



Cheat Sheet DSP

DSP fundamentals (signals II) (Technische Universiteit Eindhoven)



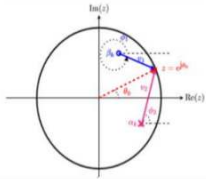
Scan to open on Studeersnel

Frequency response from pole-zero plot

$$H(z)|_{z=e^{j\theta}} = |H(e^{j\theta})| \cdot e^{j\Phi(e^{j\theta})} \text{ with } H(z) = A \cdot z^{p-q} \frac{\prod_{k=1}^q (z - \beta_k)}{\prod_{k=1}^p (z - \alpha_k)}$$

$$|H(e^{j\theta})| = |A| \times \left(\prod_{k=1}^q \text{length}(e^{j\theta} - \beta_k) \right) / \left(\prod_{k=1}^p \text{length}(e^{j\theta} - \alpha_k) \right)$$

$$\Phi(e^{j\theta}) = (p-q) \cdot \theta + \sum_{k=1}^q \arg(e^{j\theta} - \beta_k) - \sum_{k=1}^p \arg(e^{j\theta} - \alpha_k)$$



$$|H(e^{j\theta_0})| = v_1/v_2$$

$$\Phi(e^{j\theta_0}) = \phi_1 - \phi_2$$

Common Z-transform pairs

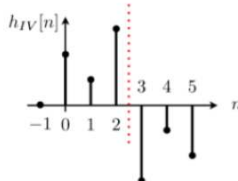
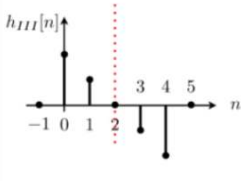
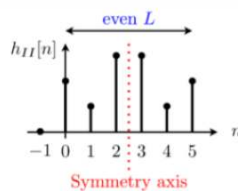
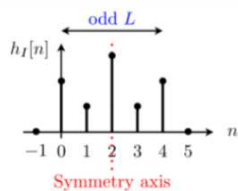
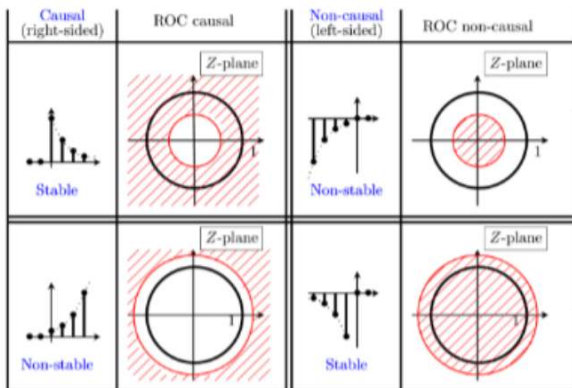
Sequence	Z-transform	ROC
$\delta[n]$	1	all z
$\delta[n-i]$	z^{-i}	$z \neq 0, \infty$
$a^n u[n]$	$\frac{1}{1-az^{-1}}$	$ z > a$
$-a^n u[-n-1]$	$\frac{1}{1-az^{-1}}$	$ z < a$
$na^n u[n]$	$\frac{az^{-1}}{(1-az^{-1})^2}$	$ z > a$
$-na^n u[-n-1]$	$\frac{az^{-1}}{(1-az^{-1})^2}$	$ z < a$
$a^n \cos(n\theta_0) u[n]$	$\frac{1-az^{-1}\cos(\theta_0)}{1-2az^{-1}\cos(\theta_0)+a^2z^{-2}}$	$ z > a$
$a^n \sin(n\theta_0) u[n]$	$\frac{az^{-1}\sin(\theta_0)}{1-2az^{-1}\cos(\theta_0)+a^2z^{-2}}$	$ z > a$
$a^n \sin(n\theta_0 + \phi) u[n]$	$\frac{\sin(\phi) + az^{-1}\sin(\theta_0 - \phi)}{1-2az^{-1}\cos(\theta_0) + a^2z^{-2}}$	$ z > a$

$$y[n] = \sum_{k=0}^q b_k x[n-k] - \sum_{k=1}^p a_k y[n-k]$$

$$H(z) = \frac{\sum_{k=0}^q b_k z^{-k}}{1 + \sum_{k=1}^p a_k z^{-k}} = b_0 \frac{\prod_{k=1}^q (1 - \beta_k z^{-1})}{\prod_{k=1}^p (1 - \alpha_k z^{-1})} = b_0 \cdot z^{p-q} \frac{\prod_{k=1}^q (z - \beta_k)}{\prod_{k=1}^p (z - \alpha_k)}$$

