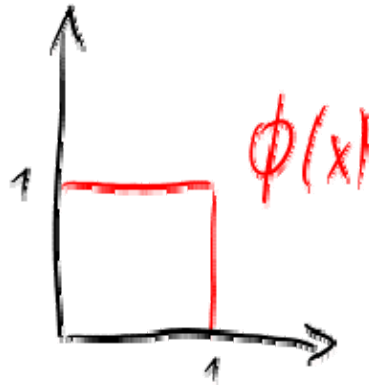
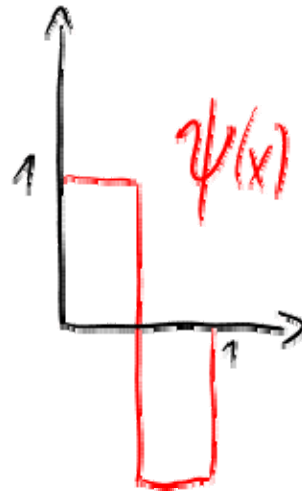


Haar wavelets

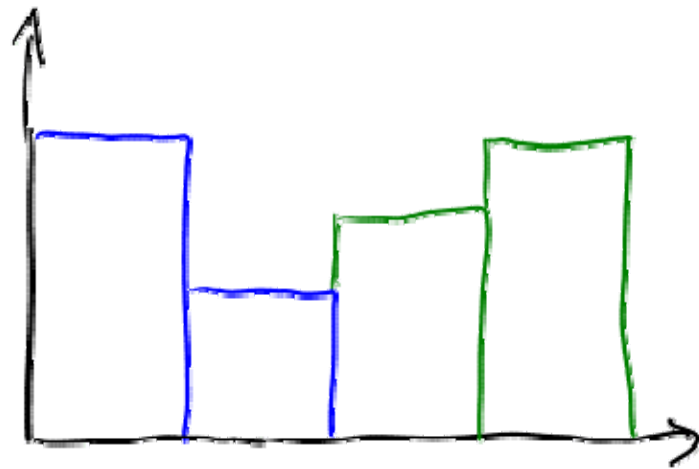
Basis function



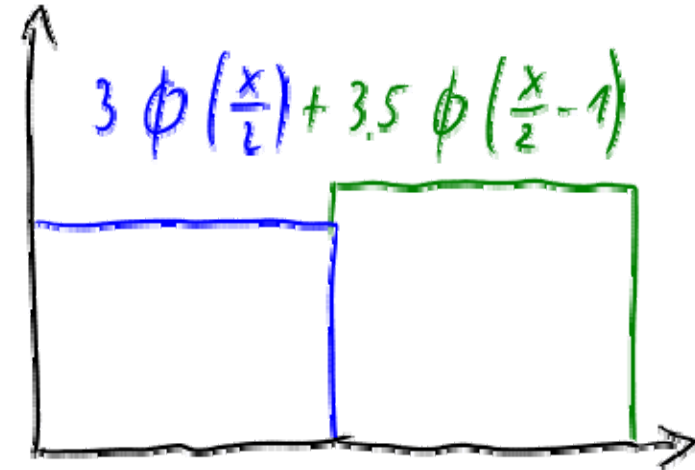
Wavelet function



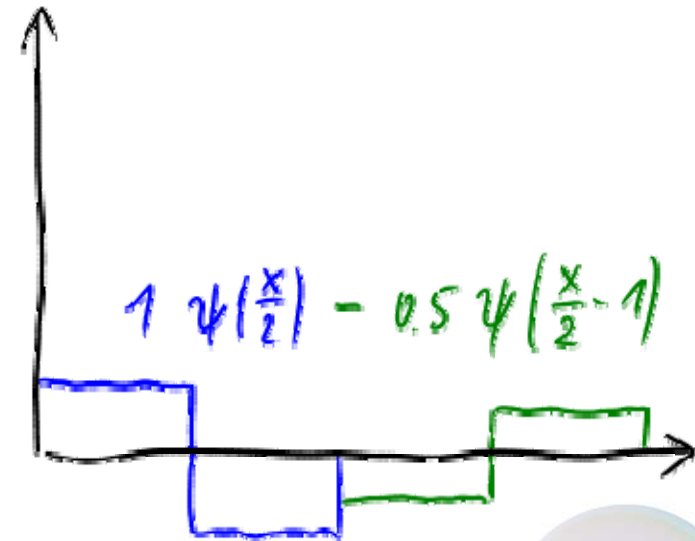
Haar wavelets



$$4\phi(x) + 2\phi(x-1) + 3\phi(x-2) + 4\phi(x-3)$$



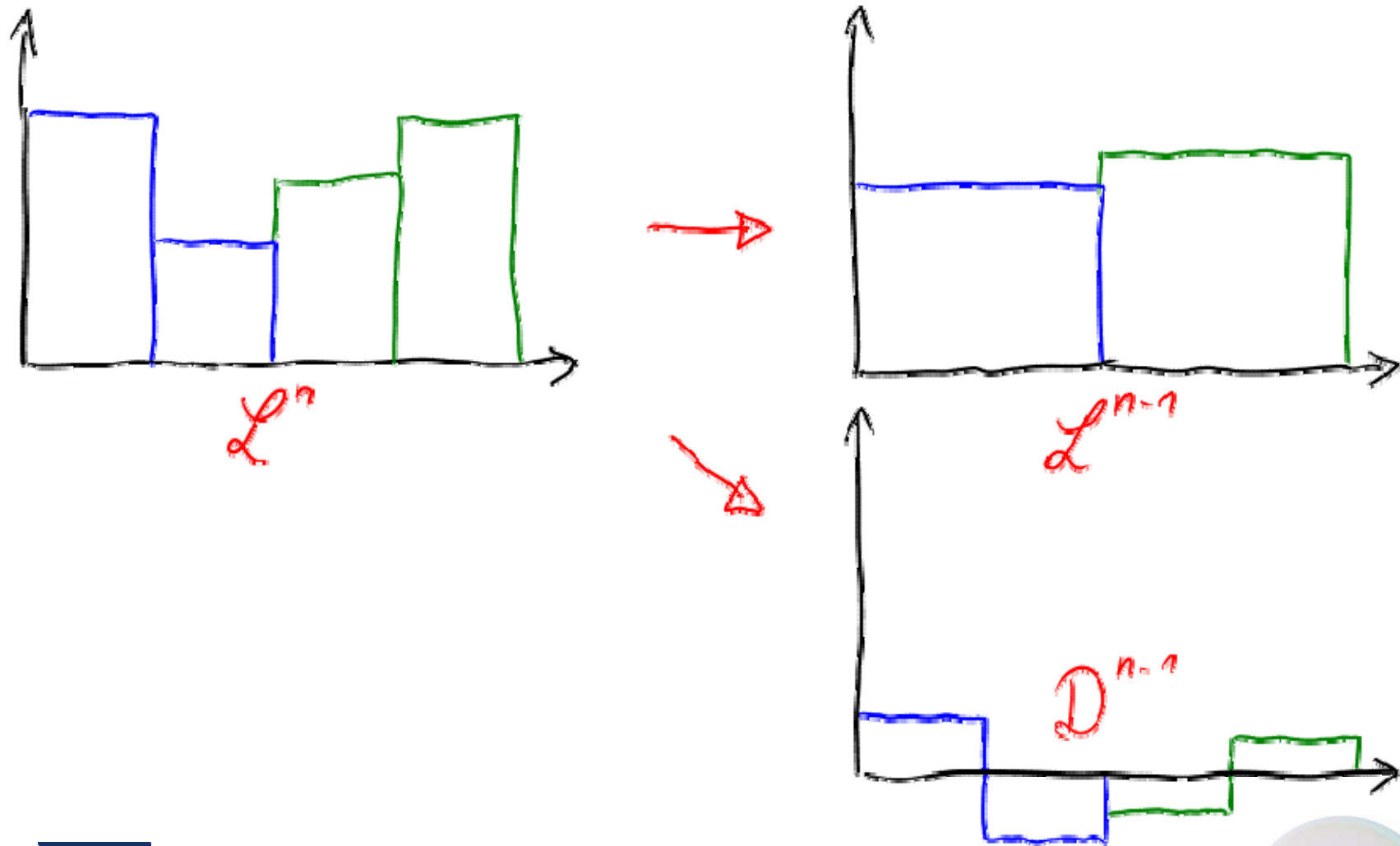
$$3\phi\left(\frac{x}{2}\right) + 3.5\phi\left(\frac{x}{2}-1\right)$$



