

# Last Time: MRA of $L^2(\mathbb{R})$ and Haar Wavelets

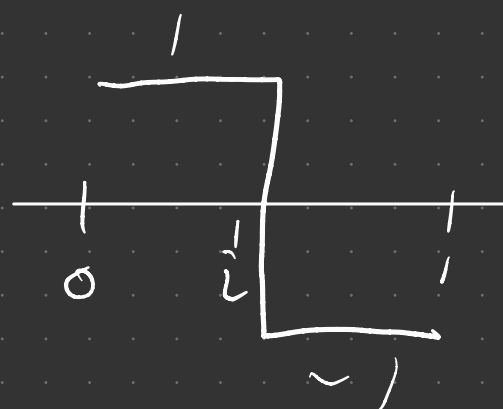
Haar scaling function:

$$\varphi(t) = \begin{cases} 1, & 0 \leq t \leq 1 \\ 0, & \text{else} \end{cases}$$



Haar wavelet function:

$$\psi(t) = \begin{cases} 1, & 0 \leq t \leq \frac{1}{2} \\ -1, & \frac{1}{2} < t \leq 1 \\ 0, & \text{else} \end{cases}$$



Notation:

$$\varphi_{i,n}(t) = 2^{\frac{i}{2}} \varphi(2^i t - n)$$

$$\psi_{i,n}(t) = 2^{\frac{i}{2}} \psi(2^i t - n)$$

translates and dilates of the scaling and wavelet function

Approximation and Wavelet Spaces:

$$V_i = \text{Span} \{ \varphi_{i,n} \}_{n \in \mathbb{Z}}$$

$$W_i = \text{Span} \{ \psi_{i,n} \}_{n \in \mathbb{Z}}$$

The collection  $\{V_i\}_{i \in \mathbb{Z}}$  form a nested sequence of subspaces of  $L^2(\mathbb{R})$

$$\{0\} \subset \dots \subset V_{-1} \subset V_0 \subset V_1 \subset \dots \subset L^2(\mathbb{R})$$

called a multiresolution analysis (MRA).

Recall:  $V_0 \subset V_1$ ,  $W_0 \subset V_1$

$$V_1 = V_0 \oplus W_0$$

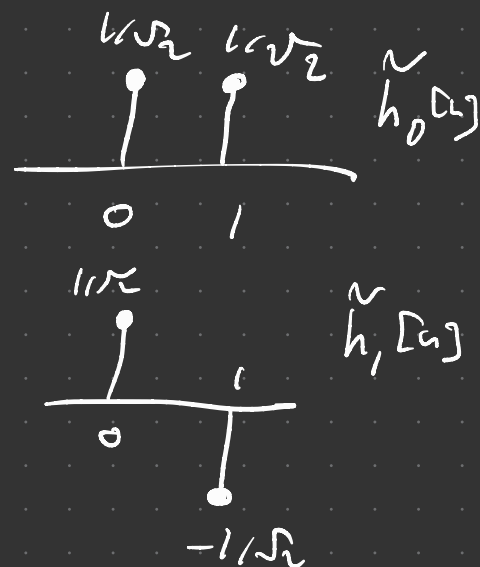
→  $W_0$  is the orthogonal complement of  $V_0$  in  $V_1$ .

$$\rightarrow W_0 = \{f \in V_1 : \langle f, g \rangle = 0, \forall g \in V_0\}$$

Two-Scale Equations:

$$\varphi(t) = \sum_{n \in \mathbb{Z}} \tilde{h}_0[n] \sqrt{2} \varphi(2t - n)$$

$$\psi(t) = \sum_{n \in \mathbb{Z}} \tilde{h}_1[n] \sqrt{2} \varphi(2t - n)$$



Obs:  $\tilde{h}_0[n]$  &  $\tilde{h}_1[n]$  are the Haar FB analysis filters.

Last time, we did this by inspection.

Q: Is there a more direct way to do this?  
(i.e., for other wavelets?)

