

$$\underset{\uparrow A}{1} \cos(\underset{\uparrow \phi}{2\pi f_0 n} + \underset{\uparrow A}{0}) = -1 \cos(\underset{\uparrow A}{2\pi f_0 n} + \underset{\uparrow \phi}{\pi})$$

WILL NEED MLE OF A, ϕ . CAN BE SHOWN TO BE (FOR LARGE N AND f_0 NOT NEAR 0 OR $1/2$)

$$\hat{A} = \sqrt{\hat{\alpha}_1^2 + \hat{\alpha}_2^2}$$

$$\hat{\phi} = \text{ARCTAN}\left(-\frac{\hat{\alpha}_2}{\hat{\alpha}_1}\right)$$

$$\hat{\alpha}_1 = \frac{2}{N} \sum_{n=0}^{N-1} x(n) \cos 2\pi f_0 n$$

$$\hat{\alpha}_2 = \frac{2}{N} \sum_{n=0}^{N-1} x(n) \sin 2\pi f_0 n$$

$$\text{OR } \hat{A} e^{j\hat{\phi}} = \frac{2}{N} \sum_{n=0}^{N-1} x(n) e^{-j2\pi f_0 n}$$

$$= \hat{\alpha}_1 - j\hat{\alpha}_2$$

GLRT DECIDES H_1 IF

$$L_G(x) = \frac{p(x; \hat{A}, \hat{\phi}, H_1)}{p(x; H_0)} > \gamma$$

$$L_G(x) = \frac{1}{(2\pi\sigma^2)^{N/2}} e^{-\frac{1}{2\sigma^2} \sum_{n=0}^{N-1} [x(n) - \hat{A} \cos(2\pi f_0 n + \hat{\phi})]^2}$$

$$\frac{1}{(2\pi\sigma^2)^{N/2}} e^{-\frac{1}{2\sigma^2} \sum_{n=0}^{N-1} x^2(n)}$$

$$LN L_G(x) = -\frac{1}{2\sigma^2} \left[\sum_n -2x(n) \hat{A} \cos(2\pi f_0 n + \hat{\phi}) + \sum_n \hat{A}^2 \cos^2(2\pi f_0 n + \hat{\phi}) \right]$$

BUT $\sum_{n=0}^{N-1} \cos^2(2\pi f_0 n + \hat{\phi}) \approx N/2$

$$\begin{aligned} \sum_{n=0}^{N-1} x(n) \hat{A} \cos(2\pi f_0 n + \hat{\phi}) &= \\ \underbrace{\sum_n x(n) \cos 2\pi f_0 n}_{\frac{N}{2} \hat{\alpha}_1} \underbrace{\hat{A} \cos \hat{\phi}}_{\hat{\alpha}_1} &- \\ \underbrace{\sum_n x(n) \sin 2\pi f_0 n}_{\frac{N}{2} \hat{\alpha}_2} \underbrace{\hat{A} \sin \hat{\phi}}_{-\hat{\alpha}_2} & \\ &= N/2 (\hat{\alpha}_1^2 + \hat{\alpha}_2^2) \end{aligned}$$

$$\Rightarrow LN L_G(x) \approx -\frac{1}{2\sigma^2} \left[-2(N/2) (\hat{\alpha}_1^2 + \hat{\alpha}_2^2) + N/2 \hat{A}^2 \right]$$

$\uparrow \hat{\alpha}_1^2 + \hat{\alpha}_2^2$

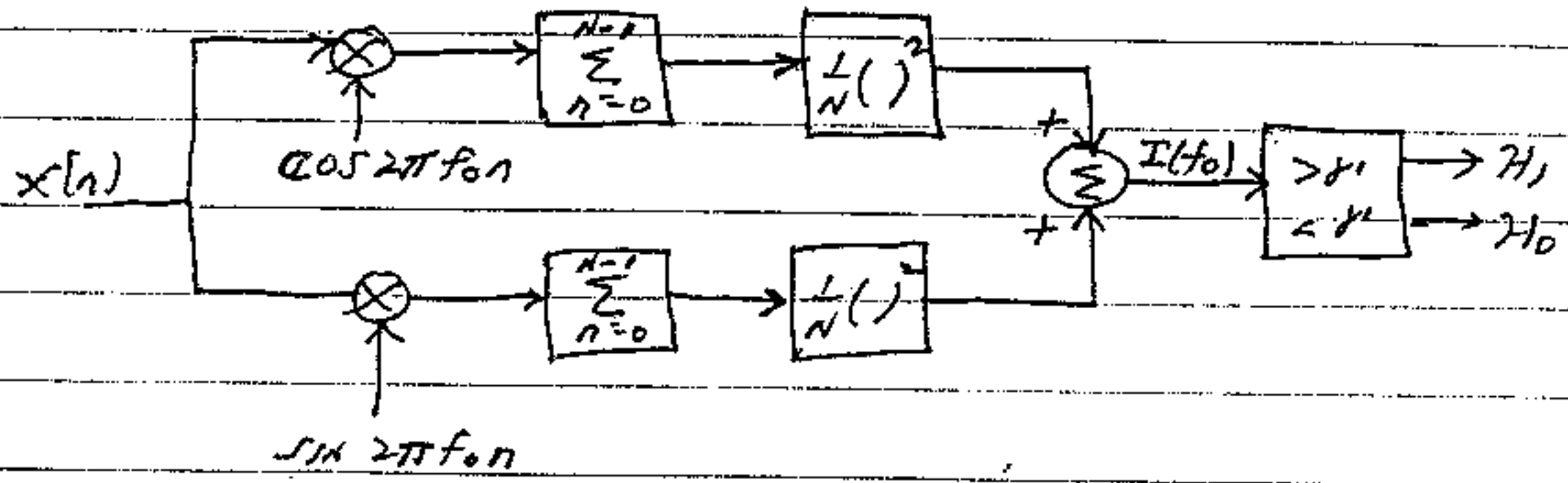
$$= -\frac{1}{2\sigma^2} \left[-N/2 (\hat{\alpha}_1^2 + \hat{\alpha}_2^2) \right]$$

$$= \frac{N}{4\sigma^2} (\hat{\alpha}_1^2 + \hat{\alpha}_2^2) = \frac{N}{4\sigma^2} \hat{A}^2$$