$$1\psi\left(\frac{x}{2}\right) - 0.5\psi\left(\frac{x}{2}-1\right)$$



Multiresolution representation

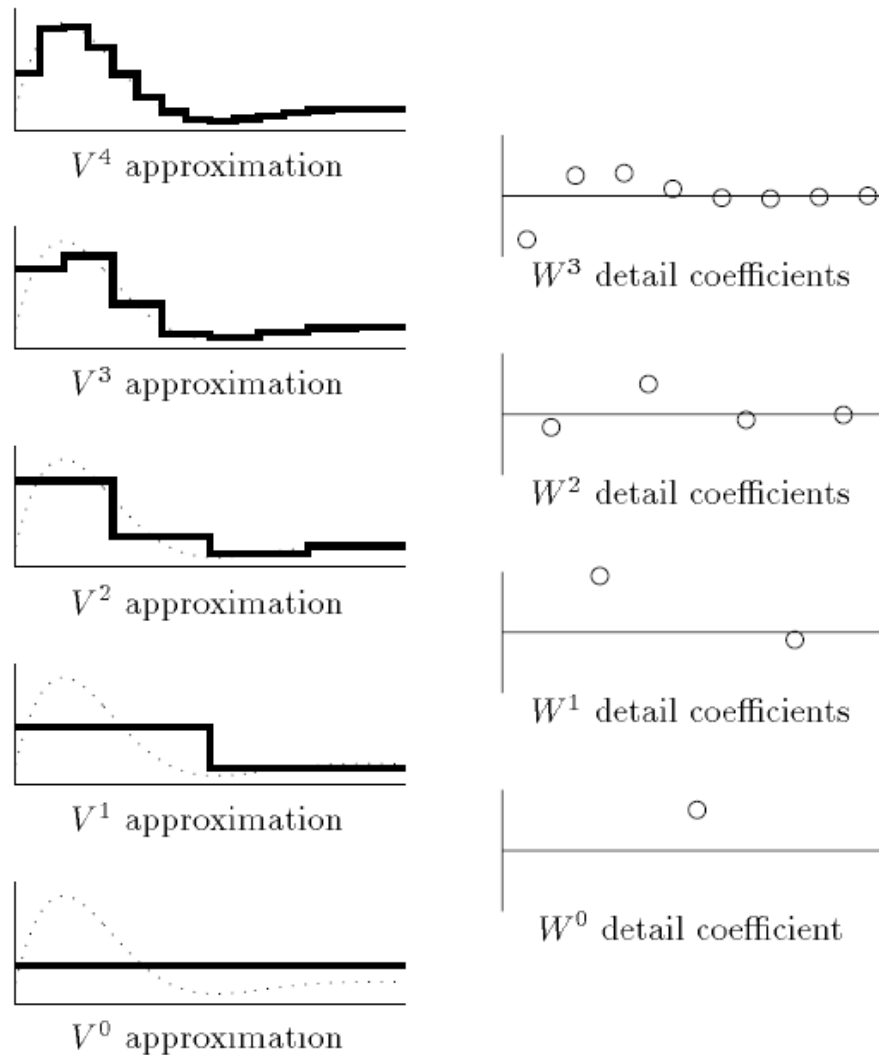


Multiresolution representation

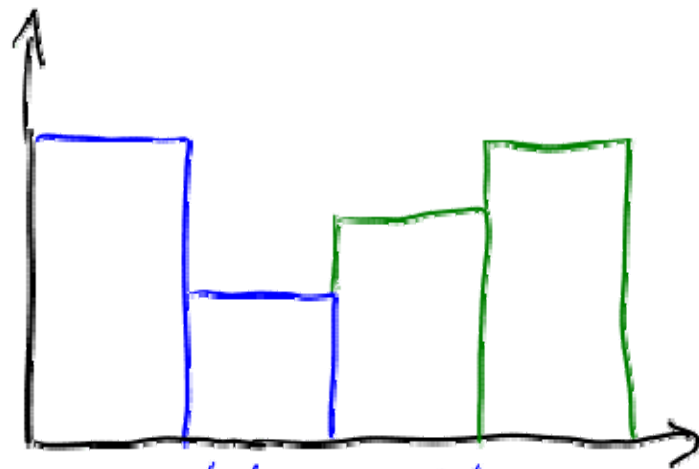
- Object is represented as a sequence of resolutions.
- The resolutions \mathcal{L}^n are called levels (levels of detail, LOD)
- The differences \mathcal{D}^{n-1} are called detail coefficients.
- The levels build a multiresolution hierarchy:
$$\mathcal{L}^n \supset \mathcal{L}^{n-1} \supset \dots \supset \mathcal{L}^0$$
- The level \mathcal{L}^0 is the base level.
- The base level does not need to be represented by a regular mesh. All levels use then semi-regular meshes.



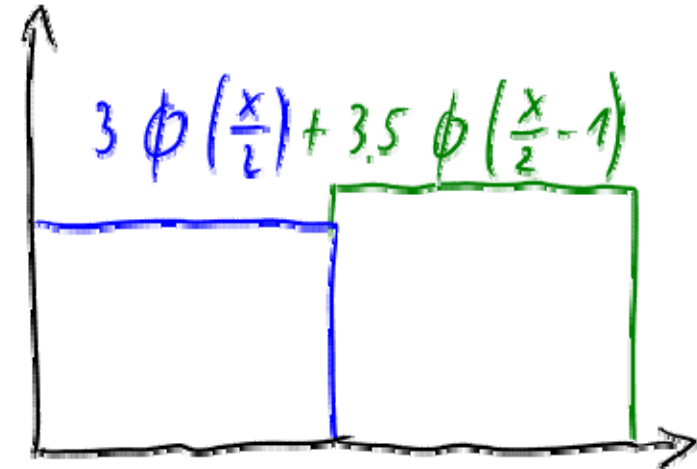
Multiresolution representation with Haar wavelets



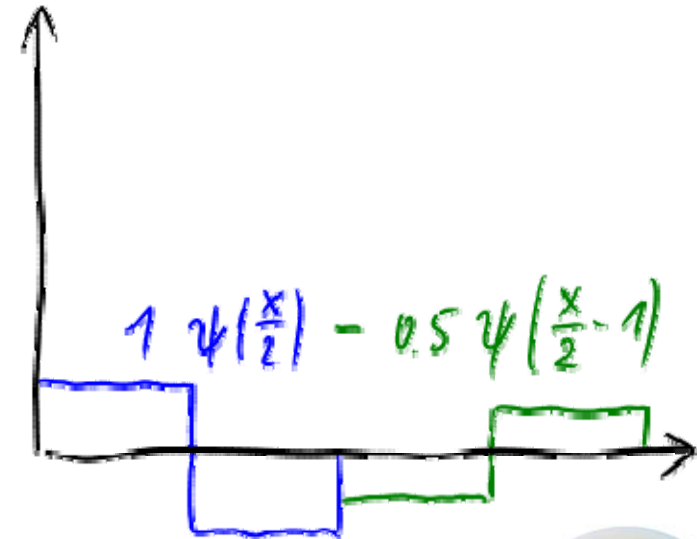
Memory requirements



$$4\phi(x) + 2\phi(x-1) + 3\phi(x-2) + 4\phi(x-3)$$



$$3\phi\left(\frac{x}{2}\right) + 3.5\phi\left(\frac{x}{2} - 1\right)$$



$$1\psi\left(\frac{x}{2}\right) - 0.5\psi\left(\frac{x}{2} - 1\right)$$

Storing the highest resolution L^n requires the same amount of storage as storing the coarsest resolution L^0 and all detail coefficients

$$D^{n-1}, D^{n-2}, \dots, D^0$$



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1.4.2 Wavelets



References

- [Wavelets for Computer Graphics: A Primer.](#)
Eric J. Stollnitz, Tony D. DeRose, David H. Salesin
University of Washington, technical report,
September 1994.

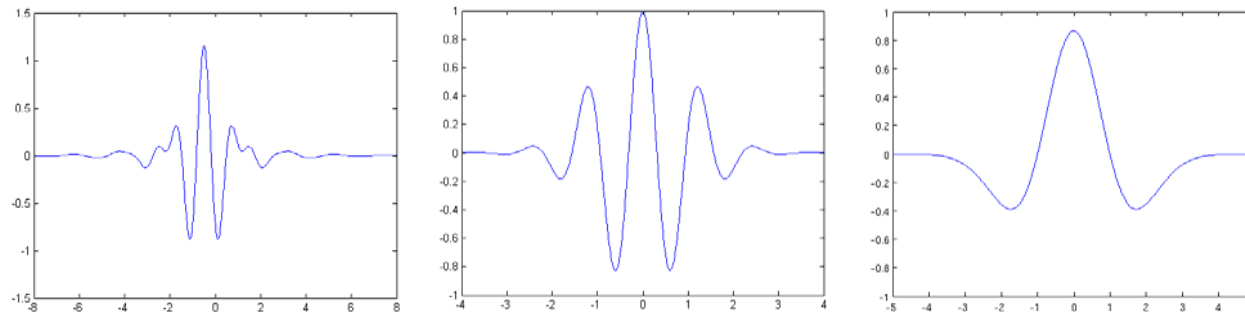


Wavelets

- Haar wavelets are the simplest and oldest wavelets (Alfred Haar, 1909).
- Wavelet theory started much later (boom in the 80s).
- Wavelet functions must fulfill the criteria:

$$\int_{-\infty}^{\infty} \psi(t) dt = 0$$
$$\int_{-\infty}^{\infty} |\psi(t)|^2 dt = 1$$

- Examples:

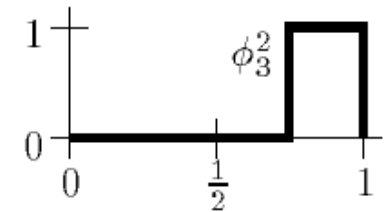
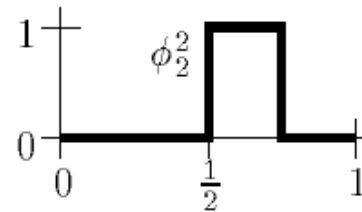
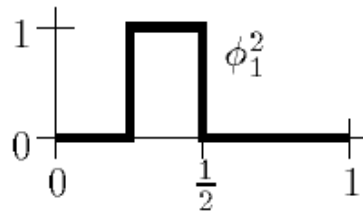
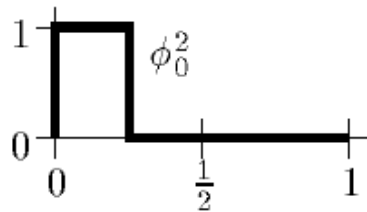


Multiresolution discrete wavelet transform

- Basis and wavelet functions span spaces:

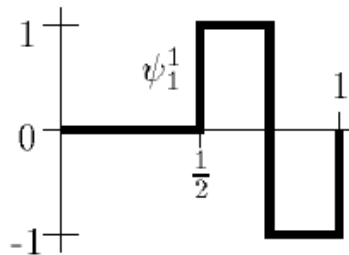
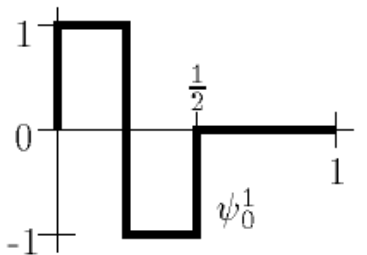
$$V_m = \text{span}(\phi_{m,n} : n \in \mathbb{Z})$$

$$\phi_{m,n}(t) = 2^{-m/2} \phi(2^{-m}t - n)$$



$$W_m = \text{span}(\psi_{m,n} : n \in \mathbb{Z})$$

$$\psi_{m,n}(t) = 2^{-m/2} \psi(2^{-m}t - n)$$



Multiresolution discrete wavelet transform

- The spaces V_m form a multiresolution analysis:

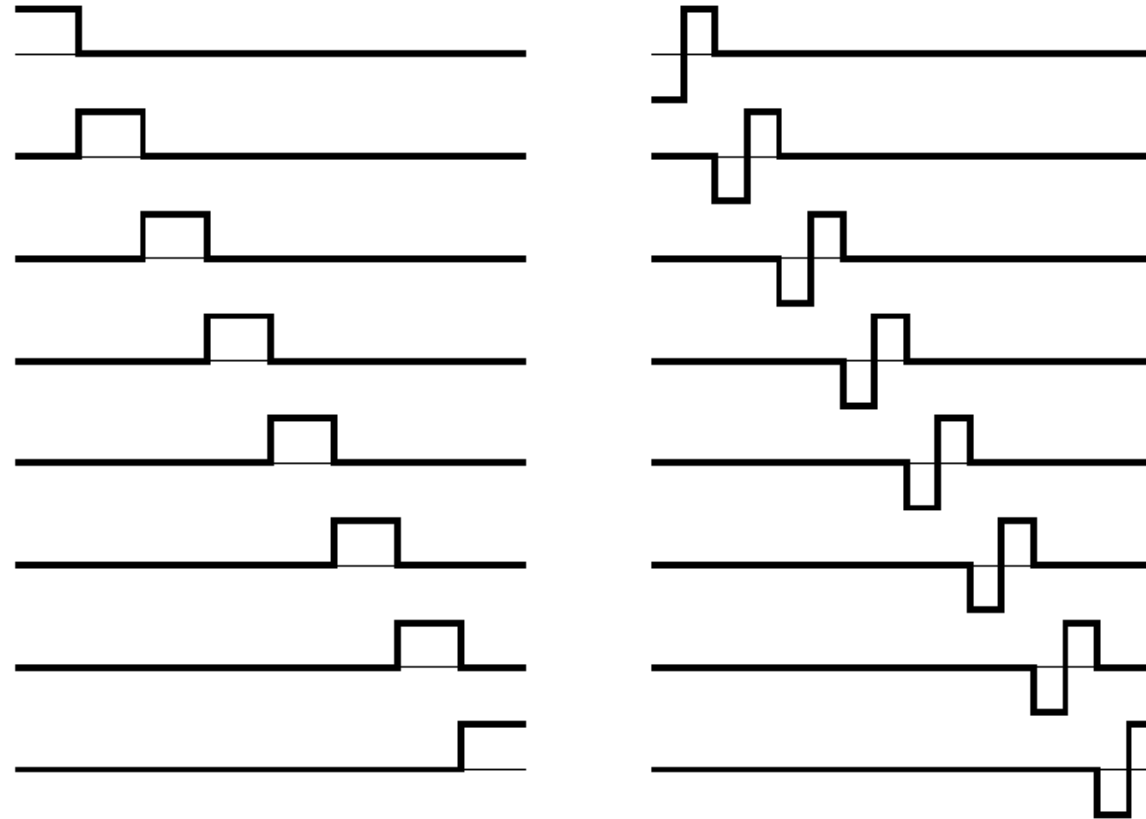
$$V_m \supset V_{m-1} \supset \dots \supset V_0$$

