

Filters in Image Processing

2nd Generation of Wavelets – Lifting Scheme

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Outline

- 1 Motivation
- 2 Introduction to Z-transform
- 3 Analysis of FWT
- 4 Lifting scheme (LS)
- 5 Integer Wavelet transform
- 6 Applications

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Practical use of DWT or FWT?

- Complexity of:
 - DWT: $O(n^2)$
 - FWT: $O(cn)$

Can we do it faster?

- DWT/FWT are computed in floating point arithmetic.

Can we restrict ourselves to integer number?

- DWT/FWT are computed *out-of-place*.

Can we reduce the memory usage?

- Some basis functions in DWT are not orthogonal \rightarrow dual basis must be found.

Can we avoid this issue?

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Z-Transform

Definition

(bilateral) Z-Transform

$$\mathcal{F}(z) = \sum_{n=-\infty}^{\infty} f(n)z^{-n}$$

(unilateral) Z-Transform

$$\mathcal{F}(z) = \sum_{n=0}^{\infty} f(n)z^{-n}$$

Notice: Here $z \in \mathbb{C}$, i.e. for $z = e^{i\omega}$ we get a special case of Z-transform, which is DFT.

Z-Transform

Definition

Forward transform

- converts discrete series into continuous signal (Z-plane)

$$\mathcal{F}(z) = \sum_{n=0}^{\infty} f(n)z^{-n}$$

Inverse transform

- maps continuous signal into discrete series

$$f(n) = Z^{-1}(\mathcal{F}(z)) = \frac{1}{2\pi i} \oint_C \mathcal{F}(z)z^{n-1} dz$$

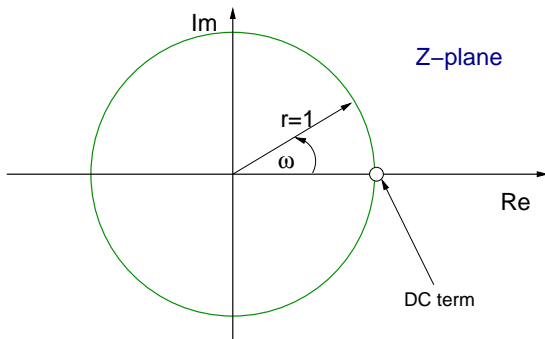
where C is a counterclockwise closed path encircling (with radius 1) the origin

Z-Transform

Properties

Important notes:

- green circle ($z = e^{i\varphi} \Rightarrow |z| = 1$) reduces Z-transform simply to discrete Fourier transform
- DC (direct current) term is DC term from DFT



List of the most important properties

- linearity:

$$Z\{af(n) + bg(n)\} = a\mathcal{F}(z) + b\mathcal{G}(z)$$

- delay (shift):

$$Z\{f(n - k)\} = \mathcal{F}(z)z^{-k}$$

- convolution theorem:

$$Z\{f(n) * g(n)\} = \mathcal{F}(z) \cdot \mathcal{G}(z)$$

- symmetry:

$$Z\{f(-n)\} = \mathcal{F}(z^{-1})$$

Z-Transform

Some examples

- Z-transform of a constant signal:

$$f(n) = [1, 1, 1, 1, \dots]$$
$$\mathcal{F}(z) = 1 + z^{-1} + z^{-2} + z^{-3} + \dots = \frac{1}{1 - z^{-1}} = \frac{z}{z - 1}$$

- Z-transform of a linear filter:

