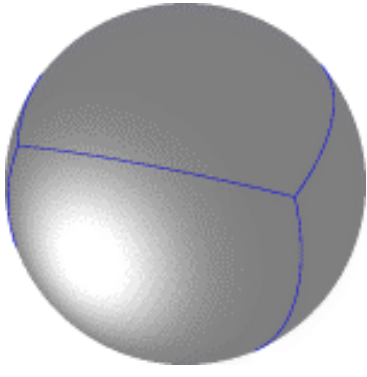
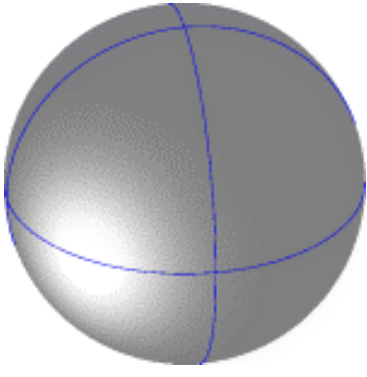
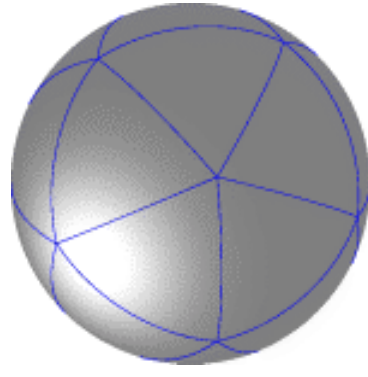

Optimal Triangular Haar Bases for Spherical Data

- Triangular Haar basis
- Orthogonality properties
- Previous bases
- New families of nearly-orthogonal bases
- Optimal basis
- Results

Triangular Haar basis: domain definition

Domain mesh = recursive 4-1 split of a base mesh

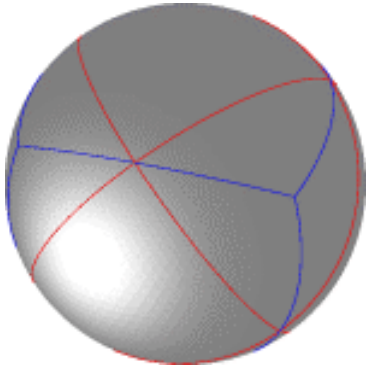
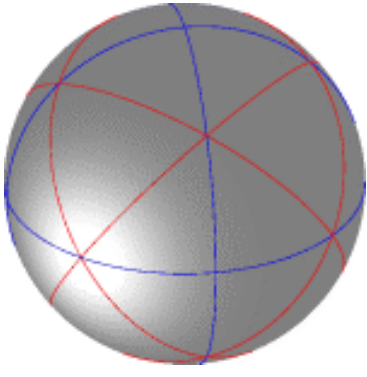
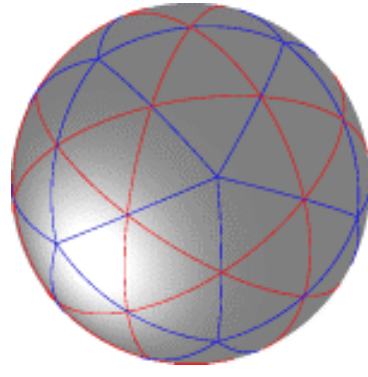
Level 0

σ (area (Δ_i))	0	0	0
			
	tetraedron	octaedron	icosaedron

Triangular Haar basis: domain definition

Domain mesh = recursive 4-1 split of a base mesh

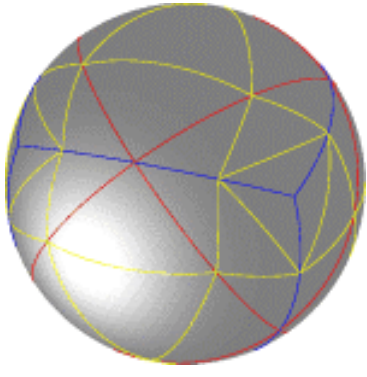
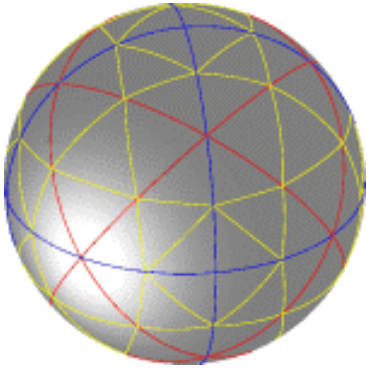
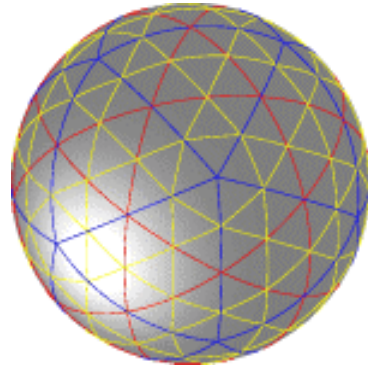
Level 1

σ (area (Δ_i))	0,1443	0,0412	0,0093
			
	tetraedron	octaedron	icosaedron

Triangular Haar basis: domain definition

Domain mesh = recursive 4-1 split of a base mesh

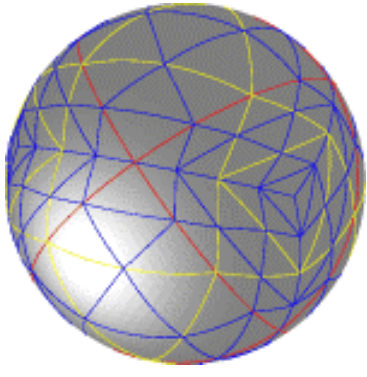
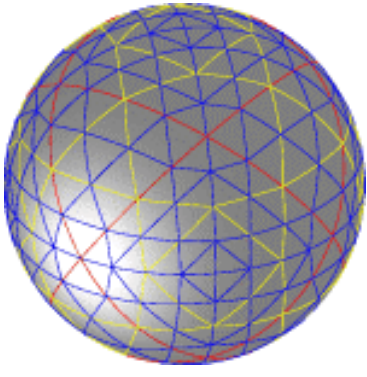
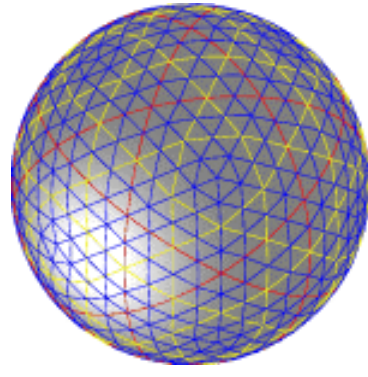
Level 2