- The spaces W_m are the orthogonal complements:

$$V_m \oplus W_m = V_{m+1}$$



Constant B-spline wavelets (Haar wavelets)



Constant B-spline wavelets (Haar wavelets)

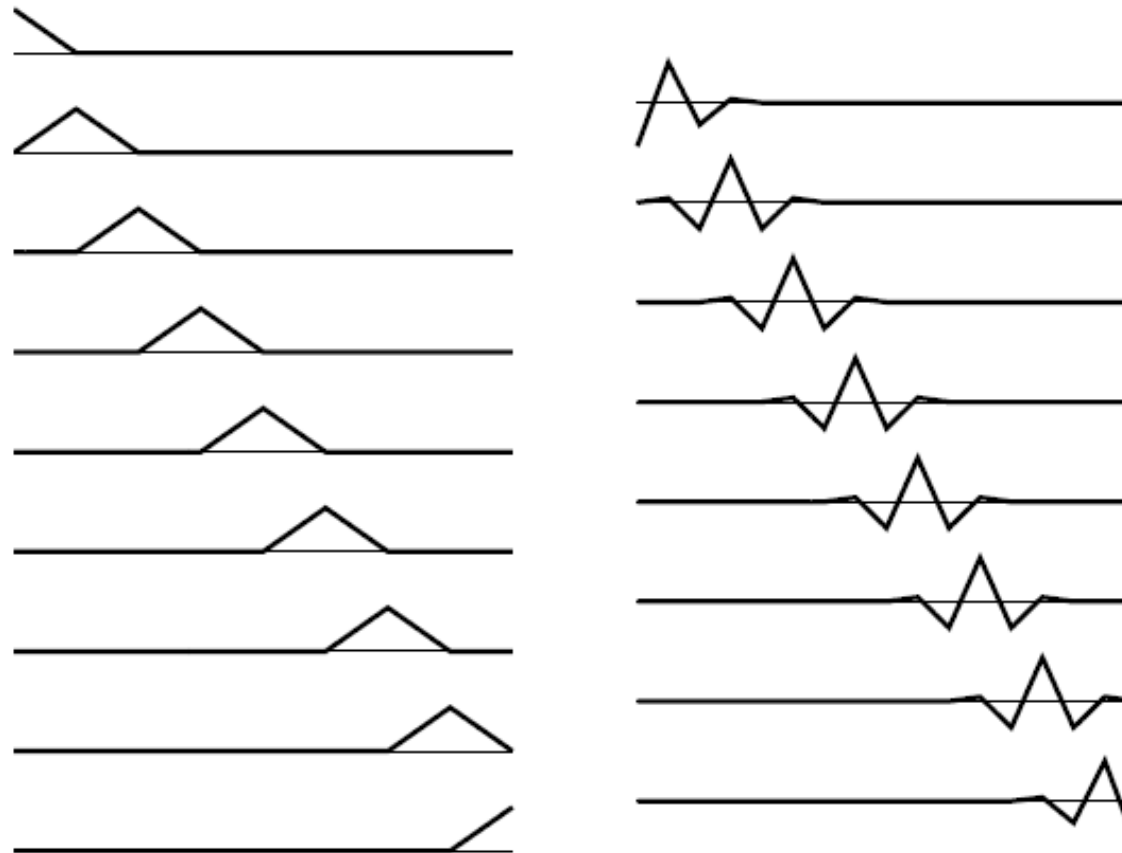
The wavelet transform can also be written in matrix form:

$$P^j = \begin{bmatrix} 1 & & & & & \\ 1 & & & & & \\ & 1 & & & & \\ & 1 & \cdot & & & \\ & & \cdot & \cdot & & \\ & & & \cdot & 1 & \\ & & & & 1 & \\ & & & & & 1 \\ & & & & & 1 \end{bmatrix} \quad Q^j = \begin{bmatrix} 1 & & & & & \\ -1 & & & & & \\ & 1 & & & & \\ & -1 & \cdot & & & \\ & & \cdot & \cdot & & \\ & & & \cdot & 1 & \\ & & & & -1 & \\ & & & & & 1 \\ & & & & & -1 \end{bmatrix}$$

The matrices are called synthesis matrices.



Linear B-spline wavelets



Linear B-spline wavelets

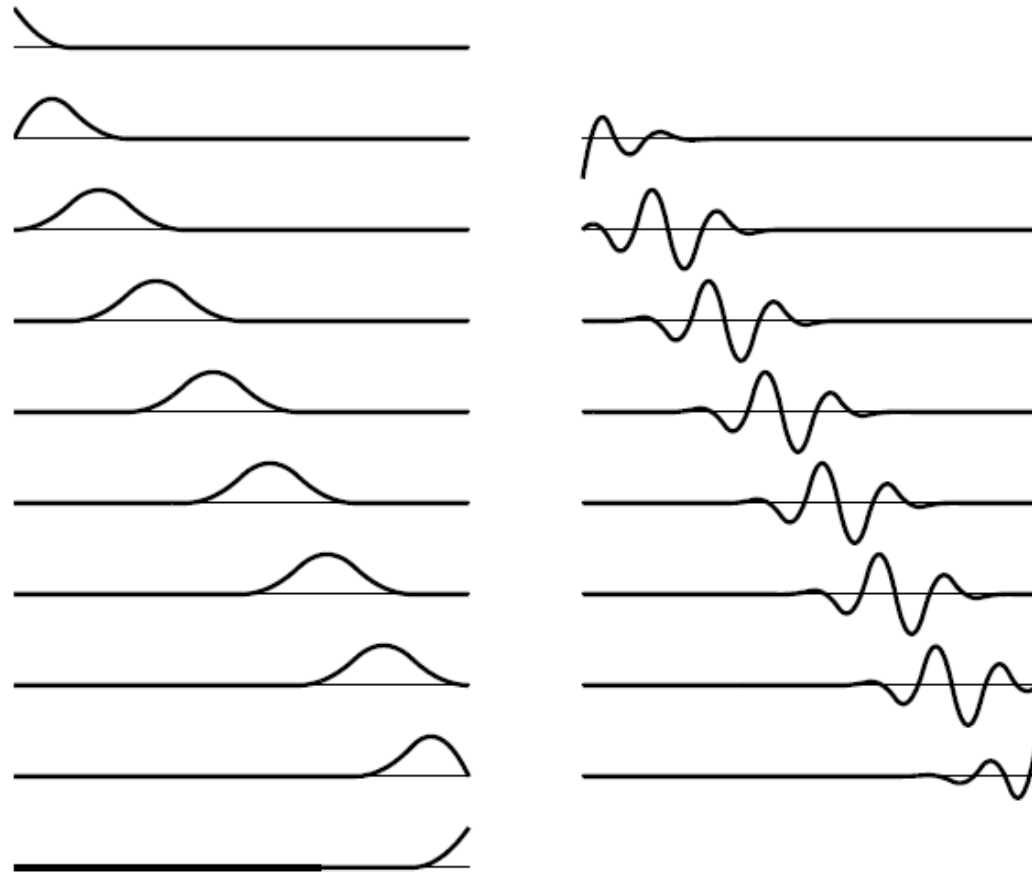
Synthesis matrices:

$$P^{j \geq 3} = \frac{1}{2} \begin{bmatrix} 2 & & & & & & & \\ 1 & 1 & & & & & & \\ & 2 & & & & & & \\ & 1 & 1 & & & & & \\ & & 2 & & & & & \\ & & 1 & \cdot & & & & \\ & & & \cdot & & & & \\ & & & & 1 & & & \\ & & & & 2 & & & \\ & & & & 1 & 1 & & \\ & & & & & 2 & & \end{bmatrix}$$

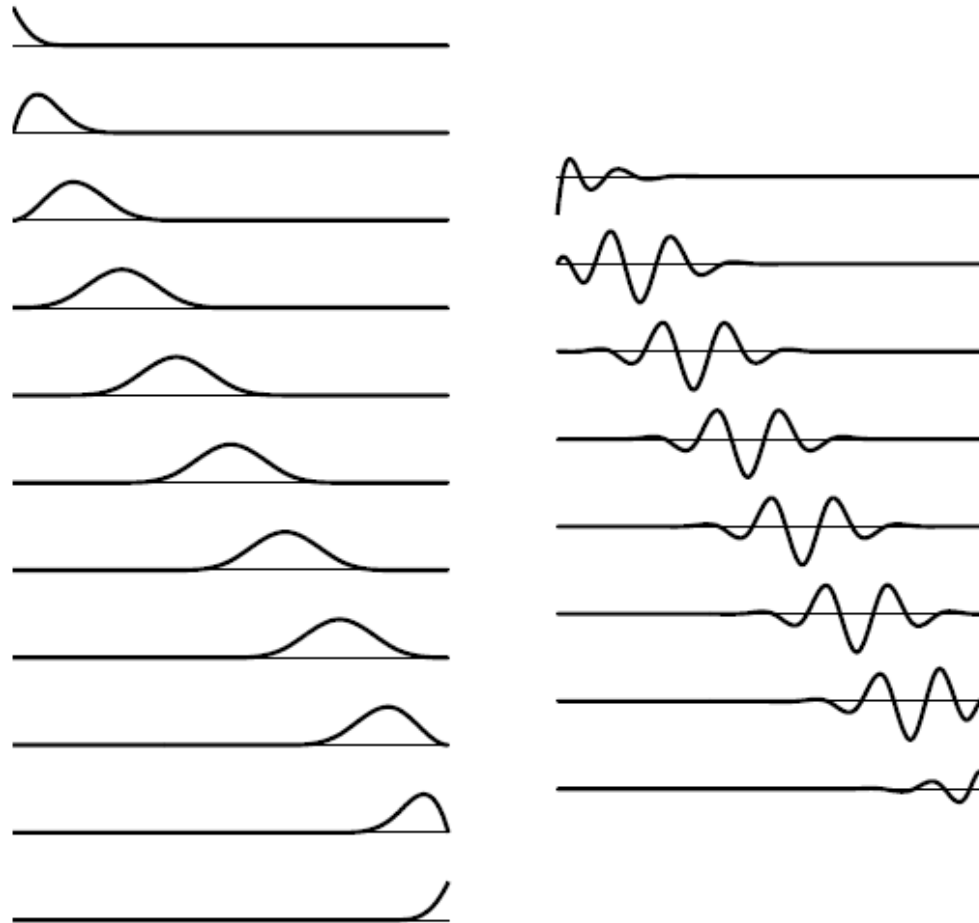
$$Q^{j \geq 3} = \begin{bmatrix} -12 & & & & & & & \\ 11 & 1 & & & & & & \\ -6 & -6 & & & & & & \\ 1 & 10 & 1 & & & & & \\ & -6 & -6 & & & & & \\ & 1 & 10 & & & & & \\ & & -6 & \cdot & & & & \\ & & 1 & \cdot & & 1 & & \\ & & & \cdot & & -6 & & \\ & & & & 10 & 1 & & \\ & & & & -6 & -6 & & \\ & & & & 1 & 11 & & \\ & & & & & -12 & & \end{bmatrix}$$



