

Last Time: General Wavelet Analysis Pipeline

Setup:

$f(t)$: underlying analog signal

$a_I[n]$: discrete approximation at some high resolution I that you actually get to work with (e.g., image, audio signal, etc.)

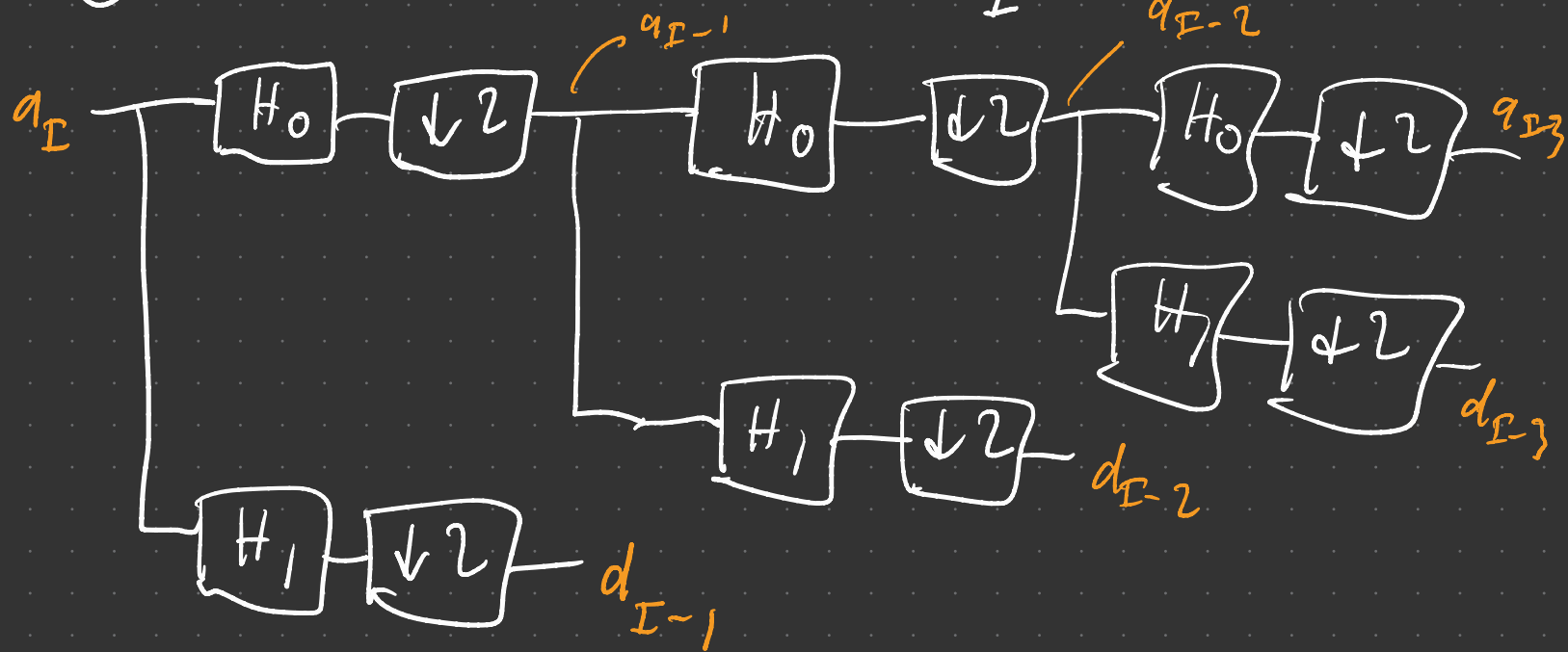
$\phi(t), \psi(t)$: scaling and wavelet functions that give rise to a MRA.

Fundamental assumption:

$$f(t) \approx \sum_{n \in \mathbb{Z}} a_I[n] \phi_{I,n}(t)$$

Pipeline:

① Perform a DWT on $a_I[n]$



→ Iterate as many times as you want

→ The filters come from the scaling and wavelet functions

$$\bullet h_0[n] = \tilde{h}_0[-n], \quad \tilde{h}_0[n] = \langle \phi, e_{1,n} \rangle$$

$$\bullet h_1[n] = \tilde{h}_1[-n], \quad \tilde{h}_1[n] = \langle \psi, e_{1,n} \rangle$$

② Do some (typically nonlinear) processing on, e.g.,

$$a_{I-3}, d_{I-3}, d_{I-2}, d_{I-1}$$

→ For thresholding procedures people typically only threshold the detail coefficients

③ Perform an IDWT on the processed signals, resulting in $\hat{d}_I[n]$.