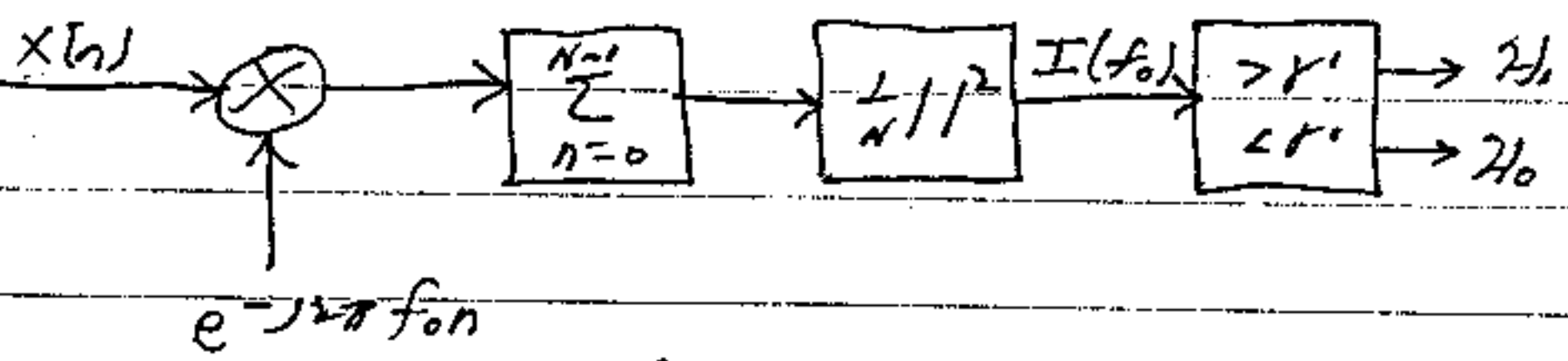
$$= \frac{N}{4\sigma^2} \left| \frac{2}{N} \sum_{n=0}^{N-1} x(n) e^{-j2\pi f_0 n} \right|^2$$

$$= \frac{1}{\sigma^2} I(f_0)$$

\uparrow PERIODOGRAM!



QUADRATURE MATCHED FILTER



PERIODOGRAM

b) UNKNOWN AMPLITUDE AND PHASE

PERFORMANCE WILL BE DIFFERENT
THAN FOR RAYLEIGH FADING

HERE: A, ϕ DETERMINISTIC AND UNKNOWN
FADING: $A \sim$ RAYLEIGH
 $\phi \sim$ UNIFORM

TO FIND PERFORMANCE:

$$I(f_0) = Z_1^2 + Z_2^2$$

$$Z_1 = \frac{1}{\sqrt{N}} \sum_{n=0}^{N-1} x(n) \cos 2\pi f_0 n$$

$$Z_2 = \frac{1}{\sqrt{N}} \sum_{n=0}^{N-1} x(n) \sin 2\pi f_0 n$$

OR $\underline{Z} = \underline{H}^T \underline{x}$ $\underline{H} = \begin{bmatrix} \frac{1}{\sqrt{N}} & 0 \\ \frac{1}{\sqrt{N}} \cos 2\pi f_0 & \frac{1}{\sqrt{N}} \sin 2\pi f_0 \\ \vdots & \vdots \end{bmatrix}$

$\underline{Z} \sim$ GAUSSIAN

SINCE \underline{x} IS GAUSSIAN

$$\underline{H}^T \underline{H} \approx \frac{1}{2} \underline{I}$$

UNDER H_0 :

$$\underline{x} \sim N(\underline{0}, \sigma^2 \underline{I})$$

$$\Rightarrow \underline{Z} \sim N(\underline{0}, \underbrace{\underline{H}^T \sigma^2 \underline{I} \underline{H}}_{\frac{\sigma^2}{2} \underline{I}})$$

UNDER \mathcal{H}_1 :

$$\underline{z} \sim N(\underline{H}^T \underline{E}(x), \frac{\sigma^2}{2} \underline{I})$$

$$\underline{E}(x) = \begin{bmatrix} A \cos \phi \\ A \cos(2\pi f_0 + \phi) \\ \vdots \\ A \cos(2\pi f_0(N-1) + \phi) \end{bmatrix}$$

$$\begin{aligned} \Rightarrow E(z_1) &= \frac{A}{\sqrt{N}} \sum_{n=0}^{N-1} \cos 2\pi f_0 n \cos(2\pi f_0 n + \phi) \\ &= \frac{A}{2\sqrt{N}} \sum_n [\cos \phi + \cos(4\pi f_0 n + \phi)] \\ &\approx \frac{\sqrt{N} A}{2} \cos \phi \end{aligned}$$

SIMILARLY $E(z_2) \approx -\frac{\sqrt{N} A}{2} \sin \phi$

$$\underline{z} \sim N \left(\begin{bmatrix} \frac{\sqrt{N} A}{2} \cos \phi \\ -\frac{\sqrt{N} A}{2} \sin \phi \end{bmatrix}, \frac{\sigma^2}{2} \underline{I} \right)$$

NOTE z_1, z_2 ARE INDEPENDENT UNDER \mathcal{H}_0 AND \mathcal{H}_1 .

