

# Generalized Sampling

Given an analog signal  $f: \mathbb{R} \rightarrow \mathbb{R}$ ,

We can construct a discrete approximation

$$a[n] = \langle f, e_n \rangle = \int_{-\infty}^{\infty} f(t) \underbrace{e_n(t)}_{\mathcal{E}(t-n)} dt$$

n \in \mathbb{Z}

- $e$  models the impulse response of the acquisition system.

Q: What's the simplest choice of  $\epsilon$ ?

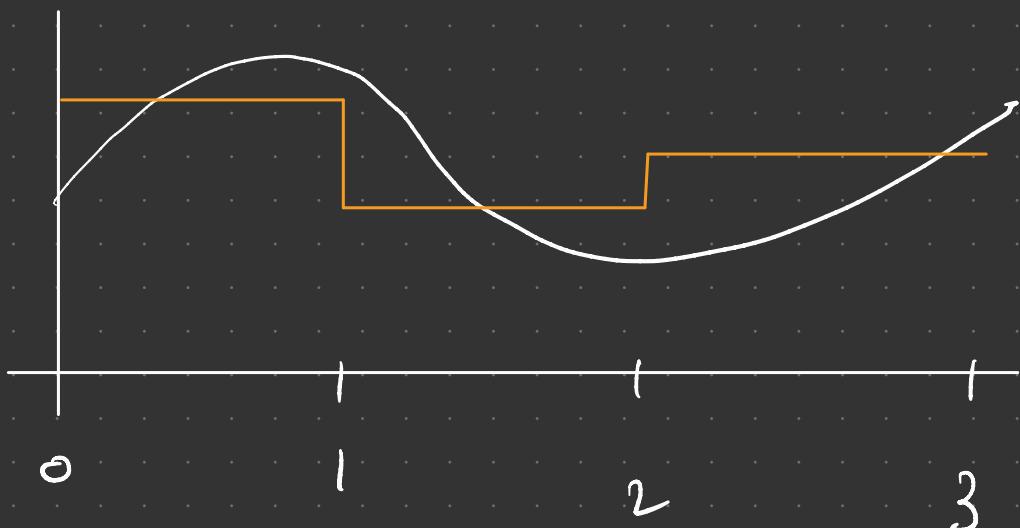
A: Box (or rect) function:

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



Q: What happens if we sample an analog signal with  $\{\epsilon(t-n)\}_{n \in \mathbb{Z}}$ ?

A: Piecewise constant approx.

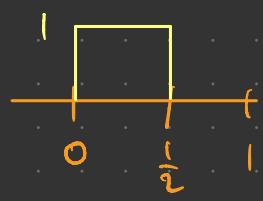


Q: What if we want a higher resolution?

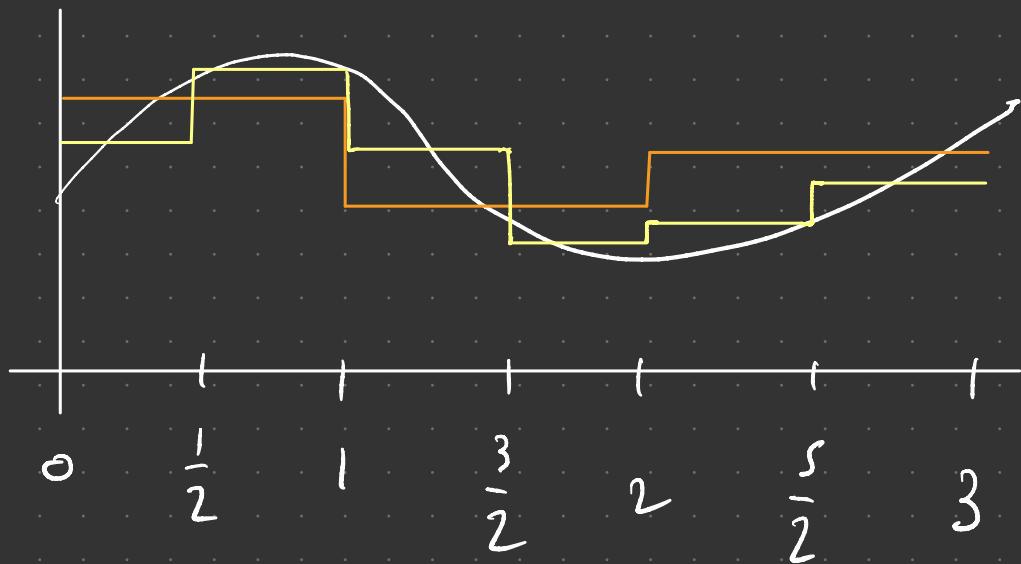
How do we double the resolution?

