

TSDT14 Signal Theory

Lecture 8

Analog Modulation – Part 2

Repetition of Sampling and PAM (deterministically) Sampling and PAM of Stochastic Processes

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Broadband Angle Modulation

Modulation of a Gaussian process

Assumptions: $\Phi_d(t)$ is an ergodic Gaussian process, mean 0.

Ψ is uniform on $[0, 2\pi)$ and indep of $\Phi_d(t)$.

Signal: $X(t) = A \cos(2\pi f_c t + \Phi_d(t) + \Psi) \Rightarrow m_X = 0$

Phase dev.: $\Phi_d(t)$

Freq. dev.: $F_d(t) = \frac{1}{2\pi} \cdot \frac{d}{dt} \Phi_d(t)$

Relation: $R_{F_d}(f) = f^2 R_{\Phi_d}(f). \Rightarrow r_{F_d}(\tau) = -\frac{1}{(2\pi)^2} \cdot \frac{d^2}{d\tau^2} r_{\Phi_d}(\tau)$

Definitions: $\phi_d^2 = 2r_{\Phi_d}(0)$ $f_d^2 = 2r_{F_d}(0)$,

ACF of Angle Modulation 1(3)

$$\begin{aligned} r_X(t + \tau, t) &= E \left\{ A \cos(2\pi f_c(t + \tau) + \Phi_d(t + \tau) + \Psi) \cdot A \cos(2\pi f_c t + \Phi_d(t) + \Psi) \right\} \\ &= \frac{A^2}{2} E \left\{ \cos(2\pi f_c(2t + \tau) + \Phi_d(t + \tau) + \Phi_d(t) + 2\Psi) \right\} \\ &\quad + \frac{A^2}{2} E \left\{ \cos(2\pi f_c \tau + \Phi_d(t + \tau) - \Phi_d(t)) \right\}, \\ &= \frac{A^2}{2} E \left\{ \cos(2\pi f_c \tau + \Phi_d(t + \tau) - \Phi_d(t)) \right\} \\ &= \frac{A^2}{2} E \left\{ \text{Re} \left\{ e^{j[2\pi f_c \tau + \Phi_d(t + \tau) - \Phi_d(t)]} \right\} \right\} \\ &= \frac{A^2}{2} \text{Re} \left\{ E \left\{ e^{j[\Phi_d(t + \tau) - \Phi_d(t)]} \right\} e^{j2\pi f_c \tau} \right\}, \end{aligned}$$

ACF of Angle Modulation 2(3)

We had: $r_X(t + \tau, t) = \frac{A^2}{2} \text{Re} \left\{ E \left\{ e^{j[\Phi_d(t + \tau) - \Phi_d(t)]} \right\} e^{j2\pi f_c \tau} \right\}$

Temporary Gaussian variable: $Y = \Phi_d(t + \tau) - \Phi_d(t)$ (mean 0)

Then: $E \{ e^{jY} \} = E \{ \cos(Y) \} + j E \{ \sin(Y) \} = E \{ \cos(Y) \}$,

Important observation: This is real valued.

Rewrite: $r_X(t + \tau, t) = \frac{A^2}{2} E \{ e^{jY} \} \text{Re} \{ e^{j2\pi f_c \tau} \} = \frac{A^2}{2} E \{ e^{jY} \} \cos(2\pi f_c \tau)$

We have, with $f = -1/2\pi$, using the identity $\mathcal{F}_x \{ e^{-\pi(x/T)^2} \} = T e^{-\pi(fT)^2}$:

$$\begin{aligned} E \{ e^{jY} \} &= E \{ e^{-j2\pi f Y} \} = \mathcal{F}_y \{ f_Y(y) \} = \mathcal{F}_y \left\{ \frac{1}{\sqrt{2\pi}\sigma_Y} e^{-y^2/2\sigma_Y^2} \right\} \\ &= e^{-(2\pi f \sigma_Y)^2/2} = e^{-\sigma_Y^2/2} \end{aligned}$$

ACF of Angle Modulation 3(3)

We had: $r_X(t + \tau, t) = \frac{A^2}{2} e^{-\sigma_Y^2/2} \cos(2\pi f_c \tau)$

Recall: $Y = \Phi_d(t + \tau) - \Phi_d(t)$ (mean 0)

We have: $\sigma_Y^2 = E \left\{ (\Phi_d(t + \tau) - \Phi_d(t))^2 \right\} = 2r_{\Phi_d}(0) - 2r_{\Phi_d}(\tau)$.