A: Projections.

$$\begin{aligned} \varphi \in V_0 \subset V_1 &\Rightarrow \varphi(t) = \text{Proj}_{V_1} \varphi(t) \\ &= \sum_{n \in \mathbb{Z}} \underbrace{\langle \varphi, \varphi_{1,n} \rangle}_{\tilde{h}_0[n]} \varphi_{1,n}(t) \end{aligned}$$

$$\begin{aligned} \psi \in W_0 \subset V_1 &\Rightarrow \psi(t) = \text{Proj}_{V_1} \psi(t) \\ &= \sum_{n \in \mathbb{Z}} \underbrace{\langle \psi, \varphi_{1,n} \rangle}_{\tilde{h}_1[n]} \varphi_{1,n}(t) \end{aligned}$$

General: Given arbitrary scaling and wavelet function  $\varphi$  and  $\psi$ , we can derive two filters:

conjugate  
mirror  
filters

$$\begin{cases} \text{Low-pass} : \tilde{h}_0[n] = \langle \varphi, \varphi_{1,n} \rangle \\ \text{High-pass} : \tilde{h}_1[n] = \langle \psi, \varphi_{1,n} \rangle \end{cases}$$

## Last Time:

$$V_1 = V_0 \oplus W_0$$

$$V_2 = V_1 \oplus W_1 = V_0 \oplus W_0 \oplus W_1$$

⋮

$$V_i = V_0 \oplus W_1 \oplus \dots \oplus W_{i-1}$$

⋮

$$L^2(\mathbb{R}) = V_0 \oplus \left( \bigoplus_{i=0}^{\infty} W_i \right)$$

Obs:  $\underbrace{\{ \varphi_{0,n} \}_{n \in \mathbb{Z}}}_{V_0} \cup \bigcup_{i=0}^{\infty} \underbrace{\{ \psi_{i,n} \}_{n \in \mathbb{Z}}}_{W_i}$

is an ortho basis for  $L^2(\mathbb{R})$ .

Equivalently,

$$\{ \varphi_{0,n} \}_{n \in \mathbb{Z}} \cup \{ \varphi_{i,n} \}_{i \geq 0, n \in \mathbb{Z}}$$

is an ortho basis for  $L^2(\mathbb{R})$ .



Q: What does this mean for  $f \in L^2(\mathbb{R})$ ?

A:  $f$  can be written as the sum of its projections onto  $V_0, W_0, W_1, W_2, \dots$

$$f(t) = \text{Proj}_{V_0} f(t) + \sum_{i=0}^{\infty} \text{Proj}_{W_i} f(t).$$

$$\underbrace{\text{Proj}_{V_0} f(t)}_{\text{coarse approximation}} = \sum_{n \in \mathbb{Z}} \underbrace{\langle f, \phi_{0,n} \rangle}_{a_0[n]} \phi_{0,n}(t)$$

approx. coeffs.

$$\underbrace{\text{Proj}_{W_i} f(t)}_{\text{details}} = \sum_{n \in \mathbb{Z}} \underbrace{\langle f, \psi_{i,n} \rangle}_{d_i[n]} \psi_{i,n}(t)$$

detail coeffs.

$$f(t) = \sum_{n \in \mathbb{Z}} a_0[n] \phi_{0,n}(t) + \sum_{i=1}^{\infty} \sum_{n \in \mathbb{Z}} d_i[n] \psi_{i,n}(t)$$

Synthesis      Procedure

# Analysis Procedure

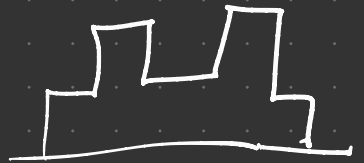
Ex: finest resolution is  $i=3$ :

$$a_3[n] = \langle f_3, e_{3,n} \rangle$$



$$f_3(t) = \sum_{n \in \mathbb{Z}} a_3[n] e_{3,n}(t)$$

$V_3$



How do we get  $a_2[n]$  &  $d_2[n]$  from  $a_3[n]$ ?

$$V_3 = V_2 \oplus W_2$$

↑ ↑ project onto these subspaces

Project onto  $V_2$

$$\text{Proj}_{V_2} f_3(t) = f_2(t) = \sum_{m \in \mathbb{Z}} \langle f_3, e_{2,m} \rangle e_{2,m}(t)$$

$$= \sum_{m \in \mathbb{Z}} \left\langle \sum_{n \in \mathbb{Z}} a_3[n] e_{3,n}, e_{2,m} \right\rangle e_{2,m}(t)$$

$$= \sum_{m \in \mathbb{Z}} \left[ \sum_{n \in \mathbb{Z}} a_3[n] \langle e_{3,n}, e_{2,m} \rangle \right] e_{2,m}(t)$$

