# Geometric and Signal 3D Processing (and some Visualization)

### Chandrajit Bajaj

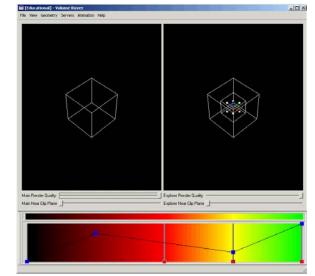


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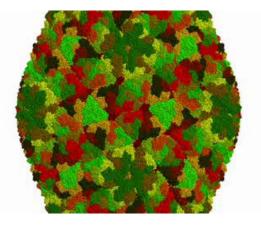
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## Algorithms & Tools

- Structure elucidation: filtering, contrast enhancement, segmentation, skeletonization, subunit identification
- > Structure Modeling: finite element meshing, spline representations(Aspline,RBF representations) for structural fitting & complementary docking
- Visualization: multi-dimensional transfer functions, surface and volume texture rendering, wavelet compression, hierarchical representations, cluster based parallelism



VolRover







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## The CVC Team and Collaborators

### Personnel

- Inderjit Dhillon (Assoc. Director)
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- Katherine Clarridge (MBE,M.S.)
- KL Chandrasekhar (ME, Ph.D.)
- Tamal Dey (OSU) \*\*
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- Insung Ihm (SNU, S. Korea)\*\*
- Sangmin Park (CS, Ph.D.)
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- Bong-Soo Sohn (Stanford U)\*\*
- Jason Sun (Res. Scientist)
- John Wiggins (Res. Scientist)
- Vinay Siddahanavalli (CS, Ph.D.)
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- Zeyun Yu (CS, Ph.D)
- Xiaoyu Zhang (CSU)\*\*
- Jessica Zhang (PostDoc, ICES)
- Wenqi Zhao (ICES, Ph.D.)

- Senior Collaborators
  - Manfred Auer (LBL)
  - Nathan Baker (Wash. U.)
  - Helen Baker, Cathy Lawson (Rutgers U)
  - Tim Baker (UCSD)
  - Tom Bartol (SALK)
  - Luis Caffarelli (ICES)
  - Wah Chiu, Matt. Baker (Baylor)
  - Leszek Demkowicz (ICES)
  - Gregory Gladish, J. Hazle (MD Anderson)
  - Tom Hughes (ICES)
  - Andy McCammon (USCD)
  - Tinsley Oden (ICES)
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  - Peter Rossky (ICES)
  - Andre Sali (UCSF)
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  - NIH: P20(planning), R01
  - NSF: ITR , DDDAS
  - Whitaker Foundation

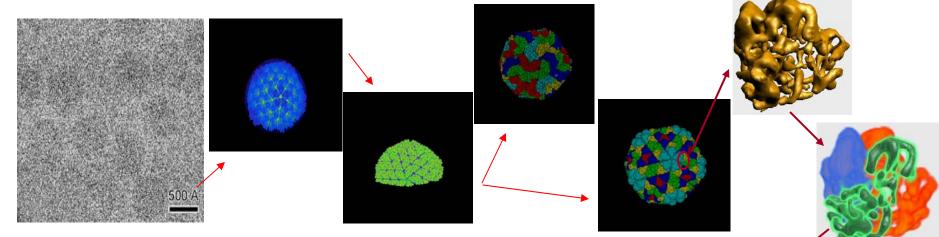


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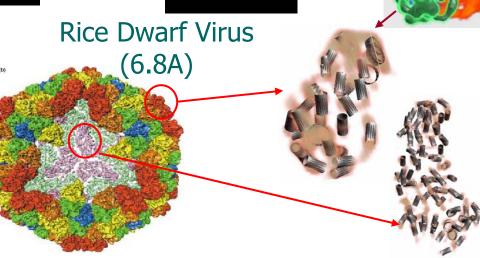
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### Sub-nanometer Structure Elucidation from 3D Cryo-EM



Cryo-EM  $\rightarrow$  FFT based 3D Reconstruction  $\rightarrow$  Anisotropic and Vector Diffusion Filtering  $\rightarrow$ Structure Segmentation  $\rightarrow$  Quasi-Atomic Modeling  $\rightarrow$  Visualization



\*\*Sponsored by NSF-ITR, NIH

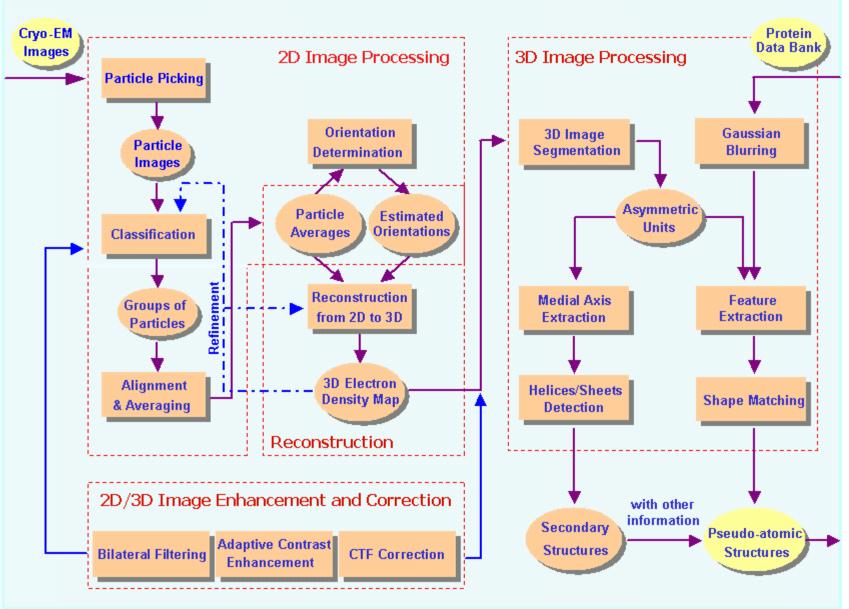


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### A Structure Determination Pipeline for single particle cryo-EM



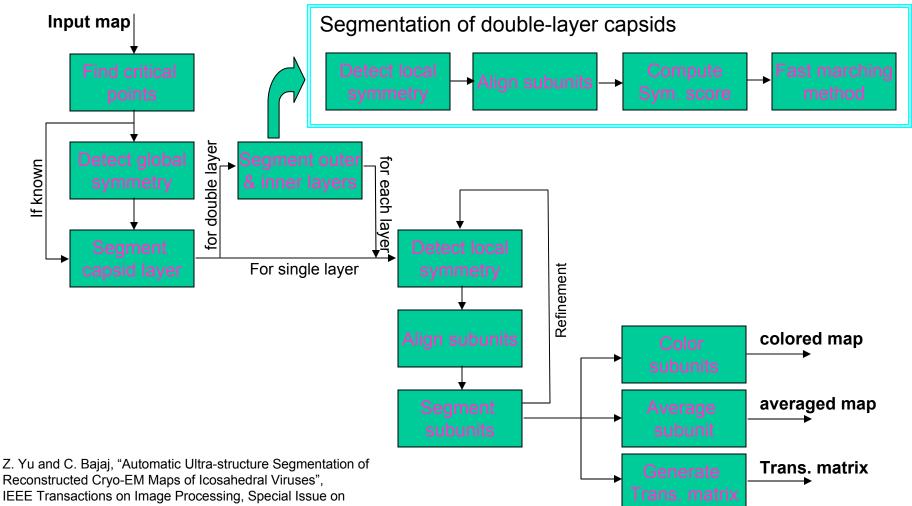


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## Structure Elucidation for Icosahedral Viruses



Cellular/Molecular Imaging, 14(9), pp. 1324-1337, 2005.



