

EE140 Introduction to Communication Systems Lecture 5

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Contents

Random signals

- Review of probability and random variables
- Random processes: basic concepts
- Gaussian white processes

Recall: Gaussian Distribution

 Gaussian or normal distribution is a continuous r.v. with pdf

$$f_X(x) = \frac{1}{\sqrt{2\pi\sigma_x^2}} \exp\left[-\frac{1}{2\sigma_x^2}(x - m_x)^2\right]$$

 A Gaussian r.v. is completely determined by its mean and variance, and hence usually denoted as

$$x \sim N(m_x, \sigma_x^2)$$

 $\boldsymbol{\mathcal{X}}$

mx

Gaussian Process

• Definition: X(t) is a Gaussian process if for all n and all t_1, t_2, \dots, t_n , the sample values $X(t_1), X(t_2), \dots, X(t_n)$ have a joint Gaussian density function

$$f_{X(t_1)X(t_2),\dots,X(t_n)}(x_1,x_2,\dots,x_n)$$

$$= \frac{1}{(2\pi)^{n/2}(\det(\mathbf{C}))^{1/2}} \exp\left[-\frac{(x-\mathbf{m})(\mathbf{C})^1(x-\mathbf{m})}{2}\right]$$

$$(x_1,x_2,\dots,x_n)$$

$$= \frac{1}{(2\pi)^{n/2}(\det(\mathbf{C}))^{1/2}} \exp\left[-\frac{(x-\mathbf{m})(\mathbf{C})^1(x-\mathbf{m})}{2}\right]$$

$$(x_1,x_2,\dots,x_n)$$

$$= \frac{1}{(2\pi)^{n/2}(\det(\mathbf{C}))^{1/2}} \exp\left[-\frac{(x-\mathbf{m})(\mathbf{C})^1(x-\mathbf{m})}{2}\right]$$

• Properties:

 If it is wide-sense stationary, it is also strictly stationary (Gaussian process is completely defined by its first order statistics m and second order statistics C.)

Gaussian Process

Properties:

- If the samples of Gaussian process $X(t_1), X(t_2), ..., X(t_n)$ are uncorrelated in time, they are also independent

$$f_{X(t_1)X(t_2),\dots,X(t_n)}(x_1,x_2,\dots,x_n) = \prod_{k=1}^n \frac{1}{\sqrt{2\pi}\sigma_k} \exp\left[-\frac{(x_k - a_k)^2}{2\sigma_k^2}\right]$$

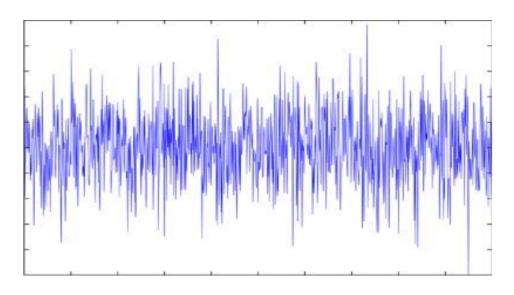
= $f_{X(t_1)}(x_1) \cdot f_{X(t_2)}(x_2) \cdot \dots \cdot f_{X(t_n)}(x_n)$

 If the input to a linear system is a Gaussian process, the output is also a Gaussian process

$$Y_o(t) = \int_{-\infty}^{\infty} h(\tau) X_i(t - \tau) d\tau$$

Noise

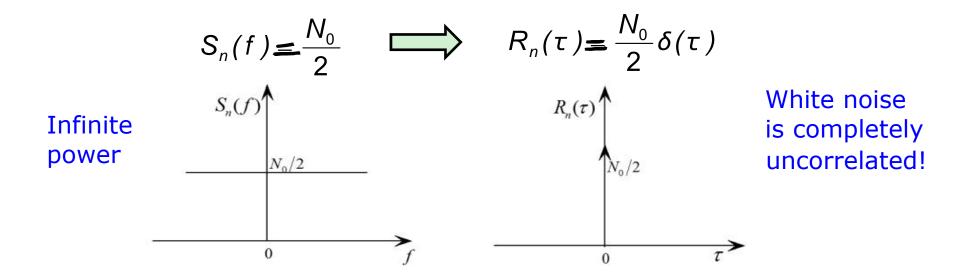
- Gaussian Noise:
 - often modeled as Gaussian and stationary with 0 mean



