

Today

- (1) White Noise
- (2) White Gaussian Noise

3) Introduction to Markov Chains.

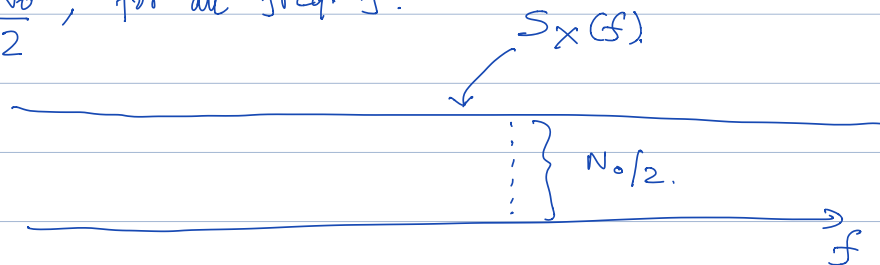
Recap. → (SS) Stationary, WSS,  $\xrightarrow[\text{I/P}]{\text{WSS}} \boxed{\text{LTI}} \rightarrow \text{O/P}$   
Jointly WSS

White Noise → <sup>to model</sup> thermal noise in electronic systems.

$X(t)$  is White Noise if.

PSD of  $X(t)$ ,  $S_X(f)$  is constant for all frequencies.

$$S_X(f) = \frac{N_0}{2}, \text{ for all freq. } f.$$



Power of White Noise <sup>Random Process</sup> ?

$$= \int_{-\infty}^{\infty} \frac{N_0}{2} df = \boxed{\infty}$$

"in reality"



$$\begin{aligned} P_X &= R_X(0) \\ &= E[X^2(t)] \\ &= \int_{-\infty}^{\infty} \underbrace{S_X(f)}_{\text{PSD}} df. \end{aligned}$$

Power of

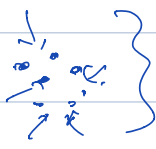
→  
true white noise.

# White Gaussian Noise (WGN).

$X(t) \rightarrow$  (i) Stationary Gaussian random process.

$\rightarrow$  (ii) Zero mean ( $\mu_x = 0$ )

(iii) Flat (constant) PSD,  
 $S_x(f) = \frac{N_0}{2}$  for all freq.  $f$ .

  $\rightarrow \frac{\sum_i \text{disturbances.}}{\# \text{ of items.}} \Rightarrow \frac{X_1 + X_2 + \dots + X_n}{n}$   
 by CLT  $\rightarrow N(\mu, \sigma^2)$

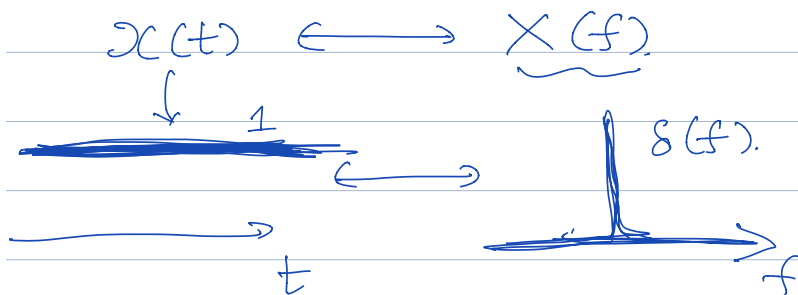
ACF of WGN : (F.T. pair)

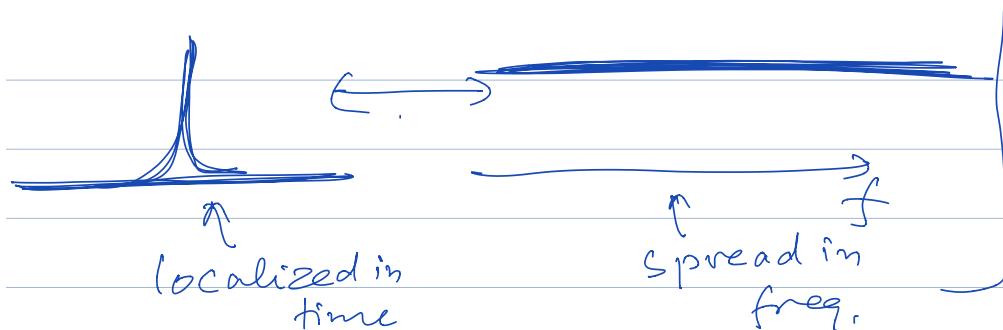
ACF  $\longleftrightarrow$  PSD  
 $R_x(\tau) \longleftrightarrow S_x(f)$

