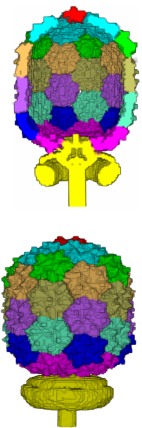
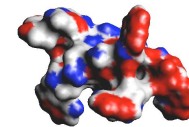
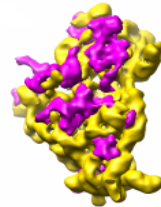
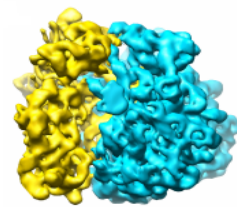


Molecular Electron Microscopy to Biophysical Analysis

Chandrajit Bajaj

<http://www.cs.utexas.edu/~bajaj>



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January 2008

Acknowledgements

- **Group Members**

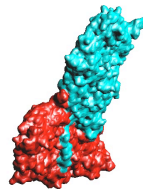
- Albert Chen (CS, Ph.D)
- Rezaul Chowdhury (PostDoc)
- Ajay Gopinath (BME,Ph.D)
- Andrew Gillete (Math, Ph.D.)
- R. Inkulu (PostDoc)
- Alex Kim (CS, Ugrad)
- Bongjune Kwon (BME,Ph.D)
- Oliver Martin (CS, Ugrad)
- Maysam Mousallem (CS,PhD)
- Joe Rivera (Res. Scientist)
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- Jesse Sweet (Math, MS)
- Dongjin Suh (CS, MS)
- Samrat Goswami (PostDoc)
- Zeyun Yu (UCSD)**
- Xiaoyu Zhang (CSU)**
- Wenqi Zhao (CAM, Ph.D.)

- **Senior Collaborators**

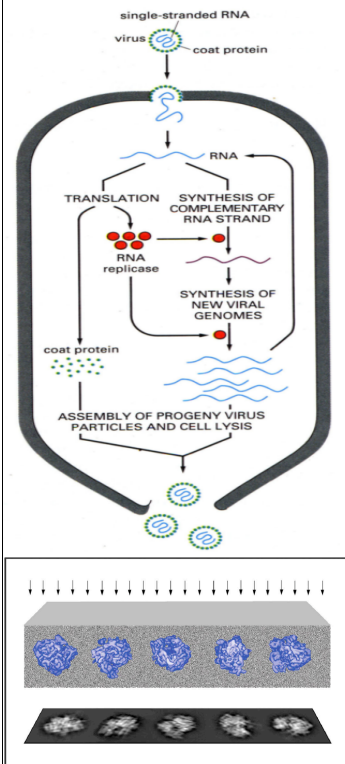
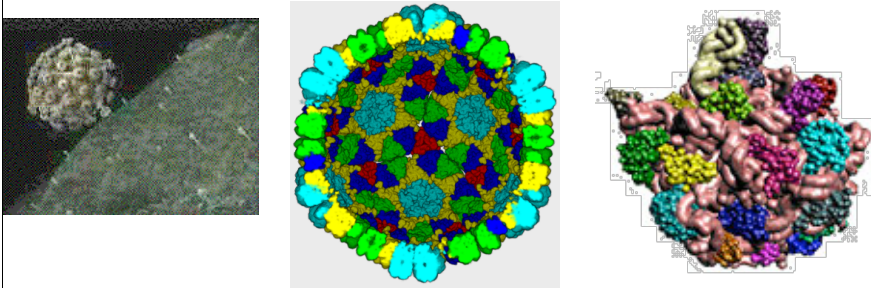
- Manfred Auer (LBL,UCB)
- Tim Baker, M. Ellisman (UCSD)
- Nathan Baker (WashU)
- Wah Chiu, Steve Ludtke (Baylor)
- Inderjit Dhillon (UT, CS)
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- Joachim Frank (Suny Albany, Columbia)
- Kristen Harris , Dan Johnston (UT, ILM)
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- Justin Kiney, Tom Bartol, Terry Sejnowski (Salk)
- Andy McCammon, Michael Holst (UCSD)
- Art Olson, Michel Sanner, David Goodsell (TSRI)
- Alberto Paoluzzi, Antonio DiCarlo(Roma)
- Sriram Subramaniam (NIH-NCI)

- **Sponsored by**

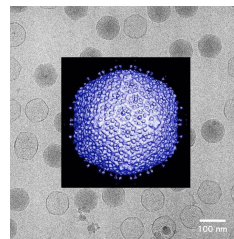
- **NIH: P20-RR020647, R01-GM074258, R01-0783087, R01-EB004873**
- **NSF: ITR-IIS-032550, DDDAS-CNS-0540033**
- **UT: HTDS**



Seek, Analyze and Destroy: Imaging to BioPhysical Modeling to Drug Discovery

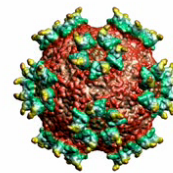


Single Particle Cryo-EM

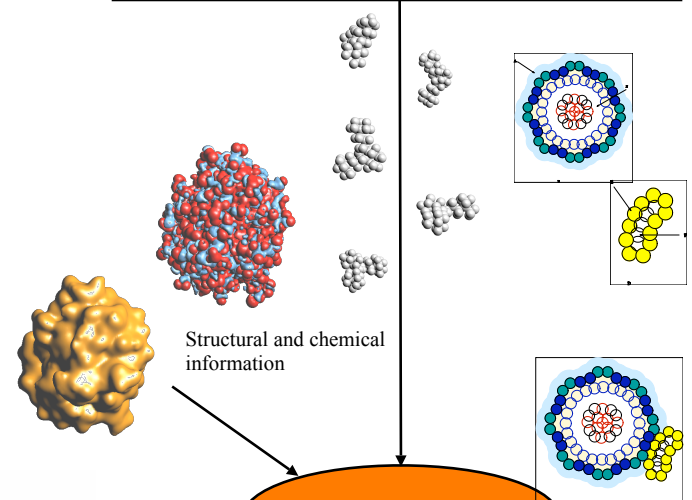


QUESTIONS ARE ASKED TO SEE THE PICTURE.

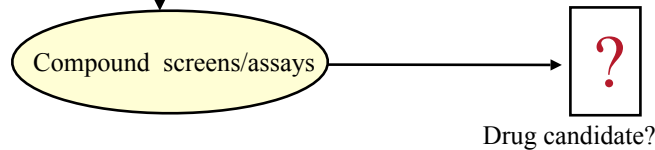
HRV with IG



Database of drug like compounds



Potential complex conformations



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Collaborators: BCM, TSRI ****Sponsored by NIH, TI-3D**

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Molecular Structure Elucidation from Electron Microscopy

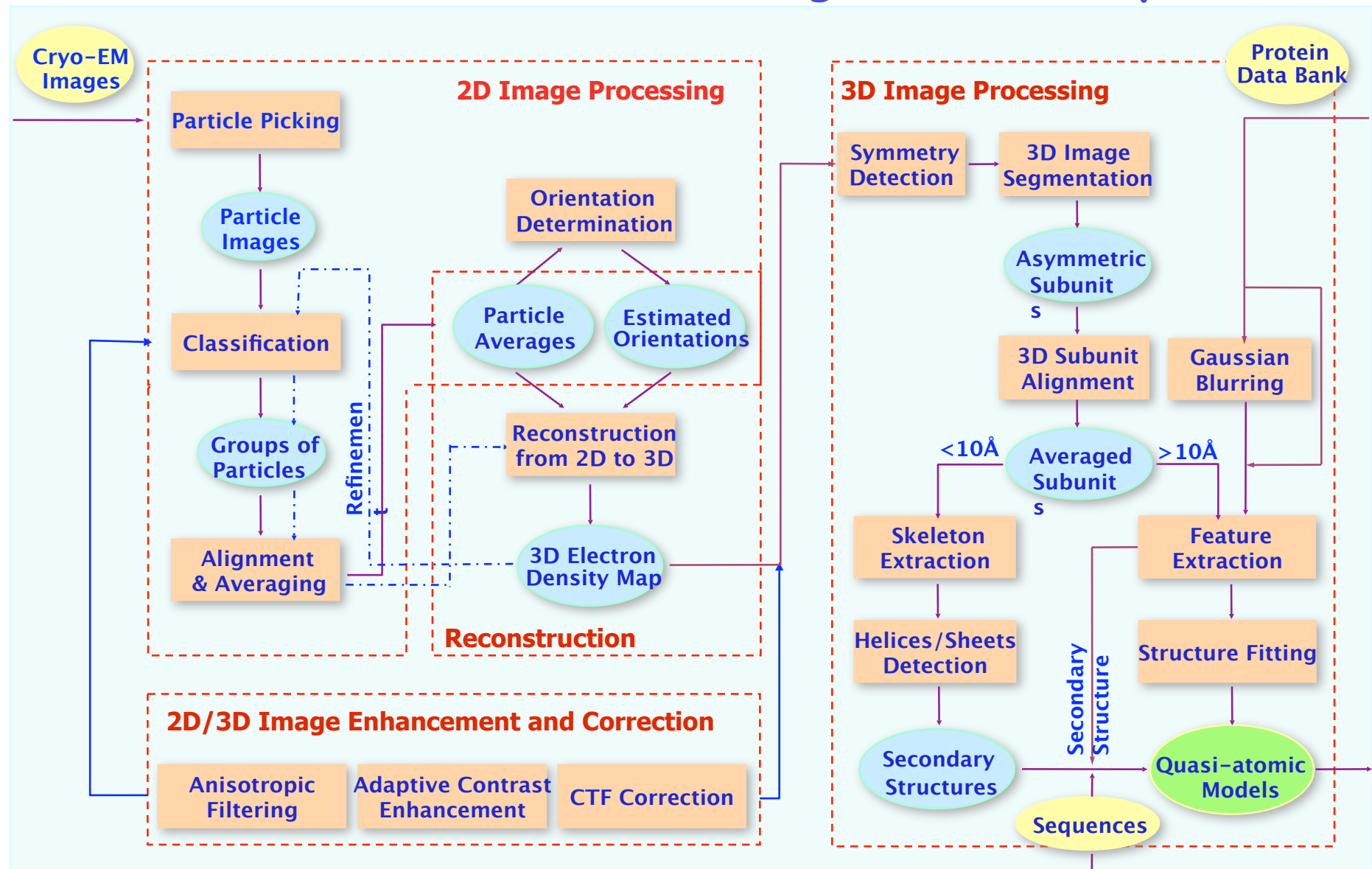


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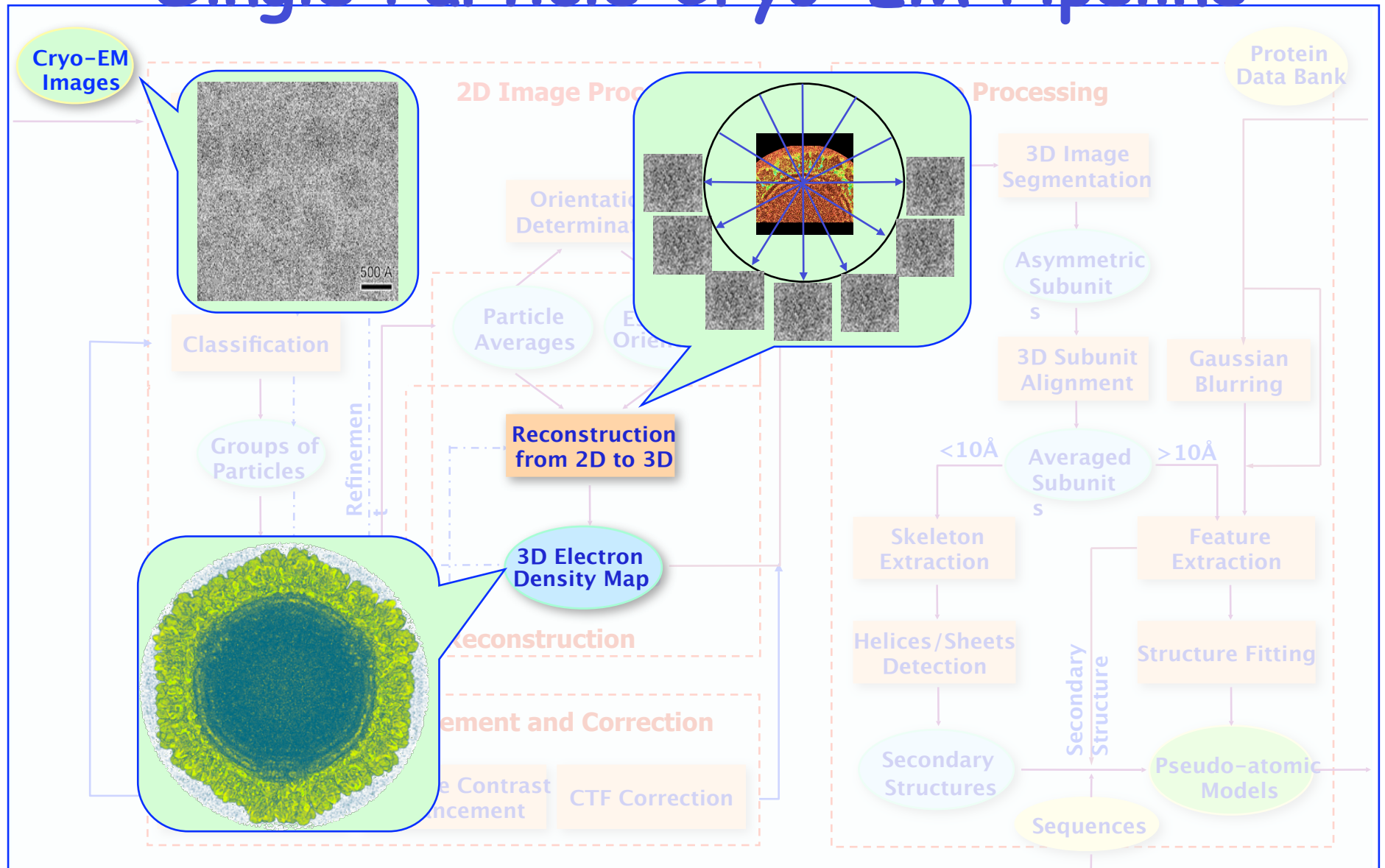
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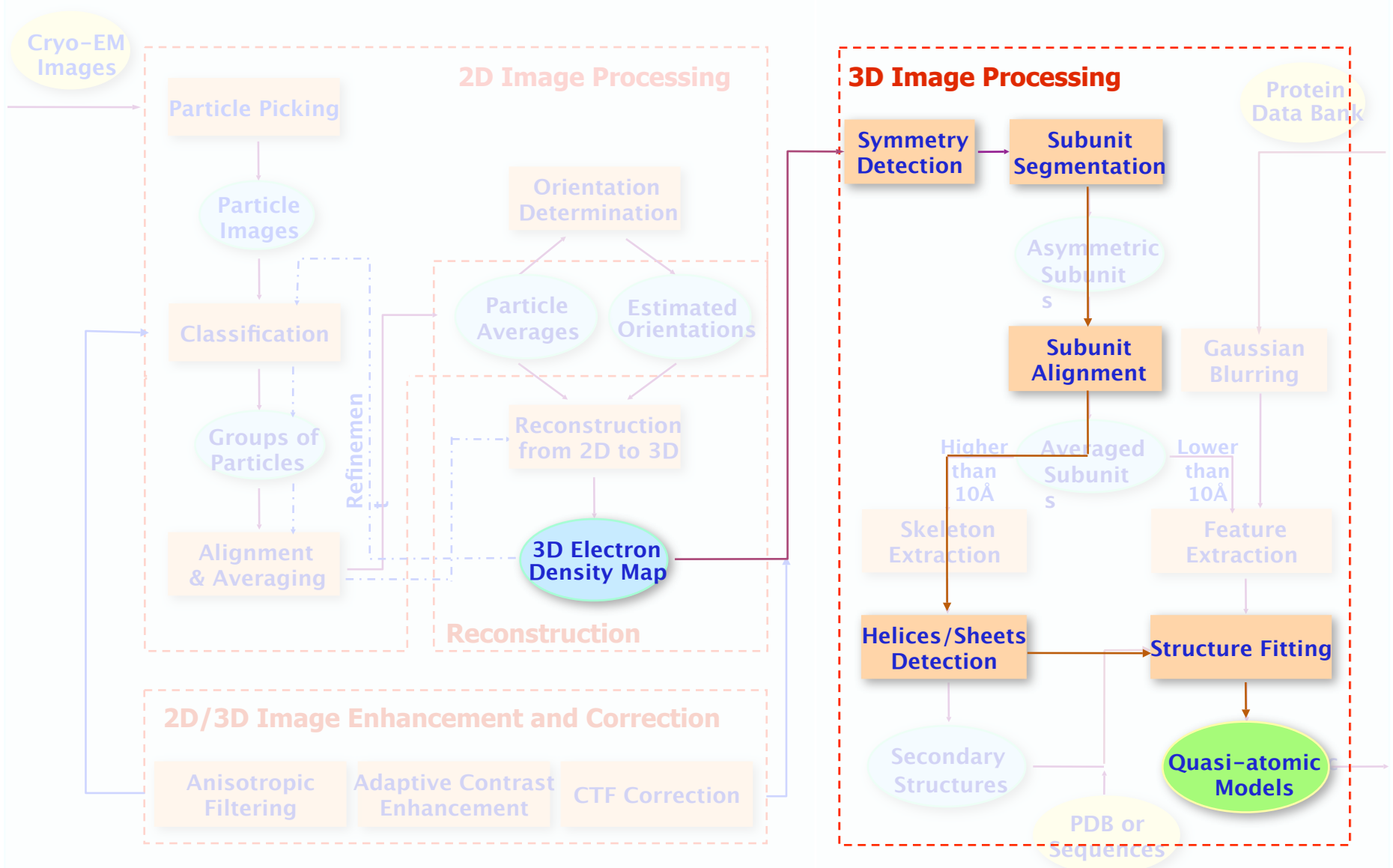
Quasi Atomic Models from Single Particle Cryo-EM



Single Particle Cryo-EM Pipeline



Single Particle Cryo-EM: Automatic Structure Analysis

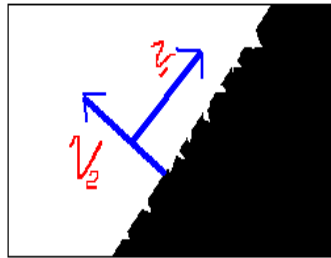
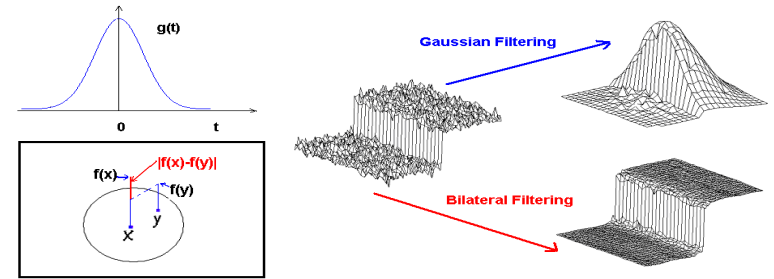


Step #1 : Anisotropic Filtering

Bilateral filtering

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

where σ_d and σ_r are parameters and $f(\cdot)$ is the image intensity value.

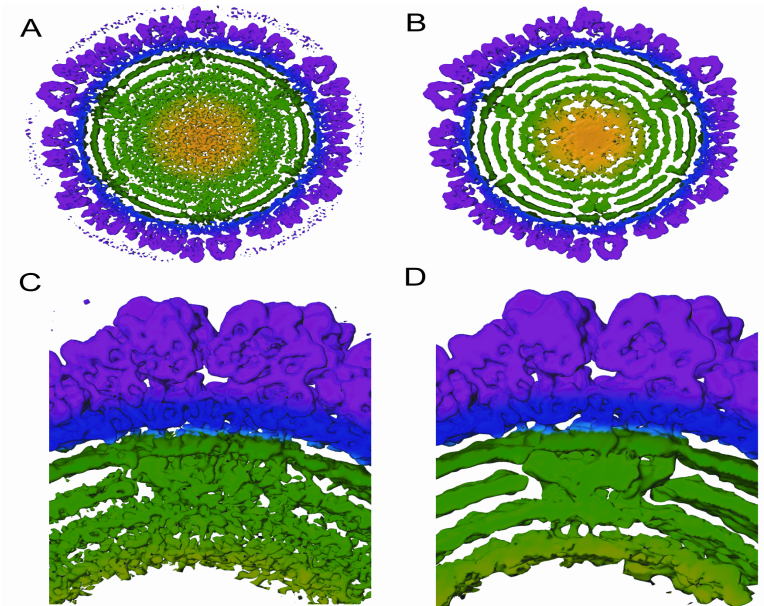


C. Bajaj, G. Xu, ACM Transactions on Graphics, (2003),22(1), pp. 4- 32.

Anisotropic diffusion filtering

$$\partial_t \phi - \text{div}(a(|\nabla \phi|) \nabla \phi) = 0$$

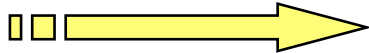
where \mathbf{a} stands for the diffusion tensor determined by local curvature estimation.

