

Deriving 3D Shape Properties by Using Backward Wavelet Remesher

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1

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Motivation: Brain Image Atlasing



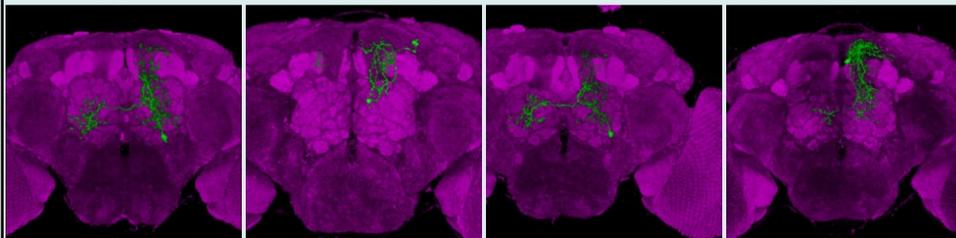
Similar neurons (less neurons, more neurons)

5HT1A-M-700001
(r: 59.6%, pValue: 2.2e-16)

fru-M-400310
(r: 58.45%, pValue: 2.2e-16)

5HT1A-M-100029
(r: 54.5%, pValue: 2.2e-16)

fru-M-800082
(r: 52.49%, pValue: 2.2e-16)



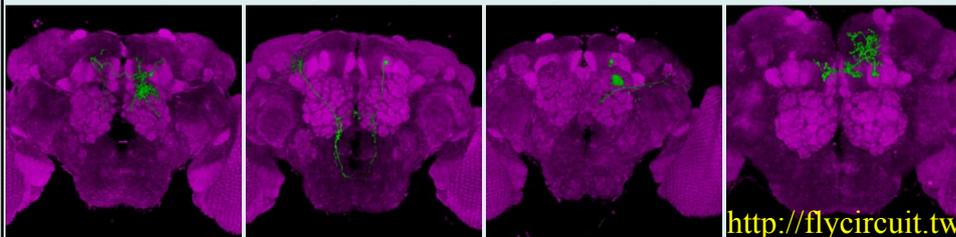
Cell body neighborhood (less neurons, more neurons)

5-HT1B-M-000000
(Distance: 0 um)

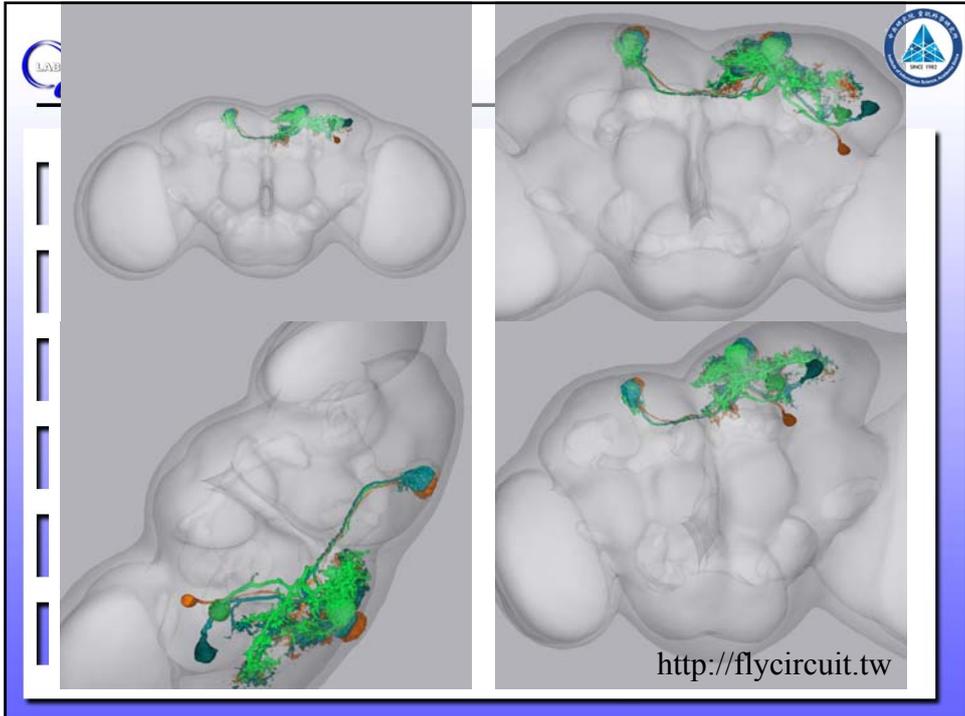
fru-M-500110
(Distance: 6.13 um)

npf-M-300052
(Distance: 6.56 um)

TH-M-500004
(Distance: 7.58 um)



<http://flycircuit.tw>



Motivation:
How to derive a 3D average?

