

SYNCHRONIZATION TECHNIQUES FOR DIGITAL DEMODULATION

SIGNAL MODEL

For a given modulation, the signal model is 'a priori' known:

$$r(t) = s(t; \underline{\phi}) + n(t) \quad \text{where } n(t) \text{ is an AWGN term}$$
$$\text{and } \underline{\phi} = (\tau_o, \theta_o, f_o, A_o, \{c_n\})$$

τ_o is the 'timing' error

θ_o is the 'carrier phase' error

f_o is the 'carrier frequency' error

A_o is the 'signal amplitude'

$\{c_n\}$ is the 'information symbol' sequence

Further on, we will assume a modulated digital PAM signal:

$$s(t; \underline{\phi}) = A_o \left[\sum_n c_n h(t - nT - \tau_o) \right] e^{j\theta_o(t)} \quad \text{where } \theta_o(t) = \theta_o + 2\pi f_o t$$

where all the conventional synchronization parameters are reflected:

MCRB FOR SYNCHRONIZATION

$$MCRB(\lambda) = \frac{1}{E_{r,\mathbf{u}} \left\{ \left(\frac{\partial \ln f(r | \mathbf{u}, \lambda)}{\partial \lambda} \right)^2 \right\}}$$

$$f(r | \mathbf{u}, \lambda) = \exp \left[-\frac{1}{2N_0} \int_{T_o} |r(t) - s(t)|^2 dt \right]$$

$$MCRB(\lambda) = \frac{N_0}{E_{\mathbf{u}} \left\{ \int_{T_o} \left| \frac{\partial s(t)}{\partial \lambda} \right|^2 dt \right\}}$$

A. Andrea, U. Mengali, "The Modified Cramer-Rao Bound and Its Application to Synchronization Problems", IEEE Trans. on Communications, vol. 42, No. 2/3/4, 1994.

MCRB FOR FREQUENCY ESTIMATION

$$\int_{T_0} \left| \frac{\partial s(t)}{\partial f_0} \right|^2 dt = 4\pi^2 \int_{T_0} t^2 \left| \sum_n c_n h(t - nT - \tau_0) \right|^2 dt$$

$$E_{c, \tau_0} \left\{ \left| \sum_n c_n h(t - nT - \tau_0) \right|^2 \right\} = E_s$$

$$E_{u_f} \left\{ \int_{T_0} \left| \frac{\partial s(t)}{\partial f_0} \right|^2 dt \right\} = \frac{2\pi^2 E_s T_0^3}{3T}$$

$$\boxed{MCRB(f_0) = \frac{3T}{2\pi^2 T_0^3} \frac{1}{E_s / N_0}}$$

MCRB FOR PHASE ESTIMATION

$$MCRB(\theta_0) = \frac{N_0}{E_{\mathbf{u}_\theta} \left\{ \int_{T_0} \left| \frac{\partial s(t)}{\partial \theta} \right|^2 dt \right\}}$$

$$\int_{T_0} \left| \frac{\partial s(t)}{\partial \theta_0} \right|^2 dt = \int_{T_0} \left| \sum_n c_n h(t - nT - \tau_0) \right|^2 dt$$

$$E_{\mathbf{u}_\theta} \left\{ \int_{T_0} \left| \frac{\partial s(t)}{\partial \theta_0} \right|^2 dt \right\} = 2E_s L \quad L = \frac{T_0}{T}$$

$$MCRB(\theta_0) = \frac{1}{2L} \frac{1}{E_s / N_0}$$

MCRB FOR TIMING ESTIMATION

$$MCRB(\tau_0) = \frac{1}{8\pi^2 \xi L} \frac{1}{E_s / N_0}$$

$$\xi = T^2 \frac{\int_{-\infty}^{\infty} f^2 |H(f)|^2 df}{\int_{-\infty}^{\infty} |H(f)|^2 df}$$

ξ is the normalized mean square bandwidth of $H(f)$

Timing estimation should be easier with wideband signals.

MAXIMUM LIKELIHOOD ESTIMATION (MLE)

Joint ML estimation/detection rule: $(\hat{c}, \hat{\theta}) = \arg \left[\max_{c, \theta} f(r | c, \theta) \right]$

- Data-aided (DA): using a preamble of known symbols.

$$\hat{\theta}(c_0) = \arg \left[\max_{\theta} f(r | c = c_0, \theta) \right]$$

- Decision-Directed (DD): using segments of decoded data as if it were the true symbols.

$$\hat{\theta}(c_L) = \arg \left[\max_{\theta} f(r | c = \hat{c}_L, \theta) \right]$$

- Non-Data-Aided (NDA): Performing the averaging operation to remove the data dependency.

$$f(r | \theta) = \sum_{\substack{\text{all possible} \\ \text{sequences } c}} f(r | c, \theta) P(c)$$

$$\hat{\theta} = \arg \left[\max_{\theta} f(r | \theta) \right]$$

MAXIMUM LIKELIHOOD ESTIMATION (II)

- Under the AWGN assumption, the ML optimal parameter estimation corresponds to the conventional Minimum Mean Square Error (M.M.S.E.), that is, the received data and the signal model fitting:

$$\min_{\phi} \mathcal{E}^2 = \min_{\phi} \int_{T_o} |r(t) - s(t; \phi)|^2 dt$$

- Thus:

$$\min_{\phi} \mathcal{E}^2 = \min_{\phi} \left[\int_{T_o} |r(t)|^2 dt + \int_{T_o} |s(t; \phi)|^2 dt - 2 \operatorname{Re} \left\{ \int_{T_o} s^*(t; \phi) r(t) dt \right\} \right]$$

?
(cross-correlator)
(constant envelope signals)

- And for ‘constant envelope signals’, it is necessary to perform the multi-dimensional search as follows:

$$\min_{\phi} \mathcal{E}^2 \equiv \max_{\phi} \operatorname{Re} \left\{ \int_{T_o} s^*(t; \phi) r(t) dt \right\}$$

DA ESTIMATION

- As we saw, the optimal parameters estimation becomes from maximizing:

$$\min_{\phi} \mathcal{E}^2 \equiv \max_{\phi} \operatorname{Re} \left\{ \int_{T_o} s^*(t; \phi) r(t) dt \right\}$$

