



Stochastic signals and systems

Lecture 2.

**Prediction, innovation
and
the Wold decomposition**

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REMINDER



Wide sense stationary processes

Random variables: defined over a probability space (Ω, \mathcal{F}, P) .

A discrete time stochastic process: $y = (y_n)$, with $-\infty < n < +\infty$.

$y = (y_n)$ is wide sense stationary, w.s.st. for short, if

$$E y_n = 0, \quad E(y_n^2) < +\infty$$

and for any fixed τ the auto-covariances below are independent of n :

$$r(\tau) = \text{Cov}(y_{n+\tau}, y_n).$$



The auto-covariance matrix R

Let $R := (R_{j,k}) = r(k-j) \quad k, l = 1, \dots, p.$

For any p the auto-covariance matrix R is a symmetric **Toeplitz matrix**:

$$R = \begin{pmatrix} r(0) & r(1) & r(2) & \dots & r(p-1) \\ r(-1) & r(0) & r(1) & \dots & r(p-2) \\ r(-2) & r(-1) & r(0) & \dots & r(p-3) \\ \vdots & \vdots & \ddots & \ddots & r(1) \\ r(-p+1) & r(-p+2) & \dots & r(-1) & r(0) \end{pmatrix}$$

Note that the R is a Toeplitz matrix.

Recall: set $Y = (y_{n-1}, \dots, y_{n-p})^T$. Show that $R = \mathbb{E}(YY^T) \geq 0$.



Prediction based on finite past

Predict y_n based on y_{n-1}, \dots, y_{n-p} . Let $r := (r(1), \dots, r(p))^T$.

Proposition. If R is nonsingular, then the LSQ linear prediction of y_n in terms of y_{n-1}, \dots, y_{n-p} is uniquely defined as

$$\hat{y}_{n,n-p} := \sum_{k=1}^p \alpha_k y_{n-k},$$

where $\alpha = (\alpha_1, \dots, \alpha_p)^T$ is the solution of the normal equation

$$R\alpha = r.$$

Exercise. Show that the prediction $\hat{y}_{n,n-p}$ is a w.s.st. process.

Question: what happens if we start increasing p and $p \rightarrow \infty$?



Linear operations on orthogonal processes

Let (e_n) be a real-valued w.s.st. orthogonal process and let us define

$$y_n = c_1 e_{n-1} + \cdots + c_p e_{n-p}, \quad \text{with } c_k \text{ real.}$$

The process (y_n) is a moving average process. It is w.s.st. and

$$\mathbb{E} y_n^2 = \sum_{k=1}^p c_k^2 \sigma^2(e).$$

What happens if $p \rightarrow \infty$?

Exercise. Show that the process

$$y_n = \sum_{k=1}^{\infty} c_k e_{n-k} \quad \text{under} \quad \sum_{k=1}^{\infty} c_k^2 < \infty$$

is well-defined and it is a w.s.st. process. (*Hint:* See next exercise).

Exercise. Let $(v_{n,p})$ be a w.s.st. process for all p , and let $v_n = \lim_p v_{n,p}$ exist for all n . Show that (v_n) is w.s.st.



PREDICTION



Prediction based on the infinite past, I.

Consider the apparently impractical prediction problem with $p = \infty$.

Consider the infinite dimensional linear space spanned by $(y_{n-1}, y_{n-2}, \dots)$:

$$\mathcal{L}_{n-1} = \left\{ u : \sum_{k=0}^p \alpha_k y_{n-k}, \quad \alpha_k \in \mathbb{R} \quad \text{for some } p \right\}$$

Let H_{n-1} be the closure of \mathcal{L}_{n-1} in the metric of $L_2(\Omega, \mathcal{F}, P)$:

$$H_{n-1} := \text{cl } \mathcal{L}_{n-1}$$

Then the one-step ahead LSQ prediction of y_n is given by

$$\hat{y}_n = (y_n | H_{n-1}),$$

the orthogonal projection of y_n on the subspace H_{n-1} .



Prediction based on the infinite past, II.

