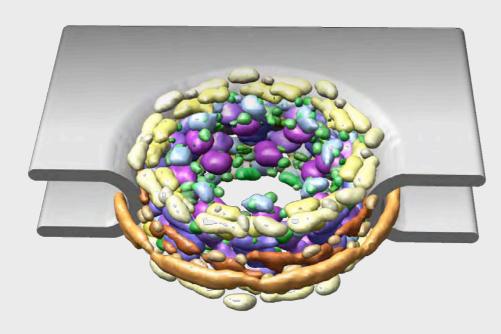
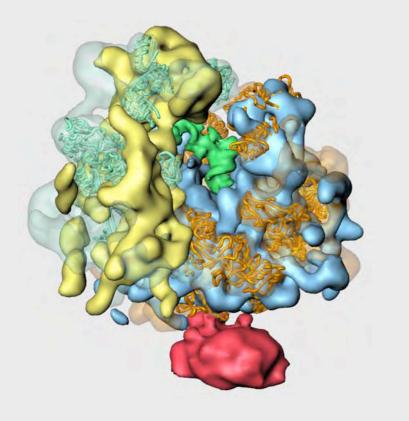
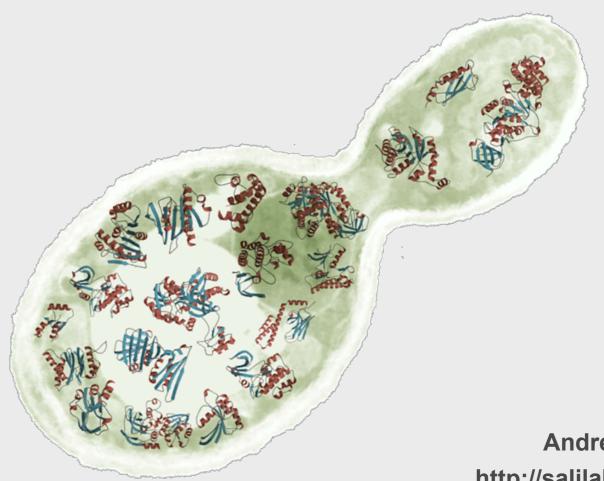
Determining the Structures of Proteins and Macromolecular Assemblies







Andrej Sali http://salilab.org/

Department of Bioengineering and Therapeutic Sciences Department of Pharmaceutical Chemistry California Institute for Quantitative Biosciences University of California at San Francisco



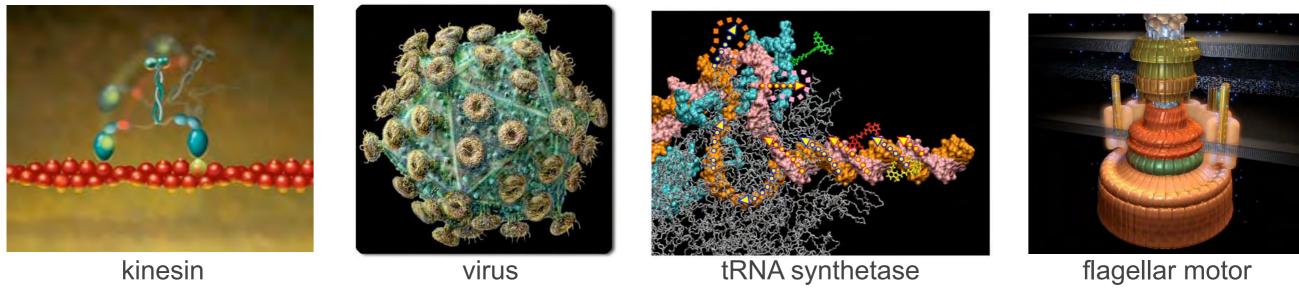
Topics

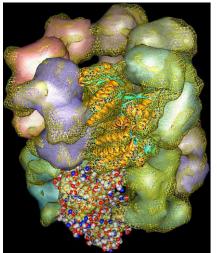
- 1. Introduction to integrative (hybrid) structure determination
- 2. Comparative model building
- 3. Predicting accuracy of atomic models
- 4. Iterative sequence-structure alignment and model building
- 5. Electron microscopy
- 6. Small angle x-ray scattering
- 7. Proteomics
- 8. Concluding Remarks

Immediate Goal

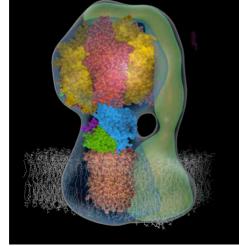
Maximize accuracy, resolution, completeness, and efficiency of the structural coverage of proteins and their assemblies (static structures).

Motivation: Structures will allow us to understand how machines work, how they evolved, how they can be controlled, modified, and perhaps even designed.

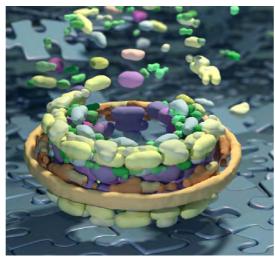




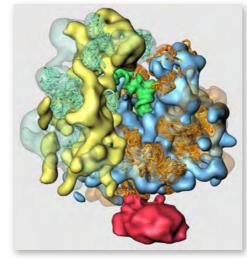
GroEL chaperonin



ATP synthase



nuclear pore complex



ribosome

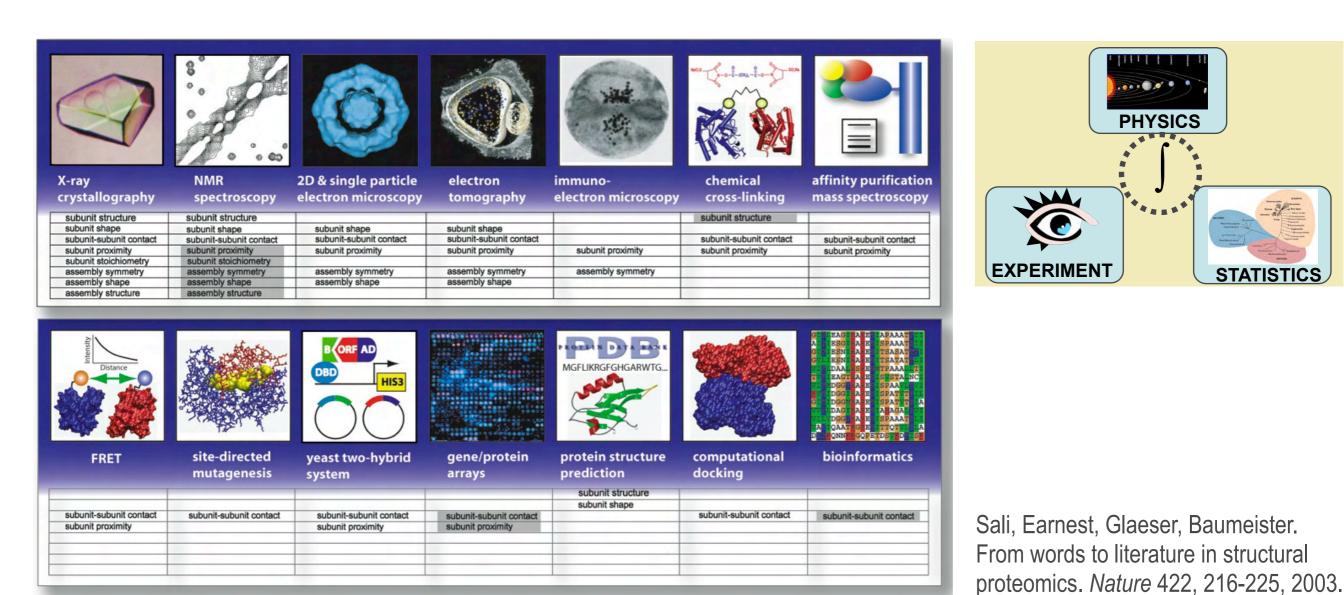
There are thousands of biologically relevant macromolecular complexes whose structures are yet to be characterized, involved in a few hundred core biological processes.

Mindset

for maximizing accuracy, resolution, completeness, and efficiency of structure determination

Use structural information from any source: measurement, first principles, rules; resolution: low or high resolution

to obtain the set of all models that are consistent with it.



Integrative (hybrid) methods for structure determination

- Integrative structure determination relies on varied types of data.
- Atomic structure determination:
 - x-ray crystallography (D. Baker ...).
 - NMR spectroscopy (M. Nilges, M. Vendruscolo, A. Bax, D. Baker, G. Montelione, ...).
- Low-resolution description of macromolecular assemblies:
 - fitting of atomic models into a cryo-EM map (M. Rossman, ...).
 - integrating proteomics data (A. Sali, ...).
- Modeling can greatly leverage experimental data in order to determine the structures and dynamics of proteins and especially macromolecular assemblies.

Characterizing Structures by Satisfaction of Spatial Restraints

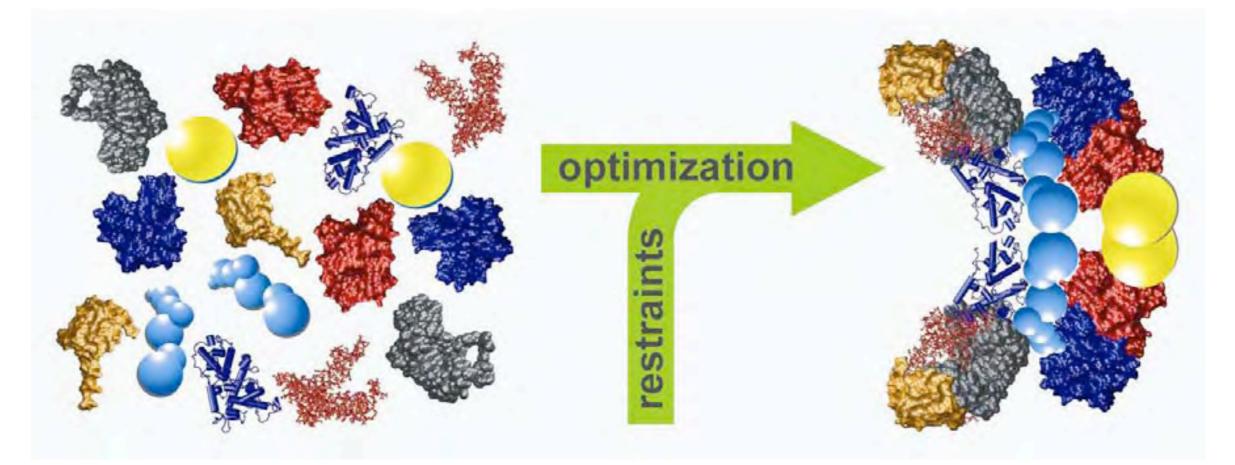
1. Representation of a system.

