

# Last Time: General Wavelet Analysis Pipeline

## Setup:

$f(t)$ : underlying analog signal

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

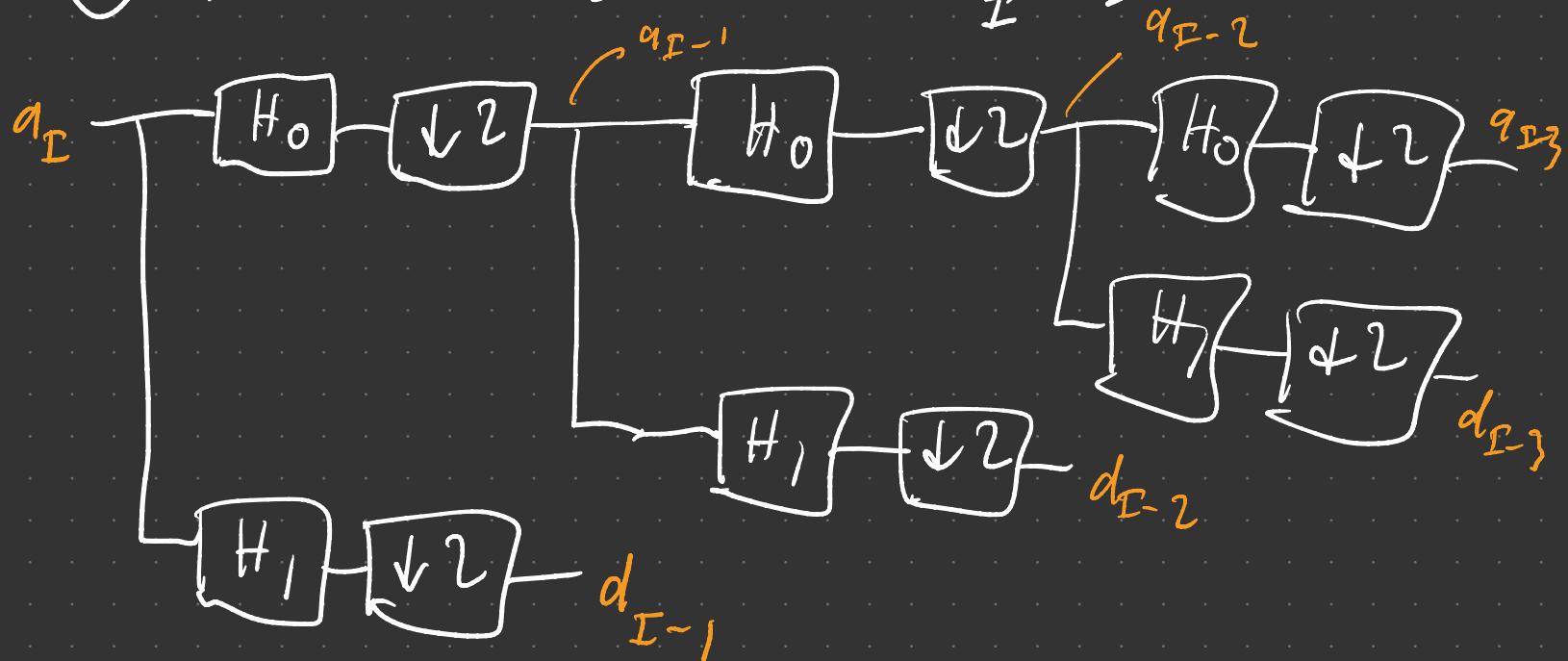
$\psi(t)$ ,  $\varphi(t)$ : scaling and wavelet functions that give rise to a MRA.

## Fundamental assumption:

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

## Pipeline :

- ① Perform a DWT on  $a_I[n]$



→ Iterate as many times as you want

→ The filters come from the scalings and wavelet functions

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

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

- 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,  $\varphi$ , and  $\psi$ .