σ (area (Δ_i))	0.0783	0.0214	0.0048
			
	tetraedron	octaedron	icosaedron

Triangular Haar basis: domain definition

Domain mesh = recursive 4-1 split of a base mesh

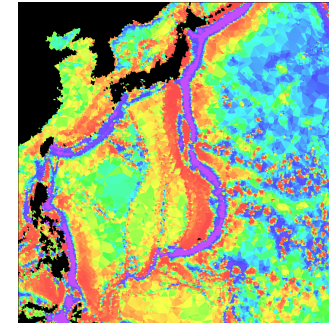
Level 3

σ (area (Δ_i))	0.0394	0.0107	0.0024
			
	tetraedron	octaedron	icosaedron

Triangular Haar Basis: Data type

Piecewise constant data: one value per face

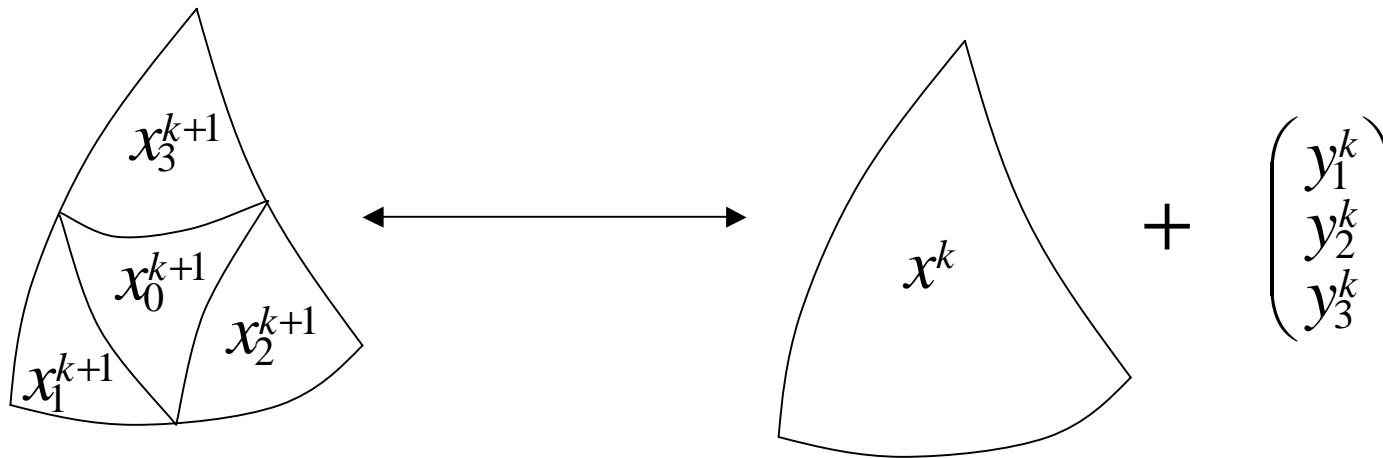
- Enough for large & complex (non-regular) data sets



- Comparison of Haar and more smoother bases in Schroeder/Sweldens
SIGGRAPH'94

=> Smoother bases are not better for non-regular data sets

Analysis & Synthesis: local scheme



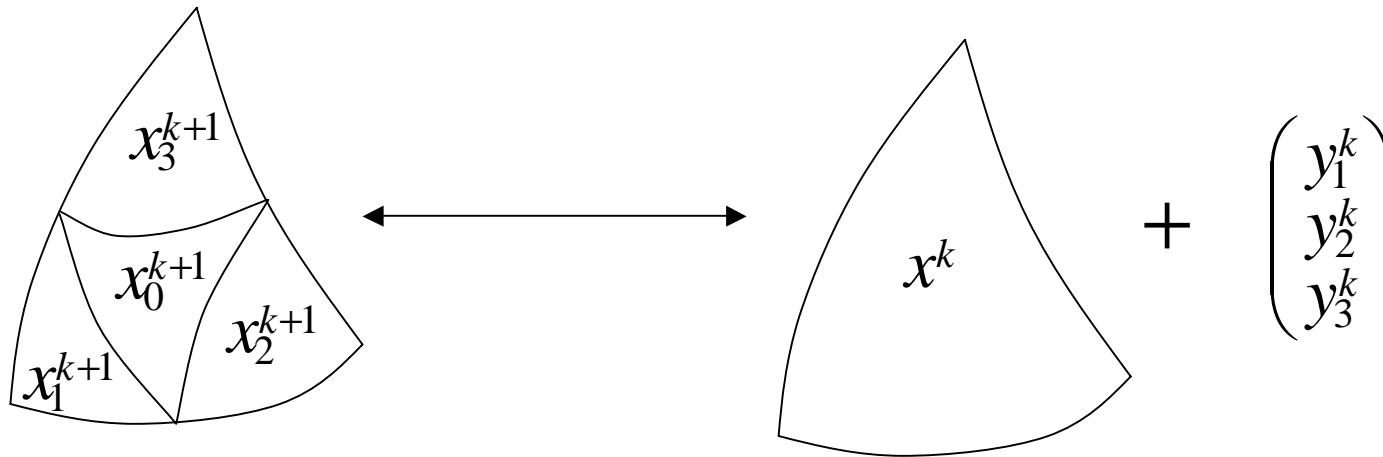
Local reconstruction

$$\begin{pmatrix} x_0^{k+1} \\ x_1^{k+1} \\ x_2^{k+1} \\ x_3^{k+1} \end{pmatrix} = R \begin{pmatrix} x_0^k \\ y_1^k \\ y_2^k \\ y_3^k \end{pmatrix}$$

Local decomposition

$$\begin{pmatrix} x_0^k \\ y_1^k \\ y_2^k \\ y_3^k \end{pmatrix} = R^{-1} \begin{pmatrix} x_0^{k+1} \\ x_1^{k+1} \\ x_2^{k+1} \\ x_3^{k+1} \end{pmatrix}$$

Analysis & Synthesis: functional point-of-view



$$x_0^{k+1}\Phi_0^{k+1} + x_1^{k+1}\Phi_1^{k+1} + x_2^{k+1}\Phi_2^{k+1} + x_3^{k+1}\Phi_3^{k+1}$$

=

$$x^k\Phi^k + y_1^k\Psi_1^k + y_2^k\Psi_2^k + y_3^k\Psi_3^k$$

Analysis & Synthesis: global scheme

Global decomposition (analysis)

```
for k=K-1 to 0
  for all triangles  $T_k$  at level k
    perform local decomposition in  $T_k$  // previous slide
     $(x_k, y_{1k}, y_{2k}, y_{3k}, y_{4k}) \rightarrow (x_{0k+1}, x_{1k+1}, x_{2k+1}, x_{3k+1})$ 
```

Global reconstruction (synthesis)

```
for k=0 to K-1
  for all triangles  $T_k$  at level k
    perform local reconstruction in  $T_k$  // previous slide
     $(x_{0k+1}, x_{1k+1}, x_{2k+1}, x_{3k+1}) \rightarrow (x_k, y_{1k}, y_{2k}, y_{3k}, y_{4k})$ 
```