- Thus, we have to maximize:

$$\max_{\theta_o} \operatorname{Re} \left\{ \underbrace{\sum_n c_n^* e^{-j\hat{\theta}_o} \int_{T_o} r(t) h^*(t - nT - \tau_o) dt}_{x(n)} \right\} = \max_{\theta_o} \operatorname{Re} \left\{ e^{-j\hat{\theta}_o} \sum_n c_n^* x(n) \right\}$$

$$\hat{\theta}_o = \arg \left\{ \sum_n c_n^* x(n) \right\}$$

DD ESTIMATION

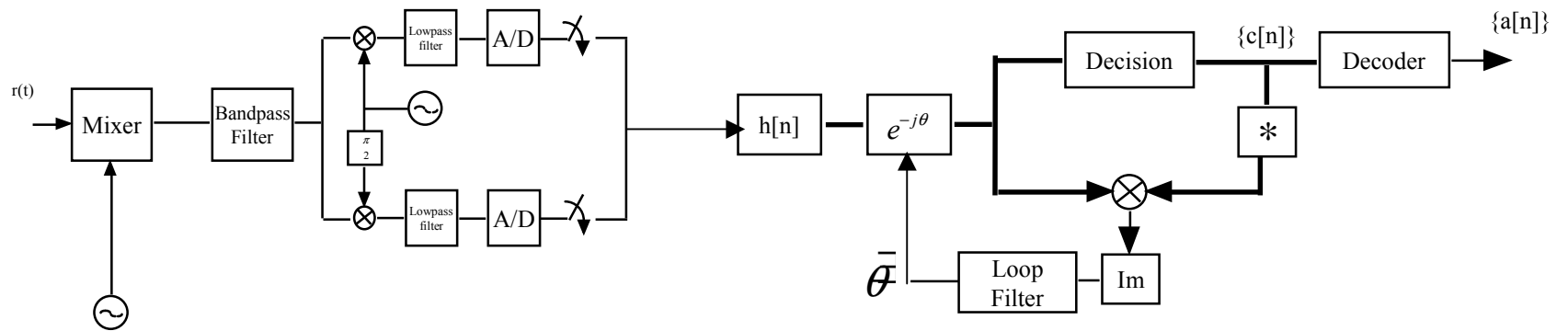
$$\max_{\theta_o} \mu^2 = \max_{\theta_o} \operatorname{Re} \left\{ \sum_n \hat{c}_n^* e^{-j\hat{\theta}_o} \int_{T_o} r(t) h^*(t - nT - \tau_o) dt \right\}$$

$$\frac{\partial}{\partial \hat{\theta}_o} \mu^2 = 0 \quad \text{where} \quad u_\theta \equiv \frac{\partial}{\partial \hat{\theta}_o} \mu^2 = \frac{\partial}{\partial \hat{\theta}_o} \operatorname{Re} \left\{ \sum_n \hat{c}_n^* e^{-j\hat{\theta}_o} \int_{T_o} r(t) h^*(t - nT - \tau_o) dt \right\}$$

ML Estimation:
$$u_\theta \equiv \operatorname{Im} \left\{ \sum_n \hat{c}_n^* e^{-j\hat{\theta}_o} \int_{T_o} r(t) h^*(t - nT - \tau_o) dt \right\}$$

Loop equations:
$$\hat{\theta}(n+1) = \hat{\theta}(n) + k_\theta u_\theta(n)$$

CLOSED LOOP SCHEME



DD Schemes: BPSK/QPSK

Sampled Matched Filter output:

$$I_k + jQ_k = \int_{T_o} r(t)h^*(t-nT)dt \Big|_{t=kT}$$

$$u_\theta \equiv \frac{\partial}{\partial \hat{\theta}_o} \mu^2 = \text{Im} \left\{ \hat{c}_n^* e^{-j\hat{\theta}_o} (I_k + jQ_k) \right\}$$

$$(I'_k + jQ'_k) = e^{-j\hat{\theta}_o} (I_k + jQ_k)$$

$$\hat{c}_n = \begin{cases} \text{BPSK:} & \text{sign}[I'_k] \\ \text{QPSK:} & \text{sign}[I'_k] + j\text{sign}[Q'_k] \end{cases}$$

BPSK

$$u_\theta = \text{Im} \left\{ \text{sign}[I'_k] (I'_k + jQ'_k) \right\}$$

$$u_\theta = \text{sign}[I'_k] Q'_k$$

QPSK

$$u_\theta = \text{Im} \left\{ (\text{sign}[I'_k] - j\text{sign}[Q'_k]) (I'_k + jQ'_k) \right\}$$

$$u_\theta = \text{sign}[I'_k] Q'_k - \text{sign}[Q'_k] I'_k$$

NDA Schemes: BPSK/QPSK

Sampled Matched Filter output:

$$I_k + jQ_k = \int_{T_o} r(t)h^*(t-nT)dt \Big|_{t=kT}$$

$$u_\theta \equiv \frac{\partial}{\partial \hat{\theta}_o} \mu^2 = \text{Im} \left\{ \hat{c}_n^* e^{-j\hat{\theta}_o} (I_k + jQ_k) \right\}$$

$$(I'_k + jQ'_k) = e^{-j\hat{\theta}_o} (I_k + jQ_k)$$

USE OF A NON-LINEARITY

$$(I''_k + jQ''_k) = [I'_k + jQ'_k]^M \text{ (for MPSK)}$$

$$c = e^{j2\pi l/M} \rightarrow c^M = 1$$

$$\max_{\theta_o} \text{Re} \left\{ \sum_n c_n^* e^{-j\hat{\theta}_o} \int_{T_o} r(t)h^*(t-nT-\tau_o)dt \right\} = \max_{\theta_o} \text{Re} \left\{ e^{-j\hat{\theta}_o} \sum_n c_n^* x(n) \right\}$$

$$\max_{\theta_o} \text{Re} \left\{ e^{-j\hat{\theta}_o M} \sum_n x(n)^M \right\} \rightarrow e^{-j\hat{\theta}_o M} = \exp \left[j \arg \sum_n x(n)^M \right]$$