2. Scoring function (spatial restraints).

3. Optimization / sampling.

There is nothing but points and restraints on them. We seek joint pdf for **R**, given information **I**:

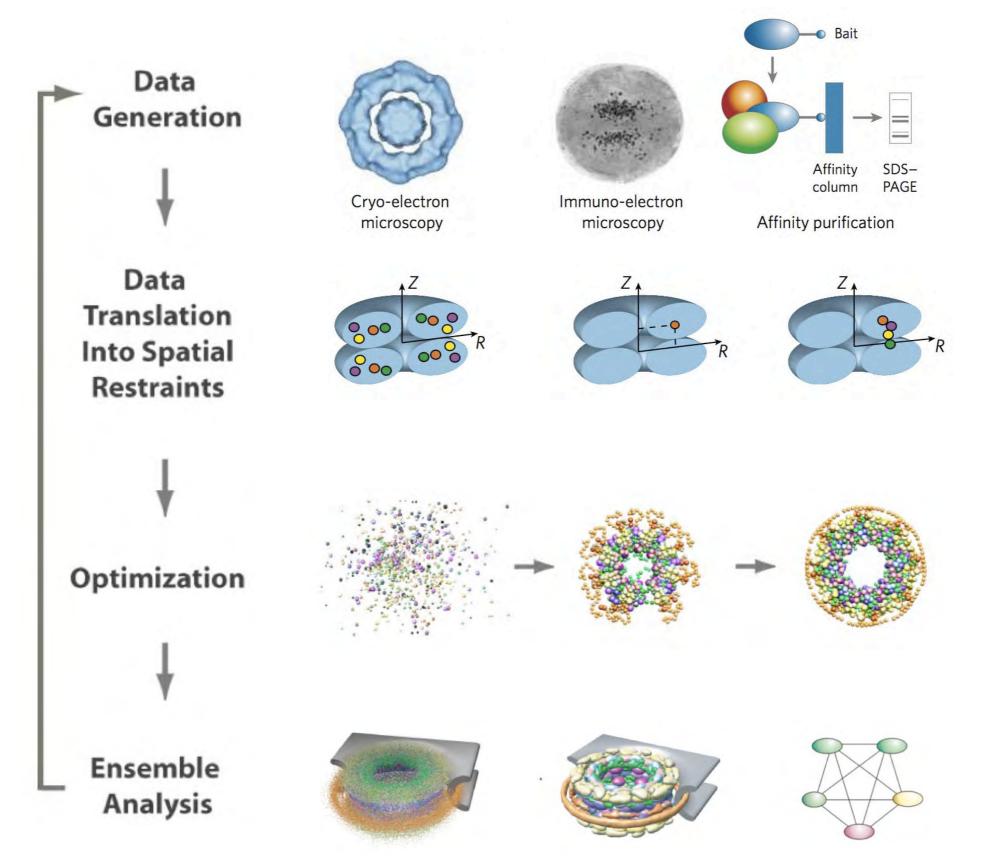
 $P(\mathbf{R} / \mathbf{I}) \approx \prod_{i} p_i (\mathbf{r}_i / \mathbf{I}_i)$



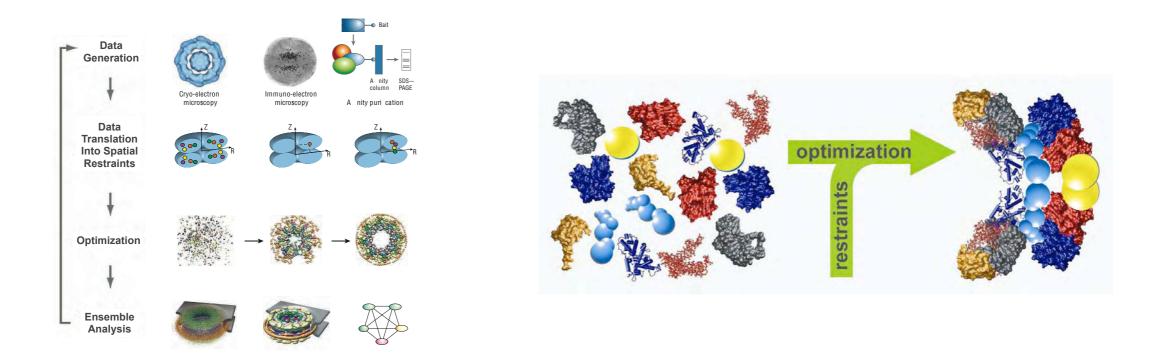
Integrative Modeling Platform (IMP): http://salilab.org/imp

Using All Spatial Information

Alber *et al. Nature* 450, 683-694, 2007. Robinson, Sali, Baumeister. *Nature* 450, 974-982, 2007. Alber, Foerster, Korkin, Topf, Sali. *Annual Reviews in Biochemistry* 77, 11.1–11.35, 2008.



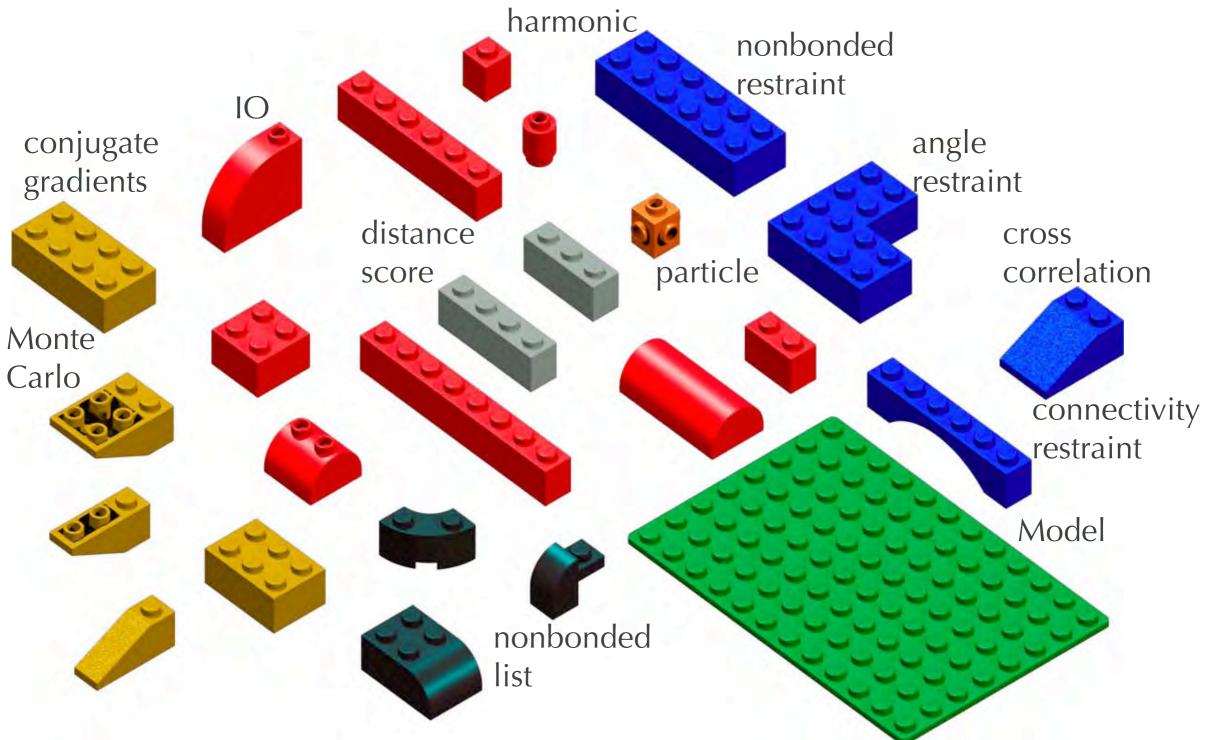
Why Integrative Modeling?



- 1. Benefits from the **synergy** among the input data, maximizing accuracy, resolution, completness, and efficiency of structure characterization.
- 2. Finds "all" models consistent with the data, not just one.
- 3. Facilitates **assessing** the results in terms of precision and accuracy.
- 4. Provides feedback to **guide** future experiments (*eg*, "what if", ...).

Integrative Modeling Platform (IMP): Building blocks for modeling

http://salilab.org/imp



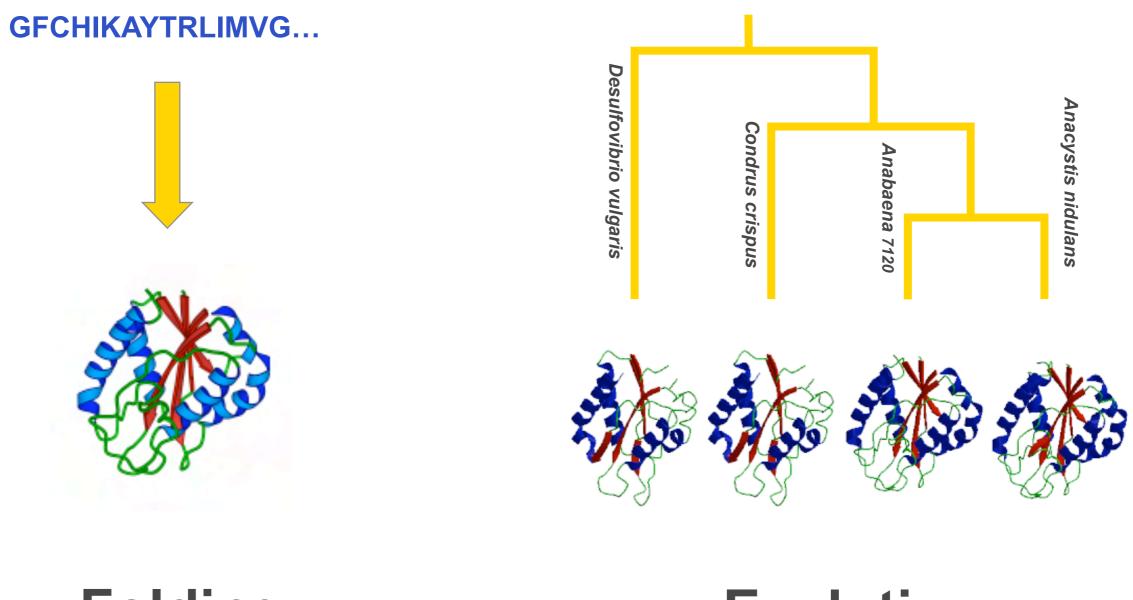
D. Russel, B. Webb, K. Lasker, F. Alber, B. Peterson

Topics

- 1. Introduction to integrative (hybrid) structure determination
- 2. Comparative model building
- 3. Predicting accuracy of atomic models
- 4. Iterative sequence-structure alignment and model building
- 5. Electron microscopy
- 6. Small angle x-ray scattering
- 7. Proteomics
- 8. Concluding Remarks