Analysis & Synthesis: Complexity

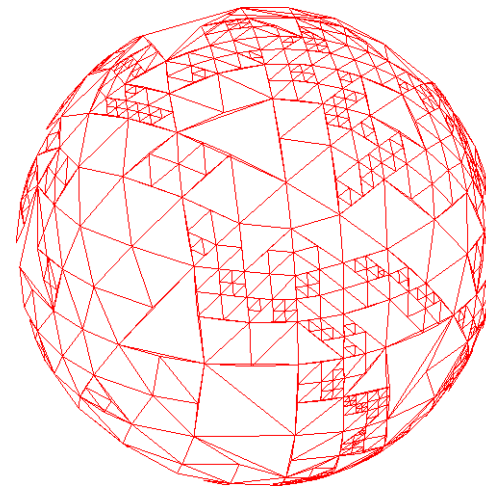
- Hierarchy initialization:
 - Linear time required
 - Linear space required
- Both global decomposition and reconstruction:
 - Linear time required
 - Linear space required
- Sort wavelet coefficients: $N \log N$

Application: progressive transmission

Transmit & insert wavelet coefficients by decreasing order of magnitude



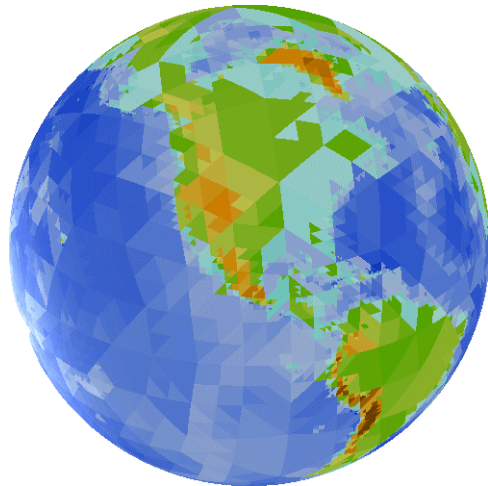
1000 wav. Coeff.



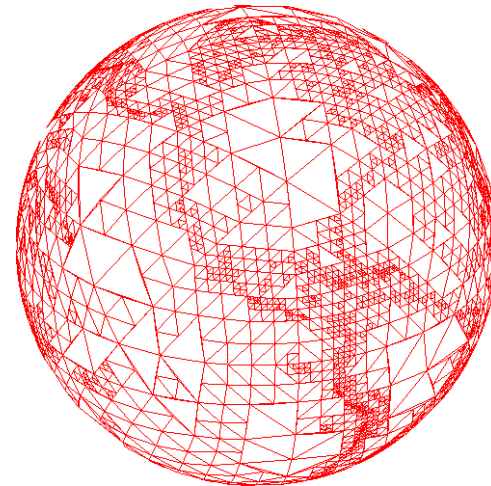
Rel. L2 error = 0.199

Application: progressive transmission

Transmit & insert wavelet coefficients by decreasing order of magnitude



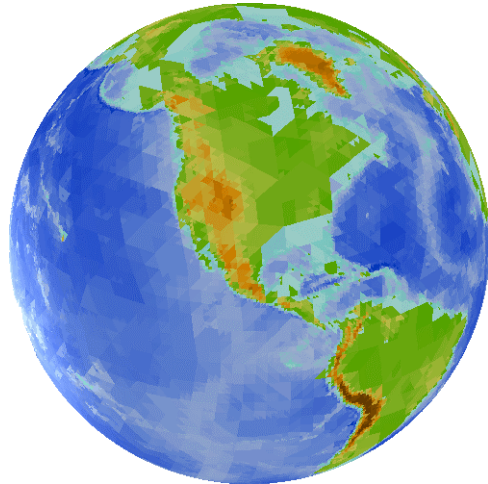
5000 wav. Coeff.



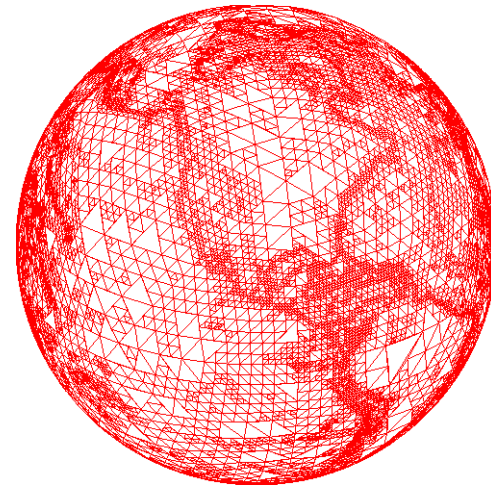
Rel. L2 error = 0.191

Application: progressive transmission

Transmit & insert wavelet coefficients by decreasing order of magnitude



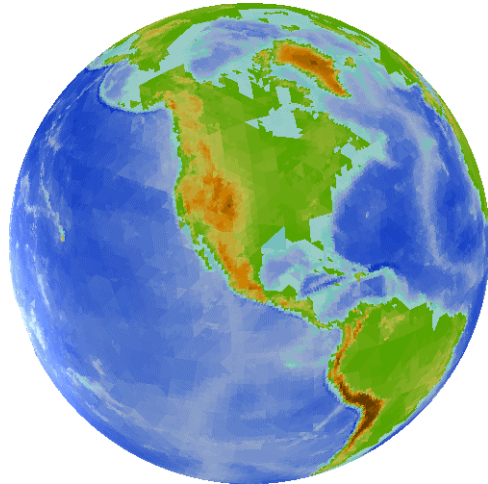
25000 wav. Coeff.



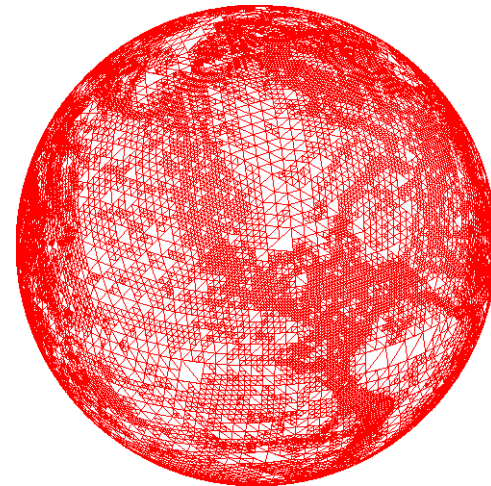
Rel. L2 error = 0.040

Application: progressive transmission

Transmit & insert wavelet coefficients by decreasing order of magnitude



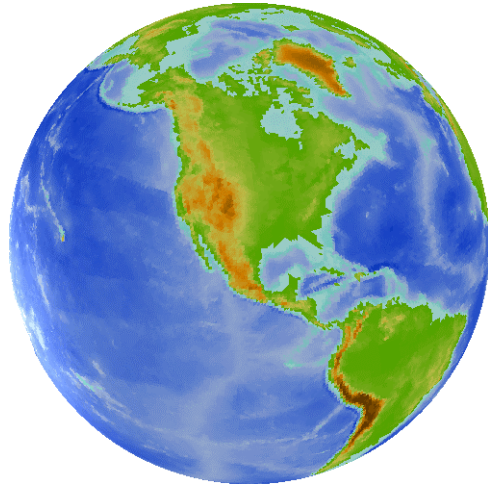
50000 wav. Coeff.



