

# EE269

## Signal Processing for Machine Learning

**Wavelets, Discrete Wavelet Transform and Short-Time Fourier Transform**

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Stanford University

# Continuous Wavelet Transform

- ▶ Define a function  $\psi(t)$
- ▶ Create scaled and shifted versions of  $\psi(t)$

$$\psi_{s,\tau} = \frac{1}{\sqrt{s}}\psi\left(\frac{t-\tau}{s}\right)$$

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$$W(s, \tau) = \int_{-\infty}^{\infty} f(t)\psi_{s,\tau}^* dt = \langle f(t), \psi_{s,\tau} \rangle$$

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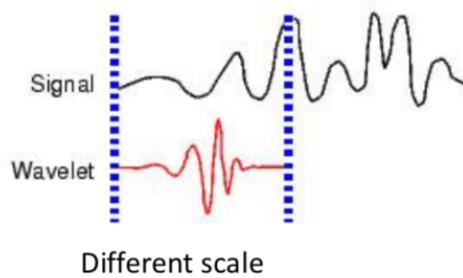
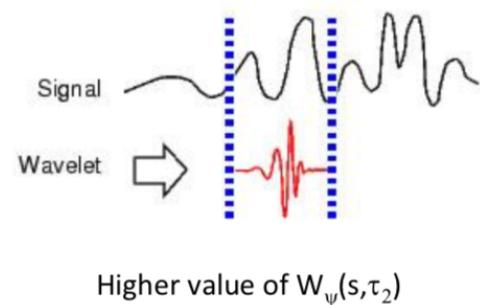
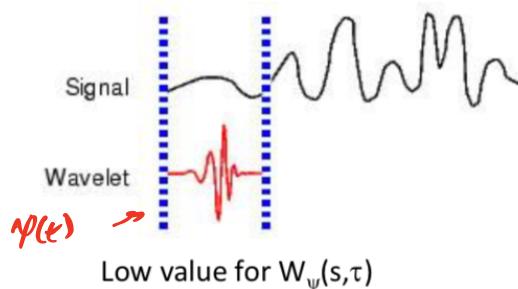
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- ▶ Transforms a continuous function of one variable into a continuous function of two variables : **translation** and **scale**
- ▶ For a compact representation, we can choose a mother wavelet  $\psi(t)$  that matches the signal shape
- ▶ Inverse Wavelet Transform

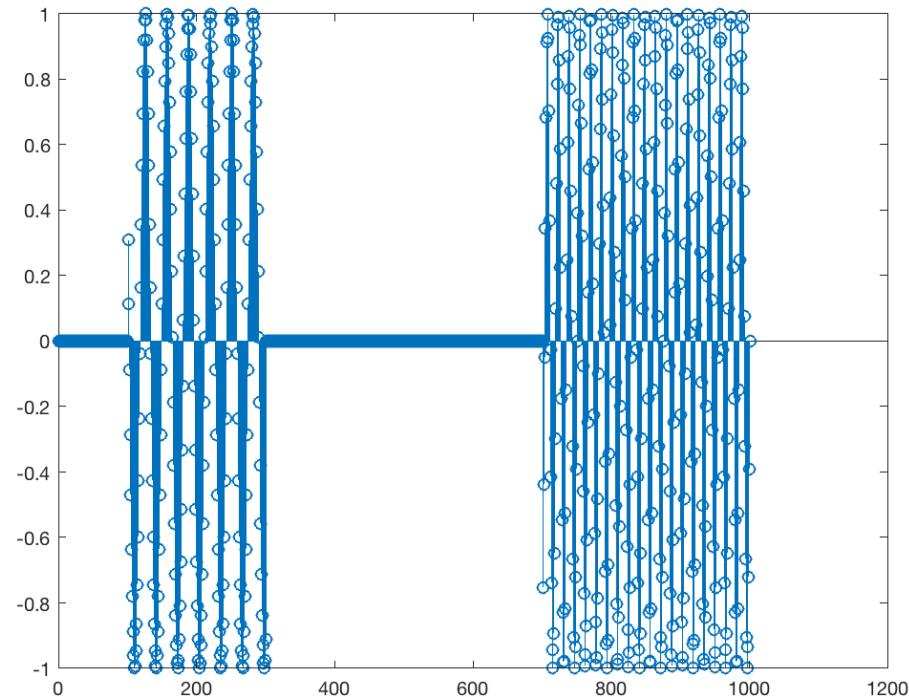
$$f(t) = \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} W(s, \tau) \psi_{s,\tau} d\tau ds$$