$i=1$

A: Sample with  $\{\epsilon(2t-n)\}_{n \in \mathbb{Z}}$

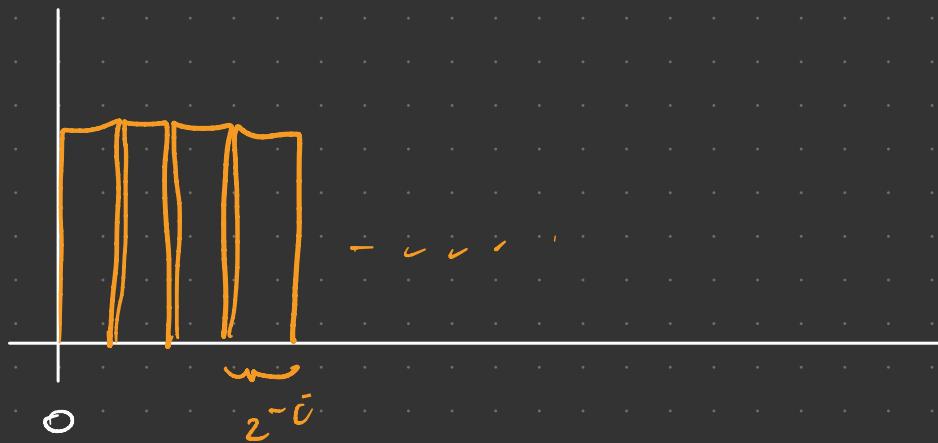


We get a higher-resolution approximation.



Obs: If we keep doubling the resolution,  
we can recover the original  
analog signal !!!

Remark: For a general resolution  $i$ ,  
we sample with  $\{e(2^i t - n)\}_{n \in \mathbb{Z}}$



Q: Is there a problem?

What if  $i \rightarrow \infty$ ?

$\varphi(2^i t - n) \rightarrow 0$  for almost every  $t$

Fix: Normalize the sampling functions

to have unit energy:

$$\left\{ 2^{\frac{i}{2}} \varphi(2^i t - n) \right\}_{n \in \mathbb{Z}, i \in \mathbb{Z}}$$

$$\int_{-\infty}^{\infty} \left[ 2^{\frac{i}{2}} \varphi(2^i t - n) \right]^2 dt$$

$$= 2^i \int_{-\infty}^{\infty} \varphi(2^i t - n)^2 dt \quad u = 2^i t - n \\ du = 2^i dt$$

$$= \underbrace{\int_{-\infty}^{\infty} \varphi(u)^2 du}_v = 1$$

"Lebesgue space"  $\|\varphi\|_{L^2}^2 = \langle \varphi, \varphi \rangle$

Def<sup>n</sup>: The space of finite-energy signals is

$$L^2(\mathbb{R}) = \{ f : \mathbb{R} \rightarrow \mathbb{R} : \|f\|_{L^2} < \infty \}$$

# PR FBs vs. DWTs

PR FB:



DWT:



- Iterated structure
- Multiscale decomposition of the input signal.

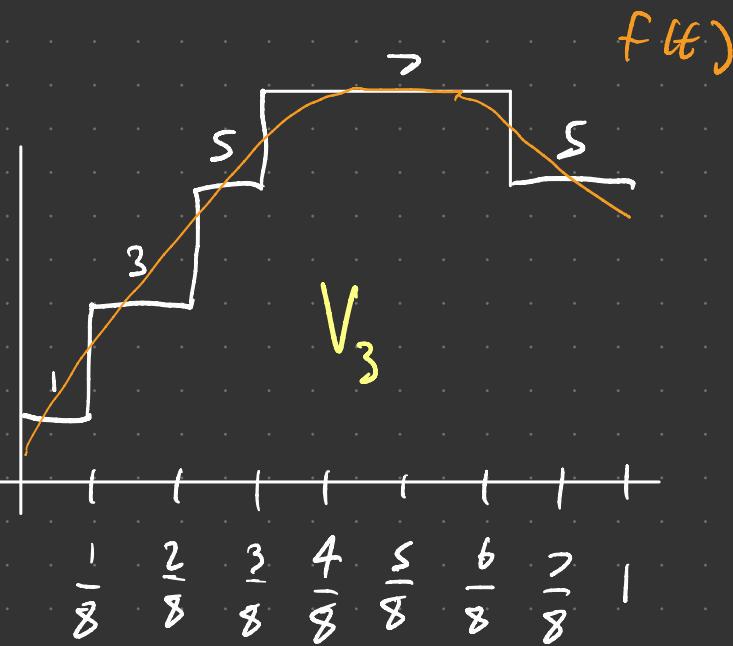
Remark: All of this processing is in the discrete domain.

