

16/05/24

## Lecture - 11

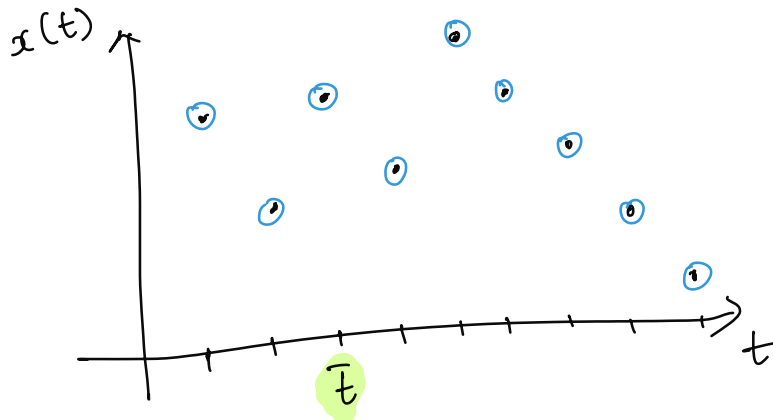
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### Definition & Properties of Gaussian Process (Continued)

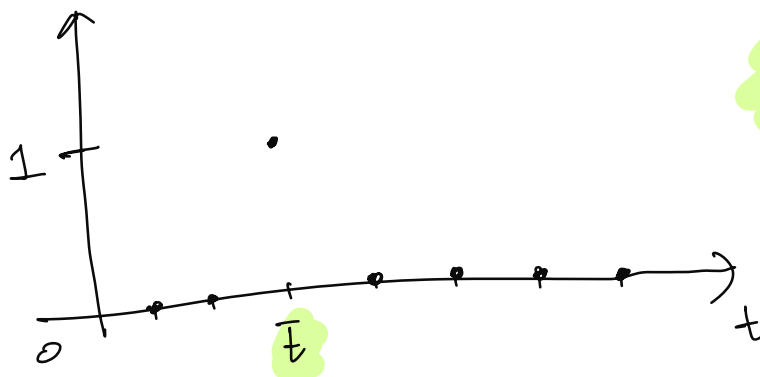
- We are studying, how to extend the properties of Gaussian RVs to the Gaussian processes
- we were able to understand that our definition of Gaussian process goes through the use of projections and if a process obeys the assumption of being a Gaussian process then its projections when I align them in a single vector will form a Random Vectors with its elements being Jointly Gaussian.
- ① What about the samples of a Gaussian process?  
Are the samples we get are Gaussian?  
(scalars)
- We would like the samples to have a Gaussian distribution and this requires a little attention, because this is true for

# Discrete time sampling of DT Gaussian processes

Discrete time case



process



Delta function  
 $\delta_{t, \bar{t}}$

$$p = \sum_{t=-\infty}^{+\infty} \delta_{t, \bar{t}} x[t] = x[\bar{t}]$$

When we project the instance of the process onto the Delta centered in  $\bar{t}$