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

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

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

must converge.  $h[n]$

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

Def<sup>n</sup>:  $x$  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  $F$ .

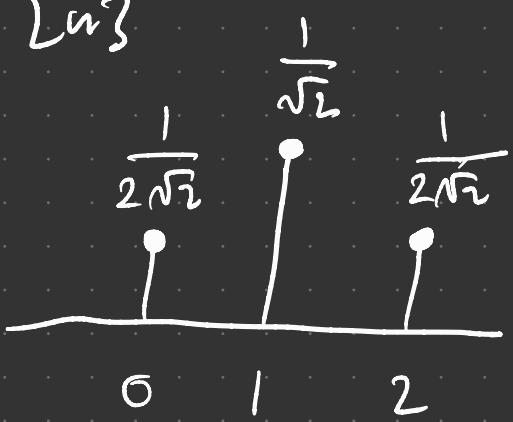
Ex: Does the sequence  $x_{k+1} = \frac{x_k}{2}$  converge? What is its fixed point?

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

This sequence converges to 0 for any choice of  $x_0$ .

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

Exer: h [n3]

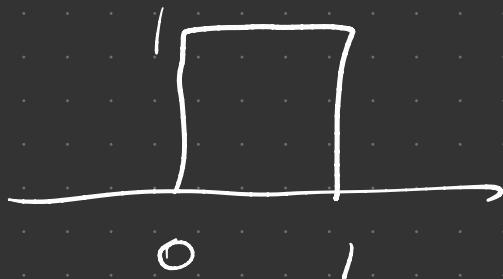


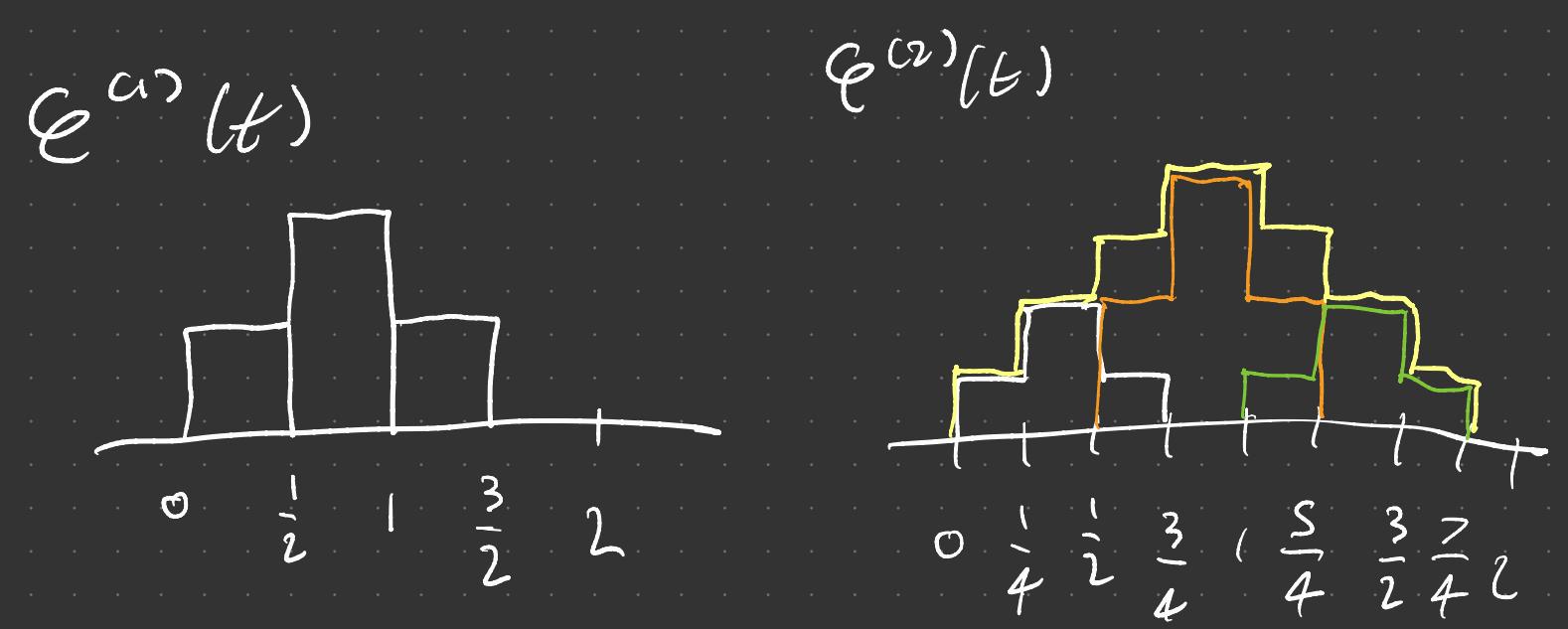
What is the fixed point of the cascade algorithm?

