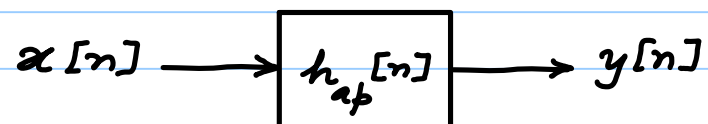


Consider the following causal and stable all-pass filter:



We showed earlier that $\sum_{n=-\infty}^{\infty} |x[n]|^2 = \sum_{n=-\infty}^{\infty} |y[n]|^2$

Since the filter is causal, and the i/p is applied at $n=0$, the lower limit can be replaced by $n=0$.

Now consider the following i/p: $x_1[n] = \begin{cases} x[n] & n \leq n_0 \\ 0 & n > n_0 \end{cases}$

Let $y_1[n]$ be the corresponding output. Then,

$$\sum_{n=0}^{\infty} |x_1[n]|^2 = \sum_{n=0}^{\infty} |y_1[n]|^2$$

$$\sum_{n=0}^{n_0} |x_1[n]|^2 = \sum_{n=0}^{n_0} |y_1[n]|^2 + \sum_{n=n_0+1}^{\infty} |y_1[n]|^2$$

$$= \sum_{n=0}^{n_0} |y[n]|^2 + \sum_{n=n_0+1}^{\infty} |y_1[n]|^2 \quad [\text{since } y_1[n] = y[n] \text{ for } n \leq n_0]$$

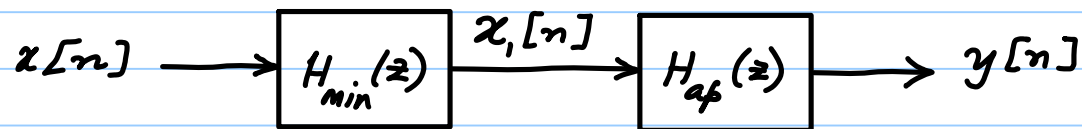
$$\geq \sum_{n=0}^{n_0} |y[n]|^2$$

Thus, for an all-pass filter,

$$\sum_{n=0}^{n_0} |x[n]|^2 \geq \sum_{n=0}^{n_0} |y[n]|^2$$

The above result can be used to show that min^m phase filters have the least energy delay.

Recall that any $H(z)$ can be decomposed as follows:



$y[n]$ is the o/p of an arbitrary, causal, stable filter

$x_1[n]$ is the o/p of the minimum phase counterpart of $H(z)$.

Using the previous result,

$$\sum_{n=0}^{n_0} |x_1[n]|^2 \geq \sum_{n=0}^{n_0} |y[n]|^2$$

That is, the *minimum-phase filter* has the *least energy lag*.

Hence the term "*minimum lag*" is more accurate than the well-entrenched "*minimum phase*" terminology.

"Causal" DTFT and its implications

Recall that $x[n] = 0$ for $n < 0$ imposed restrictions on the corresponding transform's real and imaginary parts.

Suppose now that $X(e^{j\omega}) = 0$ for $\omega < 0$, i.e., $X(e^{j\omega})$ is "causal".

Since $X(e^{j\omega})$ is periodic, "causal" here means $X(e^{j\omega}) = 0$ for

$-\pi < \omega < 0$. Similar to expressing $x[n]$ as $x_e[n] + x_o[n]$, consider

$$X_e(e^{j\omega}) = \frac{1}{2} [X(e^{j\omega}) + X^*(e^{-j\omega})]$$

$$X_o(e^{j\omega}) = \frac{1}{2j} [X(e^{j\omega}) - X^*(e^{-j\omega})]$$

We can recover $X(e^{j\omega})$ over $0 < \omega < \pi$ from either $X_e(e^{j\omega})$ or $X_o(e^{j\omega})$:

$$X(e^{j\omega}) = \begin{cases} 2X_e(e^{j\omega}) & 0 < \omega < \pi \\ 0 & -\pi < \omega < 0 \end{cases}$$

$$X(e^{j\omega}) = \begin{cases} 2jX_o(e^{j\omega}) & 0 < \omega < \pi \\ 0 & -\pi < \omega < 0 \end{cases}$$

One can also relate $X_e(e^{j\omega})$ and $X_o(e^{j\omega})$.