→ There is an underlying sampling procedure.

Ex: Suppose we have a resolution 3 approx. of an analog signal.

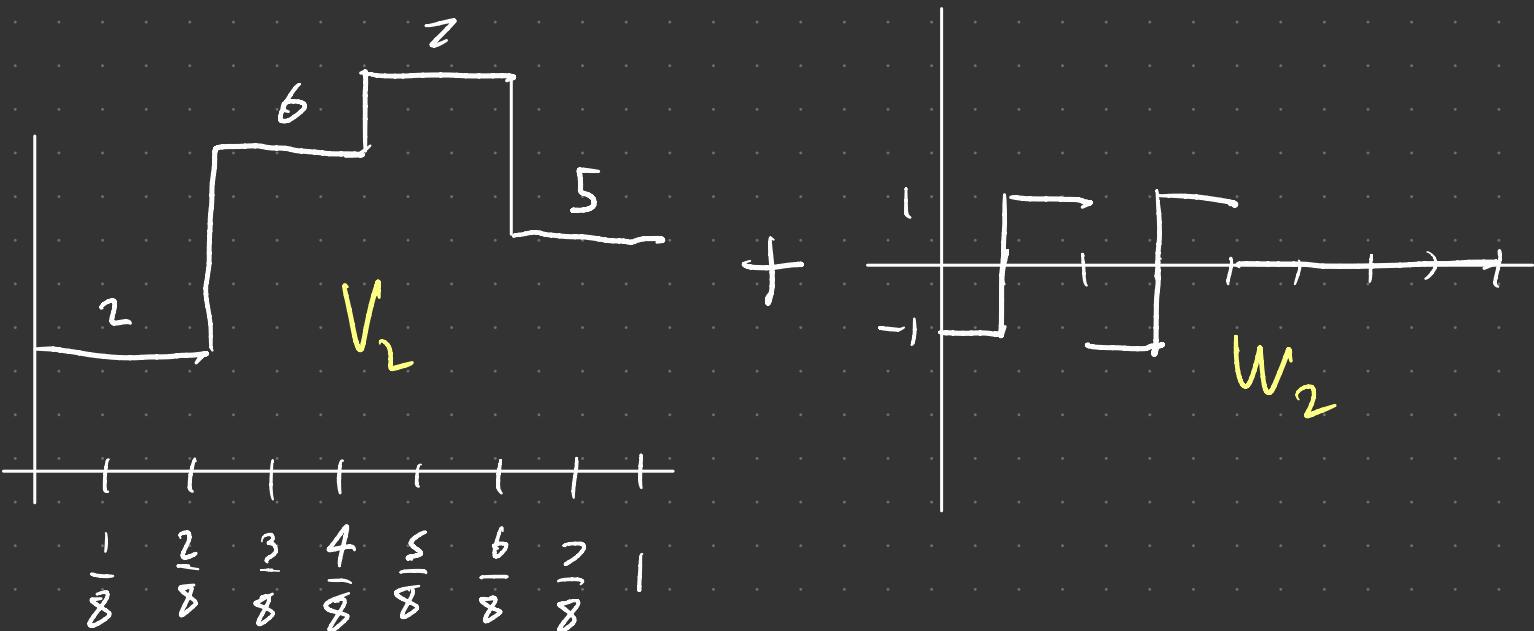
Q: What is the bin width?