LAB *asp*

NSC 100

minimize $\sum_{i=1}^K \|S - S_i\|^2$

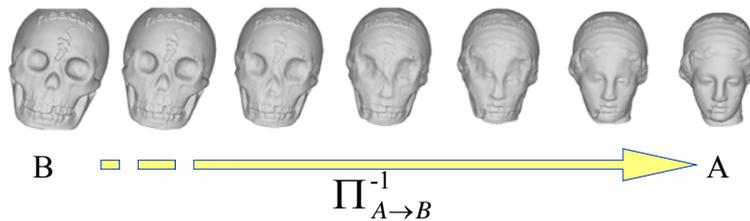
S_i + S_i $\xrightarrow{?}$ $S^* = \frac{1}{K} \sum_{i=1}^K S_i$



Motivation:

1. How to derive a 3D standard average?
2. How to construct a 2D+Z deformation field?
3. How to define 3D boundary condition?

→ Analyze 3D shape properties first.



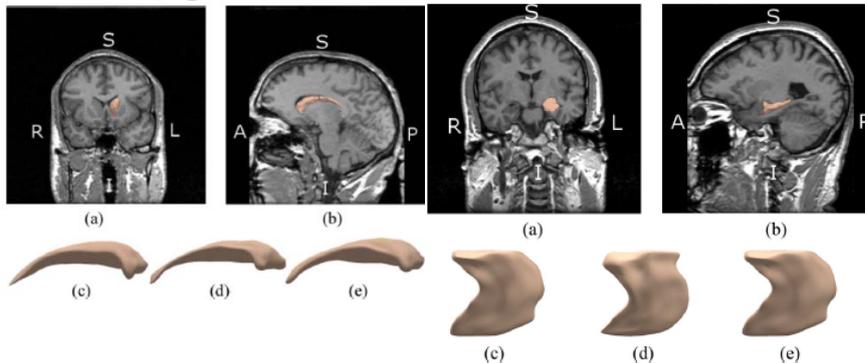
5



3D Shape Analysis:

An Example (Nain et al.)

■ Schizophrenia VS Hippocampus



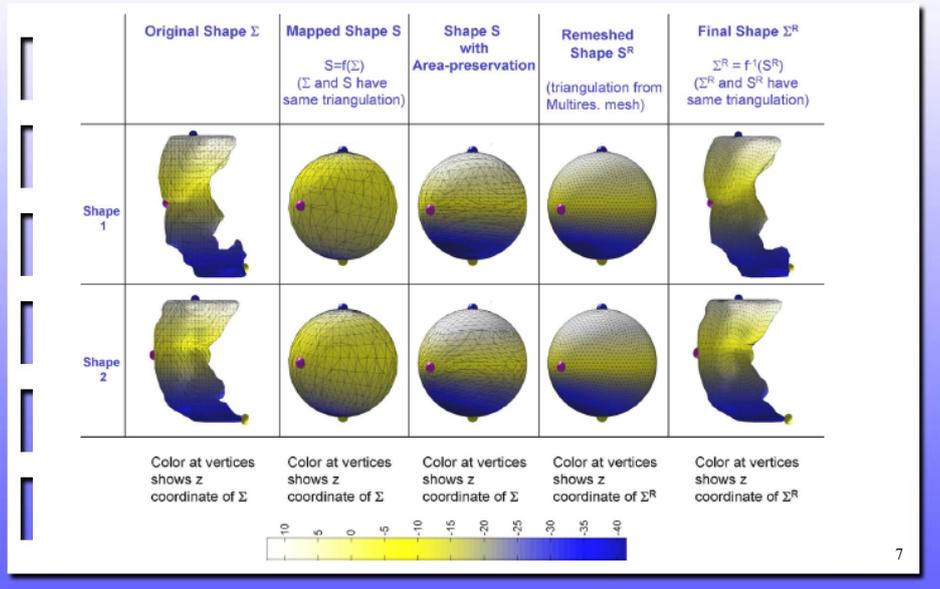
Nain et al., "Multiscale 3-d shape representation and segmentation using spherical wavelets",
IEEE Trans. Med. Imaging, pp.598-618, vol.26(4), 2007.

6

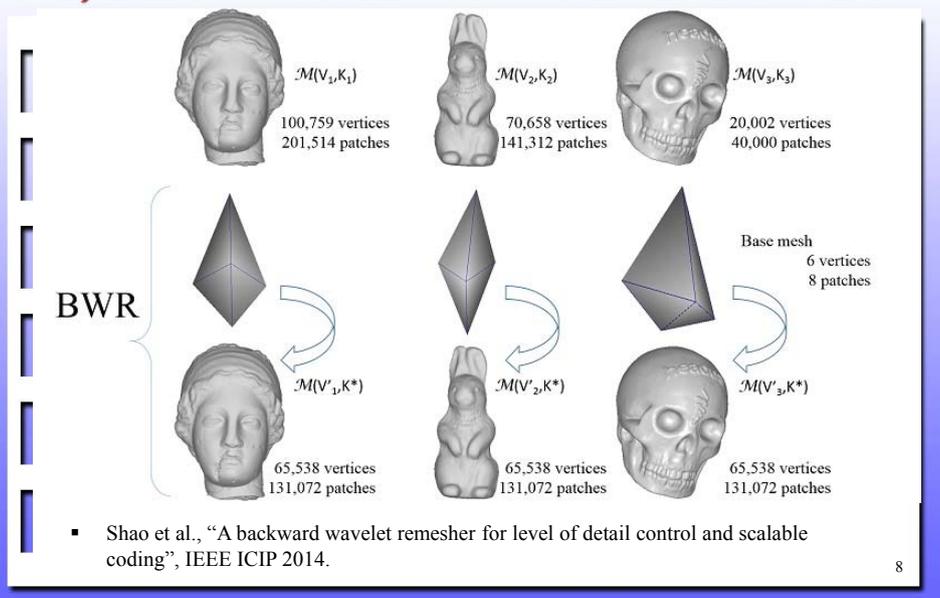


3D Shape Analysis:

An Example (Nain et al.)