# Wavelets



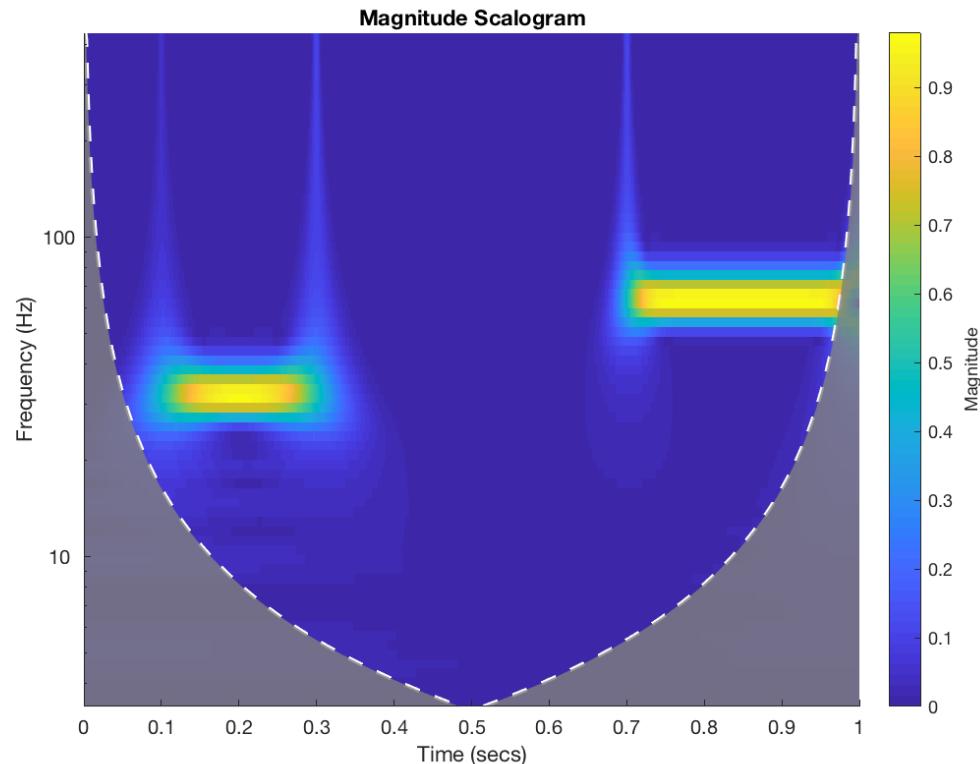
# Continuous Wavelet Transform

```
Fs = 1e3;  
t = 0:1/Fs:1;  
x = cos(2*pi*32*t).*(t>=0.1 & t<0.3) + sin(2*pi*64*t).*(t>0.7);  
stem(x);
```



# Continuous Wavelet Transform

`cwt(x,1000)`

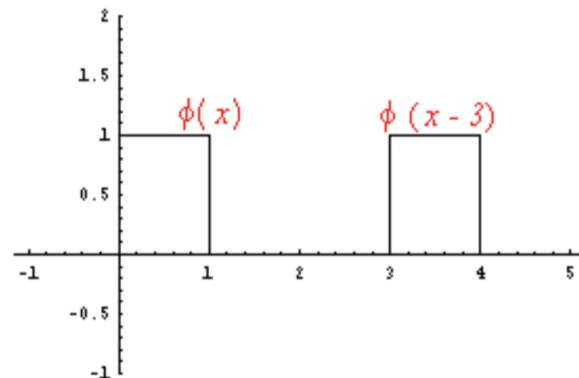


# Continuous Haar Wavelets

- ▶ Consider the function

$$\phi(x) = \begin{cases} 1 & \text{if } 0 \leq x \leq 1 \\ 0 & \text{otherwise} \end{cases}$$