Quadratic B-spline wavelets



Cubic B-spline wavelets



Cubic B-spline wavelets

Synthesis matrices:

$$P^{j \geq 3} = \frac{1}{16} \begin{bmatrix} 16 & & & & & & & & & & & & & & & \\ & 8 & & & & & & & & & & & & & & \\ & & 12 & 4 & & & & & & & & & & & & \\ & & & 3 & 11 & 2 & & & & & & & & & & \\ & & & & 8 & 8 & & & & & & & & & & \\ & & & & 2 & 12 & 2 & & & & & & & & & \\ & & & & & 8 & 8 & & & & & & & & & \\ & & & & & 2 & 12 & & & & & & & & & \\ & & & & & & 8 & \cdot & & & & & & & & \\ & & & & & & 2 & \cdot & & 2 & & & & & & \\ & & & & & & & \cdot & & 8 & & & & & & \\ & & & & & & & & 12 & 2 & & & & & & \\ & & & & & & & & 8 & 8 & & & & & & \\ & & & & & & & & 2 & 11 & 3 & & & & & \\ & & & & & & & & & 4 & 12 & & & & & \\ & & & & & & & & & & 8 & 8 & & & & \\ & & & & & & & & & & & 16 & & & & \end{bmatrix}$$



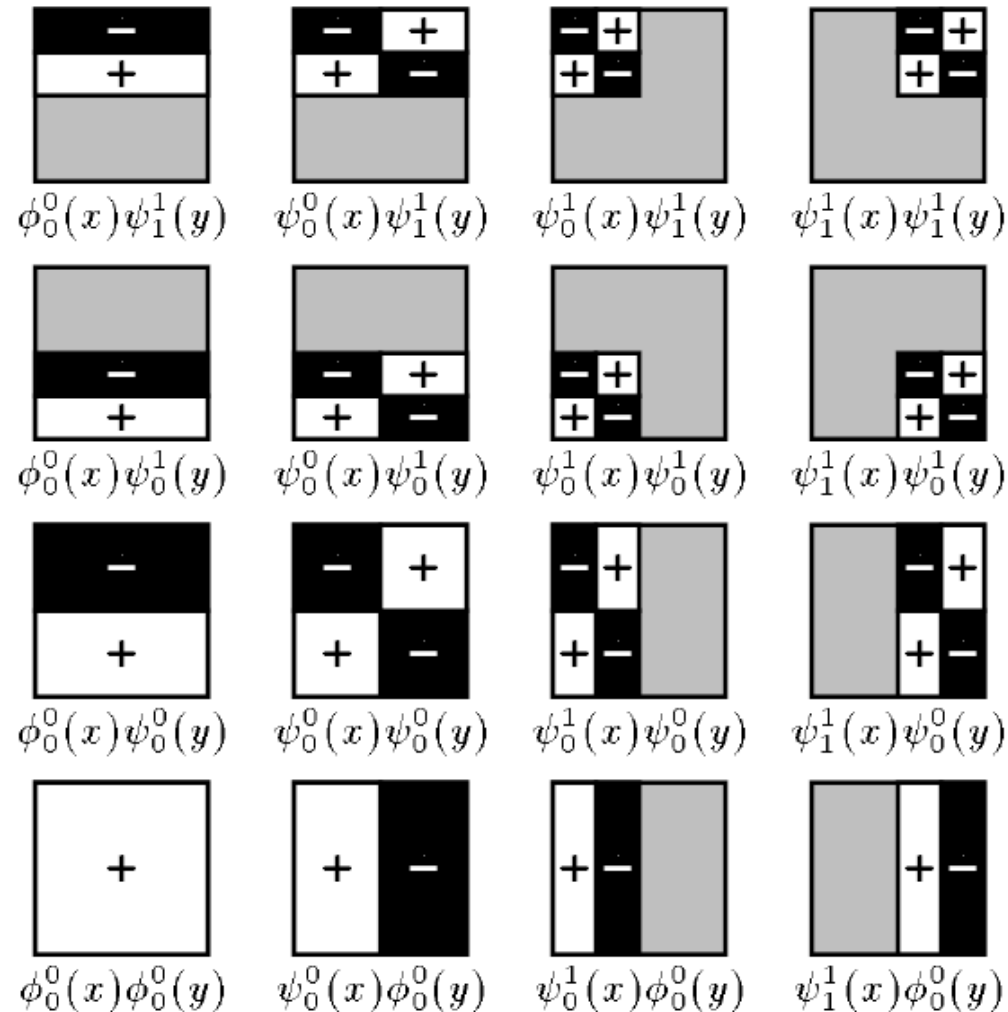
2D Haar wavelet transform

- 2D basis and wavelet functions are **tensor products** of 1D basis and wavelet functions.

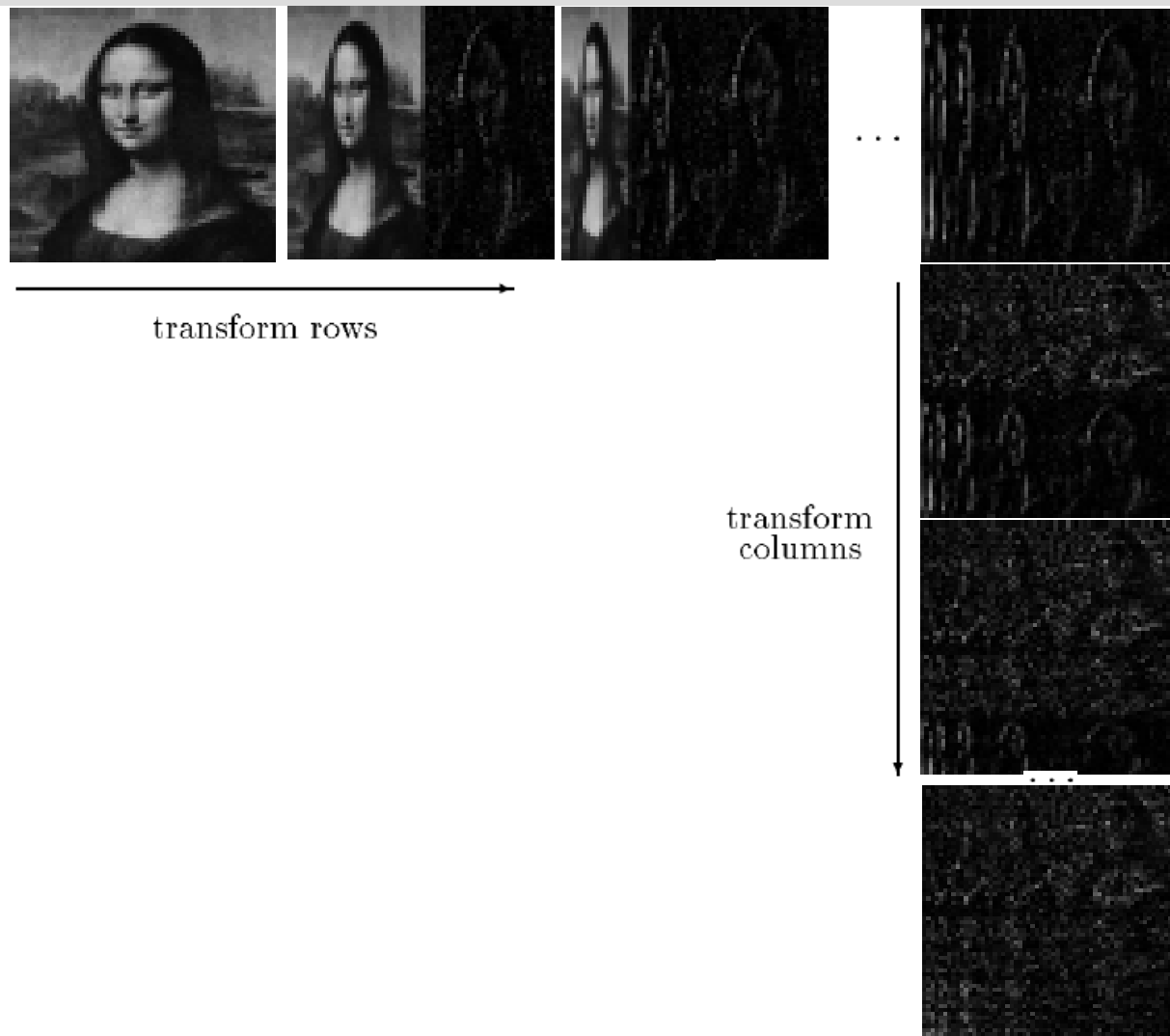


2D Haar wavelet transform

Basis:



2D Haar wavelet transform



2D Haar wavelet transform

Alternative construction:

Use 2D basis function

$$\phi\phi(x, y) := \phi(x) \phi(y)$$

and three 2D wavelet functions

$$\phi\psi(x, y) := \phi(x) \psi(y)$$

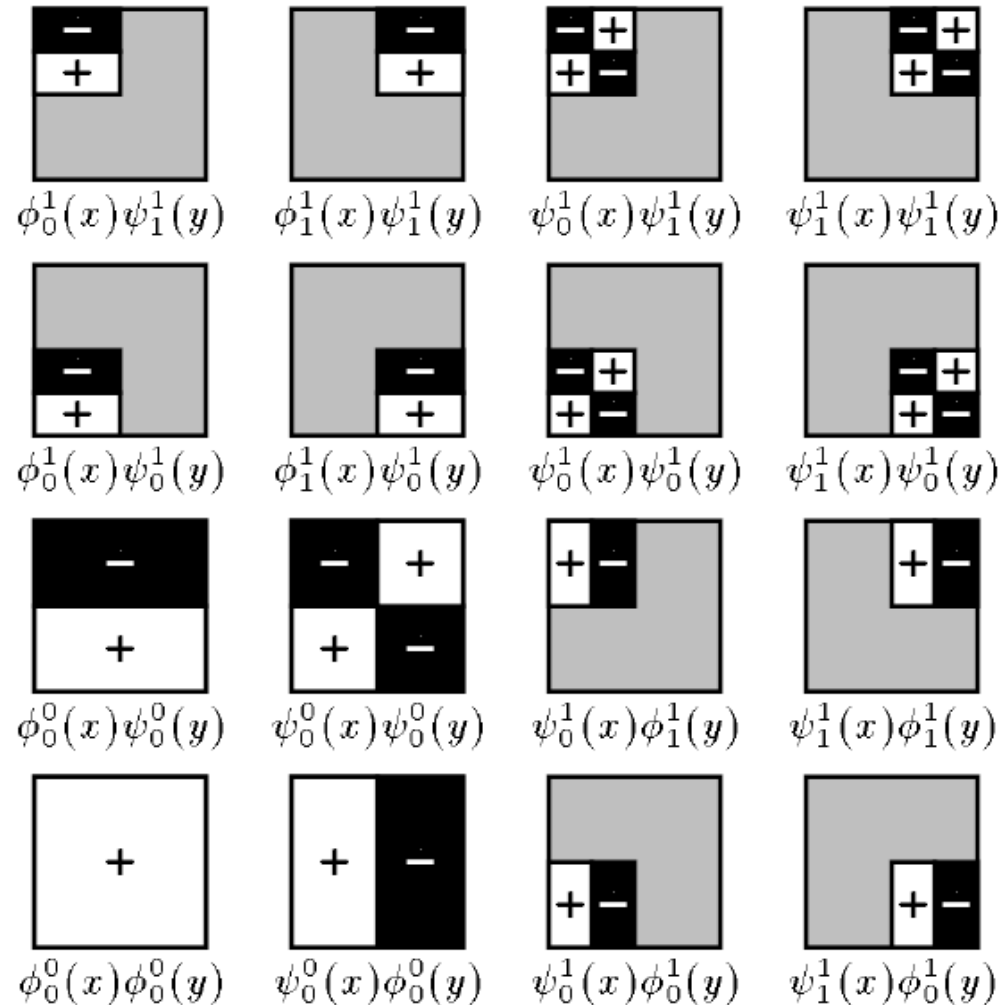
$$\psi\phi(x, y) := \psi(x) \phi(y)$$

$$\psi\psi(x, y) := \psi(x) \psi(y)$$

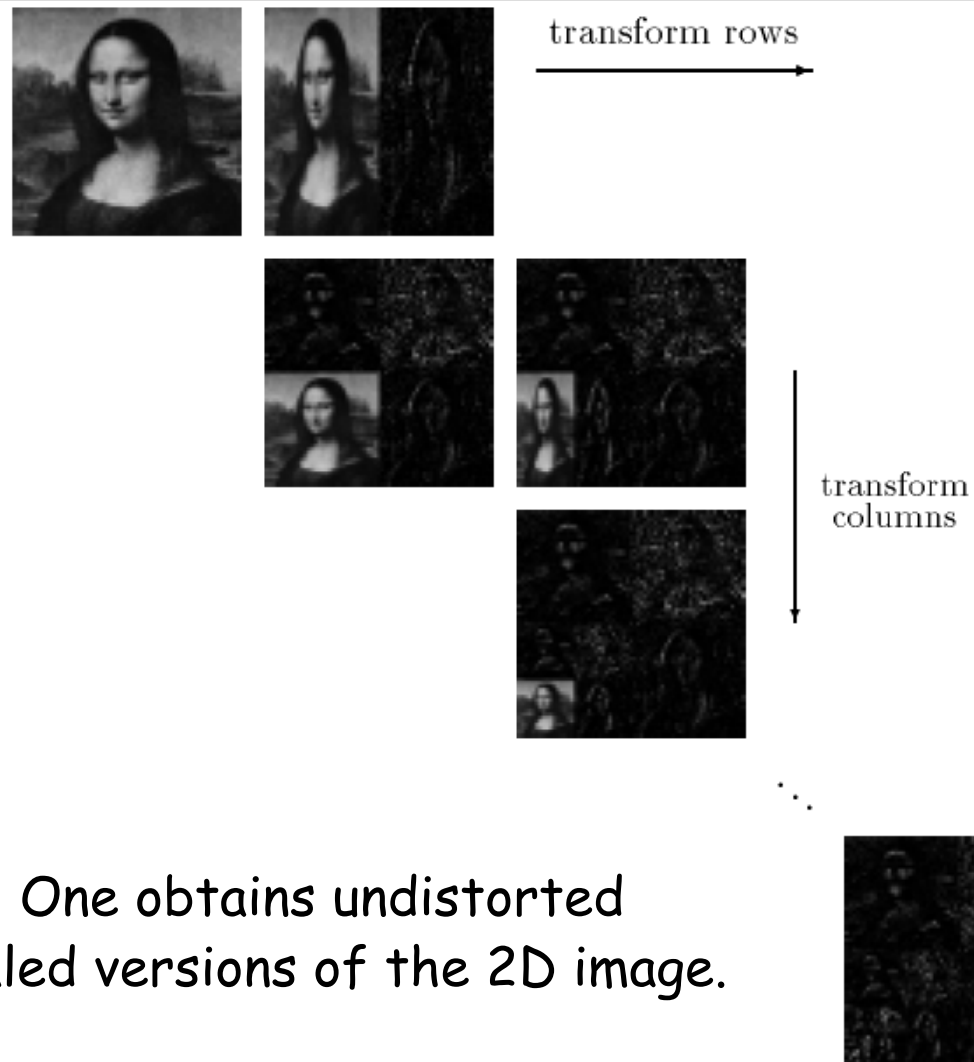


2D Haar wavelet transform

Basis:



2D Haar wavelet transform



Advantage: One obtains undistorted
downscaled versions of the 2D image.



2D wavelet transform in RGB space

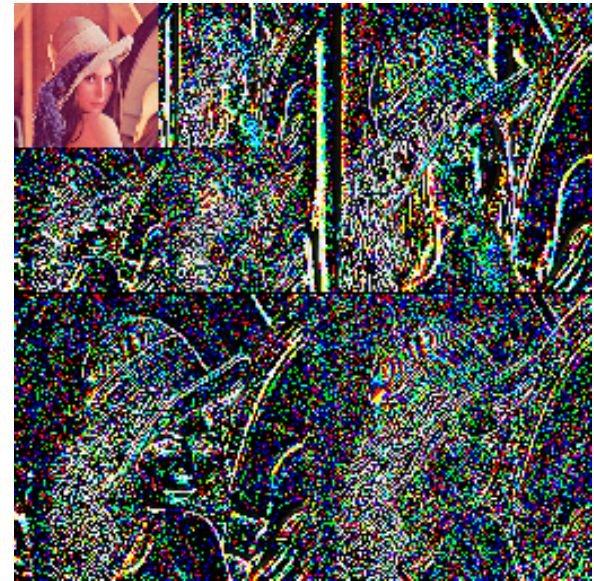
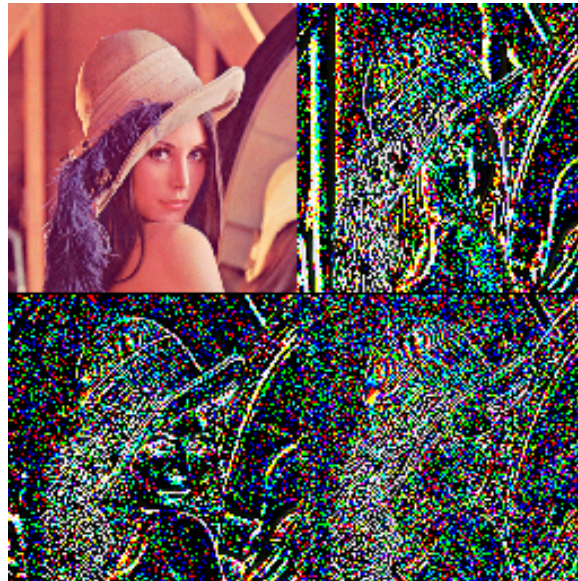


Image compression

- Loss-less compression
 - Do not store detail coefficients that are 0.
 - Constant regions are stored by 1 value only.
- Lossy compression
 - Set detail coefficients with small absolute values to 0.
 - A threshold determines the compression rate.



Image compression

Haar wavelets:



100%

Error: 0%



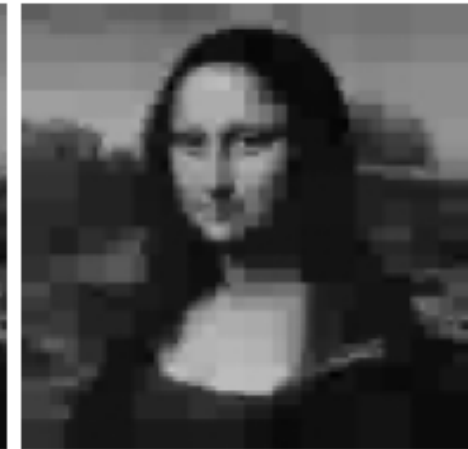
21%

5%



4%

10%



1%

15%



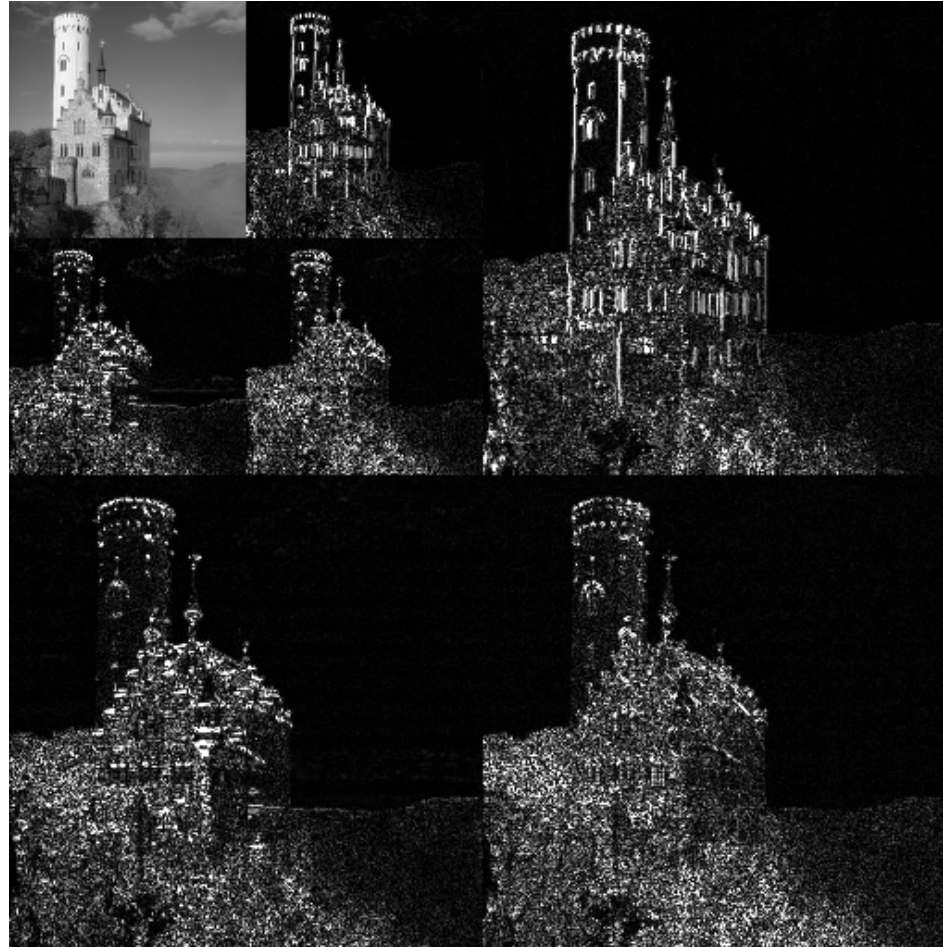
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Image compression

JPEG 2000: Cohen-Daubechies-Feauveau wavelets



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Image compression

JPEG 2000: lossy compression leads to blurring.