Mean and ACF independent of time t . Wide sense stationary!

$$m_X = 0 \quad r_X(\tau) = \frac{A^2}{2} \cdot r_{\tilde{X}}(\tau) \cdot \cos(2\pi f_c \tau) \quad r_{\tilde{X}}(\tau) = e^{r_{\Phi_d}(\tau) - r_{\Phi_d}(0)}$$

The introduced process $\tilde{X}(t)$ is a normalized baseband representation of $X(t)$.

Bandwidth of Angle Modulation

Define bandwidth B_X as the width of the frequency band, symmetrically around the carrier frequency that contains 90% of the power.

$$\int_{-B_X/2}^{B_X/2} R_{\tilde{X}}(f) df = 0.9.$$

Broadband $\Rightarrow m_{\tilde{X}}^2$ very small. Ignore!

$$\int_{-B_X/2}^{B_X/2} \frac{1}{\sqrt{\pi} \cdot f_d} \cdot e^{-(f/f_d)^2} df \approx 0.9.$$

We get:

$$\int_{-B_X/2}^{\infty} \frac{1}{\sqrt{\pi} \cdot f_d} \cdot e^{-(f/f_d)^2} df = Q\left(\frac{B_X/2}{f_d/\sqrt{2}}\right) \approx 0.05. \quad \Rightarrow \quad B_X \approx 2.33 f_d.$$

Compare Carson's rule: Bandwidth is slightly more than $2 \cdot f_{d,\max}$.

PSD of Angle Modulation

We had: $r_X(\tau) = \frac{A^2}{2} \cdot r_{\tilde{X}}(\tau) \cdot \cos(2\pi f_c \tau) \quad r_{\tilde{X}}(\tau) = e^{r_{\Phi_d}(\tau) - r_{\Phi_d}(0)}$

Recall: $\phi_d^2 = 2r_{\Phi_d}(0) \quad f_d^2 = 2r_{F_d}(0),$

This holds if $r_{\Phi_d}(\tau)$ has no periodic components and $\Phi_d(t)$ has mean 0:

$$r_{\tilde{X}}(\tau) \approx m_{\tilde{X}}^2 + (1 - m_{\tilde{X}}^2) \cdot e^{-2\pi^2 r_{F_d}(0) \tau^2} = m_{\tilde{X}}^2 + (1 - m_{\tilde{X}}^2) \cdot e^{-(\pi f_d \tau)^2}$$

$$m_{\tilde{X}}^2 = e^{-r_{\Phi_d}(0)} = e^{-\phi_d^2/2}.$$

Spectrum: $R_{\tilde{X}}(f) \approx m_{\tilde{X}}^2 \delta(f) + \frac{1 - m_{\tilde{X}}^2}{\sqrt{\pi} \cdot f_d} \cdot e^{-(f/f_d)^2}.$

↑ Carrier ↑ Sidebands

The only relation to $R_{\Phi_d}(f)$ is through $r_{F_d}(0)$ and $r_{\Phi_d}(0)$.

Special Cases: Broadband PM and FM

Phase modulation:

$$\Phi_d(t) = a M(t),$$

$$r_{\Phi_d}(\tau) = a^2 r_M(\tau) = \frac{\phi_d^2}{2} \cdot \frac{r_M(\tau)}{r_M(0)},$$

$$R_{\Phi_d}(f) = a^2 R_M(f) = \frac{\phi_d^2}{2} \cdot \frac{R_M(f)}{r_M(0)}.$$

Frequency modulation

$$\Phi_d(t) = a \int M(t) dt,$$

$$F_d(t) = \frac{1}{2\pi} \cdot \frac{d}{dt} \Phi(t) = \frac{a}{2\pi} M(t),$$

$$r_{F_d}(\tau) = \left(\frac{a}{2\pi}\right)^2 r_M(\tau) = \frac{f_d^2}{2} \cdot \frac{r_M(\tau)}{r_M(0)},$$

$$R_{F_d}(f) = \left(\frac{a}{2\pi}\right)^2 R_M(f) = \frac{f_d^2}{2} \cdot \frac{R_M(f)}{r_M(0)}.$$