White noise (stationary and zero mean)

Noise

White Noise (stationary and zero mean)



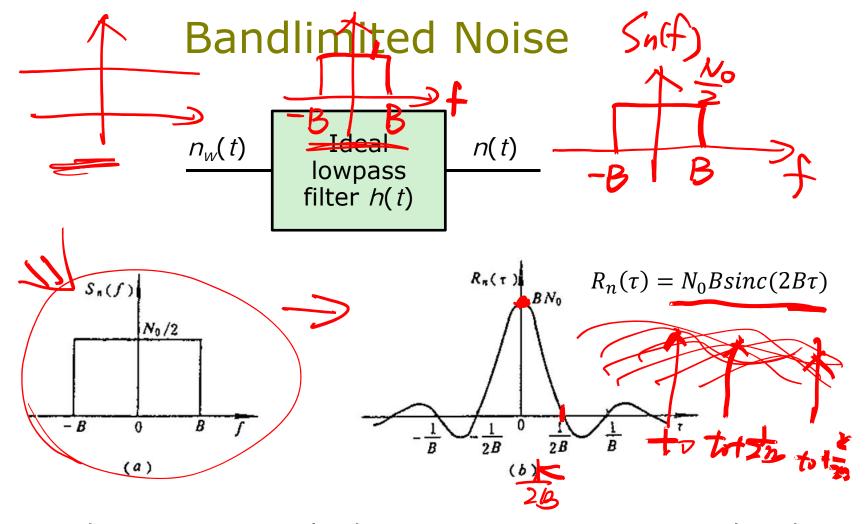
$$N_0 = KT = 4.14 \times 10^{-21}$$

= -174 dBm/Hz

 N_0 : single-sided power spectral density

 $\frac{N_0}{2}$: two-sided power spectral density

White Gaussian Noise (stationary and zero mean)



• Q1. At what rate to sample the noise can we get uncorrelated realizations? (2B/second)

• Q2. What is the power of each sample? (BN_0)

Noise Equivalent Bandwidth

- White noise: zero mean, two-sided PSD= $\frac{N_0}{2}$
- Arbitrary filter(H(f))

Average output noise powernut

$$(P_{n_0}) = \int_{-\infty}^{\infty} \frac{N_0}{2} |H(f)|^2 df = N_0 \int_0^{\infty} |H(f)|^2 df$$

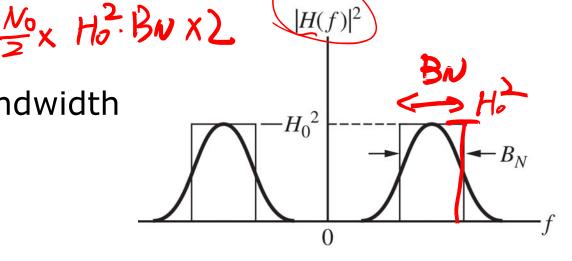
|H(f)| = |H(-f)|, if h(t) is real.

