS_n recursive $\Rightarrow \phi(z)$ has all roots outside of the unit disc $\{z \mid |z| \leq 1\}$ of order (p, q) [or: $z \rightarrow \phi(\frac{1}{z})$ has all roots inside of — " —]

and for each $u \in \ell_1$ is $x = LS_n(u)$ iff the following relation holds:

$$(3.4) \quad x_t = \underbrace{\phi_1 x_{t-1} + \dots + \phi_p x_{t-p}}_{\substack{\text{AR}(p) \text{ component} \\ \uparrow \text{Autoregression} \\ = \text{feedback}}} + \underbrace{\theta_0 u_t + \dots + \theta_q u_{t-q}}_{\substack{\text{MA}(q) \text{ component} \\ \uparrow \text{Moving Average}}}$$

In such a case $x \in \ell_1$ as well.

Proof

I. $h \in \ell_1$ by definition 3.17, $u \in \ell_1$ by assumption $\Rightarrow x \in \ell_1$ by 3.6 where $p=1$.

II. LS_n recursive $\Rightarrow LS_n$ causal and stable with $H(z) = \frac{\theta(z)}{\phi(z)} \stackrel{3.14(3)}{\Rightarrow}$ $H(z)$ is a power series absolutely convergent for $|z| \leq 1 \Rightarrow \phi(z) \neq 0$ for all $z: |z| \leq 1$ (otherwise such root would be pole of $H(z)$ which contradicts absolute convergence) \Rightarrow all roots of $\phi(z)$ lie outside of the unit disc.

Let $u \in \ell_1$ be arbitrary. As $h \in \ell_1$, both $\sum_{t=-\infty}^{\infty} |u(t)| < \infty$ and $\sum_{k=0}^{\infty} |h(k)| < \infty$ are absolutely convergent $\stackrel{3.14(2)}{\Rightarrow} U(z)$ is abs. conv. on $C(r_1, r_2)$ and $H(z)$ on $D(r)$ where we may assume without the loss of generality that $0 < r_1 < 1 < r_2 = r$. It is well-known from the calculus that their Cauchy product $X(z) := U(z) \cdot H(z)$ is a series (see 3.14(6))

$$X(z) = \sum_{t=-\infty}^{\infty} x(t) z^t \text{ absolutely convergent at least on } C(r_1, r_2)$$

where $x(t) = \sum_{k=0}^{\infty} u(t-k) h(k)$, i.e. $x = u * h$ (convolution):

$$\left(\sum_{l=-\infty}^{\infty} u(l) z^l \right) \cdot \left(\sum_{k=0}^{\infty} h(k) z^k \right) \stackrel{(*)}{=} \sum_{l=-\infty}^{\infty} \sum_{k=0}^{\infty} u(l) \cdot h(k) z^{l+k} = \sum_{t=-\infty}^{\infty} \underbrace{\left(\sum_{k=0}^{\infty} u(t-k) h(k) \right)}_{x(t)} z^t$$

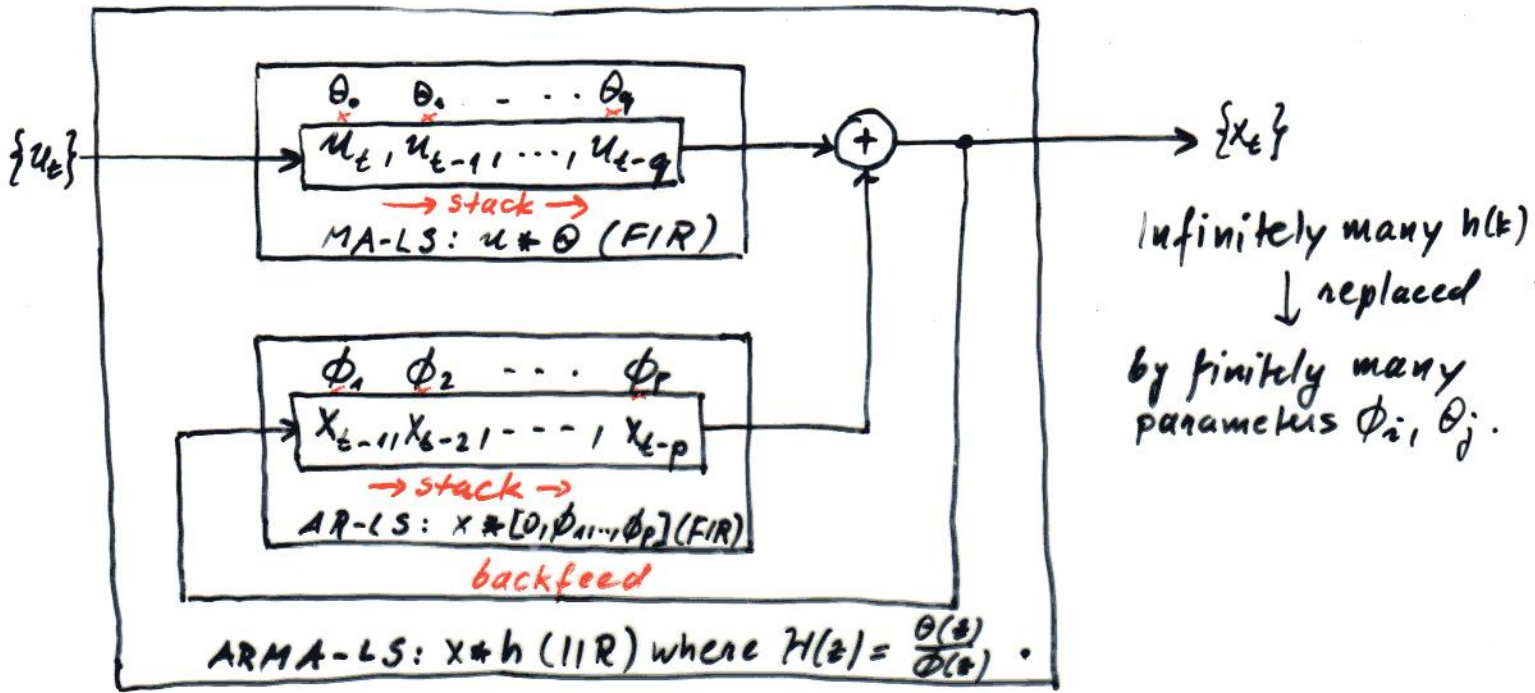
Then the following equivalence relations clearly hold for all $z \in C(r_1, r_2)$:

