

DAUBECHIES FILTERS

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In Session 1, we constructed the **orthogonal Haar Wavelet Transform**:

$$W_8 = \begin{bmatrix} h_1 & h_0 & 0 & 0 & 0 & 0 & 0 & 0 \\ 0 & 0 & h_1 & h_0 & 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 & h_1 & h_0 & 0 & 0 \\ 0 & 0 & 0 & 0 & 0 & 0 & h_1 & h_0 \\ \hline g_1 & g_0 & 0 & 0 & 0 & 0 & 0 & 0 \\ 0 & 0 & g_1 & g_0 & 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 & g_1 & g_0 & 0 & 0 \\ 0 & 0 & 0 & 0 & 0 & 0 & g_1 & g_0 \end{bmatrix}$$

where

$$\mathbf{h} = (h_0, h_1) = (\sqrt{2}/2, \sqrt{2}/2), \quad \mathbf{g} = (g_0, g_1) = (\sqrt{2}/2, -\sqrt{2}/2)$$



- ▶ $\mathbf{h} = (h_0, h_1) = (\sqrt{2}/2, \sqrt{2}/2)$ is called a **lowpass filter**.
- ▶ Lowpass filters tend to preserve non-oscillatory data and either dampen or annihilate oscillatory data.
- ▶ Lowpass filters are best analyzed using a Fourier series.



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- ▶ Lowpass filters are best analyzed using a Fourier series.



- ▶ Pretending we're engineers for the moment, let's "form" the Fourier series

$$H(\omega) = \sum_{k=0}^1 h_k e^{ik\omega} = \frac{\sqrt{2}}{2} + \frac{\sqrt{2}}{2} e^{i\omega}$$

- ▶ We wish to plot $|H(\omega)|$ so simplifying gives

$$\begin{aligned} H(\omega) &= \frac{\sqrt{2}}{2} + \frac{\sqrt{2}}{2} e^{i\omega} \\ &= \frac{\sqrt{2}}{2} e^{i\omega/2} (e^{-i\omega/2} + e^{i\omega/2}) \\ &= \sqrt{2} e^{i\omega/2} \cos(\omega/2) \end{aligned}$$

- ▶ So

$$|H(\omega)| = \sqrt{2} \cos(\omega/2)$$



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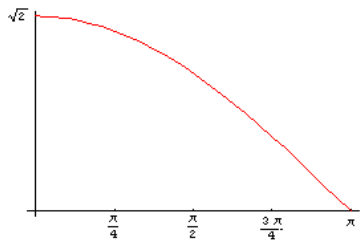
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Plotting $|H(\omega)| = \sqrt{2} \cos(\omega/2)$



- ▶ Suppose $\mathbf{h} = (h_n, \dots, h_m)$ with Fourier series

$$H(\omega) = \sum_{k=n}^m h_k e^{ik\omega}$$

- ▶ We will say \mathbf{h} is a **lowpass filter** if

$$H(0) = \sum_{k=n}^m h_k = \sqrt{2}$$

and

$$H(\pi) = \sum_{k=n}^m h_k e^{ik\pi} = \sum_{k=n}^m h_k (-1)^k = 0$$



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- ▶ If we write down the Fourier series for $\mathbf{g} = (\frac{\sqrt{2}}{2}, -\frac{\sqrt{2}}{2})$, we have

$$G(\omega) = \frac{\sqrt{2}}{2} - \frac{\sqrt{2}}{2} e^{-i\omega}$$

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$$\begin{aligned} G(\omega) &= \frac{\sqrt{2}}{2} - \frac{\sqrt{2}}{2} e^{-i\omega} \\ &= \frac{\sqrt{2}}{2} e^{-i\omega/2} (e^{i\omega/2} - e^{-i\omega/2}) \\ &= \sqrt{2} i e^{-i\omega/2} \sin(\omega/2) \end{aligned}$$

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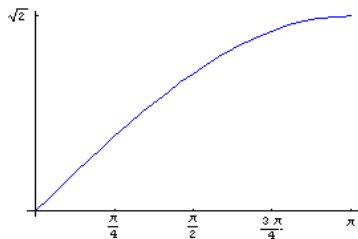
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- ▶ We can build longer **orthogonal filters** \mathbf{h} , \mathbf{g} .
- ▶ Orthogonal filters are those that give rise to an orthogonal wavelet transformation W_N (N even).
- ▶ It turns out length 3 doesn't work, so we will try length 4.
- ▶ Let lowpass $\mathbf{h} = (h_0, h_1, h_2, h_3)$ and highpass $\mathbf{g} = (g_0, g_1, g_2, g_3)$.



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We form the matrix

$$W_8 = \begin{bmatrix} h_3 & h_2 & h_1 & h_0 & 0 & 0 & 0 & 0 \\ 0 & 0 & h_3 & h_2 & h_1 & h_0 & 0 & 0 \\ 0 & 0 & 0 & 0 & h_3 & h_2 & h_1 & h_0 \\ h_1 & h_0 & 0 & 0 & 0 & 0 & h_3 & h_2 \\ \hline g_3 & g_2 & g_1 & g_0 & 0 & 0 & 0 & 0 \\ 0 & 0 & g_3 & g_2 & g_1 & g_0 & 0 & 0 \\ 0 & 0 & 0 & 0 & g_3 & g_2 & g_1 & g_0 \\ g_1 & g_0 & 0 & 0 & 0 & 0 & g_3 & g_2 \end{bmatrix}$$



- ▶ The order of the filters is not important - it turns out reflections work as well.
- ▶ The reverse order I've used reflects the fact that the wavelet matrix was first motivated using convolution (see Section 5.1 of the text).
- ▶ Note the “wrapping of rows”. This makes it easy to set up orthogonality conditions for W_8 , but is not practical in applications.
- ▶ In this case, the fourth and eighth elements of the transformed data will be built using elements from the beginning and end of the input vector.
- ▶ This makes sense if the data are periodic ...
- ▶ But data are not typically periodic!
- ▶ We can periodize input data, but then we need more from our filters ... that's coming!