Ambiguity: if $\hat{\theta}$ is a solution so is $\hat{\theta} + \frac{2\pi l}{M}$ for $l = 2, \dots, M-1$

$$e^{j\left(\hat{\theta} + \frac{2\pi l}{M}\right)M} = e^{j\hat{\theta}M}$$

BPSK

$$u_{\theta} = \text{Im}\{I_k'' + jQ_k''\}$$

$$u_{\theta} = Q_k''$$

- Phase ambiguity: π
- Frequency error: $2f_e$

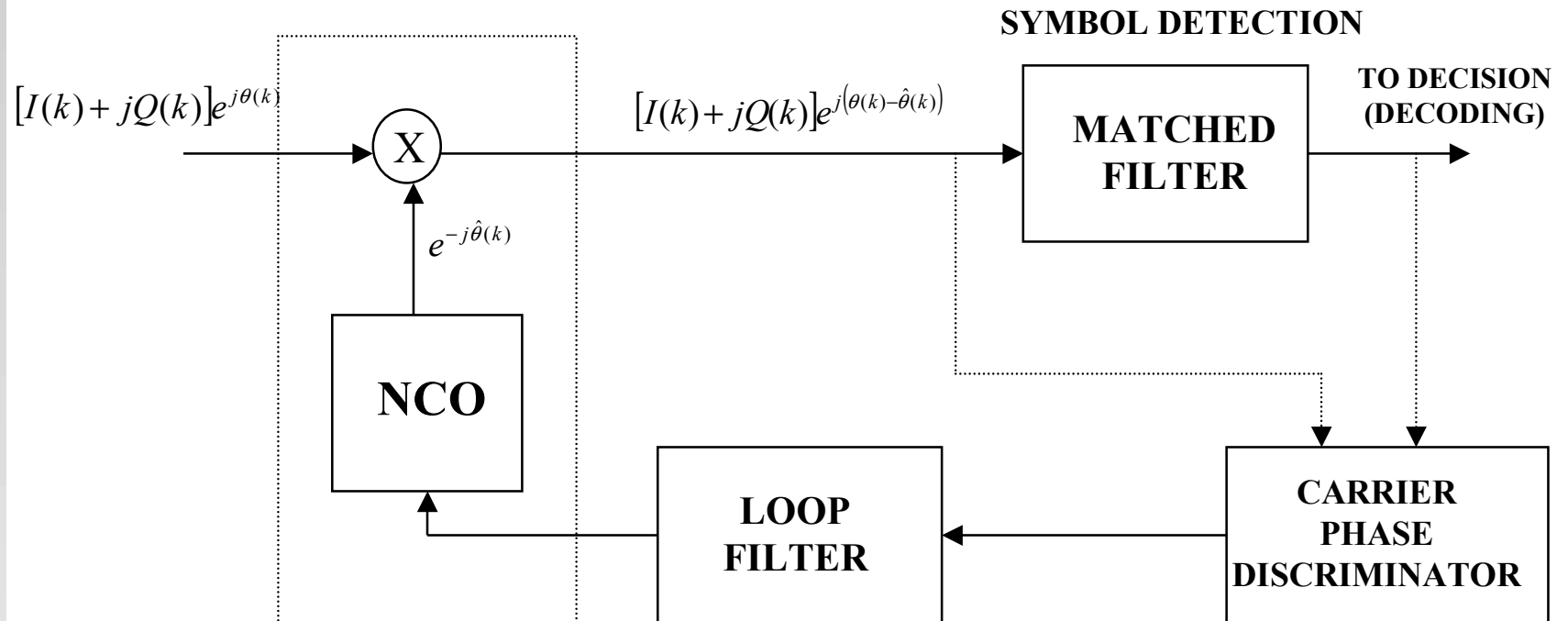
QPSK

$$u_{\theta} = \text{Im}\{I_k'' + jQ_k''\}$$

$$u_{\theta} = Q_k''$$

- Phase ambiguity: $\frac{\pi}{2}$
- Frequency error: $4f_e$

CARRIER ACQUISITION AND TRACKING



SIGNAL MODEL

- Let's consider an I&Q pure sinusoid:

$$r(t) = A_o e^{j(2\pi(f_o+f_e)t+\theta_o)} + n(t)$$

f_o is the carrier frequency

f_e is the frequency error (or doppler error)

θ_o is the carrier phase

- The In-phase and the Quadrature components of the sampled signal becomes:

$$I(k) = \frac{A_o}{2} \cos(\theta(k) - \hat{\theta}(k)) + \frac{1}{2} n_i(t)$$

$$Q(k) = \frac{A_o}{2} \sin(\theta(k) - \hat{\theta}(k)) + \frac{1}{2} n_q(t)$$

$\omega_o = 2\pi f_o T_s$ is the normalized carrier frequency

$\omega_e = 2\pi f_e T_s$ is the normalized frequency error

$\theta(k) = (\omega_o + \omega_e)k + \theta_o$ is the carrier phase evolution

$\hat{\theta}(k)$ is the carrier phase estimate

- The *phase error* is defined by:

$$\psi(k) = \theta(k) - \hat{\theta}(k)$$

- A typical *carrier phase discriminator* is (among others)

$$e(k) = g[\psi(k)] = \arctan\left(\frac{Q(k)}{I(k)}\right) \quad |e(k)| \leq \pi$$

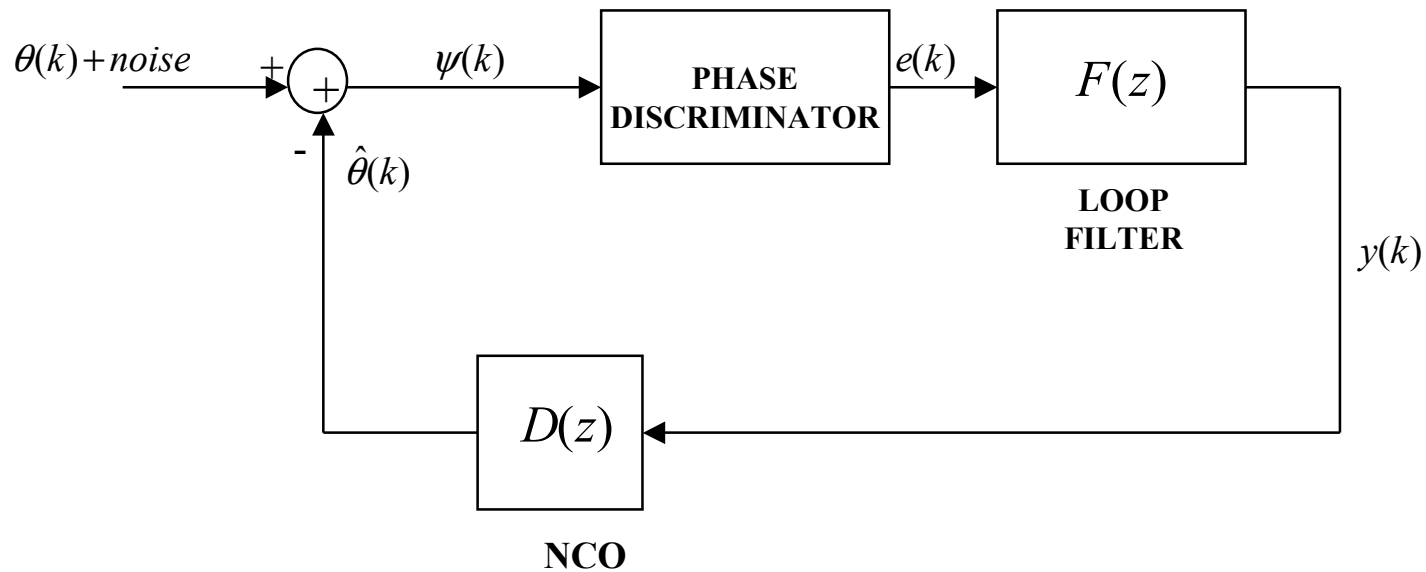
CARRIER DISCRIMINATORS (I)