Consider now the linear space spanned by $(y_{n-1}, \dots, y_{n-p})$:

$$\mathcal{L}_{n-1, n-p} = \left\{ \sum_{k=1}^p \alpha_k y_{n-k} : \alpha_1, \dots, \alpha_p \in \mathbb{R}, \right\}.$$

Obviously, $\mathcal{L}_{n-1, n-p}$ is a finite dimensional subspace of H_{n-1} .

Hence it is a Hilbert subspace of H_{n-1} . Write $\mathcal{L}_{n-1, n-p} =: H_{n-1, n-p}$.

Proposition

We have, with convergence meant in $L_2(\Omega, \mathcal{F}, P)$,

$$(y_n | H_{n-1}) = \lim_{p \rightarrow \infty} (y_n | H_{n-1, n-p}).$$

Prediction based on finite past approximates $(y_n | H_{n-1})$ arbitrarily well.



Remarks on Hilbert space geometry, I.

In a more general setting: let G be a Hilbert space and let $G_p \subset G$ be a sequence of closed sub-spaces of G s.t. $G_p \subset G_{p+1}$ with $-\infty < p < +\infty$. I.e. (G_p) is monotone increasing. Let

$$G_\infty = \text{cl} (\cup_p G_p).$$

with cl denoting the closure.

Lemma

Under the conditions above for any $x \in G$ we have

$$\lim_{p \rightarrow \infty} (x|G_p) = (x|G_\infty).$$

Apply: $G_p = H_{n-1, n-p}$ and $G_\infty = H_{n-1}$. Projection on **expanding** past.



Remarks on Hilbert space geometry, II.

Proof. First we show, replacing p by n , that for any $x \in G$

$$(x|G_n) = ((x|G_\infty)|G_n). \quad (1)$$

Cf. a projection on the (x, y) plane followed by projection to the x axis.

Indeed, let $(x|G_\infty) = y$, and take an orthogonal decomposition

$$x = y + \Delta x, \quad (2)$$

where $\Delta x \perp G_\infty$. Then also $\Delta x \perp G_n$ for any n .

Projecting both sides of the equality (2) onto G_n we get

$$(x|G_n) = (y|G_n).$$



Remarks on Hilbert space geometry, III.

Repeat: projecting both sides of the equality (2) onto G_n we get

$$(x|G_n) = (y|G_n).$$

But $y \in G_\infty = \text{cl}(\cup G_n)$ implies that

$$\lim_{n \rightarrow \infty} (y - (y|G_n)) = 0.$$

Substituting $y = (x|G_\infty)$ and $(y|G_n) = (x|G_n)$ we get the claim:

$$\lim_{n \rightarrow \infty} ((x|G_\infty) - (x|G_n)) = 0.$$



A dual result, I.

Let G be and $G_m \subset G$ be as above, so that $G_m \subset G_{m+1}$, and let

$$G_{-\infty} = \bigcap_m G_{-m}.$$

Recall that $-\infty < m < +\infty$. Note that $G_{-\infty}$ is automatically closed.

Lemma

Under the conditions above for any $x \in G$ we have

$$\lim_{m \rightarrow \infty} (x|G_{-m}) = (x|G_{-\infty}).$$

Apply with $G = L_2(\Omega, \mathcal{F}, P)$, and $G_m = H_m$. Projection on **shrinking** past:

$$\lim_{p \rightarrow \infty} (x|H_{n-p}) = (x|H_{-\infty}). \quad (3)$$



Proof of the dual result

Let us denote the orthogonal complement of G_{-m} in G by G_{-m}^\perp , i.e.

$$G = G_{-m} \oplus G_{-m}^\perp.$$

From here we get

$$(x|G_{-m}) = x - (x|G_{-m}^\perp).$$

Obviously, G_{-m}^\perp is monotone increasing. Let $F = \text{cl}(\cup_m G_{-m}^\perp)$. Then

$$\lim_{m \rightarrow \infty} (x|G_{-m}^\perp) = (x|F),$$

by Lemma 1. The lemma then follows from the next exercise.

Exercise. Show that $F = G_{-\infty}^\perp$. (*Hint:* Verify inclusion both way).



The innovation process, I.

Question: express the new information in y_n not contained in H_{n-1}

This is a key object in the theory of w.s.st. processes given below.

From now on we will use superscripts such as H_{n-1}^y for H_{n-1} .