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What orthogonality conditions can we impose on W_8 ?

$$W_8 = \begin{bmatrix} h_3 & h_2 & h_1 & h_0 & 0 & 0 & 0 & 0 \\ 0 & 0 & h_3 & h_2 & h_1 & h_0 & 0 & 0 \\ 0 & 0 & 0 & 0 & h_3 & h_2 & h_1 & h_0 \\ h_1 & h_0 & 0 & 0 & 0 & 0 & h_3 & h_2 \\ \hline g_3 & g_2 & g_1 & g_0 & 0 & 0 & 0 & 0 \\ 0 & 0 & g_3 & g_2 & g_1 & g_0 & 0 & 0 \\ 0 & 0 & 0 & 0 & g_3 & g_2 & g_1 & g_0 \\ g_1 & g_0 & 0 & 0 & 0 & 0 & g_3 & g_2 \end{bmatrix}$$



Orthogonality Conditions

$$h_0^2 + h_1^2 + h_2^2 + h_3^2 = 1$$

$$h_0 h_2 + h_1 h_3 = 0$$

$$g_0^2 + g_1^2 + g_2^2 + g_3^2 = 1$$

$$g_0 g_2 + g_1 g_3 = 0$$

$$h_0 g_0 + h_1 g_1 + h_2 g_2 + h_3 g_3 = 0$$

$$h_0 g_2 + h_1 g_3 = 0$$

$$h_2 g_0 + h_3 g_1 = 0$$



and Lowpass/Highpass Conditions

$$h_0 + h_1 + h_2 + h_3 = \sqrt{2}$$

$$h_0 - h_1 + h_2 - h_3 = 0$$

$$g_0 + g_1 + g_2 + g_3 = 0$$

$$g_0 - g_1 + g_2 - g_3 = \sqrt{2}$$



Look at the 5th equation:

$$h_0^2 + h_1^2 + h_2^2 + h_3^2 = 1$$

$$h_0 h_2 + h_1 h_3 = 0$$

$$g_0^2 + g_1^2 + g_2^2 + g_3^2 = 1$$

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Given h_0, h_1, h_2, h_3 , can you find g_0, g_1, g_2, g_3 to satisfy it?



► How about

$$g_0 = h_3, g_1 = -h_2, g_2 = h_1, g_3 = -h_0$$

► Then

$$h_0 g_0 + h_1 g_1 + h_2 g_2 + h_3 g_3 = h_0(h_3) + h_1(-h_2) + h_2(h_1) + h_3(-h_0) = 0$$

► Moreover, if $h_0 + h_1 + h_2 + h_3 = \sqrt{2}$ and $h_0 - h_1 + h_2 - h_3 = 0$, then

$$g_0 + g_1 + g_2 + g_3 = h_3 - h_2 + h_1 - h_0 = 0$$

and

$$g_0 - g_1 + g_2 - g_3 = h_3 - (-h_2) + h_1 - (-h_0) = \sqrt{2}$$

► g is highpass!



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- ▶ \mathbf{g} is highpass!



- ▶ There's more good news ...
- ▶ The orthogonality conditions reduce to

$$\begin{aligned}h_0^2 + h_1^2 + h_2^2 + h_3^2 &= 1 \\h_0 h_2 + h_1 h_3 &= 0\end{aligned}$$

- ▶ Add to that the lowpass conditions:

$$\begin{aligned}h_0 + h_1 + h_2 + h_3 &= \sqrt{2} \\h_0 - h_1 + h_2 - h_3 &= 0\end{aligned}$$

and we have a (quadratic) system to solve.

- ▶ The moral of the story is **we can build \mathbf{g} from \mathbf{h} .**



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- ▶ The good news (for engineers) is that the system has an infinite number of solutions.
- ▶ We can add a condition!
- ▶ The condition Ingrid Daubechies added:
- ▶ “Flatten” $H(\omega)$ at $\omega = \pi$.
- ▶ Set $H'(\pi) = 0$!



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► Differentiating

$$H(\omega) = h_0 + h_1 e^{i\omega} + h_2 e^{2i\omega} + h_3 e^{3i\omega}$$

gives

$$H'(\omega) = ih_1 e^{i\omega} + 2ih_2 e^{2i\omega} + 3ih_3 e^{3i\omega}$$

► Plugging in $\omega = \pi$ and simplifying gives the condition

$$h_1 - 2h_2 + 3h_3 = 0$$



► Differentiating

$$H(\omega) = h_0 + h_1 e^{i\omega} + h_2 e^{2i\omega} + h_3 e^{3i\omega}$$

gives

$$H'(\omega) = ih_1 e^{i\omega} + 2ih_2 e^{2i\omega} + 3ih_3 e^{3i\omega}$$

► Plugging in $\omega = \pi$ and simplifying gives the condition

$$h_1 - 2h_2 + 3h_3 = 0$$



So we want to solve

$$h_0^2 + h_1^2 + h_2^2 + h_3^2 = 1$$

$$h_0 h_2 + h_1 h_3 = 0$$

$$h_0 + h_1 + h_2 + h_3 = \sqrt{2}$$

$$h_0 - h_1 + h_2 - h_3 = 0$$

$$h_1 - 2h_2 + 3h_3 = 0$$



Let's look at the notebook

```
DaubechiesFilters.nb
```