$a_2[m]$

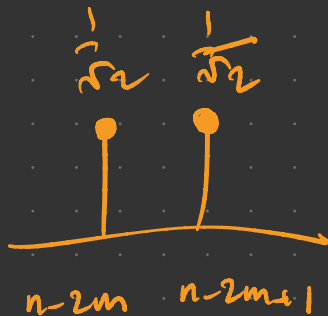
$$\langle e_{3,n}, e_{2,m} \rangle = \int_{-\infty}^{\infty} 2^{\frac{3}{2}} e(2^3 t - n) 2 e(2^2 t - m) dt$$

$$\left[ \begin{array}{l} u = 2^3 t \\ du = 2^3 dt \end{array} \right]$$

$$= \frac{2^{\frac{5}{2}}}{2^3} \int_{-\infty}^{\infty} e(u-n) e\left(\frac{u}{2} - m\right) du$$

$$\left[ \begin{array}{l} t = u - 2m \\ dt = du \end{array} \right]$$

$$= \frac{1}{\sqrt{2}} \int_{-\infty}^{\infty} e(t - (n - 2m)) e\left(\frac{t}{2}\right) dt$$



this integral is zero unless  
 $n-2m = 0$  or  $n-2m = 1$   
 in which case it is 1.

$$= \tilde{h}_0[n-2m]$$

$$f_2(t) = \sum_{m \in \mathbb{Z}} \left( \underbrace{\sum_{n \in \mathbb{Z}} a_3[n] \tilde{h}_0[n-2m]}_{a_2[m]} \right) \psi_{2,m}(t)$$



Define  $h_0[k] = \tilde{h}_0[-k]$  *time reversal*

$$\Rightarrow a_2[m] = \sum_{n \in \mathbb{Z}} a_3[n] h_0[2m-n]$$

$$= (a_3 * h_0)[2m]$$

*downsampling*



Project onto  $W_2$

$$\text{Proj}_{W_2} f_3(t) = g_2(t) = \sum_{m \in \mathbb{Z}} \langle f_3, \psi_{2,m} \rangle \psi_{2,m}(t)$$

$$= \sum_{m \in \mathbb{Z}} \left\langle \sum_{n \in \mathbb{Z}} a_3[n] \psi_{3,n}, \psi_{2,m} \right\rangle \psi_{2,m}(t)$$

$$= \sum_{m \in \mathbb{Z}} \left[ \sum_{n \in \mathbb{Z}} a_3 [n] \langle e_{3,n}, \Psi_{2,m} \rangle \right] \Psi_{2,m}(t)$$

$$\langle e_{3,n}, \Psi_{2,m} \rangle = \int_{-\infty}^{\infty} 2^{\frac{3}{2}} \varphi(2^3 t - n) 2 \Psi(2^2 t - m) dt$$

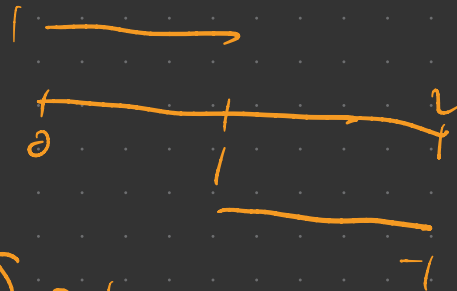
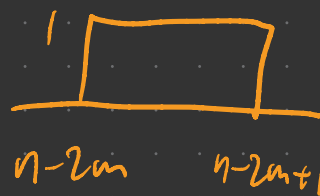
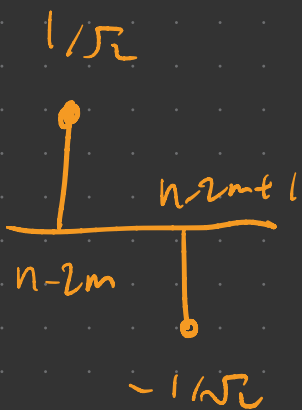
$d_2[m]$

$$\left[ \begin{array}{l} u = 2^3 t \\ du = 2^3 dt \end{array} \right]$$

$$= \frac{2^{\frac{5}{2}}}{2^3} \int_{-\infty}^{\infty} \varphi(u - n) \Psi\left(\frac{u}{2} - m\right) dt$$

$$\left[ \begin{array}{l} t = u - 2m \\ dt = du \end{array} \right]$$

$$= \frac{1}{\sqrt{2}} \int_{-\infty}^{\infty} \varphi(t - (n - 2m)) \Psi\left(\frac{t}{2}\right) dt$$

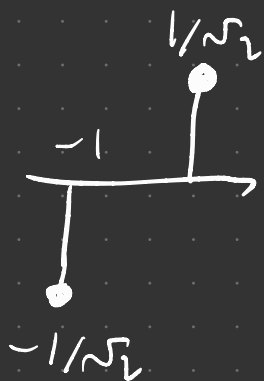


$$n - 2m = 0 \Rightarrow \int = 1$$

$$n - 2m = 1 \Rightarrow \int = -1$$

$$= \tilde{h}_1[n - 2m]$$

$$g_2(t) = \sum_{m \in \mathbb{Z}} \underbrace{\left( \sum_{n \in \mathbb{Z}} a_3[n] \tilde{h}_1[n-2m] \right)}_{d_2[m]} \psi_{2,m}(t)$$