$$p = \sum_{t=-\infty}^{\infty} \delta_{t, \bar{t}} a(t) = a[\bar{t}]$$

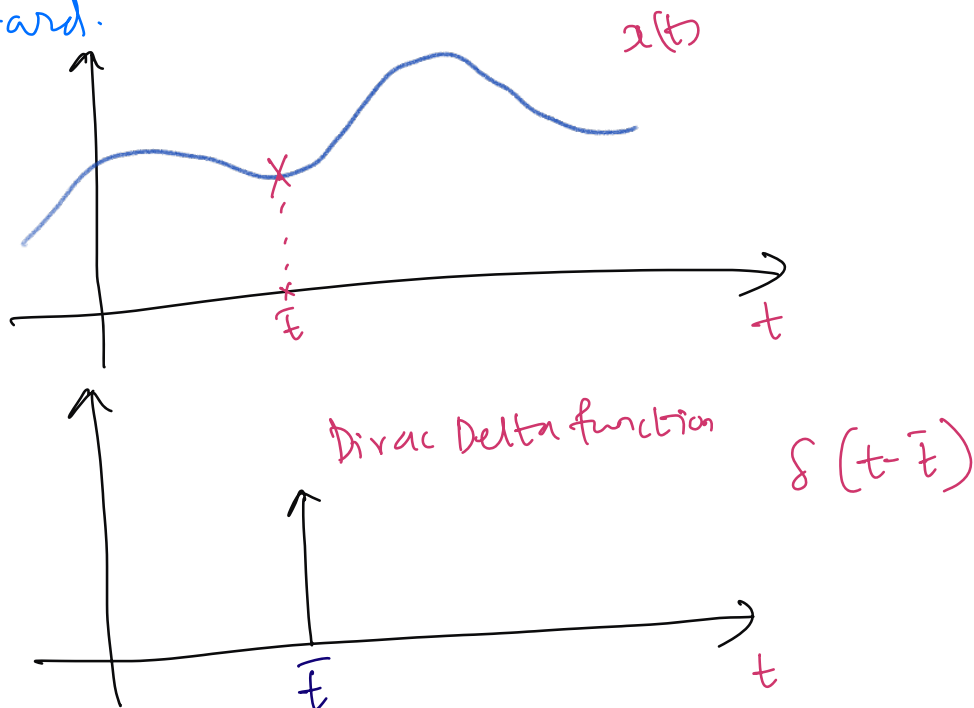
projection
sampling

$$\sum_{t=-\infty}^{+\infty} \delta_{t, \bar{t}}^2 = 1 < +\infty \quad \text{OK!}$$

This is a finite energy function

$\therefore$  There is no problem with Discrete Time

But, for the Continuous time this not so straight forward.



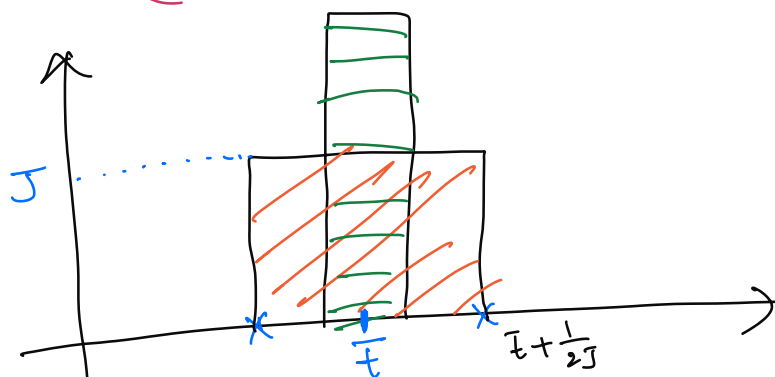
Now, if we project our <sup>CT</sup> Gaussian RP onto the Dirac Delta function. Do the samples have Finite Energy?

$$p = \underbrace{\int \delta(t - \bar{t}) x(t) dt}_{\text{projection}} = \underbrace{x(\bar{t})}_{\text{sampling}}$$

$$\int \delta^2(t - \bar{t}) dt = +\infty$$

Not OK!

\*\*\*So, we will add an additional assumption that  $k_x(t, s)$  is continuous function!  
(covariance function)



Height =  $J$   
width =  $\frac{1}{J}$

Approximation of Dirac-Delta

$$\phi_{J, \bar{t}}(t) = \begin{cases} 0 & \text{if } |t - \bar{t}| > \frac{1}{2J} \\ J & \text{if } |t - \bar{t}| \leq \frac{1}{2J} \end{cases}$$

$\phi_{J, \bar{t}}$   
finite  
energy  
function

$$\forall J \quad \int \phi_{J, \bar{t}}^2(t) dt < +\infty$$

$$P_J = \int \phi_{J, \bar{t}}(t) x(t) dt \Rightarrow \text{Gaussian}$$

if  $x(t)$  is Gaussian

→ Now we also consider the sample of our RP  $x(t)$  at  $t = \bar{t}$  i.e.  $x(\bar{t})$  and we wait to evaluate

$$E \left[ (P_J - x(\bar{t}))^2 \right] \rightarrow \text{Average Squared Error}$$

and if this value goes to zero

when  $J \rightarrow \infty$  then we know that there are sequence of <sup>Gaussian</sup> random variables that tend to be our samples.