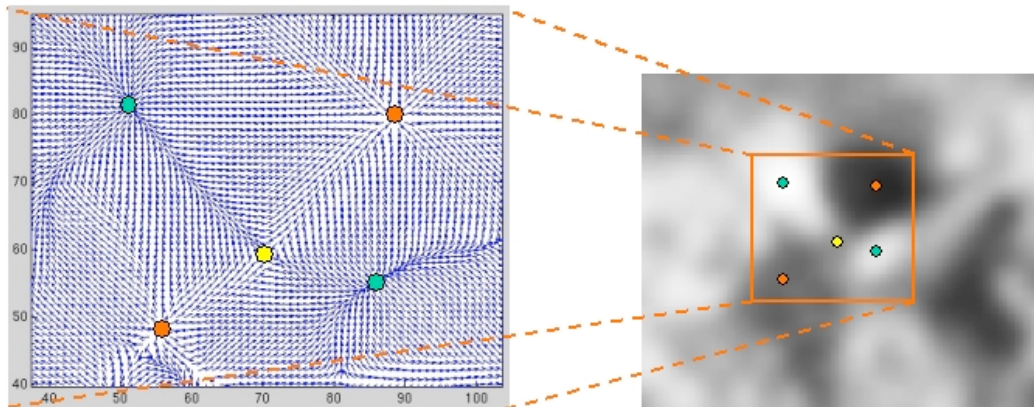


W. Jiang, M. Baker, Q. Wu, C. Bajaj, W. Chiu, Journal of Structural Biology, 144, 5,(2003), Pages 114-122



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

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

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

Z. Yu, C. Bajaj Proc. ICPR 2002, vol 2, 941- 944
 and IEEE Transactions on Image Processing, 2005,
 14, 9, 1324-1337



minimum
(0)



maximum
(3)



saddle
(1, 2)



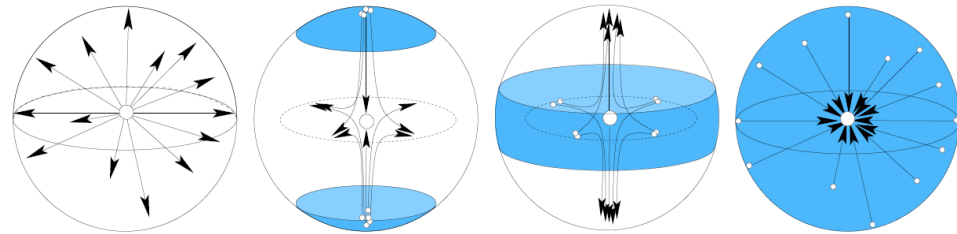
Critical Points, their Indices, and their Manifolds

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



From EHNP SoCG'03

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.

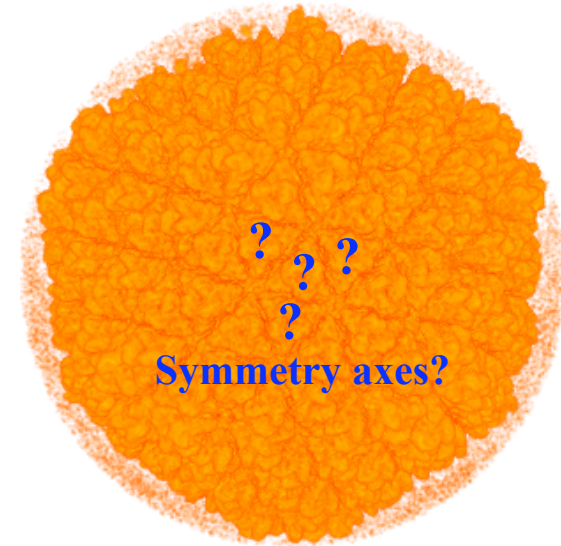
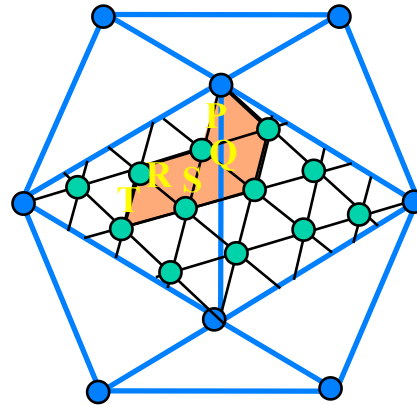
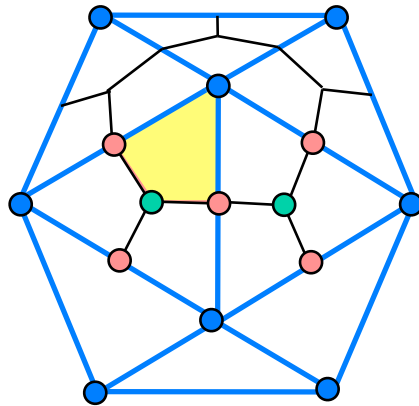
Stable Manifold of a critical point is the union of all integral curves ending at the critical point.

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



Step #3: Symmetry Detection

- Asymmetric subunits in an icosahedra



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

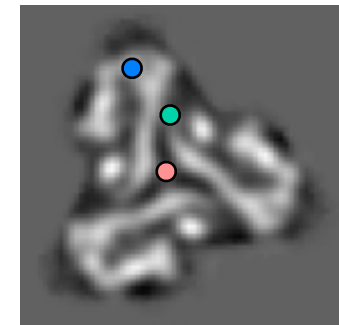
Local symmetry (RDV)
(260 trimers or 720 proteins)

- Critical Point Correlation search, addtly sped up by Spherical FFT:

Find best c , minimizing:

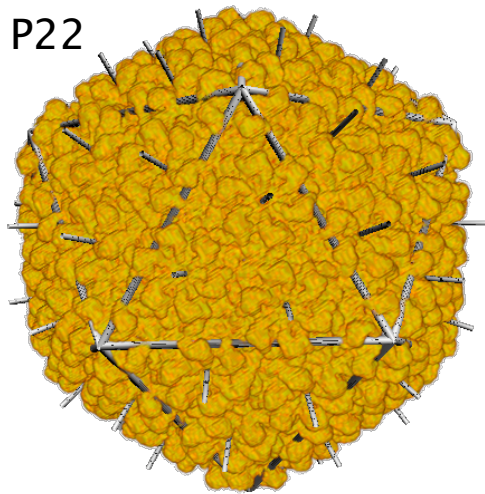
$$\sum_{\vec{r} \in D_0} (f(\vec{r}) - f(R_{2\pi/n}(c) \times \vec{r}))^2$$

Z. Yu, C. Bajaj IEEE Transactions on Image Processing, 2005, 14, 9, 1324-1337

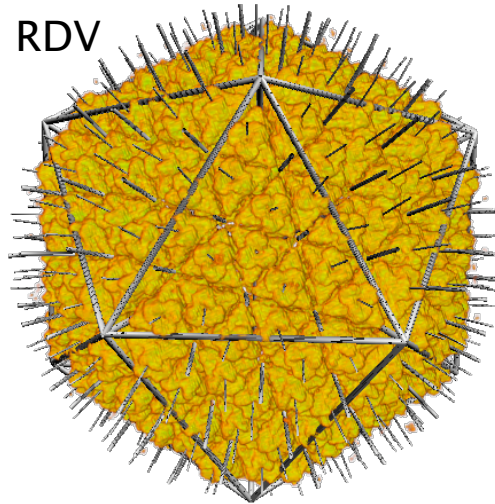


Results of Automatic Symmetry Detection in Virus and Phage Capsid Shells

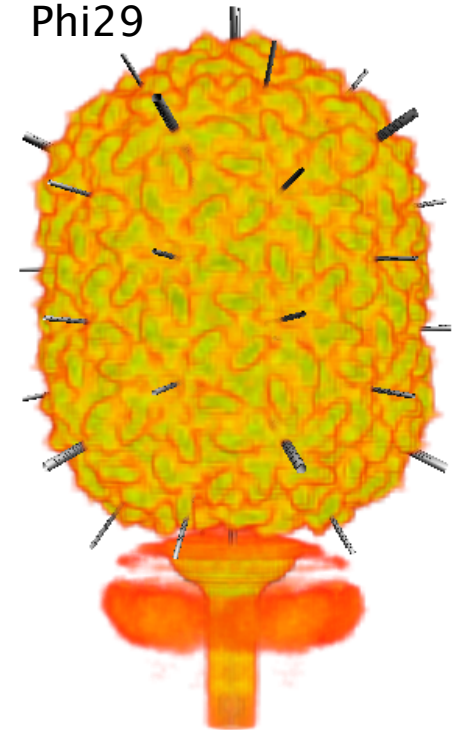
P22



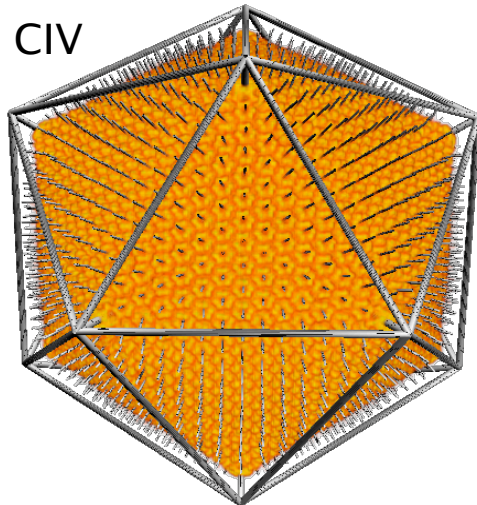
RDV



Phi29



CIV



Collaborators: Wah Chiu (BCM), Tim Baker (UCSD)



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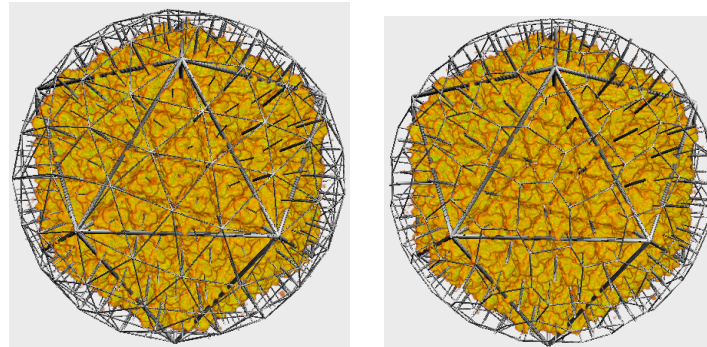
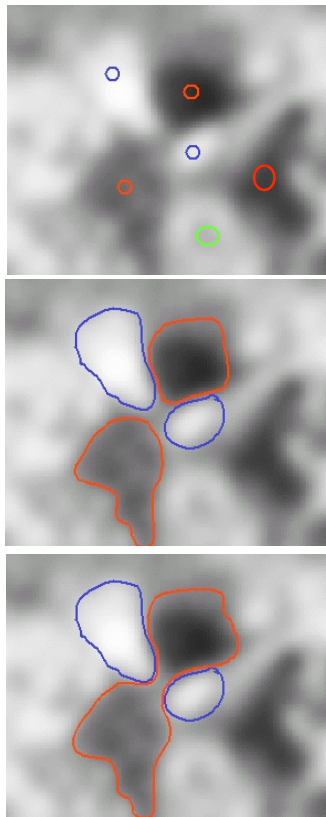
University of Texas at Austin

January 2008

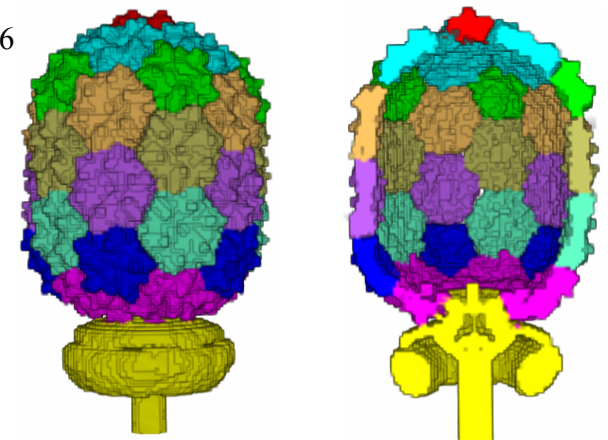
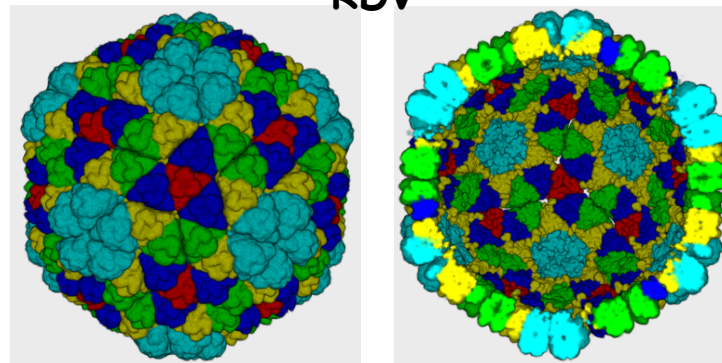
Step #4: Subunit Segmentation

- 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.

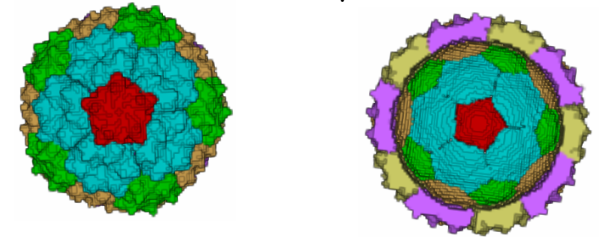
Zeyun, Bajaj IEEE Trans on Imag. Proc.,2005, Baker, Yu, Chiu, Bajaj, J S B 2006



RDV



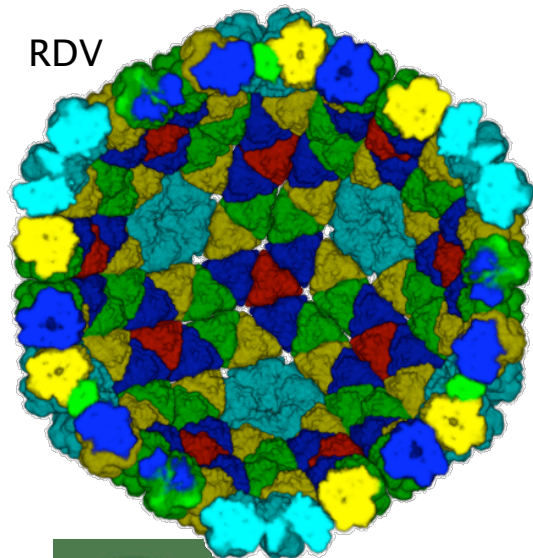
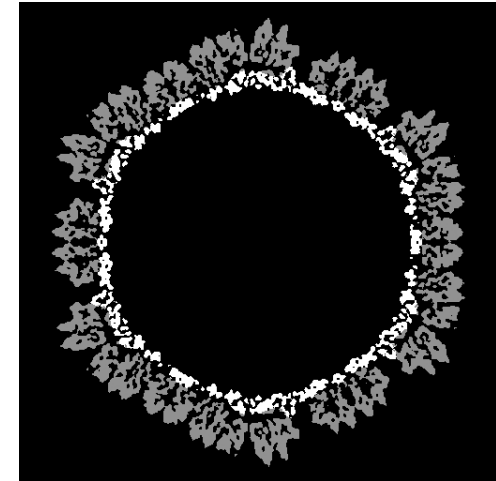
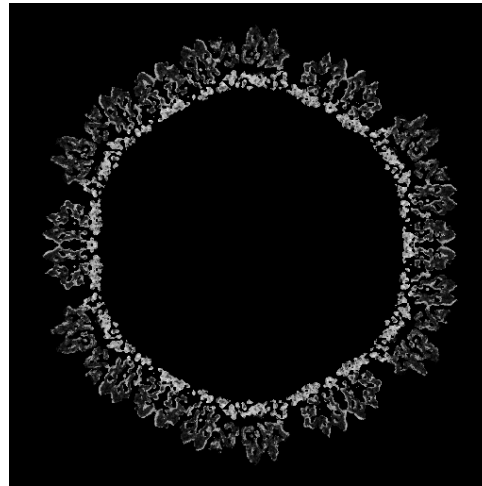
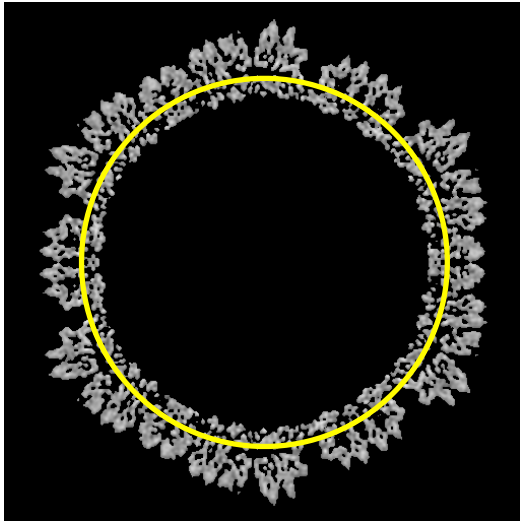
$\Phi 29$ data courtesy: Tim Baker (UCSD)



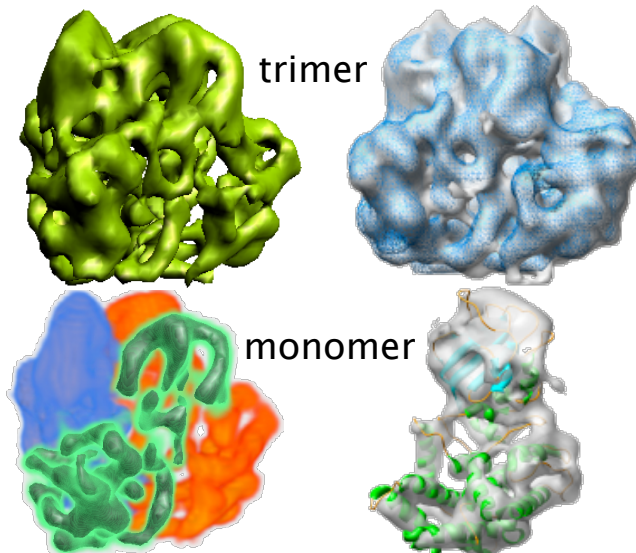
Bajaj, Xu 2007 "A Higher Order Level Set Method" ICES-TR



Capsid and Subunit Segmentation



RDV



trimer

monomer

Correlation Score with manually segmented subunits:

trimer: **0.74**
monomer: **0.80**

Correlation Score with blurred PDB density map:

trimer: **0.85**
monomer: **0.84**



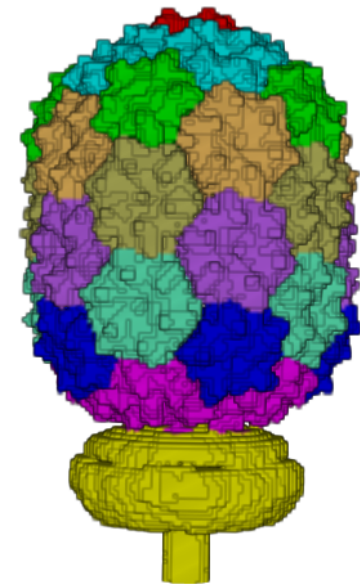
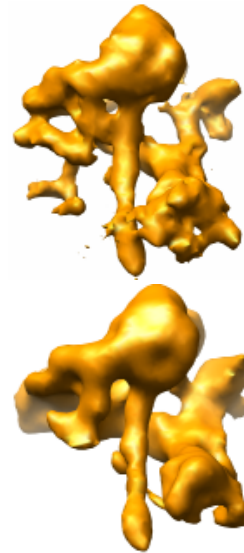
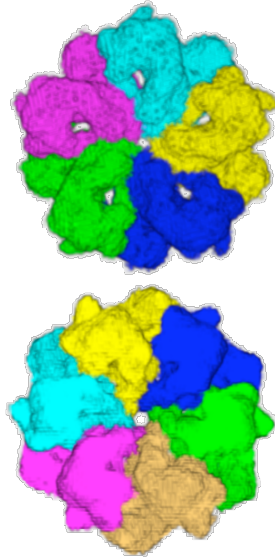
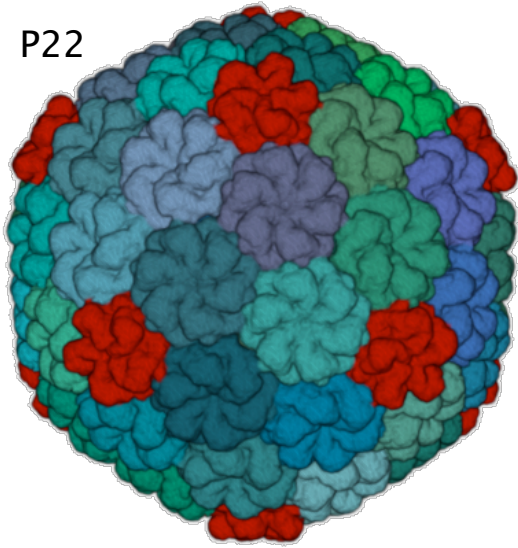
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Subunit Segmentation

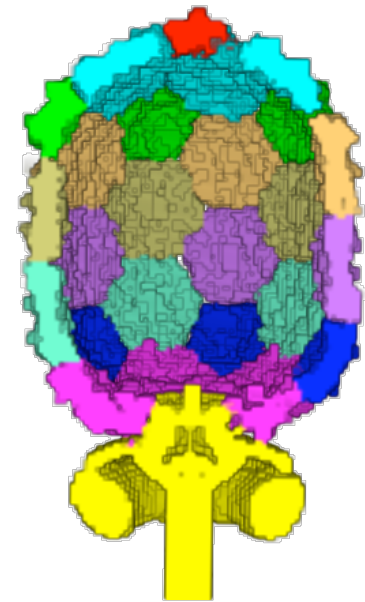
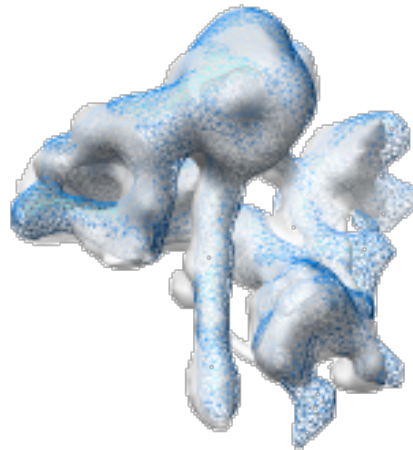
P22



Correlation Score with manually segmented subunits:

5-fold: **0.72**

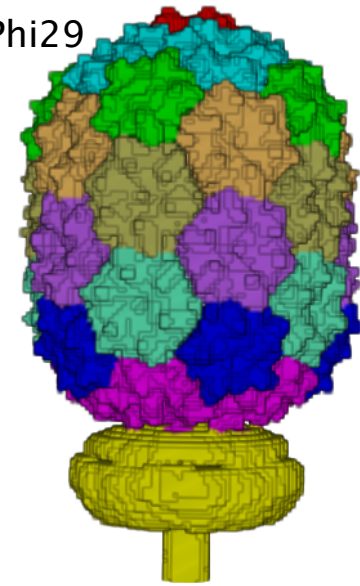
6-fold: **0.79**



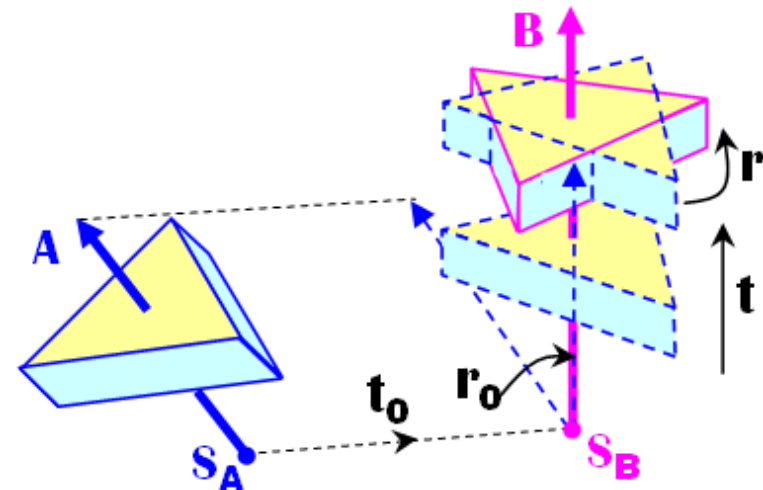
Step #4b: Subunit Alignment

- Motivation
 - Similarity scores
 - Averaged subunit
 - Map Refinement
 - Structural fitting
- General alignment problem
 - Six degrees of freedom

Phi29



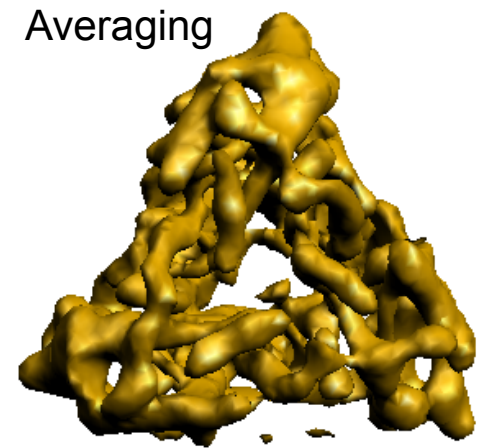
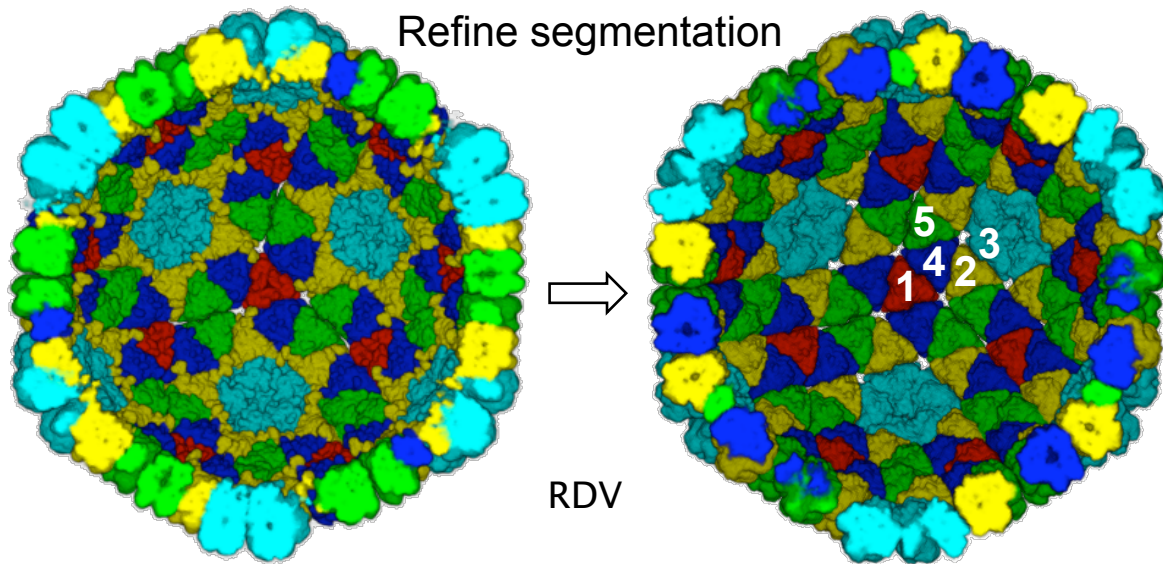
- Assumptions
 - Subunits are segmented
 - Subunits have an n-fold symmetry
 - Symmetry axes are known
- Rotation \mathbf{r} Translation \mathbf{t}
- ↓ ↓
- $0 \sim 2\pi/n$ $-5 \sim 5$ pixels



Two degrees of freedom



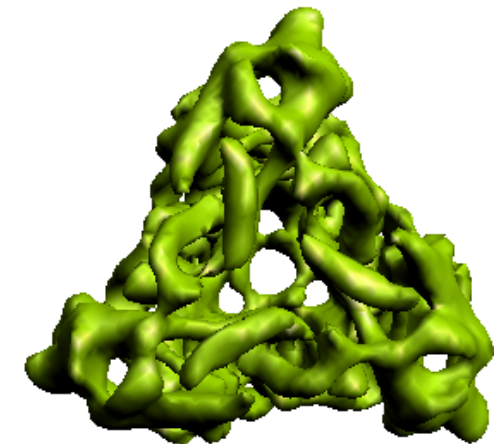
Subunit Alignment (RDV map refinement)



A single RDV trimer

Similarity table

Similarity scores	1	2	3	4	5
1	0.955	0.900	0.894	0.911	0.848
2		0.934	0.885	0.889	0.848
3			0.856	0.880	0.845
4				0.926	0.854
5					0.872

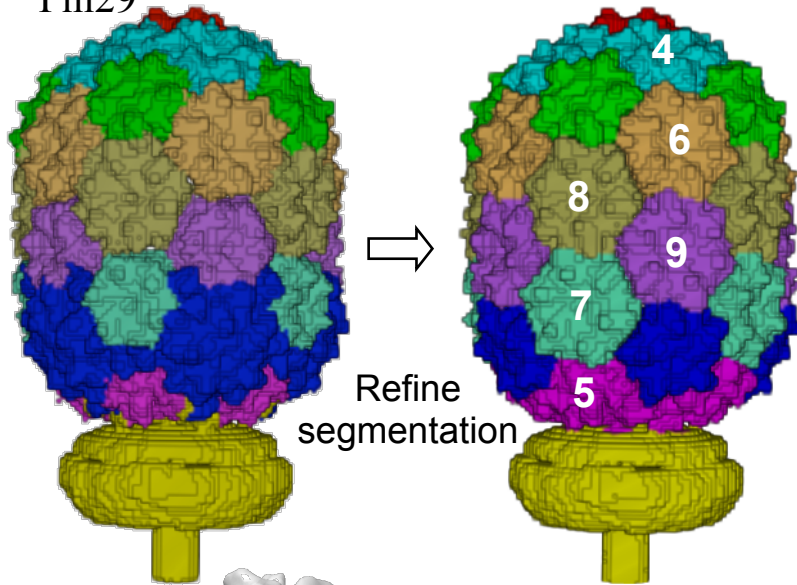


Averaged trimer



Subunit Alignment (Simul. Structure Fitting)

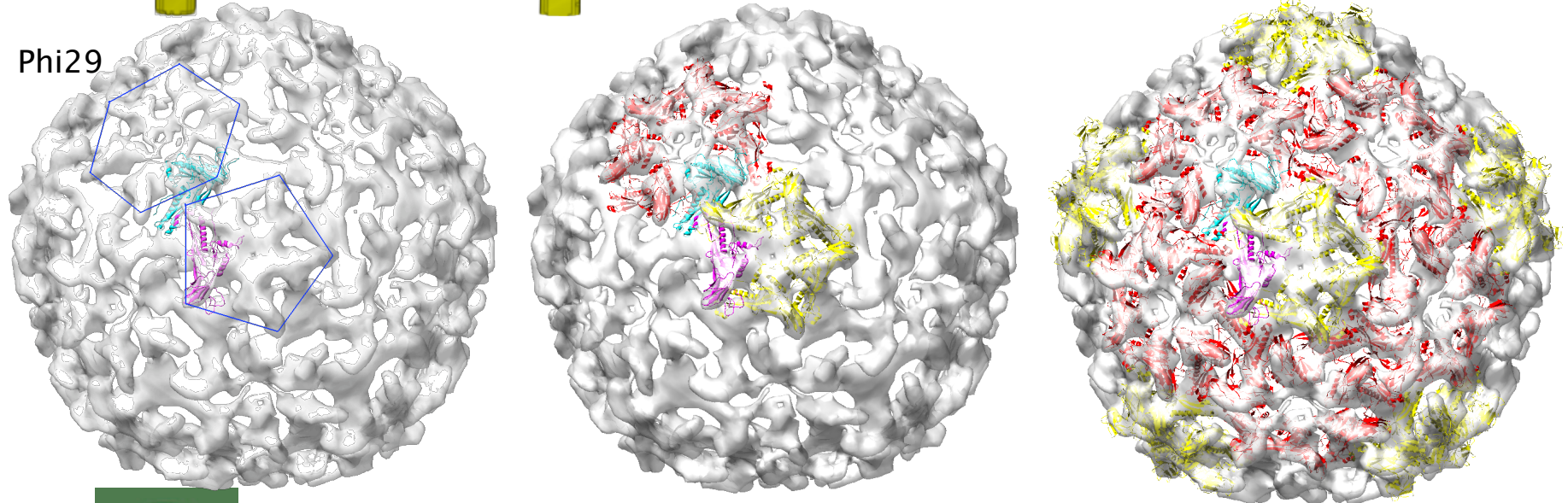
Phi29



Similarity table

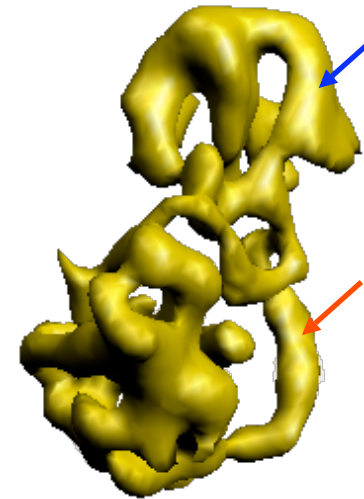
	#4	#5	#6	#7	#8	#9
#4	0.97	0.79	0.95	0.94	0.87	0.88
#5		0.81	0.79	0.78	0.77	0.79
#6			0.98	0.96	0.88	0.88
#7				0.96	0.89	0.88
#8					0.87	0.94
#9						0.81

Phi29

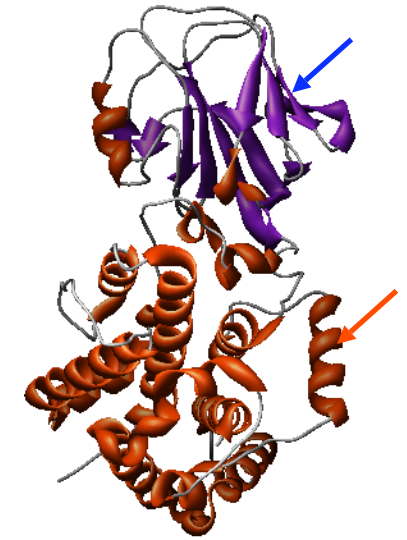


Step #5: Secondary Structure Identification

- Motivation
 - One more step towards quasi-atomic modeling
- Two types of secondary structures
 - Alpha-helices (cylindrical shape)
 - Beta-sheets (planar shape)
- Prior work
 - Alpha-helix detection (Jiang et al, 2001)
 - Beta-sheet detection (Kong et al, 2003)



RDV P8 protein
(6.8Å)



RDV P8 protein
(PDB: 1UF2)



Prototypes for alpha-helices and beta-sheets

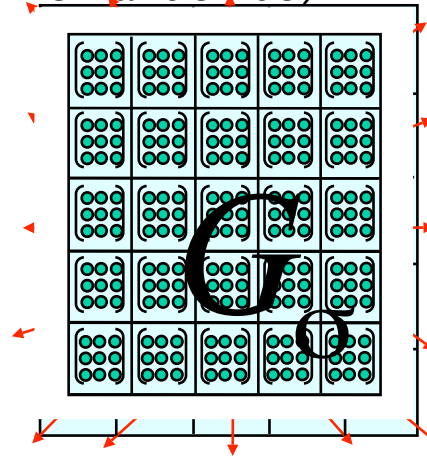


Step #5A: Secondary Structure Identification

- Gradient tensor

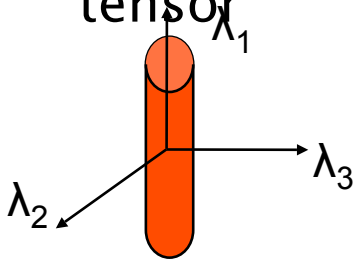
$$(I_x, I_y, I_z) \rightarrow \begin{pmatrix} I_x^2 & I_x I_y & I_x I_z \\ I_x I_y & I_y^2 & I_y I_z \\ I_x I_z & I_y I_z & I_z^2 \end{pmatrix}$$

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



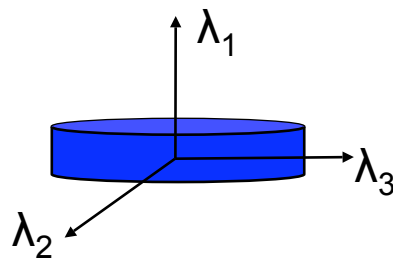
$$\begin{pmatrix} I_x^2 * G_\sigma & I_x I_y * G_\sigma & I_x I_z * G_\sigma \\ I_x I_y * G_\sigma & I_y^2 * G_\sigma & I_y I_z * G_\sigma \\ I_x I_z * G_\sigma & I_y I_z * G_\sigma & I_z^2 * G_\sigma \end{pmatrix}$$

- Property of local structure tensor



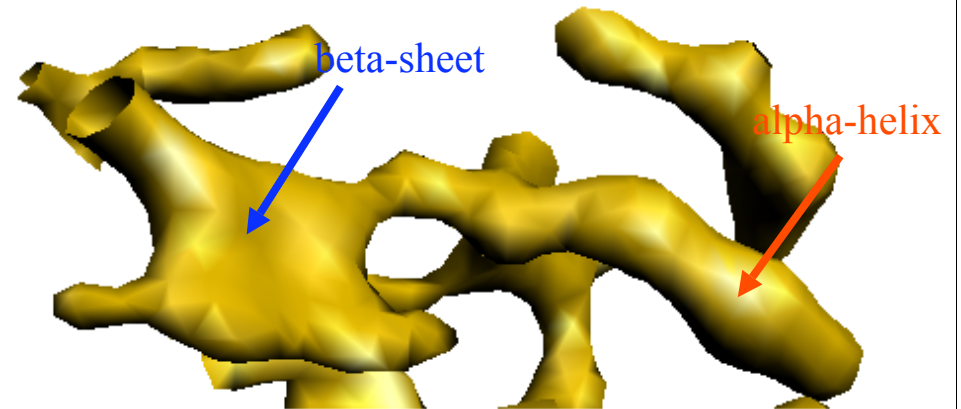
Line structure

$$\lambda_2 \approx \lambda_3 \gg \lambda_1 \approx 0$$

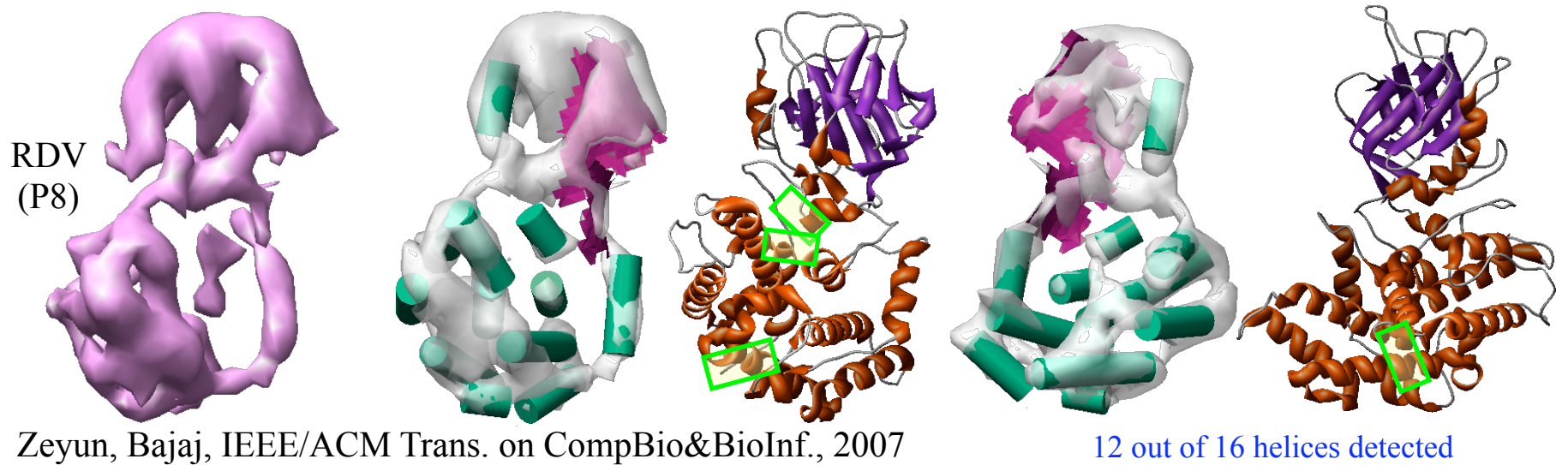
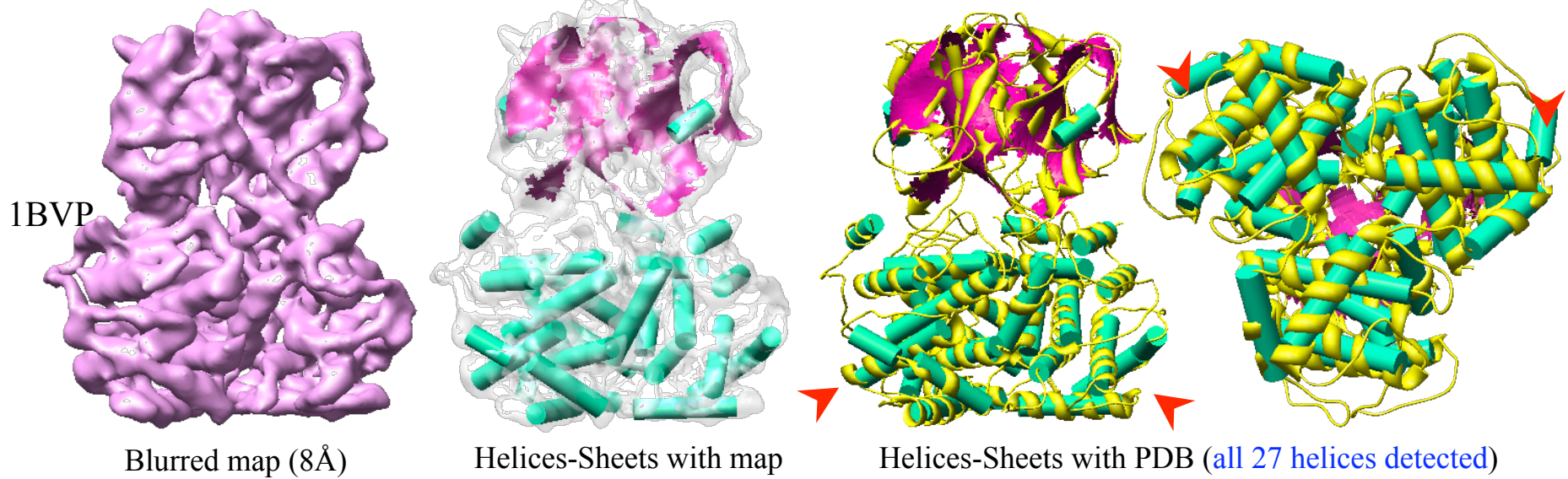


plane structure

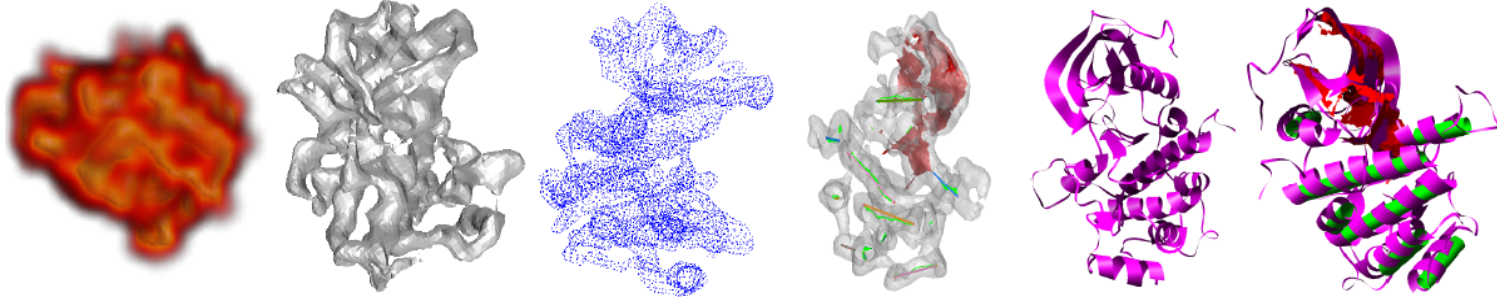
$$\lambda_1 \gg \lambda_2 \approx \lambda_3 \approx 0$$



Step #5A: Secondary Structure Identification



Step #5B: Secondary Structure Elucidation



Step 1: Vor/Del(P) computation (using CGAL).

Goswami, Dey, Bajaj, ICVGIP'2006

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.