Narrowband Angle Modulation

We have:

$$\begin{aligned} X(t) &= A \cos(2\pi f_c t + \Phi_d(t) + \Psi) \\ &= A \cos(\Phi_d(t)) \cos(2\pi f_c t + \Psi) - A \sin(\Phi_d(t)) \sin(2\pi f_c t + \Psi) \end{aligned}$$

Narrowband modulation, i.e. $\Phi_d(t)$ is small:

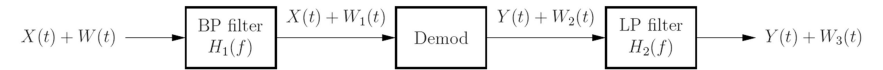
$$X(t) \approx A \cos(2\pi f_c t + \Psi) - A \Phi_d(t) \sin(2\pi f_c t + \Psi).$$

Spectrum:

$$R_X(f) = \underbrace{\frac{A^2}{4}(\delta(f + f_c) + \delta(f - f_c))}_{\text{Carrier}} + \underbrace{\frac{A^2}{4}(R_{\Phi_d}(f + f_c) + R_{\Phi_d}(f - f_c))}_{\text{Sidebands}}$$

We get AM in the quadrature component.

AWGN Impact on Phase Modulation



Output signal: $A a m(t)$ Output signal power: $P_Y = A^2 a^2 P_M$

Can be shown: $R_{W_2}(f) = \begin{cases} 2R_0, & |f| < B_X/2, \\ 0, & \text{elsewhere,} \end{cases}$

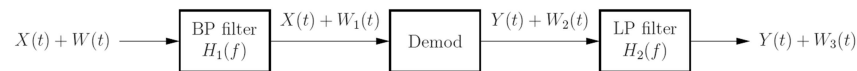
Filter: $H_2(f) = \begin{cases} 1, & |f| < B, \\ 0, & \text{elsewhere.} \end{cases}$

Output noise: $R_{W_3}(f) = |H_2(f)|^2 R_{W_2}(f) = \begin{cases} 2R_0, & |f| < B, \\ 0, & \text{elsewhere,} \end{cases}$

Noise power: $P_{W_3} = \int_{-\infty}^{\infty} R_{W_3}(f) df = \int_{-B}^B 2R_0 df = 4BR_0.$

Output SNR: $\frac{P_Y}{P_{W_3}} = \frac{A^2 a^2 P_M}{4BR_0}$

AWGN Impact on Angle Modulation



Noise PSD: $R_W(f) = R_0$

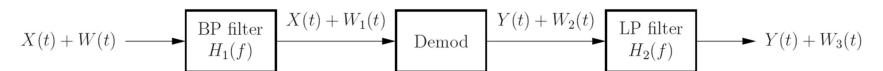
Filter: $H_1(f) = \begin{cases} 1, & ||f| - f_c| < B_X/2, \\ 0, & \text{elsewhere.} \end{cases}$

Filtered noise PSD: $R_{W_1}(f) = |H_1(f)|^2 R_W(f) = \begin{cases} R_0, & ||f| - f_c| < B_X/2, \\ 0, & \text{elsewhere,} \end{cases}$

Filtered noise power: $P_{W_1} = \int_{-\infty}^{\infty} R_{W_1}(f) df = 2 \int_{f_c - B_X/2}^{f_c + B_X/2} R_0 df = 2B_X R_0.$

Signal power: $P_X = \frac{A^2}{2}$ Input SNR: $\frac{P_X}{P_{W_1}} = \frac{A^2/2}{2B_X R_0} = \frac{A^2}{4B_X R_0}$

AWGN Impact on Frequency Modulation



Output signal: $A a m(t)$ Output signal power: $P_Y = A^2 a^2 P_M$

Can be shown: $R_{W_2}(f) = \begin{cases} 2f^2 R_0, & |f| < B_X/2, \\ 0, & \text{elsewhere,} \end{cases}$