Definition

The innovation process of (y_n) is defined as:

$$e_n := y_n - (y_n | H_{n-1}^y).$$



The innovation process, II.

Define the prediction error process based on finite past

$$e_{n,n-p} := y_n - (y_n | H_{n-1,n-p}).$$

Exercise. Prove that (e_n) is a w.s.st. process. (*Hint:* Recall that $\lim_{p \rightarrow \infty} e_{n,n-p} = e_n$).

Exercise HW Prove that (e_n) is a w.s.st. orthogonal process.



Autoregressive processes, I.

Assume that a **finite segment** of past values is sufficient to compute \hat{y}_n :

$$\hat{y}_n = (y_n | H_{n-1}^y) = \sum_{k=1}^p a_k y_{n-k}.$$

Noting that $e_n = y_n - \hat{y}_n$, we get

$$y_n = \sum_{k=1}^p a_k y_{n-k} + e_n.$$



Autoregressive processes II.

Repeat:

$$y_n = \sum_{k=1}^p a_k y_{n-k} + e_n. \quad (4)$$

Definition

A wide-sense stationary process $y = (y_n)$ satisfying (4), with (e_n) being its innovation process, is called an *autoregressive* or *AR process*.

If $a_p \neq 0$ then p is the **order** of the process, and y is an $AR(p)$ process.

Q: Under what conditions on the coefficients a_k does such a y exist ?



WOLD DECOMPOSITION



Wold decomposition, I.

Let us consider a process $y = (y_n)$ and write

$$y_n = e_n + (y_n | H_{n-1}^y).$$

Obviously, e_n and H_{n-1}^y span H_n^y , and $e_n \perp H_{n-1}^y$. Thus we have

$$H_n^y = H(e_n) \oplus H_{n-1}^y$$

$\forall n$, where $H(e_n)$ denotes the 1-dimensional linear space spanned by e_n , assuming $e_n \neq 0$, and \oplus denotes orthogonal direct sum of sub-spaces.

Consider now $(y_n | H_{n-1}^y)$. It is an element of H_{n-1}^y . Hence we can decompose it as a sum of orthogonal elements using

$$H_{n-1}^y = H(e_{n-1}) \oplus H_{n-2}^y.$$



Wold decomposition, II.

Repeat: decompose $(y_n|H_{n-1}^y)$ a sum of orthogonal elements using

$$H_{n-1}^y = H(e_{n-1}) \oplus H_{n-2}^y.$$

Note shift 1 time unit back. Now, $H(e_{n-1})$ is 1-dimensional, hence for any $v \in H(e_{n-1})$ we have $v = c e_{n-1}$ with some $c \in \mathbb{R}$.

Recall the identity on projecting on inclusive sub-spaces $H_{n-1}^y \supset H_{n-2}^y$:

$$((y_n|H_{n-1}^y)|H_{n-2}^y) = (y_n|H_{n-2}^y).$$

Combining the two remarks above we get the decomposition

$$(y_n|H_{n-1}^y) = c_1 e_{n-1} + (y_n|H_{n-2}^y).$$



Wold decomposition, III.

Repeat: we obtain the following decomposition for $(y_n | H_{n-1}^y)$:

$$(y_n | H_{n-1}^y) = c_1 e_{n-1} + (y_n | H_{n-2}^y).$$

Combining this with the definition of the innovation e_n we get

$$y_n = e_n + c_1 e_{n-1} + (y_n | H_{n-2}^y).$$

Iterating this decomposition we get, with $c_0 = 1$,

$$y_n = \sum_{k=0}^p c_k e_{n-k} + (y_n | H_{n-p-1}^y).$$



Wold decomposition, IV.

Repeat: iterating the one-step decomposition we get, with $c_0 = 1$,

$$y_n = \sum_{k=0}^p c_k e_{n-k} + (y_n | H_{n-p-1}^y).$$

To deal with the residual term define the **prehistory** of y as:

$$H_{-\infty}^y = \bigcap_{m \geq 0} H_{-m}^y.$$

Recall that by Lemma 1 we have, with convergence in $= L_2(\Omega, \mathcal{F}, P)$,

$$\lim_{m \rightarrow \infty} (y_n | H_{-m}^y) = (y_n | H_{-\infty}^y).$$

It follows that $\sum_{k=0}^p c_k e_{n-k}$ also converges when $p \rightarrow \infty$.