[By DGRS2005, Critical points are either near S or near M]

Step 4: Classification of Medial Axis via

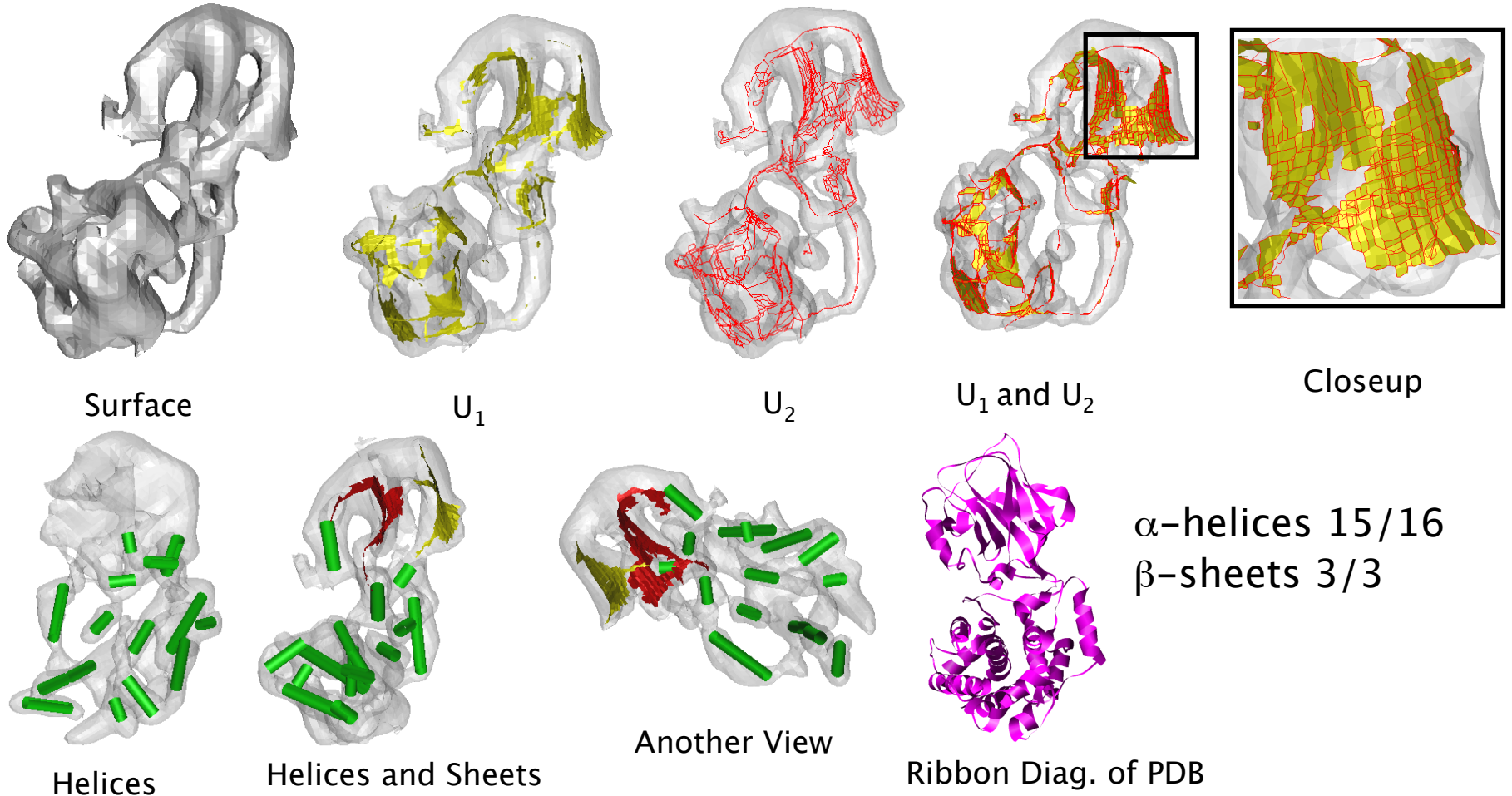
4.a: U_1 – Unstable Manifold of index 1 saddle point on M

4.b: U_2 – Unstable Manifold of index 2 saddle point on M.

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



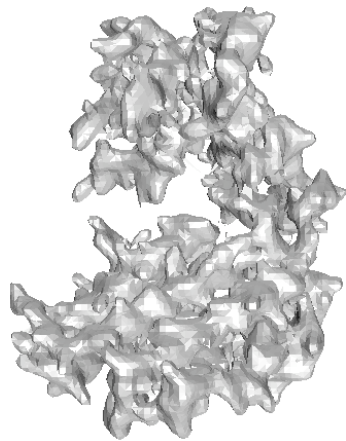
Secondary Structure of RDV Outer Capsid Coat Protein P8



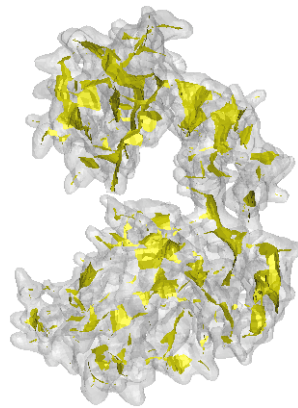
3D EM map of P8 segmented out from cryo-EM map of Rice Dwarf Virus (RDV) at 6.8 Å resolution



Secondary Structure of Bacterial Chaperonin GroEL



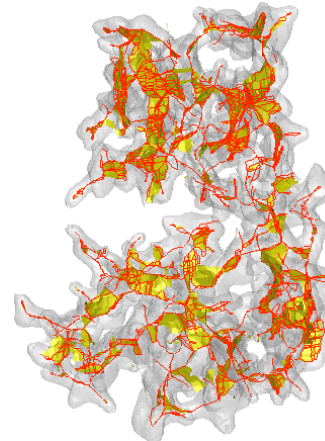
Surface



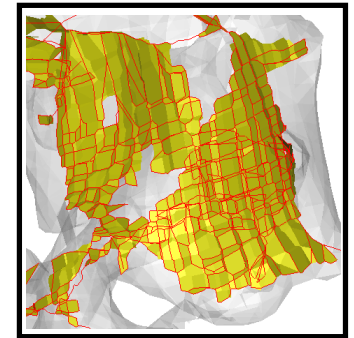
U₁



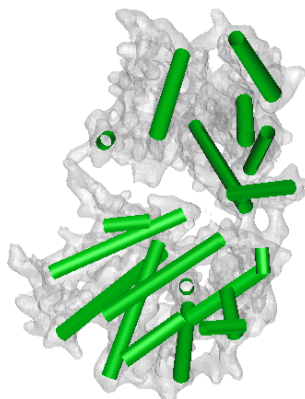
U₂



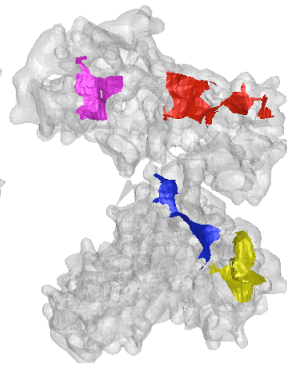
U₁ and U₂



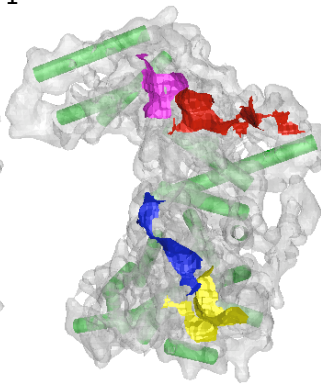
Closeup



Helices



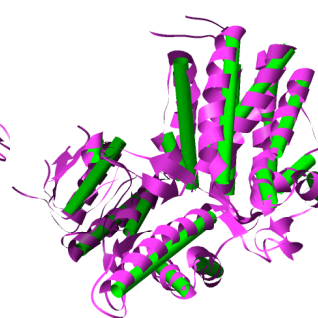
Sheets



Helices and
Sheets



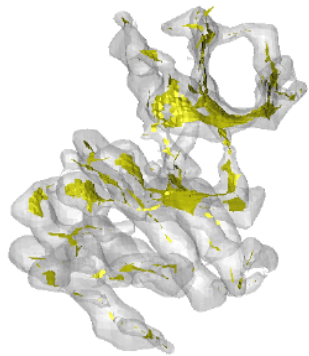
Ribbon Diag.
of PDB



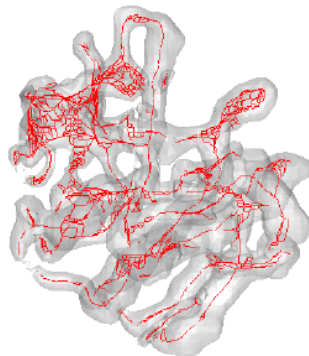
Superimposed
 α -helices 15/16
 β -sheets 3/3



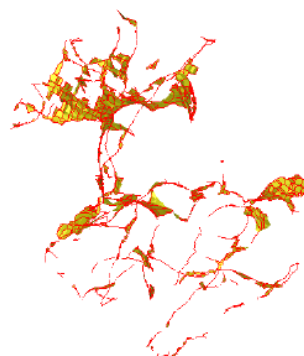
Width Test



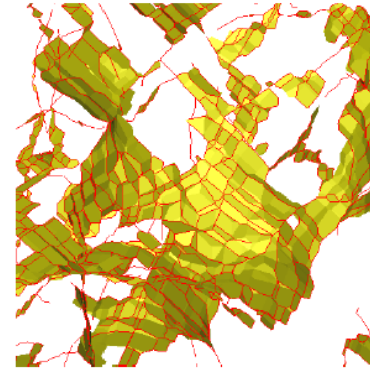
U_1



U_2

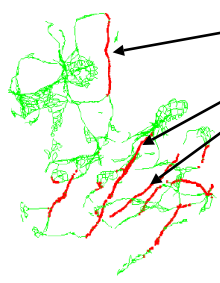
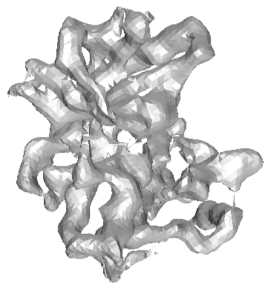


$\hat{U}_1 \text{ and } U_2$



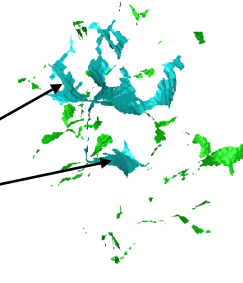
Closeup: U_1 is bounded by U_2

- 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.

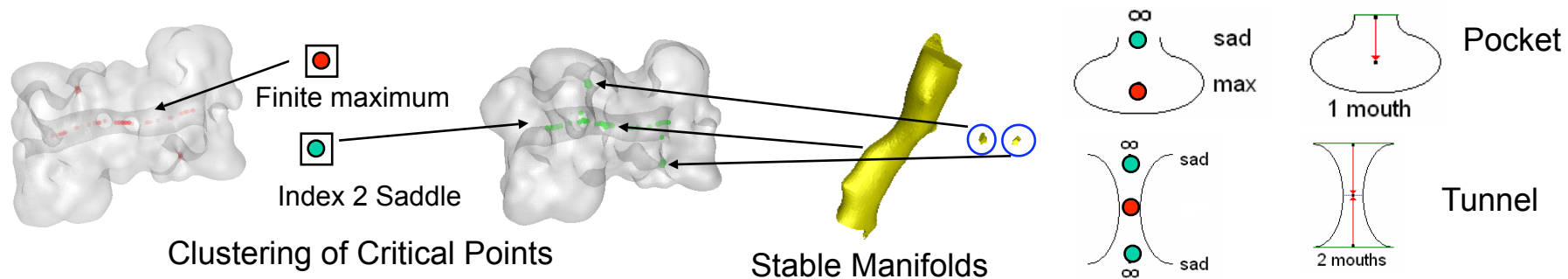


Subset of U_2 that corresponds to helices

Subset of U_1 that corresponds to sheets

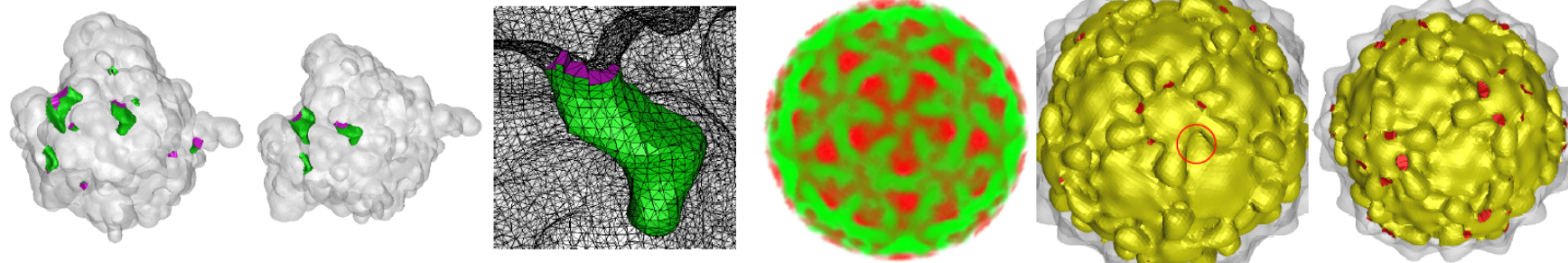


Molecular Surface Curvature & Detection of Pockets/Tunnels



- Build adjacency relation between two finite maxima if they share a common boundary.
- Compute the Transitive Closure of this relation. This gives a clustering of the maxima and the index-2 saddles.
- Compute S_3 for each cluster.

Nodavirus



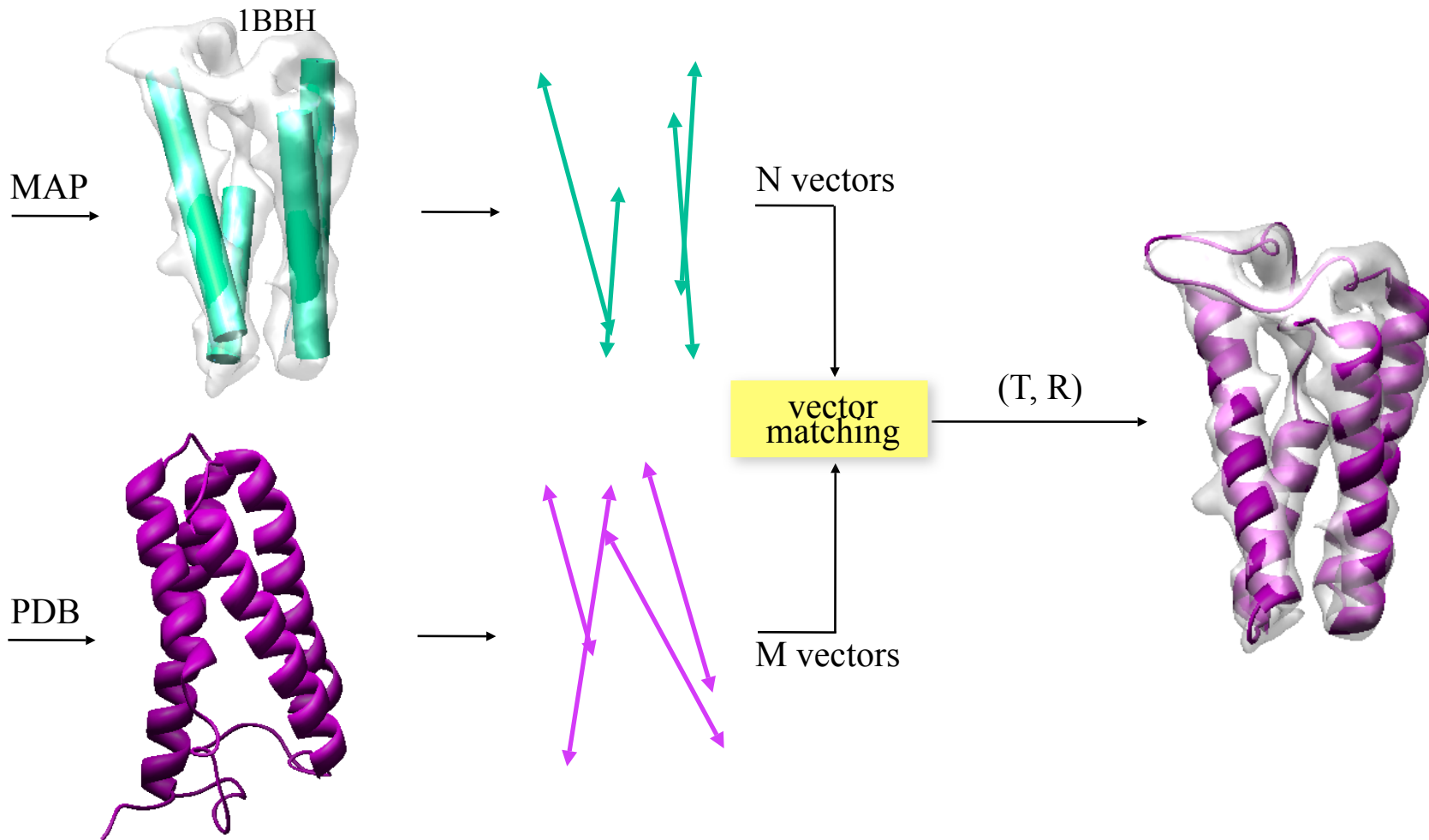
AcetylCholineEsterase

Bajaj, Gillette, Goswami
TopoInVis'2007

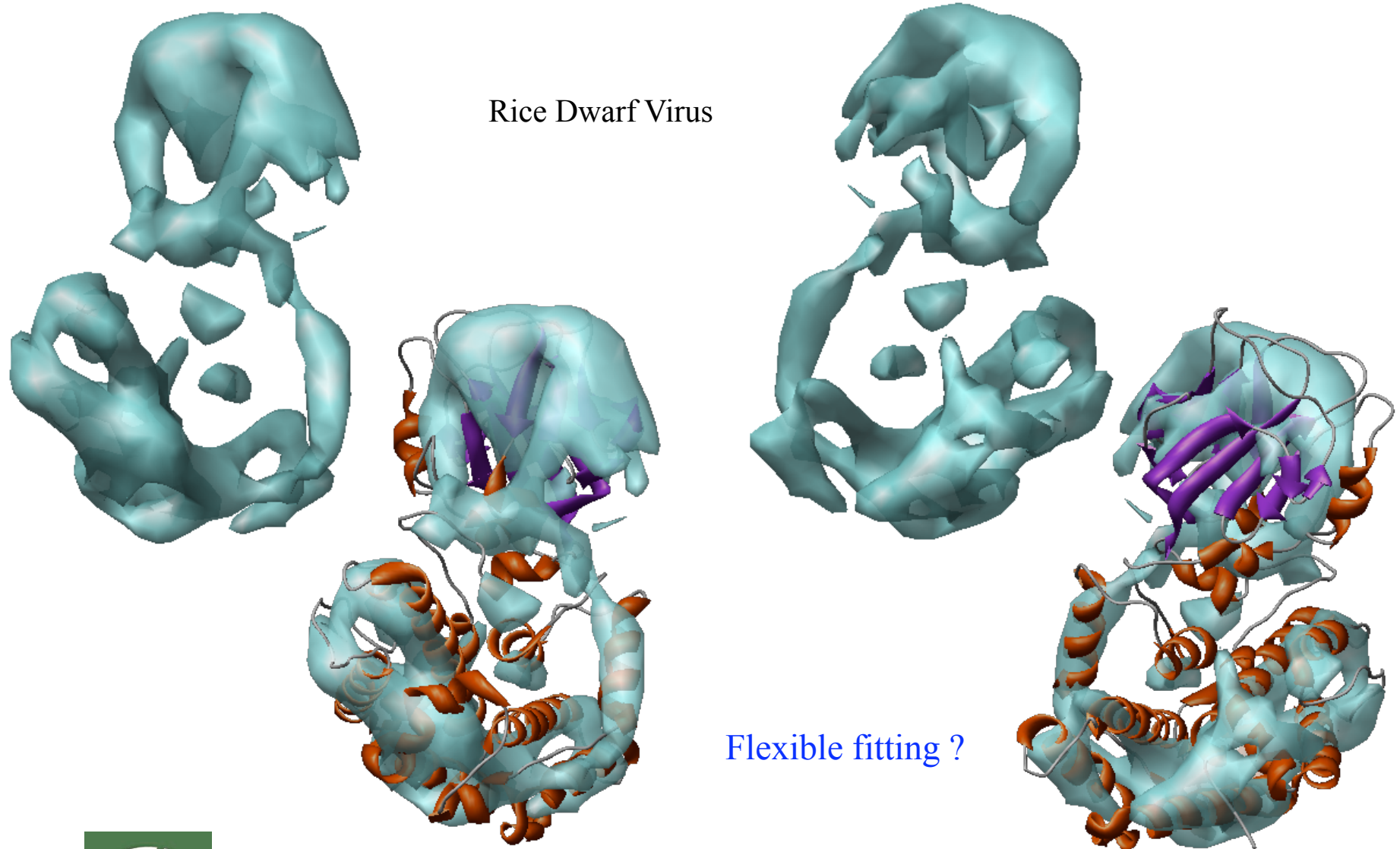


Step #6: Structure Fitting/Modeling

- PDB-based approach — vector matching



Step #6: Structure Fitting/Modeling



**Analytic Molecular Models
with extensions to
Flexible Models/Coarse Grained Models**

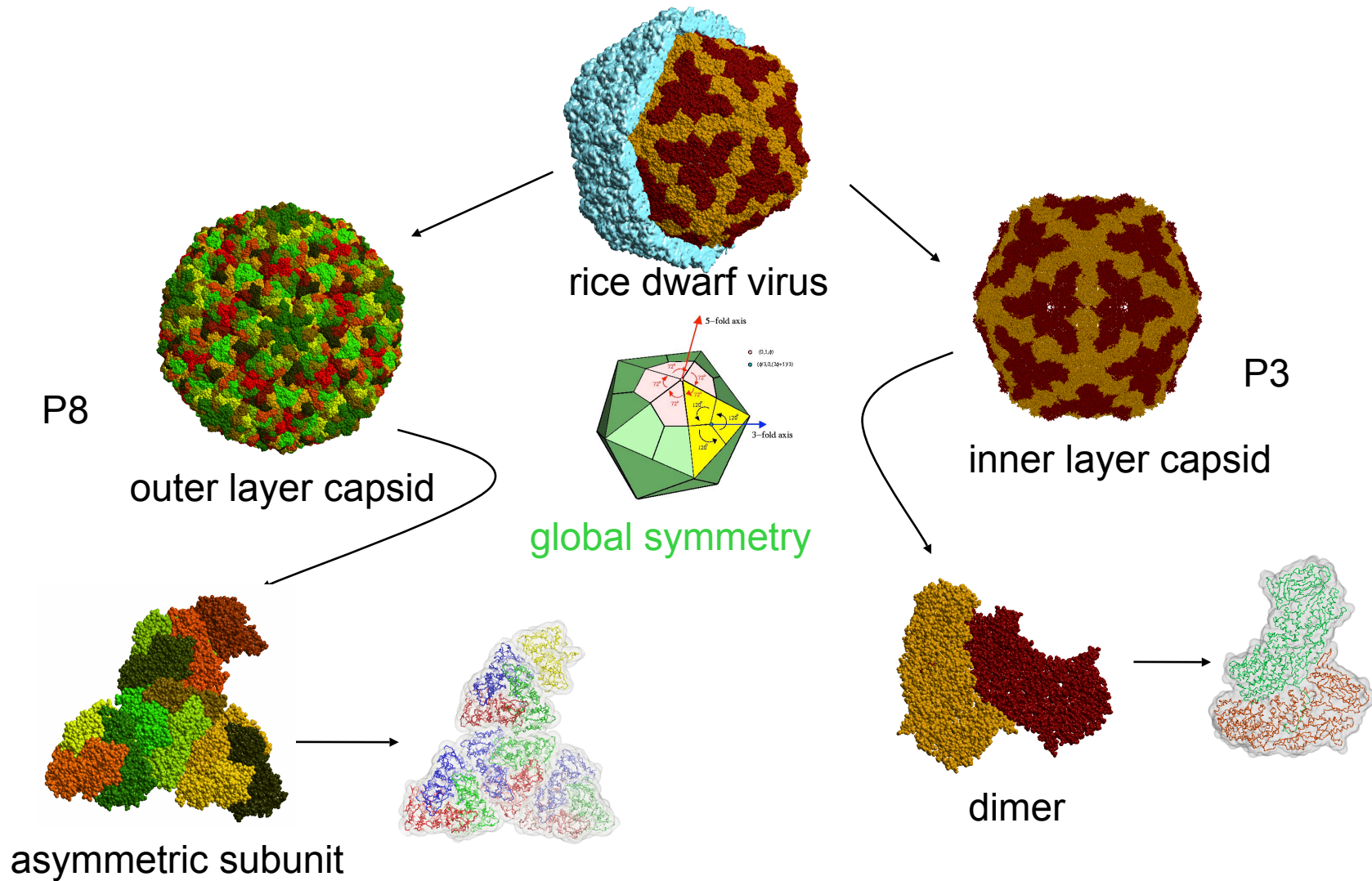


Center for Computational Visualization
Institute of Computational and Engineering Sciences
Department of Computer Sciences