- ▶ translates  $\phi(x - k)$

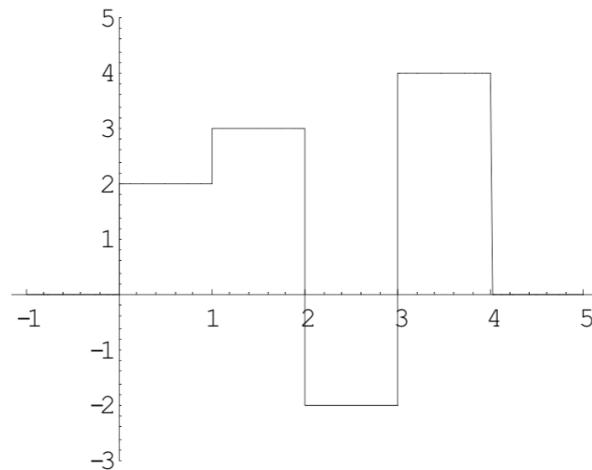


# Continuous Haar Wavelets

- ▶ Consider the function

$$\phi(x) = \begin{cases} 1 & \text{if } 0 \leq x \leq 1 \\ 0 & \text{otherwise} \end{cases}$$

- ▶ linear combination of the translates  $\phi(x - k)$



# Continuous Haar Wavelets

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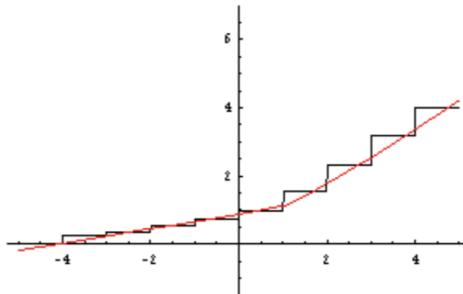
$$\phi(x) = \begin{cases} 1 & \text{if } 0 \leq x \leq 1 \\ 0 & \text{otherwise} \end{cases}$$

- ▶ Define

$V_0$  = all square integrable functions of the form

$$g(x) = \sum_k a_k \phi(x - k)$$

=all square integrable functions which are constant on integer intervals



# Continuous Haar Wavelets

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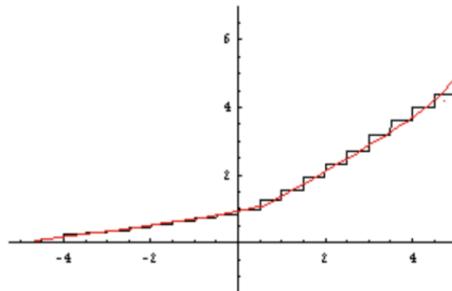
$$\phi(x) = \begin{cases} 1 & \text{if } 0 \leq x \leq 1 \\ 0 & \text{otherwise} \end{cases}$$

- ▶ Define

$V_1$  = all square integrable functions of the form

$$g(x) = \sum_k a_k \phi(2x - k)$$

=all square integrable functions which are constant on half integer intervals



# Continuous Haar Wavelets

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- ▶ Define

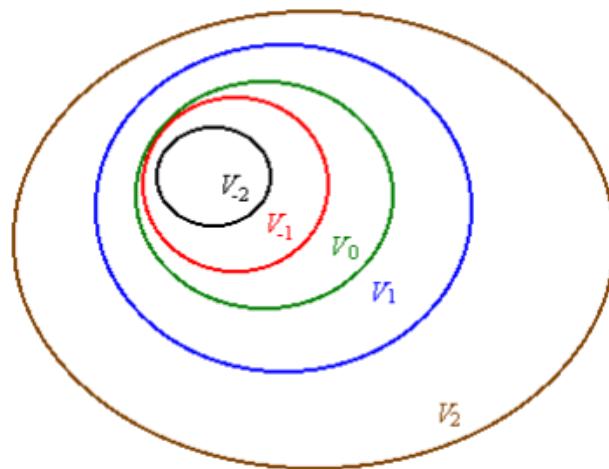
$V_j$  = all square integrable functions of the form

$$g(x) = \sum_k a_k \phi(2^j x - k)$$

=all square integrable functions which are constant on  $2^{-j}$  length intervals