1.4.3 Multiresolution Modeling



References

- Multiresolution Techniques.

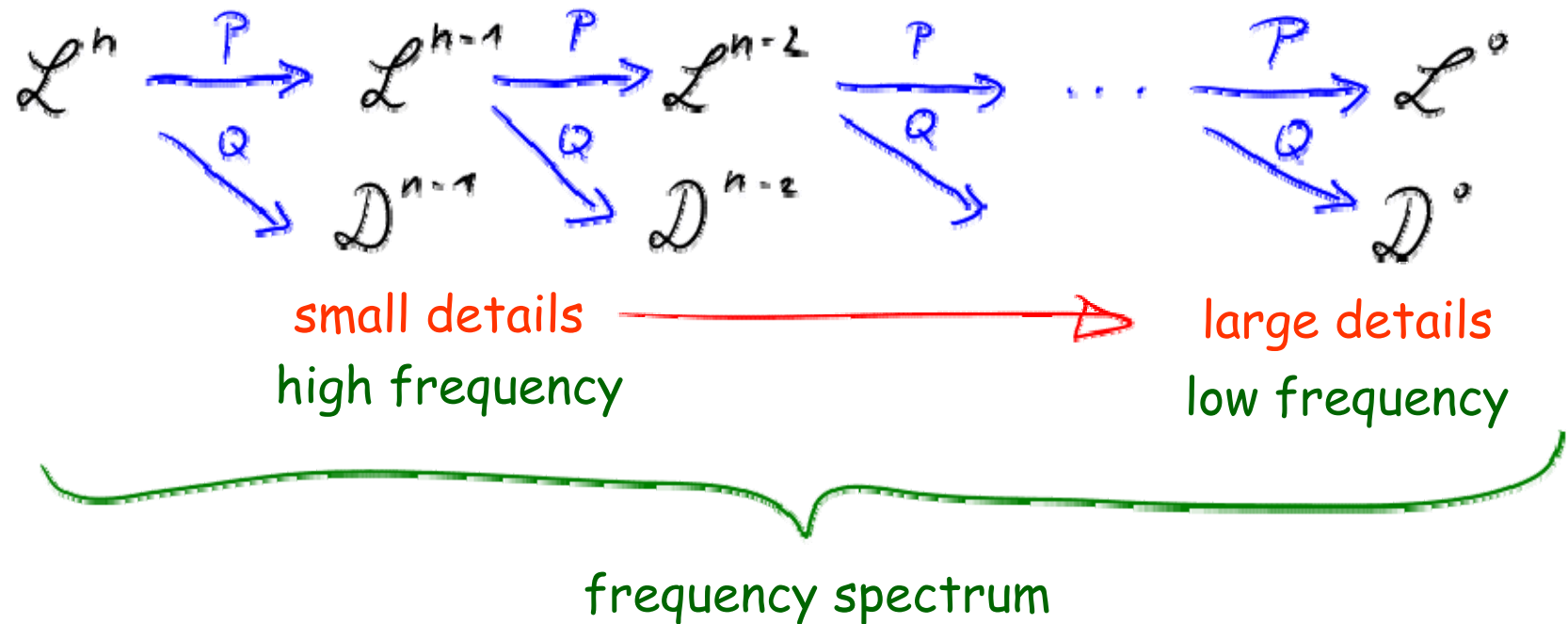
Leif P. Kobbelt.

In: Handbook of Computer Aided Geometric Design,
G. Farin, J. Hoschek, M-S. Kim (eds.), Elsevier, 2002.



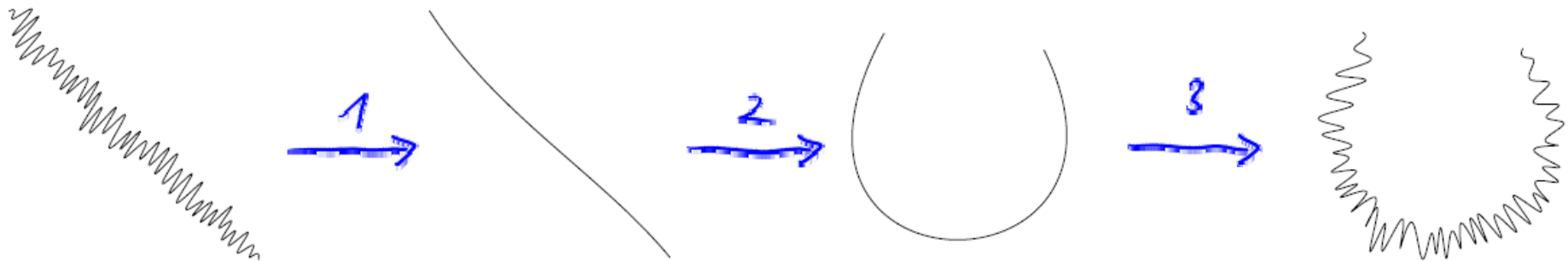
Multiresolution modeling

- Filter bank



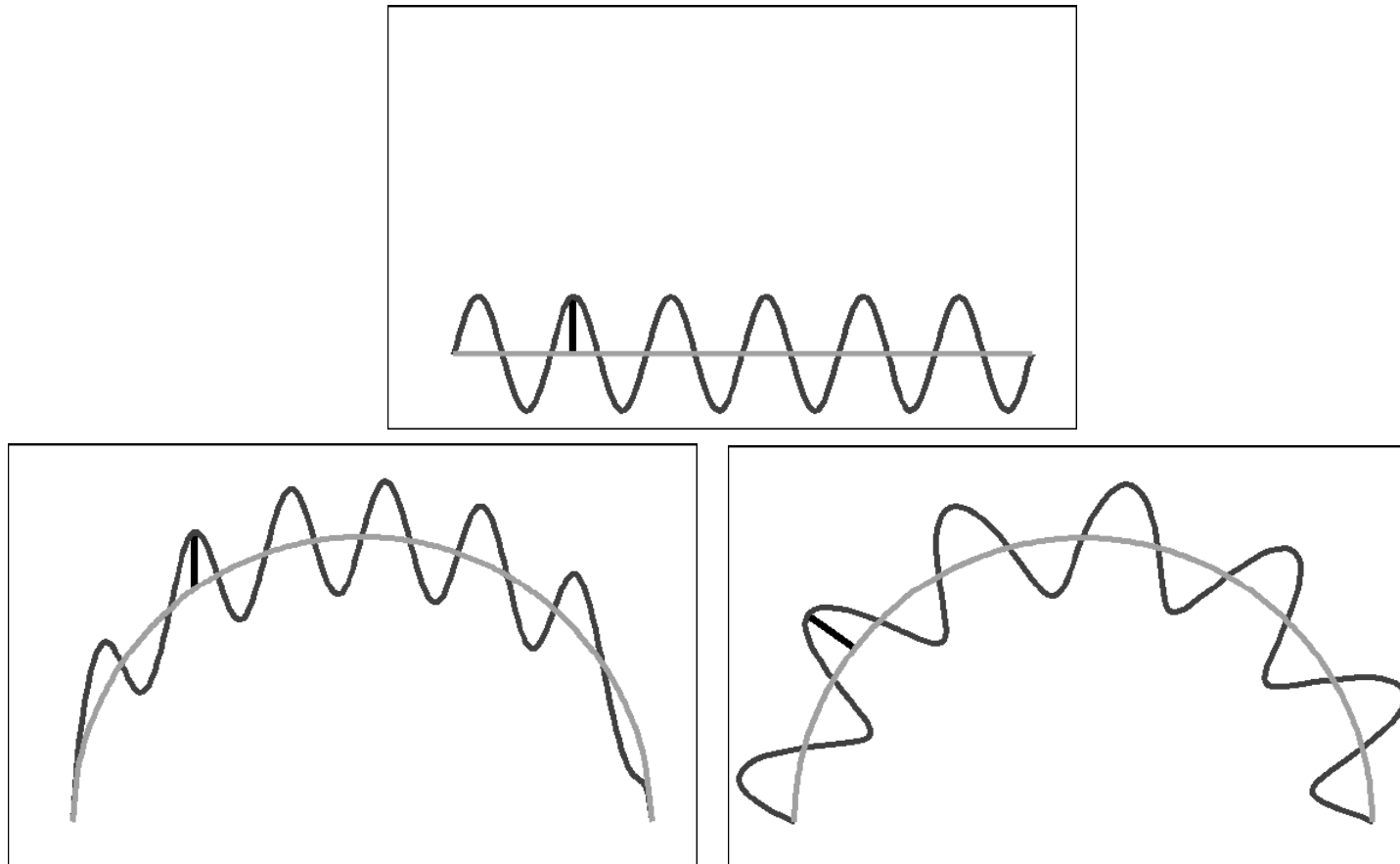
Multiresolution modeling: Example 1

1. Remove high-frequency detail by going to coarse resolution.
2. Perform modeling at coarse resolution by interactively changing the shape.
3. Reinsert high-frequency detail by going to fine resolution.



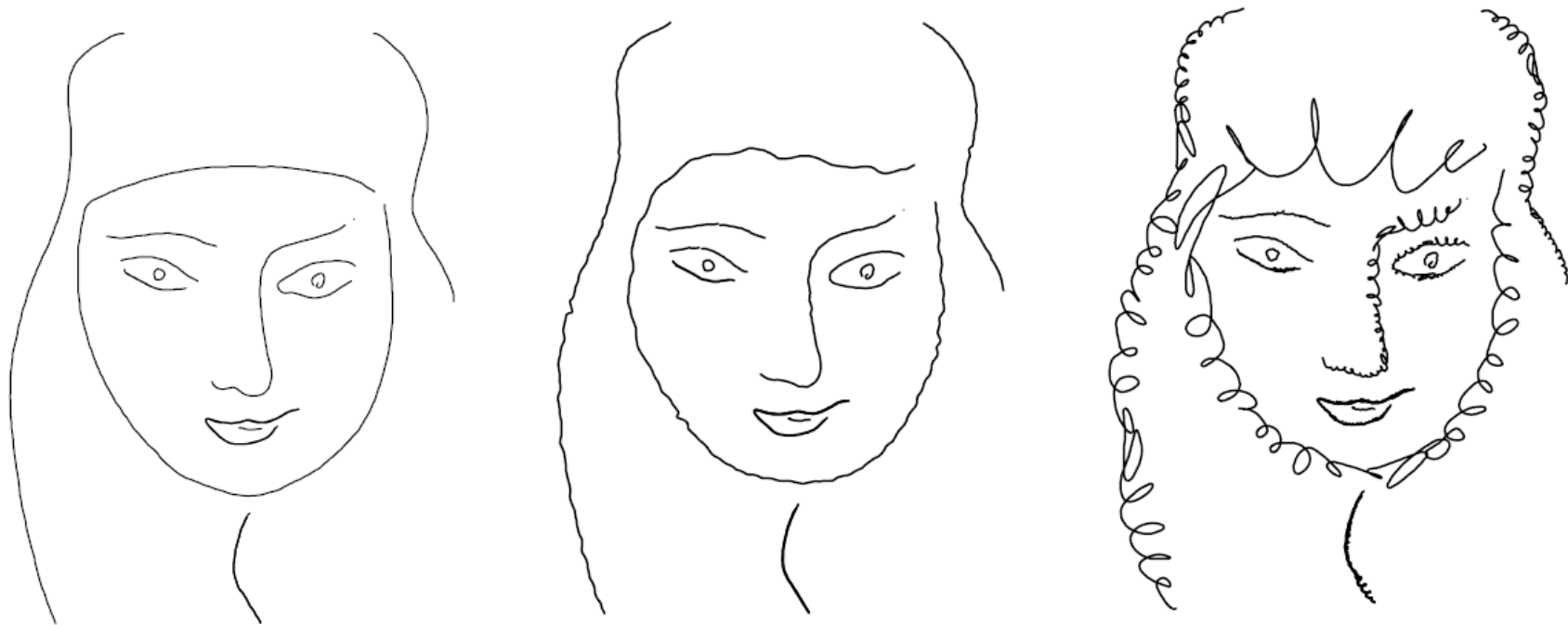
Multiresolution modeling: Example 1

- Global vs. local details:



Multiresolution modeling: Example 2

- Change the character of an object by replacing detail coefficients with new ones, e.g., by replacing small high-frequency details with large ones.