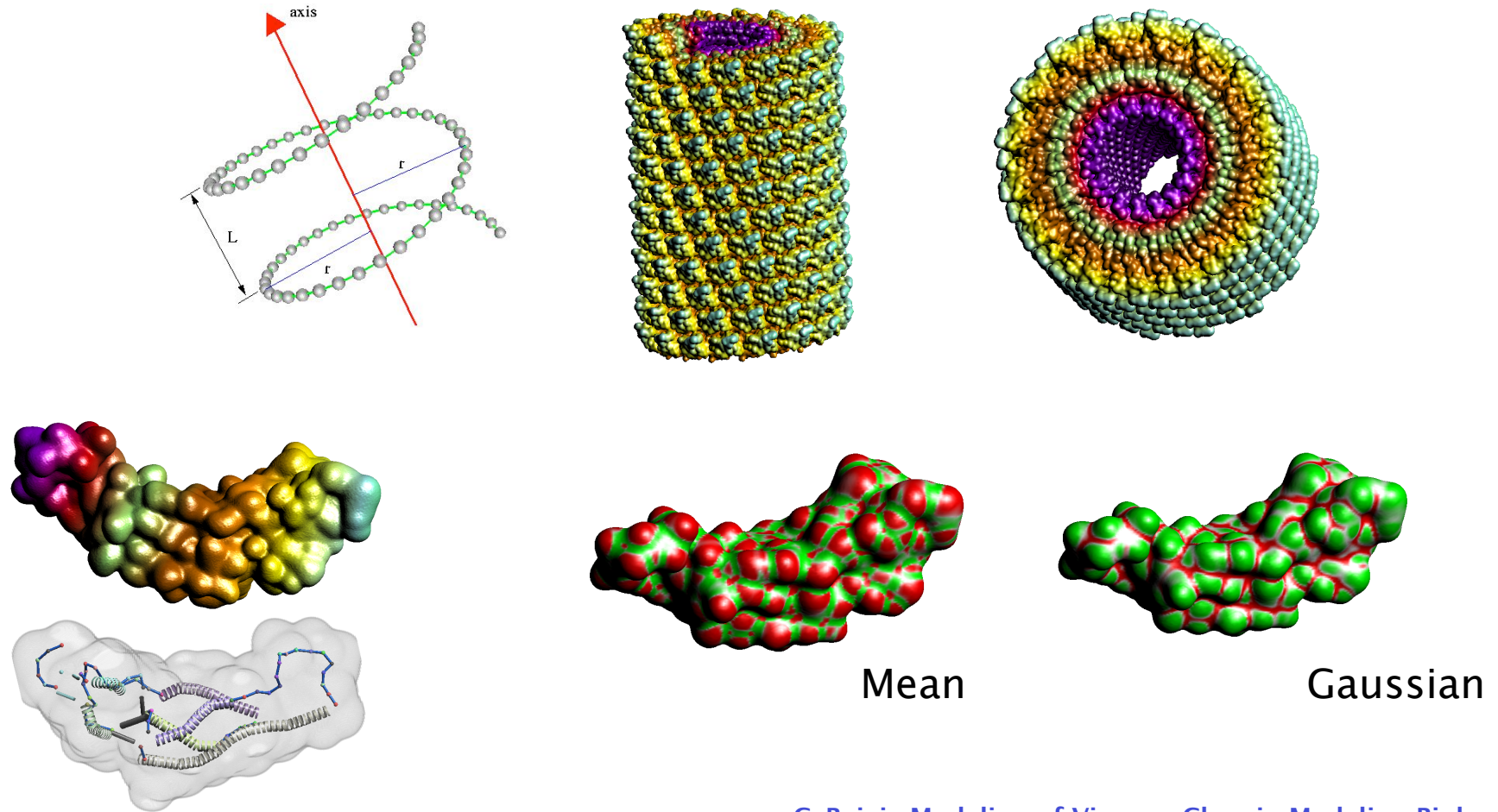
University of Texas at Austin

January 2008

Rice Dwarf Virus Capsid (1UF2)



Helical: Tobacco Mosaic Virus (1EI7)



C. Bajaj, Modeling of Viruses, Chap in Modeling Biology, MIT Press, (2007)



Center for Computational Visualization
Institute of Computational and Engineering Sciences
Department of Computer Sciences

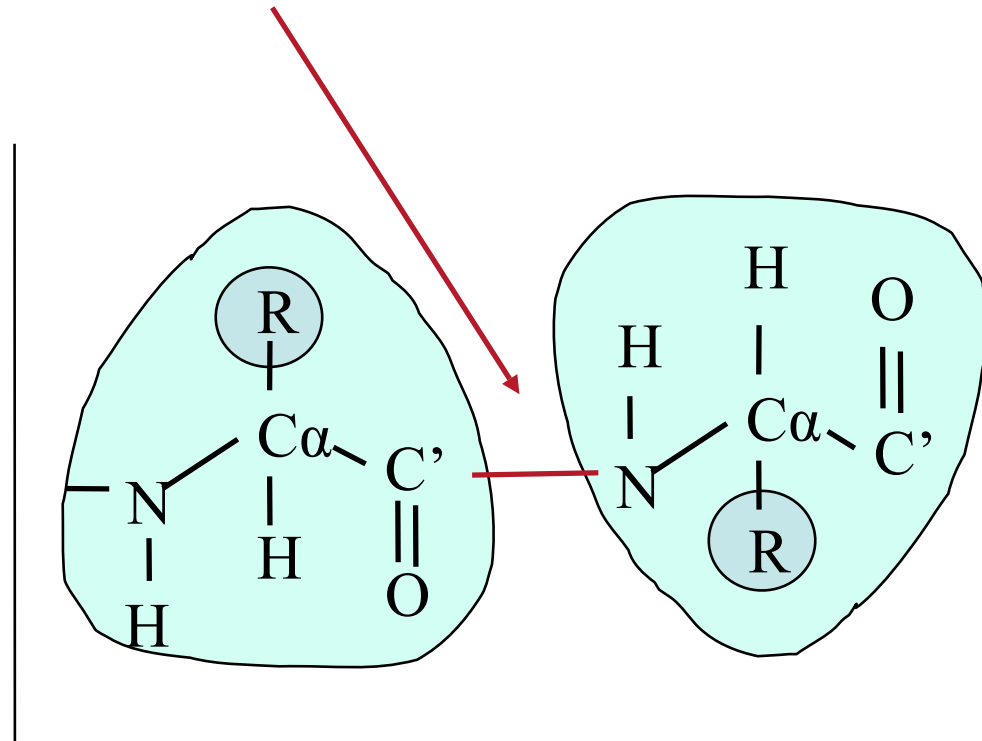
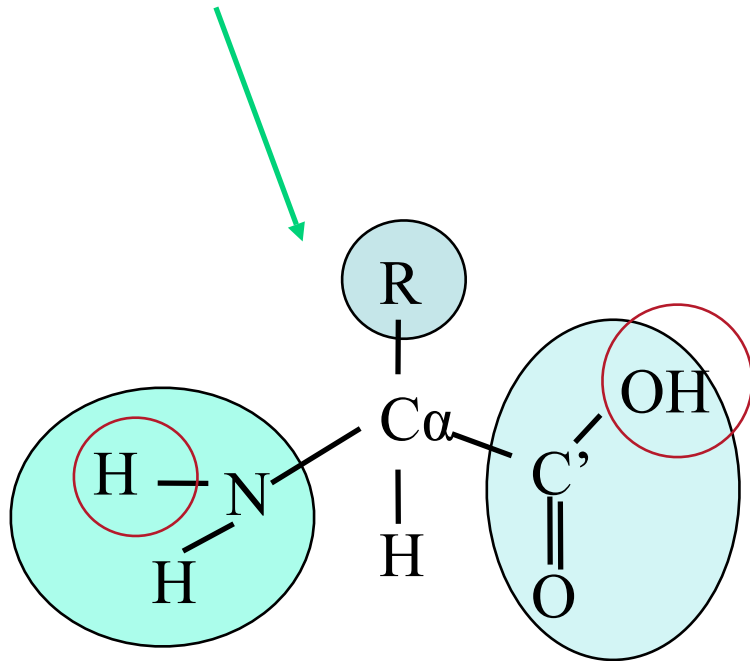
University of Texas at Austin

January 2008

Proteins

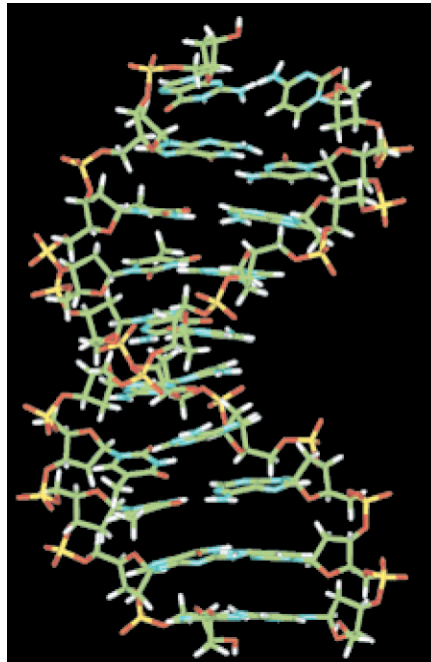


- Amino acids contain an amide, a residue and a carboxyl group
- Proteins are polypeptide chains, made from **amino acids** combined via **peptide bonds**.

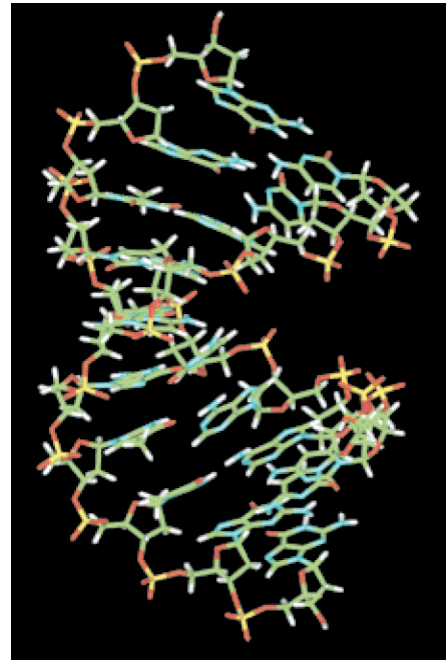


Stability of Macromolecules in Solution

d(CCAACGTTGG)



B-form DNA in water



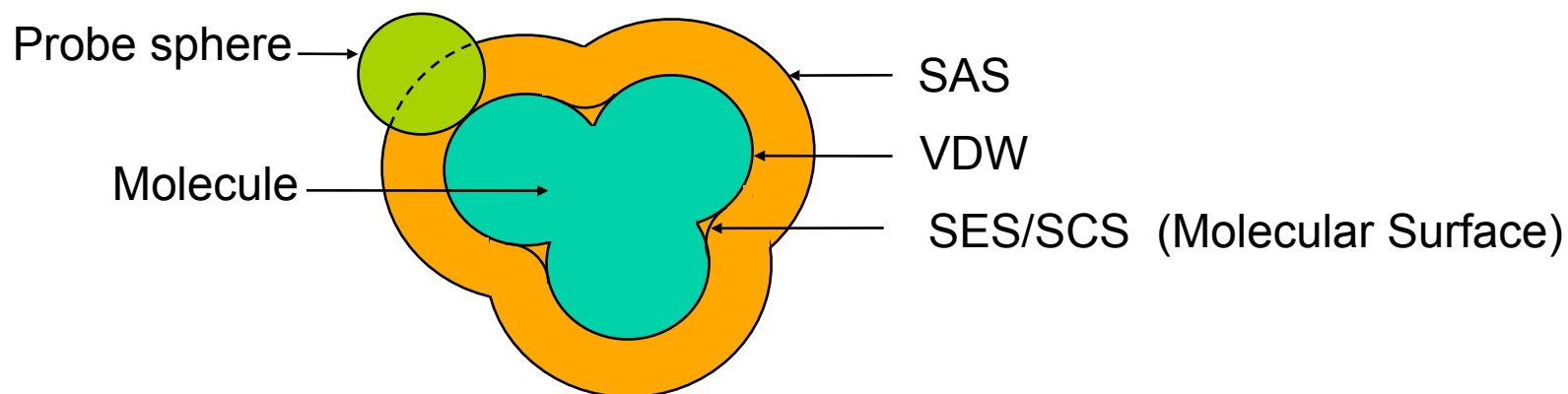
A-form DNA in ethyl

Tsui and Case, 2000



Molecules in Solvent (Implicit Model)

- Solvent molecule modeled as a sphere. Water: radius 1.4Å



SAS: solvent accessible surface: locus of probe center

VDW: van der Waals surface: Union of spheres with VDW radii

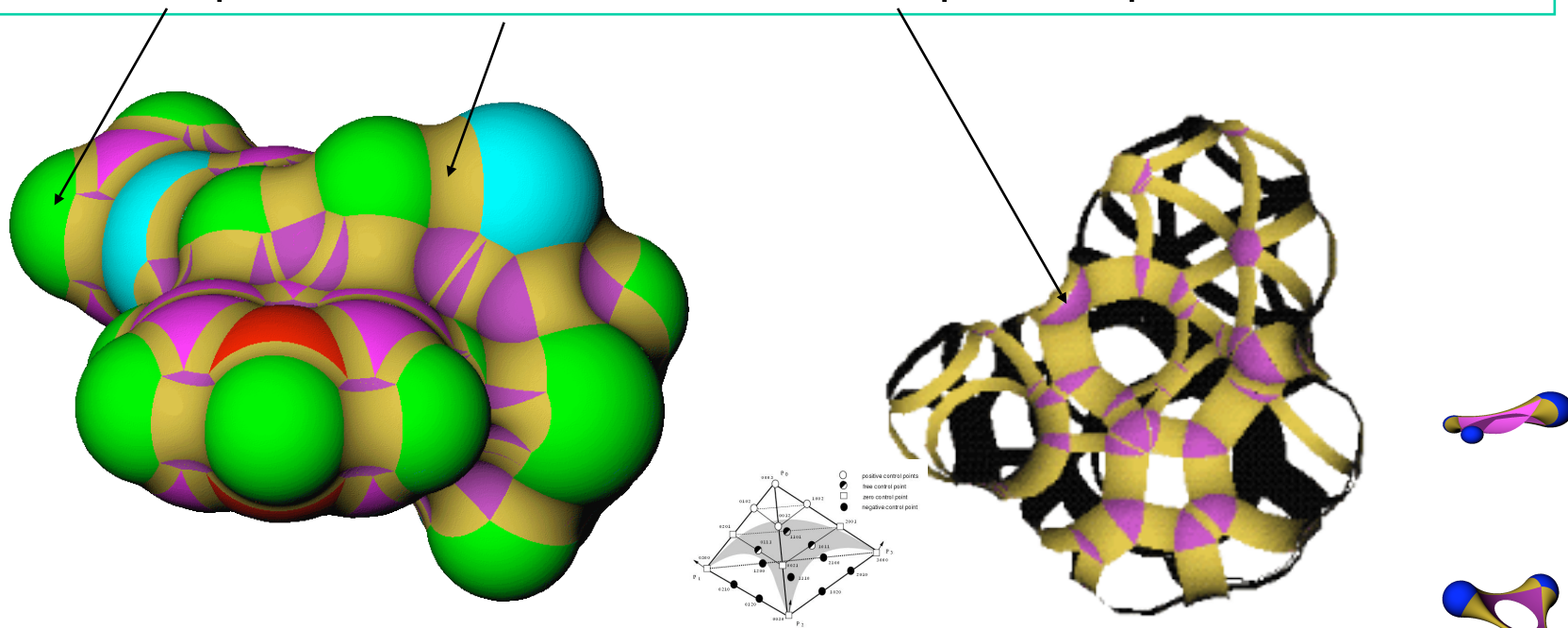
SES/SCS: solvent excluded/contact surfaces



Molecular Models I: Implicit Solvation Surface for the Hard Sphere Model

Lee-Richard (LR) surface is decomposed into three kinds of patches:

convex spherical, toroidal, and concave spherical patches



The LR surface can be represented as A-patches and NURBS

Bajaj et al, *Discrete Appl Math.* (2003), 23-51



Center for Computational Visualization
Institute of Computational and Engineering Sciences
Department of Computer Sciences

Problem with LR: Singularities

University of Texas at Austin

January 2008

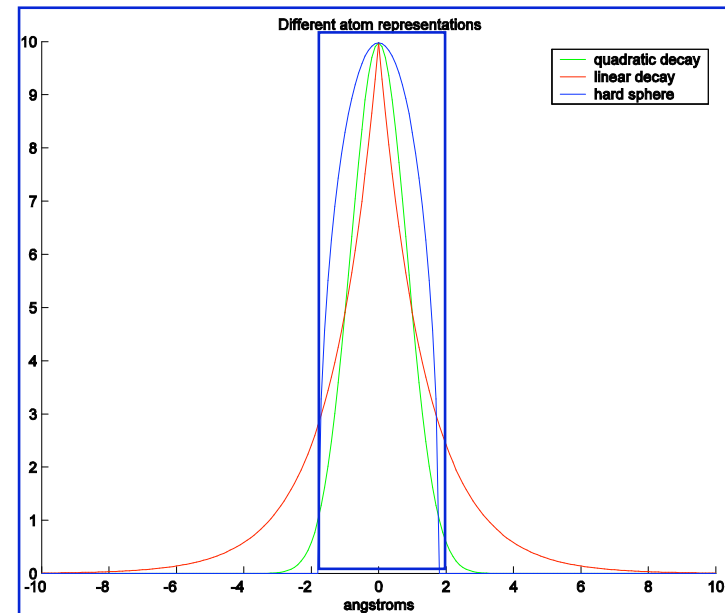
Molecular Models II: Analytic Atomic Shape Parameters

• **Isotropic Quadratic Kernel:** $G_i(\mathbf{x}) = e^{-\frac{\beta}{r_i^2}((\mathbf{x}-\mathbf{x}_i)^2 - r_i^2)}$

• **Isotropic Linear Kernel:** $G_i(\mathbf{x}) = e^{-\beta(|\mathbf{x}-\mathbf{x}_i| - r_i)}$

- where
 - The decay β controls the shape of the Gaussian function.
 - The van der Waal's radius is r_i
 - The center of the atom is \mathbf{x}_c .