DECIDE \mathcal{H}_1 IF $I(f_0) \equiv z_1^2 + z_2^2 > \gamma$

$$\text{UNDER } \mathcal{H}_0 \quad \frac{z_1^2 + z_2^2}{\sigma^2/2} = \left(\frac{z_1}{\sqrt{\sigma^2/2}} \right)^2 + \left(\frac{z_2}{\sqrt{\sigma^2/2}} \right)^2 \sim N(0,1) + N(0,1)$$

$$\sim \chi^2_2$$

$$\begin{aligned} P_{FA} &= P_r \{ I(f_0) > \gamma' \cdot H_0 \} \\ &= P_r \left\{ \frac{I(f_0)}{\sigma^2/2} > \frac{\gamma'}{\sigma^2/2} \mid H_0 \right\} \\ &= \int_{\frac{\gamma'}{\sigma^2/2}}^{\infty} \frac{1}{2} e^{-\frac{1}{2}x} dx \\ &= e^{-\gamma'/\sigma^2} \end{aligned}$$

UNDER H_1

$$\frac{z_1^2 + z_2^2}{\sigma^2/2} = \left(\frac{z_1}{\sqrt{\sigma^2/2}} \right)^2 + \left(\frac{z_2}{\sqrt{\sigma^2/2}} \right)^2$$

\uparrow $N\left(\frac{\mu_1}{\sqrt{\sigma^2/2}}, 1\right)$

$$\sim \chi^2_2(\lambda)$$

$$\text{WHERE } \lambda = \left(\frac{\mu_1}{\sqrt{\sigma^2/2}} \right)^2 + \left(\frac{\mu_2}{\sqrt{\sigma^2/2}} \right)^2$$

$$= \frac{\frac{NA^2}{4} \cos^2 \phi}{\sigma^2/2} + \frac{\frac{NA^2}{4} \sin^2 \phi}{\sigma^2/2}$$

$$= \frac{NA^2}{2\sigma^2}$$

$$\begin{aligned}
 \Rightarrow P_D &= P_r \left\{ I(f_0) > \nu' ; \mathcal{H}_1 \right\} \\
 &= P_r \left\{ \frac{I(f_0)}{\sigma^2/2} > \frac{\nu'}{\sigma^2/2} ; \mathcal{H}_1 \right\} \\
 &\quad \chi^2_2(\lambda) \\
 &= Q_{\chi^2_2(\lambda)} \left(2\nu'/\sigma^2 \right)
 \end{aligned}$$

BUT FROM CHAPTER 2

$$P_D = \sum_{k=0}^{\infty} \frac{e^{-\lambda/2} (\lambda/2)^k}{k!} Q_{\chi^2_{2+2k}} \left(\frac{2\nu'}{\sigma^2} \right)$$

OR USE CHIPR2.M

$$P = \text{CHIPR2} \left(2, \lambda, \frac{2\nu'}{\sigma^2}, \epsilon \right)$$

\uparrow $\frac{NA^2}{2\sigma^2}$ \uparrow MAX. ERROR ALLOWABLE

SEE FIG. 7.6 b FOR P_D CURVES.
 (COMPARE TO FIG. 7.6 a FOR
 ϕ KNOWN).

3) A, ϕ, f_0 UNKNOWN.

GLRT DECIDES H_1 IF

$$\frac{p(\underline{x}; \hat{A}, \hat{\Phi}, \hat{f}_0, H_1)}{p(\underline{x}; H_0)} > \gamma$$

RECALL THAT FOR θ UNKNOWN

MLE OF θ IS VALUE THAT MAXIMIZES

$$p(\underline{x}; \theta) \text{ OR } p(\underline{x}; \hat{\theta}) > p(\underline{x}; \theta)$$

FOR ALL $\theta \neq \hat{\theta}$. THUS

$$p(\underline{x}; \hat{\theta}) = \max_{\theta} p(\underline{x}; \theta)$$

$$p(\underline{x}; \hat{A}, \hat{\Phi}, \hat{f}_0, H_1) = \max_{f_0} p(\underline{x}; \hat{A}, \hat{\Phi}, f_0, H_1)$$

GLRT DECIDES H_1 IF

$$\frac{\max_{f_0} p(\underline{x}; \hat{A}, \hat{\Phi}, f_0, H_1)}{p(\underline{x}; H_0)} > \gamma$$

$$\max_{f_0} \left[\frac{p(\underline{x}; \hat{A}, \hat{\Phi}, f_0, H_1)}{p(\underline{x}; H_0)} \right] > \gamma$$