# Continuous Haar Wavelets

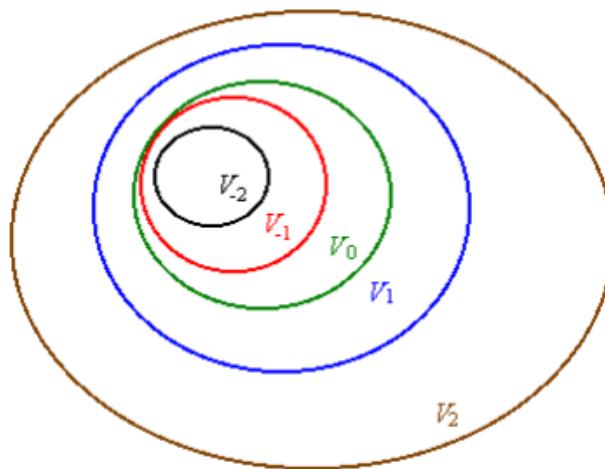
- ▶ Nested spaces:  $\dots V_{-2} \subset V_{-1} \subset V_0 \subset V_1 \subset V_2 \dots$



- ▶ There is a subspace  $W_0$  such that  $V_0 \oplus W_0 = V_1$ , i.e.  
 $W_0 := V_1 \ominus V_0$

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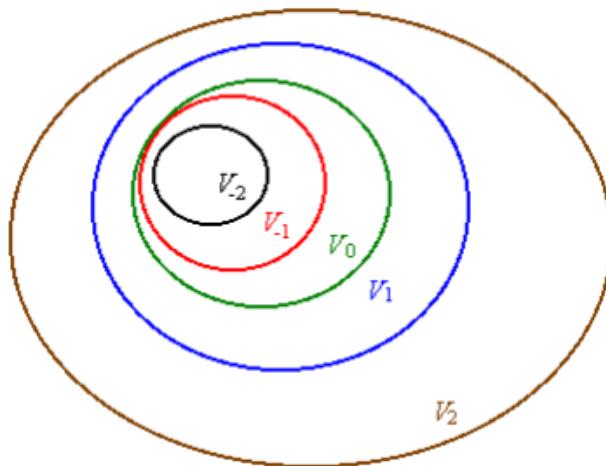
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$$W_0 := V_1 \ominus V_0$$
- ▶ In general  $W_j = V_{j+1} \ominus V_j$

# Continuous Haar Wavelets

- Nested spaces: ... $V_{-2} \subset V_{-1} \subset V_0 \subset V_1 \subset V_2 \dots$



- $W_j = V_{j+1} \ominus V_j$
- **Theorem:** Every square integrable function can be uniquely expressed as

$$\sum_{j=-\infty}^{\infty} w_j \quad \text{where } w_j \in W_j$$

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- ▶  $\left\{2^{j/2}\psi(2^j x - k)\right\}_{k=-\infty}^{\infty}$  forms an orthonormal basis for  $W_j$

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- ▶  $f = \sum_j \sum_j a_{jk} \psi_{jk}(x)$  (multiresolution analysis)

# Discrete Wavelet Transform

- ▶ Discrete shifts and scales  $\psi(\frac{t-\tau}{s})$
- ▶ Suppose we have a signal of length N

$$x = [x_1, x_2, \dots x_N]$$

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$$x_k = \frac{x_{2k-1} + x_{2k}}{2}, \quad k = 1, \dots, N/2$$

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- ▶ example

$$x = [6, 12, 15, 15, 14, 12, 120, 116] \rightarrow s = [9, 15, 13, 118]$$

- ▶ suppose that we are allowed to send  $N/2$  more numbers
- ▶ differences

$$d_k = \frac{x_{2k-1} - x_{2k}}{2}, \quad k = 1, \dots, N/2$$

- ▶ we can recover  $x$

$$\begin{aligned}x &= [6, 12, 15, 15, 14, 12, 120, 116] \rightarrow \\[s \mid d] &= [9, 15, 13, 118 \mid 3, 0, -1, -2]\end{aligned}$$

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- ▶ One step Haar Transformation  $x \rightarrow [s|d]$