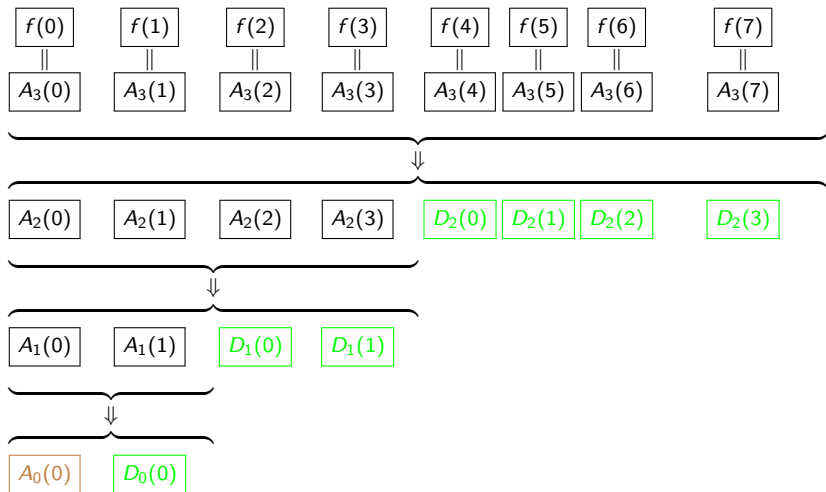
$$\begin{aligned} \text{filter}(f(k)) &= (-2)f(k-1) + 3f(k) + 5f(k+2) \\ &= [-2, \boxed{3}, 0, 5] \quad \text{/written as a kernel/} \\ \text{Filter}(z) &= (-2)z^{-1} + 3z^0 + 5z^2 \end{aligned}$$

Notice: Inverse for linear filters is straightforward.

- 1 Motivation
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Analysis of FWT

Let $|f| = N = 8 = 2^3 = 2^J$ and $j_0 = 0$



Analysis of FWT

An example (Haar wavelets)

Let $f = A_2 = [1 \ 4 \ -3 \ 0]$ be an input signal:

$$A_1 = \left(f * \left[\frac{\sqrt{2}}{2}, \frac{\sqrt{2}}{2} \right] \right) (\downarrow 2 \times)$$

$$D_1 = \left(f * \left[\frac{\sqrt{2}}{2}, -\frac{\sqrt{2}}{2} \right] \right) (\downarrow 2 \times)$$

During one transition, we perform only one averaging and difference:

$$A_1(k) = \left((f(k) + f(k+1)) \frac{\sqrt{2}}{2} \right) (\downarrow 2 \times)$$

$$D_1(k) = \left((f(k+1) - f(k)) \frac{\sqrt{2}}{2} \right) (\downarrow 2 \times)$$

Analysis of FWT

An example (Haar wavelets)

```
>> [LoD, HiD, LoR, HiR] = wfilters('haar')
```

```
LoD = 0.7071    0.7071
```

```
HiD = -0.7071   0.7071
```

```
LoR = 0.7071    0.7071
```

```
HiR = 0.7071   -0.7071
```

```
>> [A1,D1] = dwt([1 4 -3 0], LoD, HiD)
```

```
A1 = 3.5355    -2.1213
```

```
D1 = -2.1213   -2.1213
```

```
>> [A0,D0] = dwt(A1, LoD, HiD)
```

```
A0 = 1.0000
```

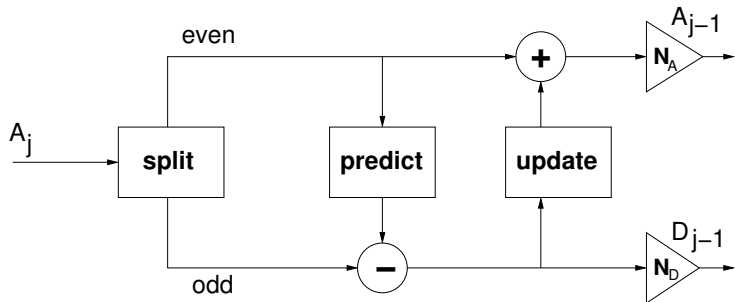
```
D0 = 4.0000
```

Here $c = |LoD| = |HiD| = 2$.

- 1 Motivation
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Lifting scheme (LS)

- Developed in 1996 by Wim Sweldens
- Adopted idea of FastDWT
- The transition from level j to $j - 1$ is computed efficiently
- Basic idea: **split** \rightarrow **predict** \rightarrow **update** \rightarrow **normalize**