$$R_x(\tau) = \mathcal{F}^{-1}(\text{PSD})$$

$$= \mathcal{F}^{-1}\left(\frac{N_0}{2}\right) = \frac{N_0}{2} \cdot \mathcal{F}^{-1}(1)$$

$\mathcal{F}^{-1}(1)$  is the Dirac delta function.





WGN Are  $X(t_1)$ ,  $X(t_2)$  independent of each other?

is  $\rho = 0$ ? or  $\text{cov}(X(t_1), X(t_2)) \stackrel{?}{=} 0$

$$E[(X(t_1) - 0)(X(t_2) - 0)]$$

$$R_X(\tau) = \frac{N_0}{2} \delta(\tau)$$

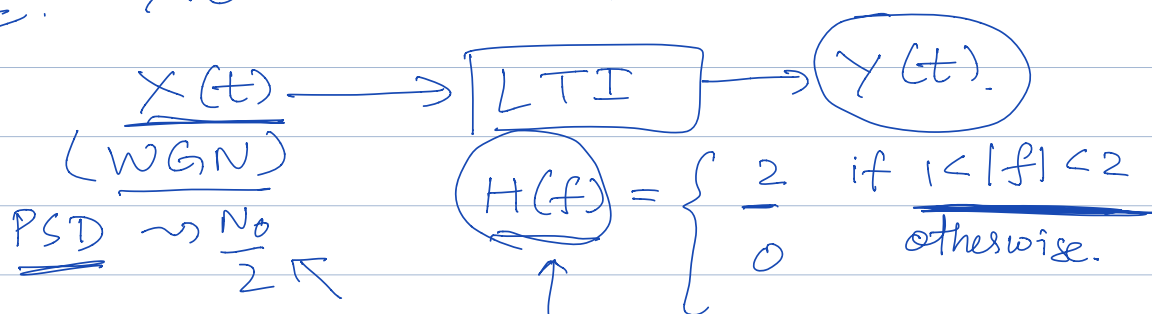
$$= E[X(t_1)X(t_2)]$$

$$= R_X(t_1 - t_2) = \frac{N_0}{2} \delta(t_1 - t_2)$$

for all  $t_1 \neq t_2$ .

Eg.:  $X(t)$  is a WGN

$$\mu_Y = \mu_X H(0)$$



Find  $P(Y(5) < \sqrt{N_0})$   $\leftarrow \leftarrow \leftarrow \leftarrow$

PSD of  $Y(t)$ ?

$$S_Y(f) = |H(f)|^2 \times \underbrace{S_X(f)}_{\text{input PSD.}}$$

$$= 4 \times \frac{N_0}{2} = 2N_0.$$

$$S_Y(f) = \begin{cases} 2N_0, & 1 < |f| < 2 \\ 0 & \text{otherwise.} \end{cases}$$

PSD of  $Y(t)$ .

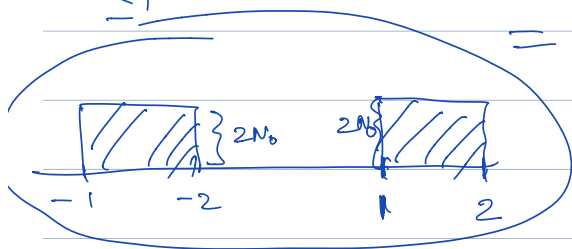
Distribution of  $Y(t)$ ?