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# Structure Elucidation 1(A)

- Adaptive contrast enhancement
- 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  $\sigma_d$  and  $\sigma_r$  are parameters and f(.) is the image intensity value.
- Anisotropic diffusion filtering

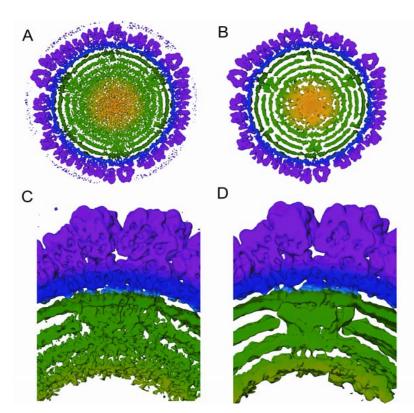
$$\partial_t \phi - \operatorname{div}(a(|\nabla \phi_\sigma|) \nabla \phi) = 0$$

where **a** stands for the diffusion tensor determined by local curvature estimation.

• Anisotropic gradient vector diffusion

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

- Z. Yu & C. Bajaj, Proc. Int'l Conf. Image Processing, 2002. pp. 1001-1004.
- Z. Yu & C. Bajaj, Proc. Int'l Conf. Computer Vision and Pattern Recognition, 2004. 415-420.



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

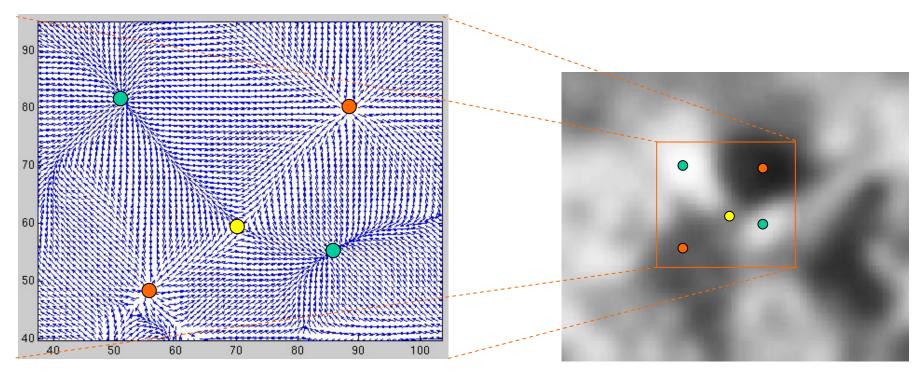


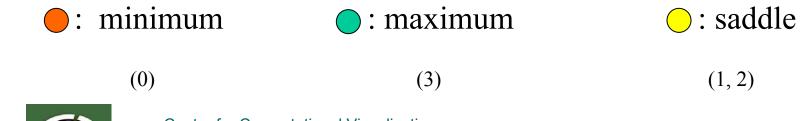
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# Compute Critical Points Using AGVD







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# Anisotropic Gradient Vector Diffusion (AGVD)

Isotropic Diffusion (Xu et al., 1998)

$$\frac{\partial u}{\partial t} = \mu \nabla^2 u - (u - f_x)(f_x^2 + f_y^2)$$
$$\frac{\partial v}{\partial t} = \mu \nabla^2 v - (v - f_y)(f_x^2 + f_y^2)$$

Where:

(u(t), v(t)) stands for the evolving vector field;  $\mu$  is a constant;

f is the original image to be diffused;

 $(f_x, f_y) = (u(0), v(0)).$ 

Anisotropic Diffusion (Yu & Bajaj ICPR'02)  $\begin{cases}
\frac{\partial u}{\partial t} = \mu \nabla (g(\alpha) \cdot \nabla u) - (u - f_x)(f_x^2 + f_y^2) \\
\frac{\partial v}{\partial t} = \mu \nabla (g(\alpha) \cdot \nabla v) - (v - f_y)(f_x^2 + f_y^2)
\end{cases}$ 

#### Where

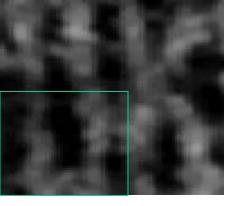
(u(t), v(t)) stands for vector field;  $\mu$  is a constant;  $(f_x, f_y) = (u(0), v(0))$ . *f* is the original image to be diffused; g(.) is the angle between two vectors



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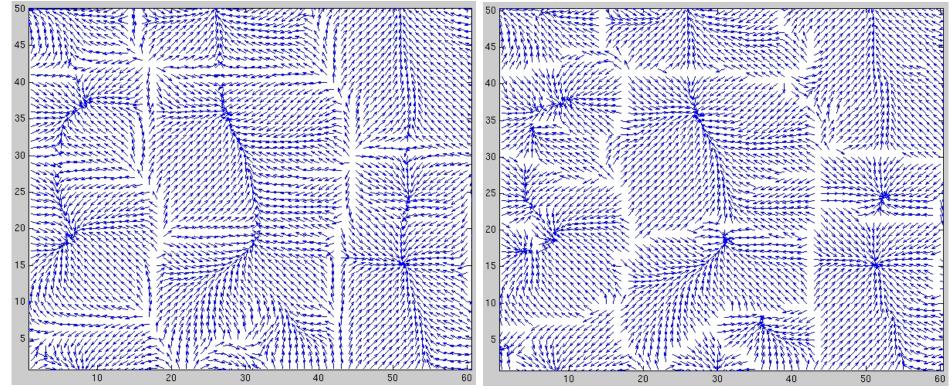
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# GVD v.s. AGVD