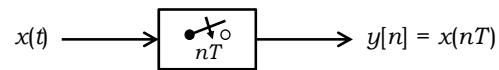
Filter: $H_2(f) = \begin{cases} 1, & |f| < B, \\ 0, & \text{elsewhere.} \end{cases}$

Output noise: $R_{W_3}(f) = \begin{cases} 2f^2 R_0, & |f| < B, \\ 0, & \text{elsewhere.} \end{cases}$

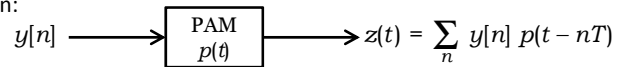
Noise power: $P_{W_3} = \frac{4B^3 R_0}{3}.$ Output SNR: $\frac{P_Y}{P_{W_3}} = \frac{3A^2 a^2 P_M}{4B^3 R_0},$

Linear Mappings

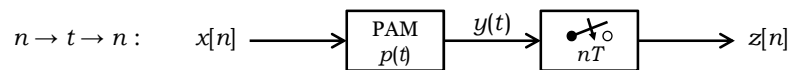
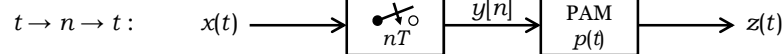
Sampling:



Pulse-Amplitude Modulation:
(PAM)



Reconstruction:



Sampling – Frequency Domain

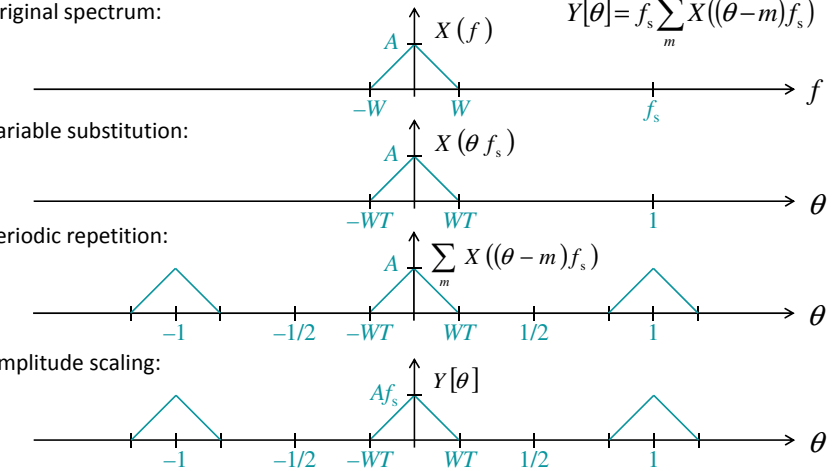
Original spectrum:

$$Y[\theta] = f_s \sum_m X((\theta - m)f_s)$$

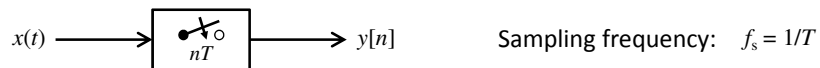
Variable substitution:

Periodic repetition:

Amplitude scaling:



Sampling – Deterministically (from S&S)



Time domain: $y[n] = x(nT)$

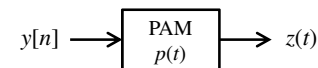
Frequency domain: $Y[\theta] = \frac{1}{T} \sum_m X\left(\frac{\theta - m}{T}\right) = f_s \sum_m X((\theta - m)f_s)$

If $X(f) = 0$ for $|f| \geq f_s/2$:

$$Y[\theta] = f_s X(\theta f_s) \quad \text{for } |\theta| < 1/2$$

$$Y[\theta] = Y[\theta + k] \quad \text{for } k \text{ integer}$$

PAM – Deterministically (from S&S)