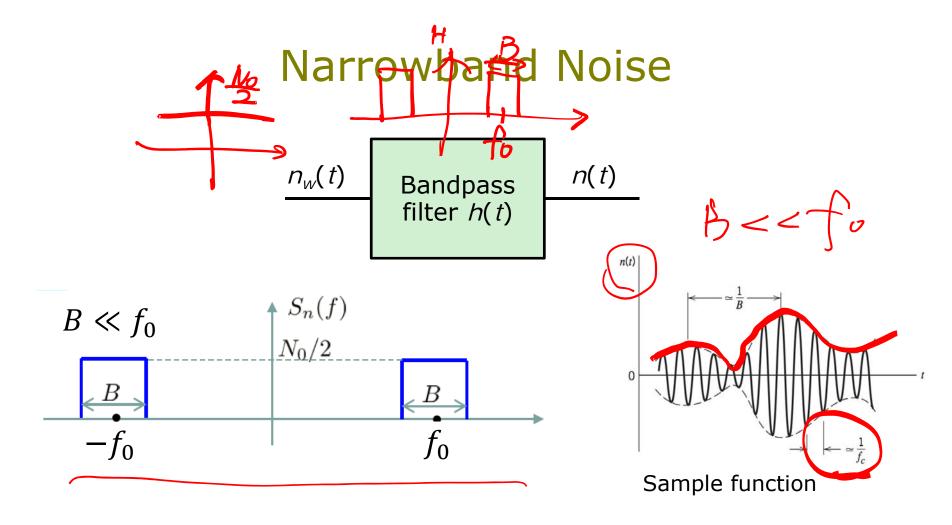
• Ideal filter: B_N, H_0

$$P_{n_0} = N_0 H_0^2 B_N$$

Noise equivalent bandwidth

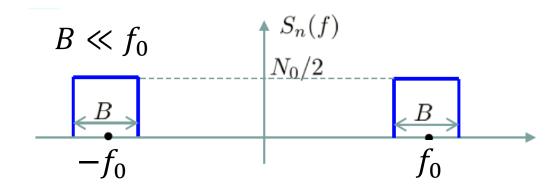
$$B_N = \frac{\int_0^\infty |H(f)|^2 df}{H_0^2}$$



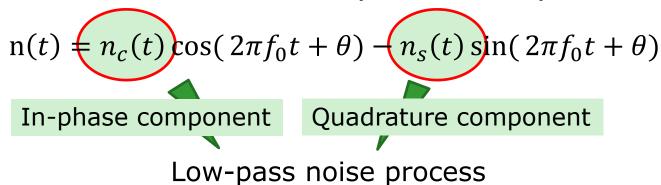


- Two specific representation of narrowband noise
 - In-phase and quadrature components
 - Envelope and phase

Narrowband Noise



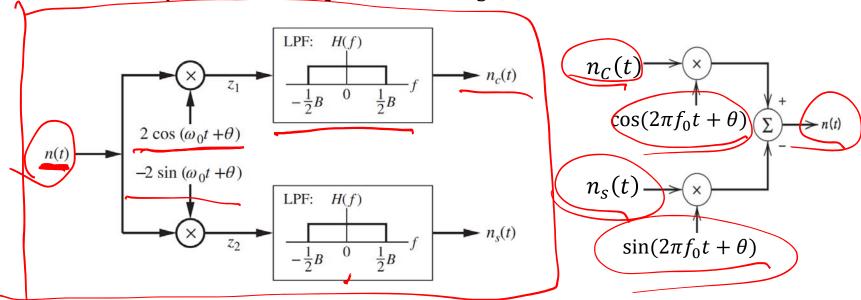
Canonical form of a band-pass noise process



 θ is an arbitrary phase angle

Narrowband Noise

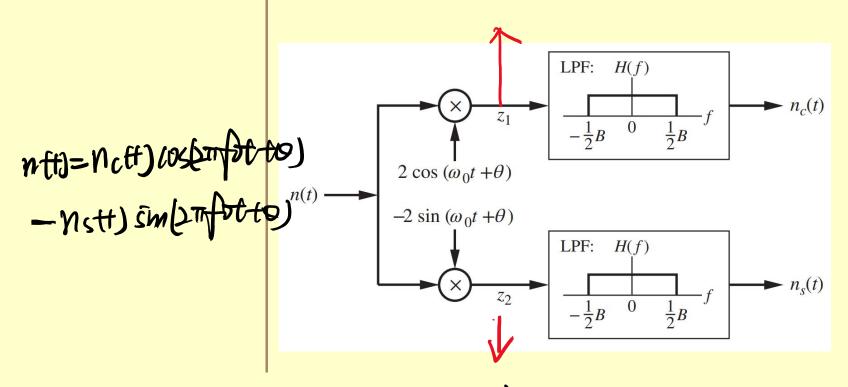
• How to produce $n_c(t)$ and $n_s(t)$



Why equality holds? (Proof: Page 712, Appendix C)

$$E\left\{\left[n(t) - \left[n_c\left(t\right)\cos(2\pi f_0 t + \theta) - n_s\left(t\right)\sin(2\pi f_0 t + \theta)\right]\right]^2\right\} = 0$$

Zitt) = n(+) -2005 (Wott+Q) = 2nctt) (052 (211 fot+0) -2nstt) &m(27 fot+0) (05/24 fot+0) = nctt) + nctt) cos (471 f.t + 20) - nctt) sim (471 fot + 20)