Principles of protein structure

D. Baker & A. Sali. Science 294, 93-97, 2001.



Folding

(physics)

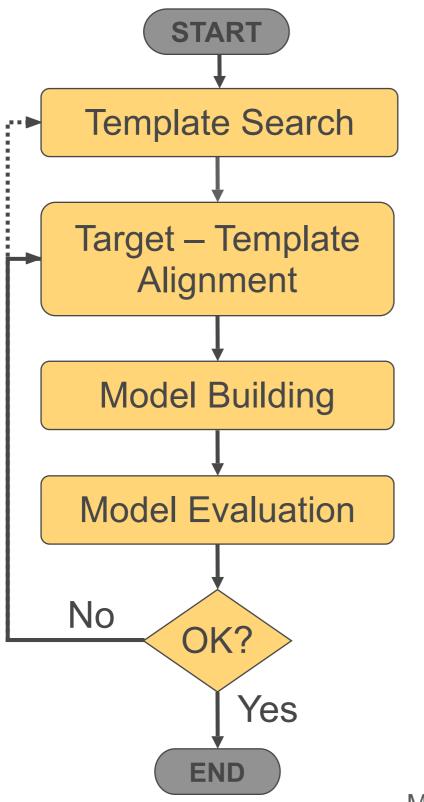
Ab initio prediction

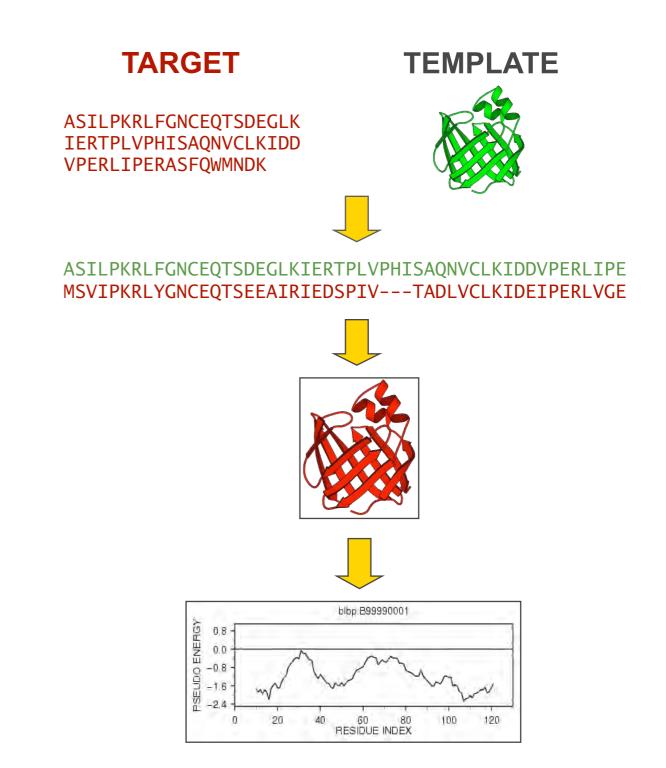
Evolution

("statistical" rules)

Threading Comparative Modeling

Steps in Comparative Protein Structure Modeling

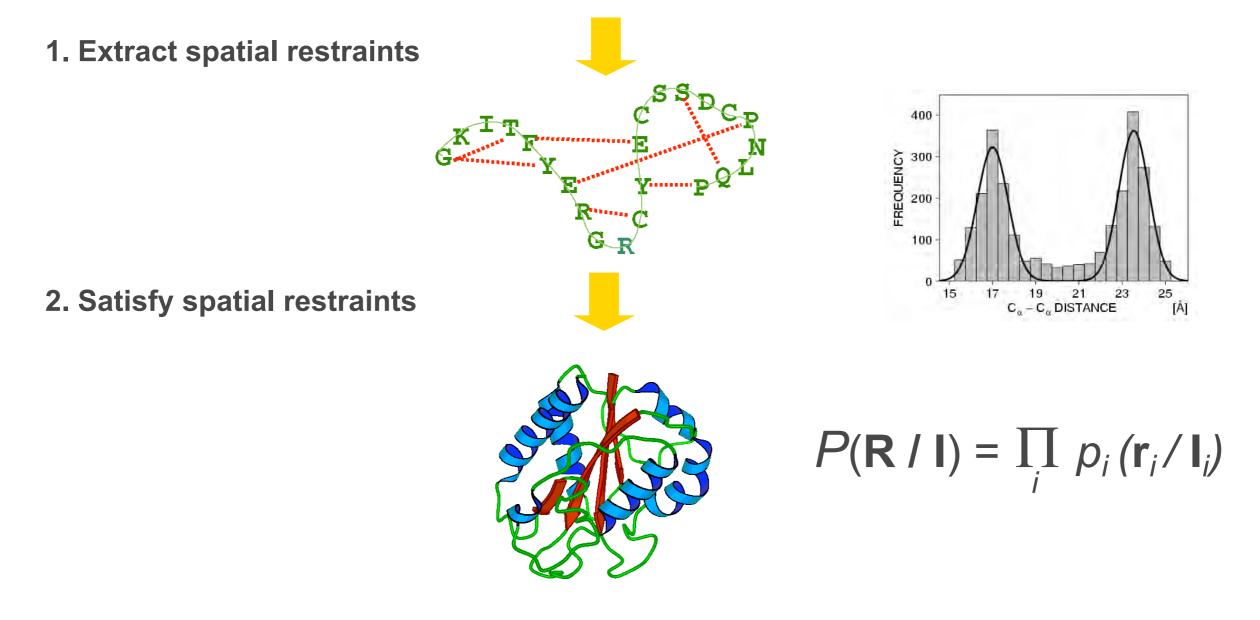




M. Marti-Renom *et al. Ann. Rev. Biophys. Biomolec. Struct.* **29**, 291, 2000. <u>http://salilab.org</u>/

Comparative modeling by satisfaction of spatial restraints MODELLER

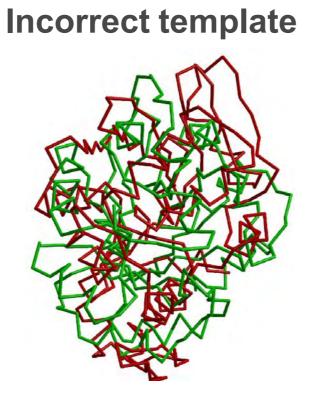
3DGKITFYERGFQGHCYESDC-NLQP...SEQGKITFYERG---RCYESDCPNLQP...



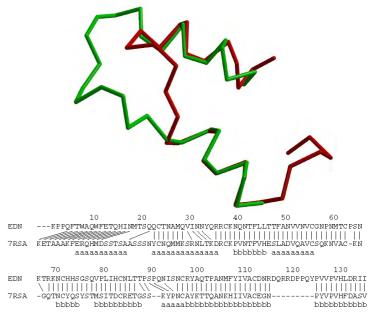
A. Šali & T. Blundell. *J. Mol. Biol.* 234, 779, 1993.
J.P. Overington & A. Šali. *Prot. Sci.* 3, 1582, 1994.
A. Fiser, R. Do & A. Šali, *Prot. Sci.*, 9, 1753, 2000.

http://salilab.org/

Typical errors in comparative models



Misalignment

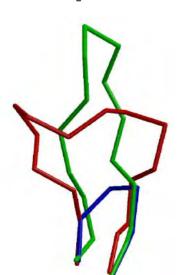


Region without a template

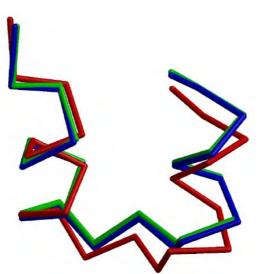
MODEL

X-RAY

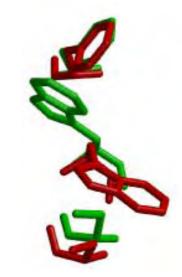
TEMPLATE



Distortion/shifts in aligned regions

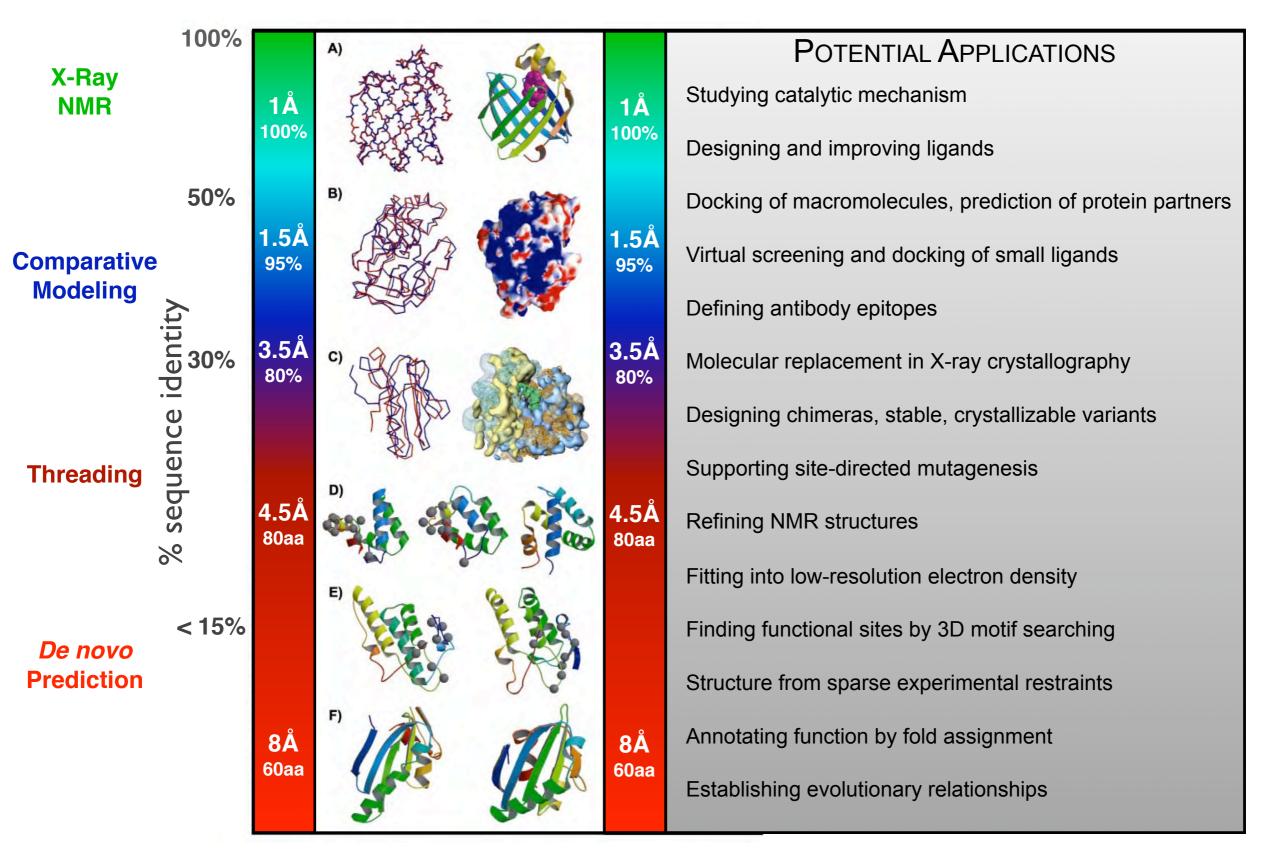


Sidechain packing



Marti-Renom et al. Annu.Rev.Biophys.Biomol.Struct. 29, 291-325, 2000.