$$\downarrow$$

$$\text{LN MAX}_{f_0} \left[\quad \right] > \text{LN } \gamma$$

$$\text{BUT } \text{LN } \max_x g(x) = \max_x \text{LN } g(x)$$

SINCE LN IS MONOTONICALLY
INCREASING

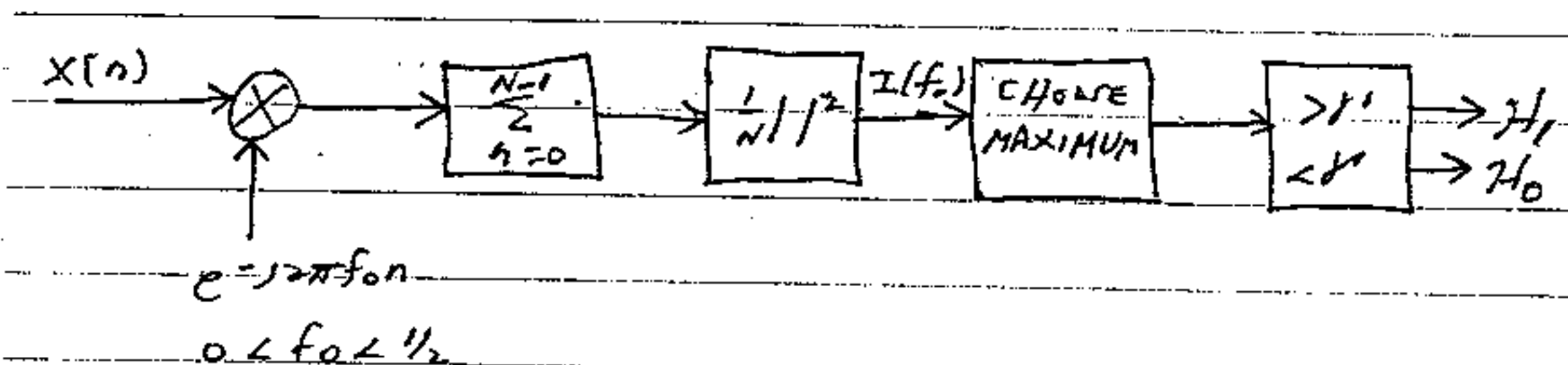
$$\Rightarrow \max_{f_0} \text{LN } \frac{p(x; \hat{A}, \hat{\phi}, f_0, H_1)}{p(x; H_0)} > \text{LN } \gamma$$

$I(f_0)/\sigma^2$ A, ϕ UNKNOWN CASE

WE DECIDE H_1 IF

$$\max_{f_0} I(f_0) > \sigma^2 \text{LN } \gamma = \gamma'$$

CONVENTIONAL NARROWBAND OR
FFT DETECTOR



c) UNKNOWN AMPLITUDE, PHASE, FREQUENCY

WILL DISCUSS PERFORMANCE LATER.

4) A, ϕ, f_0, n_0 UNKNOWN

GLRT DECIDES \mathcal{H}_1 IF

$$L_G(x) = \frac{p(x; \hat{A}, \hat{\phi}, \hat{f}_0, \hat{n}_0; \mathcal{H}_1)}{p(x; \mathcal{H}_0)} > \gamma$$

OR

$$L_G(x) = \max_{n_0} \frac{p(x; \hat{A}, \hat{\phi}, \hat{f}_0, n_0; \mathcal{H}_1)}{p(x; \mathcal{H}_0)}$$

EXCEPT FOR MAXIMIZATION OVER n_0

WE HAVE SAME GLRT AS FOR

A, ϕ, f_0 UNKNOWN. NEED ONLY

MODIFY OBSERVATION INTERVAL

TO $[n_0, n_0 + M - 1]$ (BEFORE $n_0 = 0$, $M = N$)

$$\Rightarrow I_{n_0}(f_0) = \frac{1}{M} \left| \sum_{n=n_0}^{n_0+M-1} x[n] e^{-j2\pi f_0 n} \right|^2$$

$$\ln \frac{p(x; \hat{A}, \hat{\phi}, \hat{f}_0, n_0; \mathcal{H}_1)}{p(x; \mathcal{H}_0)} = \frac{I_{n_0}(\hat{f}_0)}{\sigma^2}$$

FROM UNKNOWN A, ϕ, f_0 CASE.

$$\begin{aligned}
 LG(\underline{x}) &= \max_{n_0} \frac{I_{n_0}(\hat{f}_0)}{\sigma^2} \\
 &= \max_{n_0, f_0} \frac{I_{n_0}(f_0)}{\sigma^2}
 \end{aligned}$$

WHERE

$$I_{n_0}(f_0) = \frac{1}{M} \left| \sum_{n=n_0}^{n_0+M-1} x[n] e^{-j2\pi f_0 n} \right|^2$$