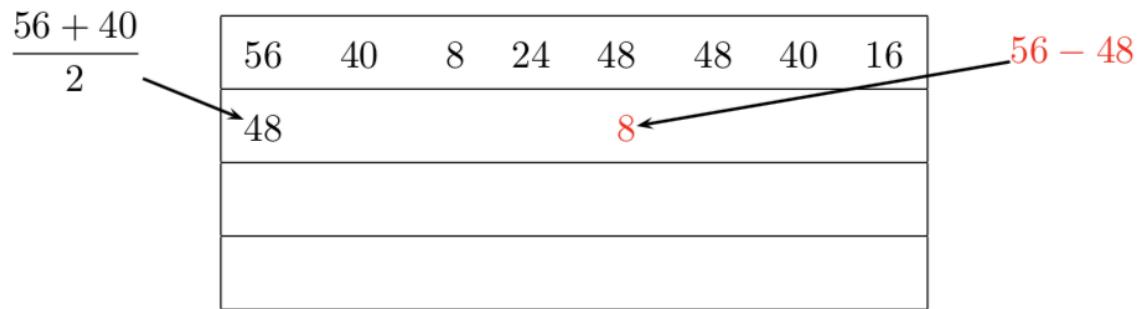
# One Step Haar Transformation

$$\begin{bmatrix} \frac{1}{2} & \frac{1}{2} & 0 & 0 & 0 & 0 & 0 & 0 \\ 0 & 0 & \frac{1}{2} & \frac{1}{2} & 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 & \frac{1}{2} & \frac{1}{2} & 0 & 0 \\ 0 & 0 & 0 & 0 & 0 & 0 & \frac{1}{2} & \frac{1}{2} \\ \hline -\frac{1}{2} & \frac{1}{2} & 0 & 0 & 0 & 0 & 0 & 0 \\ 0 & 0 & -\frac{1}{2} & \frac{1}{2} & 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 & -\frac{1}{2} & \frac{1}{2} & 0 & 0 \\ 0 & 0 & 0 & 0 & 0 & 0 & -\frac{1}{2} & \frac{1}{2} \end{bmatrix} \cdot \begin{bmatrix} x_1 \\ x_2 \\ x_3 \\ x_4 \\ x_5 \\ x_6 \\ x_7 \\ x_8 \end{bmatrix} = \frac{1}{2} \begin{bmatrix} x_1 + x_2 \\ x_3 + x_4 \\ x_5 + x_6 \\ x_7 + x_8 \\ \hline x_2 - x_1 \\ x_4 - x_3 \\ x_6 - x_5 \\ x_8 - x_7 \end{bmatrix}$$

# Discrete Haar Wavelet Transform

56	40	8	24	48	48	40	16

# Discrete Haar Wavelet Transform



# Discrete Haar Wavelet Transform

56	40	8	24	48	48	40	16
48	16			8	-8		

# Discrete Haar Wavelet Transform

56	40	8	24	48	48	40	16
48	16	48		8	-8	0	

# Discrete Haar Wavelet Transform

56	40	8	24	48	48	40	16
48	16	48	28	8	-8	0	12

# Discrete Haar Wavelet Transform

56	40	8	24	48	48	40	16
48	16	48	28	8	-8	0	12
				8	-8	0	12

## Repeating the same process on the averages

56	40	8	24	48	48	40	16
48	16	48	28	8	-8	0	12
32		16		8	-8	0	12

# Discrete Haar Wavelet Transform

56	40	8	24	48	48	40	16
48	16	48	28	8	-8	0	12
32	38	16	10	8	-8	0	12

# Discrete Haar Wavelet Transform

56	40	8	24	48	48	40	16
48	16	48	28	8	-8	0	12
32	38	16	10	8	-8	0	12
		16	10	8	-8	0	12

# Discrete Haar Wavelet Transform

56	40	8	24	48	48	40	16
48	16	48	28	8	-8	0	12
32	38	16	10	8	-8	0	12
35	-3	16	10	8	-8	0	12

# Discrete Haar Wavelet Transform

## Discrete Wavelet Transform : Haar wavelet

```
s = 1;
h = [1 1]/2;
g = [1 -1]/2;

y = [2 4 6 8]
a1 = filter (h, 1, y) % level 1 approximation
d1 = filter (g, 1, y) % level 1 detail

a1 = a1 (2:2:end) % downsample
d1 = d1 (2:2:end)

a2 = filter (h, 1, a1) % level 2 approximation
d2 = filter (g, 1, a1) % level 2 detail
a2 = a2 (2:2:end)
d2 = d2 (2:2:end) % downsample

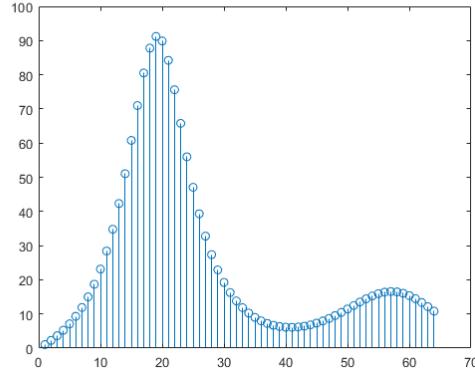
dwty = [a2 d2 d1]
```

	y	a1	d1	a1	d1	a2	d2	dwty
1	2	1	1	3	1	1.5000	1.5000	5
2	4	3	1	5	1	5.0000	2.0000	2
3	6	1	1	7	1			1
4	8	1	1					1

