

16/05/24

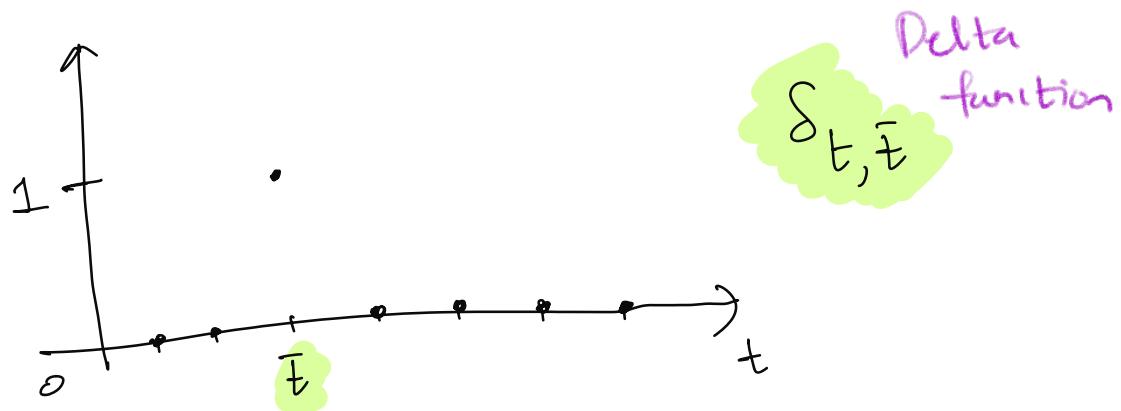
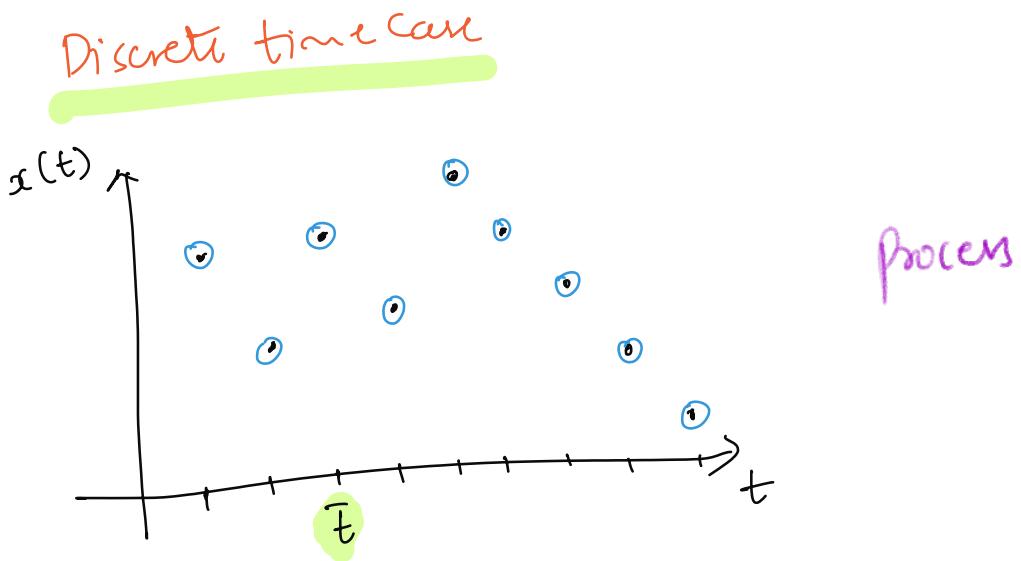
Lecture-11

12/05/21

Definition & Properties of Gaussian process (Continued)

- We are studying, How to extend the properties of Gaussian RV's 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 Random vectors with its elements being Jointly Gaussian.
- Q) 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 process



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

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

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

$\underbrace{\phantom{\sum_{t=-\infty}^{\infty}}}_{\text{projection}}$ $\underbrace{}_{\text{sampling}}$

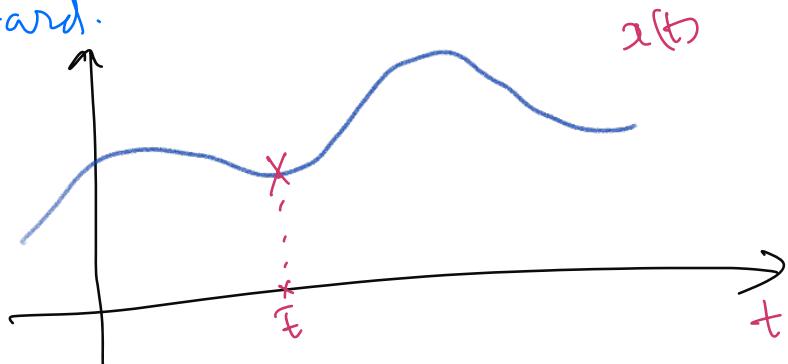
$$\sum_{t=-\infty}^{+\infty} \delta_{t,\bar{t}}^2 = 1 < +\infty$$

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 Gaussian RP onto the Dirac Delta function. Do the samples have Finite Energy?

