

CS448f: Image Processing For Photography and Vision

Wavelets Continued

Last Time:

- Last time we saw the Daubechies filter satisfied the following:
 - fully orthogonal
 - four taps
 - as smooth as possible
 - wavelet filter a simple modification of the scaling filter
- Why did we care about our wavelet basis functions being orthogonal?

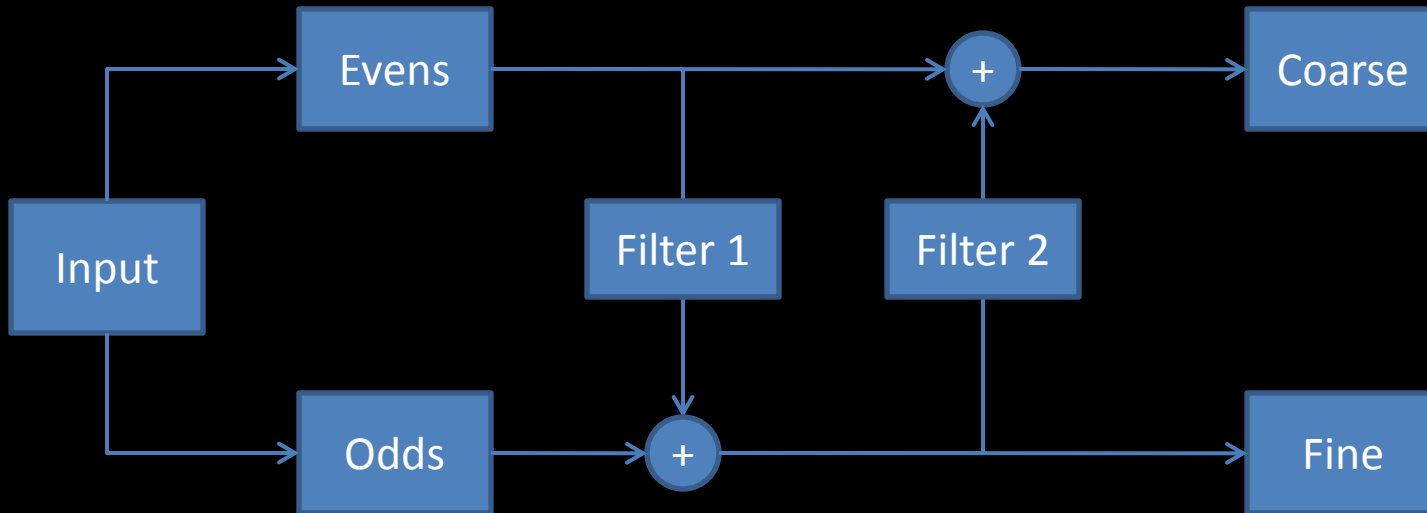
Orthogonality

- Why did we care about our wavelet basis functions being orthogonal?
 - Easy to invert
 - Orthogonal transforms preserve distance
- We can probably relax this requirement, provided we get something that's still easy to invert