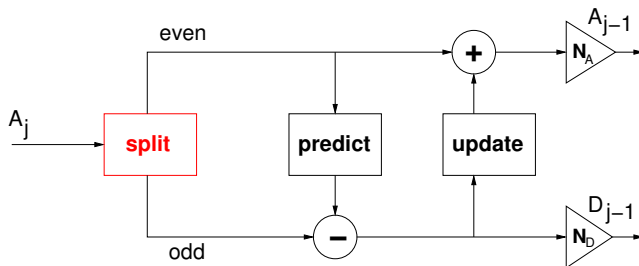


Lifting scheme

Split step

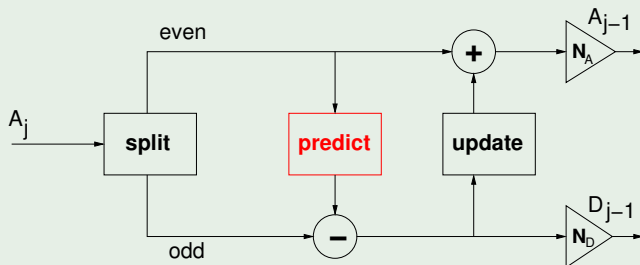
- Also known as *lazy wavelet*.
- The signal A_j is split into odd and even samples:

$$(A_{j-1}, D_{j-1}) = \text{split}(A_j)$$



Lifting scheme

Prediction step

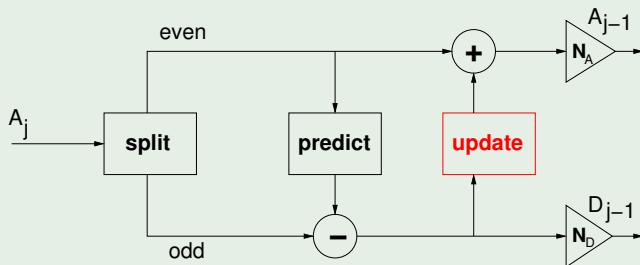


$$D_{j-1}(k) = D_{j-1}(k) - \text{predict}(A_{j-1}(k))$$

- Also known as *dual lifting step*
- When using Haar wavelets the neighbouring samples are supposed to be equal, i.e. the predictor is simple: $\text{predict}_{\text{Haar}}(f(k)) = f(k)$
 $D_{j-1}(k) = D_{j-1}(k) - A_{j-1}(k)$

Lifting scheme

Update step

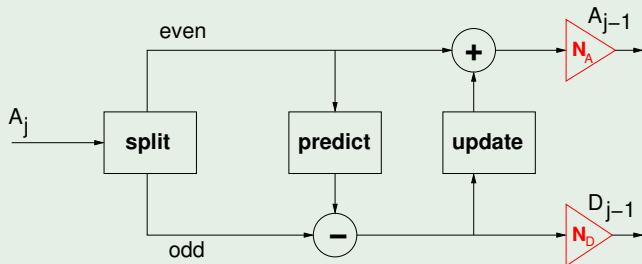


$$A_{j-1}(k) = A_{j-1}(k) + \text{update}(D_{j-1}(k))$$

- Also known as *primal lifting step*
- Update step repairs the wrong estimate of the prediction step.
- When using Haar wavelet, we use: $\text{update}_{\text{Haar}}(f(k)) = \frac{1}{2}f(k)$
 $A_{j-1}(k) = A_{j-1}(k) + \frac{1}{2}D_{j-1}(k)$

Lifting scheme

Normalization step



$$A_{j-1}(k) = A_{j-1}(k) \cdot N_A$$

$$D_{j-1}(k) = D_{j-1}(k) \cdot N_D$$

$$N_A \cdot N_D = 1$$

- The output signals are normalized to avoid boosting the signal.

Lifting scheme

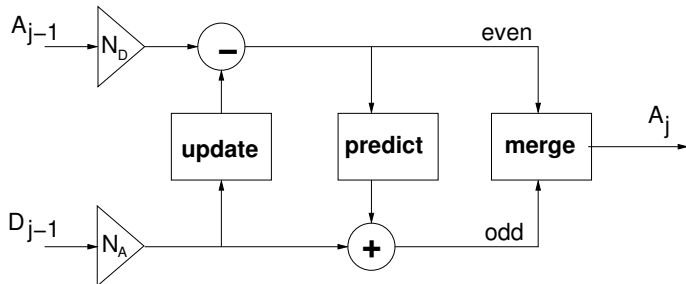
Inverse lifting

- The forward algorithm can be simply inverted:

$$A_{j-1}(k) = A_{j-1}(k) - \text{update}(D_{j-1}(k))$$

$$D_{j-1}(k) = D_{j-1}(k) + \text{predict}(A_{j-1}(k))$$

$$A_j = \text{merge}(A_{j-1}, D_{j-1})$$



Lifting scheme

Inverse lifting (example)