$$E \left[ \left( \int \phi_{J, \bar{t}}(t) x(t) dt - \underbrace{x(\bar{t})}_{\downarrow} \right)^2 \right]$$

$$\int \phi_{J, \bar{t}}(t) x(\bar{t}) dt$$

$$\Rightarrow x(\bar{t}) \underbrace{\int \phi_{J, \bar{t}}(t) dt}_{\text{This is equal to 1}}$$

$$\Rightarrow E \left[ \left( \int \phi_{J, \bar{t}}(t) (x(t) - x(\bar{t})) dt \right)^2 \right]$$

The square of an integral is a double integral and in this case we use two variables  $t, s$

$$\Rightarrow E \left[ \iint \phi_{J, \bar{t}}(t) [x(t) - x(\bar{t})] \phi_{J, \bar{t}}(s) [x(s) - x(\bar{t})] dt ds \right]$$

$$\Rightarrow \iint \phi_{J, \bar{t}}(t) \phi_{J, \bar{t}}(s) E \left[ (x(t) - x(\bar{t})) (x(s) - x(\bar{t})) \right] dt ds$$

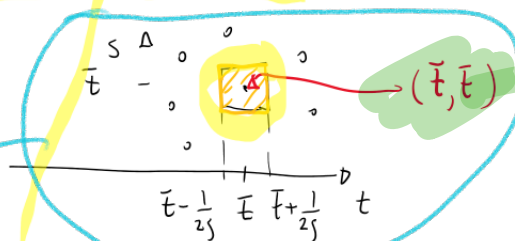
$$\Rightarrow \int_{-\infty}^{+\infty} \int_{-\infty}^{+\infty} \phi_{J, \bar{t}}(t) \phi_{J, \bar{t}}(s) \left\{ E[x(t)x(s)] - E[x(t)x(\bar{t})] - E[x(\bar{t})x(s)] + E[x(\bar{t})x(\bar{t})] \right\} dt ds$$

⇒

Here we are evaluating the Correlation function at two different time instances

Here we have correlation function at the same time instat

$$= \int_{-\infty}^{+\infty} \int_{-\infty}^{+\infty} \rho_{s,\bar{T}}(t) \rho_{s,\bar{T}}(s) \left\{ E[x(t)x(s)] - E[x(t)x(\bar{T})] - E[x(\bar{T})x(s)] + E[x(\bar{T})x(\bar{T})] \right\} dt ds$$



The square shrinks as  $J \rightarrow \infty$  and  $t^*, s^*$  are inside the square centered at  $\bar{T}$

$$= \int_{\bar{T} - \frac{1}{2J}}^{\bar{T} + \frac{1}{2J}} \int_{\bar{T} - \frac{1}{2J}}^{\bar{T} + \frac{1}{2J}} J^2 [C_x(t,s) - C_x(t,\bar{T}) - C_x(\bar{T},s) + C_x(\bar{T},\bar{T})] dt ds$$

$K_x$  is continuous

⇓

$C_x$  is continuous

⇓

the integrand is continuous

intermediate value theorem



$$= \frac{1}{J^2} J^2 [C_x(t^*, s^*) - C_x(t^*, \bar{T}) - C_x(\bar{T}, s^*) + C_x(\bar{T}, \bar{T})]$$

$$= \frac{1}{\delta^2} \delta^2 [C_x(t^*, s^*) - C_x(t^*, \bar{t}) - C_x(\bar{t}, s^*) + C_x(\bar{t}, \bar{t})]$$

$$\delta \rightarrow \infty \quad \frac{1}{\delta} \rightarrow 0 \quad t^*, s^* \rightarrow \bar{t}$$