Multiresolution modeling: Example 3

- Performing interactive changes of the shape at different levels of resolution.



fine level



intermediate level

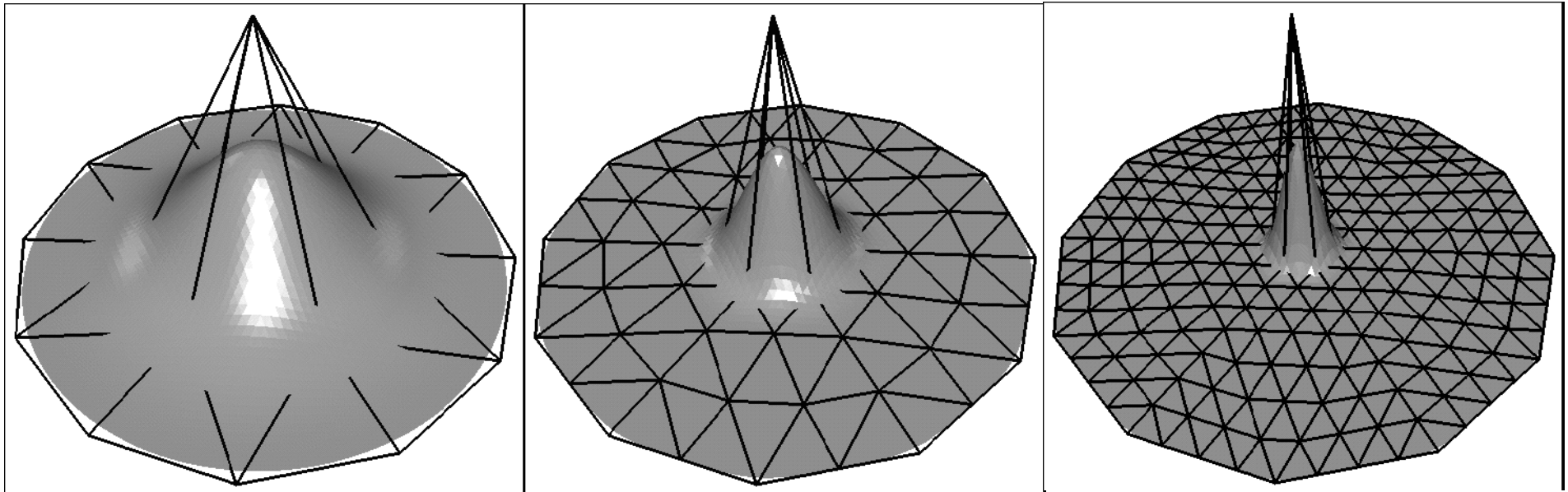


coarse level



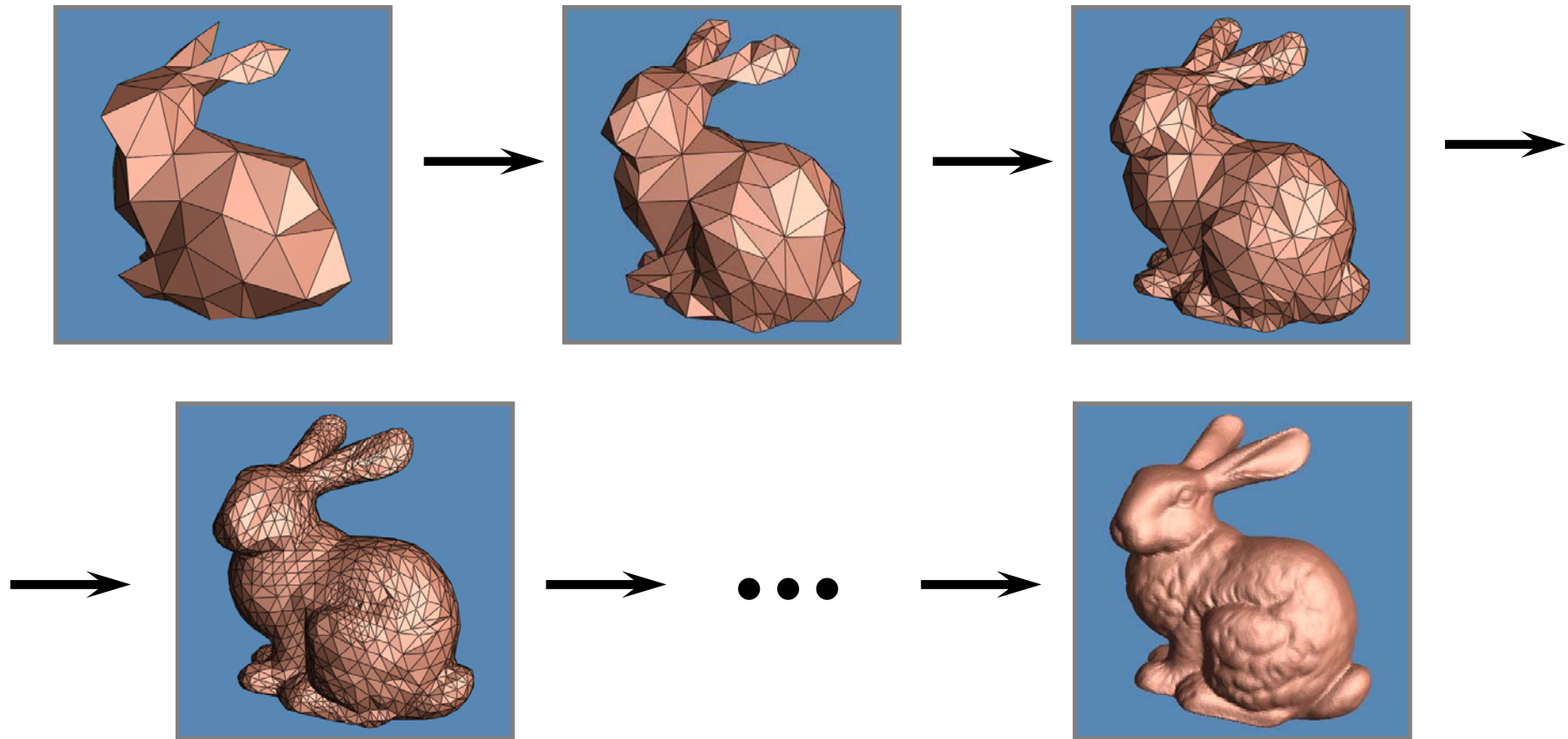
Multiresolution modeling: Example 3

- The finer the level of resolution, the more local the change.



Applications

Hierarchy:



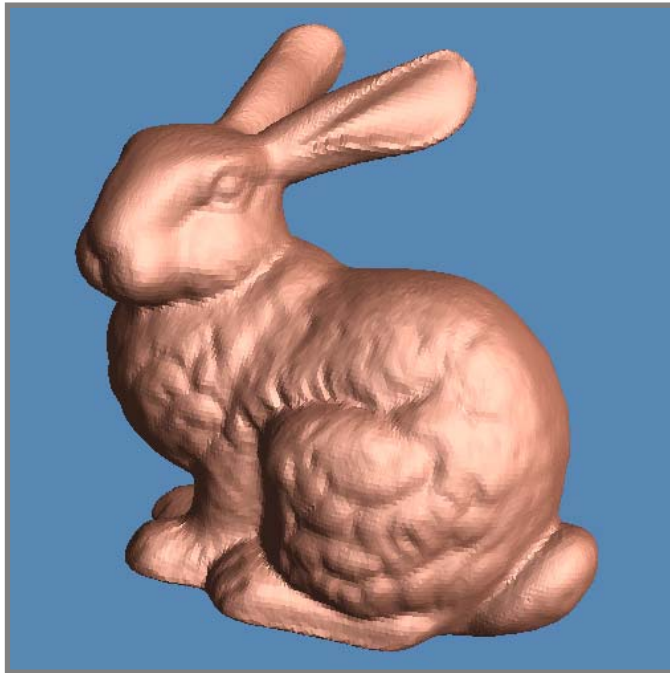
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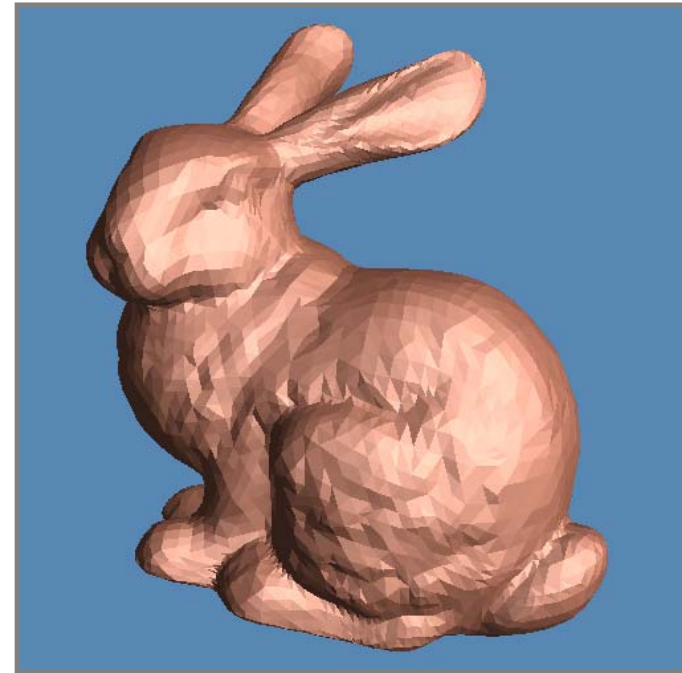
Compression

Reduction of storage requirements with error bounds:



