Molecular Structure and Properties **Elucidation from 3D Electron Microscopy**

Chandrajit Bajaj





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Multi-scale Multi-Modal Imaging

- X-ray crystallography (diffraction)
 - Atomic resolution
 - Difficulties (experimental, computational)
- Nuclear magnetic resonance (NMR)
 - Atomic resolution
 - Limited to small structures

- **Electron Microscopy**
 - Cryo-electron tomography
 - Low resolution (20Å 200Å)
 - · Good for whole cell or cell organelles
 - Single particle cryo-EM
 - Intermediate resolution (5Å 20Å)
 - Computationally more demanding





Electron tomography (Picture from A.J. Koster et al, JSB, 1997)





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Image Acquisition



schematic of possible orientations

2D Cryo-EM map

- Images collected via electron microscopy
- Lowest possible radiation used to limit damage to sample

Annu. Rev. Biophys. Biomol. Struct. 2002. 31:303-19



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Quasi Atomic Models from Single Particle Cryo-EM



Single Particle Cryo-EM Pipeline





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Sub-nanometer Structure Elucidation from 3D-EM



Single Particle Cryo-EM: Automatic Structure Analysis



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Step #1 : Anisotropic Filtering

Bilateral filtering

$$h(x,\xi) = e^{-\frac{(x-\xi)^2}{2\sigma_d^2}} \cdot e^{-\frac{(f(x)-f(\xi))^2}{2\sigma_r^2}}$$

where σ_d and σ_r are parameters and *f*(.) is the image intensity value.



Anisotropic diffusion filtering

 $\partial_t \phi - \operatorname{div}(a(|\nabla \phi|) | \nabla \phi) = 0$ where **a** stands for the diffusion tensor determined by local curvature estimation.

C. Bajaj, G. Xu, ACM Transactions on Graphics, (2003),22(1), pp. 4- 32.



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W. Jiang, M. Baker, Q. Wu, C. Bajaj, W. Chiu, Journal of Structural Biology, 144, 5,(2003), Pages 114-122

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Step #2: Critical Point Detection

- For smooth data:
 - zeroes of the gradient vector field
 - simple, easy to implement
- For noisy data:
 - Gradient vector diffusion
 - higher time complexity but robust to noise



- Gradient vector diffusion:
 - smoothing the vector fields
 - diffusion to flat regions

$$\begin{cases} \frac{\partial u}{\partial t} = \mu \cdot div(g(\alpha)\nabla u) \\ \frac{\partial v}{\partial t} = \mu \cdot div(g(\alpha)\nabla v) \end{cases}$$

where $g(\alpha)$ is a decreasing function α is the angle between the central pixel and its surrounding pixels.









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Step #3: Symmetry Detection

• Asymmetric subunits in an icosahedra



- Two-fold vertices
- Three-fold vertices
- Five-fold vertices



Local symmetry (RDV) (260 trimers or 720 proteins)



• Correlation search, addtly sped up by Spherical FFT:

Find best c, minimizing:

 $\sum_{\vec{r} \in \mathcal{D}} \left(f(\vec{r}) - f(R_{2\pi/n}(c) \cdot \vec{r}) \right)^2$

Y.Zu, C. BajajIEEE Transactions on Image Processin, 2005, 14, 9, 1324-1337



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Results of Automatic Symmetry Detection in Virus and Phage Capsid Shells



Step #4: Subunit Segmentation

- Multi-seed Fast Marching Method
 - Classify the critical points based on local symmetry into separate groups.
 - Each seed initializes one contour, with its group's membership.
 - Contours march simultaneously. Contours with same membership are merged, while contours with different membership stop each other.



Zeyun, Bajaj IEEE Trans on Imag. Proc.,2005

Bacteriophage P22





Subunit about global **Icosahedral 5 fold axis**

Subunit about local 6 fold axis



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silor

GroEL

- Chaperonin responsible for protein folding
- Composed of 14 identical subunits
- Subunits have D7 symmetry





- Segmentation from 6 angstrom and and 11.5 angstrom Cryo-EM structures yielded 14 monomers as expected
- Segmentation of 25 angstrom Cryo-EM structure yielded seven segments, each composed of 2 monomers



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- Gray= automated segmentation
- Green= crystal structure
- Correlation value generated against x-ray structure
- Accuracy of segmentation relies on resolution of Cryo-EM map
 - -Segments don't always correspond to subunits from low resolution maps