### Isotropic diffusion

### Anisotropic diffusion





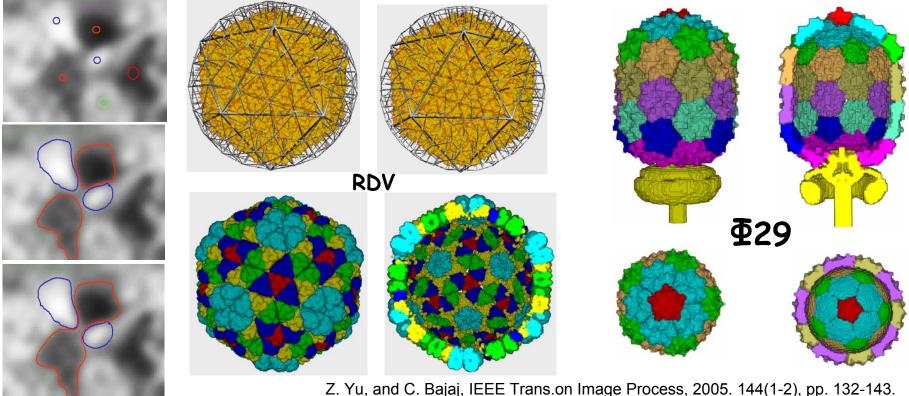
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# Structure Elucidation 1(B)

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

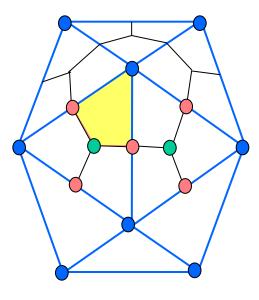


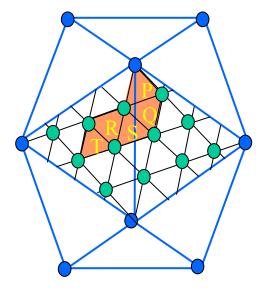


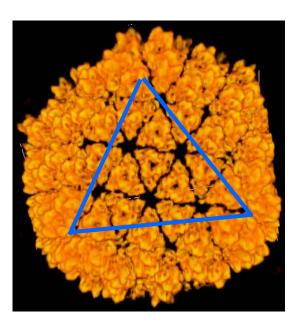
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# **Global and Local Symmetry**

• Automatic structure unit identification in a 3D Map







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

Example: RDV



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### **Symmetry Detection: Correlation Search**

$$C(\theta, \varphi) = \sum_{\vec{r} \in V} f(\vec{r}) f(R_{(\theta, \varphi, 2\pi/5)} \cdot \vec{r})$$

- Algorithm: *detect 5-fold rotation symmetry* 
  - Compute the scoring function
    - For every angular bin B<sub>j</sub>, compute θ<sub>j</sub>,φ<sub>j</sub> { For every critical point C<sub>i</sub> {

$$\vec{r}_{k}(C_{i}, B_{j}) = R_{(\theta_{j}, \phi_{j}, 2k\pi/5)} \cdot C_{i}, \quad k = 0, 1, 2, 3, 4$$
$$Dev(C_{i}, B_{j}) = \frac{1}{5} \sum_{k=0}^{4} (f(\vec{r}_{k}) - \bar{f}) \}$$

$$SF(B_j) = \frac{1}{p} \sum_{i=0}^{p} Dev(C_i, B_j)$$

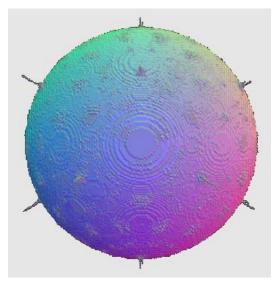
- Locate the symmetry axes
  - The 12 peaks
- Refine the symmetry axes
  - In order to locate a perfect icosahedron

(rotate the axes by  $0^0$ , 63.43<sup>0</sup>, 116.57<sup>0</sup>, 180<sup>0</sup>)

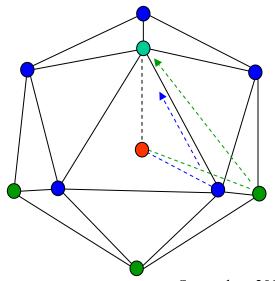


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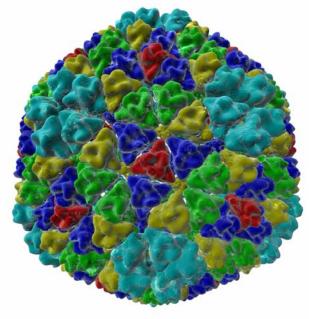
Inverted and normalized SF(Bj)



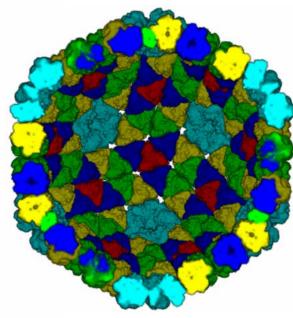
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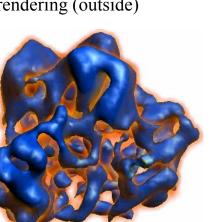
### Structure Elucidation Results: RDV (Bakeoff)

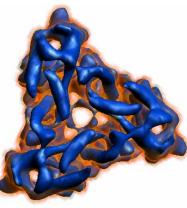


surface rendering (outside)

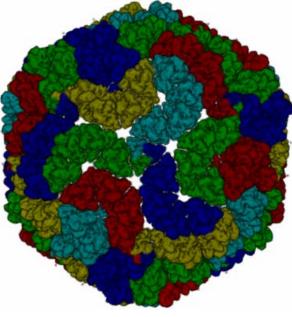


volume rendering (inside)

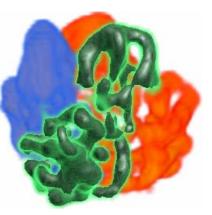




averaged trimer (bottom)



volume rendering (asymmetric unit)



segmented monomers

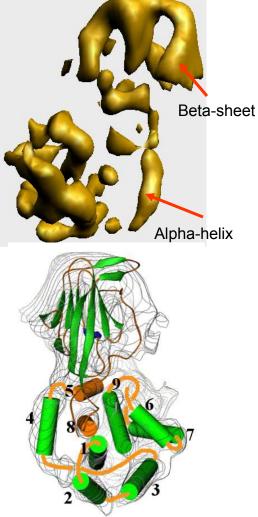
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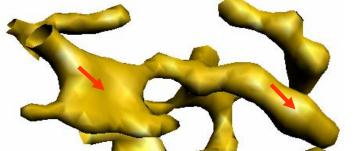
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averaged trimer (side) Center for Computational Visualization Institute of Computational and Engineering Sciences **Department of Computer Sciences** 

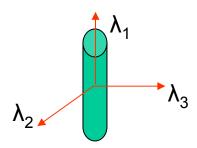
### Structure Elucidation 1(C): Secondary Structure Identification