A:  $2^{-3} = \frac{1}{8}$



$i = 3$  approximation

Basis:  $\left\{ 2^{\frac{3}{2}} \varphi(2^3 t - n) \right\}_{n \in \mathbb{Z}}$

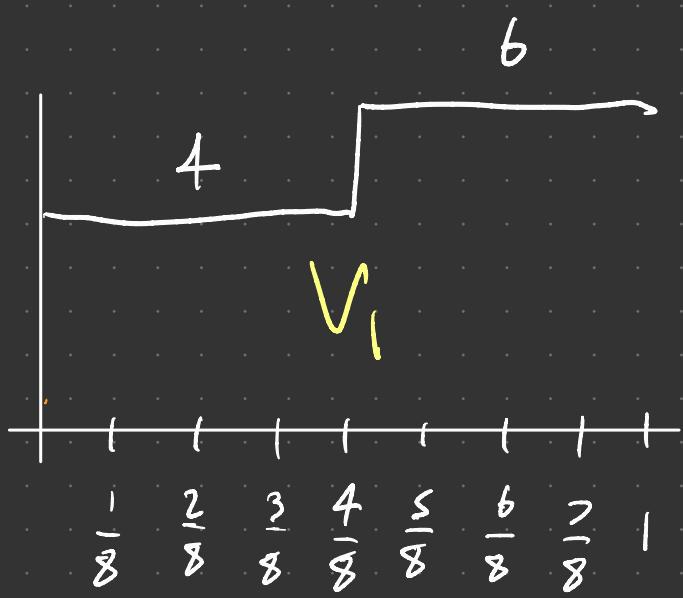


$i = 2$  approximation

Basis:  $\left\{ 2 \varphi(2^2 t - n) \right\}_{n \in \mathbb{Z}}$

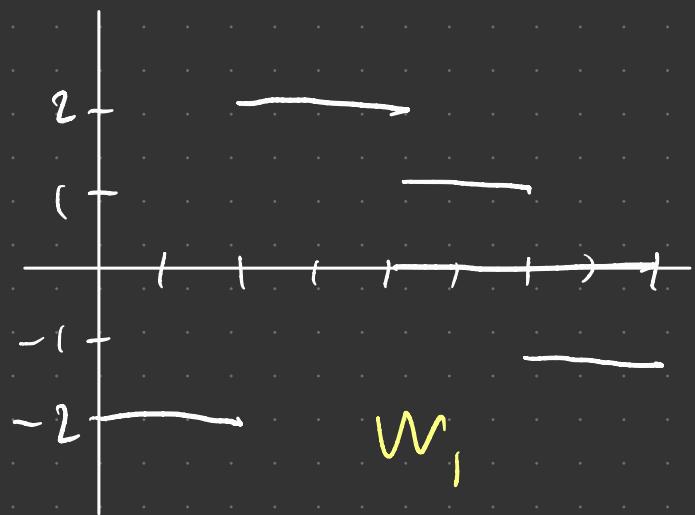
$i = 2$  detail

Basis:  $\left\{ 2 \psi(2^2 t - n) \right\}_{n \in \mathbb{Z}}$



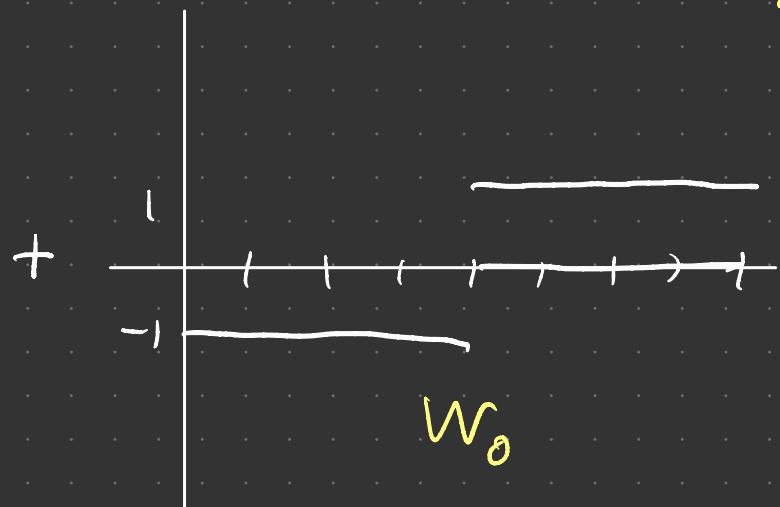
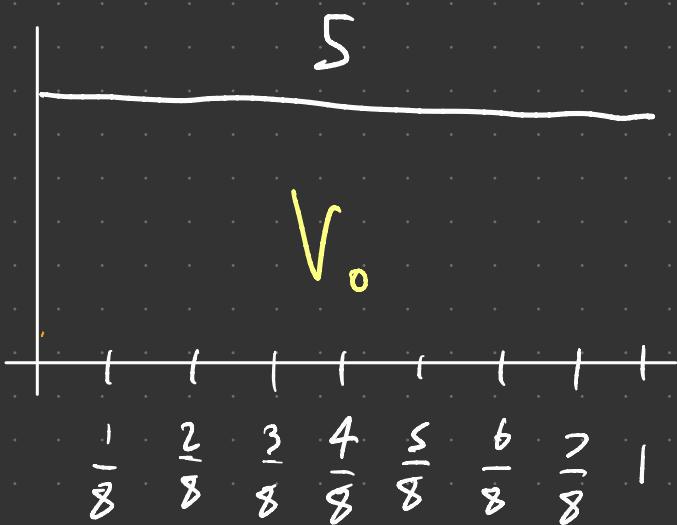
$i=1$  approximation