- Namely for Haar wavelets we get:

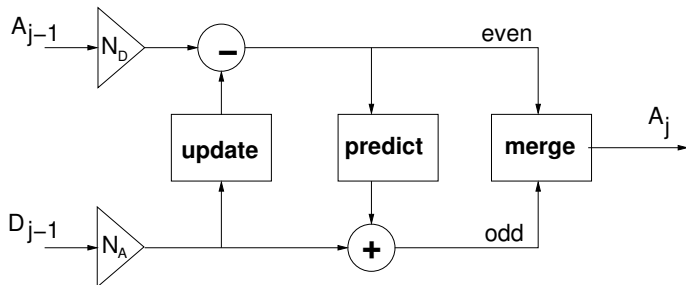
$$A_{j-1}(k) = A_{j-1}(k) \cdot N_D$$

$$D_{j-1}(k) = D_{j-1}(k) \cdot N_A$$

$$A_{j-1}(k) = A_{j-1}(k) - (D_{j-1}(k)/2)$$

$$D_{j-1}(k) = D_{j-1}(k) + A_{j-1}(k)$$

$$A_j = \text{merge}(A_{j-1}, D_{j-1})$$



Lifting scheme

From Filters to Lifting

There exists algorithm invented by Wim Sweldens (1996):

- 1 Input: either 'LoD' and 'HiD' filters or ϕ and φ functions
- 2 Convert both filters 'LoD' and 'HiD' to Z-domain
- 3 Create *polyphase matrix* (2×2)
- 4 Factorize matrix into simple (lower and upper diagonal) matrices
- 5 Each simple matrix correspond either to one update or prediction step
- 6 Convert each matrix from Z-domain to time domain

The algorithm is quite tricky \rightarrow we utilize Matlab function 'liftwave' that performs steps 1-5. All we have to do is to enjoy step 6 😊

Lifting scheme

From Filters to Lifting (Haar)

```
>> h = liftwave('haar')
h = 'd'      [ -1]   [0]
      'p'      [0.5000] [0]
      [1.4142] [0.7071] []
```

- 'd' ... **d**ual lifting step (predict)
 $Predict(z) = (-1) \cdot z^0 \rightarrow predict(f(k)) = (-1) \cdot f(k)$
- 'p' ... **p**rimal lifting step (update)
 $Update(z) = 0.5 \cdot z^0 \rightarrow update(f(k)) = 0.5 \cdot f(k)$
- [1.4142] ... N_A
- [0.7071] ... N_D

Lifting scheme

From Filters to Lifting (Daubechies-2)

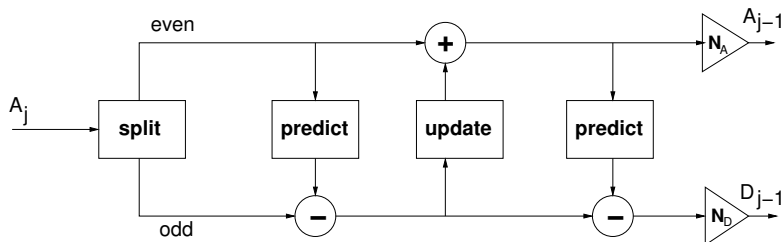
In general, there may appear two or more *predict(.)* and *update(.)* steps:

```
>> h = liftwave('db2')
h = { ...
    'd'          [ -1.732]          [0]
    'p'          [ -0.067  0.433]    [1]
    'd'          [  1.000]          [-1]
    [  1.932]    [  0.518]          []
```

- 'd' ... **dual** lifting step (predict)
 $Predict_1(z) = (-1.732) \cdot z^0 \rightarrow predict_1(f(k)) = (-1.732) \cdot f(k)$
- 'p' ... **primal** lifting step (update)
 $Update_1(z) = (-0.067) \cdot z^1 + 0.433 \cdot z^0 \rightarrow$
 $update_1(f(k)) = (-0.067) \cdot f(k+1) + 0.433 \cdot f(k)$
- 'd' ... **dual** lifting step (predict)
 $Predict_2(z) = 1.000 \cdot z^{-1} \rightarrow predict_2(f(k)) = 1.000 \cdot f(k-1)$
- [1.932] ... N_A
- [0.518] ... N_D