Wold decomposition, V.

Letting $p \rightarrow \infty$ we arrive at the following decomposition:

$$y_n = \sum_{k=0}^{\infty} c_k e_{n-k} + (y_n | H_{-\infty}^y). \quad (5)$$

Exercise. Show that $\sum_{k=0}^{\infty} c_k^2 < \infty$.

Consider the r.h.s. of (5) and define the processes

$$y_n^r = \sum_{k=0}^{\infty} c_k e_{n-k} \quad \text{and} \quad y_n^s = (y_n | H_{-\infty}^y).$$

Exercise. Show that both (y_n^r) and (y_n^s) are w.s.st. processes. (*Hint:* recall that $y_n^s = \lim_{p \rightarrow \infty} (y_n | H_{n-p})$).



Wold decomposition, VI.

Claim: the processes (y_n^r) and (y_n^s) are **orthogonal**, $y^s \perp y^r$, meaning:

$$y_n^r \perp y_m^s \quad \text{for all } n, m.$$

Proof: For any fixed p we have

$$\sum_{k=0}^p c_k e_{n-k} \perp H_{n-m}^y \quad \text{for } m > p.$$

But $H_{n-m}^y \supset H_{-\infty}^y$, hence we get for any p :

$$\sum_{k=0}^p c_k e_{n-k} \perp H_{-\infty}^y.$$

Letting $p \rightarrow \infty$ we get the claim.



Wold decomposition, VII.

What can we say about $y_n^s = (y_n | H_{-\infty}^y)$?

Proposition

The history $H_n^{y^s} =: H_n^s$ is independent of n , hence $H_n^s = H_{n-1}^s$ for all n .

Proof: Fix n . Note that $e_n \perp H_{-\infty}^y$. (Why ?) But then the known equality $H_n^y = H(e_n) \oplus H_{n-1}^y$, implies, by projecting all three terms onto $H_{-\infty}^y$,

$$(H_n^y | H_{-\infty}^y) = (H_{n-1}^y | H_{-\infty}^y).$$

To end the proof it is sufficient to show that for any m

$$(H_m^y | H_{-\infty}^y) = H_m^s.$$



Wold decomposition, VIII.

Repeat: to end the proof it is sufficient to show that for any m

$$(H_m^y | H_{-\infty}^y) = H_m^s.$$

Recall the definition of H_m^y :

$$H_m^y = \text{cl } \mathcal{L}_m^y = \left\{ u : \sum_{k=0}^p \alpha_k y_{m-k}, \quad \alpha_k \in \mathbb{R} \quad \text{for some } p \right\}.$$

Projecting all the elements of \mathcal{L}_m^y onto $H_{-\infty}^y$ we get, due to linearity:

$$(\mathcal{L}_m^y | H_{-\infty}^y) = \left\{ u : \sum_{k=0}^p \alpha_k y_{m-k}^s, \quad \alpha_k \in \mathbb{R} \quad \text{for some } p \right\} = \mathcal{L}_m^s !$$

Take the closure of both sides. On the r.h.s we get $\text{cl } \mathcal{L}_m^s = H_m^s$.

On the l.h.s. we get by the continuity of projections:

$$\text{cl } (\mathcal{L}_m^y | H_{-\infty}^y) = (\text{cl } \mathcal{L}_m^y | H_{-\infty}^y) = (H_m^y | H_{-\infty}^y), \quad \text{q.e.d.}$$



SINGULAR PROCESSES



Singular processes, I.

Let us have a closer look at $y_n^s = (y_n | H_{-\infty}^y)$. We have seen that for all n .

$$H_n^s = H_{n-1}^s.$$

Thus we get that the innovation process of y_n^s is identically 0:

$$e_n^s := y_n^s - (y_n^s | H_{n-1}^s) = 0!$$

Exercise. Verify the above equality. (*Hint:* Note that $y_n^s \in H_n^s$).

Intuition: a truly random process must have non-trivial innovations.