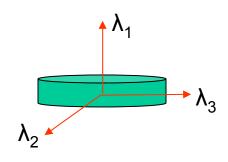




 $G_{\sigma} * \begin{pmatrix} I_x^2 & I_x I_y & I_x I_z \\ I_x I_y & I_y^2 & I_y I_z \\ I_x I_z & I_y I_z & I_z^2 \end{pmatrix}$ 

The eigenvectors of the local structure tensor give the principal directions of the local features:





Line structure (alpha-helix)

 $\lambda_2 \approx \lambda_3 >> \lambda_1 \approx 0$ 

plane structure (beta-sheet)

 $\lambda_1 >> \lambda_2 \approx \lambda_3 \approx 0$ 

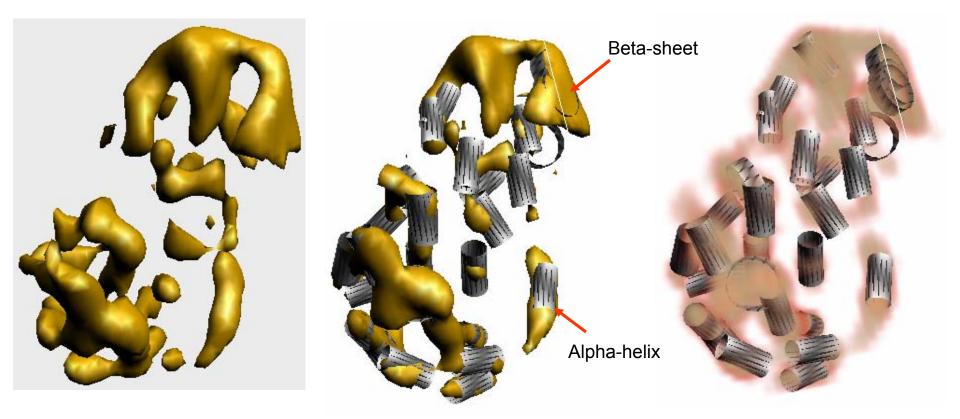


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## Monomeric Unit of Outer Capsid of RDV

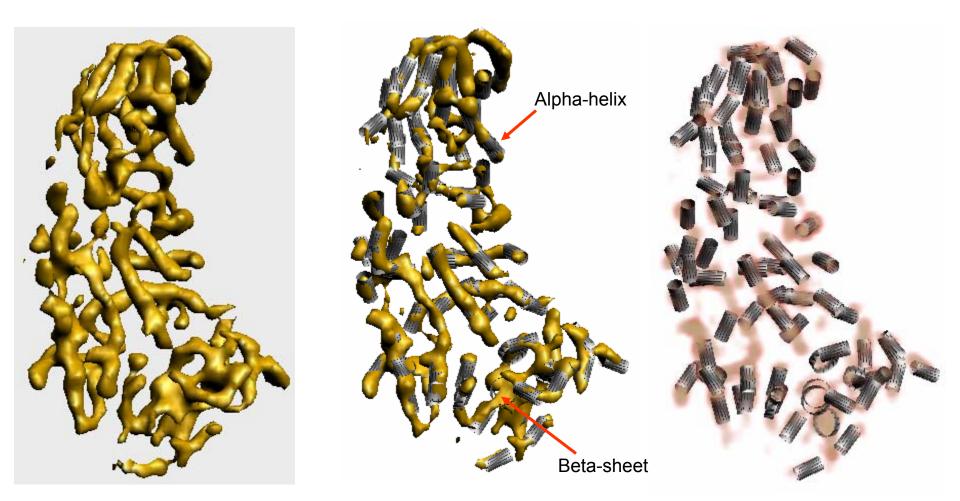




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### Monomeric Unit of Inner Capsid of RDV

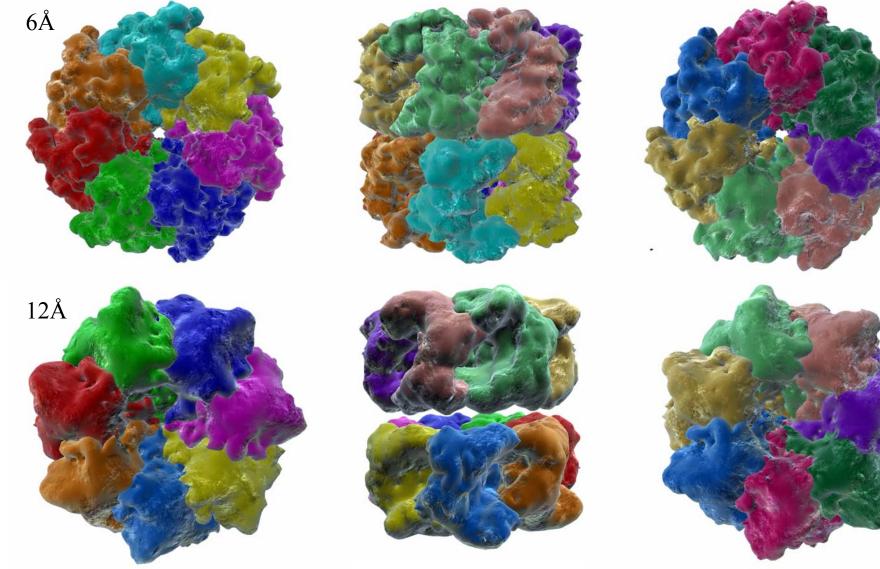




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### Structure Elucidation Results: GroEL (Bakeoff)

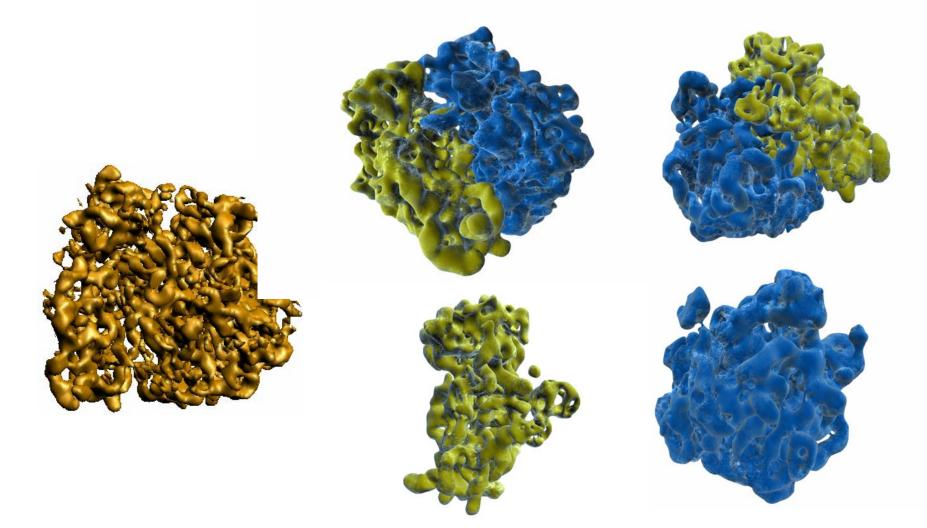




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Data courtesy: Dr. Wah Chiu University of Texas at Austin