Additional Results

Subunit	Model	Correlation
RDV P8 trimer	Manual segmentation, P8 trimer	0.74
RDV P8 trimer	X-ray structure, P8 trimer	0.85
RDV P8 monomer	Manual segmentation, P8 monomer	0.80
RDV P8 monomer	X-ray structure, P8 monomer	0.84
P22 tail machine	X-ray structure, P22 trimeric tail spike	0.76
70S ribosome, 50S subunit	X-ray structure, 50S subunit	0.63
70S ribosome, 30S subunit	X-ray structure, 30S subunit	0.73
30S subunit, RNA	X-ray structure, 30S subunit RNA	0.66



M. Baker, Z. Yu, W. chiu, C. Bajaj, J of Structural Biology, 2006



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Step #5A: Secondary Structure Identification

Gradient tensor

$$(\mathbf{I}_{\mathbf{x}}, \mathbf{I}_{\mathbf{y}}, \mathbf{I}_{\mathbf{z}}) = \begin{pmatrix} I_{x}^{2} & I_{x}I_{y} & I_{x}I_{z} \\ I_{x}I_{y} & I_{y}^{2} & I_{y}I_{z} \\ I_{x}I_{z} & I_{y}I_{z} & I_{z}^{2} \end{pmatrix}$$

• Local structure tensor (Weickert'98, Fernandez'03)



$$\begin{pmatrix} I_x^2 * G_\sigma & I_x I_y * G_\sigma & I_x I_z * G_\sigma \\ I_x I_y * G_\sigma & I_y^2 * G_\sigma & I_y I_z * G_\sigma \\ I_x I_z * G_\sigma & I_y I_z * G_\sigma & I_z^2 * G_\sigma \end{pmatrix}$$

Property of local structure tensor eta-sheet •λ₃ ha-helix λ_3 λ_{2} λ_2 plane structure Line structure $\lambda_2 \approx \lambda_3 >> \lambda_1 \approx 0$ $\lambda_1 >> \lambda_2 \approx \lambda_3 \approx 0$ Center for Computational Visualization http://www.ices.utexas.edu/CCV Institute of Computational and Engineering Sciences May 2007 University of Texas at Austin **Department of Computer Sciences**

Step #5A: Quasi-Atomic Models



Zeyun, Bajaj, IEEE/ACM Tran on CompBio&BioInf., 2007



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Critical Points, their Indices, and their Manifolds

Critical Point of a smooth function is a point where the gradient of the function vanishes.

Index of a critical point is the number of independent directions in which the function decreases.

In 3D, four types of critical points

- 1. Minima index 0
- 2. Saddle of index 1
- 3. Saddle of index 2
- 4. Maxima index 3



Integral curve : A path in the domain of the function on which at every point the tangent to the curve equals the gradient of the function.

Stable Manifold of a critical point is the union of all integral curves ending at the critical point.

Unstable Manifold of a critical point is the union of all integral curves starting at the critical point.



Medial Axis and Distance Functions

<u>Medial Axis</u> M of a shape S is defined as a set of points which has more than one nearest point on S.



Distance Function h_S assigns every point x the nearest distance to S.

Approximation of h_S is done via h_P when S is known only via a finite set of points P on S.

$$h_P : \mathbb{R}^3 \to \mathbb{R}, \ x \mapsto \min_{p \in P} \|x - p\|$$



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#5B-I: Secondary Structure Elucidation



Step 1: Vor/Del(P) computation

Step 2: Identification of Interior Medial Axis M.

Step 3:

3.a: Identification of Critical points of distance function from Vor/Del(P).

3.b: Selection of Critical points only on M.

Step 4: Classification of Medial Axis via
4.a: U₁ – <u>Unstable Manifold of index 1 saddle</u> point on *M* 4.b: U₂ – <u>Unstable Manifold of index 2 saddle</u> point on *M*.

Step 5: Width Test to select the subsets of U_1 (β -sheets) and U_2 (α -helices).



#5B-II: Tertiary Structure Elucidation



Step 1: Vor/Del(P) computation

Step 2: Identification of Interior Medial Axis M

Step 3:

- 3.a: Identification of Critical points of h_P from Vor/Del(P).
- 3.b: Selection of Critical points only on M.

Step 4: Decomposition of shape via S3 stable manifold of maxima on M

Step 5: Width Test to select the subsets of S3.



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Secondary Structure of RDV Outer Capsid Coat Protein P8 Closeup U_1 and U_2 Surface U₁ α -helices 15/16 β -sheets 3/3 Another View Ribbon Diag. of PDB Helices and Sheets Helices 3D EM map of P8 segmented out from cryo-EM map of

Rice Dwarf Virus (RDV) at 6.8 A resolution



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Secondary Structure of Bacterial Chaperonin GroEL





- U_1 and U_2 give superset of sheets and helices.
- α -helix: width 2.5 A and pitch-length 1.5 A [Branden-Tooze]
- β-sheet: thickness 1.5 A [Branden-Tooze]
- h_P values of Voronoi elements constituting U_1 and U_2 help select the subset that passes the width and thickness test.