• **Anisotropic Kernels**



$$\beta = 2.3, r = 1.8 \text{ \AA}$$

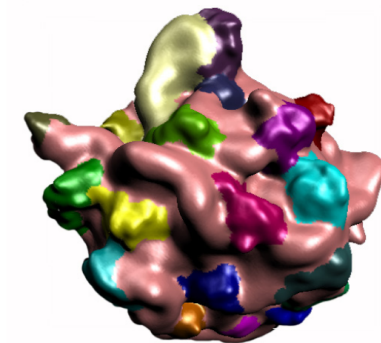
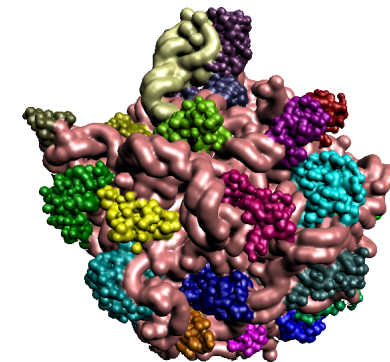
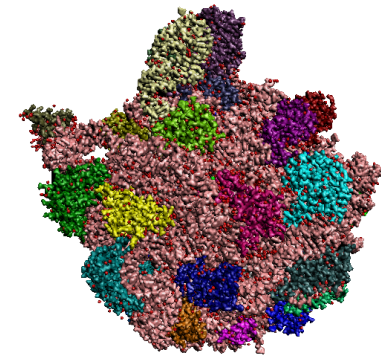
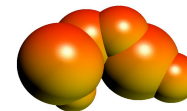
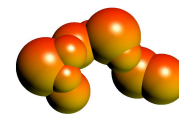
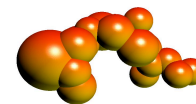
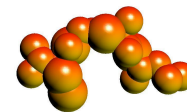
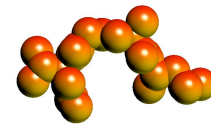
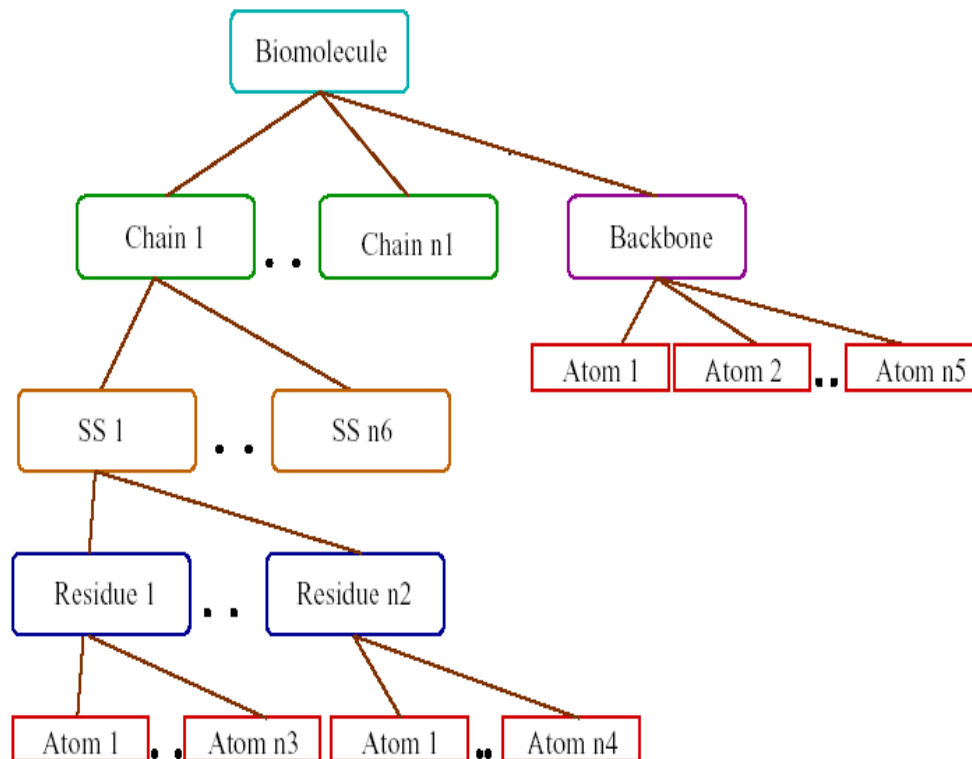
β values suggested in (Boys 50), (Grant Pickup 99)



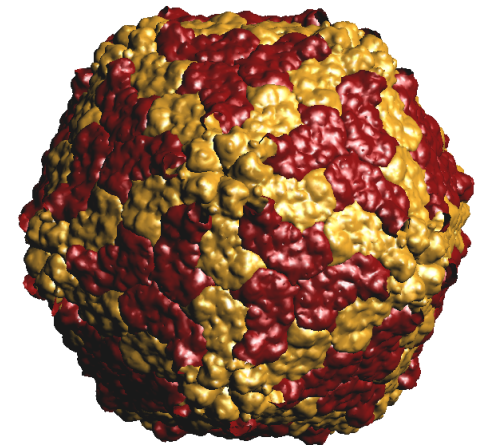
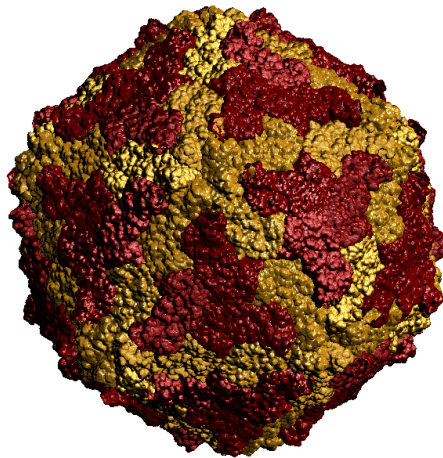
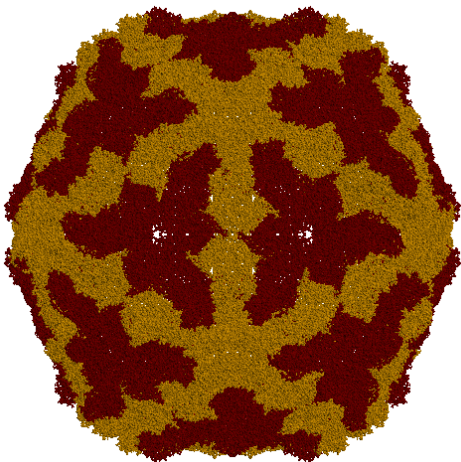
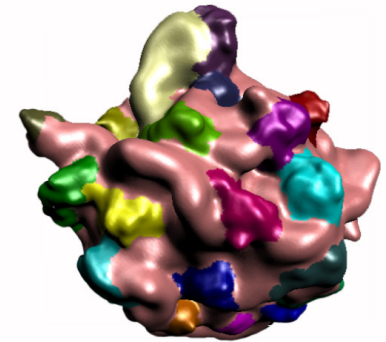
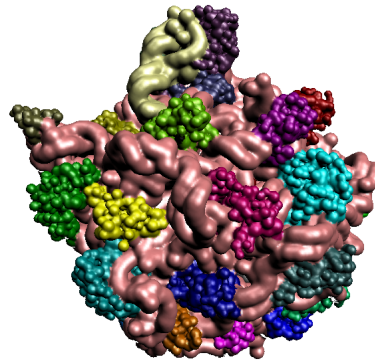
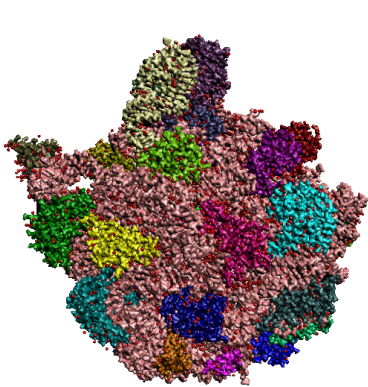
Modeling Flexibility: FCC Cluster Hierarchy

Clustering of atoms is based on

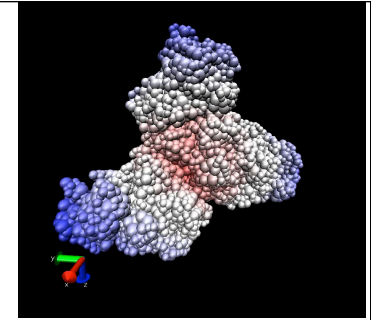
- biochemical units
- preserving molecular shape feature



FCC Models of Ribosome and Viral Capsid

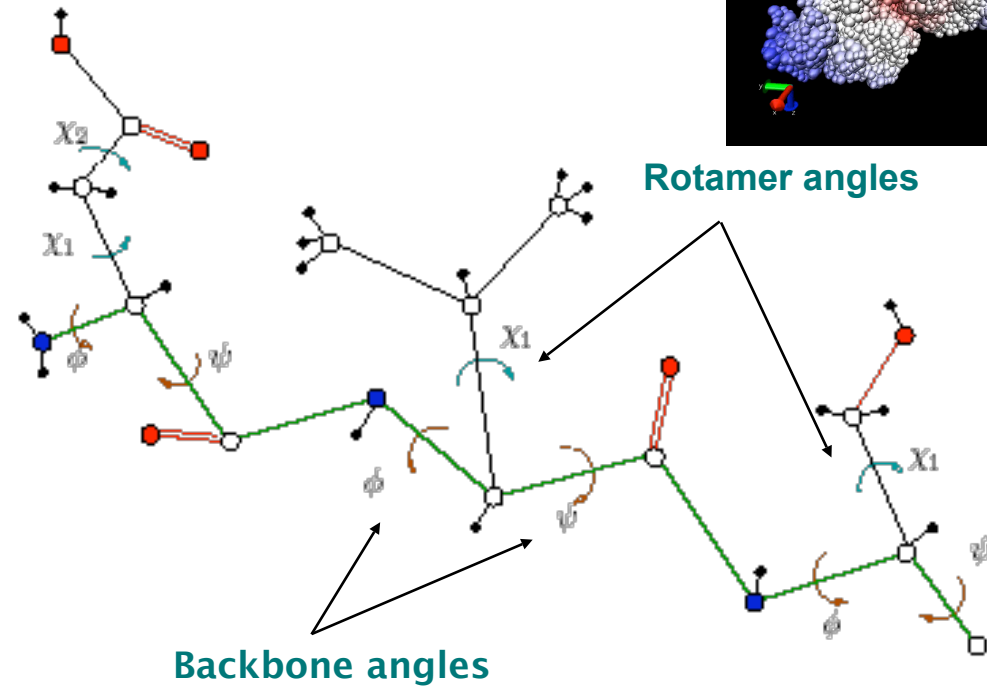


Modeling Flexibility

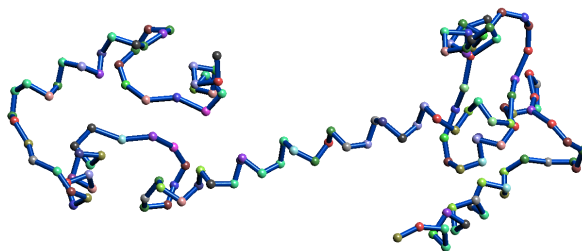


Conformational changes occur due to

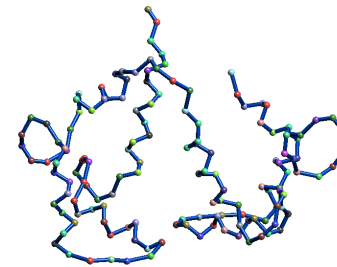
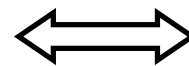
- changes in torsional angles: ϕ , ψ and χ
- hinge-type bending and shearing movements



Conformational changes in Calmodulin:



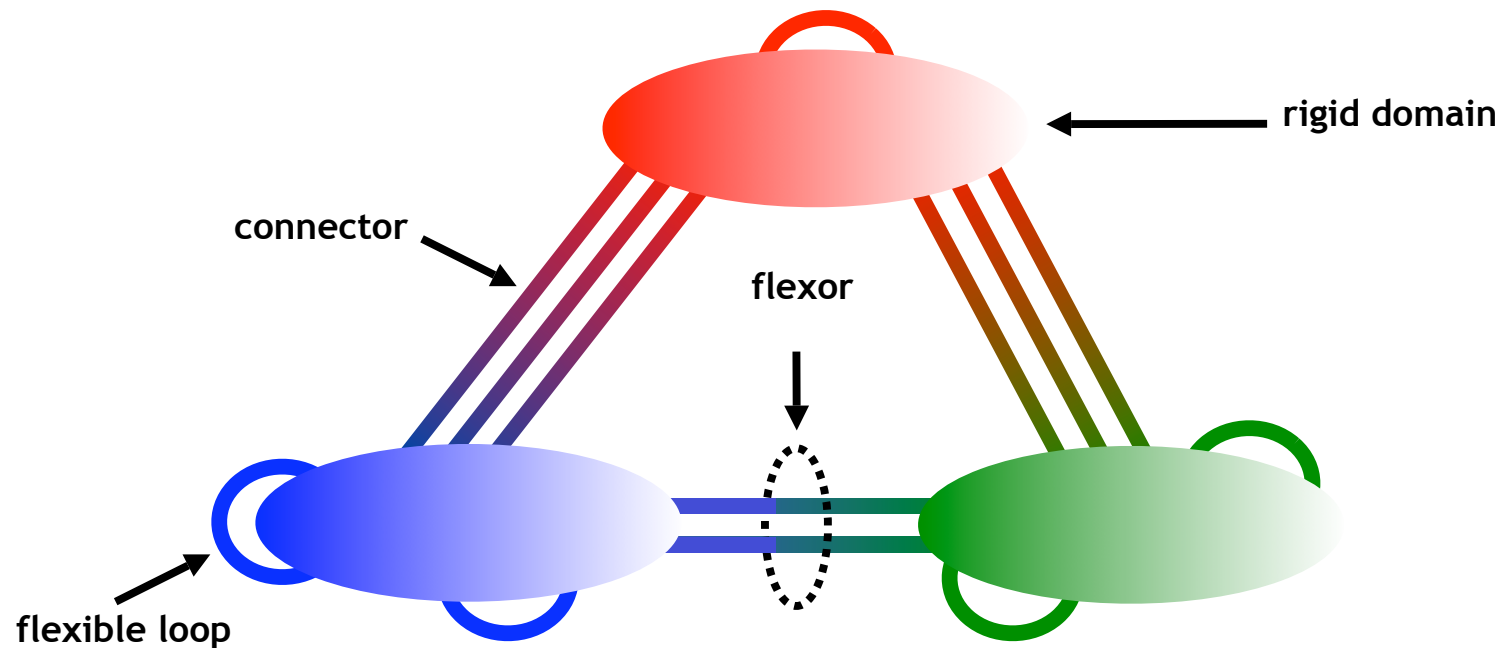
open (1CLL.PDB)



closed (2BBM.PDB)



Modeling Flexibility: The Flexible Chain Complex (FCC)



Flexible Chain Complex (FCC)

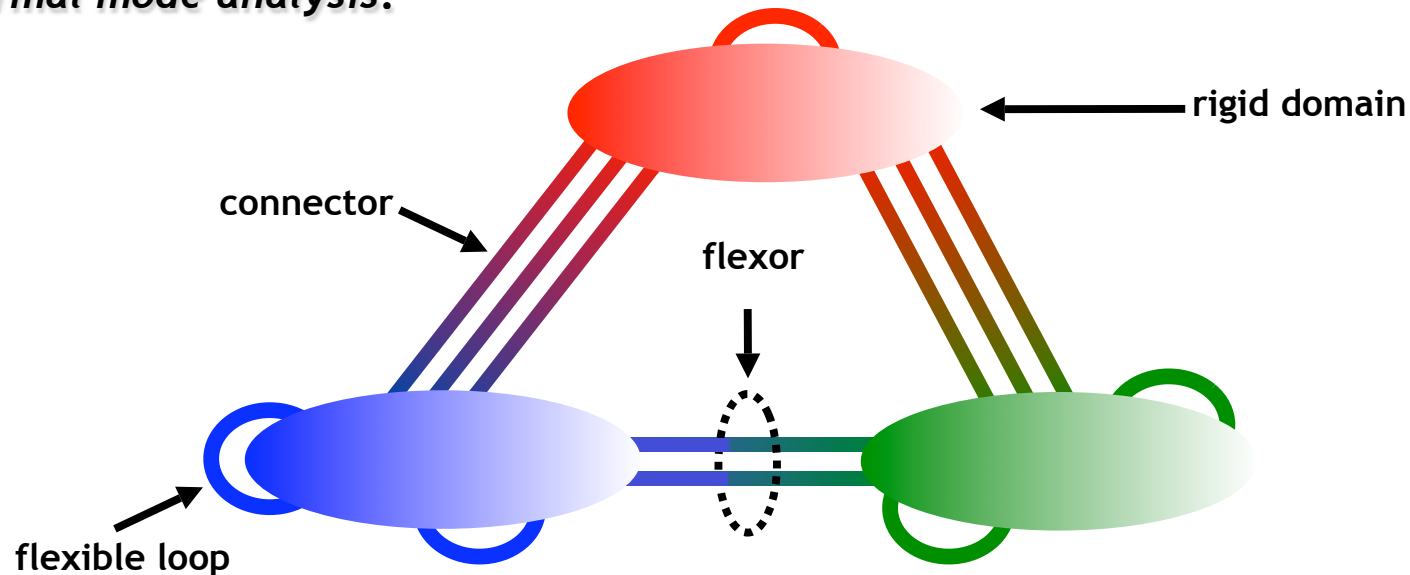
Hierarchical

By relaxing the threshold of rigidity domains can be hierarchically decomposed into subdomains



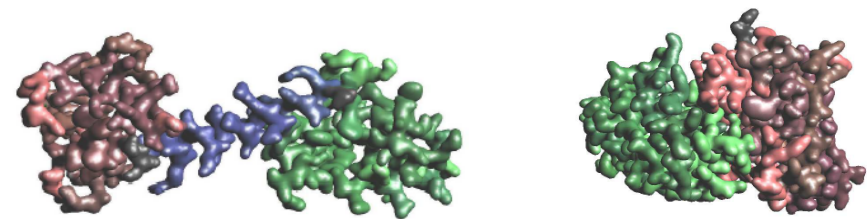
Modeling Flexibility: Identifying Domains using Normal Mode Analysis

Domains are identified by capturing domain movements using *normal mode analysis*.



Computing normal modes:

- Minimum energy conformation
- Hessian of internal energy
- Eigen-decomposition gives motion



Calmodulin : 2BBM

Immunoglobulin: 1VFB



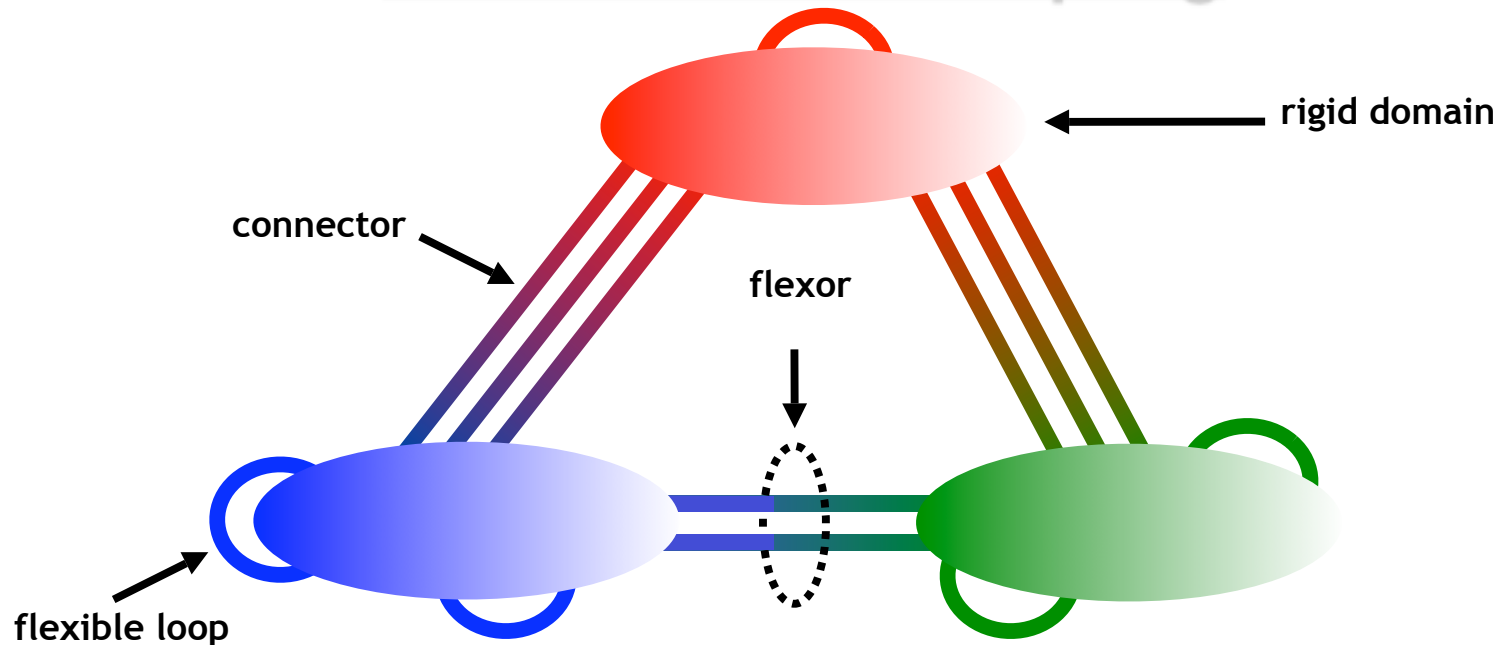
Modeling Flexibility: Motions at Flexors

We apply three types of motions at flexors:

- **Shear:** a lateral movement applied along the interface of two domains if
 - they share a large number of short connectors, and / or
 - they have a large interface area
- **Bending:** applied when the domains are connected by at least one connector
 - applied around three orthogonal axes
 - hinge point: geometric center of the shortest connector
 - primary axis: normal to the plane containing the geometric centers of the two domains and of the shortest connector
 - secondary axis: orthogonal to the primary axis and the line connecting the two domain centers
 - third axis: line through the domain centers
- **Twist:** applied when only a single connector exists between the domains
 - the torsion angles along the backbone are updated



Modeling Flexibility: Conformation Sampling

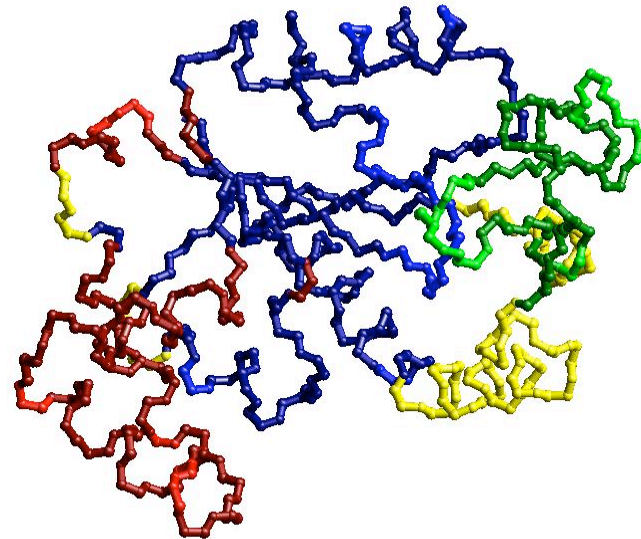
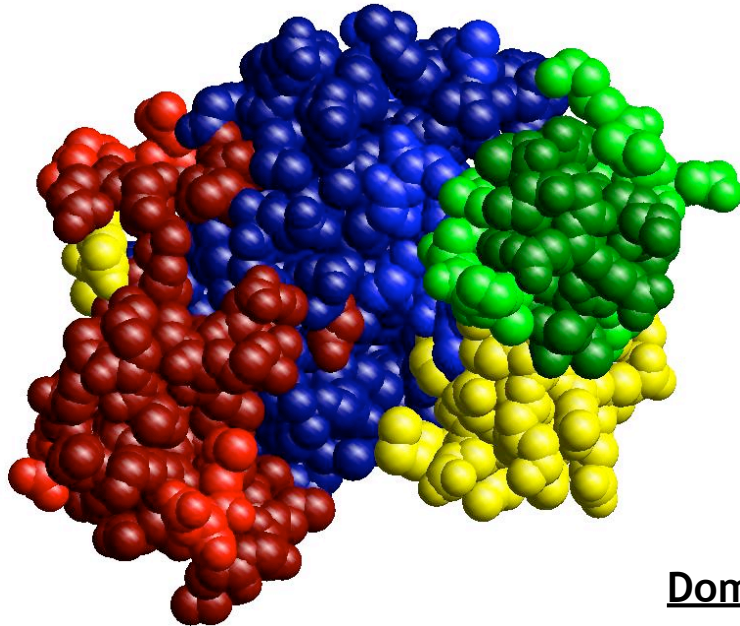


Suppose we want to generate N global conformations.