## Segmentation Results: Ribosome (Bakeoff)



70S ribosome from E. coli complex. 70S-tRNAfMet-MF-tRNAPhe. Data courtesy: EBI & J.Frank

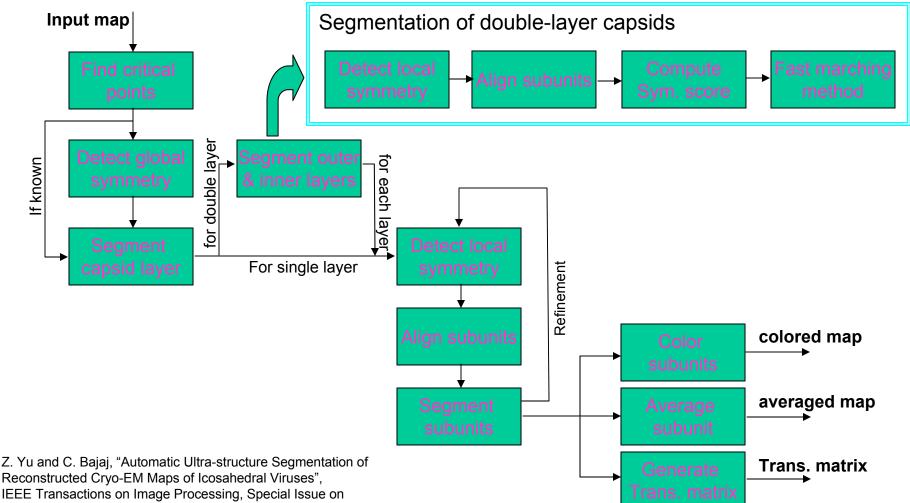


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### Structure Elucidation for Symmetric Capsid Viruses



Cellular/Molecular Imaging, 14(9), pp. 1324-1337, 2005.

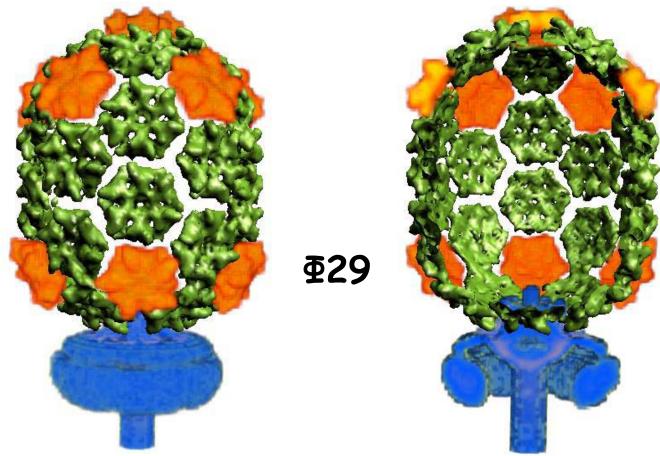


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### Subunit alignment (1): averaging



The above two pictures (left: outer; right: inner) show the averaged capsid layer, calculated from one 5-fold subunit (orange) and one 6-fold subunit (green). The tail structure (blue) is augmented after the averaging.



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#### Data courtesy: Tim Baker

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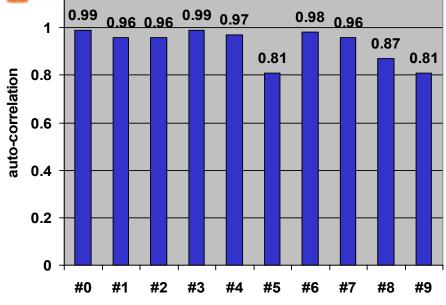


• Cross-correlation

	#0	#1	#2	#3
#0	1	0.95	0.95	0.34
#1	0.95	1	0.96	0.31
#2	0.95	0.96	1	0.31
#3	0.35	0.31	0.32	1

ſ		#4	#5	#6	#7	#8	#9
		<i></i>					
	#4	1	0.79	0.95	0.94	0.87	0.88
	#5	0.79	1	0.79	0.78	0.77	0.79
	#6	0.95	0.79	1	0.96	0.88	0.88
	#7	0.94	0.78	0.96	1	0.89	0.88
	#8	0.87	0.77	0.88	0.89	1	0.94
	#9	0.88	0.79	0.88	0.88	0.94	1





~ 9

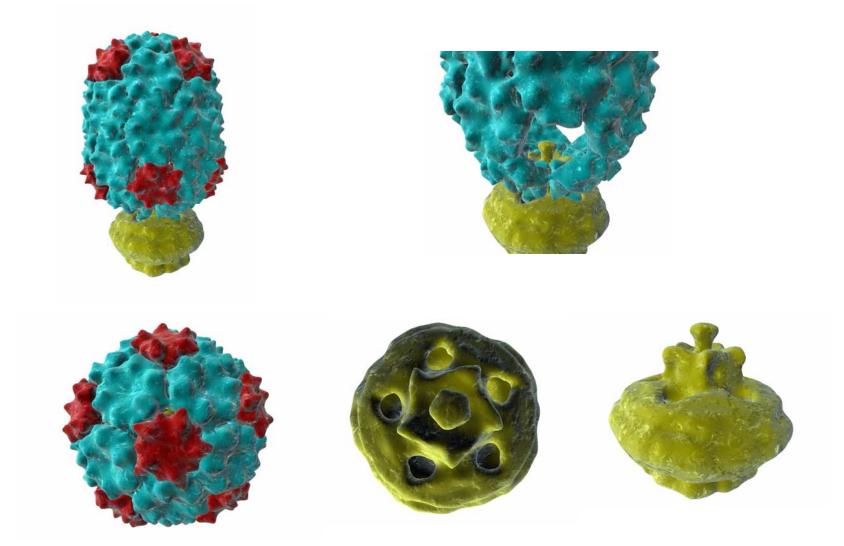
**<b><b><b><b>2<b>9** 

8-

1.2

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

### Structure Elucidation Results: **Ф29**

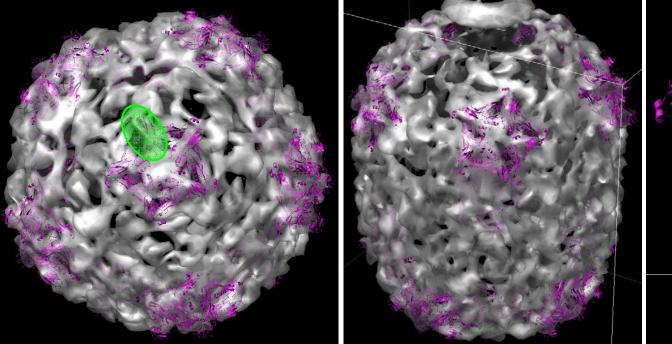


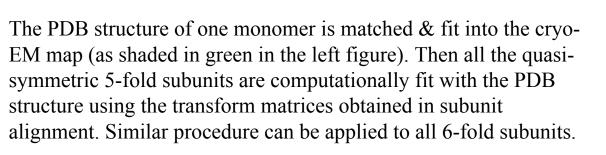


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Data courtesy:Tim BakerUniversity of Texas at AustinSeptemb