④ Processed analog signal:

$$\hat{f}(t) = \sum_{n \in \mathbb{Z}} \hat{d}_I[n] \varphi_{I,n}(t)$$

Slogan: "Think analog, act digital"

MRA

DWT

Obs: This whole pipeline is "blind" to the underlying MRA, φ , and ψ .

- We only need the filters h_0 & h_1 (or, equivalently \tilde{h}_0 & \tilde{h}_1).

Remark: Last time we saw how ϕ & ψ determine \tilde{h}_0 & \tilde{h}_1 . Today we will see how \tilde{h}_0 & \tilde{h}_1 determine ϕ & ψ .

Last Time: "filters from wavelets"

Today: "wavelets from filters"

Idea: We can iterate the DWT as many times as we want. Therefore, we should be able to iterate it infinitely many times.

This procedure needs to converge

Equivalently, given $\tilde{h}_0[n]$, the iteration

cascade algorithm

$$\phi^{(k+1)}(t) = \sum_{n \in \mathbb{Z}} \underbrace{\tilde{h}_0[n]}_{h[n]} \sqrt{2} \phi^{(k)}(2t-n)$$

must converge.

Obs: We are looking for the fixed point of the two-scale equation.

Defⁿ: x is called the fixed point of

$$x_{k+1} = F(x_k)$$

if $x = F(x)$.

Obs: Given an initial value x_0 , if the sequence $\{x_k\}_{k=0}^{\infty}$ converges to x , i.e., $\lim_{k \rightarrow \infty} x_k = x$, then x is a fixed point of F .

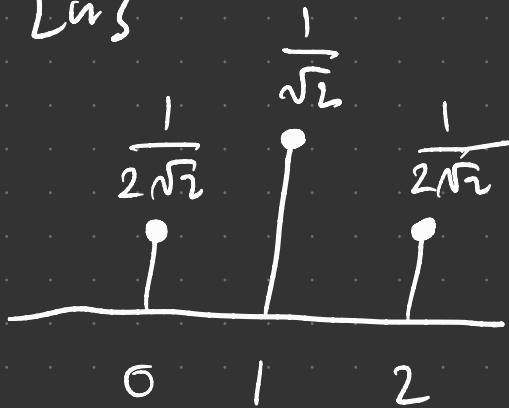
Ex: Does the sequence $x_{k+1} = \frac{x_k}{2}$ converge? What is its fixed point?

$$x_1 = \frac{x_0}{2}, x_2 = \frac{x_1}{2} = \frac{x_0}{4}, \dots, x_k = \frac{x_0}{2^k}$$

This sequence converges to 0 for any choice of x_0 .

$$\theta = \frac{0}{2} \quad \text{fixed point}$$

Exer: $h[n]$



What is the fixed point of the cascade algorithm?

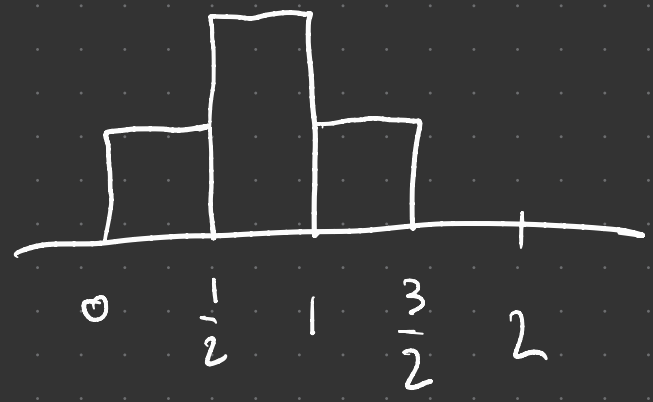
$$e(t) = \frac{1}{2} e(2t) + e(2t-1) + \frac{1}{2} e(2t-2)$$