Basis:  $\{2^{n/2} \varphi(2t-n)\}_{n \in \mathbb{Z}}$



$i=1$  detail

Basis:  $\{2^{n/2} \psi(2t-n)\}_{n \in \mathbb{Z}}$



$i=0$  approximation

Basis:  $\{\varphi(t-n)\}_{n \in \mathbb{Z}}$

$i=0$  detail

Basis:  $\{\psi(t-n)\}_{n \in \mathbb{Z}}$

Obs: This is a multiresolution decomp.

Obs: The approximation bases are multiscale versions of  $\varphi(t)$ .

Q: What basis functions are we using  
for the details?

A: Multiscale versions of

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



Remark:  $\Psi(t)$  is the Haar (mother) wavelet  
 $\varphi(t)$  is the Haar scaling function  
(father wavelet)

Def<sup>n</sup>: The approximation space  
at resolution i is the space

$$V_i = \overline{\text{span}} \left\{ 2^{\frac{i}{2}} \varphi(2^i t - n) \right\}_{n \in \mathbb{Z}}$$

Def<sup>i</sup>: The wavelet (detail) Space  
at resolution i is the space

$$W_i = \overline{\text{span}} \left\{ 2^{\frac{i}{2}} \Psi(2^i t - n) \right\}_{n \in \mathbb{Z}}$$

Q: Which is bigger:  $V_0$  or  $V_1$ ?

A:  $V_0 \subset V_1$

Q: Which is bigger:  $W_0$  or  $V_1$ ?

A:  $W_0 \subset V_1$

What does this mean?

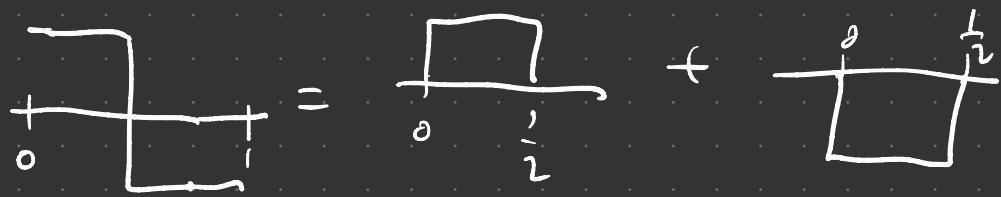
$$\underbrace{\varphi(t)}_{V_0} = \sum_{n \in \mathbb{Z}} g_0[n] \underbrace{\sqrt{2} \varphi(2t-n)}_{V_1}$$

$$\begin{array}{c} \text{square wave} \\ 0 \quad 1 \end{array} = \begin{array}{c} \text{square wave} \\ 0 \quad \frac{1}{2} \end{array} + \begin{array}{c} \text{square wave} \\ \frac{1}{2} \quad 1 \end{array}$$

$$g_0[n] \quad \begin{array}{c} \frac{1}{\sqrt{2}} \\ \frac{1}{\sqrt{2}} \\ \hline 0 \quad 1 \end{array}$$

Haar FB  
low-pass  
filter

$$\bullet \Psi(t) = \sum_{n \in \mathbb{Z}} g_1[n] \sqrt{2} e^{j(2t-n)}$$



$$g_1[n] \frac{1}{\sqrt{2}}$$

Haar FB  
high-pass  
filter

Obs:  $g_0[n]$  &  $g_1[n]$  are the Haar wavelet filters!

Remark: These equations are called the two-scale equations

Q: How do we get  $V_1$  from  $V_0$  &  $W_0$ ?

A:  $V_1 = V_0 \oplus W_0$ .

↳ direct sum

- $V_i = \{f+g : f \in V_0 \text{ and } g \in W_0\}$  "sum"
- $V_0 \cap W_0 = \{\emptyset\}$  "direct"

Q: What happens to  $V_i$  as  $i$  becomes large?

A:  $V_i \rightarrow L^2(\mathbb{R})$  as  $i \rightarrow \infty$

Q: What happens to  $V_i$  as  $i$  becomes small?

A:  $V_i \rightarrow \{\emptyset\}$  as  $i \rightarrow -\infty$

Obs: We have constructed a nested sequence of subspaces of  $L^2(\mathbb{R})$ :

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

Remark: This sequence is called a multi-resolution analysis (MRA) of  $L^2(\mathbb{R})$ .