β_k values are the zeros of $H(z)$

α_k are the poles of $H(z)$



The definition of the Hanning window:

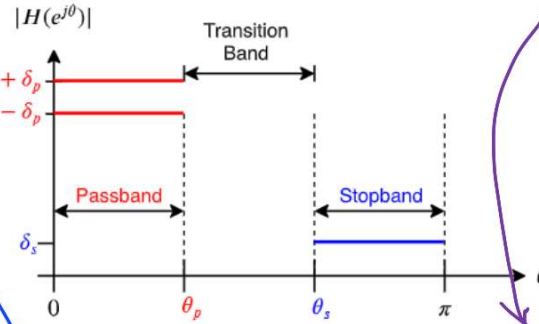
$$w[n] = \begin{cases} 0.5 - 0.5 \cos\left(\frac{2\pi n}{N}\right), & \text{if } 0 \leq n \leq N \\ 0, & \text{otherwise} \end{cases}$$

Window	A [dB]	Transition Δ	Stopband [dB]
Rectangular	-13	$1 \times (2\pi/N)$	-21
Hanning	-31	$3.1 \times (2\pi/N)$	-44
Hamming	-41	$3.3 \times (2\pi/N)$	-53
Blackman	-57	$5.5 \times (2\pi/N)$	-74

Circular shift: With DFT pair $x_p[n] \leftrightarrow X_p[k]$

Time shift: $x_p[n-i] \leftrightarrow e^{-j\frac{2\pi}{N}ki} \cdot X_p[k]$

Frequency shift: $e^{j\frac{2\pi}{N}ni} \cdot x_p[n] \leftrightarrow X_p[k-i]$



δ_p passband ripple
 δ_s stopband ripple
 ω cutoff freq

$$N = \frac{c}{\Delta f}$$

$$\Delta \theta = \theta_s - \theta_p$$

$$\Delta f = \frac{\Delta \theta}{2\pi}$$

$$h_d[n] = \frac{\sin(n-\alpha)\omega_c}{\pi(n-\alpha)} \quad \alpha = \frac{N}{2}$$

General specification LPT:

$$0 \leq \omega \leq \omega_p \quad 1 - \delta_p < |H(e^{j\omega})| \leq 1 + \delta_p$$

$$\omega_s \leq \omega < \pi$$

$$|H(e^{j\omega})| \leq \delta_s$$

$$\text{Gain} = \frac{v_1}{v_2} = \frac{\sqrt{\text{Real}^2 + \text{Complex}^2}}{\sqrt{\text{Real}^2 + \text{Complex}^2}}$$

to unit circle
zeros
pole

Definitions 2nd - order statistics: Real and WSS

- * Mean: $\mu_x[n] = E\{x[n]\}$
- * Variance: $\sigma_x^2 = E\{(x[n] - \mu_x)^2\} = E\{x^2[n]\} - \mu_x^2$
- * Autocorrelation: $r_x[l] = E\{x[n] \cdot x[n-l]\} = E\{x[n+l] \cdot x[n]\}$
 Notes: Power $E\{x^2[n]\} = r_x[0] \geq 0$ and $r_x[0] \geq r_x[l]$ $\forall l$
 Symmetry $r_x[l] = r_x[-l]$
- * Autocovariance: $\gamma_x[l] = E\{(x[n] - \mu_x) \cdot (x[n-l] - \mu_x)\}$
 $= E\{(x[n+l] - \mu_x) \cdot (x[n] - \mu_x)\} = r_x[l] - \mu_x^2$
- * Cross-correlation: $r_{xy}[l] = E\{x[n] \cdot y[n-l]\} = E\{x[n+l] \cdot y[n]\}$
- * Cross-covariance: $\gamma_{xy}[l] = E\{(x[n] - \mu_x) \cdot (y[n-l] - \mu_y)\}$
 $= r_{xy}[l] - \mu_x \cdot \mu_y$
- * Normalized γ_{xy} : $\rho_{xy}[l] = \frac{\gamma_{xy}[l]}{\sigma_x \cdot \sigma_y}$