Z2H)= n(t). (-15m67+fot+10)) = -> nc(+) Sm(>17/5t+10) COS (>17/5t+10) +> nc(+) Sm2(>17/5t+10) = - nctt) Sm (47970t+20) + nctt) - nctt) Cos(47770t+0)

Properties

If n(t) is Gaussian RP $n_{\rm S}(t)$ and $n_{\rm C}(t)$ are joint Gaussian RP

If n(t) is stationary $n_s(t)$ and $n_c(t)$ are \mathcal{R}_{n_s} , \mathcal{R}_{n_c} jointly stationary \mathcal{R}_{n_s} , \mathcal{R}_{n_c}

 $n_{S}(t)$ and $n_{C}(t)$

Zero mean

Same PSD (autocorrelation, variance)

$$S_{n_c}(f) = S_{n_s}(f)$$

= Lp[$S_n(f - f_0) + S_n(f + f_0)$]

Cross-PSD (odd Cross-correlation)

$$S_{n_c n_s}(f) = j \text{Lp}[S_n(f - f_0) - S_n(f + f_0)]$$

$$n(t) \longrightarrow \begin{bmatrix} \sum_{t=1}^{LPF: H(f)} \\ -\frac{1}{2}B & 0 & \frac{1}{2}B \end{bmatrix} \longrightarrow n_{c}(t)$$

$$2 \times (t+1) \longrightarrow 2 \times (t+1) \longrightarrow n_{c}(t+1) \longrightarrow n_{c}(t+1)$$

$$R_{z,(t,t+z)} = E \left[\frac{4 \text{ net}}{mt+z} \right] \cos(w_0 t + \theta) \cos(w_0 t + z) + \theta$$

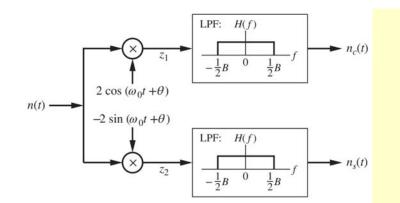
$$= 2 R_n(z) \cos(2\pi f_0 z)$$

$$Rz_2(t, t+z) = E[4 n(t)n(t+z) Sm(2\pi \beta t+0) Sm(2\pi \beta (t+2)+10)]$$

= $2Rn(z)(cos(2\pi \beta z) - E[cos(2\pi \beta (t+2)+10)])$
= $2Rn(z)(cos(2\pi \beta z))$

$$R_{2,122}$$
 (t.t+z)= $E[2,tt) \ge (t+z)$]
$$= -2R_{11}(z) E[2\cos(wot+0) \le m(wdt+z)+0)$$

$$= -2R_{11}(z) \le m \ne T_{11}(z)$$



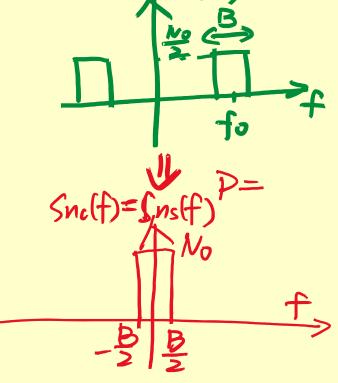
$$z_1(t) = \sum_{i=1}^{n} l(t) los(wot+10)$$
 $h_i(t) = \sum_{i=1}^{n} l(t) = \sum_{i=1}^{n} l(t) = \sum_{i=1}^{n} l(t) = \sum_{i=1}^{n} l(t) = l(t) =$