Model accuracy determines utility

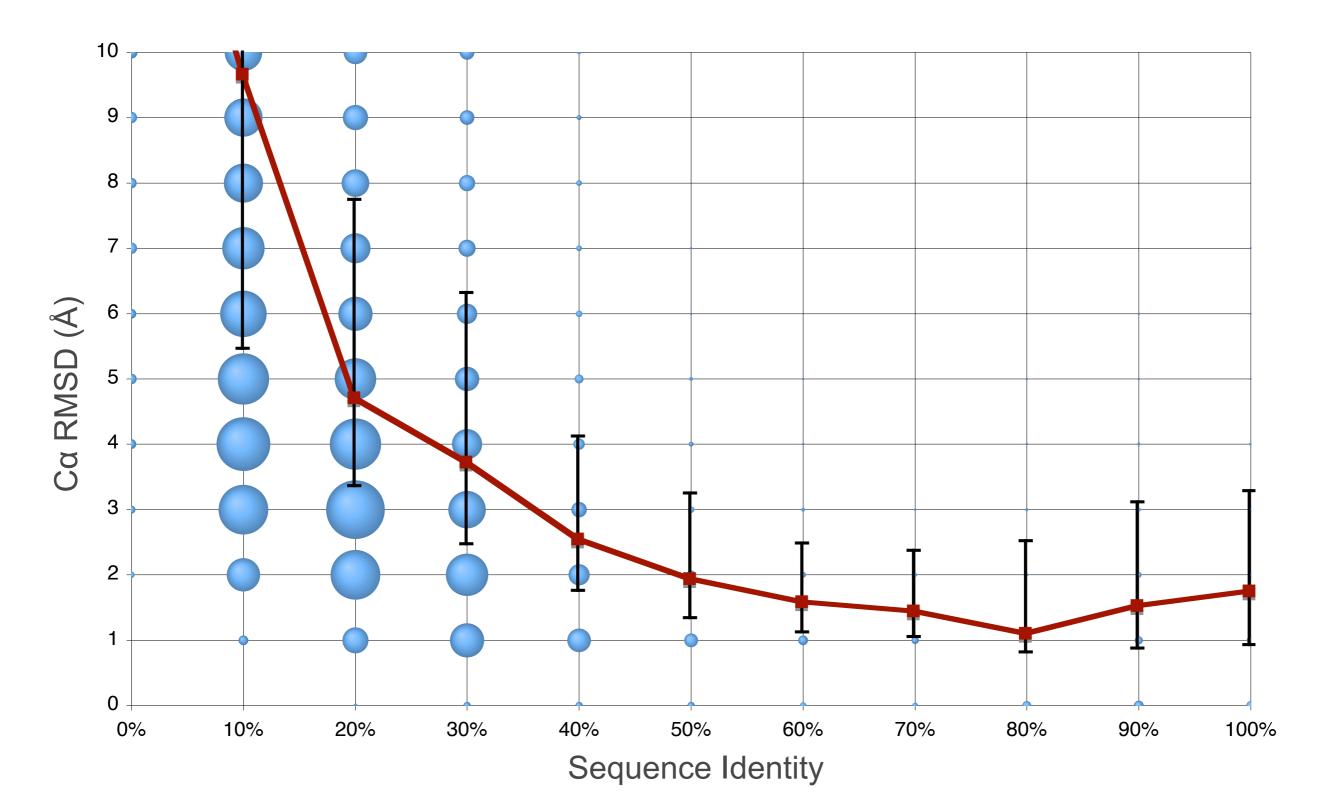


D. Baker & A. Sali. Science 294, 93-97, 2001.

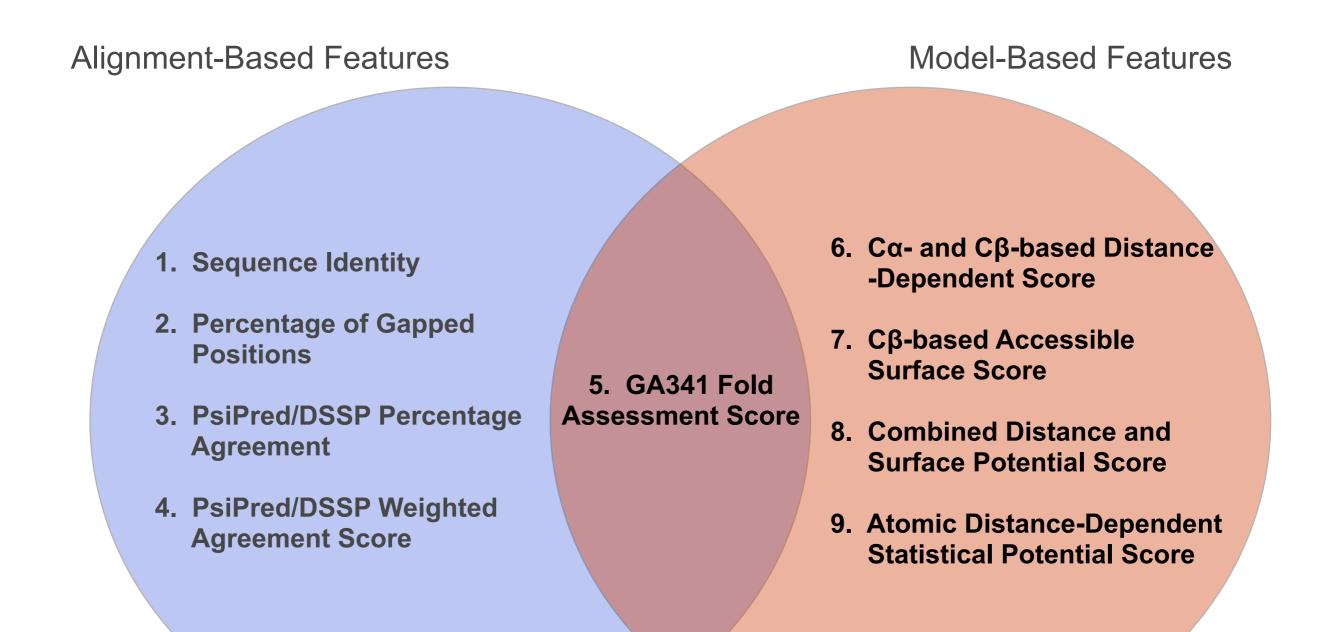
Topics

- 1. Introduction to integrative (hybrid) structure determination
- 2. Comparative model building
- 3. Predicting accuracy of atomic models
- 4. Iterative sequence-structure alignment and model building
- 5. Electron microscopy
- 6. Small angle x-ray scattering
- 7. Proteomics
- 8. Concluding Remarks

Comparative model accuracy varies widely with decreasing sequence identity



Model Assessment Scores



Outline of TSVMod

Input: Atomic model and optional alignment.

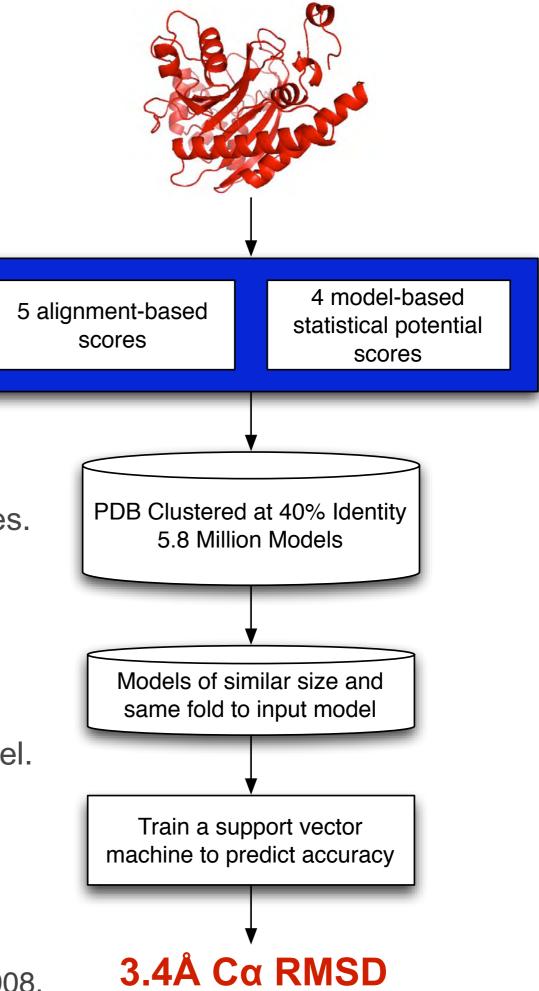
Output: Predicted Cα RMSD error.