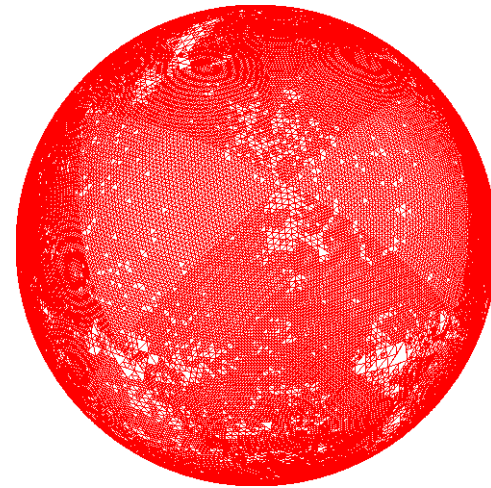
Rel. L2 error = 0.017

Application: progressive transmission

Transmit & insert wavelet coefficients by decreasing order of magnitude



100000 wav. Coeff.



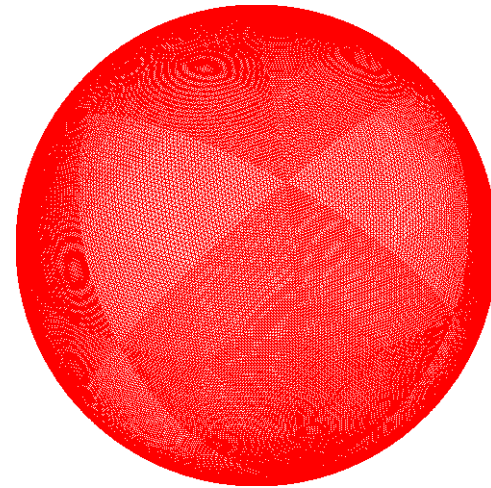
Rel. L2 error = 0.002

Application: progressive transmission

Transmit & insert wavelet coefficients by decreasing order of magnitude



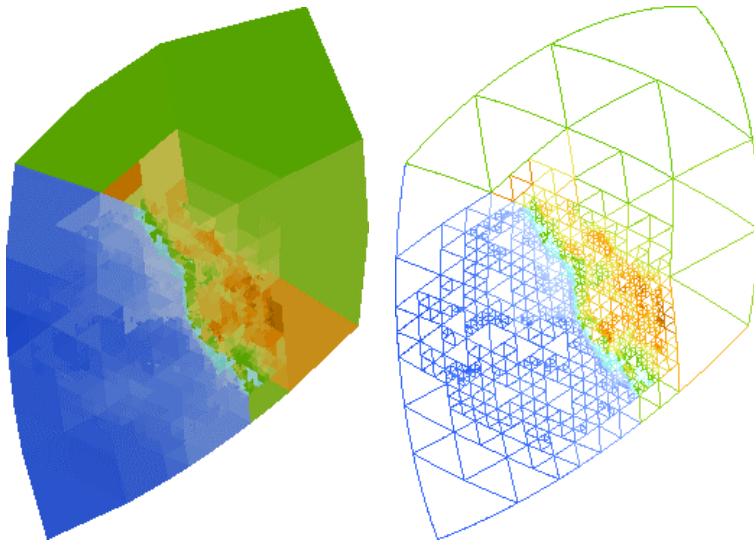
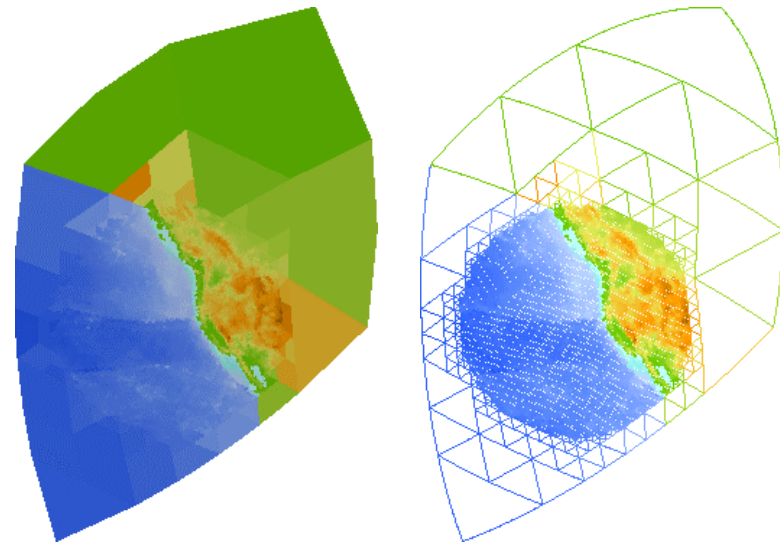
All wav. Coeff.



Rel. L2 error = 0

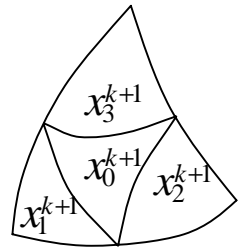
Application: view-dependent reconstruction

Selection of all wav. coeff.
inside a region



Selection of biggest wav. coeff.
inside a region

Orthogonality properties


$$= x^k \Phi^k + y_1^k \Psi_1^k + y_2^k \Psi_2^k + y_3^k \Psi_3^k$$

■ Semi-orthogonal:

- required: $\Phi^k \Psi_1^k = \Phi^k \Psi_2^k = \Phi^k \Psi_3^k = 0$
- prop:
 - discarding all three wavelet coefficients leads to best L2 approx.

■ Orthogonal:

- required: $\Psi_1^k \Psi_2^k = \Psi_1^k \Psi_3^k = \Psi_2^k \Psi_3^k = 0$ + semi-orthogonality
- prop:
 - discarding the any wavelet coefficients leads to best L2 approx.
 - exact L2 error using wavelet coefficients (squared rooted sum)
 - no matrix inversion needed

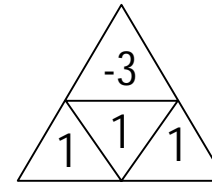
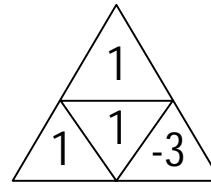
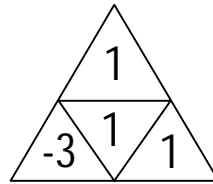
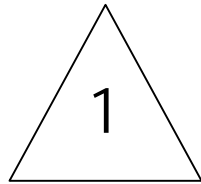
Nearly orthogonality

- Nielson, Vis'97
- specific to spherical wavelets:
 - no uniform triangulation of the sphere
 - no orthogonal triangular spherical wavelets
- subd. depth increase => local uniform planar triang.
- **Nearly orthogonality:**
 - required: orthogonal in the limit case of uniform areas