$$R_{z,l}(Z) = 2R_{l}(Z)(0)(2000)$$

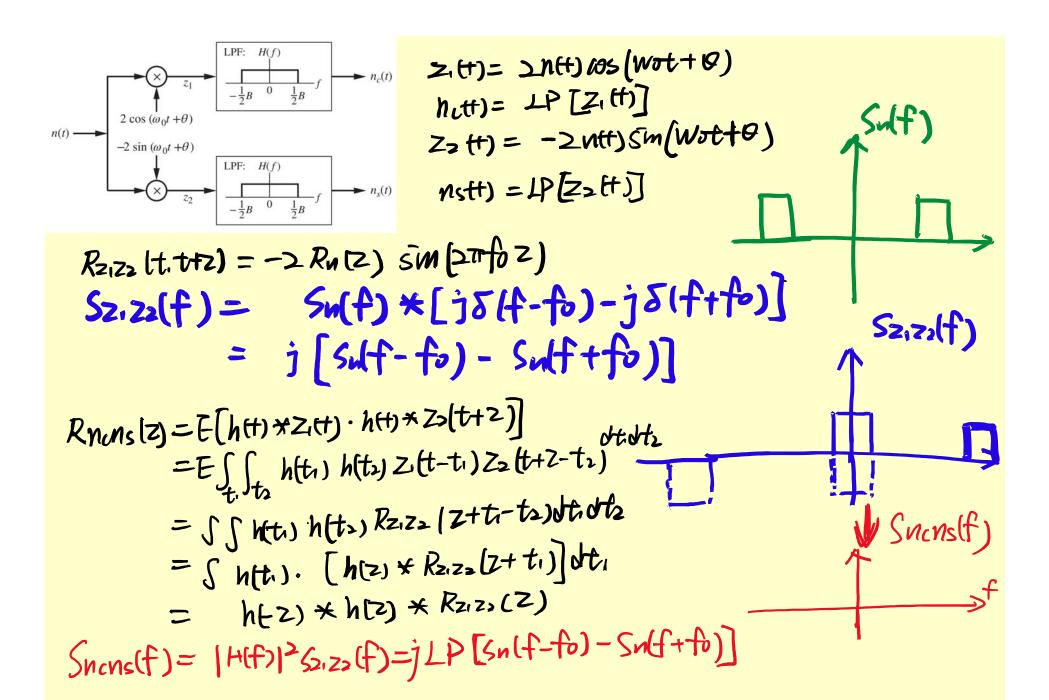
 $S_{z,l}(f) = S_{l}(f-f_{0}) + S_{l}(f+f_{0})$
 $S_{l}(f) = LP[S_{l}(f-f_{0}) + S_{l}(f+f_{0})]$

Rz2(Z) = 2 Rn(Z) cos
$$e^{\pi f_0 Z}$$

Sz2(f) = Sn (f-fo) + Sn(f+fo)
Sns(f) = LP [Sn (f-fo) + Sn (f+fo)]



P=NOB



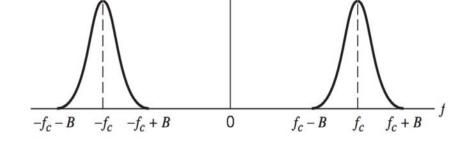
Properties

- Let n(t) be a zero-mean, stationary and Gaussian noise, then n_c(t) and n_s(t) satisfy the following properties
 - $n_c(t)$ and $n_s(t)$ are zero-mean, jointly stationary and jointly Gaussian process
 - Means: $E[n(t)] = E[n_c(t)] = E[n_s(t)] = 0$
 - PSD: $S_{n_c}(f) = S_{n_s}(f) = \text{Lp}[S_n(f f_0) + S_n(f + f_0)]$
 - Variances(power): $E[n^2(t)] = E[n_c^2(t)] = E[n_s^2(t)] = N_0 B \triangleq \sigma^2$
 - Correlation function:
 - $R_{n_c}(\tau) = R_{n_S}(\tau)$, $R_n(0) = R_{n_c}(0) = R_{n_S}(0)$
 - $R_{n_c n_s}(\tau) = -R_{n_c n_s}(-\tau)$ (odd), $R_{sc}(0) = R_{cs}(0) = 0$.
 - Cross-PSD: $S_{n_c n_s}(f) = j \text{Lp}[S_n(f f_0) S_n(f + f_0)]$
 - $R_{n_c n_s}(\tau) \equiv 0, \forall \tau$, if $Lp[S_n(f f_0) S_n(f + f_0)] = 0$.

Properties

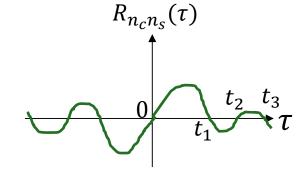
- Let n(t) be a zero-mean, stationary and Gaussian noise, then n_c(t) and n_s(t) satisfy the following properties
 - $n_c(t)$ and $n_s(t)$ are uncorrelated (independent) Gaussian process
 - 1. If PSD of n(t) is symmetric about f_0