Definition

A w.s.st. stochastic process (v_n) is called singular, if its innovation process is identically 0: $v_n - (v_n | H_{n-1}^v) = 0$ for all n .



Singular processes, II.

Exercise. Show that if $H_n^y = H_{n-1}^y$ for a **single** n then it is true **for all** n .

How can we construct a singular process? Example: consider the process

$$y_n = \xi e^{in\omega}, \quad -\infty < n < +\infty,$$

where $\omega \in (0, 2\pi)$ is a **fixed** frequency, ξ is a **complex-valued r.v.** with

$$E\xi = 0, \quad E|\xi|^2 = \sigma^2 < +\infty.$$

Exercise. Show that the **complex-valued** process $y = (y_n)$ is w.s.st.:

$$r(\tau) := Ey_{n+\tau}\bar{y}_n = \sigma^2 e^{i\tau\omega}.$$

Note that $r(\tau) = \sigma^2 e^{i\tau\omega}$ does not decay in absolute value, as τ increases.



Singular processes III.

Consider now a finite sum of the above complex-valued w.s.st. processes:

$$y_n = \sum_{k=1}^m \xi_k e^{in\omega_k}. \quad (6)$$

Here $\omega_k \neq \omega_j$ for $k \neq j$, and

$$\mathbb{E}\xi_k = 0, \quad \mathbb{E}|\xi_k|^2 = \sigma_k^2 < +\infty.$$

Assume in addition that $\xi_j \perp \xi_k$ for $j \neq k$, i.e.

$$\mathbb{E}\xi_j \bar{\xi}_k = 0, \quad \text{for } k \neq j.$$



Singular processes, IV.

Repeat: consider the process $y = (y_n)$ given by $y_n = \sum_{k=1}^m \xi_k e^{in\omega_k}$.

Exercise. Show that the above **complex-valued** process y is w.s.st.:

$$r(\tau) := \mathbb{E}y_{n+\tau}\bar{y}_n = \sum_{k=1}^m \sigma_k^2 e^{i\tau\omega_k}.$$

The variance of y_n is obtained by setting $\tau = 0$:

$$\mathbb{E}|y_n|^2 = \sum_{k=1}^n \sigma_k^2.$$

In telecommunication $\mathbb{E}|y_n|^2$ is the energy of the signal.

The values σ_k^2 show how the energy of y_n is spread among frequencies.



Singular processes, V.

Proposition

The complex-valued process $y_n = \sum_{k=1}^m \xi_k e^{in\omega_k}$, given above, is singular.

We note in passing: all the arguments of the present lecture carry over to **complex-valued** processes with evident modifications.

An elegant way to establish singularity is given by the following exercise:

Exercise. Show that if $\dim(H_n^y) < \infty$ for some n , then y is singular.

(*Hint:* Prove that if y is **not** singular, then $\dim(H_n^y) = \infty$ for all n .)

Exercise. Using the last exercise prove the proposition above.



Singular processes, VI.

How can we construct a **real-valued** singular process?

Let us consider the complex-valued w.s.st. process $y = (y_n)$. Let

$$v_n := \operatorname{Re} y_n.$$

Now H_n^y is a subspace of $L^2(\Omega, \mathcal{F}, \mathcal{P})$, the set of **complex-valued** random variables ξ with $\mathbb{E}|\xi|^2 < \infty$. And $H_n^v \subset L^2(\Omega, \mathcal{F}, \mathcal{P})$.

Exercise. Prove that $\dim(H_n^y) < \infty$ for some n implies $\dim(H_n^v) < \infty$.



Singular processes, VII.

Example: Consider the w.s.st. processes as above:

$$y_n = \sum_{k=1}^m \xi_k e^{in\omega_k}. \quad \text{with} \quad \xi_k = \sigma_k e^{i\varphi_k}, \quad (7)$$

where the phases φ_k have uniform distribution on $[0, 2\pi]$, and are i.i.d.

We conclude that the real part of the process is given by

$$v_n = \sum_{k=1}^m \sigma_k \cos(in\omega_k + \varphi_k).$$

is singular. A simple special case for a real-valued singular process is:

$$y_n = \cos(\omega n + \varphi) \quad \omega \neq 0,$$

where φ is a random phase with uniform distribution on $[0, 2\pi]$.