

## Lecture 4 Gaussian Processes

Def A SP  $(X_t)_{t \in \mathbb{T}}$  is a Gaussian process if  $\forall t_1, \dots, t_n \in \mathbb{T}, \forall n$   
 $(X_{t_1}, \dots, X_{t_n})$  is a Gaussian vector  
(centered if  $\mathbb{E}X_t = 0 \forall t$ )  $\square$

Def  $c: \mathbb{T} \times \mathbb{T} \rightarrow \mathbb{R}$   $c(t, s) = c(s, t) \forall s, t \in \mathbb{T}$ .  
we say  $c$  is PSD if  $\forall t_1, \dots, t_n \in \mathbb{T}$

$$C_n = (c(t_i, t_j))_{i, j \leq n} \text{ is PSD}$$

Rmk 1 If  $(X_t)_{t \in \mathbb{T}}$  is a SP  $\mathbb{E}(X_t^2) < \infty \forall t$   
then  $c(t, s) = \text{Cov}(X_t, X_s)$  is PSD.

Rmk 2 Given  $c: \mathbb{T} \times \mathbb{T} \rightarrow \mathbb{R}$  PSD,  $m: \mathbb{T} \rightarrow \mathbb{R}$   
there exists a GP  $(X_t)$  with covariance  $c$  and mean  $m$  ( $\mathbb{E}X_t = m(t)$ ).  
The law of  $(X_t)$  (on  $\mathbb{R}^{\mathbb{T}}$ ) is unique.

Proof  $\forall t_1, \dots, t_n$   $\underline{m}_n = (m(t_1), \dots, m(t_n))$   
 $C_n = (c(t_i, t_j))_{i, j \leq n}$

$$\mu_{t_1, \dots, t_n} := N(\underline{m}_n, C_n).$$

Easy to check  $(\mu)$  are consistent  $\square$

Rmk 3 If  $\mathcal{H}$  is a Hilbert space.

$\forall t_1, t_2 \in \mathcal{H}$  define  $c(t_1, t_2) = \langle t_1, t_2 \rangle_{\mathcal{H}}$

$$\sum_{i,j=1}^n c(t_i, t_j) v_i v_j = \left\| \sum_{i=1}^n v_i t_i \right\|_{\mathcal{H}}^2 \geq 0$$

so  $c$  is PSD.

$\Rightarrow \exists$  GP indexed by  $\Pi = \mathcal{H}$   
with covariance  $c$ .

$\mathcal{H} = (\mathbb{R}^d, \langle \cdot, \cdot \rangle)$  consider  $m=0$

$$t \in \mathbb{R}^d \quad \mathbb{E}(X_{t_1} X_{t_2}) = \langle t_1, t_2 \rangle$$

$$t = e_i \quad \mathbb{E}(X_{e_i} X_{e_j}) = 1_{i=j}$$

$$X_t = \langle t, G \rangle \quad G \sim N(0, I_d)$$

[  $\mathcal{H} = L^2([0,1], \text{Leb})$   
White noise process -

Lemma Assume  $m=0$ .  $\Pi = [a, b]^d$ . If

$$(*) \quad c(t,t) + c(s,s) - 2c(t,s) \leq C \|t-s\|^{2\bar{\gamma}}$$

Then  $\exists (X_t)$  GP with cov  $c$  which  
is  $\bar{\gamma}$ -Hölder  $\forall \gamma < \bar{\gamma}$ .

Rmk (\*) means

$$\mathbb{E}[|X_t - X_s|^2] \leq C \|t-s\|^{2\bar{\gamma}}$$

□

Proof Apply KC

$$\stackrel{\text{wts}}{\hookrightarrow} \mathbb{E}[|X_t - X_s|^\alpha] \leq C \|t-s\|^{\beta+d}$$

$$\alpha = 2k$$

$$\gamma < \frac{\beta}{\alpha} = \frac{\beta+d}{\alpha} - \frac{d}{\alpha}$$

$$\mathbb{E}[|X_t - X_s|^{2k}] \leq C_k \mathbb{E}[|X_t - X_s|^2]^k$$

$$\leq C_k \cdot [c(t,t) + c(s,s) - 2c(t,s)]^k$$

$$\leq C_k' \|t-s\|^{2\bar{\gamma} \cdot k}$$

$$\gamma < \frac{2\bar{\gamma}k}{2k} - \frac{d}{2k} = \bar{\gamma} - \frac{d}{2k} \quad \text{take } k \text{ large enough}$$