$$Y(t) \sim \mathcal{N}(0, 4N_0).$$

$$\text{Var}[Y(t)] = E[Y(t)^2]$$

$$= R_Y(0) = \text{Power of } Y(t)$$

$$= \int_{-\infty}^{\infty} S_Y(f) df = 4N_0.$$

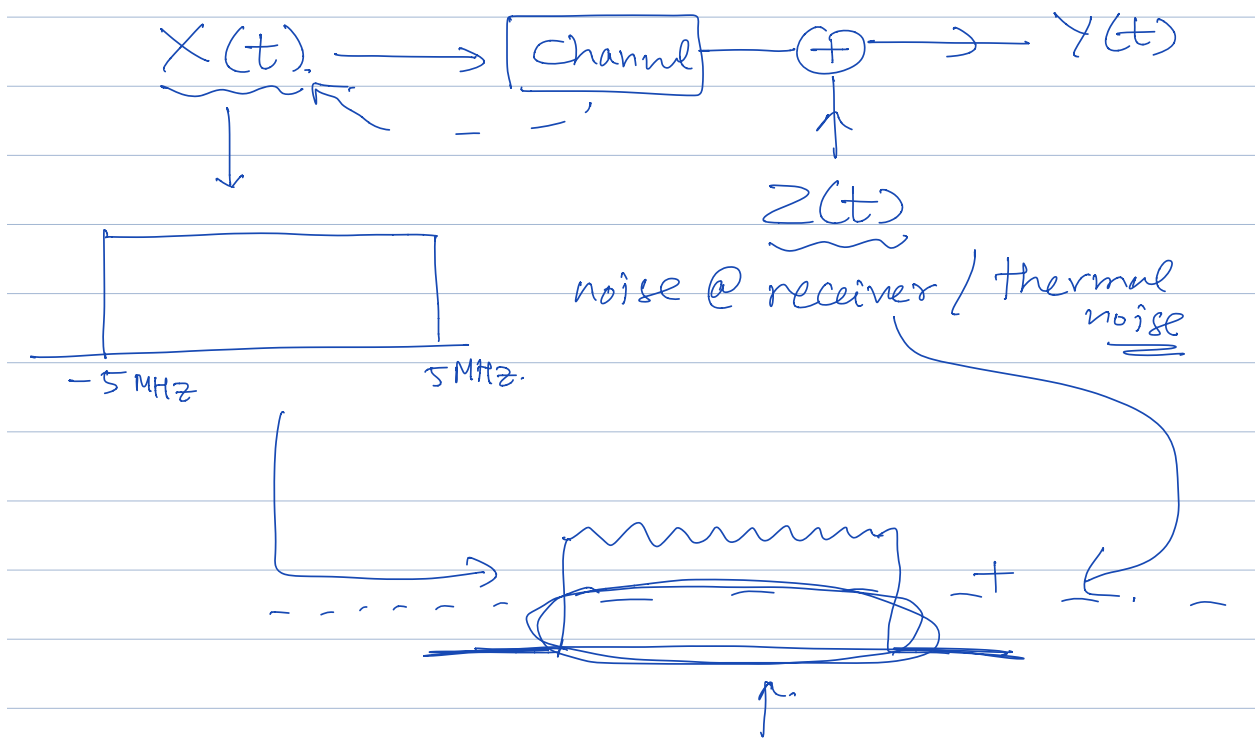


$$P(Y(5) < \sqrt{N_0}) = \Phi\left(\frac{\sqrt{N_0} - 0}{\sqrt{4N_0}}\right)$$

$$\downarrow$$

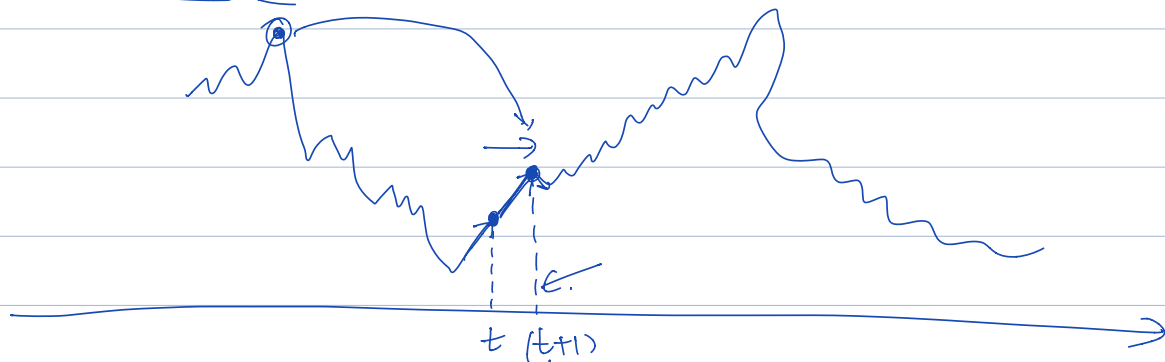
$$\mathcal{N}(0, 4N_0).$$

$$= \Phi(1/2) \approx 0.69.$$



## Markov Random Processes

Stocks (prices).