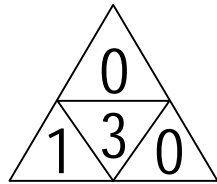
Orthogonality / Semi-orthogonality

Bio-haar basis
[Sch/Sw94],
unif. areas=1

semi-orthogonal

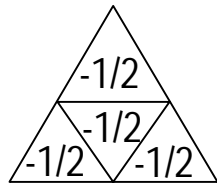


Original
function



$$= 1 \cdot \Phi + \frac{1}{2} \Psi_1 + \frac{3}{4} \Psi_2 + \frac{3}{4} \Psi_3$$

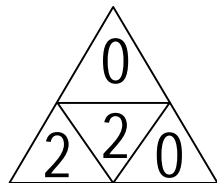
Discarding the
lowest wav. coef.



$$= 1 \cdot \Phi + \frac{3}{4} \Psi_2 + \frac{3}{4} \Psi_3$$

L2 error = $\sqrt{3}$

Best app. with
same basis
functions



$$= 1 \cdot \Phi + \frac{1}{2} \Psi_2 + \frac{1}{2} \Psi_3$$

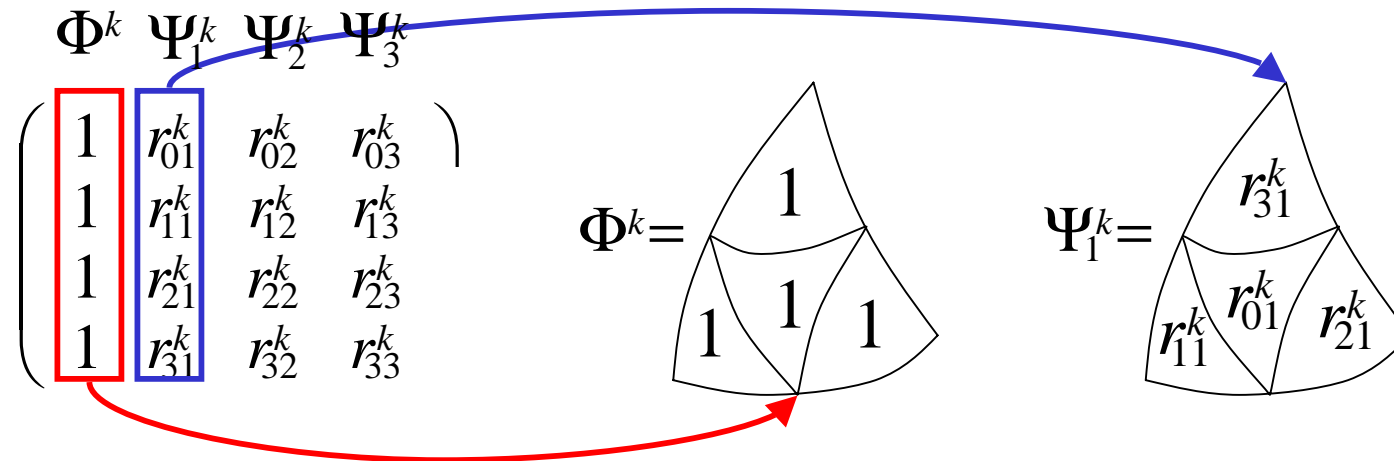
22% error loss

L2 error = $\sqrt{2}$

Orthogonal reconstruction matrices

Columns of 4x4 reconstruction matrix R

= value of functions $\Phi^k, \Psi_1^k, \Psi_2^k, \Psi_3^k$ on the 4 sub-triangles



$$\Phi^k \Psi_1^k = \alpha_0^k \cdot \widehat{1} \times \underbrace{r_{01}^k}_{\text{blue}} + \alpha_1^k \cdot \widehat{1} \times \underbrace{r_{11}^k}_{\text{blue}} + \alpha_2^k \cdot \widehat{1} \times \underbrace{r_{21}^k}_{\text{blue}} + \alpha_3^k \cdot \widehat{1} \times \underbrace{r_{31}^k}_{\text{blue}}$$

Ortho. conditions \longleftrightarrow weighted scalar product of matrix columns

Previous spherical triangular Haar bases

- Bio-Haar [Sch/Sw94]

$$\begin{pmatrix} 1 & -\alpha_1 & -\alpha_2 & -\alpha_3 \\ 1 & \alpha_0 + \alpha_2 + \alpha_3 & -\alpha_2 & -\alpha_3 \\ 1 & -\alpha_1 & \alpha_0 + \alpha_1 + \alpha_3 & -\alpha_3 \\ 1 & -\alpha_1 & -\alpha_2 & \alpha_0 + \alpha_1 + \alpha_2 \end{pmatrix} \quad \alpha_0, \alpha_1, \alpha_2, \alpha_3 : \text{triangle areas}$$

- Semi-orthogonal: Yes

$$C1 \times C2 = \alpha_0 \cdot 1 \times (-\alpha_1) + \alpha_1 \cdot 1 \times (\alpha_0 + \alpha_2 + \alpha_3) + \alpha_2 \cdot 1 \times (-\alpha_1) + \alpha_3 \cdot 1 \times (-\alpha_1) = 0$$

$$C1 \times C3 = 0$$

$$C1 \times C4 = 0$$

- Nearly-orthogonal: No

$$\alpha_0 = \alpha_1 = \alpha_2 = \alpha_3 = \alpha \quad C2 \times C3 = C2 \times C4 = C3 \times C4 = -4\alpha \neq 0$$

Previous spherical triangular Haar basis

- Nielson Vis'97

$$\left(\begin{array}{ccc} 1 & \Delta - \alpha_{0_2}^2 - \alpha_{0_1} \alpha_{0_1} & \Delta - \alpha_{0_2}^2 - \alpha_{0_1} \alpha_{0_2} & \Delta - \alpha_{0_2}^2 - \alpha_{0_1} \alpha_{0_3} \\ 1 & \Delta - \alpha_{1_0}^2 - \alpha_{0_1} \alpha_{0_1} & -\alpha_{0_1} \alpha_{0_2} - \alpha_{0_1} \alpha_{0_2} & -\alpha_{0_1} \alpha_{0_1} - \alpha_{0_1} \alpha_{0_3} \\ 1 & -\alpha_{0_2} \alpha_{0_2} - \alpha_{0_1} \alpha_{0_2} & \Delta - \alpha_{2_2}^2 - \alpha_{0_2} \alpha_{0_2} & -\alpha_{0_2} \alpha_{0_2} - \alpha_{0_2} \alpha_{0_3} \\ 1 & -\alpha_{0_3} \alpha_{0_3} - \alpha_{0_1} \alpha_{0_3} & -\alpha_{0_3} \alpha_{0_3} - \alpha_{0_2} \alpha_{0_3} & \Delta - \alpha_{3_3}^2 - \alpha_{0_3} \alpha_{0_3} \end{array} \right)$$