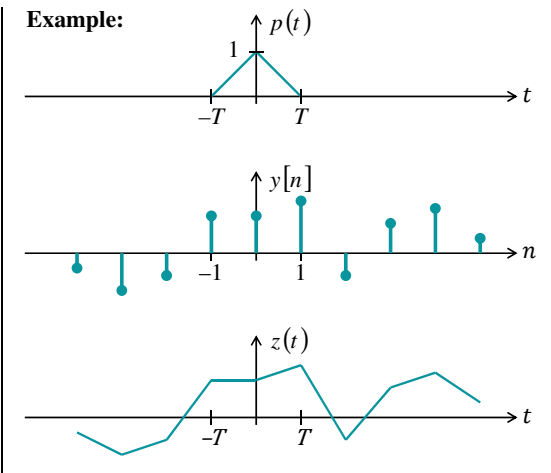
Time domain:

$$z(t) = \sum_n y[n] p(t - nT)$$

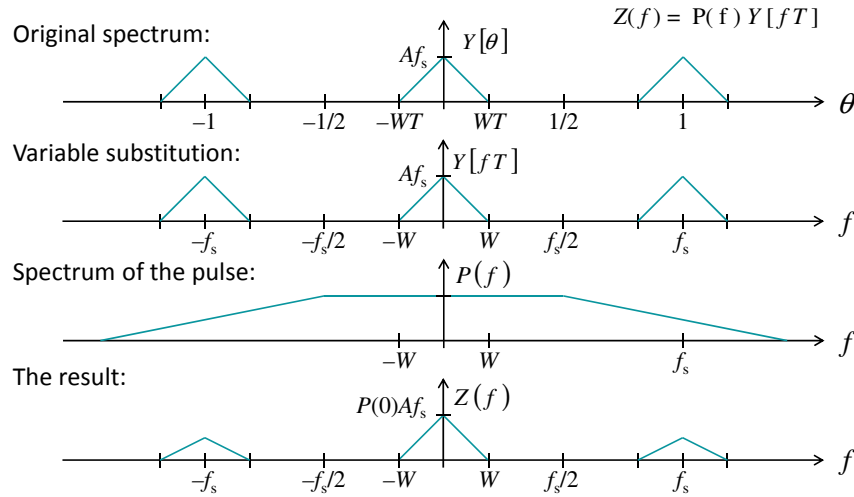
Frequency domain:

$$Z(f) = P(f) Y[fT]$$

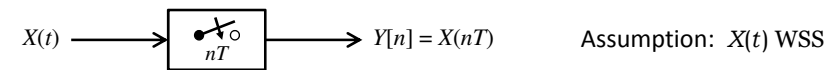
Example:



PAM – Frequency Domain



Sampling of a Stochastic Process 2(2)



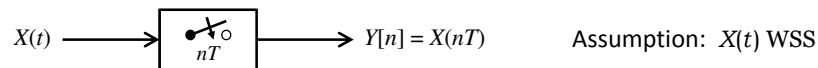
$$R_Y[\theta] = \frac{1}{T} \sum_m R_X\left(\frac{\theta - m}{T}\right) = f_s \sum_m R_X((\theta - m)f_s)$$

If $R_X(f) = 0$ for $|f| \geq f_s/2$:

$$R_Y[\theta] = f_s R_X(\theta f_s) \text{ for } |\theta| < 1/2$$

$$R_Y[\theta] = R_Y[\theta + k] \text{ for } k \text{ integer}$$

Sampling of a Stochastic Process 1(2)



Mean: $m_Y[n] = E\{Y[n]\} = E\{X(nT)\} = m_X$

Definition

Sampling

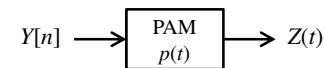
$X(t)$ WSS

ACF: $r_Y[n, n+k] = E\{Y[n]Y[n+k]\} = E\{X(nT)X((n+k)T)\} = r_X(kT)$

This is sampling of the deterministic function $r_X(\tau)$.

PSD: $R_Y[\theta] = \frac{1}{T} \sum_m R_X\left(\frac{\theta - m}{T}\right) = f_s \sum_m R_X((\theta - m)f_s)$

PAM of a Stochastic Process 1(3)



Time domain (naïve approach):

$$Z(t) = \sum_n Y[n] p(t - nT)$$

Problem: Does not retain WSS.

$$m_Z(t) = m_Y \sum_n p(t - nT)$$

$$r_Z(t+\tau) = \sum_n \sum_m p(t-nT) p(t+\tau-mT) r_Y[n-m]$$

Time domain (alternative approach): Ψ unif. $[0, T]$

$$Z(t) = \sum_n Y[n] p(t - nT - \Psi) \quad \Psi \text{ \& } Y[n] \text{ indep.}$$