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

$$\Rightarrow d_2[m] = \sum_{n \in \mathbb{Z}} a_3[n] h_1[2m-n]$$

$$= (a_3 * h_1)[2m]$$

downsampling



Obs: Given an approximation  $a_i[n]$ ,  
we can get all other coefficients  
 $a_k[n]$ ,  $k \leq i$ , and  $d_l[n]$ ,  $l \leq i$ ,  
Extremely fast: Fast Wavelet Transform

concrete implementation: DWT

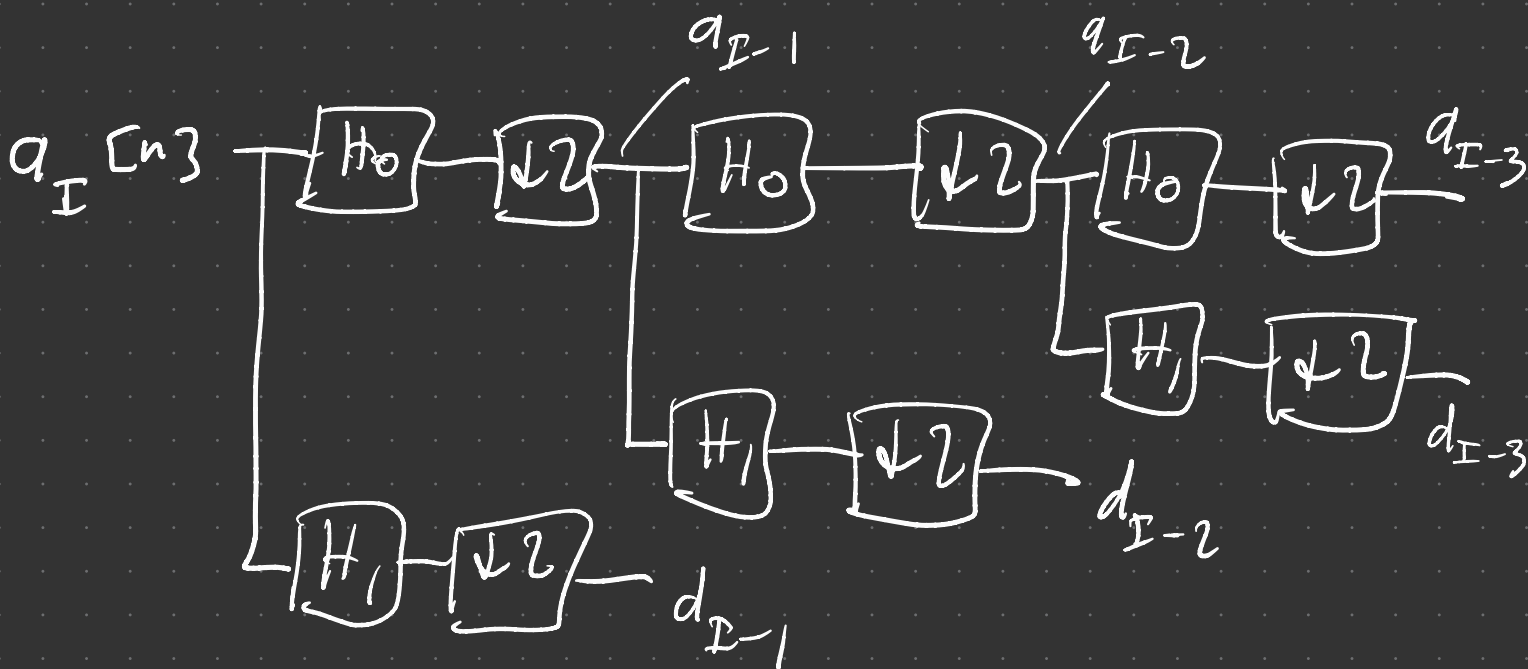
# General DWT Pipeline:

$f(t)$ : underlying analog signal

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

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

• Perform a DWT:



- Do some (nonlinear) processing on, e.g.,  
 $d_{I-3}, d_{I-3}, d_{I-2}, d_{I-1}$
- Perform an IDWT on the processed signals,  
 resulting in  $\hat{q}_I[n]$
- Processed analog signal:

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

"Think analog, act digital!"

MRA

DWT