□

Lemma  $(X^{(k)})_{t \in \Pi}$ ,  $(X_t)_{t \in \Pi}$  with  $(X_t^{(k)})$  GP

If  $\forall t \in \Pi$

$$\lim_{n \rightarrow \infty} \mathbb{E}[(X_t^{(k)} - X_t)^2] = 0$$

then  $(X_t)$  is a GP

$$m(t) = \lim_{k \rightarrow \infty} m_k(t)$$

$$c(t,s) = \lim_{k \rightarrow \infty} c_k(t,s) \quad \square$$

Proof Fix  $n$ ,  $t_1, \dots, t_n$ ,  $\mu_{t_1, \dots, t_n}$  fdd of  $(X_t)$

$$\mu_{t_1, \dots, t_n}^{(k)} = N(\underline{m}_n^{(k)}, \underline{C}_n^{(k)})$$

$\hookrightarrow (c^{(k)}(t_i, t_j))_{i,j \leq n}$

$$\underline{X}_n^{(k)} = \pi_{t_1, \dots, t_n}(X^{(k)}) \sim \mu_{t_1, \dots, t_n}^{(k)}$$

$$\underline{X}_n = \pi_{t_1, \dots, t_n}(X)$$

By assumption

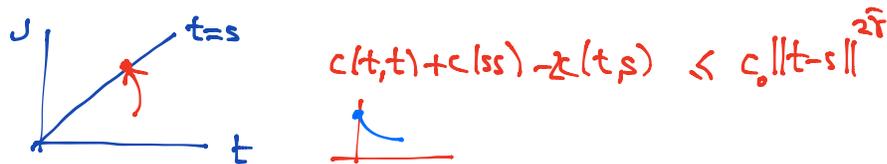
$$\mathbb{E}[\|X_n^{(e)} - \underline{X}_n\|_2^2] = \sum_{i=1}^n \mathbb{E}[(X_{t_i}^{(e)} - X_{t_i})^2] \xrightarrow{t \rightarrow \infty} 0.$$

$\Rightarrow$  sequence of G vectors converging in  $L^2$ . The limit is Gaussian with covariance = limit of covariances.  $\square$

$$\gamma_k \uparrow \bar{\gamma} \quad \mathcal{U}_k \quad \mathcal{U} = \bigcup_k \mathcal{U}_k$$

For each  $k$  construct  $X_k$   $\gamma_k$  Hölder and  $X_k(\omega) = \bar{X}(\omega) \forall \omega \in \mathcal{U}_k$

Set  $X = \bar{X}$  for  $\omega \in \mathcal{U}$   
 $X = 0$  otherwise.  $\square$



Rmk A GP  $(X_t)_{t \geq 0}$  has indep increments iff  $\forall t_1 \leq t_2 < t_3$

$$\text{Cov}(X_{t_1}, X_{t_3} - X_{t_2}) = 0 \quad \square$$

[General def was

$$X_{t_3} - X_{t_2} \text{ indep of } \mathcal{F}_{t_2}^X = \sigma(\{X_t : t \leq t_2\}) \\ = \sigma(\{X_{s_i} \in A_i, X_{t_n} \in A_n : n, s_i, A_i\})$$

sufficient to check  $\forall n, s_1, \dots, s_n \leq t_2$

$$X_{t_3} - X_{t_2} \text{ indep of } (X_{s_1}, \dots, X_{s_n})$$

For GP uncorrelated  $\Rightarrow$  indep.  
suff to check

$$X_{t_3} - X_{t_1} \text{ uncorrel with } X_{s_i} \quad i \leq n \quad \square ]$$

$$\text{Cov}(X_{t_1}, X_{t_3} - X_{t_2}) = 0 \quad \forall t_1 \leq t_2 < t_3$$

$$(*) \quad c(t_1, t_3) = c(t_1, t_2) \quad \text{set } \begin{matrix} t_1 = t_2 = t \\ t_3 = t' \end{matrix}$$

$$c(t, t') = \underbrace{c(t, t)}_{g(t)} \quad \forall t' > t$$

$$\boxed{c(t, t') = g(t \wedge t')} \quad \forall t, t' \in \mathbb{R}_{\geq 0}$$

viceversa, this implies indep increm.

Definition Wiener process / Brownian motion is

GP on  $\Pi = \mathbb{R}_{\geq 0}$  with covariance

$$c(t, t') = t \wedge t', \quad m(t) = x \in \mathbb{R} \quad \forall t$$

with cont. sample paths  $\square$

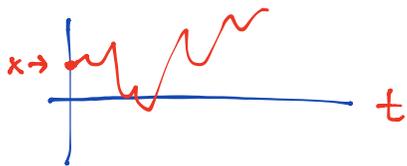
Rmk - Has indep increments

-  $c(t, t')$  1-Lipschitz as  $t \rightarrow t'$

$2\bar{\gamma}$ -Hölder  $\bar{\gamma} = 1/2$

$\Rightarrow$  BM is  $\gamma$ -Hölder  $\forall \gamma < 1/2$ .

⇒ Can be shown not  $\gamma$ -Hölder  $\forall \gamma \geq 1/2$ .