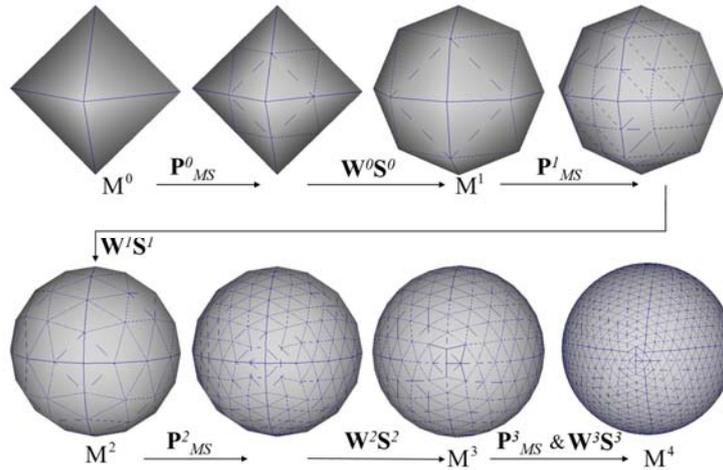
Our Design Goal & Our Prototype





Concept:

Coarse-to-fine (backward) wavelet representation

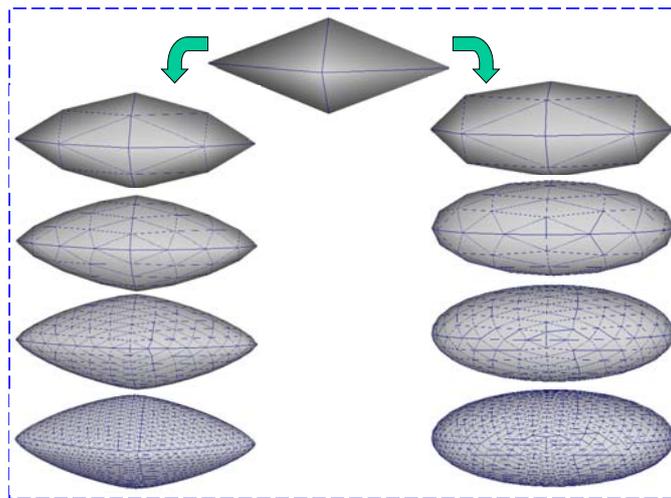


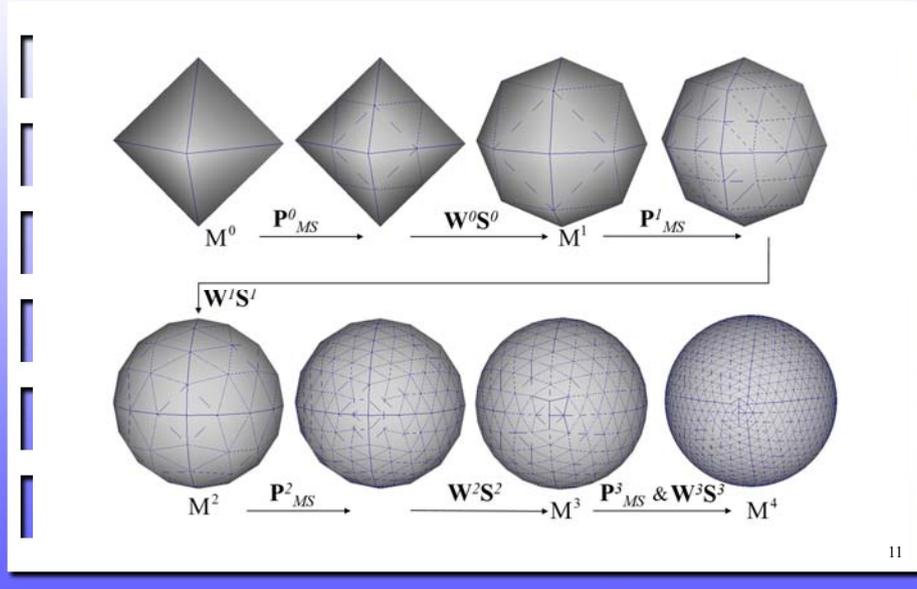
9



Concept:

Coarse-to-fine (backward) wavelet representation





1D DWT

$$\begin{bmatrix} L_x \\ H_x \end{bmatrix} = \begin{bmatrix} A \\ B \end{bmatrix} [x]$$



$$\begin{bmatrix} A \\ B \end{bmatrix}^{-1} \begin{bmatrix} L_x \\ H_x \end{bmatrix} = [x]$$

$$= [P \quad Q] \begin{bmatrix} L_x \\ H_x \end{bmatrix}$$

$$\frac{1}{\sqrt{2}} \begin{bmatrix} 1.5 \\ 3.5 \\ 5.5 \\ 7.5 \\ 9.5 \\ -0.5 \\ -0.5 \\ -0.5 \\ -0.5 \\ -0.5 \end{bmatrix}_{10 \times 1} = \begin{bmatrix} \gamma_{FE} & \gamma_{FE} & & & & & & & & & \\ & \gamma_{FE} & \gamma_{FE} & & & & & & & & \\ & & \gamma_{FE} & \gamma_{FE} & & & & & & & \\ & & & \gamma_{FE} & \gamma_{FE} & & & & & & \\ & & & & \gamma_{FE} & \gamma_{FE} & & & & & \\ \gamma_{FE} & \gamma_{FE} & & & & & & & & & \gamma_{FE} & \gamma_{FE} \\ & & \gamma_{FE} & \gamma_{FE} & & & & & & & & \\ & & & \gamma_{FE} & \gamma_{FE} & & & & & & & \\ & & & & \gamma_{FE} & \gamma_{FE} & & & & & & \\ & & & & & \gamma_{FE} & \gamma_{FE} & & & & & \\ & & & & & & \gamma_{FE} & \gamma_{FE} & & & & \\ & & & & & & & \gamma_{FE} & \gamma_{FE} & & & \\ & & & & & & & & \gamma_{FE} & \gamma_{FE} & & \\ & & & & & & & & & \gamma_{FE} & \gamma_{FE} & \\ & & & & & & & & & & \gamma_{FE} & \gamma_{FE} \end{bmatrix}_{10 \times 10} \begin{bmatrix} 1 \\ 2 \\ 3 \\ 4 \\ 5 \\ 6 \\ 7 \\ 8 \\ 9 \\ 10 \end{bmatrix}_{10 \times 1}$$



$$\frac{1}{\sqrt{2}} \begin{bmatrix} \sqrt{2} & & & & & & & & & & \\ & \sqrt{2} & & & & & & & & & \\ & & \sqrt{2} & & & & & & & & \\ & & & \sqrt{2} & & & & & & & \\ & & & & \sqrt{2} & & & & & & \\ & & & & & \sqrt{2} & & & & & \\ & & & & & & \sqrt{2} & & & & \\ & & & & & & & \sqrt{2} & & & \\ & & & & & & & & \sqrt{2} & & \\ & & & & & & & & & \sqrt{2} & \\ & & & & & & & & & & \sqrt{2} \\ & & & & & & & & & & & \sqrt{2} \\ & & & & & & & & & & & & \sqrt{2} \\ & & & & & & & & & & & & & \sqrt{2} \\ & & & & & & & & & & & & & & \sqrt{2} \end{bmatrix}_{10 \times 10} \begin{bmatrix} 1.5 \\ 3.5 \\ 5.5 \\ 7.5 \\ 9.5 \\ -0.5 \\ -0.5 \\ -0.5 \\ -0.5 \\ -0.5 \end{bmatrix}_{10 \times 1} = \begin{bmatrix} 1 \\ 2 \\ 3 \\ 4 \\ 5 \\ 6 \\ 7 \\ 8 \\ 9 \\ 10 \end{bmatrix}_{10 \times 1}$$

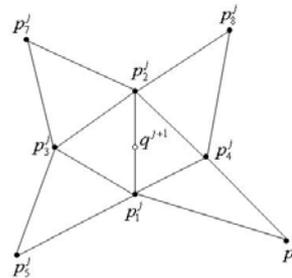
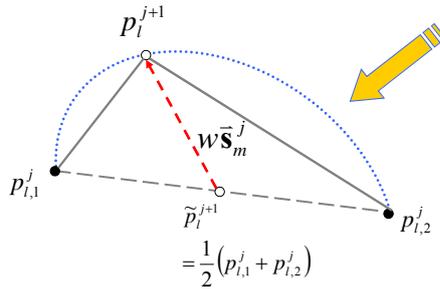
- What will happen if butterfly subdivision scheme is adopted?

$$\mathbf{q}_m^{j+1} = \tilde{\mathbf{q}}_m^{j+1} + \mathbf{w}_m^j$$



$$p_i^{j+1} = \frac{1}{2}(p_{i,1}^j + p_{i,2}^j) + w\bar{s}_i$$

$$\bar{s}_i = 2(p_{i,3}^j + p_{i,4}^j) - (p_{i,5}^j + p_{i,6}^j + p_{i,7}^j + p_{i,8}^j)$$



N. Dyn, D. Levine, and J. A. Gregory, "A butterfly subdivision scheme for surface interpolation with tension control," ACM transactions on Graphics (TOG), vol. 9, no. 2, pp. 160–169, 1990.