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

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

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



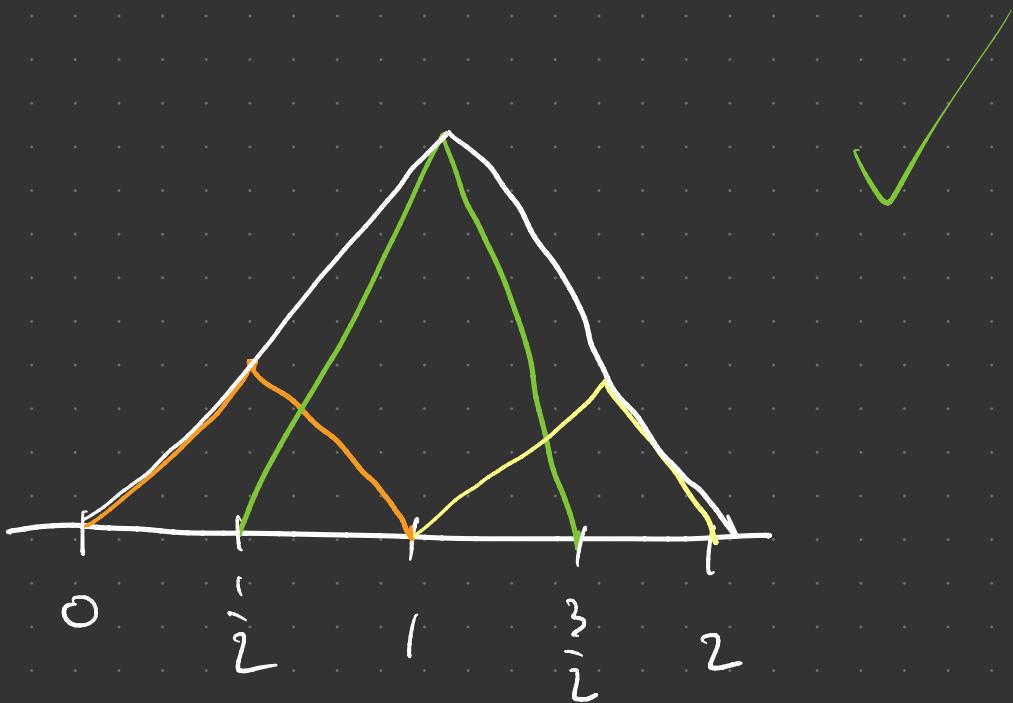


keep iterating ...  $\varphi(t) = \varphi^{(\infty)}(t)$

Exer: Do one more iteration.



Check:



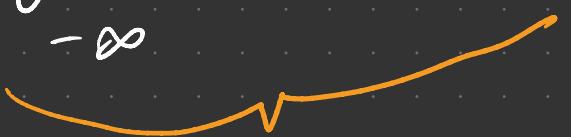
Q: How do we understand this in general?