■ Matrices are both semi- and nearly-orthogonal

where $\Delta = \alpha_{0_0}^2 + \alpha_{1_1}^2 + \alpha_{2_2}^2 + \alpha_{3_3}^2$

$$\left(\begin{array}{ccc} 1 & \Delta - \alpha_{0_2}^2 + 3\alpha_{0_1} \alpha_{0_1} & \Delta - \alpha_{0_2}^2 + 3\alpha_{0_1} \alpha_{0_2} & \Delta - \alpha_{0_2}^2 + 3\alpha_{0_1} \alpha_{0_3} \\ 1 & -3(\Delta - \alpha_{1_0}^2) - \alpha_{0_1} \alpha_{0_1} & -\alpha_{0_1} \alpha_{0_1} + 3\alpha_{0_1} \alpha_{0_2} & -\alpha_{0_1} \alpha_{0_1} + 3\alpha_{0_1} \alpha_{0_3} \\ 1 & -\alpha_{0_2} \alpha_{0_2} + 3\alpha_{0_1} \alpha_{0_2} & -3(\Delta - \alpha_{2_2}^2) - \alpha_{0_2} \alpha_{0_2} & -\alpha_{0_2} \alpha_{0_2} + 3\alpha_{0_2} \alpha_{0_3} \\ 1 & -\alpha_{0_3} \alpha_{0_3} + 3\alpha_{0_1} \alpha_{0_3} & -\alpha_{0_3} \alpha_{0_3} + 3\alpha_{0_2} \alpha_{0_3} & -3(\Delta - \alpha_{3_3}^2) - \alpha_{0_3} \alpha_{0_3} \end{array} \right)$$

New families of triangular Haar bases

- Previous reconstruction matrices:
 - degree 1 (Sch/Sw94) and 2 (Nielson97) in the triangle areas
- Systematic investigation of possible bases
 - Use any polynomial function of the areas in the reconstruction matrix
- Low degree \implies look for degree 1 polynomial functions in $\alpha_0, \alpha_1, \alpha_2, \alpha_3$

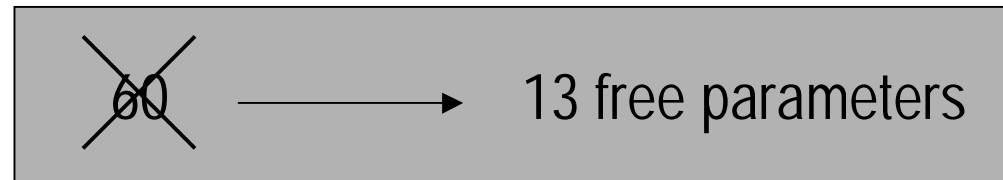
$$\begin{pmatrix} 1 \\ 1 \\ 1 \\ 1 \end{pmatrix} + \sum_{l=0}^3 r_{ij}^l \alpha_l \quad 4 \times 3 \times 5 = 60 \text{ free parameters}$$

Required condition: Symmetry

- Symmetry \iff affine invariance
 \iff invariance when permuting the indices (1,2,3)

$$\begin{array}{c}
 1 \longleftrightarrow 2 \\
 \begin{array}{c} 1 \\ 2 \end{array} \updownarrow
 \end{array}
 \left(
 \begin{array}{ccc}
 1 & a + a \alpha_0 + a \alpha_1 + a \alpha_2 + a \alpha_3 & a + a \alpha_0 + a \alpha_1 + a \alpha_2 + a \alpha_3 & a + a \alpha_0 + a \alpha_1 + a \alpha_2 + a \alpha_3 \\
 1 & b + b \alpha_0 + b \alpha_1 + b \alpha_2 + b \alpha_3 & c + c \alpha_0 + c \alpha_1 + c \alpha_2 + c \alpha_3 & c + c \alpha_0 + c \alpha_1 + c \alpha_2 + c \alpha_3 \\
 1 & c + c \alpha_0 + c \alpha_1 + c \alpha_2 + c \alpha_3 & b + b \alpha_0 + b \alpha_1 + b \alpha_2 + b \alpha_3 & c + c \alpha_0 + c \alpha_1 + c \alpha_2 + c \alpha_3 \\
 1 & c + c \alpha_0 + c \alpha_1 + c \alpha_2 + c \alpha_3 & c + c \alpha_0 + c \alpha_1 + c \alpha_2 + c \alpha_3 & b + b \alpha_0 + b \alpha_1 + b \alpha_2 + b \alpha_3
 \end{array}
 \right)$$

- free parameters: $a_{0-3}, b_{0-3}, c_{0-4}$

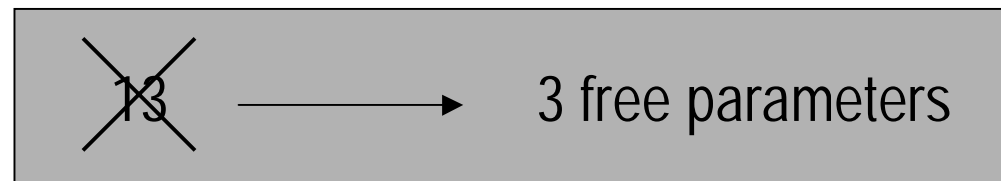


Required condition: semi-orthogonality

- $C1 \times C2 = C1 \times C3 = C1 \times C4 = 0$

$$\begin{pmatrix} 1 & a\alpha_1 + b\alpha_2 + b\alpha_3 & b\alpha_1 + a\alpha_2 + b\alpha_3 & b\alpha_1 + b\alpha_2 + a\alpha_3 \\ 1 & -a\alpha_0 + c\alpha_2 + c\alpha_3 & -b\alpha_0 - c\alpha_2 & -b\alpha_0 - c\alpha_3 \\ 1 & -b\alpha_0 - c\alpha_1 & -a\alpha_0 + c\alpha_1 + c\alpha_3 & -b\alpha_0 - c\alpha_3 \\ 1 & -b\alpha_0 - c\alpha_1 & -b\alpha_0 - c\alpha_2 & -a\alpha_0 + c\alpha_1 + c\alpha_2 \end{pmatrix}$$

- free parameters: a, b, c



Required condition: nearly-orthogonality

- $C_2 \times C_3 = C_2 \times C_4 = C_3 \times C_4 = 0$ if all triangle areas are equal
- Same matrix as in previous slide, with