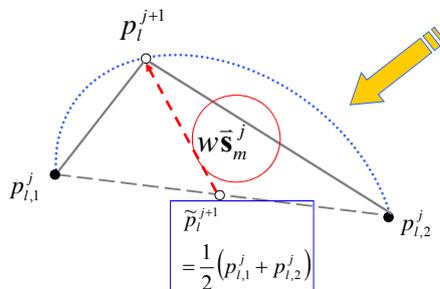
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$$\mathbf{c}^{j+1} = \begin{bmatrix} \mathbf{P}^j & \mathbf{Q}^j \end{bmatrix} \begin{bmatrix} \mathbf{c}^j \\ \mathbf{d}^j \end{bmatrix}$$

$$= \mathbf{P}^j \mathbf{c}^j + \mathbf{Q}^j \mathbf{d}^j$$

N. Dyn, D. Levine, and J. A. Gregory, "A butterfly subdivision scheme for surface interpolation with tension control," ACM transactions on Graphics (TOG), vol. 9, no. 2, pp. 160–169, 1990.



Coarse-to-fine (backward) wavelet representation

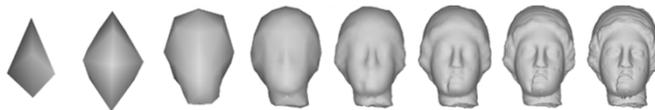
$$[x] = [P \quad Q] \begin{bmatrix} L_x \\ H_x \end{bmatrix} = \begin{bmatrix} P_{upper} & Q_{upper} \\ P_{lower} & Q_{lower} \end{bmatrix} \begin{bmatrix} L_x \\ H_x \end{bmatrix}$$

$$\begin{bmatrix} c_n^{j+1} \\ c_m^{j+1} \end{bmatrix} = \begin{bmatrix} P_n^j & \mathbf{0}_{n \times m} \\ P_m^j & I_m \end{bmatrix} \begin{bmatrix} c_n^j \\ d_m^j \end{bmatrix} \quad (\text{lazy wavelet})$$

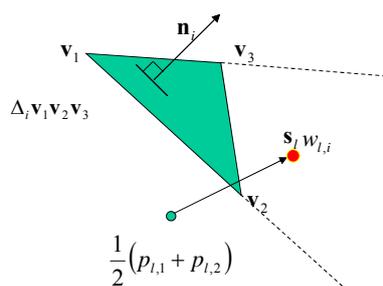
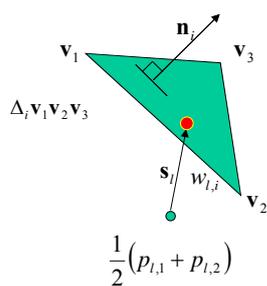
$$\begin{bmatrix} c_n^{j+1} \\ c_m^{j+1} \end{bmatrix} = \begin{bmatrix} P_n^j \\ P_m^j \end{bmatrix} c_n^j + \begin{bmatrix} \mathbf{0}_{n \times 3} \\ d_m^j \end{bmatrix}$$

$$\begin{bmatrix} c_n^{j+1} \\ c_m^{j+1} \end{bmatrix} = \begin{bmatrix} I_n \\ P_m^j \end{bmatrix} c_n^j + \begin{bmatrix} \mathbf{0}_{n \times 3} \\ d_m^j \end{bmatrix} \quad (\text{interpolating subdivision})$$

$$\begin{bmatrix} c_n^{j+1} \\ c_m^{j+1} \end{bmatrix} = P_{mid}^j c_n^j + \begin{bmatrix} \mathbf{0}_{n \times 3} \\ W^j S^j + d^j \end{bmatrix}$$



How to find "w"?



$$w_{l,i} = \frac{\langle \mathbf{n}_i, \mathbf{v}_1 \rangle - \langle \mathbf{n}_i, \frac{1}{2}(p_{l,1} + p_{l,2}) \rangle}{\langle \mathbf{n}_i, \mathbf{s}_l \rangle} \quad \begin{bmatrix} \lambda_1 \\ \lambda_2 \end{bmatrix} = (\mathbf{T}^t \mathbf{T})^{-1} \mathbf{T}^t (\mathbf{p}_l - \mathbf{v}_3), \text{ where}$$

$$\mathbf{T} = \begin{bmatrix} x_1 - x_3 & x_2 - x_3 \\ y_1 - y_3 & y_2 - y_3 \\ z_1 - z_3 & z_2 - z_3 \end{bmatrix}$$