# Subunit alignment (2): Fitting

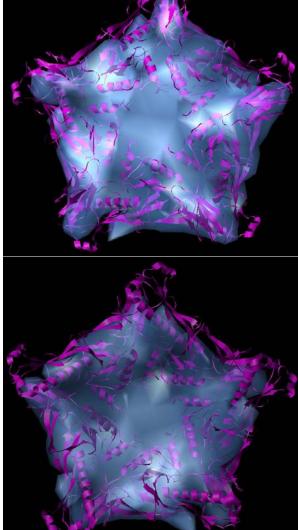






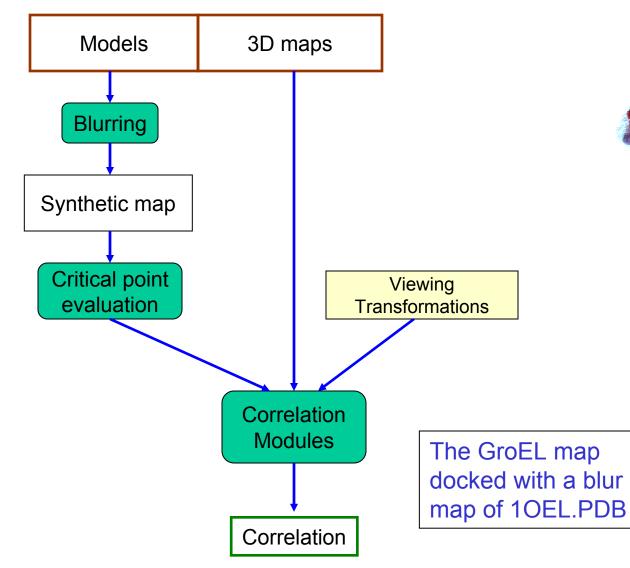
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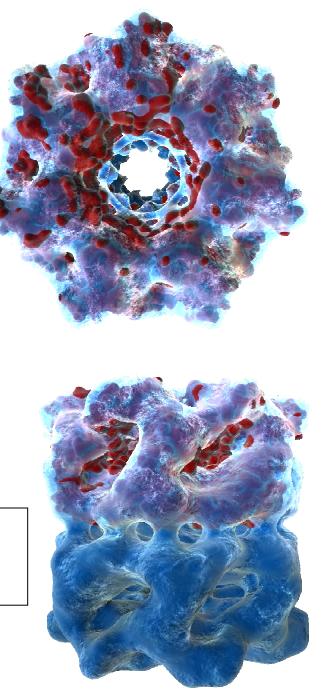
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### "Interactive" Fitting







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### Gro-EL: X-ray structures docked in Cryo-EM

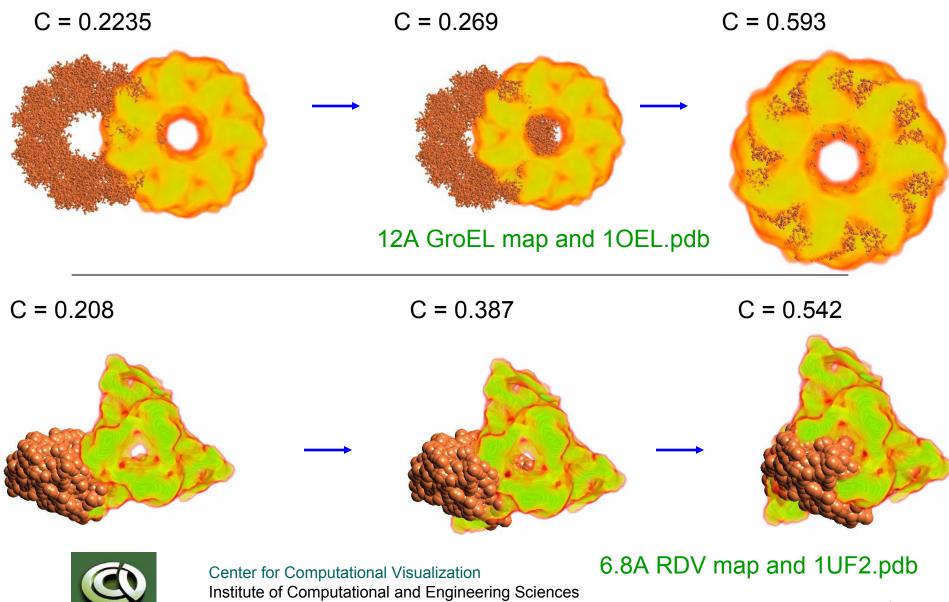




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### **Interactive Correlation Analysis**



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### **Approximate Correlation Analysis**

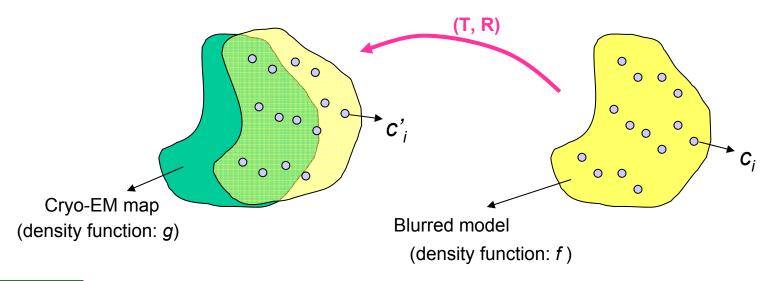
$$score = 1 - \frac{\sum_{i=1}^{N} |f(c_i) - g(c'_i)|}{\sum_{i=1}^{N} \max(f(c_i), g(c'_i))}$$

Where *f* is the normalized density function of the blurred crystal structure;

*g* is the normalized density function of the cryo-EM map;

 $c_i$ , i=1,2,...N, are the critical points of the blurred crystal structure;

 $c'_i$ , i=1,2,...N, are the transformations of the critical points.