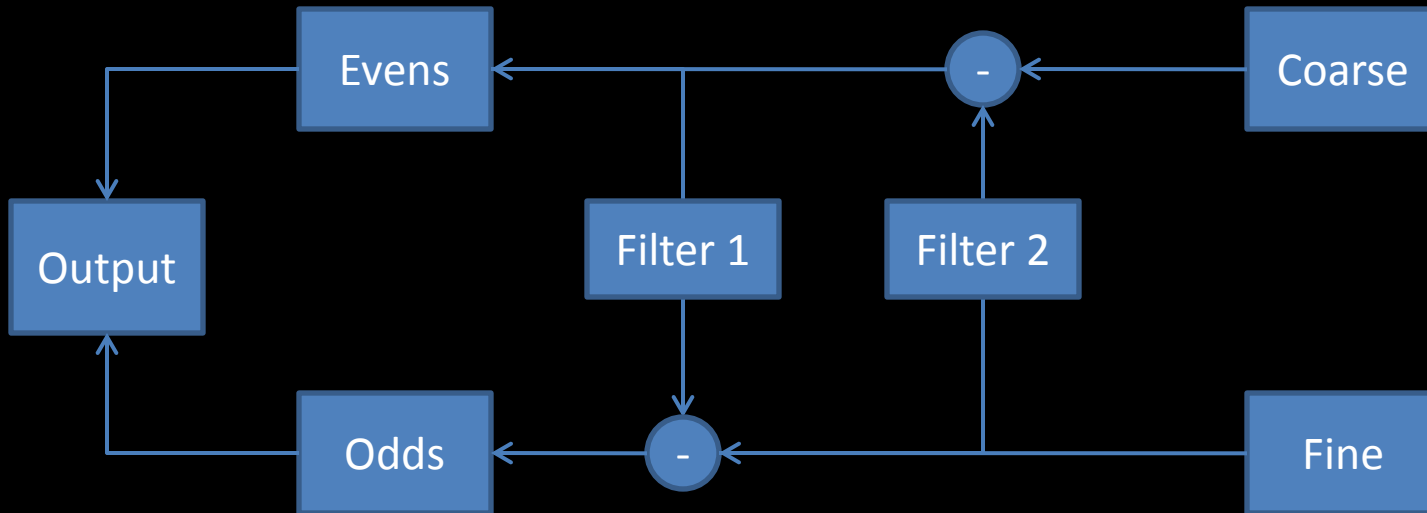
Lifting

- Let's construct our filters using the following sequence:
 - Divide the inputs into evens and odds
 - Add some function of the odds to the evens
 - Add some function of the evens to the odds