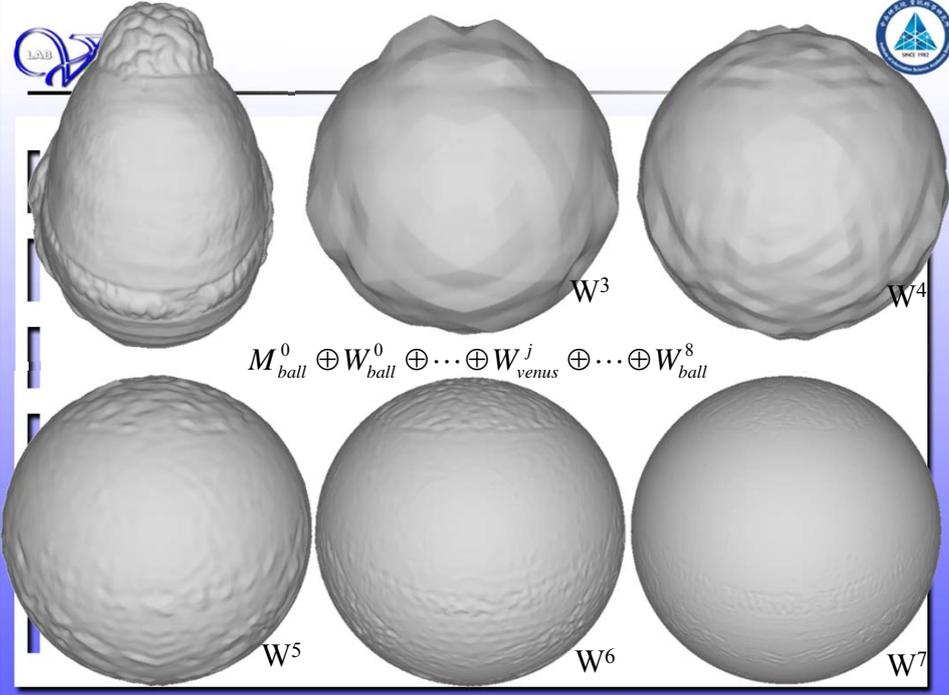


Experiment Results



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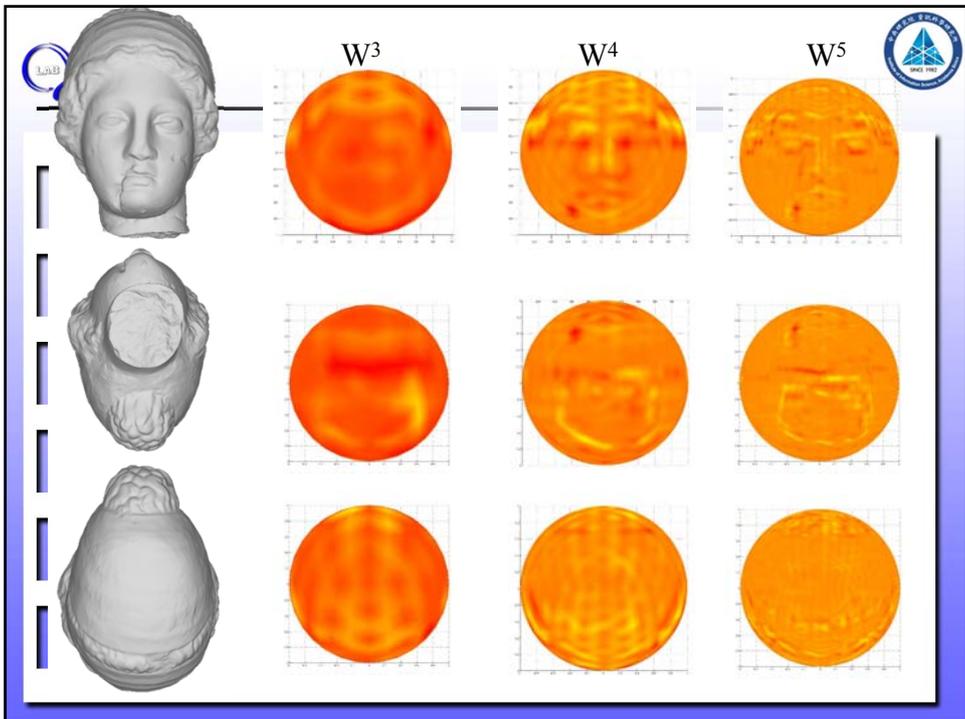
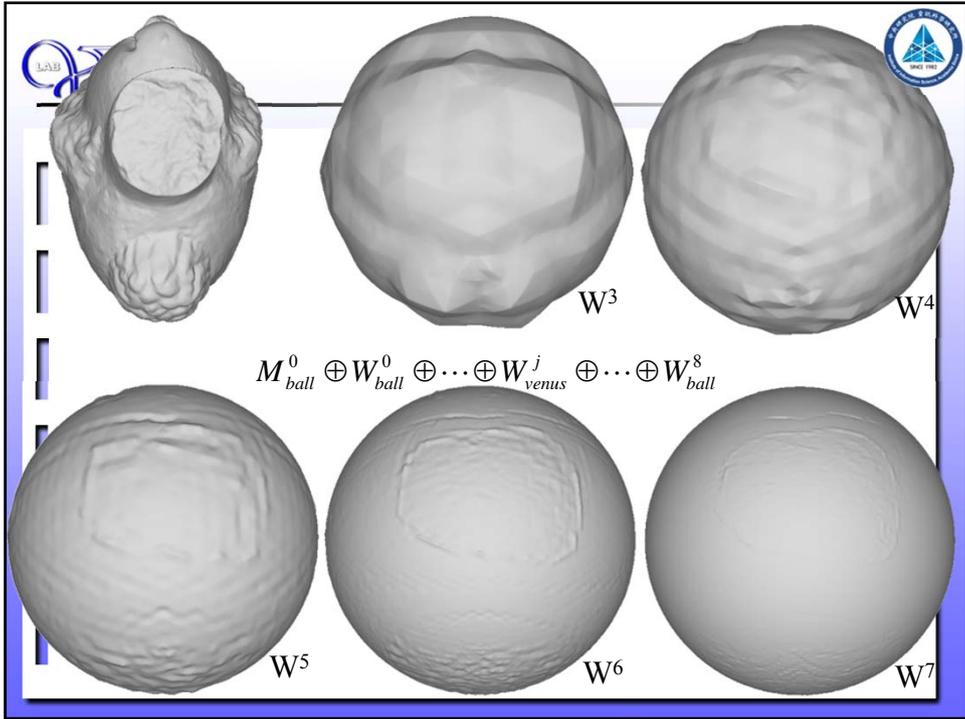