- A typical *carrier phase discriminator* is:

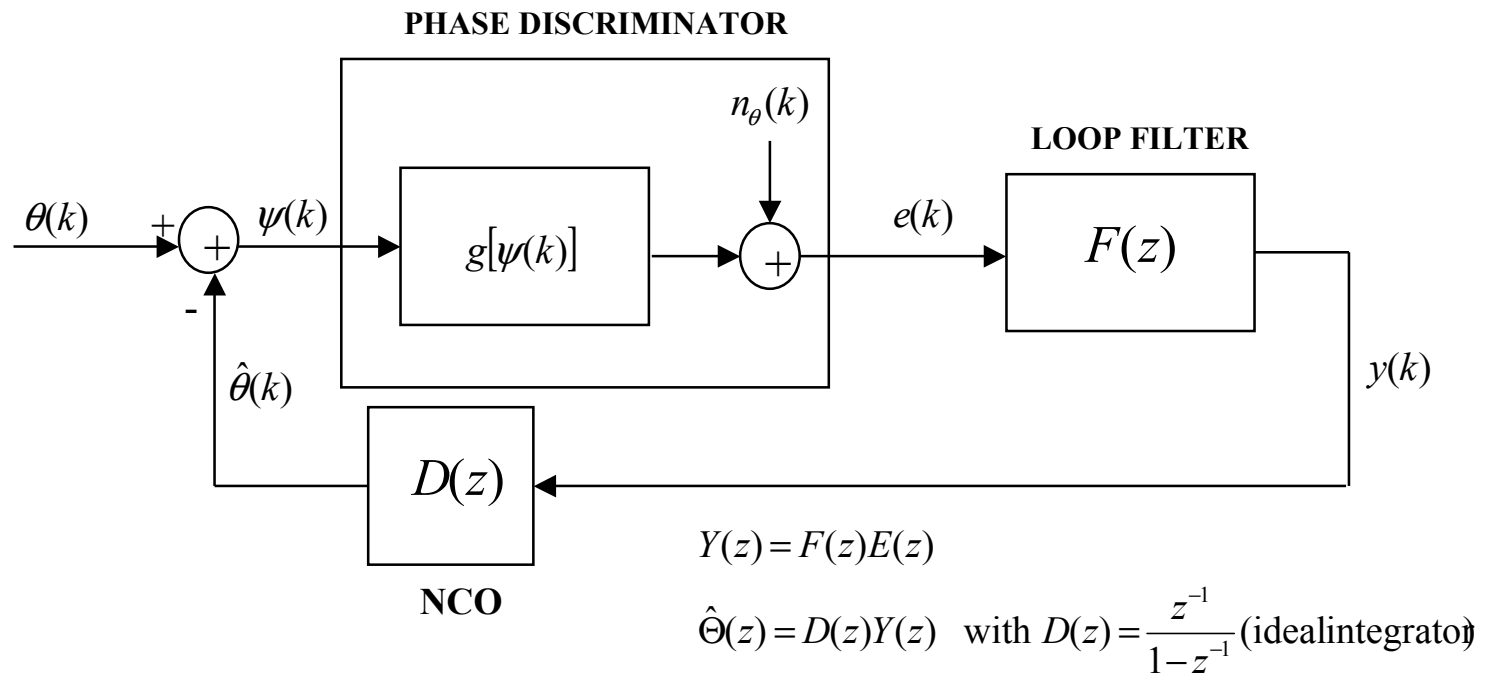
$$e(k) = g[\psi(k)] = \arctan\left(\frac{Q(k)}{I(k)}\right) \quad |e(k)| \leq \pi$$

or:

$$e(k) = g[\psi(k)] = \text{imag}[I(k) + jQ(k)] = Q(k)$$

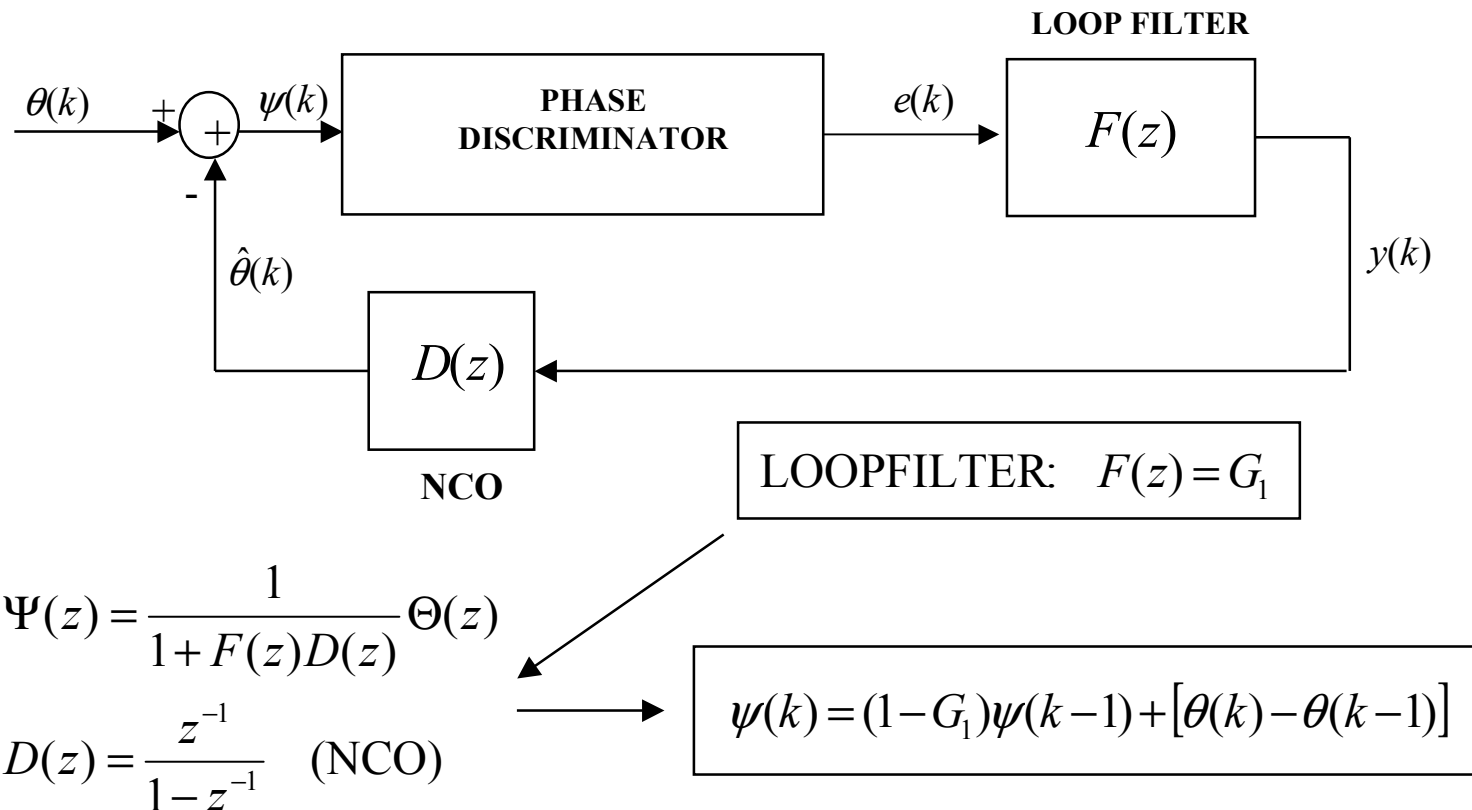


CARRIER DISCRIMINATORS (II)



- NOISEFREE: $e(k) = g[\psi(k)] = \psi(k) = \theta(k) - \hat{\theta}(k) \Rightarrow |e(k)| = |\psi(k)| \leq \pi$
- WITH NOISE: $e(k) = g[\psi(k)] + n_\theta(k) \quad |e(k)| \leq \pi$

FIRST ORDER LOOP FILTER (NOISE FREE)



FIRST ORDER: TRACKING ERROR

$$\psi(k) = (1 - G_1)\psi(k-1) + [\theta(k) - \theta(k-1)]$$

Filtrado/integración

Error/innovación

- We will consider a linear phase error evolution (carrier phase and frequency errors):

$$\theta(k) = \underbrace{a_1 k}_{\text{Frequency error}} + \underbrace{a_0}_{\text{Phase error}}$$

- The basic difference equation becomes:

$$\psi(k) = (1 - G_1)\psi(k-1) + a_1$$

which defines the system transient response for a given initial $\psi(0) = \psi_0$

- Under steady-state conditions:

$$\psi(k) = \psi(k-1) = \psi_{ss} \Rightarrow \psi_{ss} = (1 - G_1)\psi_{ss} + a_1 \Rightarrow \boxed{\psi_{ss} = \frac{a_1}{G_1}} \text{ TRACKING ERROR}$$

FIRST ORDER: LOCK-IN

$$\psi(k) = (1 - G_1)\psi(k-1) + a_1$$

- To ensure the system Lock-In, it is necessary to avoid any phase aliasing (wrapping) in the loop:

$$|\psi(k)| \leq \pi \quad \forall k$$

- The worst cases are the following: $\rightarrow a_1 > 0$ (positive frequency error) with $\psi(k-1) = \pm\pi$

$$\psi(k) = a_1 + (1 - G_1)\pi < \pi \quad \Rightarrow \quad G_1 > \frac{a_1}{\pi}$$

$$\psi(k) = a_1 - (1 - G_1)\pi > -\pi \quad \Rightarrow \quad G_1 < 2 - \frac{a_1}{\pi}$$

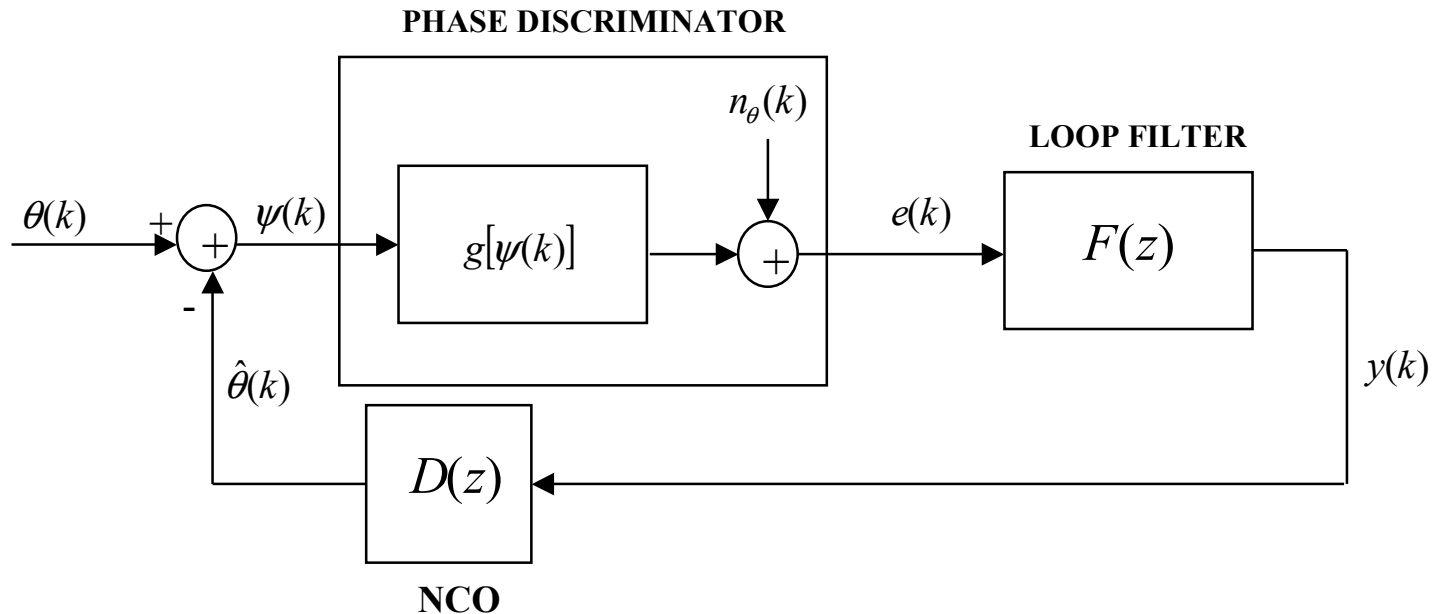
$\rightarrow a_1 < 0$ (negative frequency error) with $\psi(k-1) = \pm\pi$

$$G_1 > \frac{-a_1}{\pi}$$

$$G_1 < 2 + \frac{a_1}{\pi}$$

$$\text{LOCK-IN: } \frac{|a_1|}{\pi} < G_1 < 2 - \frac{|a_1|}{\pi} \Rightarrow |a_1| < \pi \quad (\text{Nyquist}) \quad |1 - G_1| < 1$$