Exercise 8 From $P_x(e^{j\theta}) = \frac{1}{4} + \cos(\theta)$ it follows: $r_x[l] = \frac{1}{4}\delta[l+1] + \frac{1}{2}\delta[l] + \frac{1}{4}\delta[l-1]$. Furthermore from the given impulse response we obtain: $y[n] = x[n] - \frac{1}{2}x[n-1]$. Thus:

$$r_y[l] = E\{y[n]y[n-l]\} = E\{(x[n] - \frac{1}{2}x[n-1])(x[n-l] - \frac{1}{2}x[n-l-1])\}$$

$$= r_x[l] - \frac{1}{2}r_x[l-1] - \frac{1}{2}r_x[l+1] + \frac{1}{4}r_x[l] = \frac{17}{16}\delta[l] - \frac{1}{4}\delta[l-2] - \frac{1}{4}\delta[l+2]$$

$l =$	-4	-3	-2	-1	0	1	2	3	4
$r_y[l]$	0	0	0	0	$\frac{17}{16}$	0	$-\frac{1}{4}$	0	0

$$= P_x(e^{j\theta}) \cdot \frac{1}{9} (e^{3j\theta} + 2e^{j\theta} + 3 + 2e^{-j\theta} + e^{-3j\theta})$$

4.1 a

$$r_x[l] = E\{x[n]x[n-l]\}$$

$$r_y[l] = E\{y[n]y[n-l]\}$$

$$r_y[l] = E\{y[n]y[n-l]\}$$

$$= E\{\frac{1}{3}(x[n] + x[n-1] + x[n-2]) \cdot \frac{1}{3}(x[n-l] + x[n-l-1] + x[n-l-2])\}$$

$$= (E\{x[n]x[n-l]\} + E\{x[n-1]x[n-l]\} + E\{x[n-2]x[n-l]\})$$

$$+ (E\{x[n]x[n-l-1]\} + E\{x[n-1]x[n-l-1]\} + E\{x[n-2]x[n-l-1]\})$$

$$+ (E\{x[n]x[n-l-2]\} + E\{x[n-1]x[n-l-2]\} + E\{x[n-2]x[n-l-2]\})$$

Since $r_x[l] = E\{x[n]x[n-l]\}$ we obtain:

$$r_y[l] = \frac{1}{9} (r_x[l] + r_x[l-1] + r_x[l-2]$$

$$+ r_x[l+1] + r_x[l] + r_x[l-1]$$

$$+ r_x[l+2] + r_x[l+1] + r_x[l])$$

$$= \frac{1}{9} (3r_x[l] + 2r_x[l-1] + 2r_x[l+1] + r_x[l+2] + r_x[l-2])$$

$$= \frac{1}{9} (3(\frac{17}{16}\delta[l] - \frac{1}{4}\delta[l-2] - \frac{1}{4}\delta[l+2]) + 2(\frac{17}{16}\delta[l-1] - \frac{1}{4}\delta[l-3] - \frac{1}{4}\delta[l+3]) + 2(\frac{17}{16}\delta[l+1] - \frac{1}{4}\delta[l-1] - \frac{1}{4}\delta[l+3]))$$

9
-2
12

$H(z) = \frac{1}{N} (1 - z^{-N}) \cdot \left(\frac{1}{1 - z^{-1}} - \frac{\frac{1}{2}}{1 - e^{j\frac{2\pi}{N}} z^{-1}} - \frac{\frac{1}{2}}{1 - e^{j\frac{2\pi}{N}(N-1)} z^{-1}} \right)$
 $= \frac{1}{N} (1 - z^{-N}) \cdot \left(\frac{1}{1 - z^{-1}} - \frac{1 - \cos(\frac{2\pi}{N}) z^{-1}}{1 - 2 \cos(\frac{2\pi}{N}) z^{-1} + z^{-2}} \right)$