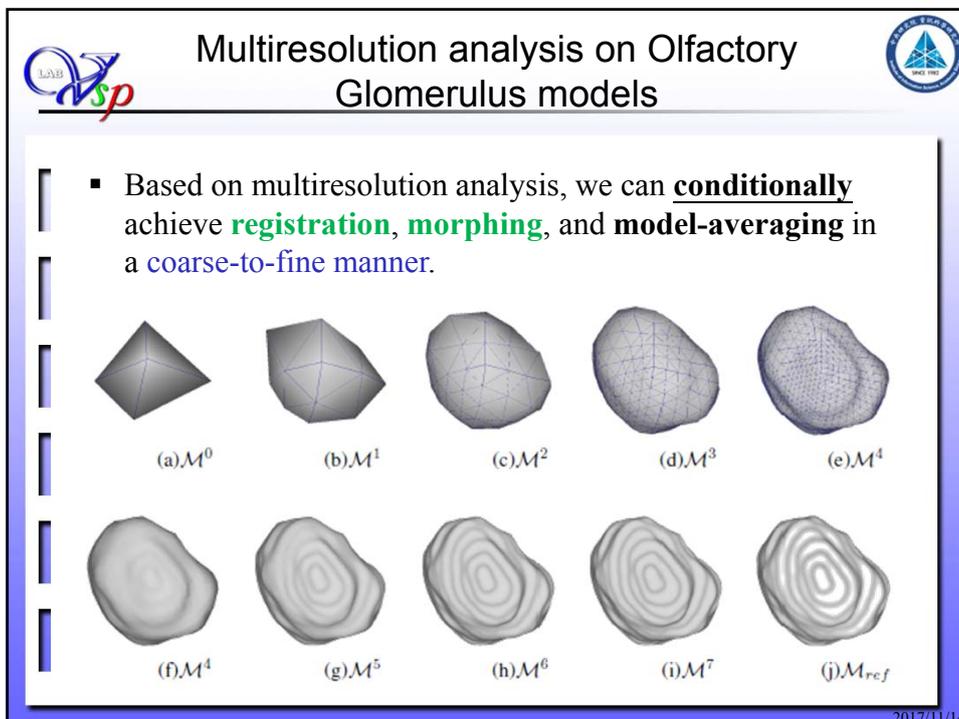
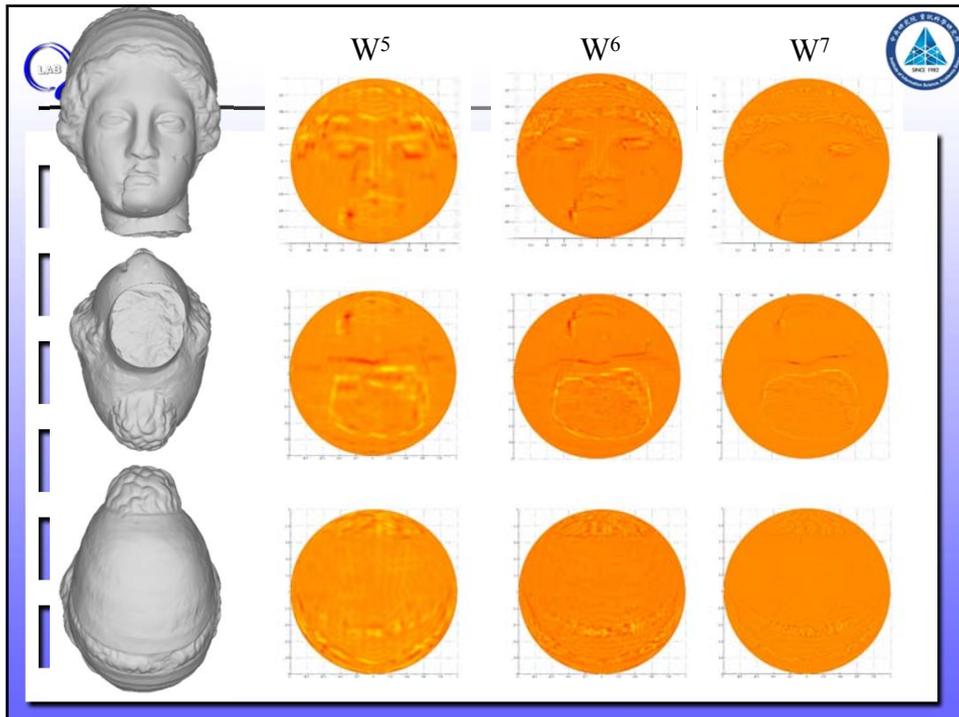