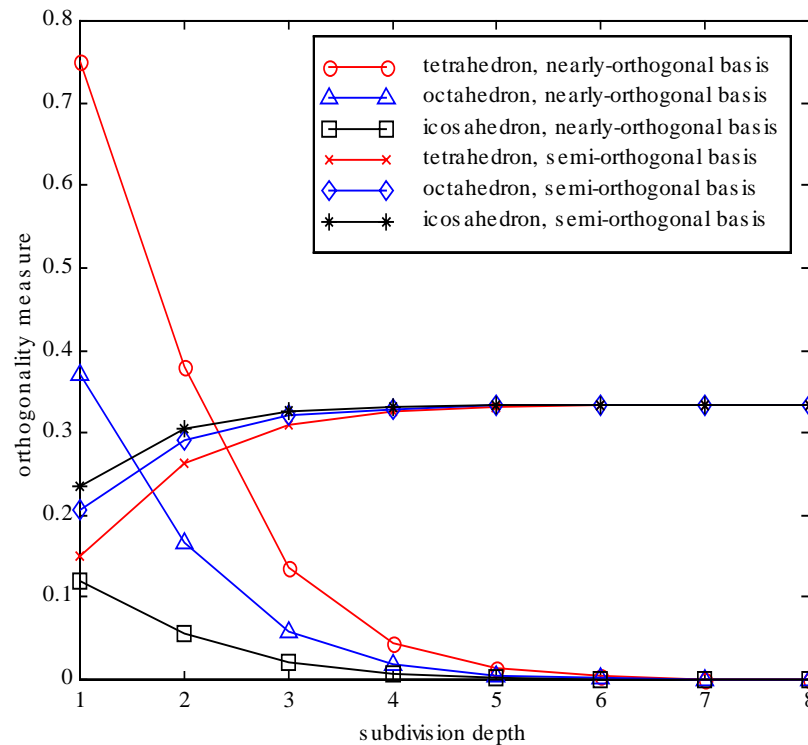
$$\left\{ \begin{array}{l} \text{or} \\ c = a+b \\ c = -\left(\frac{a+5b}{3}\right) \end{array} \right. \quad \begin{array}{l} \longleftarrow \\ \longleftarrow \end{array} \quad \begin{array}{l} \text{Basis family (I)} \\ \text{Basis family (II)} \end{array}$$

- free parameters: a, b
- normalization \Rightarrow free parameter b , ($a = 1$)

$$\cancel{3} \longrightarrow 1 \text{ free parameter}$$

Orthogonality measure

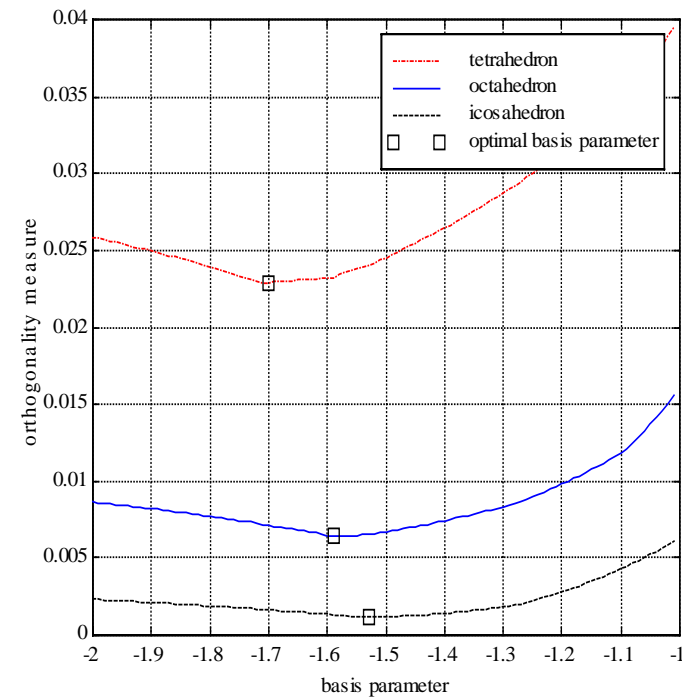
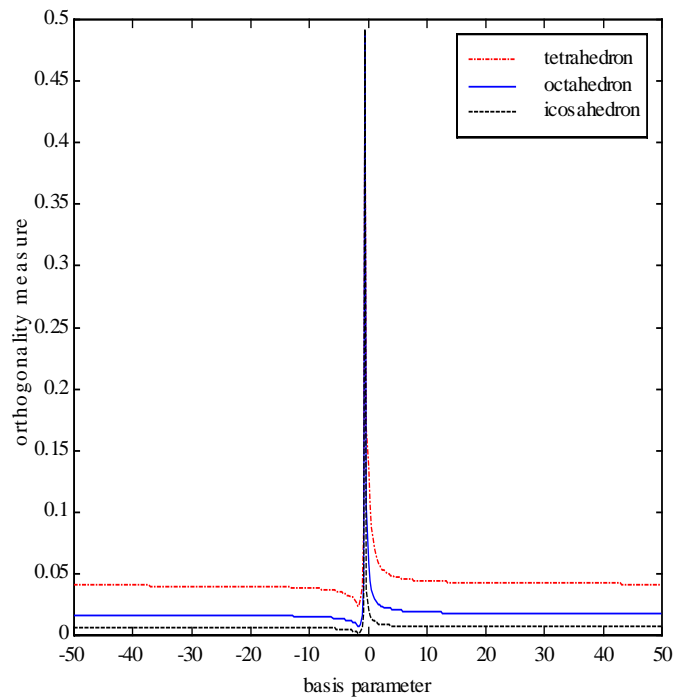
$$M(K) = \frac{1}{n} \sum_{k=1}^K \sum_{T^k} \left| \Psi_1^k \Psi_2^k \right| + \left| \Psi_1^k \Psi_3^k \right| + \left| \Psi_2^k \Psi_3^k \right|$$