Algorithm:

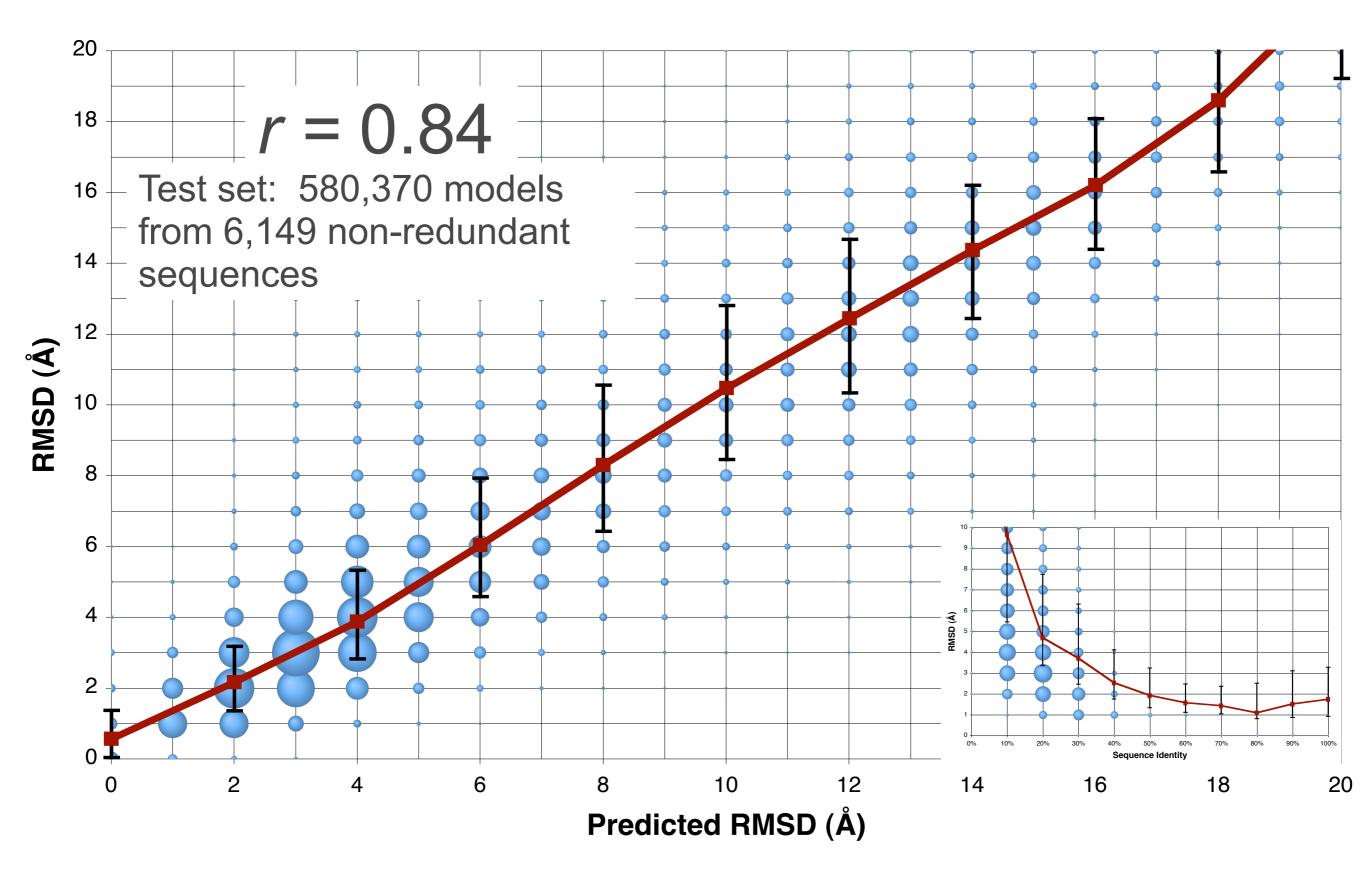
1. Calculate individual model assessment scores.

- 2.Scan through PDB model database.
- 3. Construct a tailored model training set.
- 4. Train a specialized Support Vector Machine.
- 5. Run the Support Vector Machine on the model.

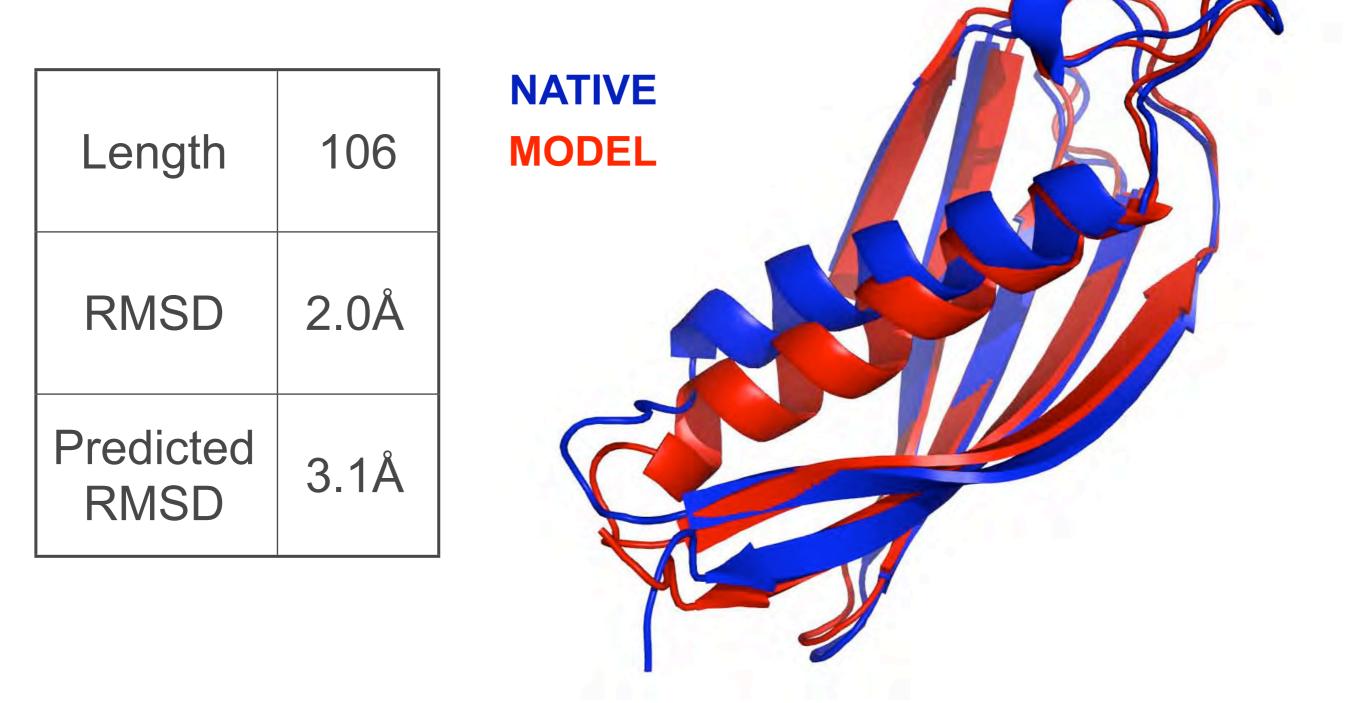
D. Eramian, N. Eswar, M-Y Shen, A. Sali, Prot. Sci, 2008.



Predicted versus actual Cα RMSD error



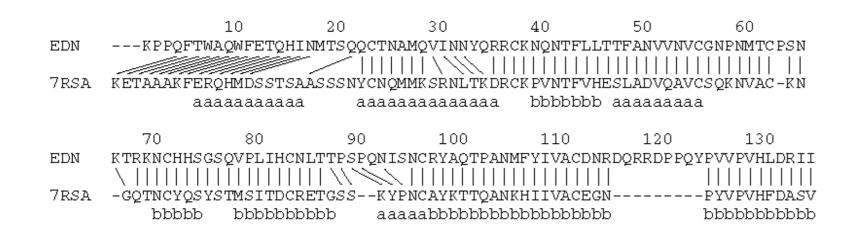
Good Model in the Midnight Zone: 12.3% Sequence Identity

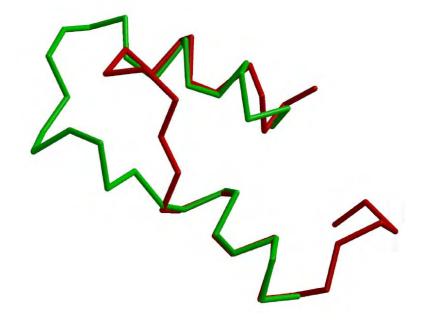


Topics

- 1. Introduction to integrative (hybrid) structure determination
- 2. Comparative model building
- 3. Predicting accuracy of atomic models
- 4. Iterative sequence-structure alignment and model building
- 5. Electron microscopy
- 6. Small angle x-ray scattering
- 7. Proteomics
- 8. Concluding Remarks

Minimizing errors in sequence-structure alignment

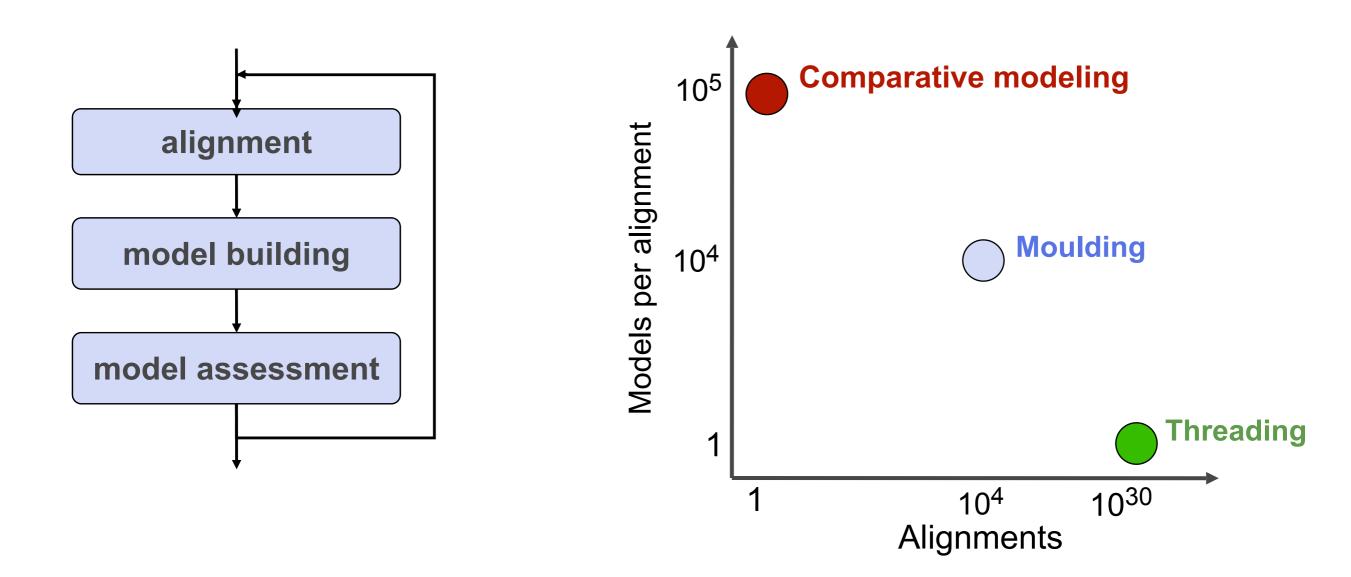




- Complex gap penalty functions.
- Multiple sequence profiles.
- Hidden Markov Models.
- Threading.