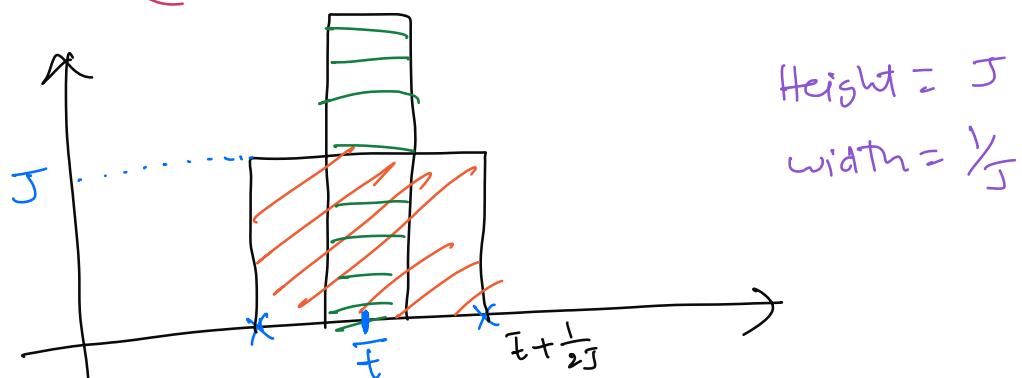
$$p = \int \delta(t-\bar{t}) x(t) dt = \tilde{x}(\bar{t})$$

projection sampling

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

Not OK!

So, we will add an additional assumption that $\kappa_x(t, s)$ is continuous function!
 (Covariance function)



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

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

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

is
Gaussian
if $x(t)$ is Gaussian

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

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

Average
Squared Error

and if this value goes to zero
when $J \rightarrow \infty$ then we know that
there are sequence of ^{Gaussian} Random variables that
tend to be our samples.

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

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

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

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

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

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

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

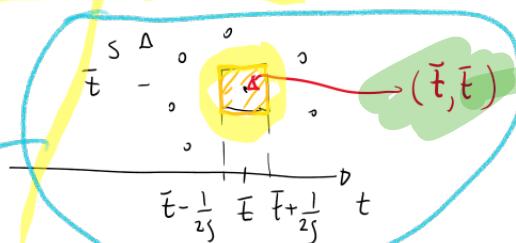
$$\Rightarrow \iint_{-\infty}^{+\infty} \phi_{J, \bar{t}}(t) \phi_{J, \bar{t}}(s) \left\{ E[x(t)x(s)] - E[x(t)\bar{x}(\bar{t})] - E[\bar{x}(\bar{t})x(s)] + E[\bar{x}(\bar{t})\bar{x}(\bar{t})] \right\} dt ds$$



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

Here we have correlation function at the same time instant

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



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

The square shrinks as $J \rightarrow 0$
as t^*, s^* are inside the square centered at \bar{t}

k_x is continuous
↓
 c_x is continuous
↓

the integrand is continuous

Intermediate value theorem



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

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

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

C_x is continuous \rightarrow

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

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

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

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

is Gaussian

Since $x(\bar{t})$ are anyway Gaussian, and now the error b/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

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

$$\text{WGN} \Rightarrow m_p = 0 \quad \& \quad \sigma_p^2 = \frac{N_0}{2} \frac{\|\phi\|^2}{\text{NORM}}$$

Average of
The projection is
Zero

$$\Rightarrow \frac{N_0}{2} \left[\int |\phi(t)|^2 dt \right] \quad \text{energy of the } \phi(t)$$