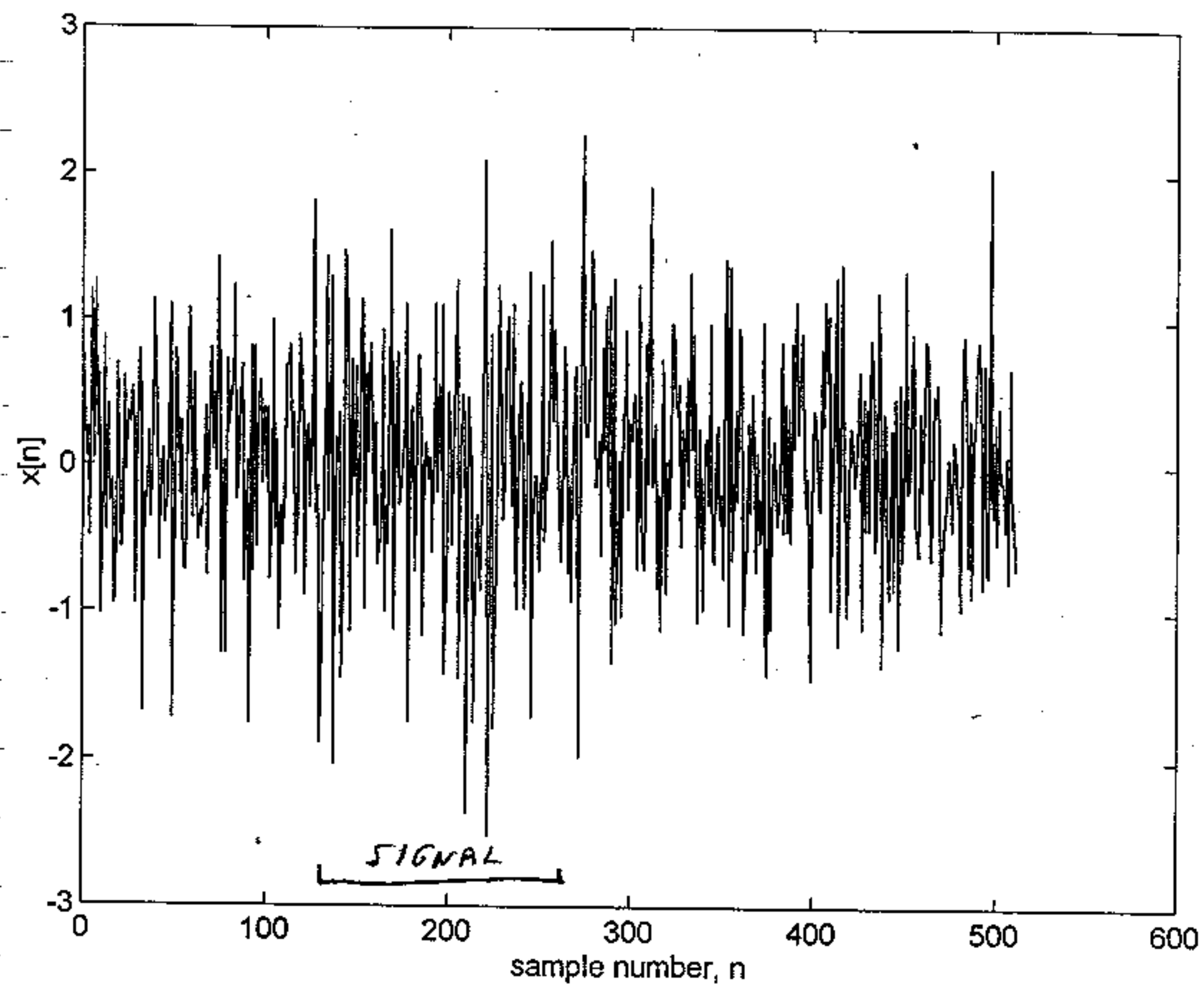
THIS IS A "RUNNING" FOURIER TRANSFORM.
CALLED SHORT-TIME PERIODOGRAM OR
SPECTROGRAM.

GLRT PICKS PEAK OF $I_{n_0}(f_0)$
AND COMPARES TO THRESHOLD.
IF EXCEEDED \Rightarrow MLE OF DELAY
AND FREQUENCY

EXAMPLE : $A=1, f_0=0.25, \phi=0, M=128,$
 $n_0=128, N=512, \sigma^2=0.5$

$$SNR = A^2 / 2\sigma^2 = 1 \text{ OR } 0 \text{ dB}$$

SIGNAL AT [128, 255]