$$x = LS_n(u) \Leftrightarrow x = u * h \Leftrightarrow X(z) = U(z) \cdot H(z) = U(z) \cdot \frac{\theta(z)}{\phi(z)} \stackrel{(\dagger)}{=} X(z) \cdot \phi(z)$$

$$X(z) \cdot \phi(z) = U(z) \cdot \theta(z) \stackrel{\text{by } (*)}{\Leftrightarrow} x * \phi = u * \theta \Leftrightarrow$$

$$\phi_0 x_t - \phi_1 x_{t-1} - \dots - \phi_p x_{t-p} = \theta_0 u_t + \theta_1 u_{t-1} + \dots + \theta_q u_{t-q} \Leftrightarrow (3.4) \quad \square$$

Note \odot We must have $\phi(z) \neq 0$ for $z \in C(r_1, r_2)$. Indeed, all roots z_0 of $\phi(z)$ satisfy $|z_0| > 1$ and as $H(z) = \frac{\theta(z)}{\phi(z)}$ has no poles in $C(r_1, r_2)$ clearly $|z_0| \geq r_2 = r > 1$ must hold.



Note: Every causal and stable IIR LS_n may be with arbitrary precision approximated by a suitable ARMA-LS.

Higher precision \Rightarrow higher orders $p, q \Rightarrow$ more parameters ϕ_i, θ_j needed.

Analogy: any real number $h \in \mathbb{R}$ with infinitely many decimal digits may be with arbitrary precision approximated by a rational number $\frac{\theta}{\phi}$ where θ, ϕ have finite number of digits (ϕ, θ are integers).

Observe that $h = \sum_{k=-\infty}^{\infty} h_k 10^k$ is Laurent expansion similar to $H(z)$, and $\theta = \sum_{i=0}^q \theta_i 10^i, \phi = 1 - \sum_{i=1}^p \phi_i 10^i$ finite power expansion similar to $\theta(z), \phi(z)$, here playing the role of decadic representations.

3.20. Alternative notation for $x_t = LS_h(u_t), h \in \ell_1$

(1) By (3.2): $x_t = \sum_{k=-\infty}^{\infty} h_k u_{t-k}$ or $x = h * u$... explicit form

(2) By 3.14(6): $X(z) = H(z) \cdot U(z)$ or with z-transforms: $X(\frac{1}{z}) = U(\frac{1}{z}) H(\frac{1}{z})$

(3) Using backward-shift operator $B: B u_t := u_{t-1}$ and forward-shift operator $B^{-1}: B^{-1} u_t := u_{t+1}$ preferred by electrical engineers

$$x_t = \sum_{k=-\infty}^{\infty} h_k u_{t-k} = \sum_{k=-\infty}^{\infty} h_k B^k u_t = \left(\sum_{k=-\infty}^{\infty} h_k B^k \right) u_t = H(B) u_t$$

$x_t = H(B) u_t$... Preferred notation in ARMA time series modeling

In particular for ARMA-LS we obtain:

- (1) $x_t - \phi_1 x_{t-1} - \dots - \phi_p x_{t-p} = \theta_0 u_t + \theta_1 u_{t-1} + \dots + \theta_q u_{t-q}$
 - (2) $\phi(z) X(z) = \theta(z) U(z)$
 - (3) $\phi(B) x_t = \theta(B) u_t$
- where $\phi(z) = 1 - \phi_1 z - \dots - \phi_p z^p$
 $\theta(z) = \theta_0 z + \theta_1 z^2 + \dots + \theta_q z^q$