•Need to choose among all possible nearly orthogonal bases
=> optimize a measure

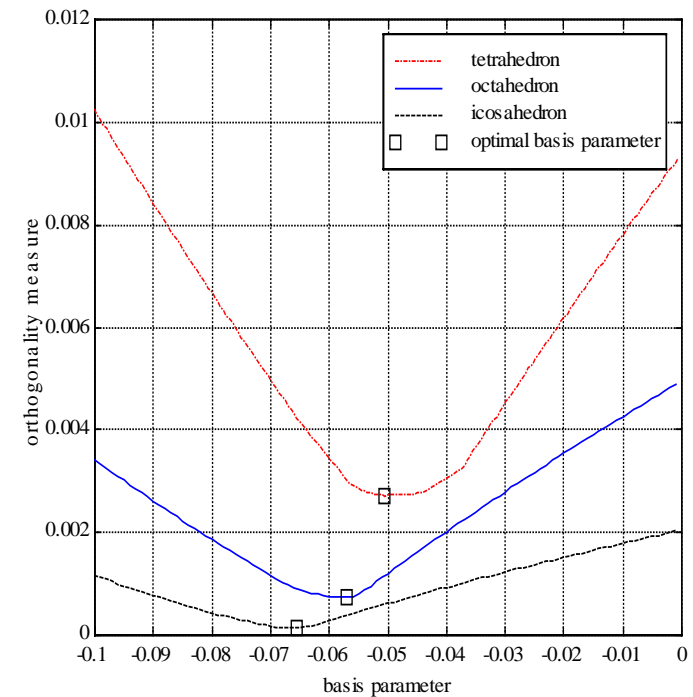
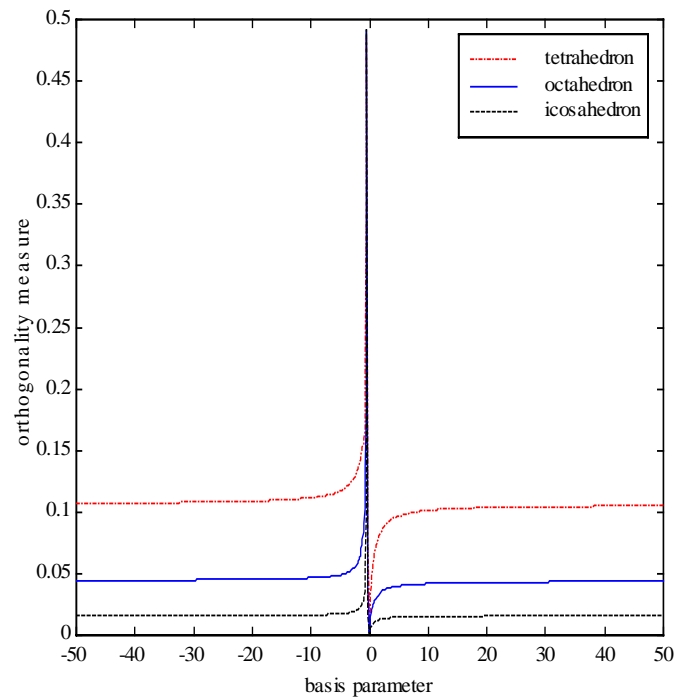
Basis optimization

Orthogonality measure at depth 3,
for different parameter values, **Basis (I)**



Basis optimization

Orthogonality measure at depth 3,
for different parameter values, **Basis (II)**



Basis optimization

Base mesh	Basis (I)		Basis (II)	
	optimal parameter values	optimal orthogonality measure	optimal parameter values	optimal orthogonality measure
tetrahedron	a=1 b=-1.707	2.28 10⁻²	a=1 b=-0.051	2.72 10⁻³
octahedron	a=1 b=-1.592	6.44 10⁻³	a=1 b=-0.061	8.99 10⁻⁵
icosahedron	a=1 b=-1.535	1.14 10⁻³	a=1 b=-0.067	1.62 10⁻⁵

Results

- Orthogonal basis \Rightarrow L2 error = squared rooted sum of squared discarded wavelet coefficients
- Measure relative difference between squared rooted sum of squared discarded wavelet coefficients and L2 error

Should be as minimal as possible (vanishes exactly for orthogonal bases)

- Test data set: global topography (NOAA), mapped on a depth 7 subdivided spherical tetrahedron, octahedron & icosahedron

Results

Base mesh: spherical tetrahedron

#wav. coeff.	comp. gain		Bio-Haar Sch/Sw94	Basis (I)	Basis (II)
500	99.2%		1.11%	0.32%	0.036%
2000	96.9%		1.76%	0.30%	0.034%
6000	90.8%		1.47%	0.30%	0.033%
10000	84.7%		1.04%	0.30%	0.033%

Results

Base mesh: spherical **octahedron**

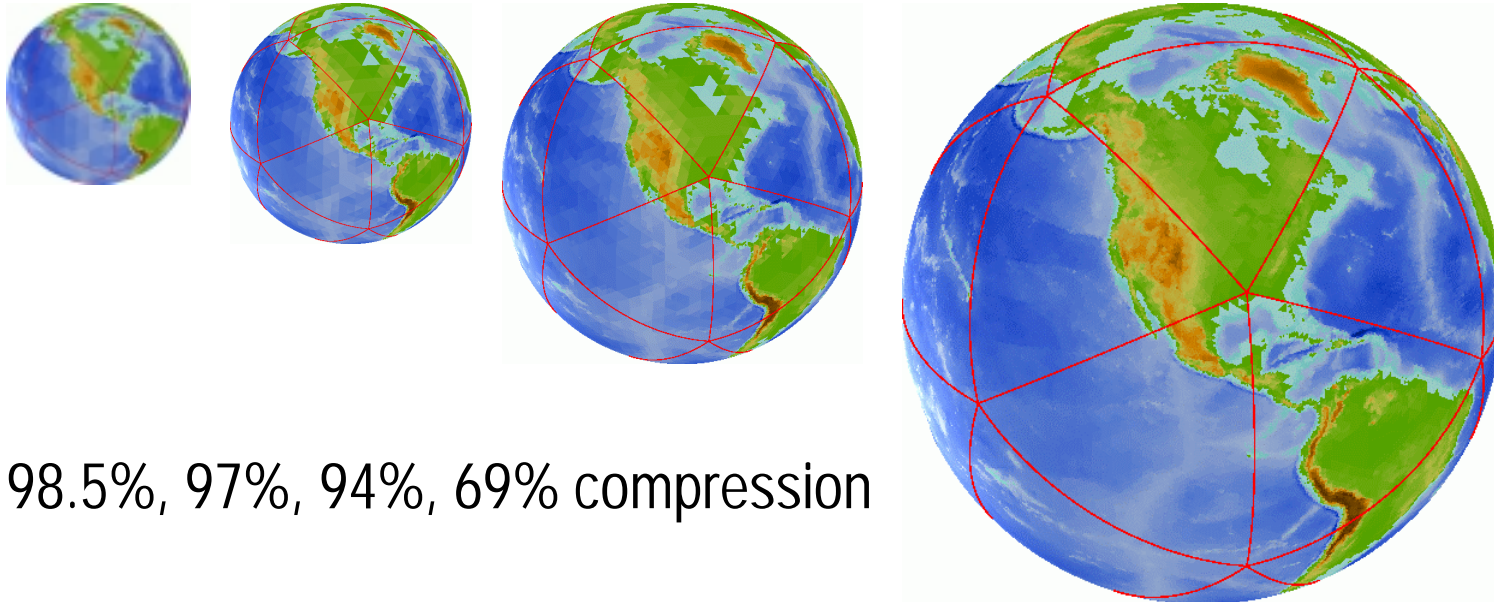
#wav. coeff.	comp. gain		Bio-Haar Sch/Sw94	Basis (I)	Basis (II)
1000	99.2%		1.27%	0.05%	0.0056%
5000	96.2%		1.36%	0.05%	0.0052%
10000	92.4%		0.78%	0.05%	0.0052%
20000	84.7%		0.58%	0.05%	0.0051%

Results

Base mesh: spherical **icosahedron**

#wav. coeff.	comp. gain		Bio-Haar Sch/Sw94	Basis (I)	Basis (II)
5000	98.5%		2.44%	0.0039%	0.00058%
10000	96.9%		2.30%	0.0040%	0.00054%
20000	94.0%		1.90%	0.0040%	0.00052%
40000	87.8%		1.59%	0.0040%	0.00051%

Results



98.5%, 97%, 94%, 69% compression