7.7(a) TIME SERIES DATA

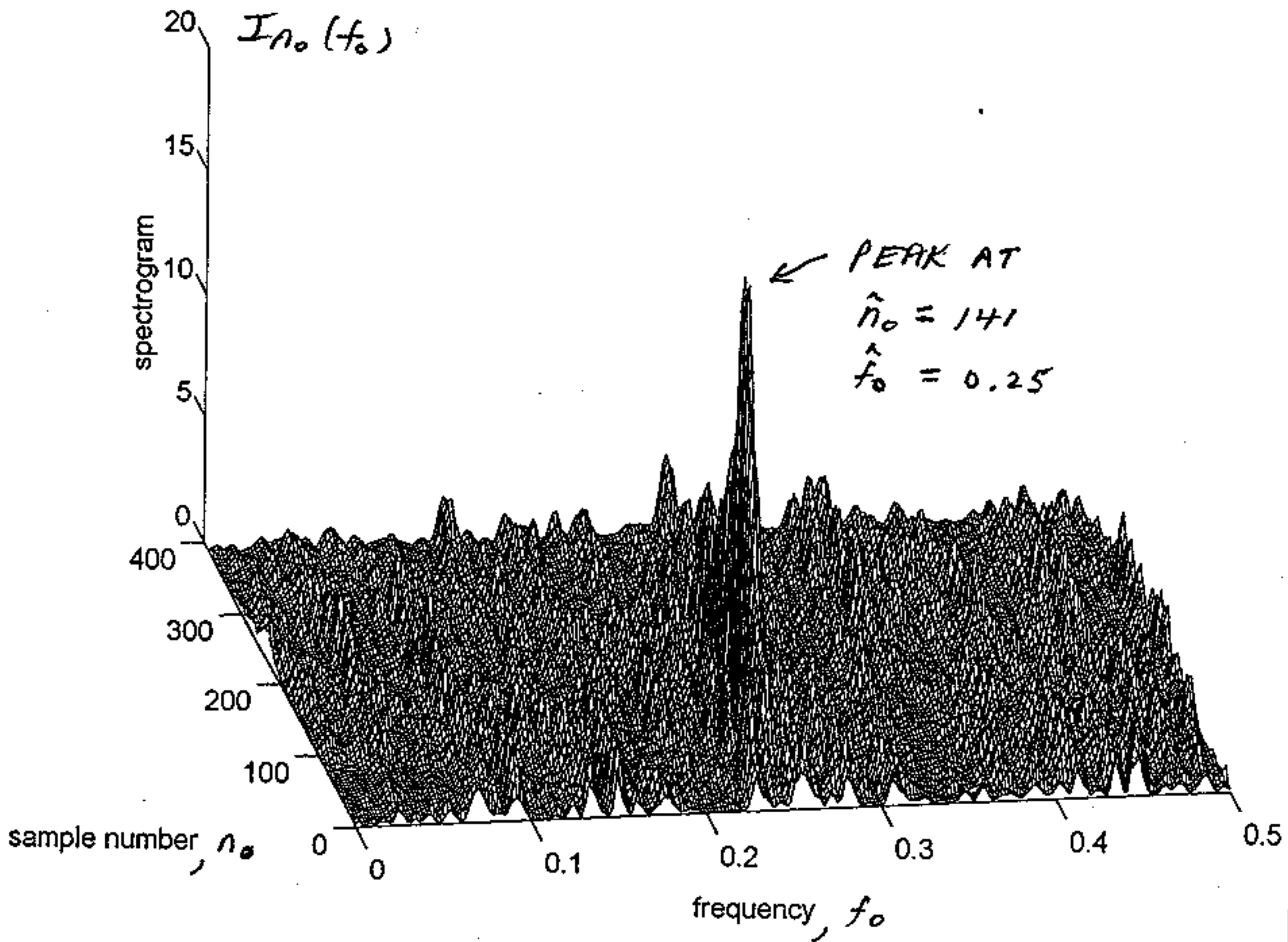
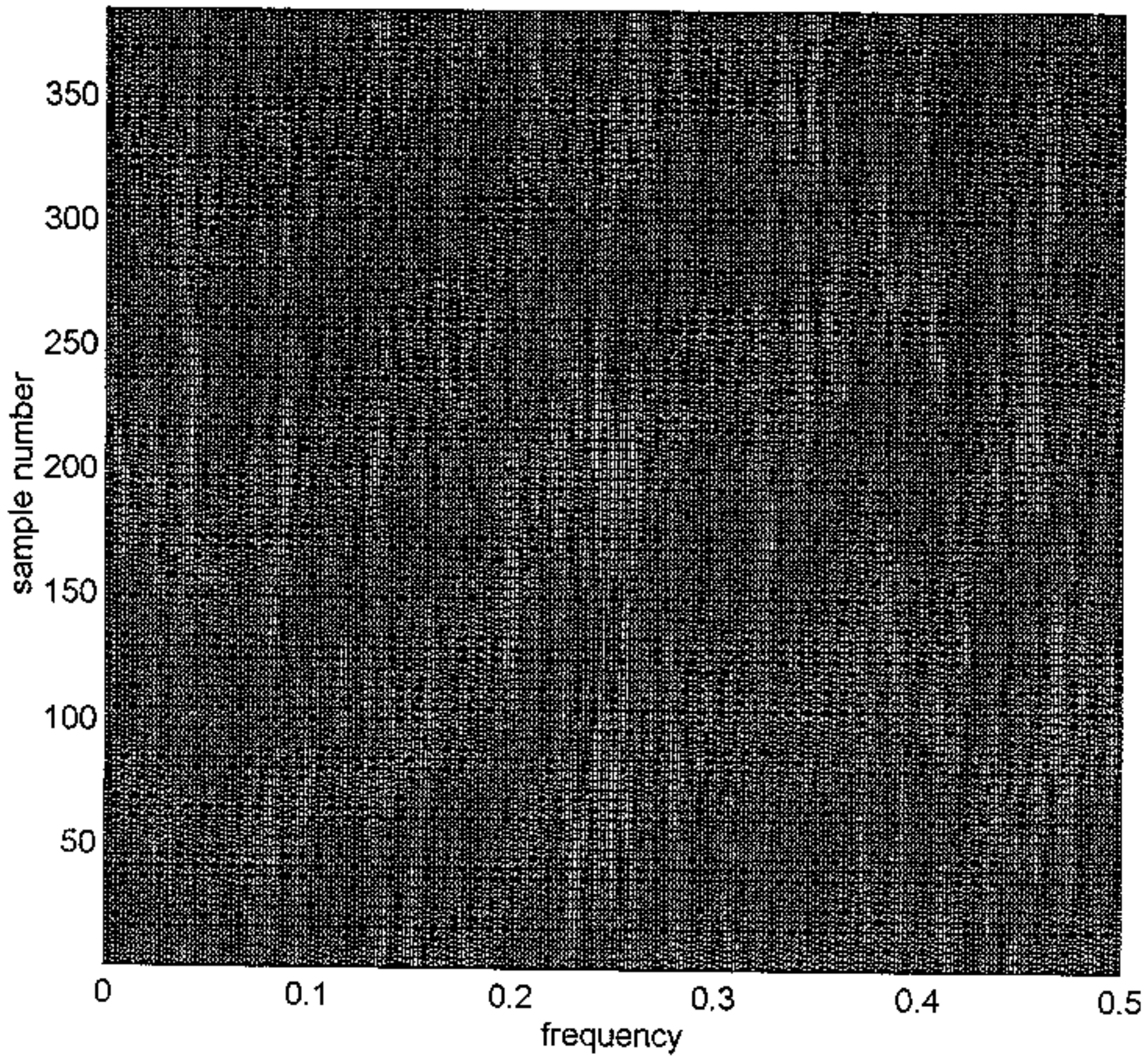