$$R_{n_c n_s}(\tau) \equiv 0, \forall \tau$$



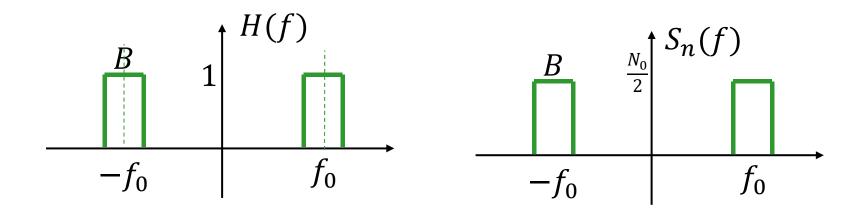
 $S_N(f)$

2. $\{\tau: R_{n_c n_s}(\tau) = 0\}$.

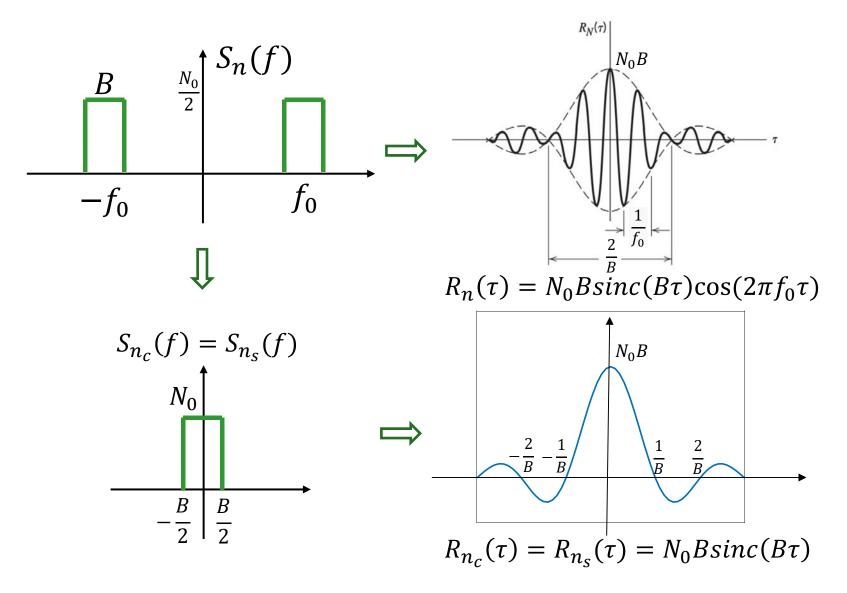


Example: Ideal band-pass filtered white noise

- Consider a white Gaussian noise of zero mean and PSD N0/2, which is passed through an ideal bandpass filter.
- Determine the autocorrelation function of n(t) and its in-phase and quadrature components.



Example: Ideal band-pass filtered white noise

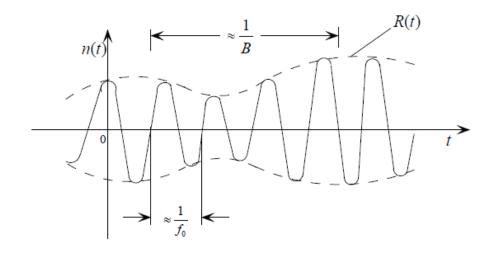


Envelope and Phase

Angular representation of n(t)

$$n(t) = R(t)\cos(\omega_0 t + \theta + \varphi(t))$$

where
$$\begin{cases} R(t) = \sqrt{n_c^2(t) + n_s^2(t)} & \text{envelope} \\ \varphi(t) = \tan^{-1} \frac{n_s(t)}{n_c(t)}, [0 \le \varphi(t) \le 2\pi] \end{cases}$$
 phase

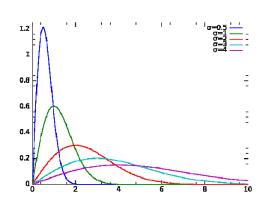


Example

- Let n(t) be a zero-mean, stationary Gaussian process, find the statistics of the envelop and phase
- Result:
 - Envelop follows Rayleigh distribution
 - Phase follows uniform distribution

$$\begin{cases} f(R) = \int_0^{2\pi} f(R,\varphi)d\varphi = \frac{R}{\sigma^2} \exp\left\{-\frac{R^2}{2\sigma^2}\right\}, R \ge 0 \end{cases}$$

$$\begin{cases} f(\varphi) = \int_0^{\infty} f(R,\varphi)dR = \frac{1}{2\pi}, 0 \le \varphi \le 2\pi \end{cases}$$



 For the same t, the envelop variable R and phase variable φ are independent (but not the two processes)

Rayleigh fading channel: model the fading channel with random scatters.