Def  $\theta_s :=$  Shift by  $s \in \mathbb{R}_{\geq 0}$

$$\theta_s: \mathbb{R}^{[0, \infty)} \rightarrow \mathbb{R}^{[0, \infty)}$$

$$x \mapsto \theta_s(x) \quad \theta_s(x)(t) = x(t+s) \quad \square$$

It is measurable on  $\mathcal{B}^{[0, \infty)}$

$(X_t)$  with law  $\mathbb{P}_x$ ,  $\mathbb{P}_x \circ \theta_s^{-1}$  law of shifted process

$$\mathbb{P}_x \circ \theta_s^{-1}(B) = \mathbb{P}(\theta_s(X) \in B)$$

Def  $(X_t)_{t \in [0, \infty)}$  is stationary if  $\mathbb{P}_x = \mathbb{P}_x \circ \theta_s^{-1}$

Rmk Stationary iff.  $\forall n, t_1, \dots, t_n$

$$(X_{t_1+s}, \dots, X_{t_n+s}) \stackrel{d}{=} (X_{t_1}, \dots, X_{t_n})$$

$$\mathbb{P}_{X_{t_1+s}, \dots, X_{t_n+s}} = \mathbb{P}_{X_{t_1}, \dots, X_{t_n}} \quad \square$$

In particular stationarity implies

$$c(t, s) = \text{Cov}(X_t, X_s) \quad m(t) = \mathbb{E} X_t$$

$$\left[ \begin{array}{l} c(t_1+s, t_2+s) = c(t_1, t_2) = r(|t_1 - t_2|) \quad (*) \\ m(t+s) = m(t) = m_0 \quad (**) \end{array} \right.$$

Def:  $(X_t)_{t \in [0, \infty)}$   $\mathbb{E} X_t^2 < \infty$  second order station.  
 if  $(*)$ ,  $(**,*)$  - ( $L^2$ -station, weakly st<sub>2</sub>)

In general  $L^2$ -stationarity  $\not\Rightarrow$  stationarity.  
 For GP two notions are equiv.

Ex: BM not stationary!

$$t > s \quad \mathbb{E}[(X_t - X_s)^2] = t - 2(t \wedge s) + s = t - s$$

$$\mathbb{E}(X_t^2) = \mathbb{E}(X_0^2) + \mathbb{E}[(X_t - X_0)^2] = 0 + t = t$$

Def  $(X_t)_{t \in [0, \infty)}$  stationary increm.

if  $\forall t, t', h \geq 0$

$$X_{t+h} - X_t \stackrel{d}{=} X_{t'+h} - X_{t'}$$

Ex For BM  $X_{t+h} - X_t \sim N(0, h)$

□

Indep incr  $c(t, t') = g(t \wedge t')$  ✓

$(W_t)$   $\geq$  BM started at  $x=0$

$h$  a non decr funct:  $\mathbb{R}_{\geq 0} \rightarrow \mathbb{R}_{\geq 0}$

$X_t := W_{h(t)}$  centered  
 Gaussian process

$$\begin{aligned} \mathbb{E}(X_{t_1} X_{t_2}) &= \mathbb{E} W_{h(t_1)} W_{h(t_2)} = h(t_1) \wedge h(t_2) \\ &= h(t_1 \wedge t_2). \end{aligned}$$

$$\begin{aligned} \underline{\mathbb{E}(X_t^2)} &= \mathbb{E}(X_t)^2 + \text{Var}(X_t) = m(t)^2 + c(t,t) \\ &= \underline{x^2 + t} \end{aligned}$$

$$\mathbb{E}[|X|^{2k}] \leq C_k \mathbb{E}(X^2)^k \quad (*)$$

$$X \sim N(0, \tau^2) \quad X = \tau \cdot G \quad G \sim N(0, 1)$$

$$\mathbb{E}(|X|^{2k}) = \mathbb{E}(\tau^{2k} |G|^{2k}) = \underbrace{\mathbb{E}(|G|^{2k})}_{C_k} \cdot \mathbb{E}(X^2)^k$$

$\tau^2$  subgaussian.  $\mathbb{E} e^{\lambda X} \leq e^{\lambda^2 \tau^2 / 2} \quad \forall \lambda \in \mathbb{R}$

$$\mathbb{P}(|X| \geq t) \leq 2e^{-t^2 / c\tau^2} \quad (e^{-t/\tau})$$

If  $X$  is  $\tau^2$  subg.  $\tau^2 \leq c \cdot \mathbb{E}(X^2) \quad (*)$

then  $(*)$  holds

$$\mathbb{E}|X|^{2k} = \int_0^\infty t^{2k-1} \mathbb{P}(|X| \geq t) dt \leq C_k \tau^{2k}$$

$$c(t,t) + c(s,s) - 2c(t,s) \leq c_0 \|t-s\|^{2\bar{\gamma}}$$

For ant.  $\mathbb{E}(|X_t - X_s|^2) \leq c_0 \|t-s\|^{2\bar{\gamma}}$

$\underbrace{\hspace{10em}}_{t-s}$