$M_{ball}^0 \oplus W_{ball}^0 \oplus \dots \oplus W_{venus}^j \oplus \dots \oplus W_{ball}^8$

W^3 W^4

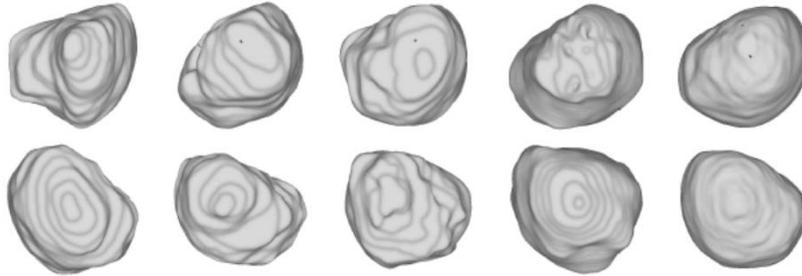
W^5 W^6 W^7







Multiresolution analysis on Olfactory Glomerulus models



(a)M-01
TABLE 1

	M-01	M-02	M-03	M-04	M-Avg.
M-01	—	0.0907	0.0864	0.1102	0.0773
M-02	0.1145	—	0.0796	0.1063	0.0783
M-03	0.0801	0.1316	—	0.1033	0.0791
M-04	0.1082	0.1192	0.1100	—	0.0740
M-Avg.	0.0767	0.1019	0.0689	0.0829	—

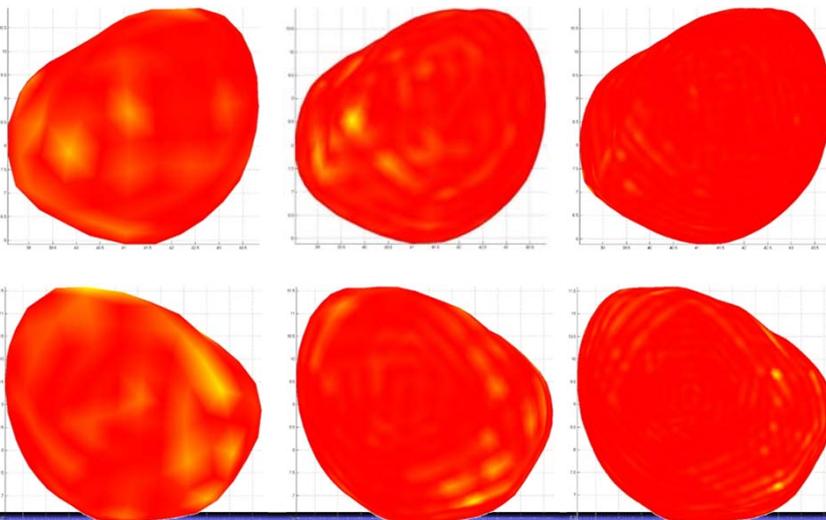
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Multiresolution analysis on Olfactory Glomerulus models



- Regions with high structural variability are highlighted in yellow.



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Concluding Remarks

- A **backward coarse-to-fine** method.
- **Tiling-invariant** → BWR acts as a transformation that can convert input meshes into a standard reference domain.
- **Simple**: remeshing only.
- Suitable for mesh **registration/editing/warping**, and also applicable to **biomedical applications**, such as tracking deformations of a 3D beating heart model.

2017/11/14



Thank you.



28

2017/11/14



Multiresolution Analysis on 2-manifold



- M. Eck, T. DeRose, T. Duchamp, H. Hoppe, M. Lounsbery, and W. Stuetzle, “**Multiresolution analysis of arbitrary meshes**”, 1995.
- M. Lounsbery, T. DeRose, and J. Warren, “**Multiresolution analysis for surfaces of arbitrary topological type**”, ACM Transactions on Graphics, 16 (1997), pp. 34-73.
- I. Guskov, W. Sweldens, and P. Schroder, “Multiresolution signal processing for meshes”, 1999.
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- **Lee et al., “Maps: Multiresolution adaptive parameterization”, 1998.**
- Stollnitz, DeRose, and Salesin, “Wavelets for computer graphics: theory and applications”, Morgan Kaufmann, 1996.
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- Schroder and Sweldens, “Spherical wavelets: Efficiently representing functions on the sphere”, 1995.
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