It is easy to see that

$$X_o(e^{j\omega}) = \begin{cases} -jX_e(e^{j\omega}) & 0 < \omega < \pi \\ jX_e(e^{j\omega}) & -\pi < \omega < 0 \end{cases}$$

That is,

$$X_o(e^{j\omega}) = X_e(e^{j\omega}) H(e^{j\omega})$$

where

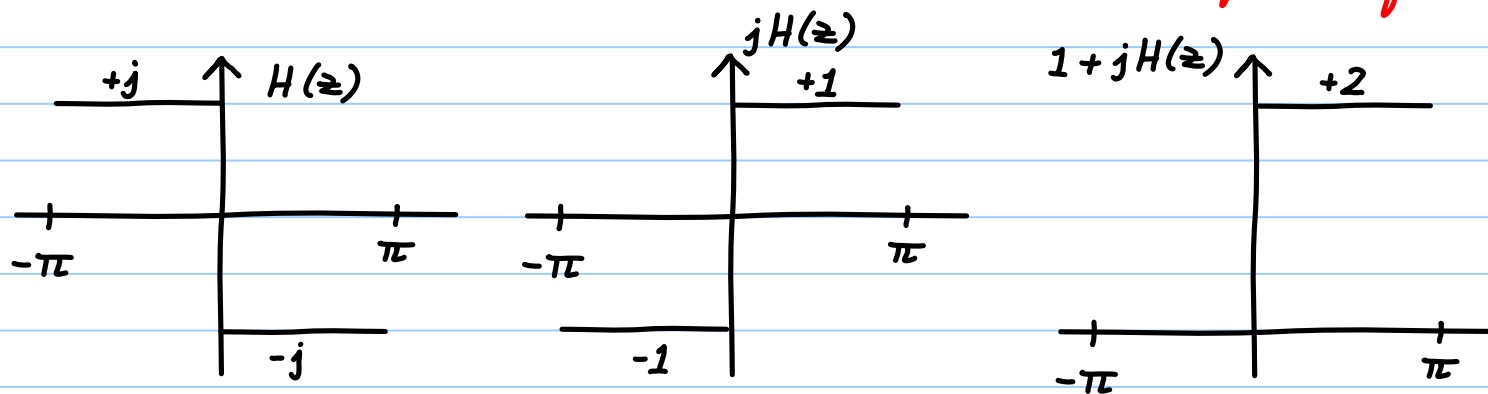
$$H(e^{j\omega}) = \begin{cases} -j & 0 < \omega < \pi \\ j & -\pi < \omega < 0 \end{cases}$$

Note that $x[n] = x_R[n] + jx_I[n]$

$$x_R[n] \leftrightarrow X_e(e^{j\omega})$$

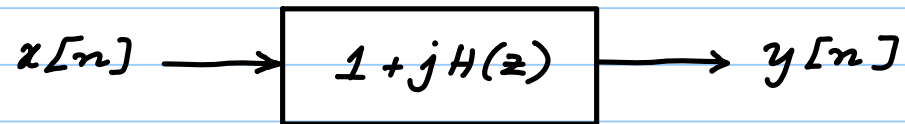
$$x_I[n] \leftrightarrow X_o(e^{j\omega})$$

Complex Half Band Filter



Let $G(z) = 1 + jH(z)$, whence it follows $G(e^{j\omega}) = \begin{cases} 2 & 0 < \omega < \pi \\ 0 & -\pi < \omega < 0 \end{cases}$

Hence if an arbitrary $x[n]$ is filtered using $G(e^{j\omega})$,
the output signal's DTFT becomes "causal" (or "one-sided").



$$g[n] = \delta[n] + jh[n]$$

Hence,

$$y[n] = x[n] * g[n]$$

$$y[n] = (\delta[n] + j h[n]) * x[n]$$

$$= x[n] + j x[n] * h[n]$$

$$= x[n] + j \hat{x}[n] = x_R[n] + j x_I[n] \Rightarrow x_R[n] \text{ \& } x_I[n] \text{ are not independent}$$

where $\hat{x}[n] = x[n] * h[n]$

Since $H(e^{j\omega}) = \begin{cases} -j & 0 < \omega < \pi \\ j & -\pi < \omega < 0 \end{cases}$, one can easily verify that

$$h[n] = \begin{cases} \frac{\sin^2(n\pi/2)}{n\pi/2} & n \neq 0 \\ 0 & n = 0 \end{cases}$$

$h[n] \leftrightarrow H(e^{j\omega})$ is called as the IDEAL HILBERT TRANSFORMER

Exercise

Explore the relationship between the real-valued halfband filter, complex halfband filter, and the Hilbert transformer. The response of a real halfband filter is given below.