FIGURE 7.7b - SPECTROGRAM OF DATA



7.7 C - GRAY SCALE SPECTROGRAM
OF DATA

CLASSICAL LINEAR MODEL

RECALL MODEL

$$\underline{x} = \underline{H} \underline{\theta} + \underline{w}$$

\uparrow \uparrow \uparrow \uparrow
 $N \times 1$ $N \times p$ $p \times 1$ $N \times 1$

$$\underline{w} \sim N(\underline{0}, \sigma^2 \underline{I})$$

\underline{H} IS KNOWN

\underline{x} IS DATA VECTOR

$\underline{\theta}$ IS TO BE TESTED

IN GENERAL WE CONSIDER

$$\mathcal{H}_0 : \underline{A} \underline{\theta} = \underline{b}$$

$$\mathcal{H}_1 : \underline{A} \underline{\theta} \neq \underline{b} \quad \Gamma \leq p$$

\uparrow \uparrow
 $\Gamma \times p$ $\Gamma \times 1$

UNDER \mathcal{H}_0 , $\underline{\theta}$ LIES IN SOME "SUBSPACE"
OF \mathbb{R}^p

EXAMPLE : SIGNAL WITH UNKNOWN
AMPLITUDE

$$\mathcal{H}_0 : x(n) = w(n)$$

$$\mathcal{H}_1 : x(n) = A s(n) + w(n)$$

$n = 0, 1, \dots, N-1$
" "

$$\text{OR } H_0: A = 0$$

$$H_1: A \neq 0$$

$$\underline{\theta} = \underline{A}, \quad \underline{A} = \underline{I}, \quad \underline{b} = \underline{0}$$

$$\underline{x} = [x(0) \dots x(N-1)]^T, \quad \underline{H} = [s(0) \dots s(N-1)]^T$$

$$\underline{w} = [w(0) \dots w(N-1)]^T$$

THEOREM : GLRT FOR CLASSICAL
LINEAR MODEL

IF $\underline{x} = \underline{H}\underline{\theta} + \underline{w}$, \underline{H} IS KNOWN $N \times p$
MATRIX, $\underline{\theta}$ IS $p \times 1$, $\underline{w} \sim N(\underline{0}, \sigma^2 \underline{I})$,
TO TEST

$$H_0: \underline{A}\underline{\theta} = \underline{b}$$

$$H_1: \underline{A}\underline{\theta} \neq \underline{b}$$

$$\underline{A} \text{ IS } r \times p \text{ (} r \leq p \text{)}$$

$$\underline{b} \text{ IS } r \times 1$$

$\underline{A}\underline{\theta} = \underline{b}$ CONSISTENT (\Rightarrow AT LEAST
ONE SOLUTION)

GLRT DECIDES H_1 IF

$$T(\underline{x}) = \frac{(\underline{A}\hat{\underline{\theta}}_1 - \underline{b})^T [\underline{A}(\underline{H}^T \underline{H})^{-1} \underline{A}^T]^{-1} (\underline{A}\hat{\underline{\theta}}_1 - \underline{b})}{\sigma^2} > \gamma$$

WHERE $\hat{\theta}_1 = (\underline{H}^T \underline{H})^{-1} \underline{H}^T \underline{x}$
 IS MLE OF θ UNDER \mathcal{H}_1
 (NO RESTRICTIONS)

DETECTION PERFORMANCE IS
 (EXACTLY, NOT AS $N \rightarrow \infty$)

$$PFA = Q_{\chi_r^2}(\gamma')$$

$$P_D = Q_{\chi_r^2(\lambda)}(\gamma')$$

$$\text{WHERE } \lambda = \frac{(\underline{A} \theta_1 - \underline{b})^T [\underline{A} (\underline{H}^T \underline{H})^{-1} \underline{A}^T]^{-1} (\underline{A} \theta_1 - \underline{b})}{\sigma^2}$$

SEE APPENDIX 7B FOR PROOF

EXAMPLE: DC OFFSET COMPENSATION,
 SIGNAL WITH UNKNOWN A

$$\mathcal{H}_0: x(n) = \mu + w(n) \quad n=0, 1, \dots, N-1$$

$$\mathcal{H}_1: x(n) = \underbrace{A s(n)}_{\text{UNKNOWN}} + \underbrace{\mu + w(n)}_{\text{DC OFFSET-UNKNOWN}}$$

$w(n)$ IS WGN WITH KNOWN σ^2

UNDER H_1 IN FORM OF LINEAR
MODEL

$$\underline{x} = \begin{bmatrix} s(0) & 1 \\ s(1) & 1 \\ \vdots & \vdots \\ s(N-1) & 1 \end{bmatrix} \underbrace{\begin{bmatrix} A \\ \mu \end{bmatrix}}_{\underline{\theta}} + \underline{w}$$

\underline{H}

DETECTION PROBLEM IS

$$H_0: A = 0, \mu$$

$$H_1: A \neq 0, \mu$$

↑ NUISANCE

PARAMETER

OR $H_0: \begin{bmatrix} 1 & 0 \end{bmatrix} \begin{bmatrix} A \\ \mu \end{bmatrix} = 0$

$$H_1: \begin{bmatrix} 1 & 0 \end{bmatrix} \begin{bmatrix} A \\ \mu \end{bmatrix} \neq 0$$

$$\underbrace{1 \times 2}_{\underline{A}} \quad \underbrace{2 \times 1}_{\underline{\theta}} \quad \underbrace{1 \times 1}_{\underline{b}} \Rightarrow r=1, p=2$$

THEOREM APPLIES:

GLRT DECIDES H_1 IF

$$T(\underline{x}) = \frac{\hat{\underline{\theta}}_1^T \underline{A}^T [\underline{A} (\underline{H}^T \underline{H})^{-1} \underline{A}^T]^{-1} \underline{A} \hat{\underline{\theta}}_1}{0.2} > \gamma$$

$$\hat{\theta}_1 = (H^T H)^{-1} H^T x$$

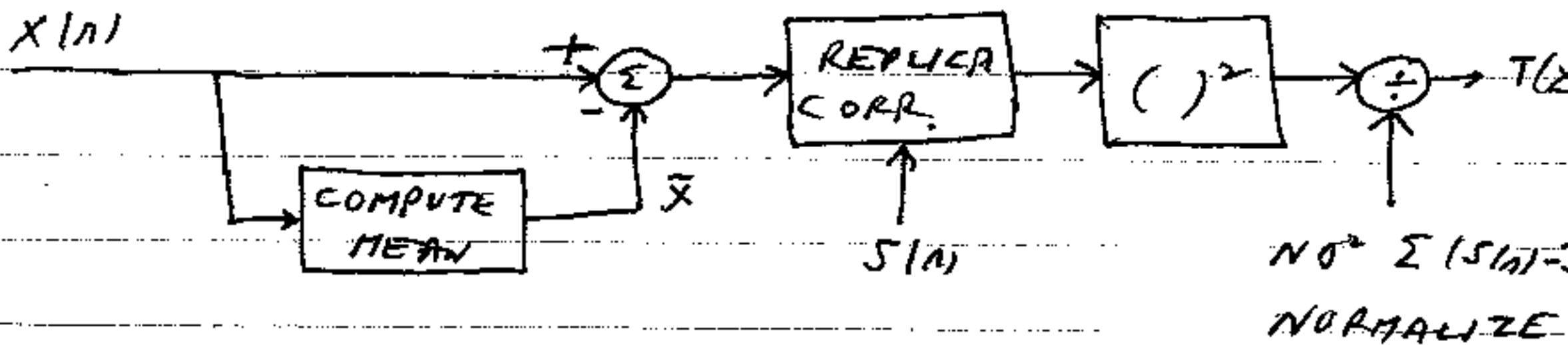
$$A = [1 \ 0]$$

AFTER SOME ALGEBRA (SEE BOOK)

$$T(x) = \frac{\left(\sum_{n=0}^{N-1} (x(n) - \bar{x}) s(n) \right)^2}{N \sigma^2}$$

$$N \sigma^2 = \sum_{n=0}^{N-1} (s(n) - \bar{s})^2$$

WHERE $\bar{s} = \frac{1}{N} \sum_{n=0}^{N-1} s(n)$



SINCE $\Gamma = 1$

$$P_{FA} = Q_{\chi^2}(\gamma) \Rightarrow \text{CAN FIND THRESHOLD}$$

NOTE THAT $T(x)$ IS INDEPENDENT OF \bar{s}
SINCE

$$T(x) = \frac{\left(\sum_{n=0}^{N-1} (x(n) - \bar{x})(s(n) - \bar{s}) \right)^2}{N \sigma^2 = \sum_{n=0}^{N-1} (s(n) - \bar{s})^2}$$

WHY?

ACTIVE SONAR / RADAR DETECTION

NOW CONSIDER A TYPICAL SONAR
NARROWBAND DETECTOR (ACTIVE)

TRANSMIT $S(t) = B \cos(2\pi f_T t + \phi_T)$

1) DUE TO ATTENUATION OF MEDIUM

RECEIVE $A \cos(2\pi f_T t + \phi_T)$ ($A \neq B$)

2) DUE TO REFLECTIONS OFF TARGET,
SEA SURFACE, ETC.

RECEIVE $A' \cos(2\pi f_T t + \phi)$ ($\phi \neq \phi_T$)

3) DUE TO DOPPLER EFFECT

RECEIVE $A \cos(2\pi f_0 t + \phi)$ ($f_0 \neq f_T$)

4) AFTER SAMPLING WE HAVE

$$A \cos(2\pi \underbrace{f_0}_{f_0} n + \phi)$$

A, f_0, ϕ UNKNOWN AS WELL

AS ARRIVAL TIME \Rightarrow GLRT

IS SPECTROGRAM OR DECIDE H_1 IF

$$\text{MAX}_{n_0, f_0} \frac{1}{M} \left| \sum_{n=n_0}^{n_0+M-1} x[n] \cdot e^{-j2\pi f_0 n} \right|^2 > \gamma$$

ASSUMING WGN OF VARIANCE σ^2 .

CHANGE NOTATION FOR SIGNAL
LENGTH TO N

TO COMPUTE SPECTROGRAM USE
FFT OR EVALUATE f_0 FOR

$$f_0 = k/N \quad k = 1, 2, \dots, N/2 - 1$$

(N EVEN)

(ASSUMES $f_0 \neq 0$ OR $1/2$)

DECIDE H_0 IF

$$\text{MAX}_{\substack{n_0 \\ k=1, 2, \dots, N/2-1}} \frac{1}{N} \left| \sum_{n=n_0}^{n_0+N-1} x[n] \cdot e^{-j \frac{2\pi}{N} k n} \right|^2$$

$\gamma \delta'$

SOME LOSS IF $f_0 \neq k/N$ -
CALLED SCALLOPING LOSS

$$\left| \sum_{n=n_0}^{n_0+N-1} A \cos(2\pi f_0 (n + \phi)) e^{-j \frac{2\pi}{N} k n} \right|^2$$

= MAGNITUDE-SQUARED OF
FOURIER TRANSFORM OF
SINUSOID EVALUATED AT $f = k/N$