Linear phase filter $h[n] = h[0]\delta[n] + h[1]\delta[n-1] + h[2]\delta[n-2] + h[3]\delta[n-3]$. Even length $L = 4$, thus two cases even- and odd-symmetry resp.:
 Case II $h[0] = h[3] = a; h[1] = h[2] = b \iff H_{II}(e^{j\theta}) = 2 \left(b \cos(\frac{1}{2}\theta) + a \cos(\frac{3}{2}\theta) \right) \cdot e^{-j\frac{3}{2}\theta}$
 Case IV $h[0] = -h[3] = a; h[1] = -h[2] = b \iff H_{IV}(e^{j\theta}) = 2 \left(b \sin(\frac{1}{2}\theta) + a \sin(\frac{3}{2}\theta) \right) \cdot e^{-j(\frac{3}{2}\theta - \frac{\pi}{2})}$
 a) Since $H(e^{j\theta})|_{\theta=0} = 0 \Rightarrow$ Case IV (odd-symmetry):
 $H_{IV}(e^{j\theta})|_{\theta=\frac{\pi}{3}} = -1 \equiv 2 \left(b \frac{1}{2} + a \cdot 0 \right) \cdot e^{-j\frac{3}{2}\pi}$
 $H_{IV}(e^{j\theta})|_{\theta=\frac{2\pi}{3}} = -j\sqrt{3} \equiv 2 \left(b \frac{1}{2} \sqrt{3} + a \cdot 0 \right) \cdot e^{-j\frac{3}{2} \cdot \frac{2\pi}{3}}$
 $\Rightarrow a = -1, b = 1$. Thus $h[n] = -\delta[n] + \delta[n-1] - \delta[n-2] + \delta[n-3]$

Discrete-time systems

$x[n] \xrightarrow{T, \{ \}} y[n] = T\{x[n]\}$

- Properties:**
 - LTI**: Linear Time-Invariance (book: "Shift"-Invariance) Additive, Homogeneous and Time-Invariant
 - Causality**: Response at n_0 depends on input up to $n = n_0$. In practice we cannot predict sample values, therefore it is important to design causal filters
 - (BIBO) Stability**: For $A, B < \infty, |x[n]| < A \Rightarrow |y[n]| < B$. Input bounded by some number A will yield output bounded by some number B
 - Invertibility**: Input may be uniquely determined from output. Only one input can be traced back from the output

The filter changes only the phase (amplitude stays the same). At frequency $\frac{\pi}{3}$ the phase is changed by $-\frac{\pi}{2}$, while at frequency $\frac{2\pi}{3}$ the phase is changed with $-\pi$. From the general expressions as given above it follows that this can only be achieved in case II.

amplitude change
 $H_{II}(e^{j\theta})|_{\theta=\frac{\pi}{3}} = 1 \cdot e^{-j\frac{3}{2}\pi} \equiv 2 \left(b \frac{1}{2} \sqrt{3} + a \cdot 0 \right) \cdot e^{-j\frac{3}{2}\pi}$
 $H_{II}(e^{j\theta})|_{\theta=\frac{2\pi}{3}} = 1 \cdot e^{-j\pi} \equiv 2 \left(b \frac{1}{2} + a \cdot (-1) \right) \cdot e^{-j\pi}$
 $\Rightarrow b = \frac{1}{3}\sqrt{3} (\approx 0.6)$ and $a = (\frac{1}{6}\sqrt{3} - \frac{1}{2}) (\approx -0.2)$. Thus $h[n] = a\delta[n] + b\delta[n-1] + b\delta[n-2] + a\delta[n-3]$

$h[n] = \frac{1}{N} \{ 1 - \cos(\frac{2\pi n}{N}) \}$
 DFT
 $= \frac{1}{N} \sum_{n=0}^{N-1} e^{-j\frac{2\pi}{N}kn} - \frac{1}{2N} \sum_{n=0}^{N-1} e^{-j\frac{2\pi}{N}(k-1)n} - \frac{1}{2N} \sum_{n=0}^{N-1} e^{-j\frac{2\pi}{N}(k+1)n}$
 $= \delta[k] - \frac{1}{2}\delta[k-1] - \frac{1}{2}\delta[k-N+1]$

* **Cauchy**: Containing spikes and therefore has high variability

* **Uniform**
 * **White noise**: uniformly distributed between -1 and 1

* **Gaussian**: Allowing for any possible (confined) value

Time domain: $h[n] = u[n] - u[n-N]$
 Frequency domain: $H(e^{j\theta}) = e^{-j\frac{N-1}{2}\theta} \cdot \frac{\sin(\frac{N}{2}\theta)}{\sin(\frac{\theta}{2})}$

Now we use 4-point DFT's. From the 4-point DFT values $X[k]$ of one period of $x[n]$ we are given:
 $X[0] = 0; X[1] = \sqrt{2} + j\sqrt{2}$ and $X[2] = 0$
 while from the 4-point DFT values of $H[k]$ of the length 4 impulse response $h[n]$ we are given:
 $H[0] = 2; H[1] = 1 + j$ and $H[2] = 0$
 Evaluate $x[n]$, $h[n]$ and the linear convolution result $y[n] = x[n] * h[n]$ all in the range for $n = -1, 0, 1, 2, 3, 4, 5, 6$.