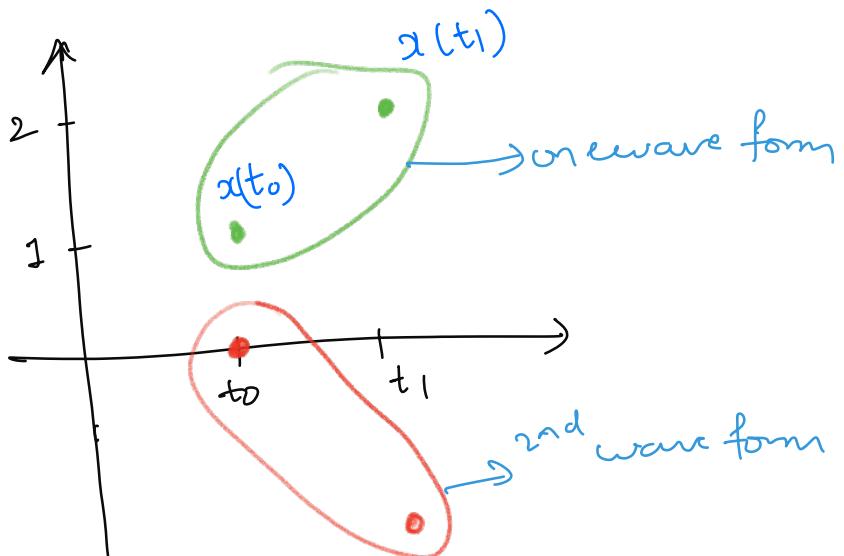
So it only depends on the
energy of the $\phi(t)$

∴ if we take two $\phi(t)$ which are sine & cosine, so the σ_p^2 of the two projections made using sine & cosine will be same because sine & cosine have equal ^{Average} energy.

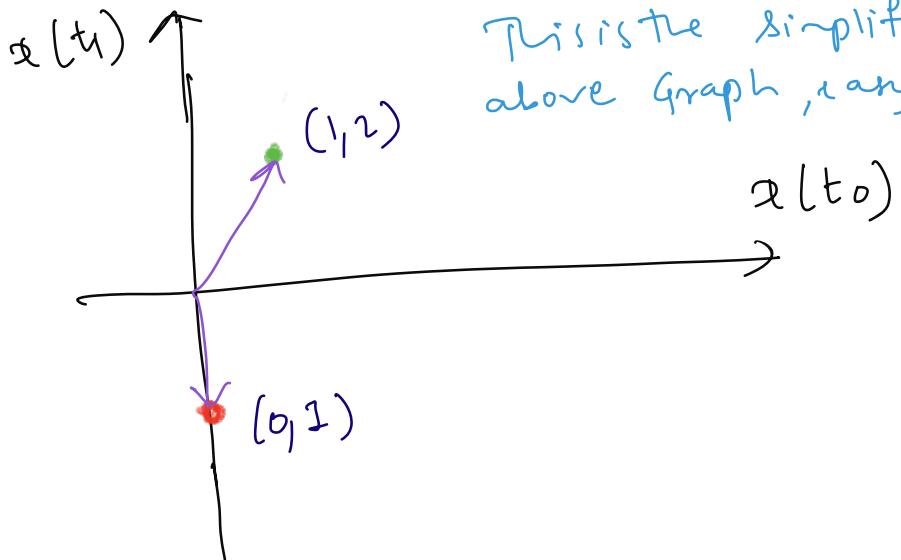
- This dependency of σ_p^2 only on the Energy makes the WGN a special object full of **virtues** & **problems**

- 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 ϕ^s in our simplified environment