 - Repeat as long as you like
 - Eventually the evens form a coarse layer and the odds form a fine layer
- This is easy to invert

Forward Transform



Inverse Transform



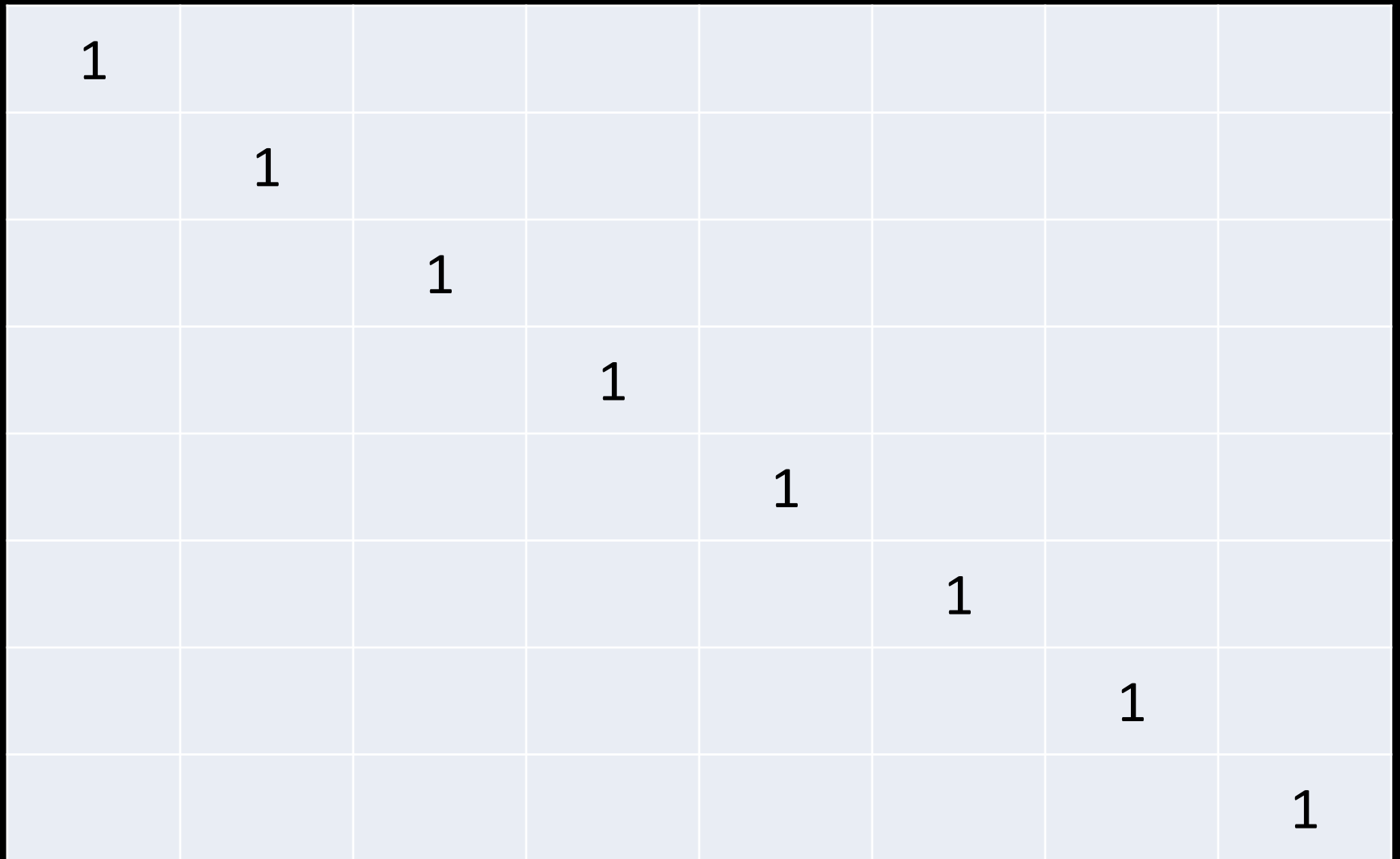
What makes a good fine layer?