Weather. (temp / Sunny / Rainy).

Close dependence of R.P.  
over time.

\* Want to incorporate dependence on recent past.

\* Ignore dependence on much later past.  
( $X_m \rightsquigarrow m$  is time index)

$X_0, X_1, X_2, X_3, \dots, X_t, X_{t+1}, \dots$

↑

1<sup>st</sup> order M.C.

$$P(X_t = j \mid X_{t-1} = i, X_{t-2} = \dots, X_1 = \dots, X_0 = \dots)$$

$$= P(X_t = j \mid X_{t-1} = i)$$

(\*) Discrete time

(\*) Time-invariant transition probabilities.

↓  
transition probabilities

(\*) State Space of a M.C. (Markov chain).

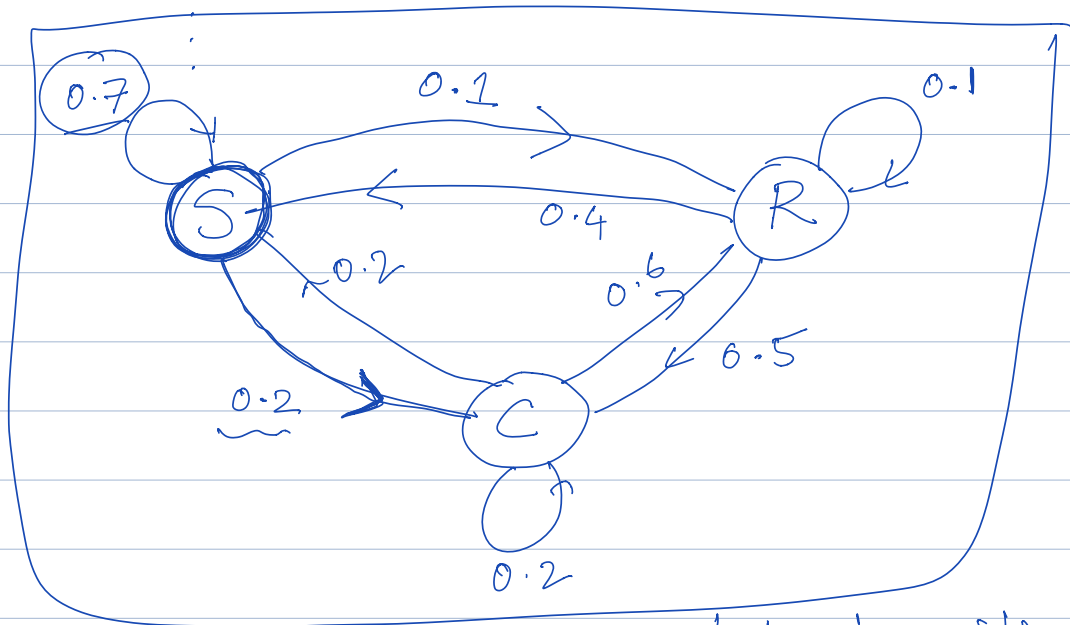
Weather  $\longrightarrow \{S, R, C\} =$  State space of M.C.

$X_t$  takes values either  $\begin{matrix} \nearrow S \\ \rightarrow R \\ \searrow C \end{matrix}$

$$P(X_t = S \mid X_{t-1} = S) = ?$$

$$P(X_t = S \mid X_{t-1} = R) = ?$$

$$P(X_t = S \mid X_{t-1} = C) = ?$$



State transition diagram

	S	R	C
S →	0.7	0.1	0.2
R →	0.4	0.1	0.5
C →	0.2	0.6	0.2

⇒ state transition matrix

P