Moulding: iterative alignment, model building, model assessment

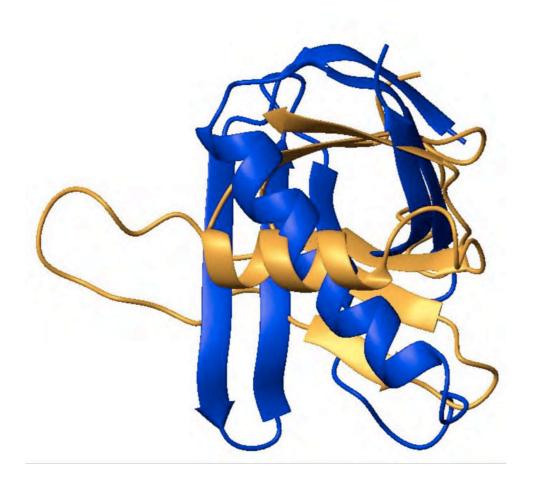
B. John, A. Sali. *Nucl. Acids Res.* **31**, 1982-1992, 2003. D. Eramian, B. Webb.



Application to a difficult modeling case 1BOV-1LTS (4.4% sequence identity)

initial

final



 $C\alpha$ RMSD 10.1 Å

 $C\alpha$ RMSD 3.6 Å

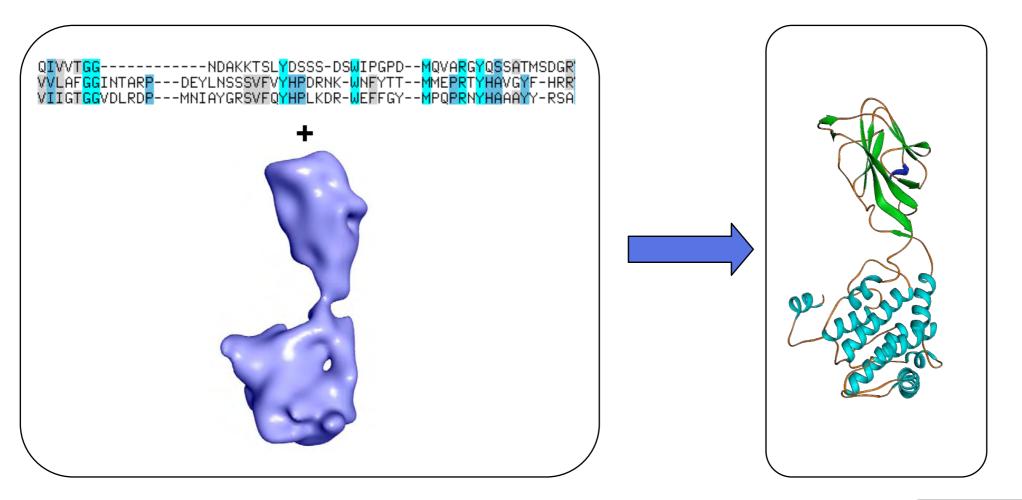
1lts structure1lts model

Topics

- 1. Introduction to integrative (hybrid) structure determination
- 2. Comparative model building
- 3. Predicting accuracy of atomic models
- 4. Iterative sequence-structure alignment and model building
- 5. Electron microscopy
- 6. Small angle x-ray scattering
- 7. Proteomics
- 8. Concluding Remarks

Comparative modeling and fitting into EM density

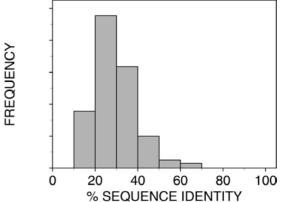
Improve comparative modeling by fitting models into the target EM density map; Improve fitting into an EM density map by simultaneous model building.



~50,000

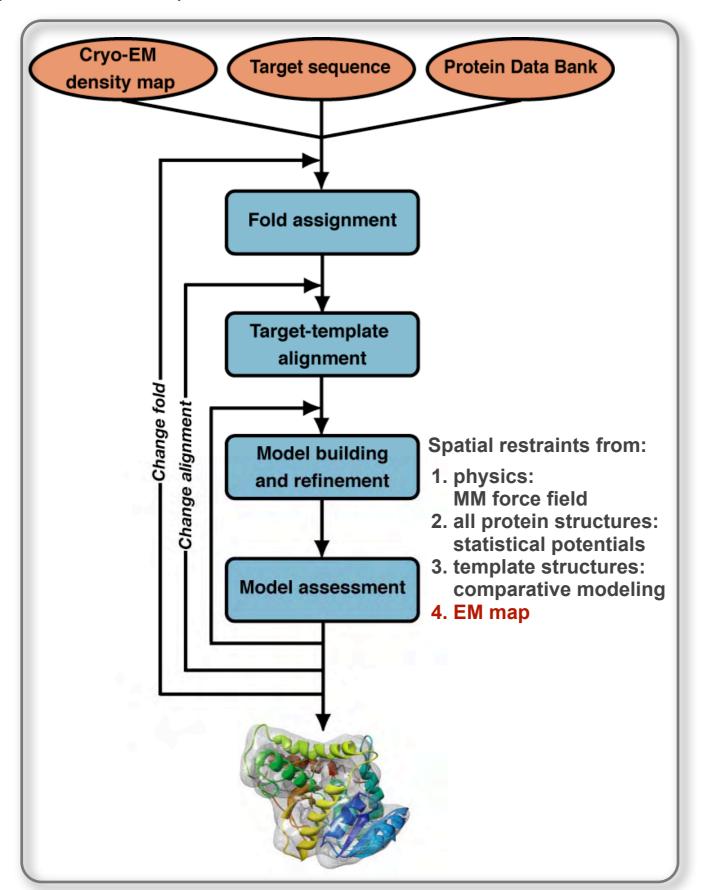
Motivation:

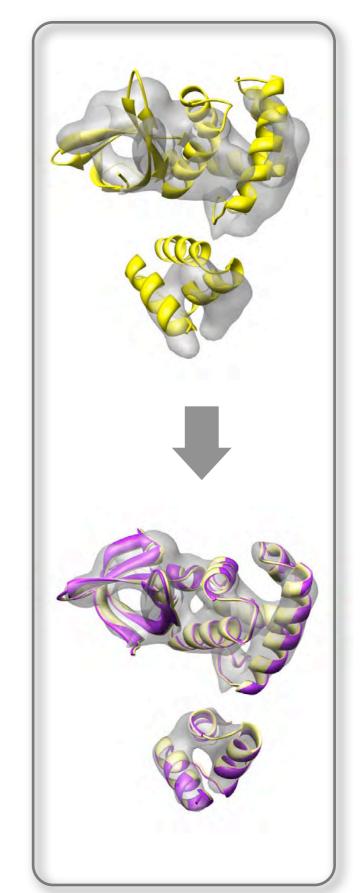
- Number of known structures in PDB:
- Number of known sequences modeled by CM: ~1,800,000
 Pieper *et al*, Nucl. Acids Res., 2006.



Protein structure modeling in an EM map

Topf, Baker, John, Chiu, Sali. *J. Struct. Biol*, 2004. Topf & Sali, *Curr. Opin. Str. Biol.*, 2005. Topf, Baker, Marti-Renom, Chiu & Sali. *J. Mol. Biol.*, 2006. Topf, Lasker, Webb, Wolfson, Chiu & Sali. *Structure*, 2008.

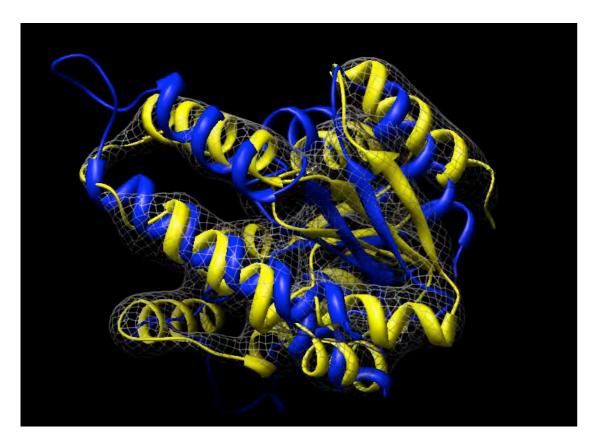


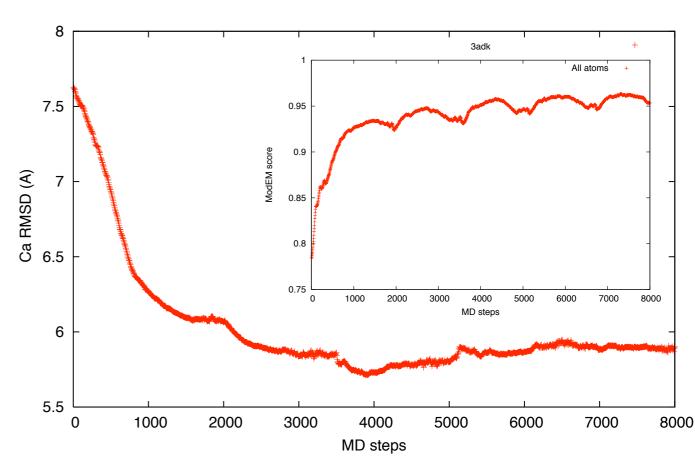


Sample refinement of 1adk

Topf, Lasker, Webb, Wolfson, Chiu & Sali. Structure, 2008.

- EM map (10 Å) from native structure;
- secondary structure segments as rigid bodies, loops flexible;
- scoring function consisting only of model-map correlation coefficient, soft-sphere atom overlap, stereochemistry;
- optimization by a combination of "molecular dynamics" with simulated annealing and conjugate gradients minimization.