Derivation

Derivation

$$f(R,\varphi) = f(n_c, n_s) \left| \frac{\partial(n_c, n_s)}{\partial(R, \varphi)} \right|$$

$$f(n_c, n_s) = f(n_c) \cdot f(n_s) = \frac{1}{2\pi\sigma^2} \exp\left[-\frac{n_c^2 + n_s^2}{2\sigma^2}\right]$$

$$\left| \frac{\partial (n_{c}, n_{s})}{\partial (R, \varphi)} \right| = \left| \frac{\partial n_{c}}{\partial R} \frac{\partial n_{s}}{\partial R} \frac{\partial n_{s}}{\partial \varphi} \right| = \left| \frac{\cos \varphi}{-R \sin \varphi} \frac{\sin \varphi}{R \cos \varphi} \right| = R$$

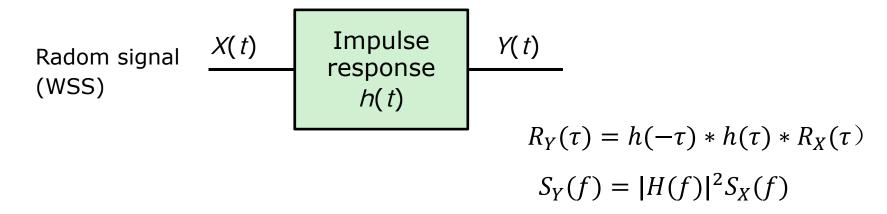
$$f(R,\varphi) = Rf(n_c, n_s) = \frac{R}{2\pi\sigma^2} \exp\left[-\frac{(R\cos\varphi)^2 + (R\sin\varphi)^2}{2\sigma^2}\right]$$
$$= \frac{R}{2\pi\sigma^2} \exp\left\{-\frac{R^2}{2\sigma^2}\right\}$$

Summary

For WSS:

$$S_X(f) \leftrightarrow R_X(\tau)$$

WSS transmission through a linear system



- White noise (zero-mean)
 - Bandlimited noise
 - Narrowband noise: Gaussian, stationary and zero-mean
 - Non-Gaussian?

Sine Wave with Bandpass Noise

Received signal

$$r(t) = A\cos(\omega_c t + \theta) + n(t)$$

where A, ω_c are deterministic, θ is random phase uniformly distributed in $[-\pi, \pi]$, n(t) is narrowband noise (zero-mean, stationary Gaussian process).

$$r(t) = [A\cos\theta + n_c(t)]\cos\omega_c t - [A\sin\theta + n_s(t)]\sin\omega_c t$$
$$= z_c(t)\cos\omega_c t - z_s(t)\sin\omega_c t$$
$$= z(t)\cos[\omega_c t + \varphi(t)]$$

$$z_c(t) = A\cos\theta + n_c(t)$$

$$z_{S}(t) = A\sin\theta + n_{S}(t)$$

Sine Wave with Bandpass Noise (cont'd)

Envelop

$$z(t) = \sqrt{z_c^2(t) + z_s^2(t)}, z \ge 0$$

Phase

$$\varphi(t) = \tan^{-1} \frac{z_s(t)}{z_c(t)}$$

• Given θ

$$E[z_c] = A \cos \theta$$

$$E[z_s] = A \sin \theta$$

$$\sigma_c^2 = \sigma_s^2 = \sigma_n^2$$

Joint distribution

$$f(z_c, z_s | \theta) = \frac{1}{2\pi\sigma_n^2} \exp\left\{-\frac{1}{2\sigma_n^2} [(z_c - A\cos\theta)^2 + (z_s - A\sin\theta)^2]\right\}$$

PDF of the Amplitude

$$f(z_c, z_s | \theta) = \frac{1}{2\pi\sigma_n^2} \exp\left\{-\frac{1}{2\sigma_n^2} \left[(z_c - A\cos\theta)^2 + (z_s - A\sin\theta)^2 \right] \right\}$$
$$f(z, \varphi | \theta) = f(z_c, z_s | \theta) \left| \frac{\partial (z_c, z_s)}{\partial (z, \varphi)} \right|$$
$$= \begin{vmatrix} \cos\varphi & \sin\varphi \\ -z\sin\varphi & z\cos\varphi \end{vmatrix} f(z_c, z_s | \theta) = z \cdot f(z_c, z_s | \theta)$$