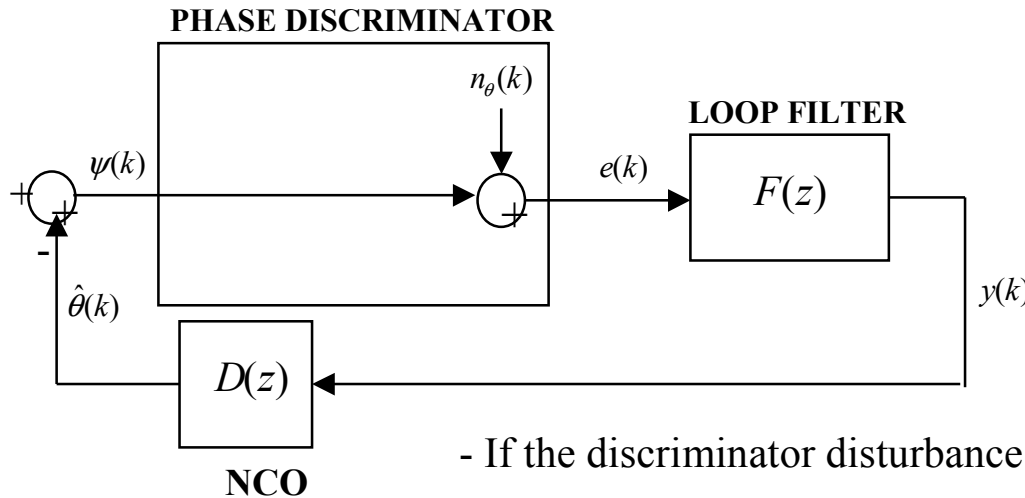
FIRST ORDER LOOP FILTER (WITH NOISE)



$$\left. \begin{aligned}
 \psi(k) &= \theta(k) - \hat{\theta}(k) \\
 e(k) &= g[\psi(k)] + n_{\theta}(k) \\
 y(k) &= G_1 e(k) \\
 \hat{\theta}(k) &= \hat{\theta}(k-1) + y(k-1)
 \end{aligned} \right\} \Rightarrow \boxed{\psi(k) = \psi(k-1) - G_1 g[\psi(k-1)] + a_1 - G_1 n_{\theta}(k-1)}$$

FIRST ORDER MARKOV PROCESS

PRAGMATIC APPROACH



$$\Psi(z) = \frac{D(z)F(z)}{1 + D(z)F(z)} N_{\theta}(z)$$

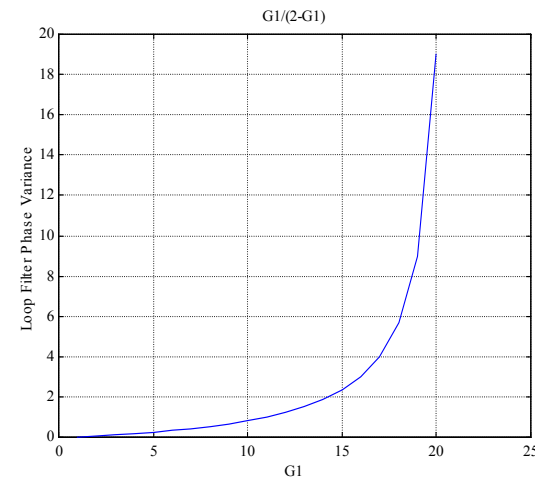
- If the discriminator disturbance noise is white: $S_{n_{\theta}n_{\theta}}(f) = \frac{\sigma_{n_{\theta}}^2}{f_s}$

$$\rightarrow E[\psi_{ss}] \approx \frac{a_1}{G_1}$$

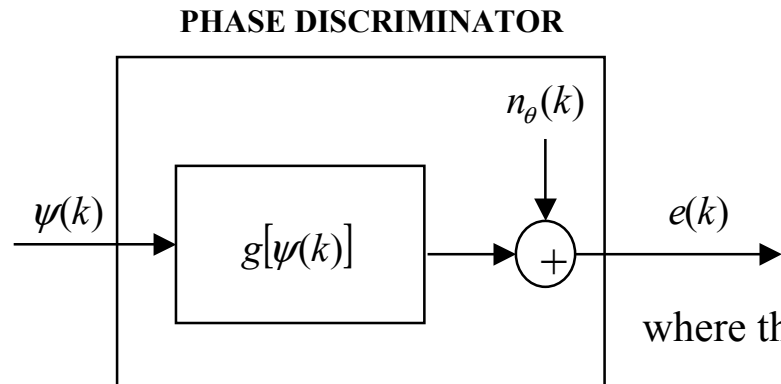
$$\rightarrow E[\psi_{ss}^2] \approx \frac{T_s}{2\pi} \int_{-\pi}^{+\pi} \left| \frac{D(z)F(z)}{1 + D(z)F(z)} \right|_{z=e^{j2\pi f T_s}}^2 \frac{\sigma_{n_{\theta}}^2}{f_s} df \quad \text{with } T_s = \frac{1}{f_s}$$