Examples: Low Pass Filter
 Time domain: Sinc function $\frac{\sin(\theta/2)}{\theta/2}$
 Frequency domain: Ideal Low Pass Filter (LTP) $H(e^{j\theta}) = 1$ for $|\theta| \leq \theta_c$, elsewhere 0.

Sequence	FTD
$\delta[n]$	1
$\delta[n - n_0]$	$e^{-jn_0\theta}$
1	$2\pi\delta(\theta)$
$e^{jn\theta_0}$	$2\pi\delta(\theta - \theta_0)$
$a^n u[n], a < 1$	$\frac{1}{1 - ae^{-j\theta}}$
$-a^n u[-n-1], a > 1$	$\frac{1}{1 - ae^{-j\theta}}$
$\cos(n\theta_0)$	$\pi\delta(\theta + \theta_0) + \pi\delta(\theta - \theta_0)$

Ergodicity

- An ensemble of measurements may not be available
- A process is **ergodic** if the statistics can be found from one **single realization**
- Only stationary signals can be ergodic
- Stationarity ensures time invariance of statistics of random signal
- Ergodicity implies that the statistics can be calculated by time-averaging over a single representative member of the ensemble. Practice: Number of measured samples is limited to, say, $N \rightarrow$ "Replace" ensemble-averaging by time-averaging:
 $E\{\cdot\} = \frac{1}{N} \sum_{n=0}^{N-1} (\cdot)$

- Minimum phase:** Zeros inside unit circle
- Maximum phase:** Mirror zeros of minimum phase

$H(z) = z^{-1} + 2z^{-2} + z^{-3} = z^{-1}(1 + z^{-1})^2 \equiv (1 - z^{-4}) \cdot G(z) \Rightarrow$
 $G(z) = \frac{z^{-1}(1 + z^{-1})^2}{(1 - z^{-4})} = \frac{z^{-1}(1 + z^{-1})}{(1 - z^{-1}) \cdot (1 - jz^{-1}) \cdot (1 + jz^{-1})}$
 $= \frac{A}{(1 - z^{-1})} + \frac{B}{(1 - jz^{-1})} + \frac{C}{(1 + jz^{-1})} = \frac{1}{(1 - z^{-1})} + \frac{-\frac{1}{2}}{(1 - jz^{-1})} + \frac{-\frac{1}{2}}{(1 + jz^{-1})}$
 $2 \sin(\theta) \cdot e^{-j\omega T_s}$

\Rightarrow **Interpolation formula (time domain):**
 $x_a(t) = \sum_{n=-\infty}^{\infty} x[n] \left(\frac{\sin(\frac{\pi}{T_s} \cdot (t - nT_s))}{\frac{\pi}{T_s} \cdot (t - nT_s)} \right)$
 $y[n \cdot T_y] = x[n \cdot (M \cdot T_x)]$
 $X(e^{j\theta}) = \frac{1}{T_x} \sum_{k=-\infty}^{\infty} X_a(\frac{\theta}{T_x} - k \frac{2\pi}{T_x})$
 $Y(e^{j\theta}) = \frac{1}{T_y} \sum_{r=-\infty}^{\infty} X_a(\frac{\theta}{T_y} - r \frac{2\pi}{T_y})$

Example 1

Let's define the system output in Z-domain:
 $Y(z) = b_0 X(z) + b_1 z^{-1} X(z) + b_2 z^{-2} X(z) - a_1 z^{-1} Y(z) - a_2 z^{-2} Y(z)$

Split last sum with $r = p + k \cdot M$ and use $T_y = M \cdot T_x$
 $Y(e^{j\theta}) = \frac{1}{M} \sum_{p=0}^{M-1} \left\{ \frac{1}{T_x} \sum_{k=-\infty}^{\infty} X_a \left(\frac{\theta - p \cdot 2\pi}{M \cdot T_x} - k \cdot \frac{2\pi}{T_x} \right) \right\} \Rightarrow$
 $Y(e^{j\theta}) = \frac{1}{M} \sum_{p=0}^{M-1} X(e^{j(\frac{\theta}{M} - p \frac{2\pi}{M})})$

