ENEL420 Formula Sheet

The bitchiest of all the exams

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1 Interpolation

Interpolation, convolution with a kernel function:

$$f(k_0) = \sum_{k} f(k) \frac{K(k_0 - k)}{\text{convolution kernel}}$$
 (1)

Where $f(k_0)$ is the interpolated signal and f(k) is the data

1.1 Kernels

Nearest Neighbour:

$$K(x) = \text{rect}(x) = \begin{cases} 1 & |x| \le \frac{1}{2} \\ 0 & \text{otherwise} \end{cases}$$
 (2)

Linear:

$$K(x) = \operatorname{tri}(x) = \begin{cases} 1 - |x| & |x| < 1\\ 0 & \text{otherwise} \end{cases}$$
 (3)

Sinc:

$$K(x) = \operatorname{sinc}(x) = \frac{\sin(\pi x)}{\pi x} \tag{4}$$

2 Aliasing

General Case to avoid aliasing:

$$\frac{2f_u}{n} < f_s < \frac{2f_l}{n-1} \tag{5}$$

Where,

$$1 \le n \le I_{MLT} \left(\frac{f_u}{B}\right) \tag{6}$$

 I_{MLT} is Maximum integer less than

3 Z-Transform

$$F(z) = \sum_{n=0}^{\infty} f(n)z^{-n}$$
(7)

Or,

$$F(s) = \sum_{n=0}^{\infty} f(nT)e^{-nTs}$$
(8)

4 Quantisation Noise

RMS quantisation noise:

$$\sigma = \frac{\Delta}{\sqrt{12}} = \frac{\Delta}{2\sqrt{3}} \tag{9}$$

Noise Power:

$$NP \propto \Delta^2 \tag{10}$$

Therefore is noise power is decreased by $\frac{1}{2}$,

$$10\log_{10}\left(\frac{1}{2}^2\right) = 6dB \tag{11}$$

5 Filters

Direct realisation:

$$H(z) = \frac{\sum_{k=0}^{K} b_k z^{-k}}{1 + \sum_{m=1}^{M} a_m z^{-m}}$$
 (12)

Linear Phase, if a filter has linear phase it can be expressed in the form:

$$H(\omega) = |H(\omega)|e^{j\theta(\omega)} \tag{13}$$

Where,

$$\theta(\omega) = -(\alpha\omega + \beta) \tag{14}$$

Windowing the impulse response is defined by the inverse Fourier Transform,

$$h_D(n) = \frac{1}{2\pi} \int_{-\pi}^{\pi} H_D(\omega) e^{j\omega n} d\omega$$
 (15)

non-recursive FIR Filters,

$$y(n) = b_0 x(n) + b_1 x(n-1)$$
(16)

Recursive,

$$y(n) = ay(n-1) + bx(n)$$
(17)

So,

$$H(z) = \sum_{n=0}^{N-1} \frac{1}{N} \sum_{k=0}^{N-1} H(k) e^{\frac{j2\pi kn}{N}} z^{-n}$$
(18)

6 Detection Theory

Likelihood Ratio,

$$\frac{p(\mathbf{x}|H_0)}{p(\mathbf{x}|H_1)}\tag{19}$$

6.1 Likelihood ratio test

Choose null hypothesis if,

$$\frac{p(\mathbf{x}|H_0)}{p(\mathbf{x}|H_1)} > \gamma \tag{20}$$

Choose alternitive hypothesis if,

$$\frac{p(\mathbf{x}|H_0)}{p(\mathbf{x}|H_1)} \le \gamma \tag{21}$$

6.2 Log Likelihood

Choose null hypothesis if,

$$-x[0] + \frac{1}{2} > 0 \leftrightarrow x[0] < \frac{1}{2} \tag{22}$$

Choose alternitive hypothesis if,

$$-x[0] + \frac{1}{2} \le 0 \leftrightarrow x[0] \ge \frac{1}{2} \tag{23}$$

Solve for γ using,

$$P_{FA} = \int_{\mathbf{x}: L(\mathbf{x}) < \gamma} p(\mathbf{x}|H_0) d\mathbf{x}$$
 (24)

6.3 Missed Detection

The probability of missed detection is,

$$P_M = p(x[0] \le \frac{1}{2}|H_1) = p(x[0] \le \frac{1}{2}|u=1)$$
(25)

6.4 Detection Error Rate

$$P_{FA}P(H_0) + P_M P(H_1) (26)$$

Where P_{FA} is the probability of falsely selecting the alternative hypothesis and P_M is the probability of falsely selecting the null hypothesis.