$$\text{var}[\psi_{ss}] = E[\psi_{ss}^2] - E^2[\psi_{ss}]$$

$$\text{var}[\psi_{ss}] \approx \frac{G_1}{2 - G_1} \sigma_{n_{\theta}}^2$$



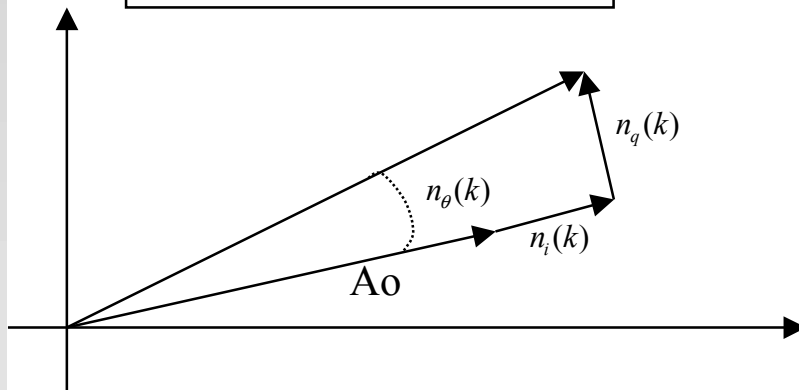
PHASE DISTURBANCE MEAN POWER



$$n_{\theta}(k) = \arctan \left[\frac{n_q(k)}{A_0 + n_i(k)} \right] \text{ for } k = 1, 2, \dots, N$$

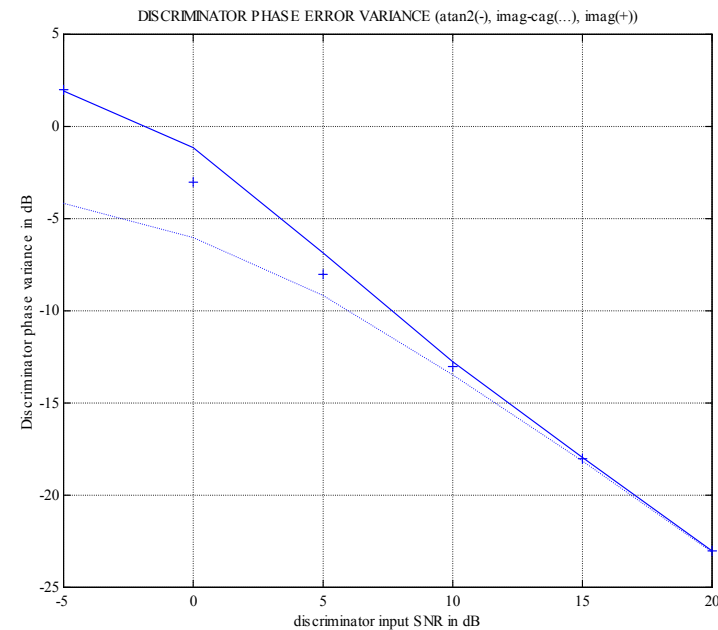
$$\hat{\sigma}_{n_{\theta}}^2 = \frac{1}{N} \sum_{k=1}^N |n_{\theta}(k)|^2$$

where the noise complex samples are zero mean and gaussian

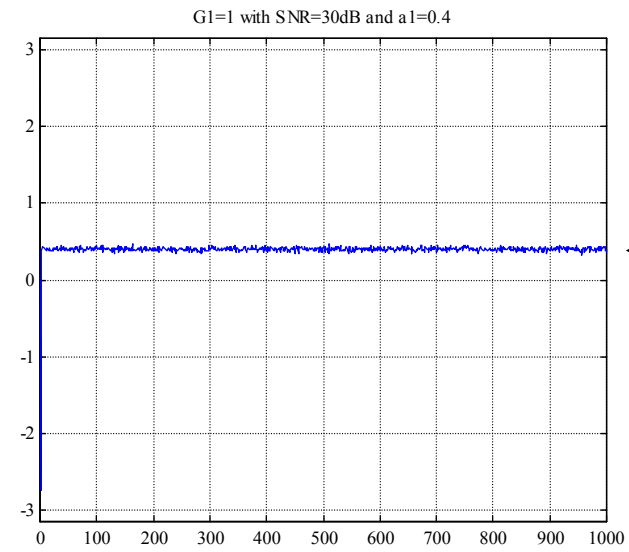
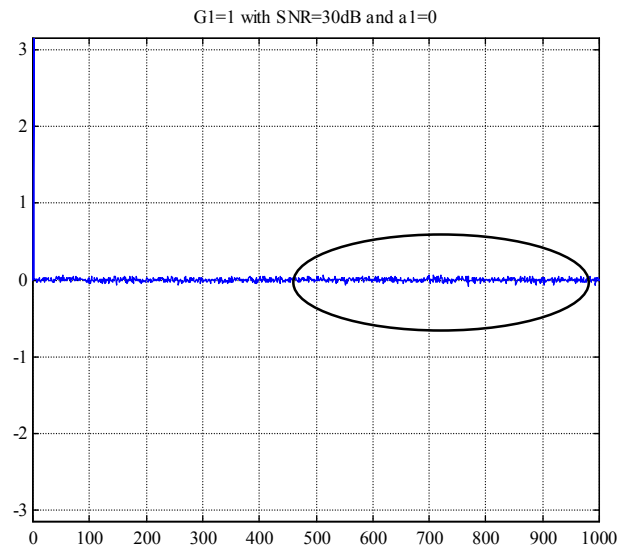
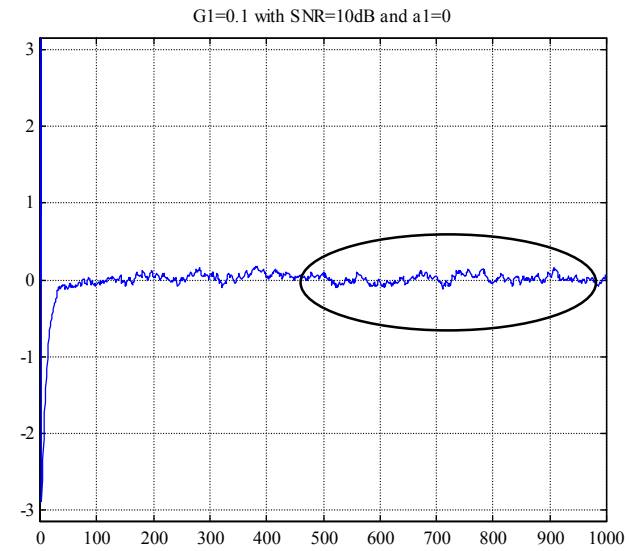
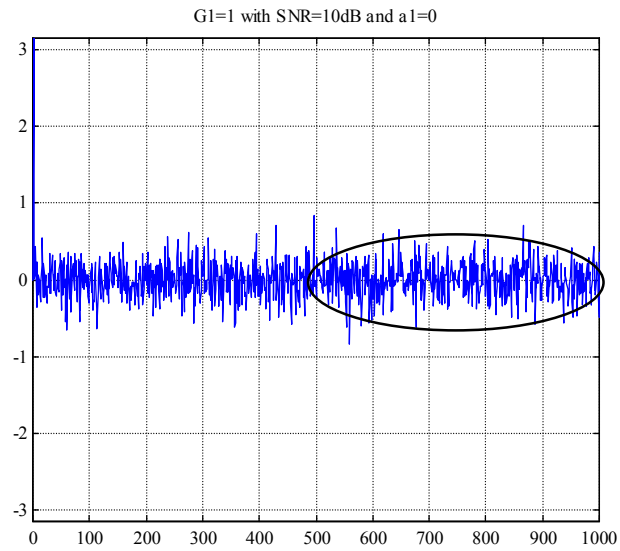