$C_x$  is continuous  $\rightarrow$

$$\delta \rightarrow \infty \quad C_x(t^*, s^*) \rightarrow C_x(\bar{t}, \bar{t})$$

$$C_x(t^*, \bar{t}) \rightarrow C_x(\bar{t}, \bar{t})$$

$$C_x(\bar{t}, s^*) \rightarrow C_x(\bar{t}, \bar{t})$$

$$= 0 = E[(p_\delta - x(\bar{t}))^2]$$

$\nwarrow$   
is Gaussian

$\rightarrow$  Since  $x(\bar{t})$  are anyway Gaussian,  $\rightarrow$  Now the error w/ the projection & the  $x(\bar{t})$  is zero our projections of CT RP  $x(t)$  are also Gaussian.

\* This comes in handy in the analysis of white Gaussian process.



## → White Gaussian Noise

WGN is a Gaussian stochastic process

$$\therefore p = \int \phi(t) x(t) \Rightarrow \text{is Gaussian}$$

WGN  $\Rightarrow$

$$m_p = 0$$

&

$$\sigma_p^2 = \frac{N_0}{2} \underbrace{\|\phi\|^2}_{\text{NORM}}$$

Average of  
The projection is  
Zero

$$\Rightarrow \frac{N_0}{2}$$

$$\int |\phi(t)|^2 dt$$

energy of the  $\phi(t)$

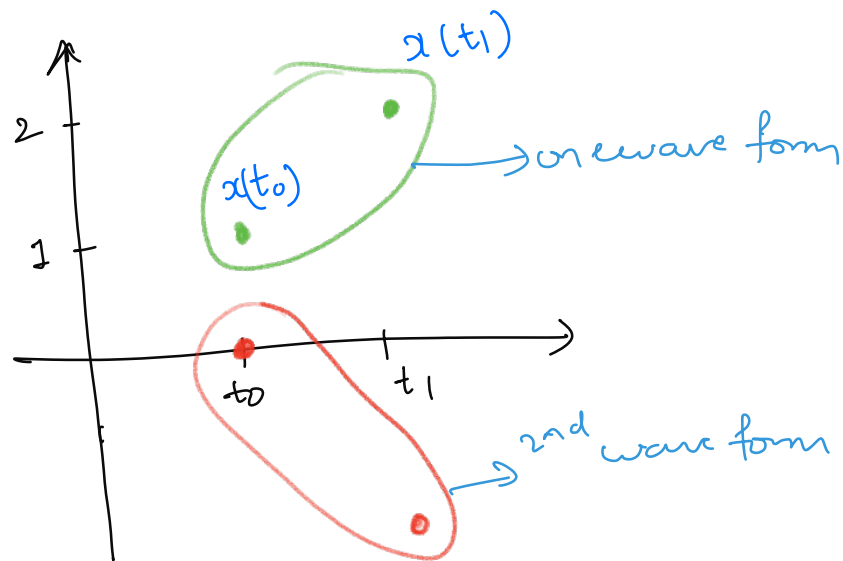
So it only depends on the  
energy of the  $\phi(t)$

$\therefore$  if we take two  $\phi(t)$  which are sine  
& cosine, so the  $\sigma_p^2$  of the two projections  
made using sine & cosine will be same  
because sine & cosine have equal <sup>Average</sup> energy.