Mean:

$$m_Z(t) = E\{Z(t)\} = E\left\{\sum_n Y[n] p(t - nT - \Psi)\right\}$$

$$= \sum_n E\{Y[n]\} E\{p(t - nT - \Psi)\} \quad (\text{indep.})$$

$$= \sum_n m_Y \int_{-\infty}^{\infty} p(t - nT - \psi) f_\Psi(\psi) d\psi$$

$$= m_Y \sum_n \int_0^T p(t - nT - \psi) \frac{1}{T} d\psi = \left/ \begin{matrix} \phi = t - nT - \psi \\ d\phi = -d\psi \end{matrix} \right/$$

$$= m_Y \frac{1}{T} \sum_n \int_{t-nT-T}^{t-nT} p(\phi) d\phi = m_Y \frac{1}{T} \int_{-\infty}^{\infty} p(\phi) d\phi$$

$$= \frac{1}{T} P(0) m_Y$$

Independent of t .

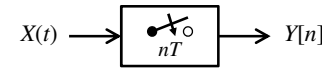
PAM of a Stochastic Process 2(3)

ACF:

$$\begin{aligned}
 r_z(t + \tau, t) &= E\{Z(t + \tau)Z(t)\} = E\left\{\sum_n Y[n]p(t + \tau - nT - \Psi)\sum_m Y[m]p(t - mT - \Psi)\right\} \\
 &= \sum_n \sum_m E\{Y[n]Y[m]\}E\{p(t + \tau - nT - \Psi)p(t - mT - \Psi)\} \\
 &= \sum_n \sum_m r_Y[n - m] \int_0^T p(t + \tau - nT - \psi)p(t - mT - \psi) \frac{1}{T} d\psi = \left/ k = n - m \right/ \\
 &= \frac{1}{T} \sum_k r_Y[k] \sum_m \int_0^T p(t + \tau - (k + m)T - \psi)p(t - mT - \psi) d\psi = \left/ \phi = \psi + mT - t \right/ \\
 &= \frac{1}{T} \sum_k r_Y[k] \sum_m \int_{mT - t}^{T + mT - t} p(\tau - kT - \phi)p(-\phi) d\phi = \left/ \text{define } \tilde{p}(t) = p(-t) \right/ \\
 &= \frac{1}{T} \sum_k r_Y[k] \int_{-\infty}^{\infty} p(\tau - kT - \phi)\tilde{p}(\phi) d\phi = \frac{1}{T} \sum_k r_Y[k] (p * \tilde{p})(\tau - kT)
 \end{aligned}$$

Independent of t .

Sampling and PAM of WSS Processes – Summary

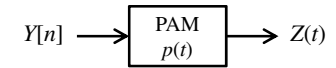


$$Y[n] = X(nT)$$

$$m_Y = m_X$$

$$r_Y[k] = r_X(kT)$$

$$R_Y[\theta] = \frac{1}{T} \sum_m R_X\left(\frac{\theta - m}{T}\right)$$



$$Z(t) = \sum_n Y[n] p(t - nT - \Psi)$$

$$m_Z = \frac{1}{T} P(0) m_Y$$

Ψ unif. $[0, T]$
 Ψ & $Y[n]$ indep.

$$r_Z(\tau) = \frac{1}{T} \sum_k r_Y[k] (p * \tilde{p})(\tau - kT)$$

$$R_Z(f) = \frac{1}{T} |P(f)|^2 R_Y[fT]$$

PAM of a Stochastic Process 3(3)

Mean: $m_Z = \frac{1}{T} P(0) m_Y$

Independent of t .
Thus WSS

ACF: $r_Z(\tau) = \frac{1}{T} \sum_k r_Y[k] (p * \tilde{p})(\tau - kT)$

PSD: $R_Z(f) = \mathcal{F} \left\{ \frac{1}{T} \sum_k r_Y[k] (p * \tilde{p})(\tau - kT) \right\}$

Notice: The sum is PAM of $r_Y[k]$ using pulse $(p * \tilde{p})(\tau)$.

$$\mathcal{F} \{(p * \tilde{p})(\tau)\} = |P(f)|^2$$

Recall deterministic PAM:

$$R_Z(f) = \frac{1}{T} |P(f)|^2 R_Y[fT]$$

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