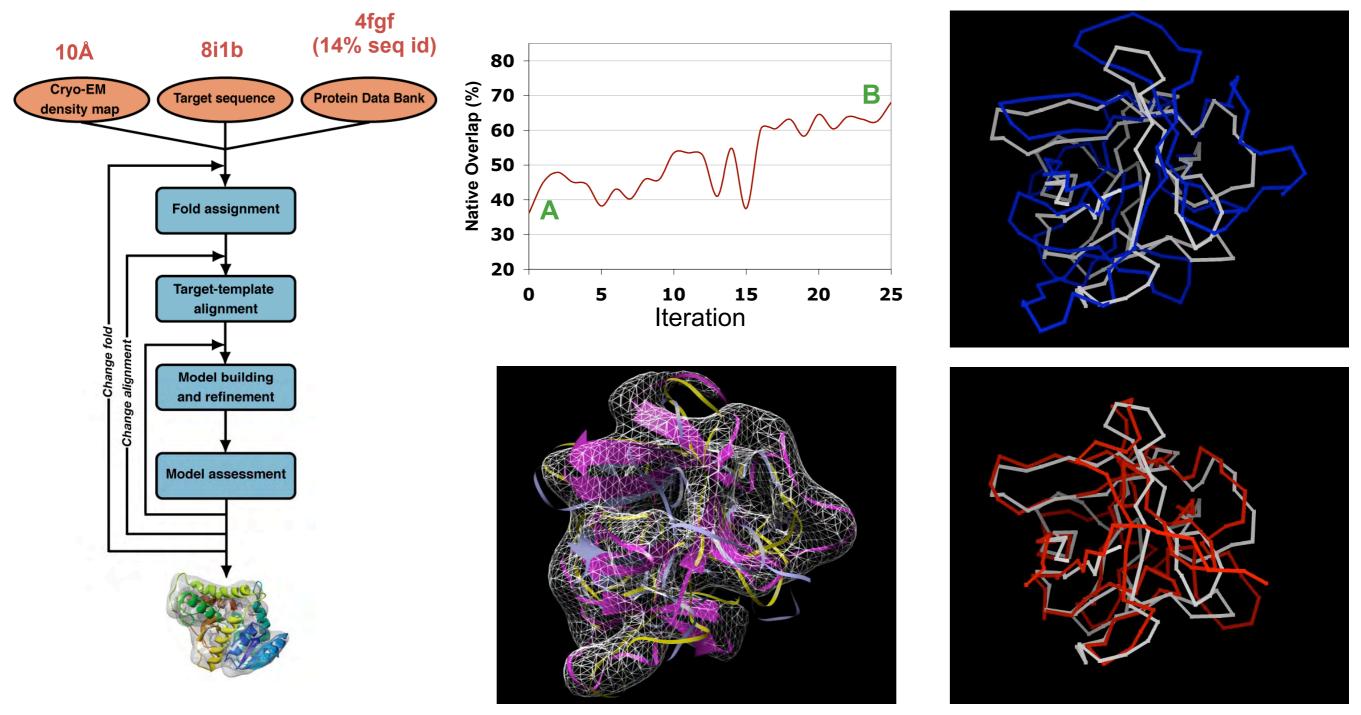




3adk

Moulding into EM maps

Topf, Baker, Marti-Renom, Chiu & Sali. J. Mol. Biol., 2006.

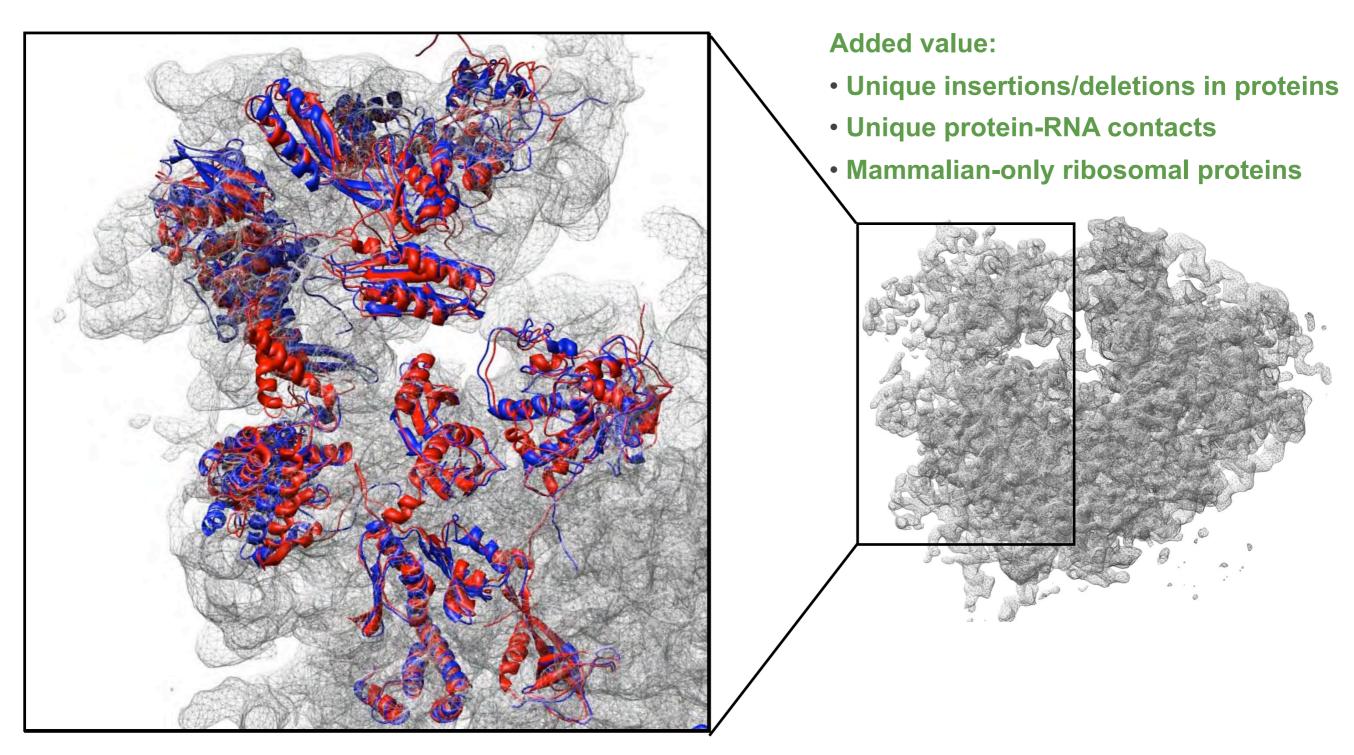


A. 37% of C_{α} within 5Å

B. 69% of C_{α} within 5Å

Dog Ribosome at 8.7 Å

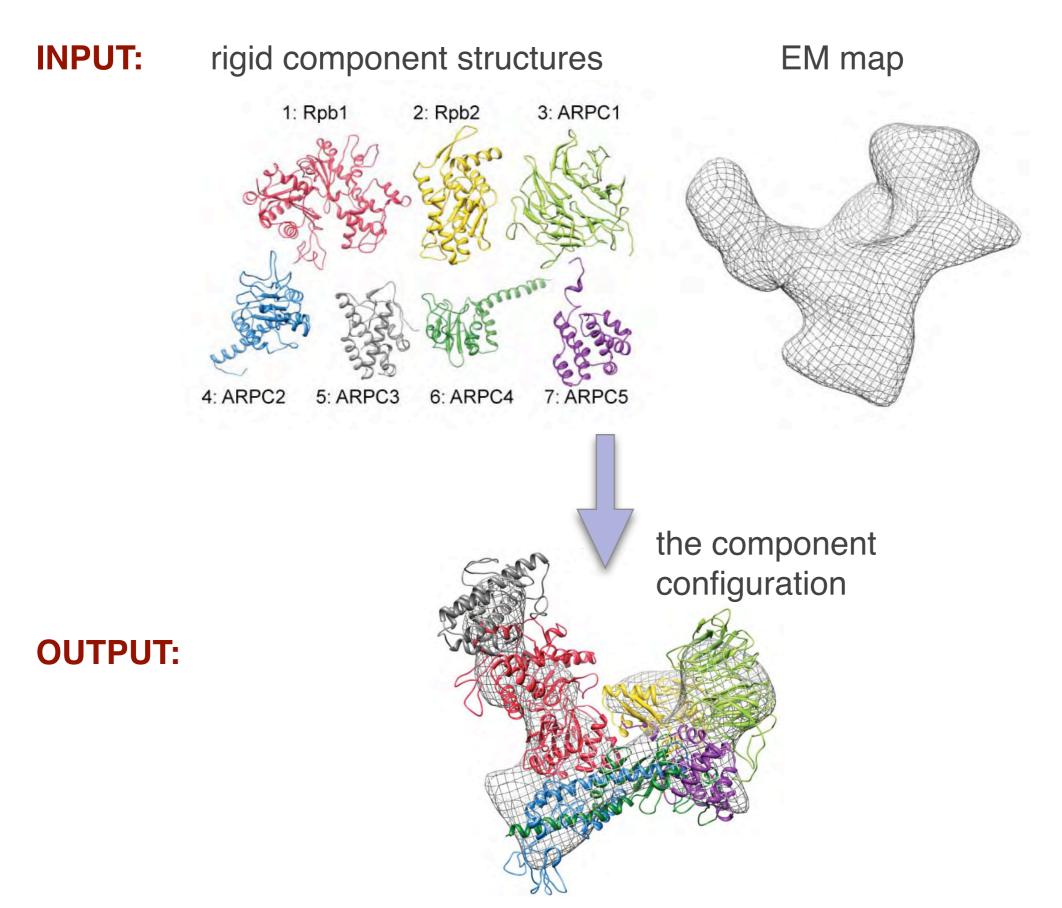
Chandramouli, Topf, Menetret, Eswar, Gutell, Sali, Akey. Structure, 2008.



Thermus Thermophilus 30S ribosomal subunit (proteins - red; RNA - yellow) Homology models of the mammalian ribosomal proteins (blue)

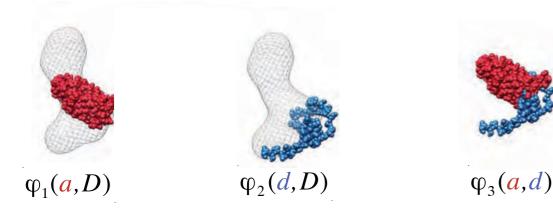
Fitting multiple components into a cryoEM map

K. Lasker, M. Topf, A. Sali, H. Wolfson



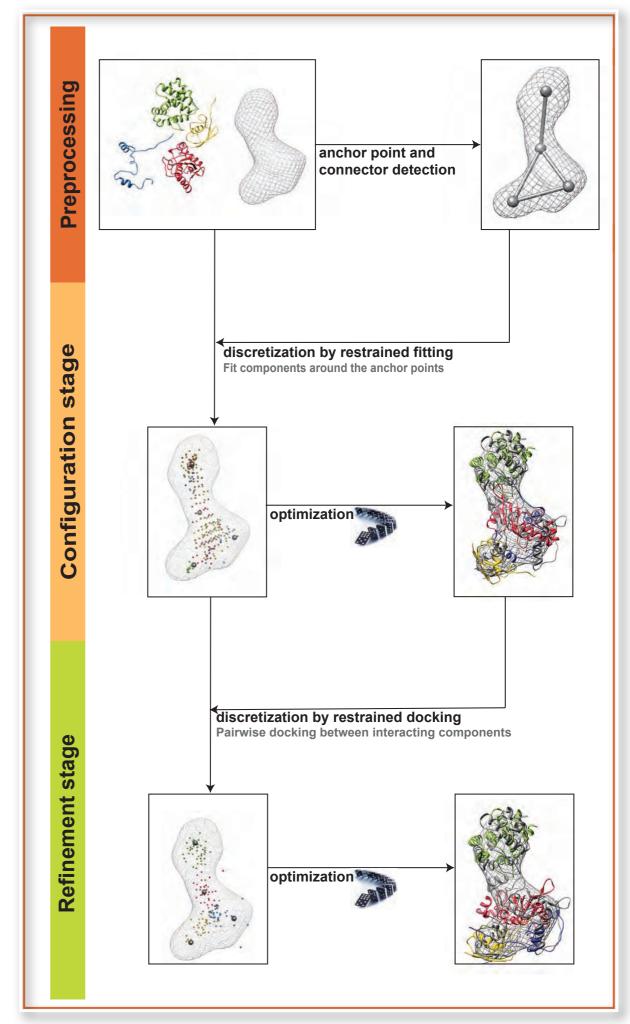
The MultiFit approach

$$\begin{split} \mathbf{S}(a, b, c, d, D) &= \sum_{x \in \{a, b, c, d\}} \{ \varphi_1(x, D) + \varphi_2(x, D) \} + \sum_{(x, y) \in B} \varphi_3(x, y) \\ B &= \{ (a, b)(a, c), (a, d), (b, c), (b, d), (c, d) \} \end{split}$$



The scoring function includes:

- quality-of-fit of components in the map.
- protrusion of components from the map.
- shape complementarity between pairs of components.