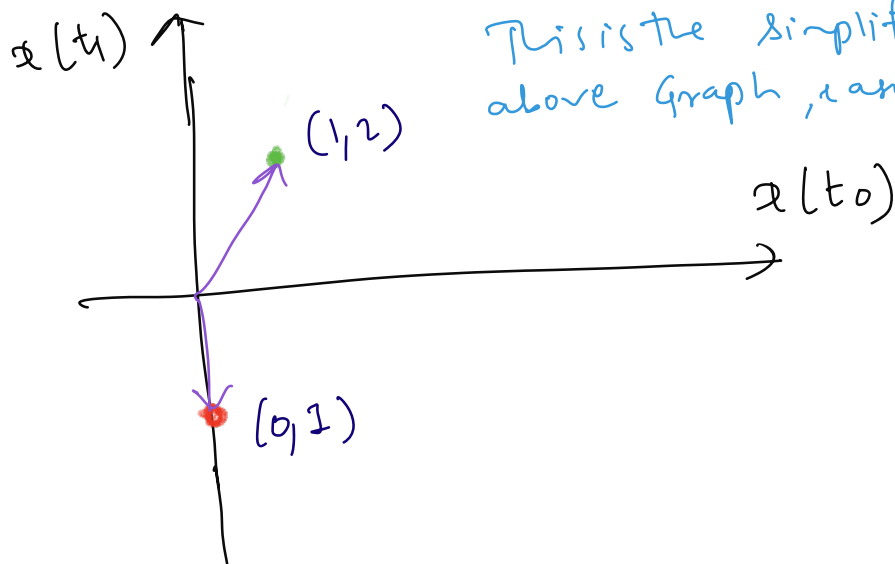
- This dependency of  $\sigma_p^2$  only on the Energy makes the WGN a special object full of virtues & problems

- From geometrical point of view WGN is seen as a sphere i.e. from any direction it looks the same from statistical point of view.

ex: WGN in Discrete Time processes



- They are discrete waveforms because they depend on different instances of time i.e.  $t_0, t_1$



This is the simplification of the above graph, easy to analyze

Q) What are  $\phi^s$ ?  
in our simplified environment

$$\phi_0 = \begin{bmatrix} \phi_{00} \\ \phi_{01} \end{bmatrix}$$

$$\|\phi_0\|^2 = 1$$

energy = 1

$$\phi_{00}^2 + \phi_{01}^2 = 1$$

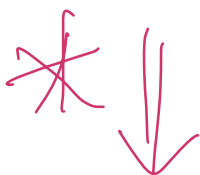
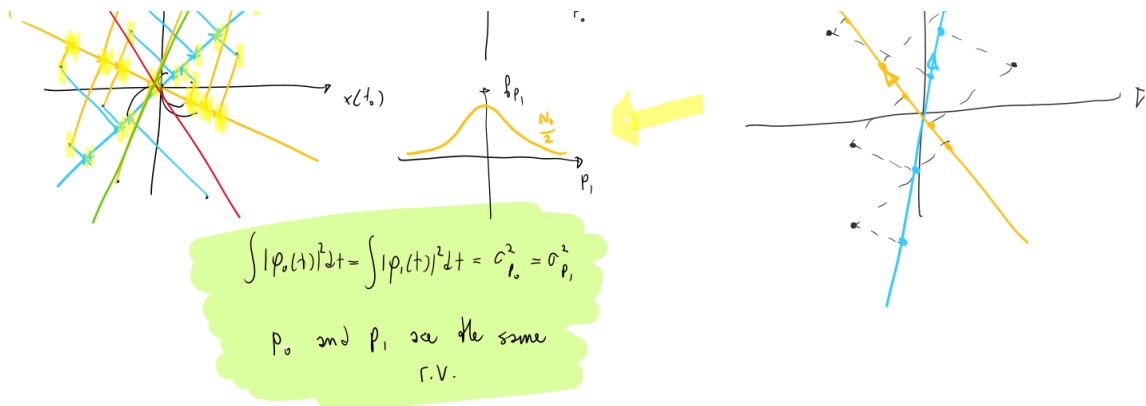
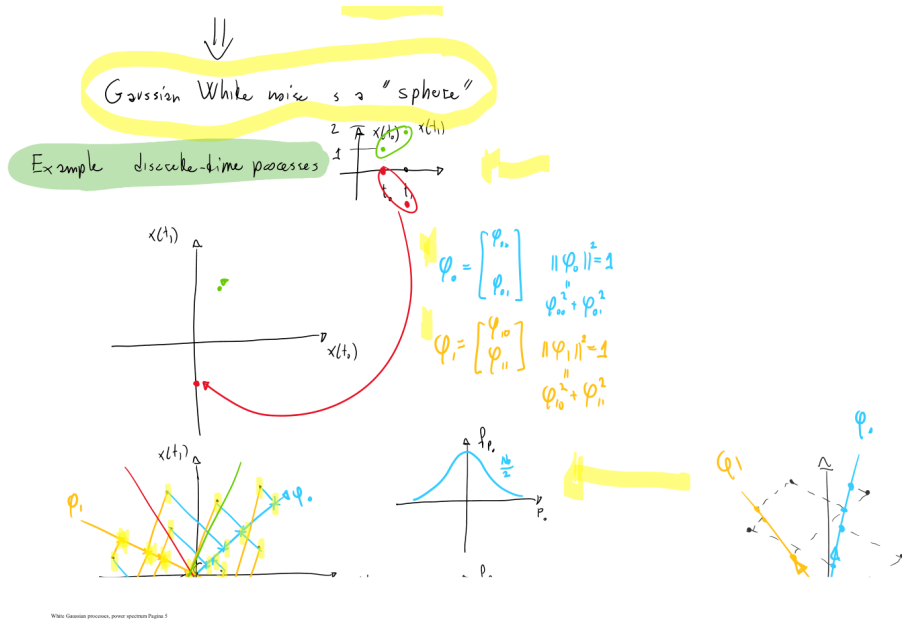
$$\phi_1 = \begin{bmatrix} \phi_{10} \\ \phi_{11} \end{bmatrix}$$