Let us try to run the cascade algorithm with the initial condition

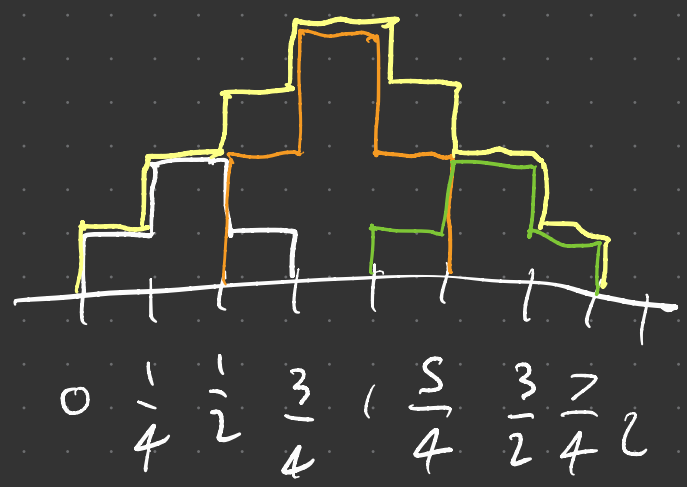
$e^{(0)}(t)$



$\varphi^{(1)}(t)$



$\varphi^{(2)}(t)$



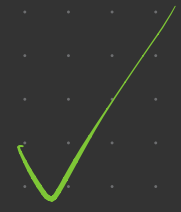
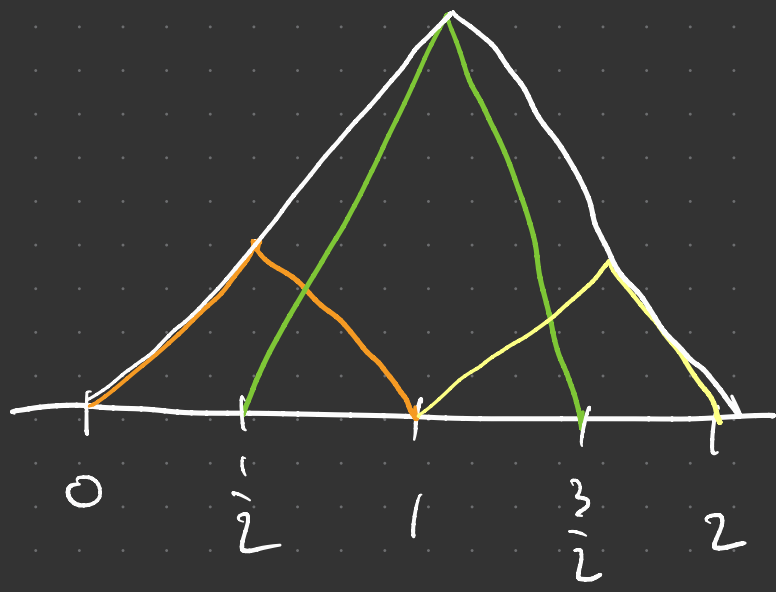
keep iterating ...

$\varphi(t) = \varphi^{(100)}(t)$



Exer: Do one more iteration.

Check:



Q: How do we understand this in general?

A: Fourier transform.

$$\mathcal{F} \left\{ \begin{aligned} & \varphi(t) = \sum_{n \in \mathbb{Z}} h[n] \sqrt{2} \varphi(2t - n) \end{aligned} \right.$$

$$\int_{-\infty}^{\infty} \varphi(t) e^{-j\omega t} dt = \int_{-\infty}^{\infty} \left(\sum_{n \in \mathbb{Z}} h[n] \sqrt{2} \varphi(2t - n) \right) e^{-j\omega t} dt$$

$\Phi(\omega)$

$$= \sum_{n \in \mathbb{Z}} h[n] \cdot \sqrt{2} \int_{-\infty}^{\infty} \varphi(2t - n) e^{-j\omega t} dt$$

$$\left[\begin{aligned} u &= 2t - n \\ du &= 2dt \end{aligned} \right]$$

$$= \frac{1}{\sqrt{2}} \sum_{n \in \mathbb{Z}} h[n] \int_{-\infty}^{\infty} \varphi(u) e^{-j\omega \left(\frac{u+n}{2} \right)} dt$$

$$= \frac{1}{\sqrt{2}} \left(\sum_{n \in \mathbb{Z}} h[n] e^{-j\omega \frac{n}{2}} \right) \int_{-\infty}^{\infty} e(\omega) e^{-j\omega u} \frac{1}{2} du$$