# Discrete Haar Wavelet Transform

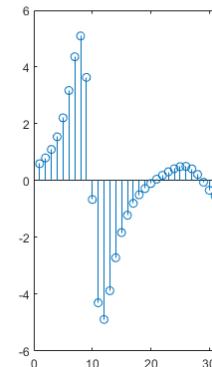
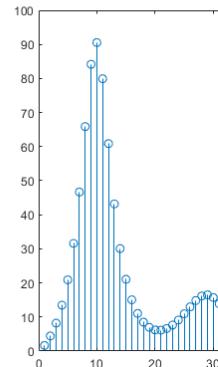
let's try another signal

```
x = (1:64) / 64;
y = humps(x) - humps(0);
stem(y)
```



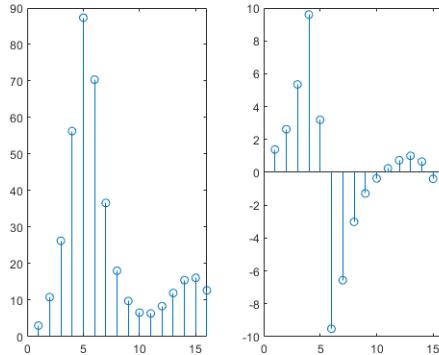
```
a1 = filter (h, 1, y); % level 1 approximation
d1 = filter (g, 1, y); % level 1 detail

a1 = a1 (2:2:end); % downsample
d1 = d1 (2:2:end);
subplot(1,2,1);stem(a1);subplot(1,2,2);stem(d1)]
```