$$\|\phi_1\|^2 = 1$$

i.e.  $\phi_0$  is a unit vector with tip of the arrow  
is on the unit circle

$\phi_1$  is also the same

∴ if project on the  $\phi_0$  (or)  $\phi_1$  we get the same Gaussian vector because both have the same Energy



Consequence of "Whiteness"

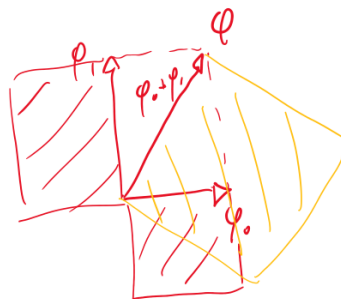
$$\varphi_0 \perp \varphi_1 \iff \int \varphi_0(t) \varphi_1(t) dt = 0$$

if the  $\varphi^i$  are 1an

$$p_0 = \int \varphi_0(t) x(t) dt \quad p_1 = \int \varphi_1(t) x(t) dt$$

$$p = \int \underbrace{[\varphi_0(t) + \varphi_1(t)]}_{\varphi(t)} x(t) dt = p_0 + p_1$$

$$\begin{aligned} \sigma_p^2 &= E[p^2] = E[(p_0 + p_1)^2] = E[p_0^2] + E[p_1^2] + 2E[p_0 p_1] \\ &= \frac{N_0}{2} \|\varphi_0 + \varphi_1\|^2 = \frac{N_0}{2} (\|\varphi_0\|^2 + \|\varphi_1\|^2) = \frac{N_0}{2} \|\varphi_0\|^2 + \frac{N_0}{2} \|\varphi_1\|^2 \end{aligned}$$



Pythagorean Theorem

$$\begin{aligned} \int |\varphi_0(t) + \varphi_1(t)|^2 dt &= \int |\varphi_0(t)|^2 dt + \int |\varphi_1(t)|^2 dt + 2 \underbrace{\int \varphi_0(t) \varphi_1(t) dt}_0 \\ &= \|\varphi_0\|^2 + \|\varphi_1\|^2 \end{aligned}$$

$$\left. \begin{aligned} \varphi_0 \perp \varphi_1 &\rightarrow E[p_0 p_1] = 0 \\ E[p_0] &= E[p_1] = 0 \end{aligned} \right\} \rightarrow \sigma_{p_0 p_1} = 0$$

$p_0, p_1$  are jointly Gaussian  
 $p_0, p_1$  have zero covariance  $\Rightarrow p_0, p_1$  are independent