- Average value is 0
- So filter 1 should probably be something that enforces that.

What makes a good coarse layer?

- Why is subsampling bad?
 - Some pixels in the input count more than others
- Each pixel in the input should count equally
 - E.g. Averaging down
- Something should sum up to 1

Let's track the linear transform



A grid representing a linear transform matrix. The grid is 8 rows by 8 columns. The diagonal elements are 1, and all other elements are 0.

1							
	1						
		1					
			1				
				1			
					1		
						1	
							1

Divide the rows into even and odd

1							
	1						
		1					
			1				
				1			
					1		
						1	
							1

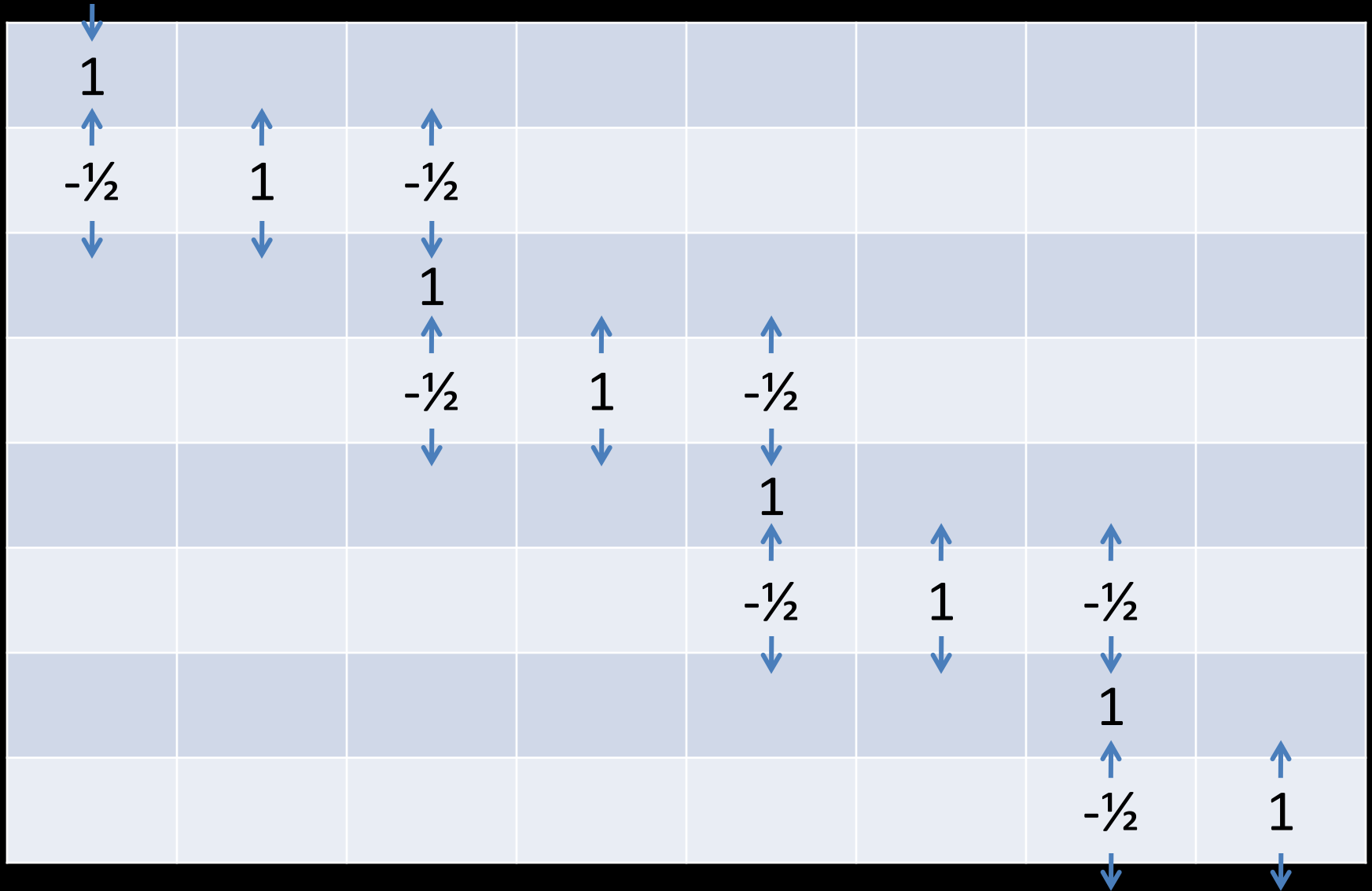
Add a filter of the even rows to the odd rows: $[-\frac{1}{2} \ 0 \ -\frac{1}{2}]$

1							
$-\frac{1}{2}$	1	$-\frac{1}{2}$					
		1					
		$-\frac{1}{2}$	1	$-\frac{1}{2}$			
				1			
				$-\frac{1}{2}$	1	$-\frac{1}{2}$	
						1	
						$-\frac{1}{2}$	1

The odd rows are now a fine layer

1							
$-\frac{1}{2}$	1	$-\frac{1}{2}$					
		1					
		$-\frac{1}{2}$	1	$-\frac{1}{2}$			
				1			
				$-\frac{1}{2}$	1	$-\frac{1}{2}$	
						1	
						$-\frac{1}{2}$	1

Add a filter of the odd rows to the
even rows: $[\frac{1}{4} \ 0 \ \frac{1}{4}]$



Add a filter of the odd rows to the even rows: $[\frac{1}{4} \ 0 \ \frac{1}{4}]$

$\frac{3}{4}$	$\frac{1}{4}$	$-\frac{1}{8}$					
$-\frac{1}{2}$	1	$-\frac{1}{2}$					
$-\frac{1}{8}$	$\frac{1}{4}$	$\frac{3}{4}$	$\frac{1}{4}$	$-\frac{1}{8}$			
		$-\frac{1}{2}$	1	$-\frac{1}{2}$			
		$-\frac{1}{8}$	$\frac{1}{4}$	$\frac{3}{4}$	$\frac{1}{4}$	$-\frac{1}{8}$	
				$-\frac{1}{2}$	1	$-\frac{1}{2}$	
				$-\frac{1}{8}$	$\frac{1}{4}$	$\frac{3}{4}$	$\frac{1}{4}$
						$-\frac{1}{2}$	1