The discriminator noise disturbance has to be generated at the working SNR, that is:

$$SNR = \frac{S}{N} = \frac{1}{E[n_i^2] + E[n_q^2]} = \frac{1}{2E[n_i^2]} = \frac{1}{2E[n_q^2]}$$



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SECOND ORDER LOOP FILTER (NOISE FREE)

LOOPFILTER: $F(z) = G_1 + \frac{G_2}{1-z^{-1}} \quad r = G_1 + \frac{G_2}{G_1}$

$$\Psi(z) = \frac{1}{1 + F(z)D(z)} \Theta(z)$$

$$D(z) = \frac{z^{-1}}{1-z^{-1}} \quad (\text{NCO})$$

$$\psi(k) = 2\psi(k-1) - \psi(k-2) + [\theta(k) - 2\theta(k-1) + \theta(k-2)] - (G_1 + G_2)\psi(k-1) + G_1\psi(k-2)$$

- We will consider a *quadratic* phase error evolution:

$$\theta(k) = a_2 k^2 + a_1 k + a_0 \rightarrow \text{Phase error}$$

Frequency Rate error \rightarrow Frequency error

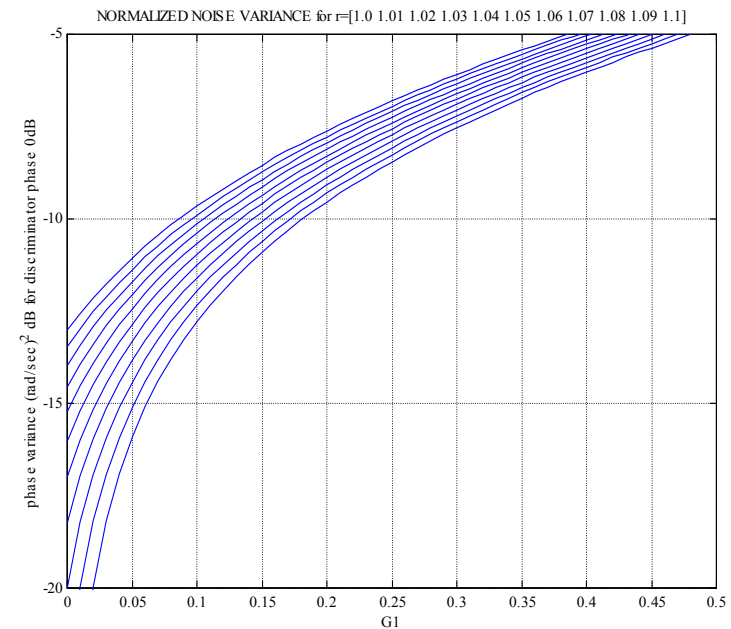
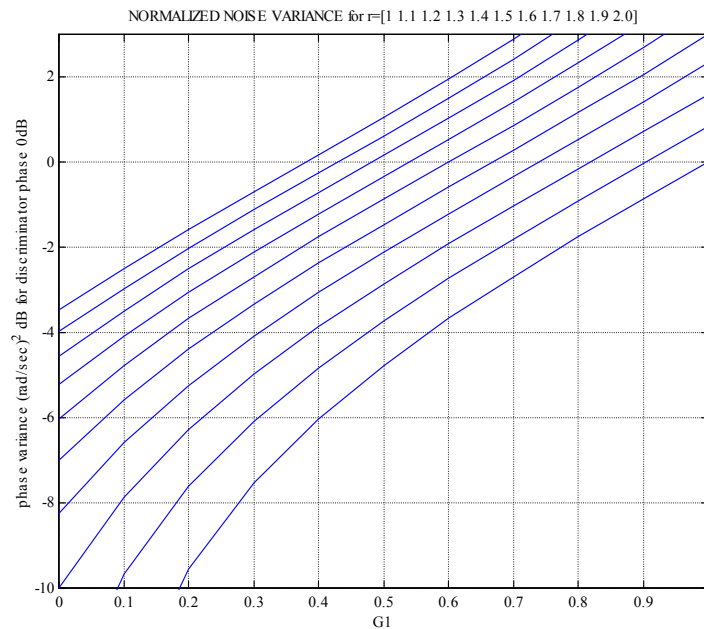
TRACKING ERROR: $\psi_{ss} = \frac{2a_2}{G_2} \quad \forall a_1!!!!$

LOCK-IN: $\left\{ \begin{array}{l} 2 + \frac{2a_2}{\pi} < G_1(r+1) < 4 - \frac{2a_2}{\pi} \\ \frac{2a_2}{\pi} < G_1(r-1) < 2 - \frac{2a_2}{\pi} \end{array} \right\} \left\{ \begin{array}{l} |a_2| < \frac{\pi}{2} \\ r > 1 \end{array} \right.$

SECOND ORDER LOOP FILTER (WITH NOISE)

$$\left. \begin{aligned} \rightarrow E[\psi_{ss}] &\approx \frac{2a_2}{G_2} \\ \rightarrow E[\psi_{ss}^2] &\approx \frac{T_s}{2\pi} \int_{-\pi}^{+\pi} \left| \frac{D(z)F(z)}{1+D(z)F(z)} \right|_{z=e^{j2\pi f T_s}}^2 \frac{\sigma_{n_\theta}^2}{f_s} df \quad \text{with } T_s = \frac{1}{f_s} \end{aligned} \right\} \text{var}[\psi_{ss}] = E[\psi_{ss}^2] - E^2[\psi_{ss}]$$

$$\text{var}[\psi_{ss}] \approx \frac{2(r-1) + G_1(r+1)}{4 - G_1(r+1)} \sigma_{n_\theta}^2$$



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