~70,000 faces

$\epsilon < 0.8\%$



~11,000 faces



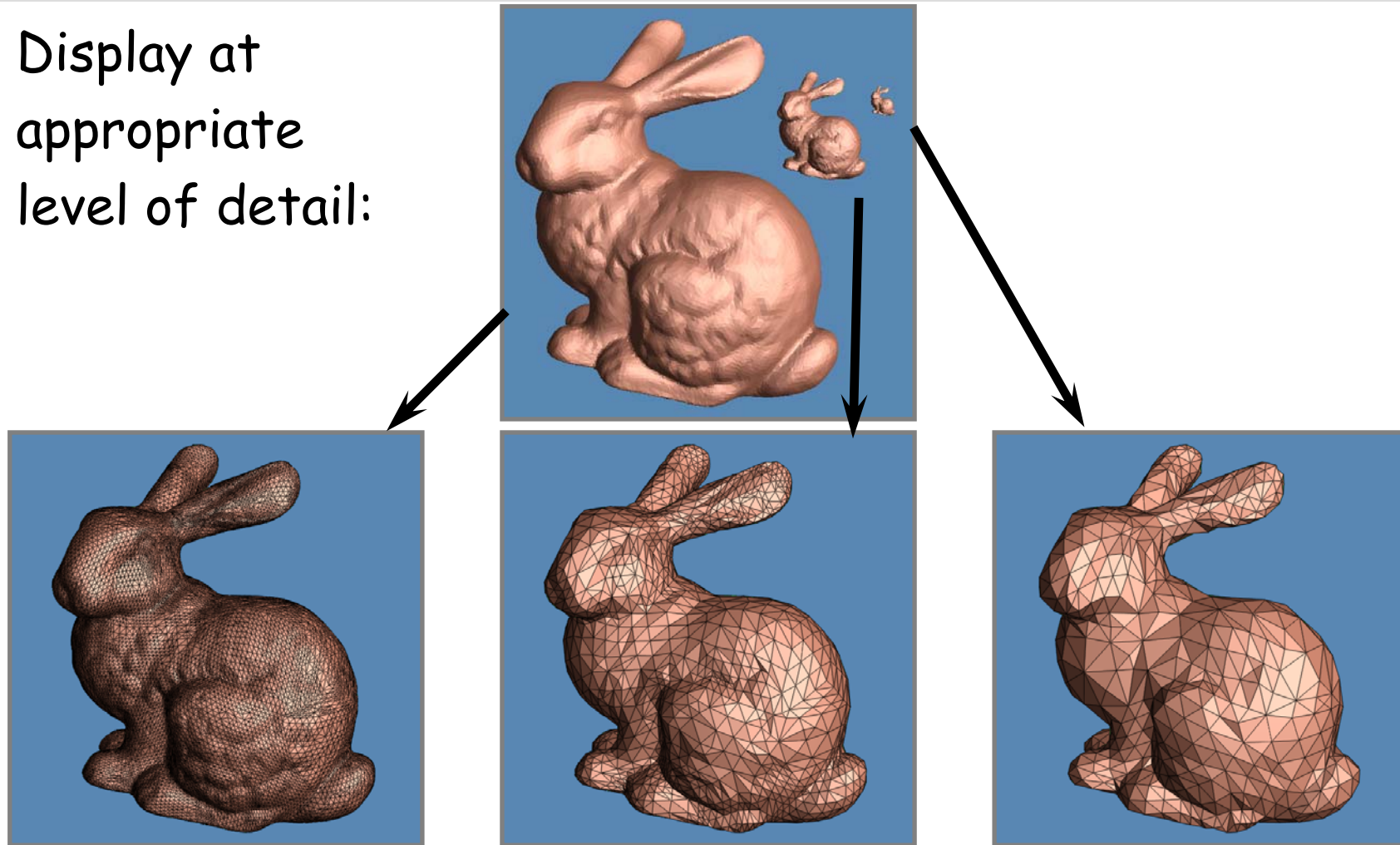
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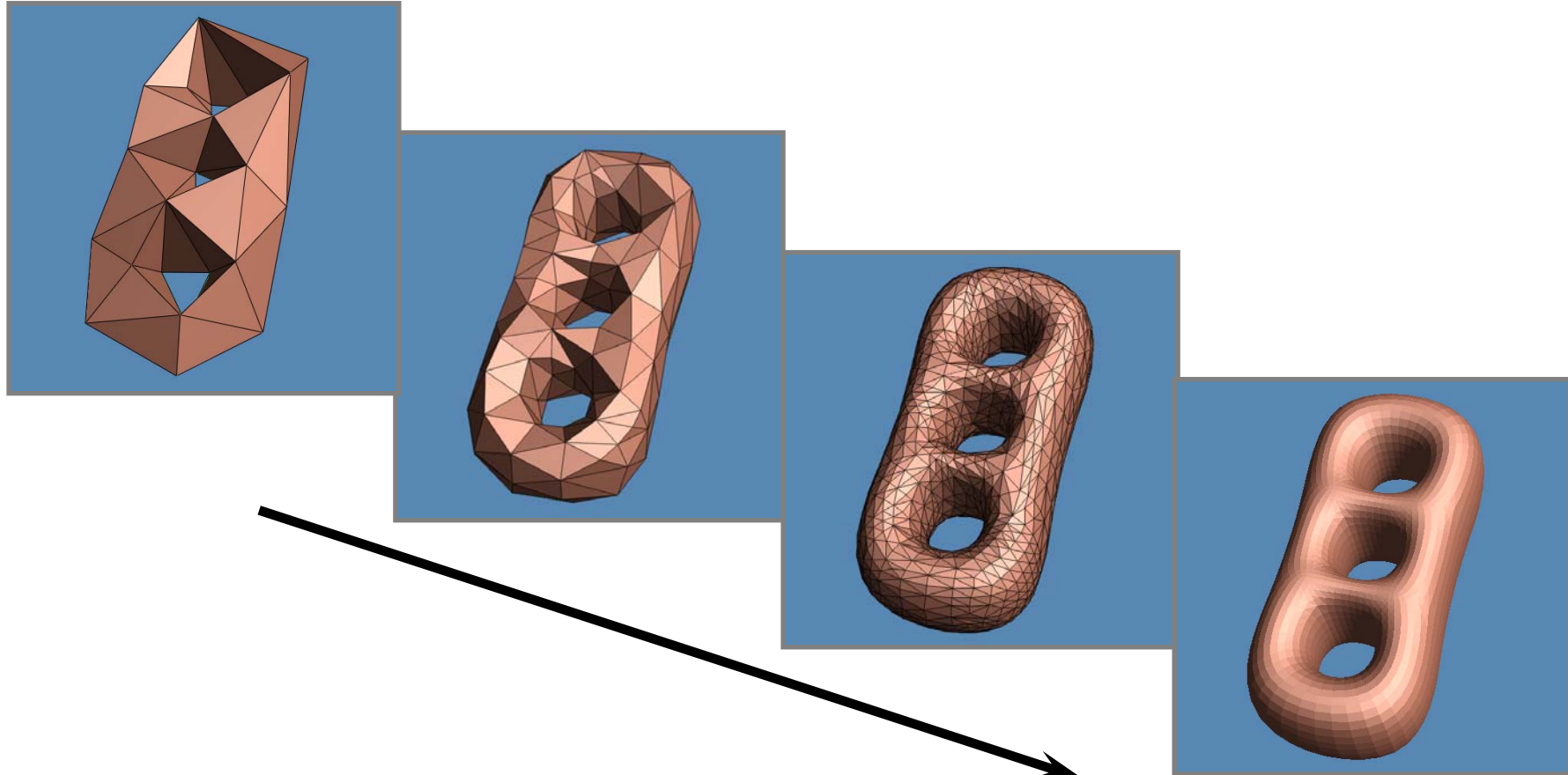
Level of detail control

Display at
appropriate
level of detail:



Progressive transmission and rendering

Progressive refinement of mesh over time:



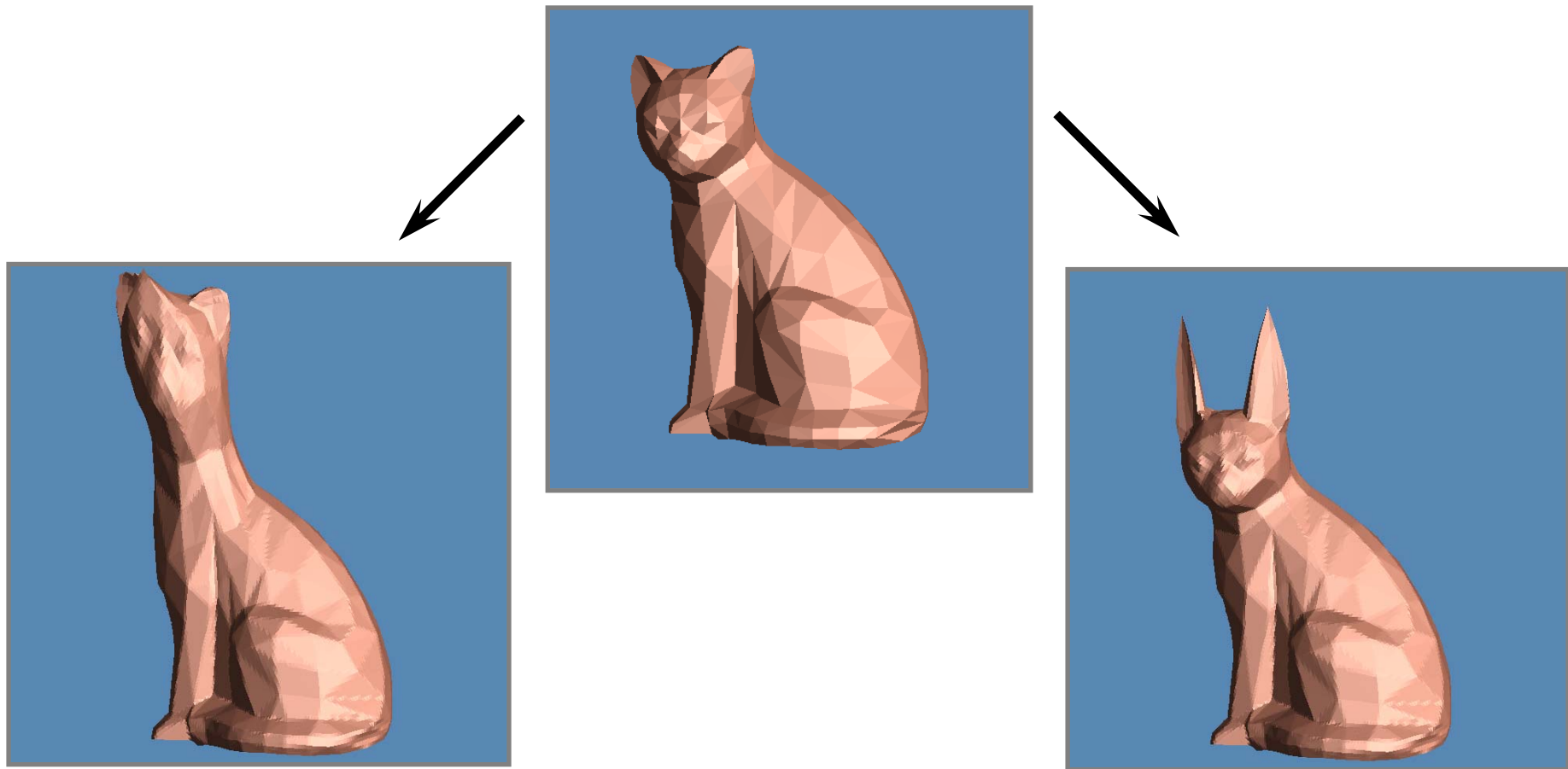
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Multiresolution editing

Shape editing at different LODs:



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