A: Fourier transform.

$$\mathcal{E}(t) = \sum_{n \in \mathbb{Z}} h[n] \sqrt{2} \mathcal{E}(2t - n)$$



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



$\hat{\mathcal{E}}(\omega)$

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

$$\begin{bmatrix} u = 2t - n \\ du = 2dt \end{bmatrix}$$

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

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

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

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

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

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

⋮

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

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:

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

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

"wavelets from filters" <sup>inverse Fourier transform</sup>

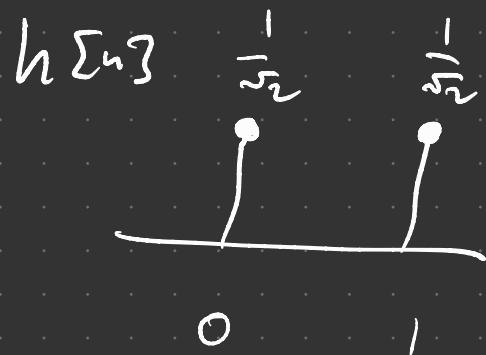
Obs: Convergence of this infinite product is the same as convergence of the cascade algorithm, which is the same as convergence of an infinite-level DWT.

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:  $\epsilon(t)$



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

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

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

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

Check at home

$Z^N$  terms

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

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

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

$$\theta = Z^N \omega$$

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

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

$$\begin{aligned} \text{modulated sinc} &= e^{-j\frac{\omega}{2}} \left( \frac{2 \sin\left(\frac{\omega}{2}\right)}{\omega} \right) \end{aligned}$$

Remark: This recaps Euler's celebrated infinite product formula for the sinc.

Fundamental Theorem of Wavelet Analysis (Mallat, 1987)

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

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

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

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

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  $\varphi$  and filters  $h$ .

Remark:

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

if you allow  
for complex coeffs.

conjugate mirror filters

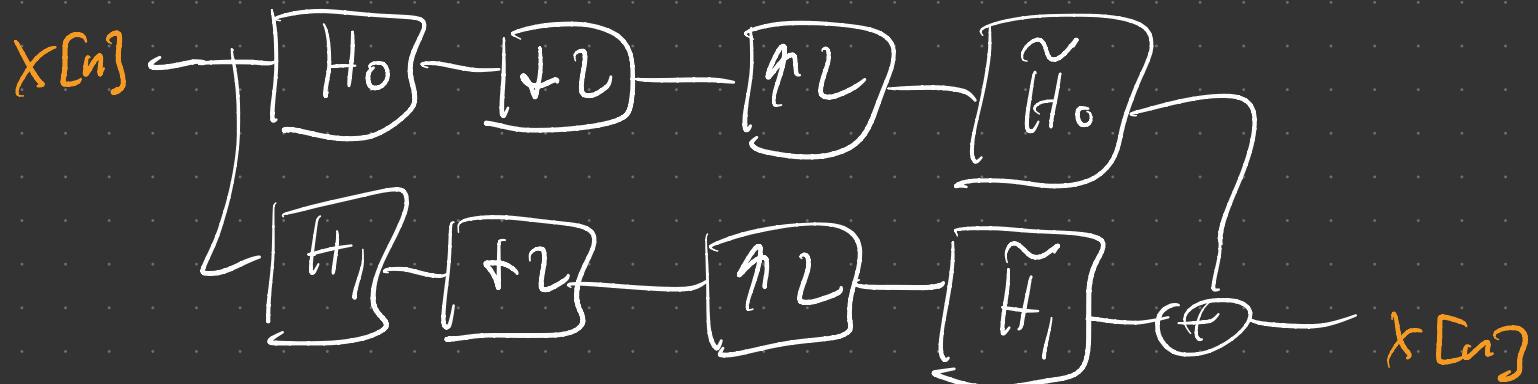
Ex: Haar

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

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

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

$$h_1[n] = \tilde{h}_1[-n]$$



Show that this is a PR FB

with  $L=0$ .

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

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

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

→ This is ① from the fundamental theorem.