The tertiary fold of 1AOR is a β -sandwich (two red sheets), which is surrounded by the differently colored helical segments.



The tertiary fold of 1TIM is .a α/β -barrel.

The β -region in the middle is segmented as red while the helical segments surrounding it are colored differently.



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Free Energy of a Macromolecule in Solvent (Towards Flexible Models of Proteins)



How do we Compute G_{pol}

$$G_{\text{pol}} = \frac{1}{2} \int [\phi_{\text{solvent}}(\mathbf{r}) - \phi_{\text{air}}(\mathbf{r})] \rho(\mathbf{r}) dV$$

(a) Poisson-Boltzmann (PB) Theory

$$-\nabla \cdot [\varepsilon(\mathbf{r})\nabla\phi(\mathbf{r})] = 4\pi\rho(\mathbf{r}) + 4\pi\lambda(\mathbf{r})\sum_{j=1}^{\infty} c_j^{\infty} q_j \exp(-q_j\phi(\mathbf{r})/k_B T)$$

Molecular volume V for charge density, dielectric interface

- ε dielectric coefficients
- ϕ electrostatic potential
- ρ solute charge density
- λ ion accessibility parameter
 - \int_{i}^{∞} ion bulk concentration
- q_i ion charge
- k_{B} Boltzmann's constant
- T temperature
- (b) Generalized Born (GB) Theory
 - Born formula (Born 1920), Generalized Born formula (Still 1990)

$$G_{\text{pol}} = -\frac{\tau}{2} \sum_{ij} \frac{q_i q_j}{[r_{ij}^2 + R_i R_j \exp(-\frac{r_{ij}^2}{4R_i R_j})]^{\frac{1}{2}}} \qquad R_i^{-1} = \frac{1}{4\pi} \int_{\text{ex}} \frac{1}{|\mathbf{r} - \mathbf{x}_i|^4} \, dV$$

$$\tau = 1 - \frac{1}{\varepsilon_{\text{solv}}} \begin{array}{c} r_{ij} \\ r_{ij}$$

Step 6b: GB based G_{pol} - Calculation of Born Radii

A single charge q_i located at the center of atom i in the molecule.

$$G_{\mathrm{pol}} = -rac{ au}{2} rac{q_i^2}{R_i}$$
 (by GB)

On the other hand, by approximating the electric field as the Columbic field,

$$G_{\text{pol}} = -\frac{1}{8\pi} \tau \int_{\text{ex}} \frac{q_i^2}{|\mathbf{r} - \mathbf{x}_i|^4} dV$$



ex: exterior of the molecule

Therefore $R_i^{-1} = \frac{1}{4\pi} \int_{ex} \frac{1}{|\mathbf{r} - \mathbf{x}_i|^4} dV$ via Gauss' Divergence Thm $1 - \int_{ex} (\mathbf{r} - \mathbf{x}_i) \cdot \mathbf{p}(\mathbf{r})$

$$R_i^{-1} = \frac{1}{4\pi} \int_{\Gamma} \frac{(\mathbf{r} - \mathbf{x}_i) \cdot \mathbf{n}(\mathbf{r})}{|\mathbf{r} - \mathbf{x}_i|^4} \ dS$$

 Γ : molecular surface



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Step 6b: GB based $G_{\rm pol}~$ - Fast Calculation of Born Radii

$$R_i^{-1} = \frac{1}{4\pi} \int_{\Gamma} \frac{(\mathbf{r} - \mathbf{x}_i) \cdot \mathbf{n}(\mathbf{r})}{|\mathbf{r} - \mathbf{x}_i|^4} dS \approx \frac{1}{4\pi} \sum_{k=1}^N w_k \frac{(\mathbf{r}_k - \mathbf{x}_i) \cdot \mathbf{n}(\mathbf{r}_k)}{|\mathbf{r}_k - \mathbf{x}_i|^4}, \quad \mathbf{r}_k \in \Gamma$$

Algorithm:

- 1. Generate a model for the molecular surface $\ \Gamma$.
- 2. Cubature: choose w_k and \mathbf{r}_k properly so that higher order accuracy can be obtained for small N.
- 3. Fast Fourier summation to evaluate R_i , i = 1, ..., M.

Bajaj, Zhao 2007



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Step 7b: Flexible Match and Fit using generalized modal analysis



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CCV Software

Open Source and in Public Domain

(http://www.ices.utexas.edu/CCV/software/)



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Senior Collaborators

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- Nathan Baker (WashU)
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