For PDF of the amplitude

$$f(z|\theta) = \int_0^{2\pi} f(z, \varphi|\theta) d\varphi$$
$$= \frac{z}{2\pi\sigma_n^2} \exp\left[-\frac{z^2 + A^2}{2\sigma_n^2}\right] \int_0^{2\pi} \exp\left[\frac{Az}{\sigma_n^2} \cos(\theta - \varphi)\right] d\varphi$$

PDF of the Amplitude (cont'd)

Amplitude

$$f(z|\theta) = \int_0^{2\pi} f(z, \varphi|\theta) d\varphi$$

$$= \frac{z}{2\pi\sigma_n^2} \exp\left[-\frac{z^2 + A^2}{2\sigma_n^2}\right] \int_0^{2\pi} \exp\left[\frac{Az}{\sigma_n^2} \cos(\theta - \varphi)\right] d\varphi$$

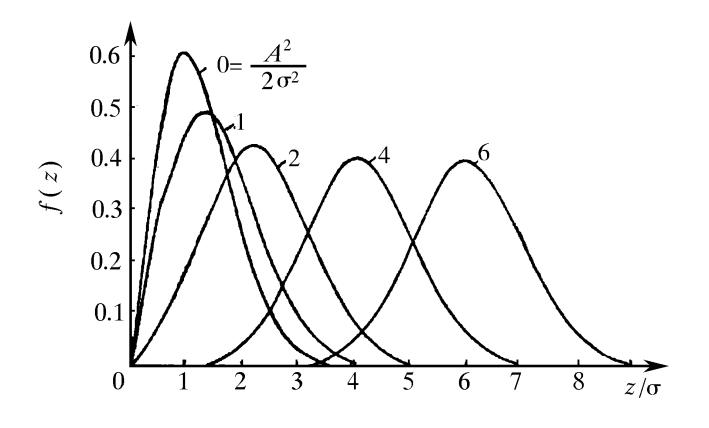
$$= \frac{z}{2\pi\sigma_n^2} \exp\left[-\frac{z^2 + A^2}{2\sigma_n^2}\right] I_0\left(\frac{Az}{\sigma_n^2}\right)$$

Ricean distribution

$$I_0\left(\frac{Az}{\sigma_n^2}\right)$$

 $I_0\left(\frac{Az}{\sigma^2}\right)$ 0th order Bessel functions of the first kind

PDF of the Amplitude (cont'd)



A small,

$$I_0\left(\frac{Az}{\sigma_n^2}\right) \approx 1$$

Rayleigh distribution

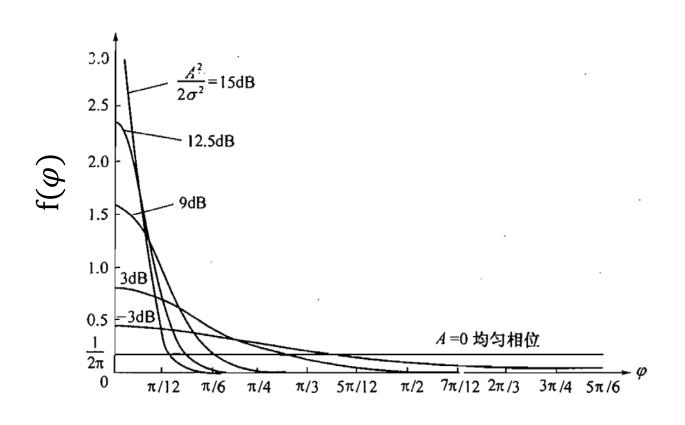
A large,

$$I_0(x) \approx \frac{e^x}{\sqrt{2\pi x}}$$

Gaussian distribution

Ricean fading channel: model the fading channel with a direct path and scatters.

PDF of the Amplitude (cont'd)



- A small,
 Uniform
 distribution
- A large, Concentrate around θ

Ricean fading channel: model the fading channel with a direct path and scatters.



Thanks for your kind attention!

Questions?