- conformations are allocated to flexors
- each flexor f is assigned a weight w_f (= sum of the domain weights)
- number of conformations allocated to f is $Nw_f / \sum w_f$
- conformations are generated by applying motions to the flexors



Modeling Flexibility: Domain Decomposition of Adenylate Kinase (4AKE.PDB)



- ❑ Domains: red / green / blue
- ❑ Flexible Loops: light domain color
- ❑ Connectors: yellow

Domains:

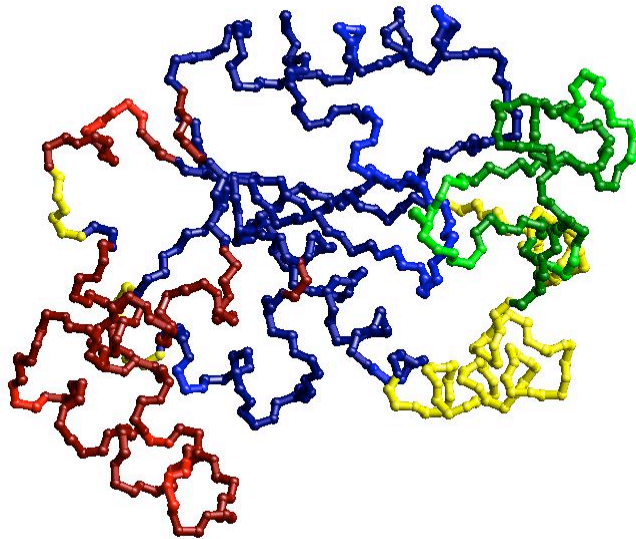
- Core: 93 residues, 12 segments, 12 loops
- AMP-binding: 52 residues, 9 segments, 4 loops
- ATP-lid: 36 residues, 4 segments, 3 loops

Connectors and Interface Area:

- Core and AMP-binding: 3, 508 Å
- Core and ATP-lid: 2, 2 Å



Modeling Flexibility: Motion Graph of Adenylate Kinase

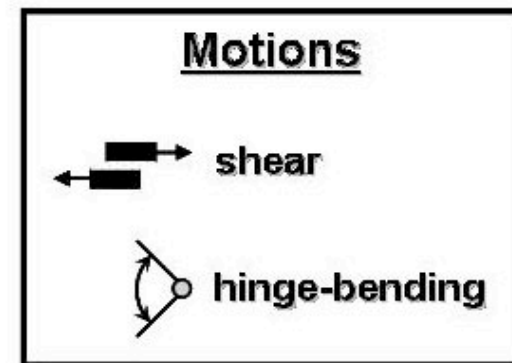
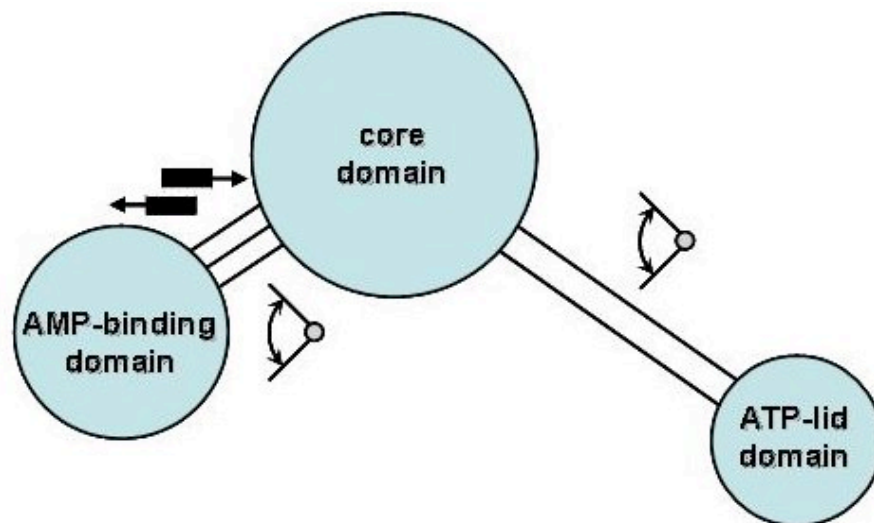


Domains:

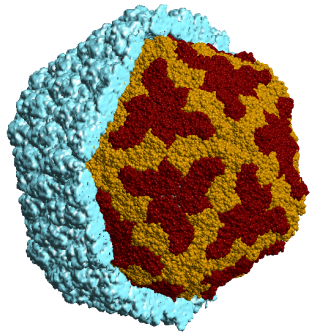
- Core: 93 residues, 12 segments, 12 loops
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Connectors and Interface Area:

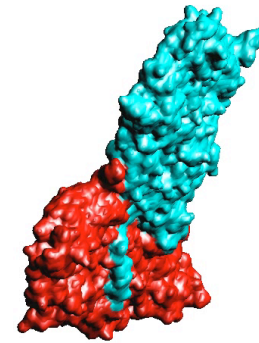
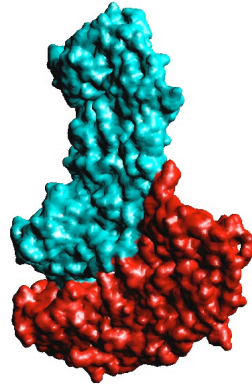
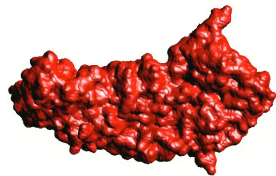
- Core and AMP-binding: 3, 508 Å
- Core and ATP-lid: 2, 2 Å



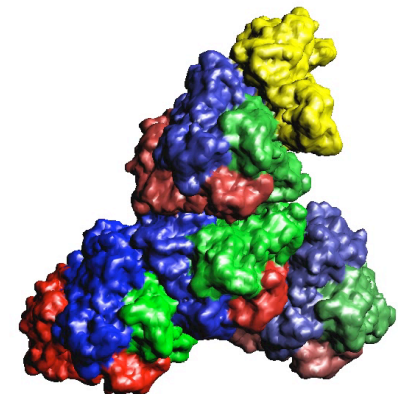
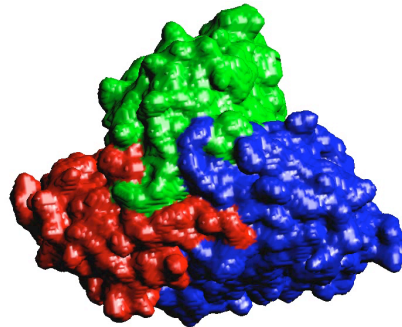
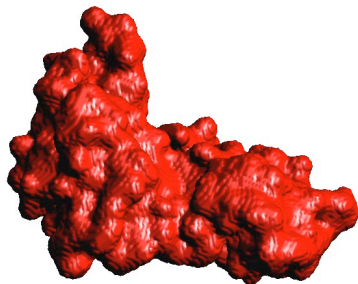
Coarse Grained Flexibility of RDV capsid Proteins



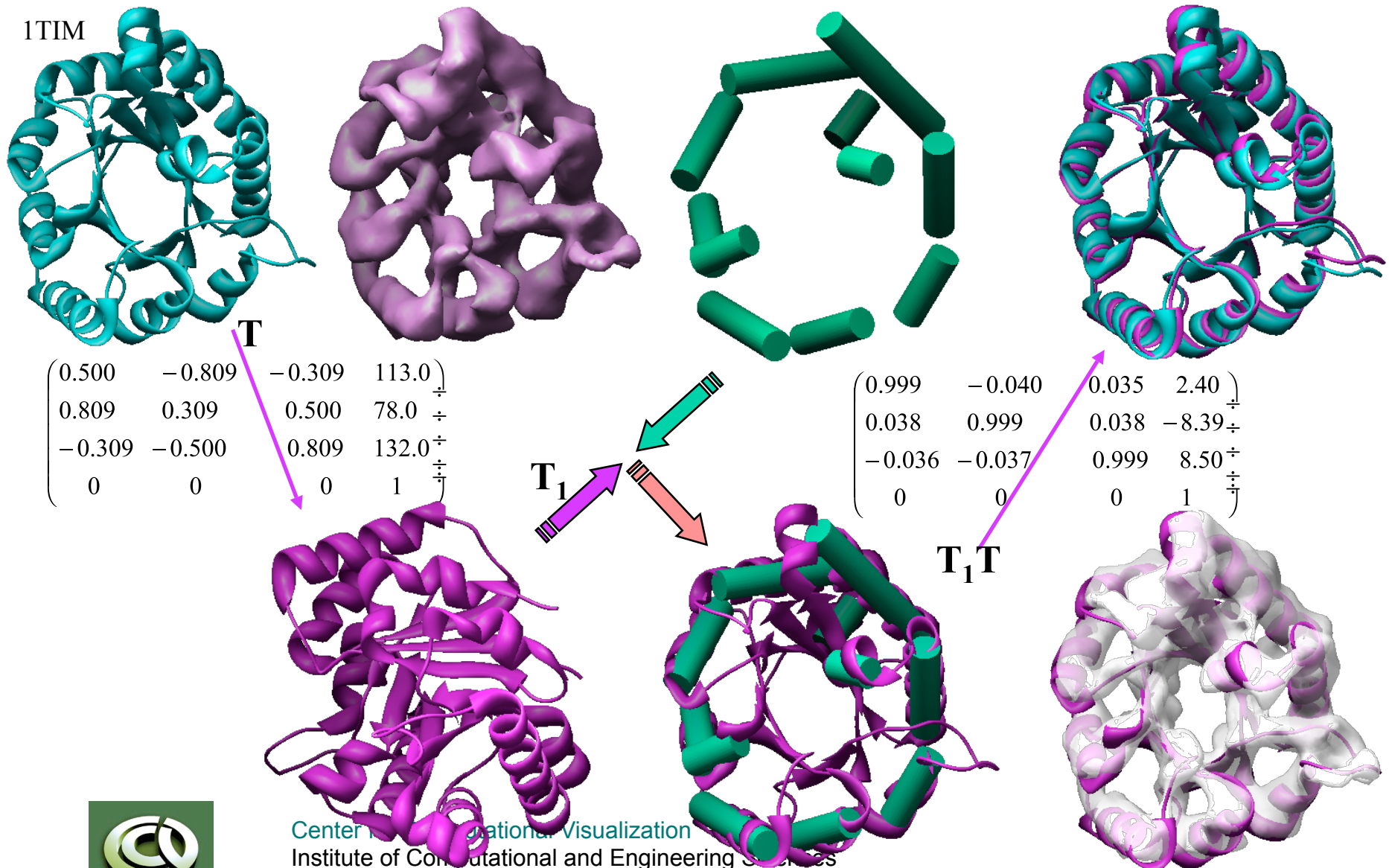
P3



P8



Low Res. Flexible Structure Fitting in 3DEM

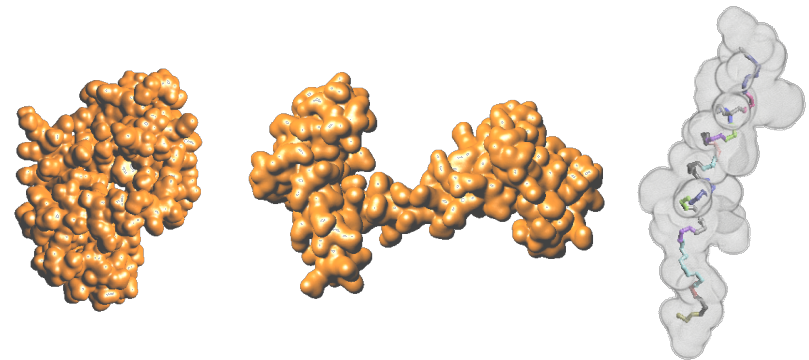


Search and Scoring for Flexible Docking

(F3Dock)

Docking score is based on

- Shape complementarity:
 - Lennard Jones potential
 - Curvature complementarity
- Electrostatic interactions
 - Long range
 - Short range
 - Hydrogen bonds
 - Salt bridges
 - Hydrophobic patches
- Desolvation free energy
- Free Energy change due to conformational changes

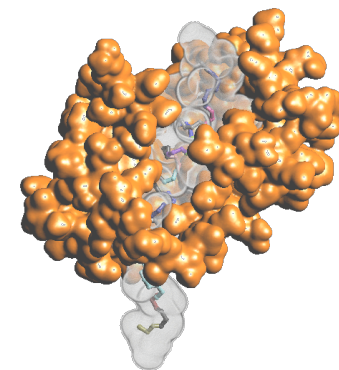


Closed

Open

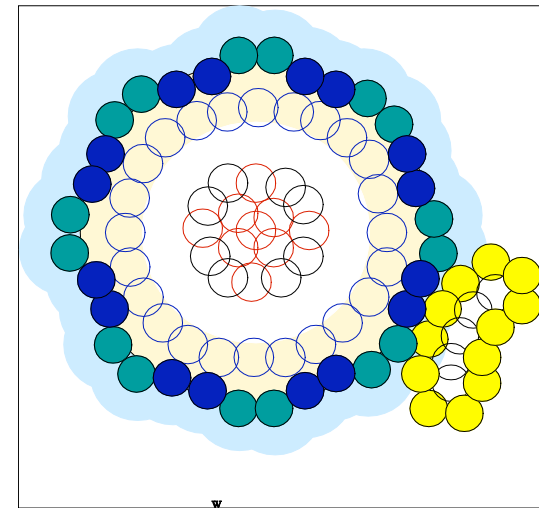
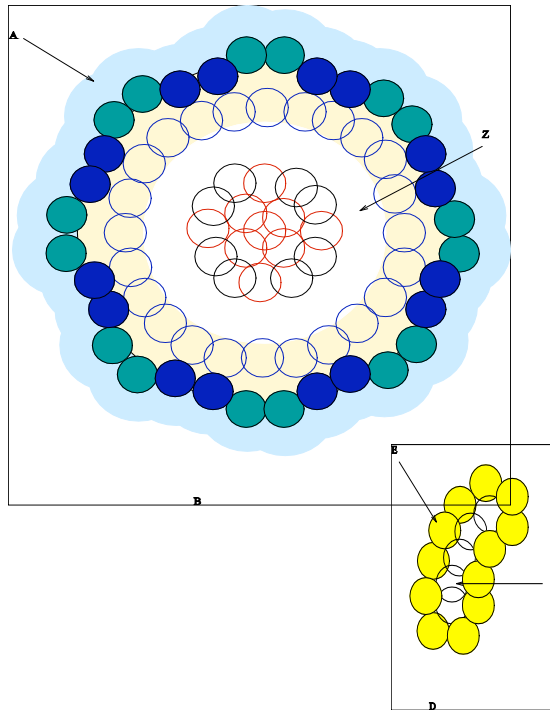
Peptide

Calmodulin
(1CLL.PDB)



Calmodulin complexed with peptide
(2BBM.PDB)

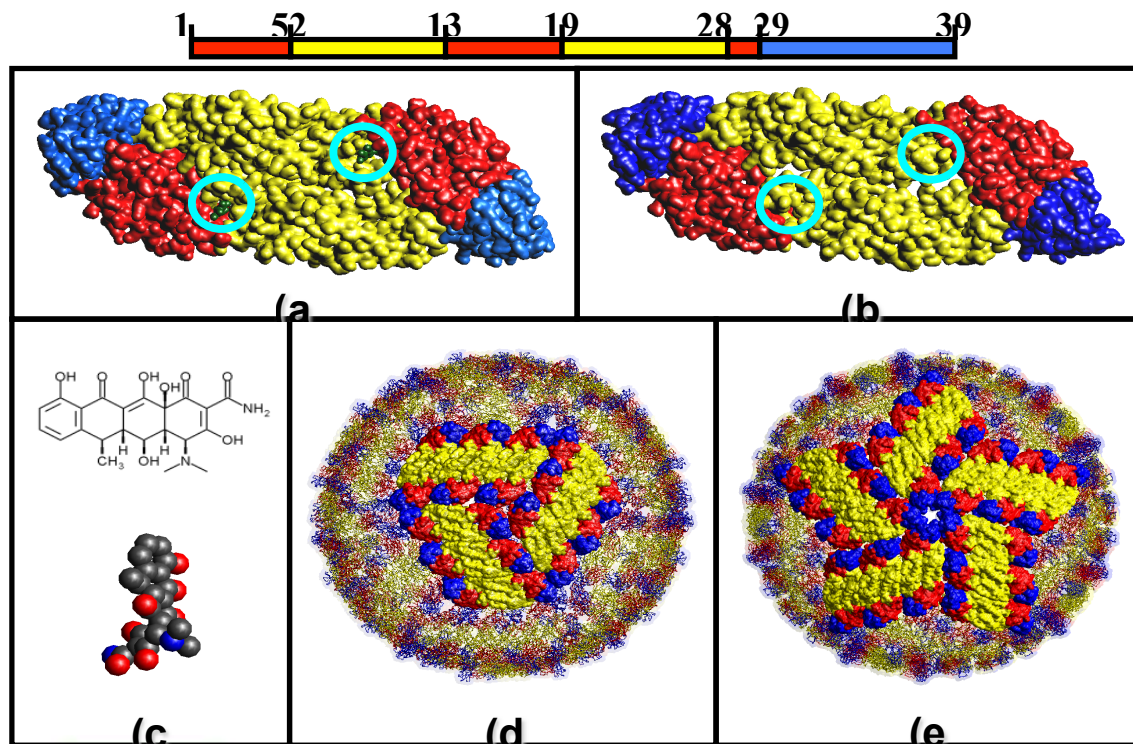
Fast Computation of Protein Binding Energetics and Forces



Binding Energy of Macromolecules

In structure based drug design, binding of a drug (ligand) to a receptor (protein/nucleic acid), usually causes the ligand to either **enhance** or **inhibit** the activity of the receptor.

Binding energy: $\Delta G_{\text{bind}} = G_{\text{protein+ligand}} - (G_{\text{protein}} + G_{\text{ligand}})$



Docking Inhibitors in a Flexible Pocket of Dengue Virus Envelope Protein

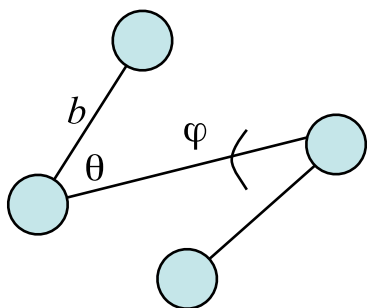


Free Energy of a Single Molecule in Solvent

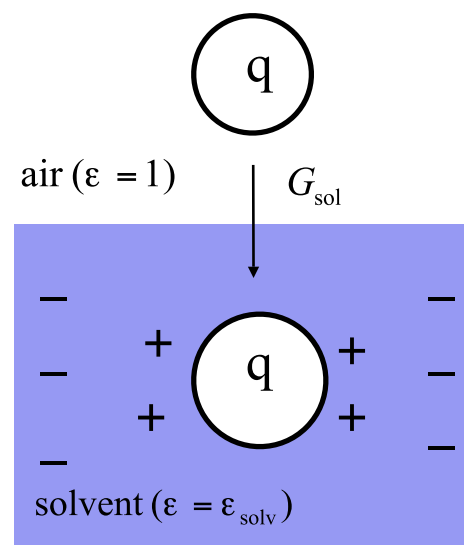
$$\text{Total free energy : } G = E_{MM} + G_{\text{sol}} - TS$$

$$E = E_b + E_\theta + E_\varphi + E_{vdw} + E_{elec}$$

└──────────┬──────────┘
bonded non-bonded



$$G_{\text{sol}} = G_{\text{cav}} + G_{\text{vdw}} + G_{\text{pol}}$$



Free Energy of a Single Molecule in Solvent

• bonded

$$\left\{ \begin{array}{l} E_b = \sum_b k_b (r_b - r_b^0)^2 \quad r, r_0 : \text{covalent bond variation} \\ E_\theta = \sum_a k_a (\theta_a - \theta_a^0)^2 \quad \theta, \theta_0 : \text{valence angle variation} \\ E_\varphi = \sum_t k_t (1 + \cos n(\varphi_t - \varphi_t^0)) \quad \varphi, \varphi_0 : \text{torsion angle variation} \end{array} \right.$$

• non-bonded

$$\left\{ \begin{array}{l} E_{vdw} = \sum_{i<j} \frac{C_{ij}}{r_{ij}^{12}} - \frac{D_{ij}}{r_{ij}^6} \quad C, D : \text{Lennard-Jones parameters} \\ \quad \quad \quad r_{ij} : \text{distance between atoms} \\ E_{elec} = \sum_{i<j} \frac{q_i q_j}{r_{ij}} \quad q : \text{atomic charge} \end{array} \right.$$

• $G_{cav} + G_{vdw} = \gamma S \quad \gamma, S : \text{surface tension and surface area}$



How to Compute G_{pol} ?