Lifting scheme

From Filters to Lifting (Daubechies-2)



$$\text{predict}_1(f(k)) = (-1.732) f(k)$$

$$\text{update}_1(f(k)) = 0.433 f(k) - 0.067 f(k + 1)$$

$$\text{predict}_2(f(k)) = f(k - 1)$$

Lifting scheme

Technical/Implementation notes

Lifting ordering (for $N = 8$)

$f(0)$	$f(1)$	$f(2)$	$f(3)$	$f(4)$	$f(5)$	$f(6)$	$f(7)$
$A_2(0)$	$D_2(0)$	$A_2(1)$	$D_2(1)$	$A_2(2)$	$D_2(2)$	$A_2(3)$	$D_2(3)$
$A_1(0)$		$D_1(0)$		$A_1(1)$		$D_1(1)$	
$A_0(0)$				$D_0(0)$			

Notice: The computation is performed completely *in-place*.

From filters to lifting scheme

- conversion 'LoD', 'HiD' \rightarrow $update(\cdot)$, $predict(\cdot)$ always exists but is not unique
- conversion is performed in frequency domain via Z -transform (decomposition of polyphase matrices)

Lifting scheme

FWT and DWT comparison (examples)

Price of one decomposition level using DWT ($|f| = N$)

family of wavelets	multiplications	additions
Haar	$4N$	$2N$
Daubechies-2	$8N$	$6N$

Extra memory usage: one memory buffer of size $O(N)$ needed for convolution.

Price of one decomposition level using LS ($|f| = N$)

family of wavelets	multiplications	additions
Haar	$2N$	N
Daubechies-2	$3N$	$2N$

Extra memory usage: \emptyset

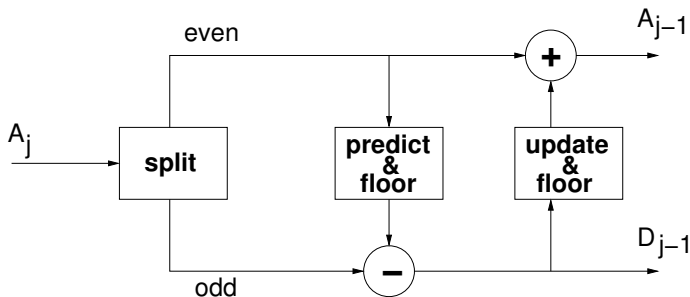
- 1 Motivation
- 2 Introduction to Z-transform
- 3 Analysis of FWT
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Basic idea & Properties

- IWT originates from lifting scheme (chain of predictions and updates).
- The fixed point arithmetic is guaranteed via 'floor' function.
- The rounding error produced in forward transform is compensated by mirror 'floor' in inverse lifting.
- The lifting is the same as the standard one but
 - each multiplication is followed but the truncation
 - no normalization is present

Integer Wavelet Transform

An example (Haar)



$$(A_{j-1}, D_{j-1}) = \text{split}(A_j)$$

$$D_{j-1}(k) = D_{j-1}(k) - \text{floor}(A_{j-1}(k))$$

$$A_{j-1}(k) = A_{j-1}(k) + \text{floor}(D_{j-1}(k)/2)$$

Integer Wavelet Transform

An example (Haar)

```
>> h2 = liftwave('haar', 'Int2Int')
```

```
h2 = 'd'      [ -1]    [0]
      'p'      [0.5000]  [0]
      [1.4142]  [0.7071]  'I'
```

```
>> [A1,D1] = lwt([1 4 -3 0], h2)
```

```
A1 = 2    -2
D1 = 3     3
```