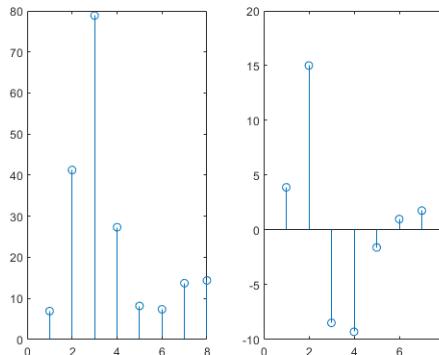


# Discrete Haar Wavelet Transform

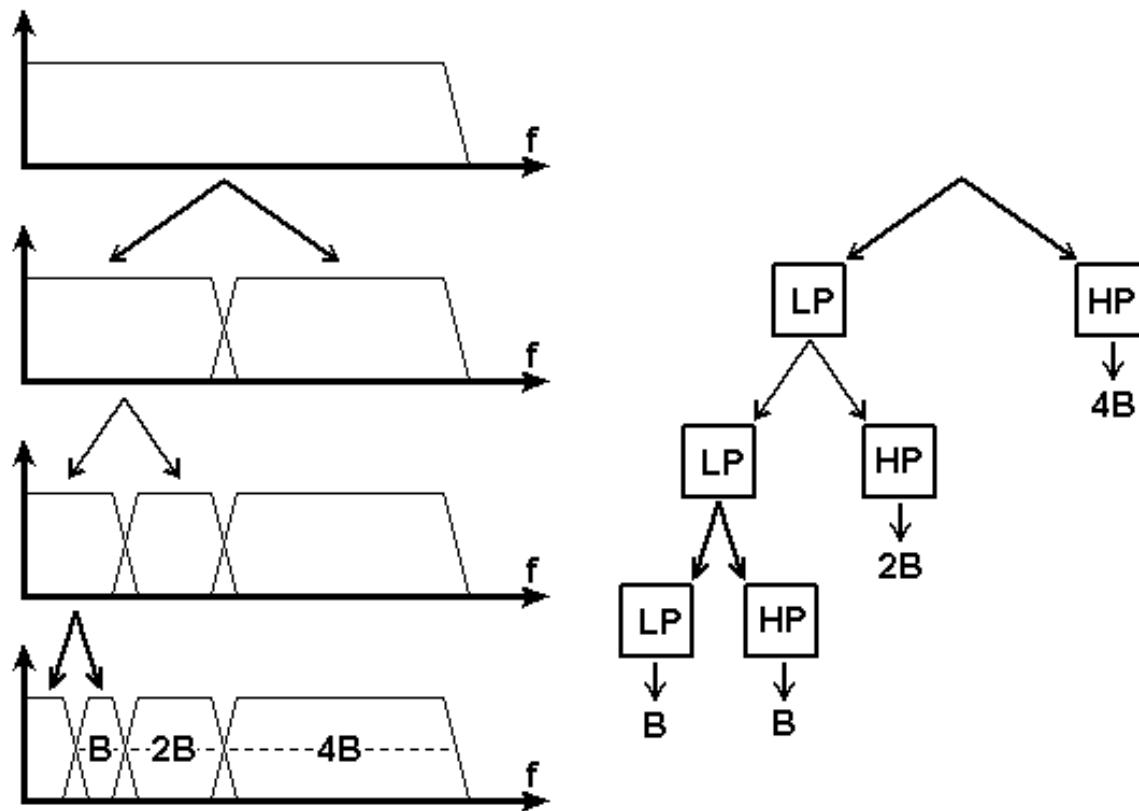
```
a2 = filter (h, 1, a1); % level 2 approximation  
d2 = filter (g, 1, a1); % level 2 detail  
a2 = a2 (2:2:end);  
d2 = d2 (2:2:end); % downsample  
figure;subplot(1,2,1);stem(a2);subplot(1,2,2);stem(d2)
```



```
a3 = filter (h, 1, a2); % level 3 approximation  
d3 = filter (g, 1, a2); % level 3 detail  
a3 = a3 (2:2:end);  
d3 = d3 (2:2:end); % downsample  
figure;subplot(1,2,1);stem(a3);subplot(1,2,2);stem(d3)
```