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

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

energy = 1

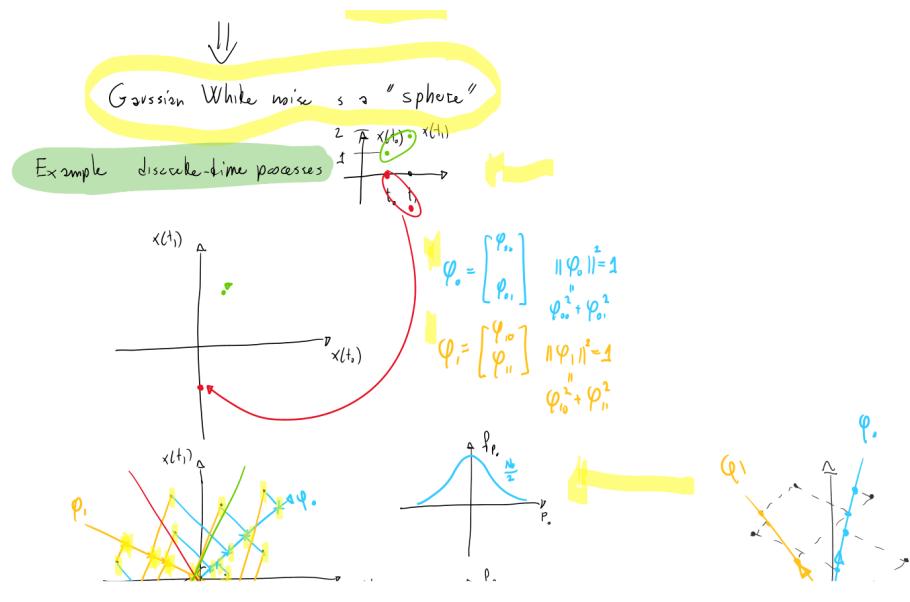
$$\phi_{00}^2 + \phi_{01}^2 = 1$$

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

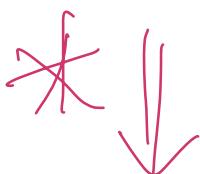
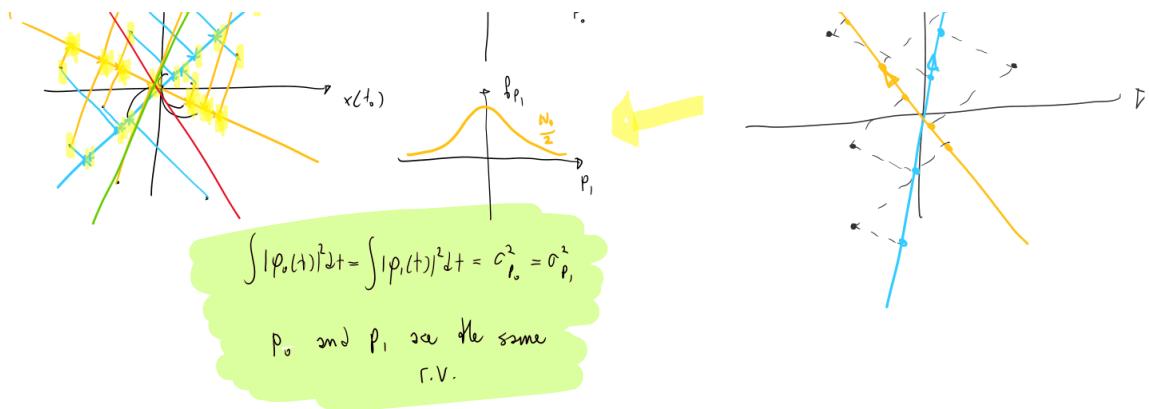
$$\|\phi_1\|^2 = 1$$

i.e ϕ_0 is a unit vector with tip of the arrow
 is on the unit circle
 ϕ_1 is also the same

\therefore if project on the ϕ_0 or ϕ_1 we get the same Gaussian vector because both have the same Energy



White Gaussian processes, power spectrum Page 5



Consequence of "Whiteness"

$$\varphi_0 \perp \varphi_1 \Leftrightarrow \int \rho_0(t) \varphi_1(t) dt = 0$$

$$\rho_0 = \int \rho_0(t) x(t) dt \quad \rho_1 = \int \varphi_1(t) x(t) dt$$

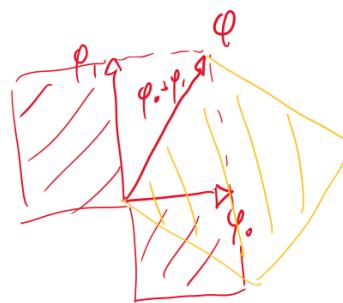
$$\rho = \int [\varphi_0(t) + \varphi_1(t)] x(t) dt = \rho_0 + \rho_1$$

$$\begin{aligned} \sigma_p^2 &= E[\rho^2] = E[(\rho_0 + \rho_1)^2] = E[\rho_0^2] + E[\rho_1^2] + 2E[\rho_0 \rho_1] \\ &= \frac{N_0}{2} \|\varphi_0 + \varphi_1\|^2 = \frac{N_0}{2} (\|\varphi_0\|^2 + \|\varphi_1\|^2) = \frac{N_0}{2} \|\rho_0\|^2 + \frac{N_0}{2} \|\rho_1\|^2 \end{aligned}$$

Ptolemy's theorem
Theorem

$$\int |\rho_0(t) + \varphi_1(t)|^2 dt = \int |\rho_0(t)|^2 dt + \int |\varphi_1(t)|^2 dt + 2 \int \rho_0(t) \varphi_1(t) dt$$

if the φ 's are 1 for



$$\varphi_0 \perp \varphi_1 \rightarrow \left. \begin{array}{l} E(\rho_0 \rho_1) = 0 \\ E(\rho_0) = E(\rho_1) = 0 \end{array} \right\} \rightarrow \delta_{\rho_0 \rho_1} = 0$$

ρ_0, ρ_1 are pointly Gaussian
 ρ_0, ρ_1 have zero covariance

$\Rightarrow \rho_0, \rho_1$ are independent