6.5 Detection Threshold

To find the detection threshold use the following,

$$P_{FA} = \int_{\mathbf{x}} 1(\log(L(\mathbf{x})) - c < \gamma) \cdot p(\log(L(\mathbf{x})) - c|H_0) d\mathbf{x}$$
 (27)

6.6 NP Detector of Known Signal in WGN

The signal likelihood under w[n] $N(0, \sigma^2)$ (Normal distribution) for the null hypothesis,

$$p(\mathbf{x}|H_0) = \prod_{n=0}^{N-1} \frac{1}{\sqrt{2\pi}\sigma} \exp\left(-\frac{x[n]^2}{2\sigma^2}\right)$$
 (28)

For the alternative hypothesis,

$$p(x|H_1) = \prod_{n=0}^{N-1} \frac{1}{\sqrt{2\pi}\sigma} \exp\left(-\frac{(x[n] - s[n])^2}{2\sigma^2}\right)$$
 (29)

Therefore the Log-Likelihood ratio test is,

$$\log(L(\mathbf{x})) = \log\left(\frac{p(\mathbf{x}|H_0)}{p(\mathbf{x}|H_1)}\right) = \frac{1}{2\sigma^2} \sum_{n=0}^{N-1} (s[n]^2 - 2s[n]x[n])$$
(30)

6.7 Multivariate Gaussian PDF

Null Hypothesis,

$$P(\mathbf{x}|H_0) = \frac{\mathbf{1}}{(\sqrt{2\pi}\sigma)^N} \exp\left(-\frac{1}{2\sigma^2}\mathbf{x}^T\mathbf{x}\right) (\delta = \sigma^2 \mathbf{I})$$
(31)

Alternative Hypothesis,

$$P(\mathbf{x}|H_1) = \frac{\mathbf{1}}{|2\pi(\mathbf{C} + \sigma^2 \mathbf{I})|} exp\left(-\frac{1}{2}\mathbf{x}^T(\mathbf{C} + \sigma^2 \mathbf{I})^{-1}\mathbf{x}\right) (\Delta = \mathbf{C} + \sigma^2 \mathbf{I})$$
(32)

LLR,

$$\log(L(\mathbf{x})) \propto -\mathbf{x}^T \mathbf{C} (\mathbf{C} + \sigma^2 \mathbf{I})^{-1} \mathbf{x}$$
(33)

Simplified,

$$\sum_{n=0}^{N=1} \frac{\delta_n}{\delta_n + \sigma^2} y[n]^2 \tag{34}$$

6.8 Mixture Model

$$p(\mathbf{x}|\theta) = \sum_{k=1}^{K} w_k p_k(\mathbf{x}|\theta)$$
 (35)

Where, $p_k(\mathbf{x}|\theta)$ is the k^{th} base distribution and is the mixing weight of $p_k(\mathbf{x}|\theta)$ Where,

$$w_k \ge 0 \quad \text{and} \quad \sum_{k=1}^K w_k = 1 \tag{36}$$

7 Fourier Transform

7.1 Properties

Shift Theorem

$$x(t-t_0) \leftrightarrow X(f)\exp(-j2\pi f t_0)$$
 (37)

Scaling

$$x(at) \leftrightarrow \frac{1}{|a|} X\left(\frac{f}{a}\right)$$
 (38)

Convolution

$$x(t) \otimes y(t) \leftrightarrow X(f)Y(f)$$
 (39)

For a real signal

$$X(-f) = X^*(f) \tag{40}$$

DFT

$$X(k) = \frac{1}{N} \sum_{n=0}^{N-1} x(n) \exp(-j2\pi nk/N)$$
 (41)

IDFT

$$x(n) = \sum_{k=0}^{N-1} X(k) \exp(j2\pi nk/N)$$
 (42)

Parseval's Theorem

$$\sum_{n=0}^{N-1} [f(n)]^2 = \sum_{k=0}^{N-1} |F(k)|^2$$
(43)

2D Fourier Transform can be expressed as a weighted sum of harmonics,

$$\exp(j2\pi(ux+vy))\tag{44}$$

7.2 2D FT

$$F(u,v) = \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} \exp(-j2\pi(ux + vy))dxdy$$
 (45)

2D DFT,

$$F(p,q) = \frac{1}{N} \sum_{m=0}^{N-1} \left(\frac{1}{N} \sum_{n=0}^{N-1} f(m,n) \exp(-j2\pi q n/N) \right) \exp(-j2\pi p m/N)$$
 (46)

Rotation of an image rotating by α ,

$$F_{\alpha}(\rho;\phi) = \int_{0}^{2\pi} \int_{0}^{\infty} f(r;\theta) \exp(-j2\pi\rho r \cos(\theta - \phi)) r dr d\theta \tag{47}$$

So,

$$F_{\alpha}(\rho;\phi) = F(\rho;\phi - \alpha) \tag{48}$$

8 Projection

$$P(\theta, x') = \int_{-\infty}^{\infty} f(x', y') dy'$$
(49)