# Discrete Haar Wavelet Transform



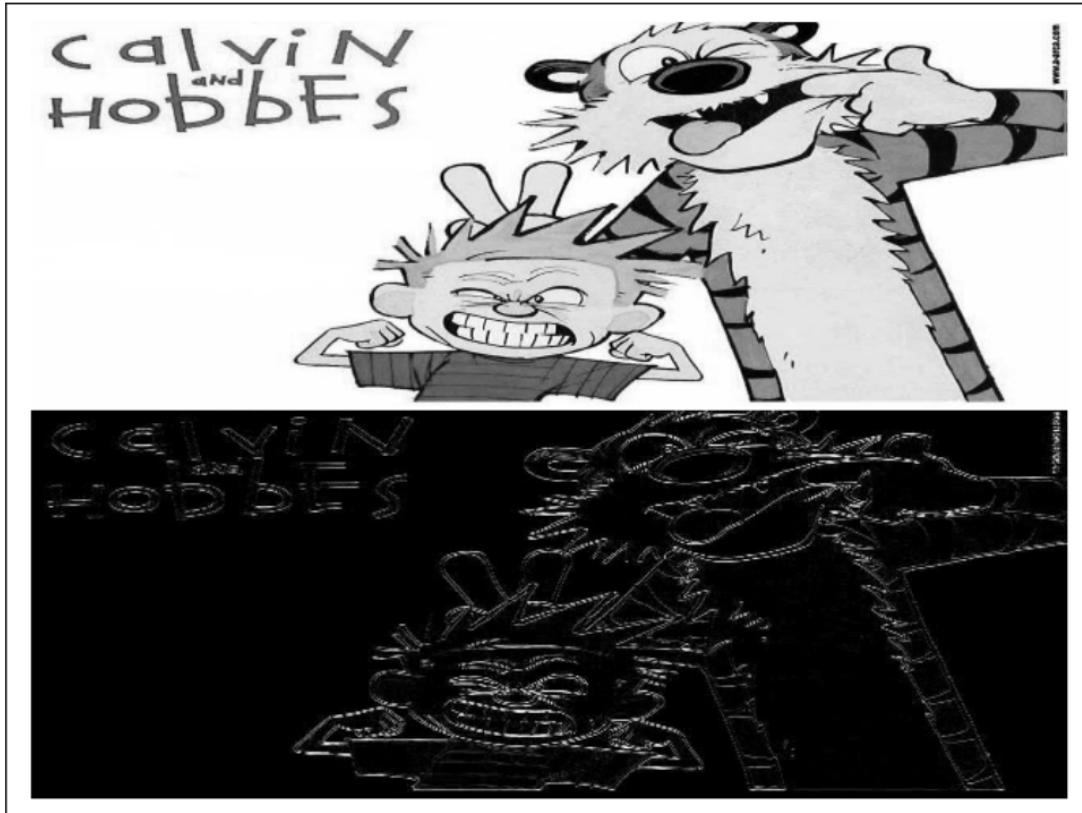
# Discrete Haar Transform Filter Bank

$$H = \begin{bmatrix} 1 & -1 & 0 & 0 & 0 & 0 & 0 & 0 \\ 0 & 0 & 1 & -1 & 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 & 1 & -1 & 0 & 0 \\ 0 & 0 & 0 & 0 & 0 & 0 & 1 & -1 \\ 1 & 1 & -1 & -1 & 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 & 1 & 1 & -1 & -1 \\ 1 & 1 & 1 & 1 & -1 & -1 & -1 & -1 \\ 1 & 1 & 1 & 1 & 1 & 1 & 1 & 1 \end{bmatrix}$$

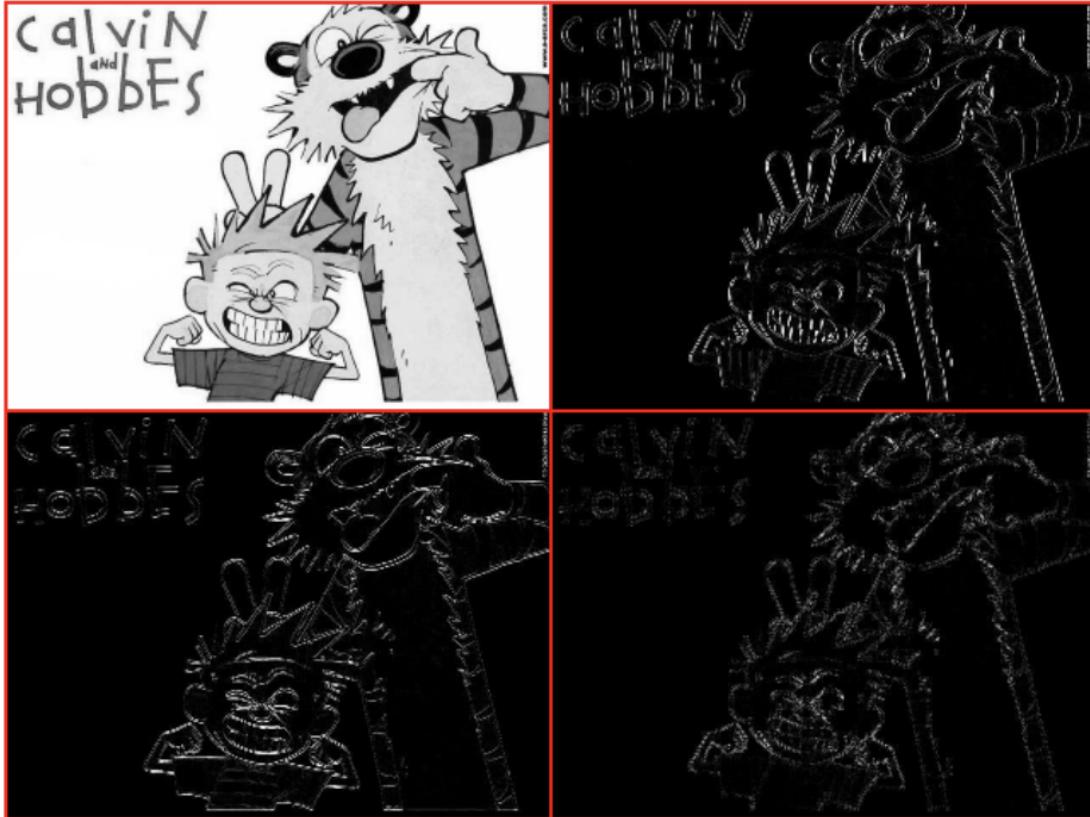
# Discrete Haar Transform Matrix

- ▶ repeat the computation on the **means**
- ▶ keep differences in each step

# 2D Discrete Haar Transform



# 2D Discrete Haar Transform



# 2D Discrete Haar Transform