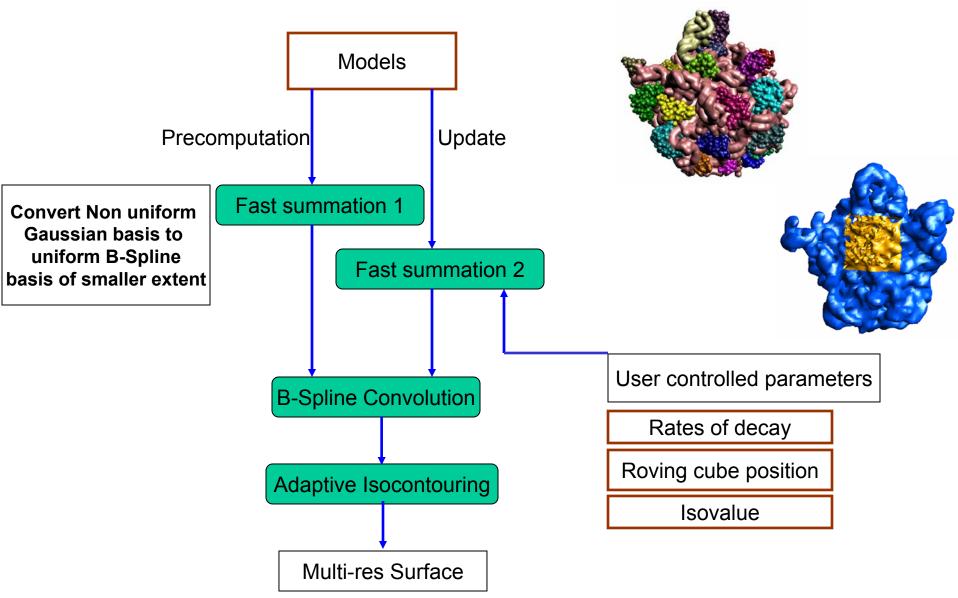


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### **Multi-resolution Molecular Surfaces**





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# Blurring I

• For a molecule with M atoms, we can define a 3D electron density map as

$$f_{elec\_dens}(\vec{\mathbf{x}}) = \sum_{i=1}^{M} G_i(\vec{\mathbf{x}}) \quad \mathbf{x} \in \mathbf{R}^3$$

• For quadratic decay kernels,  $A_i = e^d$ :

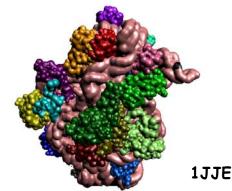
$$f_{elec\_dens}(\mathbf{x}) = \sum_{i=1}^{M} A_i e^{-\frac{d}{r^2} \mathbf{x}^2} \delta(c_i)$$

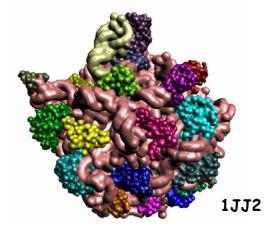
• For linear decay kernels,  $A_i = e^{dri}$ :

$$f_{elec\_dens}(\mathbf{x}) = \sum_{i=1}^{M} A_i e^{-d|\mathbf{x}|} \delta(c_i)$$



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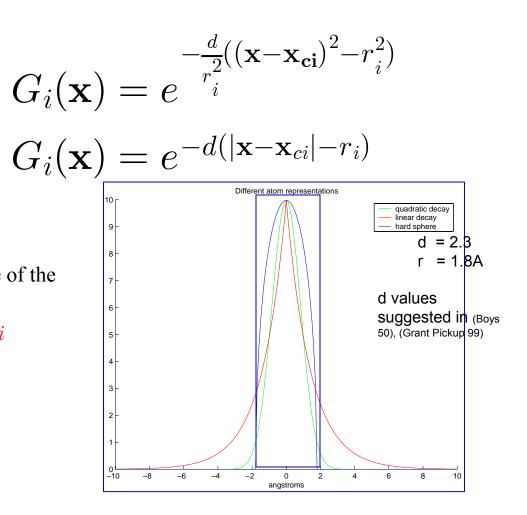




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# Atomic Shape Parameters

- Isotropic Quadratic Kernel
- Isotropic Linear Kernel
  - ➢ where
    - The decay d controls the shape of the Gaussian function.
    - $\succ$  The van der Waals radius is  $r_i$
    - > The center of the atom is  $\mathbf{x}_{c}$ .
- Anisotropic Kernels





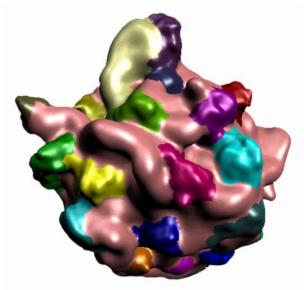
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# Blurring II

- For quadratic decay kernels,  $A_i = e^d$ :  $f_{elec\_dens}(\mathbf{x}) = \sum_{i=1}^M A_i e^{-\frac{d}{r^2}\mathbf{x}^2} \delta(c_i)$
- For linear decay kernels,  $A_i = e^{dri}$ :  $f_{elec\_dens}(\mathbf{x}) = \sum_{i=1}^{M} A_i e^{-d|\mathbf{x}|} \delta(c_i)$



• For above kernels G:

$$f_{elec\_dens}(\mathbf{x}) = G \otimes \sum_{i=1}^{M} A_i \delta(c_i)$$



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Institute of Computational and Engineering Sciences Department of Computer Sciences  $\sum_{i=1}^{M} A_i \delta(\overrightarrow{x} - \overrightarrow{x_i}) \qquad \mathbf{G}$ 

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## **FFT Solutions & Approximations**

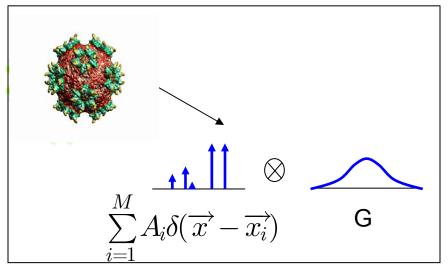
- Use the convolution theorem.
  - Convolution in spatial domain = multiplication in frequency domain.
- The 3D electron density map is a convolution of a Kernel with delta functions at the atom centers

 $- f_{elec\_dens} = FFT^{-1}(FFT(Kernel) \times FFT(atom centers))$ 

 Accuracy with Speed Tradeoff:

 The kernels are smooth functions. Hence only the first few frequencies need to be computed to obtain the summation. Also, convolution is a smoothing operator.

> - We can use an approximate FFT algorithm to calculate the low frequencies of atom centers to a high accuracy at a lower cost than a full FFT.





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# **Evaluation of 3D Electron Density Map**

Evaluation of the Electron Density of an M atom molecule at multiple points N

(typically  $N \sim 100^* M$ )

[Naiive Algorithm]

For each of the output points N, calculate the summation due to each of the M atomic kernels

• **Cost**:  $O(NM) \rightarrow$  very high!