$H(e^{j\frac{\omega}{2}})$
 $\Phi(\frac{\omega}{2})$

$$\Phi(\omega) = \frac{H(e^{j\frac{\omega}{2}})}{\sqrt{2}} \Phi(\frac{\omega}{2})$$

$$= \frac{H(e^{j\frac{\omega}{2}})}{\sqrt{2}} \cdot \frac{H(e^{j\frac{\omega}{4}})}{\sqrt{2}} \Phi(\frac{\omega}{4})$$

$$= \frac{H(e^{j\frac{\omega}{2}})}{\sqrt{2}} \cdot \frac{H(e^{j\frac{\omega}{4}})}{\sqrt{2}} \cdot \frac{H(e^{j\frac{\omega}{8}})}{\sqrt{2}} \Phi(\frac{\omega}{8})$$

$$\vdots$$

$$= \Phi(0) \prod_{i=1}^{\infty} \frac{H(e^{j2^{-i}\omega})}{\sqrt{2}}$$

Recall: $\Phi(\omega) = \int_{-\infty}^{\infty} e(t) e^{-j\omega t} dt$

$$= \int_{-\infty}^{\infty} e(t) dt$$

We often impose that $\int e(t) dt = 1$

(e.g., this holds for the box )

Infinite-Product formula:

$$\Phi(\omega) = \prod_{i=1}^{\infty} \frac{H(e^{j2^{-i}\omega})}{\sqrt{2}}$$

$$e(t) = \frac{1}{2\pi} \int_{-\infty}^{\infty} \Phi(\omega) e^{j\omega t} d\omega$$

inverse Fourier transform

"wavelets from filters"

Obs: Convergence of this infinite product is the same as convergence of the cascade algorithm, which is the same as convergence of an infinite-level DWT.

Remark: If $H(e^{j\omega})$ is continuously differentiable at $\omega=0$, and

$$\min_{\omega \in [-\frac{\pi}{2}, \frac{\pi}{2}]} |H(e^{j\omega})| > 0,$$

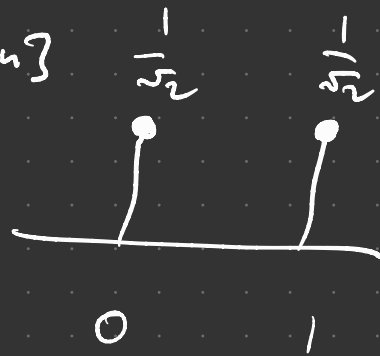
then the infinite product converges.
(Mallat, 1989).

Exer:

$e(t)$



$h(\omega)$



Compute $\Phi(\omega)$ from the infinite product formula.

$$\frac{H(e^{j\omega})}{\sqrt{2}} = \frac{1}{\sqrt{2}} \left[\frac{1}{\sqrt{2}} (1 + e^{-j\omega}) \right]$$

$$= \frac{1}{2} (1 + e^{-j\omega})$$

$$H^{(N)}(e^{j\omega}) = \frac{1}{2^N} \prod_{i=1}^N (1 + e^{-j2^i \omega})$$

[check at home]

$$= \frac{1}{2^N} \sum_{k=0}^{2^N - 1} e^{-j2^N \omega k}$$

2^N terms

$$\left[\sum_{k=0}^n r^k = \frac{1-r^{n+1}}{1-r} \right] = \frac{1}{2^N} \left(\frac{1-e^{-j\omega}}{1-e^{-j2^{-N}\omega}} \right)$$

$$\left[e^{-j\theta} = 1 - j\theta + \frac{\theta^2}{2} - \frac{j\theta^3}{3} + \dots \right]$$

$$\theta = 2^{-N}\omega$$

$$\xrightarrow{N \rightarrow \infty} \frac{1-e^{-j\omega}}{j\omega}$$

$$\Phi(\omega) = \frac{1-e^{-j\omega}}{j\omega} = e^{-j\frac{\omega}{2}} \left(\frac{e^{j\frac{\omega}{2}} - e^{-j\frac{\omega}{2}}}{j\omega} \right)$$

Modulated
Sinc

$$= e^{-j\frac{\omega}{2}} \left(\frac{2 \sin\left(\frac{\omega}{2}\right)}{\omega} \right)$$

Remark: This recovers Euler's celebrated infinite product formula for the sine.