■ Poisson-Boltzmann (PB) Theory

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

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

finite difference, finite element -- APBS, DELPHI,

- ASMS Boundary Element [Bajaj, Chen 2007]

ε	dielectric constant
ϕ	electrostatic potential
ρ	solute charge density
λ	ion accessibility parameter
c_j^{∞}	ion bulk concentration
q_j	ion charge
k_B	Boltzmann's constant
T	temperature

■ Generalized Born (GB) Theory

- Born formula (Born 1920)
- generalized Born formula (Still 1990)
- methods to compute the Born radii:

$$G_{\text{pol}} = -\frac{\tau}{2} \sum_{ij} \frac{q_i q_j}{[r_{ij}^2 + R_i R_j \exp(-\frac{r_{ij}^2}{4R_i R_j})]^{\frac{1}{2}}}$$

- pairwise summation : fast but not easy for force calculation
- ASMS, nFFT, and higher order quadrature [Bajaj, Zhao 2006]

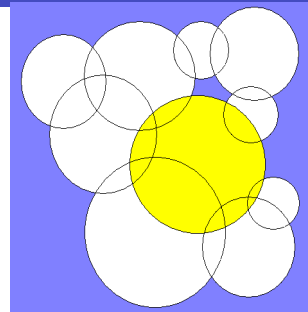
$$R_i^{-1} = \frac{1}{4\pi} \int_{\Gamma} \frac{(\mathbf{r} - \mathbf{x}_i) \cdot \mathbf{n}(\mathbf{r})}{|\mathbf{r} - \mathbf{x}_i|^4} dS$$



Generalized Born Energy

Generalized Born method

$$G_{\text{pol}} = -\frac{\tau}{2} \sum_{i,j} \frac{q_i q_j}{\left[r_{ij}^2 + R_i R_j \exp\left(-\frac{r_{ij}^2}{4R_i R_j}\right) \right]^{\frac{1}{2}}} \quad (1)$$



where q is the atomic charge, r is the distance between atom pairs, and R is the atom's **effective Born radius** which reflects how deeply the atom is buried in a molecule:

$$R_i^{-1} = \frac{1}{4\pi} \int_{\Gamma} \frac{(\mathbf{r} - \mathbf{x}_i) \cdot \mathbf{n}(\mathbf{r})}{|\mathbf{r} - \mathbf{x}_i|^4} dS \quad (2)$$

where Γ is the molecular surface of the solute, \mathbf{x} is the atomic center, and \mathbf{n} is the unit surface normal.

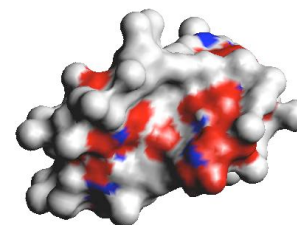
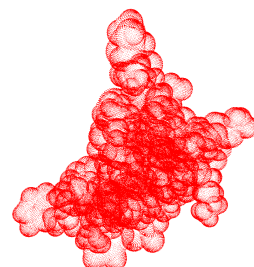
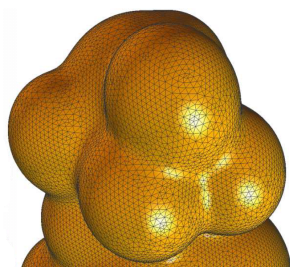
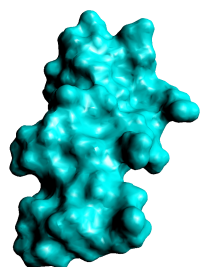


Fast Computation of Born Radii R_i

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

Algorithm:

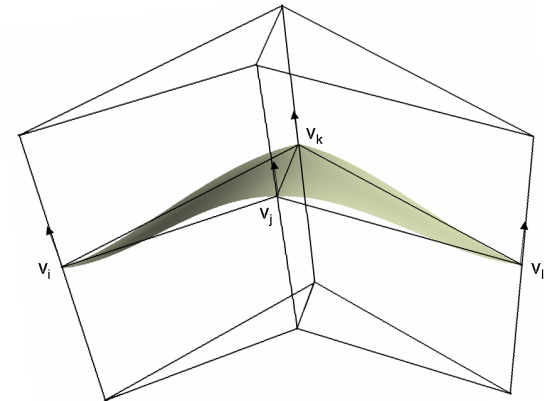
1. Generate a smooth A-spline model for the molecular surface Γ .
2. Cubature: choose w_k and \mathbf{r}_k properly so that higher order of accuracy can be obtained for small N .
3. Fast summation using non-uniform FFT to evaluate R_i , $i = 1, \dots, M$.



•[Zhao, Xu, Bajaj, ACM SPM 2006]



Algebraic Splines (A-Splines)



$S := \{\mathbf{p}(b_1, b_2, b_3, \lambda) : F(b_1, b_2, b_3, \lambda) = 0, \mathbf{p} \in D_{ijk}\}$ where

- $D_{ijk} = \{\mathbf{p}(b_1, b_2, b_3, \lambda) : \mathbf{p} = b_1 \mathbf{v}_i(\lambda) + b_2 \mathbf{v}_j(\lambda) + b_3 \mathbf{v}_k(\lambda), \lambda \in I_{ijk}\}$

- I_{ijk} is an interval containing 0

- $$F(b_1, b_2, b_3, \lambda) = \sum_{i+j+k=n} b_{ijk}(\lambda) B_{ijk}^n(b_1, b_2, b_3)$$

- $$B_{ijk}^n(b_1, b_2, b_3) = \frac{n!}{i!j!k!} b_1^i b_2^j b_3^k$$

- $n > 2$ so that S is smooth. We consider the case $n=3$. $b_{ijk}(\lambda)$ are determined so that C^1 continuity is obtained across the patch boundaries.

• [Bajaj, Xu CAGD 2002]



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Fast Summation via Error bounded Non-uniform Fast Fourier Transforms

To evaluate forms like $I(x_i) = \sum_{k=1}^N c_k K(x_i - r_k), \quad i = 1, \dots, M, \quad K \in C^p(-\frac{1}{2}, \frac{1}{2})$

$$K(x_i - r_k) \approx \sum_{\omega=-n/2}^{n/2} b_\omega e^{2\pi i \omega (x_i - r_k)} := K_F$$

$$I(x_i) \approx \sum_{k=1}^N c_k K_F(x_i - r_k) = \sum_{k=1}^N c_k \sum_{\omega=-n/2}^{n/2-1} b_\omega e^{2\pi i \omega (x_i - r_k)} = \sum_{\omega=-n/2}^{n/2-1} b_\omega a_\omega e^{2\pi i \omega x_i} \quad \text{NFFT}$$

where $a_\omega = \sum_{k=1}^N c_k e^{-2\pi i \omega r_k} \quad \text{NFFT}^\top$

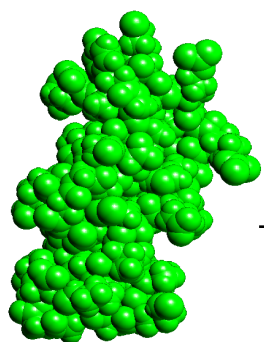
Time complexity:

$$a_\omega : O(mn + N \log N) \quad b_\omega : O(n \log n) \quad I : O(mM + n \log n) \quad m < n/2$$

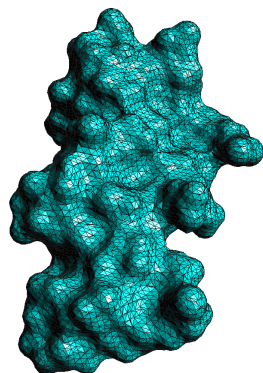
overall: $O(mM + mn + N \log N + n \log n)$ (trivial method: $O(MN)$)



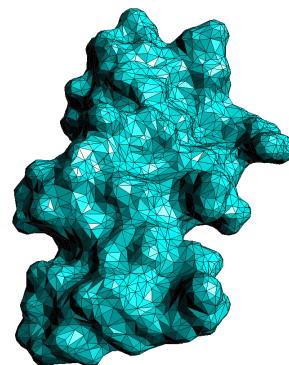
ASMS: Algebraic Spline Molecular Surfaces



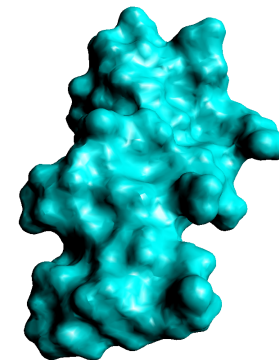
atomic/quasi-atomic structure



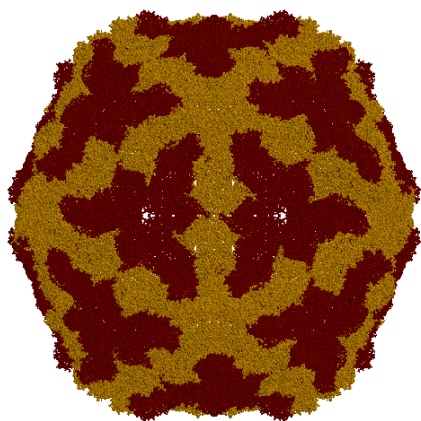
molecular surface triangulation



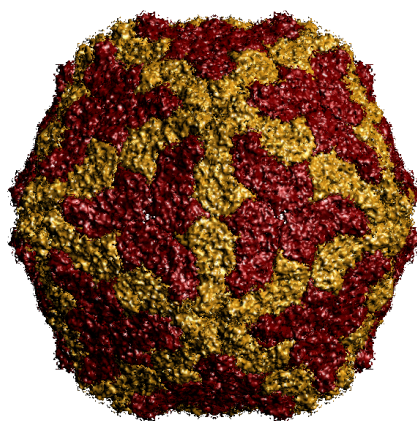
coarse mesh



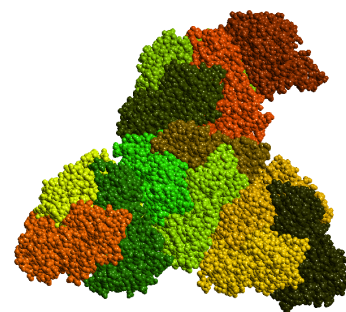
A-Spline



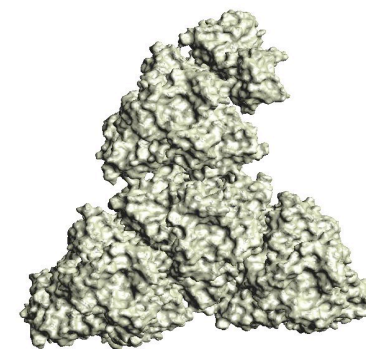
atomic structure



A-Spline



atomic structure
asymmetric subunit



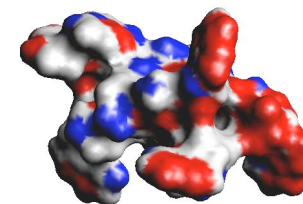
A-Spline



Force Calculation

The electrostatic force acting on atom α which is part of the forces driving Molecular dynamics is

$$\mathbf{F}_\alpha^{\text{elec}} = -\frac{\partial G_{\text{pol}}}{\partial \mathbf{x}_\alpha}$$



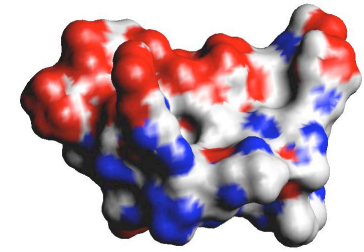
$$\frac{\partial G_{\text{pol}}}{\partial \mathbf{x}_\alpha} = \sum_i \sum_{j \neq i} \frac{\partial G_{\text{pol}}}{\partial r_{ij}} \frac{\partial r_{ij}}{\partial \mathbf{x}_\alpha} + \sum_i \frac{\partial G_{\text{pol}}}{\partial R_i} \frac{\partial R_i}{\partial \mathbf{x}_\alpha}$$

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

The integration domain depends on \mathbf{x}_α .



Force Calculation



$$-\frac{1}{R_i^2} \frac{\partial R_i}{\partial \mathbf{x}_\alpha} = \frac{1}{4\pi} \int_{\mathbb{R}^3} \frac{\partial}{\partial \mathbf{x}_\alpha} \frac{\bar{\rho}(\mathbf{r}, \{\mathbf{x}_j\})}{|\mathbf{r} - \mathbf{x}_i|^4} dV$$

$$\frac{\partial R_i}{\partial \mathbf{x}_\alpha} = -\frac{R_i^2}{4\pi} \left(\int_{\mathbb{R}^3} \frac{\frac{\partial}{\partial \mathbf{x}_\alpha} \bar{\rho}(\mathbf{r}, \{\mathbf{x}_j\})}{|\mathbf{r} - \mathbf{x}_i|^4} dV + \int_{\text{ex}} \frac{\partial}{\partial \mathbf{x}_\alpha} \frac{1}{|\mathbf{r} - \mathbf{x}_i|^4} dV \right)$$

$$\parallel$$

$$-4\mathbf{x}_i^3 \int_{\Gamma} \frac{(\mathbf{r} - \mathbf{x}_i) \cdot \mathbf{n}(\mathbf{r})}{|\mathbf{r} - \mathbf{x}_i|^6} dS$$

$$\frac{\partial}{\partial \mathbf{x}_\alpha} \bar{\rho} = -\frac{\partial}{\partial \mathbf{x}_\alpha} \tilde{\rho} = \frac{\partial \rho_\alpha}{\partial \mathbf{x}_\alpha} \left(1 - \sum_j \rho_j + \sum_{j<k} \rho_j \rho_k - \sum_{j<k<l} \rho_j \rho_k \rho_l \right)$$

Molecular surface

Since $\frac{\partial \rho_\alpha}{\partial \mathbf{x}_\alpha} \neq 0$ only if $a_\alpha - w < |\mathbf{r} - \mathbf{x}_\alpha| < a_\alpha + w$

Molecular skin

$$\int_{\mathbb{R}^3} \frac{\frac{\partial}{\partial \mathbf{x}_\alpha} \bar{\rho}(\mathbf{r}, \{\mathbf{x}_j\})}{|\mathbf{r} - \mathbf{x}_i|^4} dV = \int_{|\mathbf{r}-\mathbf{x}_\alpha|=a_\alpha-w}^{|\mathbf{r}-\mathbf{x}_\alpha|=a_\alpha+w} \frac{\frac{\partial \rho_\alpha}{\partial \mathbf{x}_\alpha} \left(1 - \sum_j \rho_j + \sum_{j<k} \rho_j \rho_k - \sum_{j<k<l} \rho_j \rho_k \rho_l \right)}{|\mathbf{r} - \mathbf{x}_i|^4} dV$$



Molecular Solvation Forces

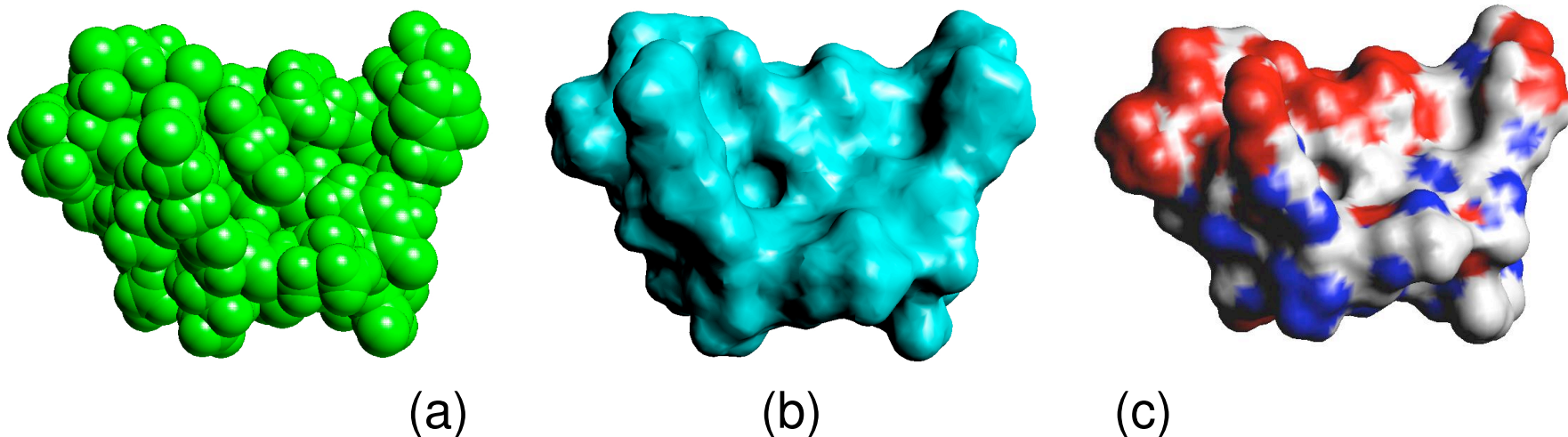


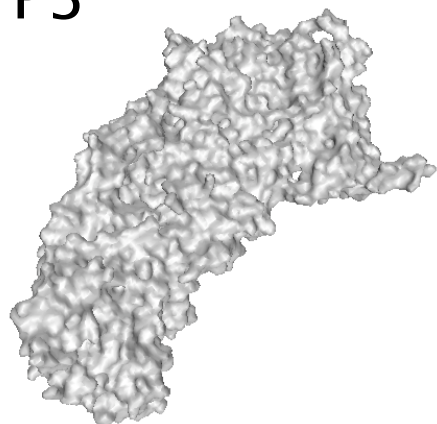
Figure: (a) The atomic model of protein 1PPE (436 atoms); (b) The piecewise algebraic surface of 1PPE. $G_{pol} = -825.33$ kcal/mol.

; (c) The solvation forces of protein 1PPE shown as function on surface

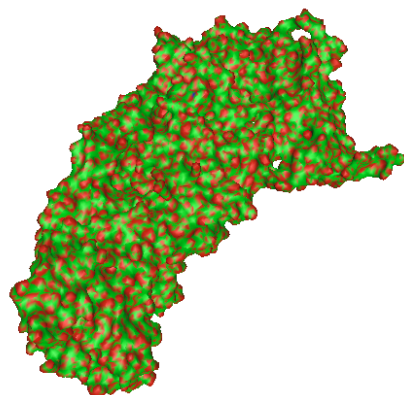


P3 and P8

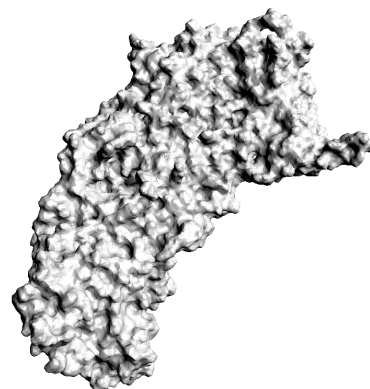
P3



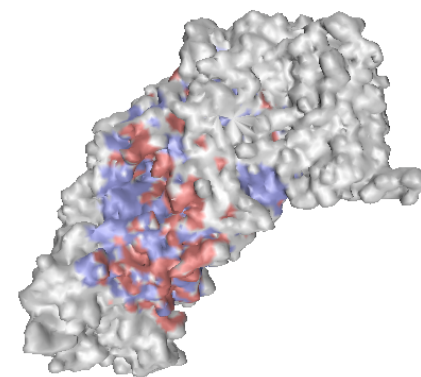
Molecular Surface



Gaussian Curvature

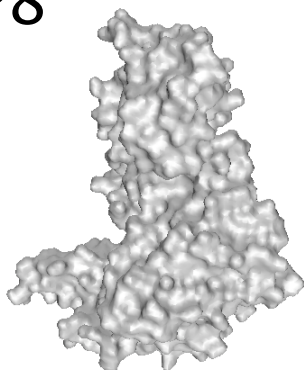


GB Energy:
-23892.33 kcal/Mol

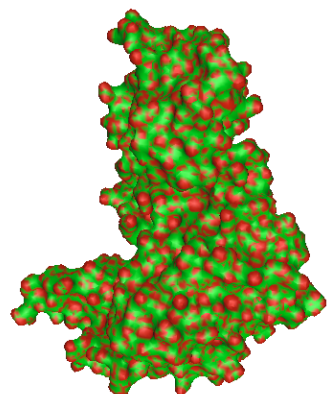


PB energy:
-17758.9 kJ/Mol

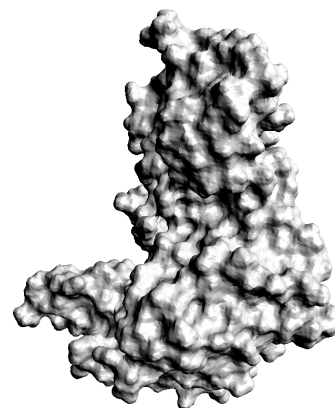
P8



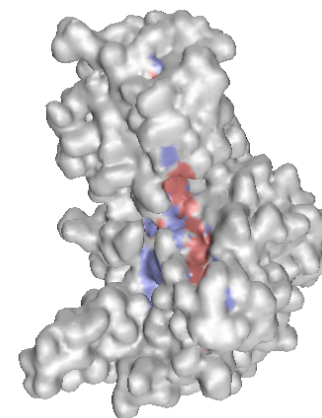
Molecular Surface



Gaussian Curvature



GB energy:
-8413.185 kcal/Mol



PB energy:
-7163.08 kJ/Mol



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