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# Fast Approximate FT Solution

-*f*<sub>elec\_dens</sub> = *FFT*^-1 (*FFT*(Kernel) x *FFT*(atom centers))

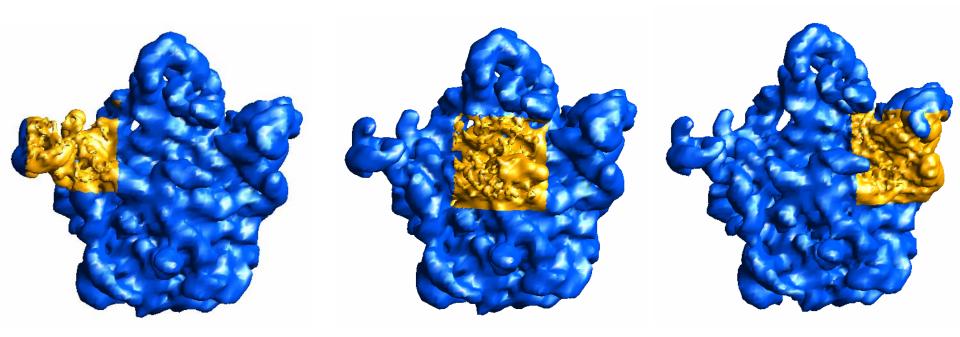
- Space complexity:
  - -O(M+N)
- Time complexity:
  - Irregularly spaced output evaluation points
    - O(MlogM + N) with larger constants
  - Regularly spaced output evaluation points
    - $O(M\log M + N \log N)$  with small constants



### Multi resolution Molecular Surfaces (Bakeoff)

1PNX.PDB

1PNY.PDB





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## Interrogative Scalable Visualization

### Techniques

- Surface/2D Textures/3D Texture based rendering exploit hardware acceleration
- Multi-dim Transfer Functions for Regions of Interest
- Multi-resolution processing
- Compressed data processing
- Parallelism (back-end PC cluster)

C.Bajaj, I. Ihm, S. Park, ACM Trans. on Graphics, 20, 1, 10-28, 2001

C Bajaj, P Djeu, V Siddavanahalli, A Thane, IEEE visualization, 2003. 243-250.

C. Bajaj, J. Castrillon-Candas, S. Vinay, A. Xu, Structure, 13, 3, 2005, 463-471

X. Zhang, C. Bajaj, IEEE Symp. On Parallel, Large Data Visualization, 2001, 51-58



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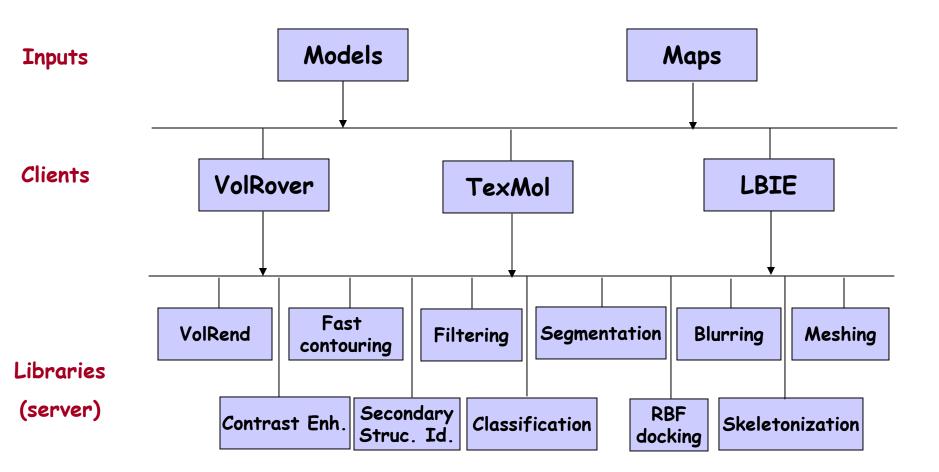


### Communication & Education: Beyond silent movies!

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### Molecular Visualization and Processing (MVP) Environment

Open Source/Public Domain under LGPL (http://ccvweb.csres.utexas.edu/software/)





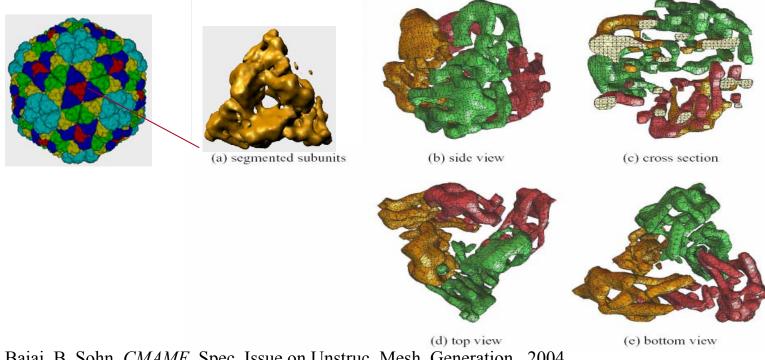
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# Modeling 2(A): Finite Element Meshing

• LBIE : also supports Integral/Derivative Property computations (areas, volumes, curvatures)



- •Y. Zhang, C. Bajaj, B. Sohn, CMAME, Spec. Issue on Unstruc. Mesh. Generation., 2004
- •Y.Zhang, C.Bajaj, G. Xu, 14th Intl Meshing Roundtable, San Diego2005.
- •Y. Zhang, C. Bajaj, CMAME,2005 (in press)



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## Summary: Algorithms & Tools

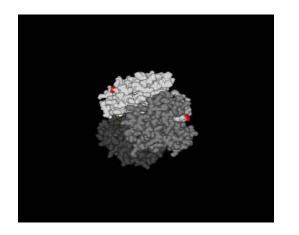
- Structure elucidation: filtering, contrast enhancement, segmentation, skeletonization, subunit identification
- Structure Modeling: finite element meshing, spline representations(Aspline,RBF representations) for structural fitting & complementary docking
- Visualization: multi-dimensional transfer functions, surface and volume texture rendering, wavelet compression, hierarchical representations, cluster based parallelism

C. Bajaj, Chap in Modeling Biology, MIT Press 2005

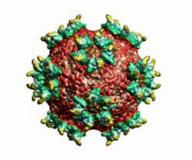
C. Bajaj, Z. Yu, Chap in



Handbook of Comp. Mol. Bio, Center for Computational And Engineering Sciences Department of Computer Sciences



VolRover



TexMol

University of Texas at Austin

## The CVC Team and Collaborators

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  - Peter Rossky (ICES)
  - Andre Sali (UCSF)
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