Fundamental Theorem of Wavelet Analysis (Mallat, 1989)

Let $\phi \in L^2(\mathbb{R})$ be a valid scaling function.

Then, the filter $h[n] = \langle \phi, \phi_{1,n} \rangle$ must satisfy

$$\textcircled{1} \quad |H(e^{j\omega})|^2 + |H(e^{j(\omega+\pi)})|^2 = 2$$

$$\textcircled{2} \quad H(e^{j0}) = \sum_{n \in \mathbb{Z}} h[n] = \sqrt{2}$$

$$\textcircled{3} \quad \min_{\omega \in [-\frac{\pi}{2}, \frac{\pi}{2}]} |H(e^{j\omega})| > 0.$$

On the other hand, given a filter $h[n]$

such that $H(e^{j\omega})$ satisfies $\textcircled{1}$, $\textcircled{2}$, & $\textcircled{3}$,

then the inverse Fourier transform of

$$\Phi(\omega) = \prod_{i=1}^{\infty} \frac{H(e^{j2^{-i}\omega})}{\sqrt{2}}$$

exists and is a valid scaling function.

Obs: This gives a one-to-one correspondence between scaling functions ϕ and filters h .

Remark:

$$\begin{aligned} \tilde{h}_0[n] &= h[n] \\ \tilde{h}_1[n] &= (-1)^{1-n} h[1-n] \end{aligned}$$

if you allow for complex coeffs.

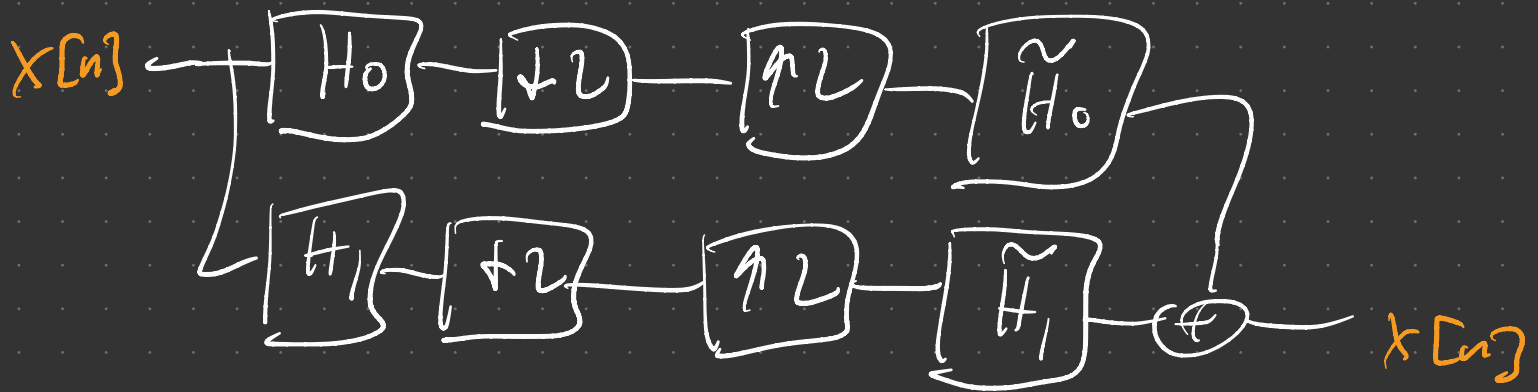
conjugate mirror filters

Ex: Haar

$$\tilde{h}_0[n] \begin{array}{c} 1 \\ 1 \\ \hline 0 \end{array}$$

$$\tilde{h}_1[n] \begin{array}{c} 1 \\ -1 \\ \hline 0 \end{array}$$

Ex: $h_0[n] = \tilde{h}_0[-n]$
 $h_1[n] = \tilde{h}_1[-n]$



Show that this is a PR FB
 with $L=0$.

$$\tilde{H}_0(z) H_0(z) + \tilde{H}_1(z) H_1(z) = 2z^{-L}$$

$$H(e^{j\omega}) H(e^{-j\omega}) + H(e^{j(\omega+\pi)}) H(e^{-j(\omega+\pi)}) = 2$$

$$|H(e^{j\omega})|^2 + |H(e^{j(\omega+\pi)})|^2 = 2$$

→ This is ① from the fundamental theorem.