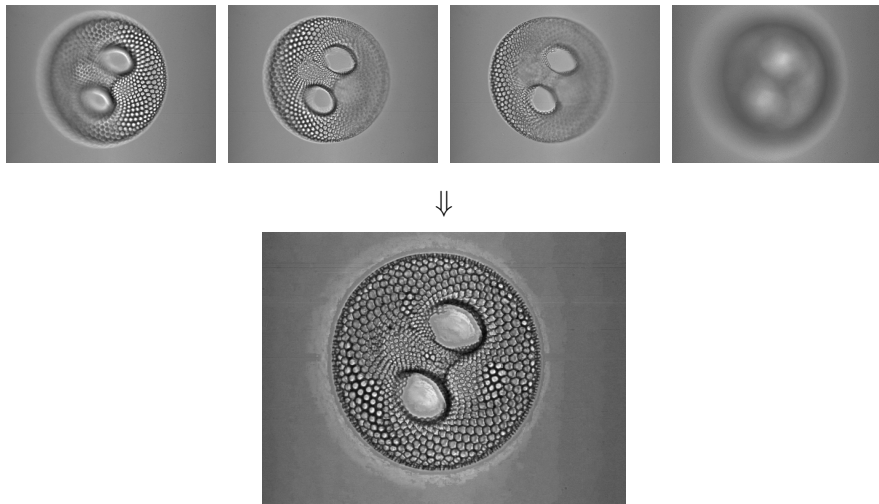
```
>> ilwt(A1,D1,h2)
```

```
ans = 1     4    -3     0
```


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Applications

Image fusion – in confocal microscopy



source: B. Zítová, UTIA, CAS

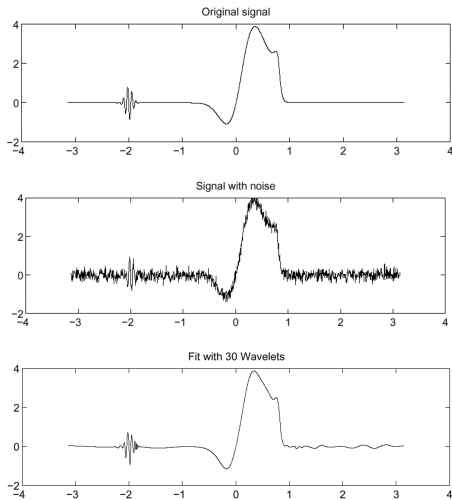
Applications

Image fusion – in photography



Applications

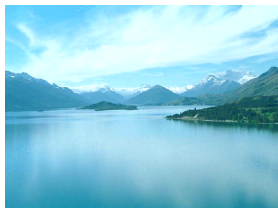
Noise removal



source: B. Zítová, UTIA, CAS

Applications

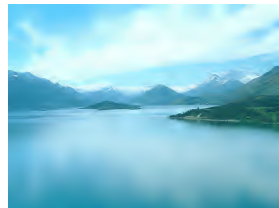
Image compression



original (979 kB)



JPEG (6.21 kB)



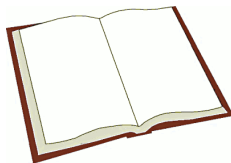
JPEG2000 (1.83 kB)

Applications

... and the others

- fusion of images with different resolution
- image registration
- edge detection
- ...

- [Li H.](#), [Manjunath B.S.](#), [Mitra S.K.](#) Multisensor Image Fusion Using the Wavelet Transform, Graphical Models and Image Processing, Volume 57, Issue 3, May 1995, Pages 235-245, ISSN 1077-3169
- [Jensen A.](#), [La Cour-Harbo A.](#) Ripples in mathematics: the discrete wavelet transform, Springer, Berlin, 2001, ISBN 3-540-41662-5
- [Sweldens W.](#) The lifting scheme: A custom-design construction of biorthogonal wavelets. Applied and Computational Harmonic Analysis. 1996, Vol 3, Issue 2, pp 186200



You should know the answers . . .

- Describe the relationship between Fourier transform and Z-transform.
- Can you apply Z-transform to a given signal or a linear filter? Give an example.
- Explain three principal phases of *lifting scheme*.
- Compare the time and spatial complexity of FWT and lifting scheme.
- Compute one level of wavelet transform (using Haar's basis) via lifting scheme for signal $f=[3 \ 5 \ 0 \ -1 \ 4 \ 2]$.
- What does *integer* wavelet transform mean?
- How does image fusion via DWT work?
- How does image denoising via DWT work?