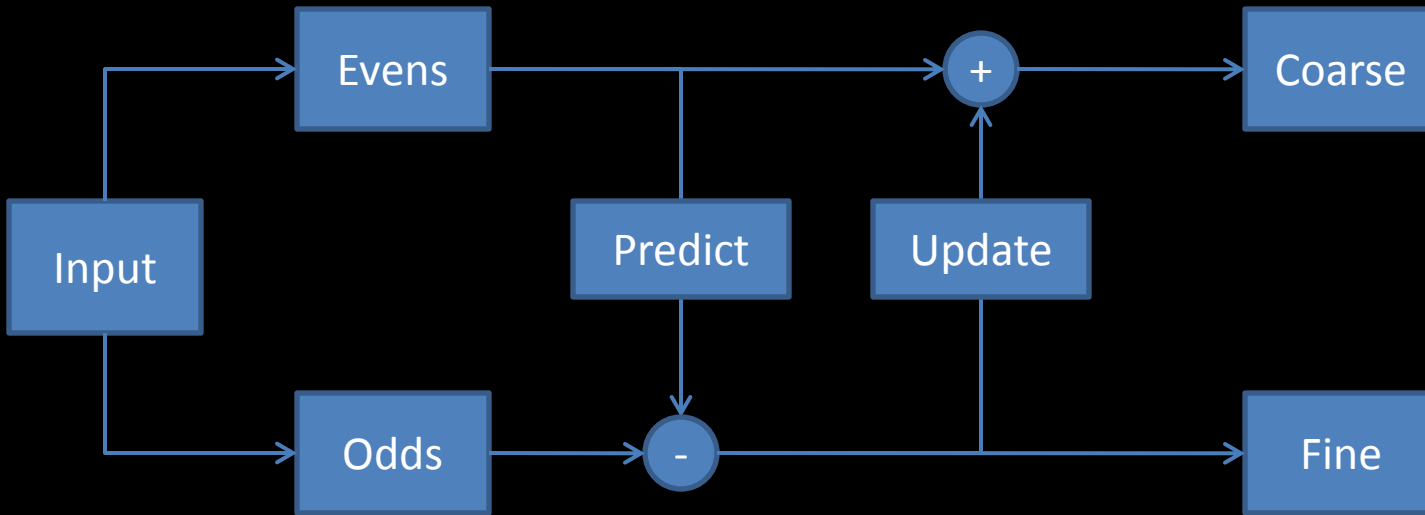
Why did I pick $\frac{1}{4}$?

$\frac{3}{4}$	$\frac{1}{4}$	$-\frac{1}{8}$					
$-\frac{1}{2}$	1	$-\frac{1}{2}$					
$-\frac{1}{8}$	$\frac{1}{4}$	$\frac{3}{4}$	$\frac{1}{4}$	$-\frac{1}{8}$			
		$-\frac{1}{2}$	1	$-\frac{1}{2}$			
		$-\frac{1}{8}$	$\frac{1}{4}$	$\frac{3}{4}$	$\frac{1}{4}$	$-\frac{1}{8}$	
				$-\frac{1}{2}$	1	$-\frac{1}{2}$	
				$-\frac{1}{8}$	$\frac{1}{4}$	$\frac{3}{4}$	$\frac{1}{4}$
						$-\frac{1}{2}$	1

In the coarse layer, each input pixel now counts equally (sum along columns is constant)

$\frac{3}{4}$	$\frac{1}{4}$	$-\frac{1}{8}$					
$-\frac{1}{2}$	1	$-\frac{1}{2}$					
$-\frac{1}{8}$	$\frac{1}{4}$	$\frac{3}{4}$	$\frac{1}{4}$	$-\frac{1}{8}$			
		$-\frac{1}{2}$	1	$-\frac{1}{2}$			
		$-\frac{1}{8}$	$\frac{1}{4}$	$\frac{3}{4}$	$\frac{1}{4}$	$-\frac{1}{8}$	
				$-\frac{1}{2}$	1	$-\frac{1}{2}$	
				$-\frac{1}{8}$	$\frac{1}{4}$	$\frac{3}{4}$	$\frac{1}{4}$
						$-\frac{1}{2}$	1

Lifting



- Using an interpolation for the predict filter gives an appropriate fine layer
- The update filter can be computed from the predict filter

Wavelets

- A coarse/fine decomposition that is fast to compute and takes no more memory than the original
- Better or worse than a Laplacian pyramid?