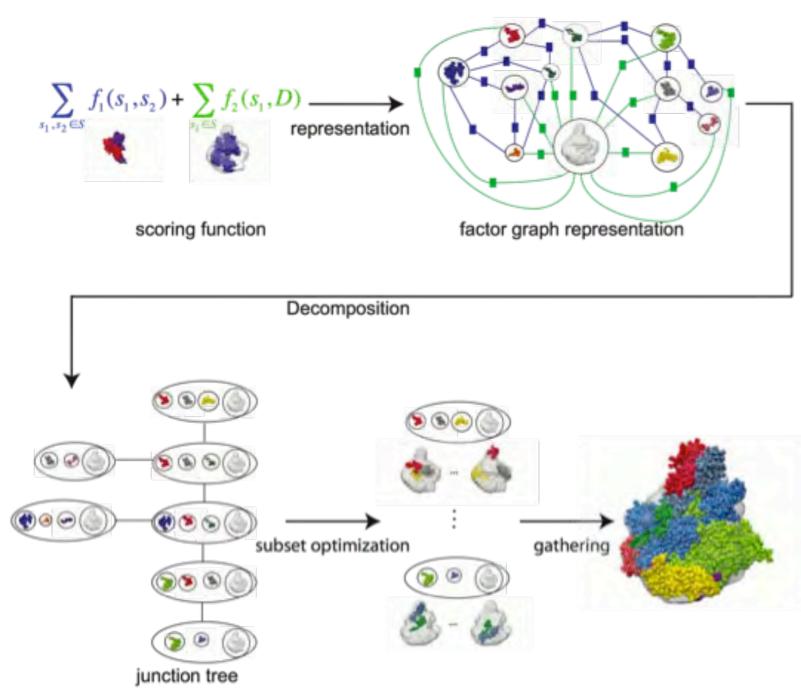


Lasker, Topf, Sali, Wolfson



DOMINO: Divide-and-Conquer

- 1. **Represent** the scoring function as a factor graph.
- 2. **Decompose** the set of components into relatively decoupled subsets (a junction tree algorithm from graph theory).
- 3. **Optimize** each subset independently by a traditional optimizer, to get the optimal and a number of suboptimal solutions (restrained fitting for configuration stage and restrained docking for refinement stage).
- 4. **Gather** subset solutions into the best possible global solutions (message passing algorithms from graph theory; *eg*, belief-propagation).

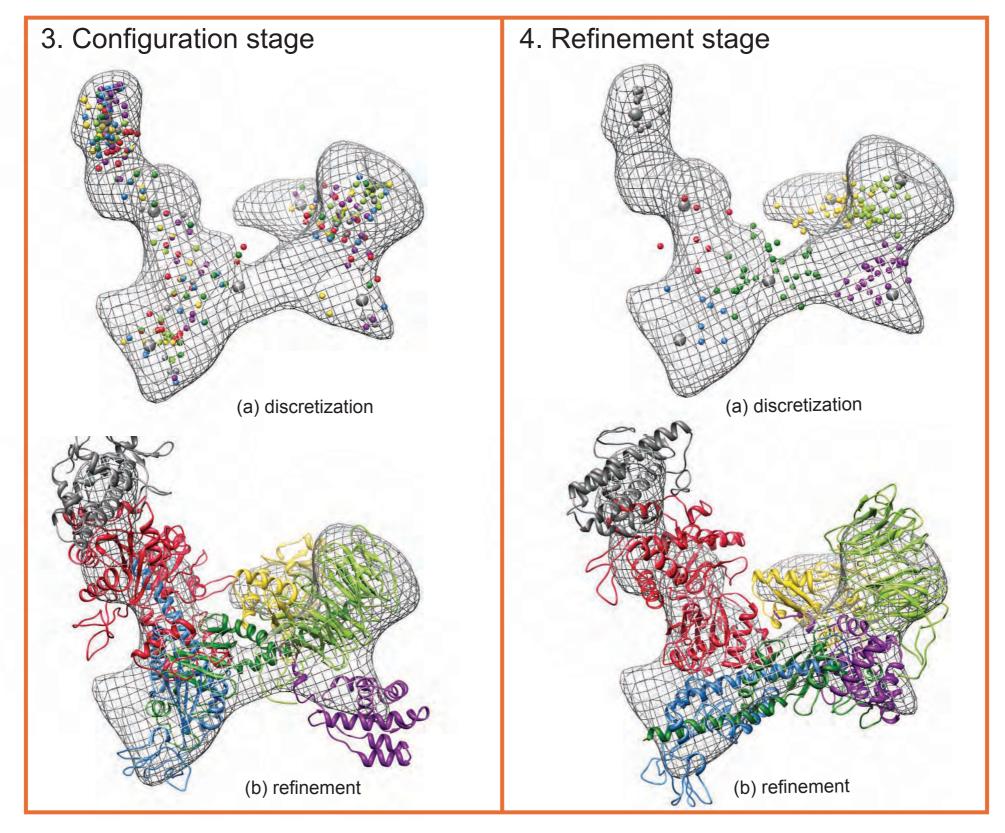


additional notes:

- factor graph simplification by eliminating terms for non-interacting components, given a component mapping to anchor points
- branch-and-bound for optimizing mappings, in configuration stage

Arp2/3 Example: Optimization stages



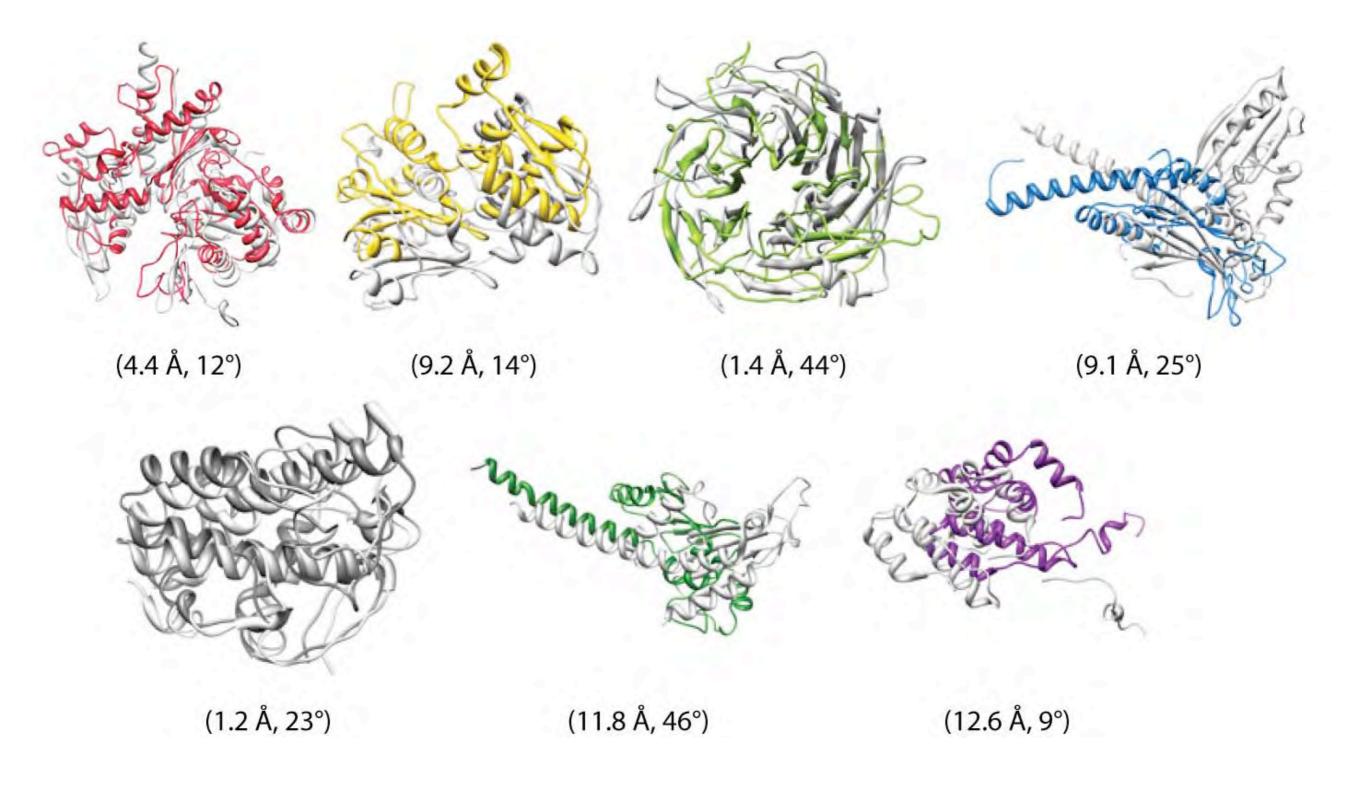


(10.8 Å, 136°) Assembly placement score

(7.1 Å, 25°)

Assembly placement score

Component displacement errors





Benchmark results

Assembly name, # components	Resolution, Å	Average sequence identity (%)	Configuration Score (Å, °), rank
groEL, 3 domains	20	65	(2.6, 13), 1
groEL/groES, 4 domains	experimental map 23.5	100	(9.3 Å, 74), 3
SUMO-RanGAP1-Ubc9- Nup358 complex, 4 proteins	20	100	(5.0, 67), 1
SUMO-RanGAP1-Ubc9- Nup358 complex, 4 proteins	20	37	(5.4, 62), 3
Dihydropyrimidine Dehydrogenase,5 domains	20	100	(2.6, 4), 1
Archaeon <i>Methanopyrus kandleri</i> , 6 proteins	20	61	(2.5, 8), 1
Arp 2/3, 7 proteins	20	51	(7.1, 25), 4

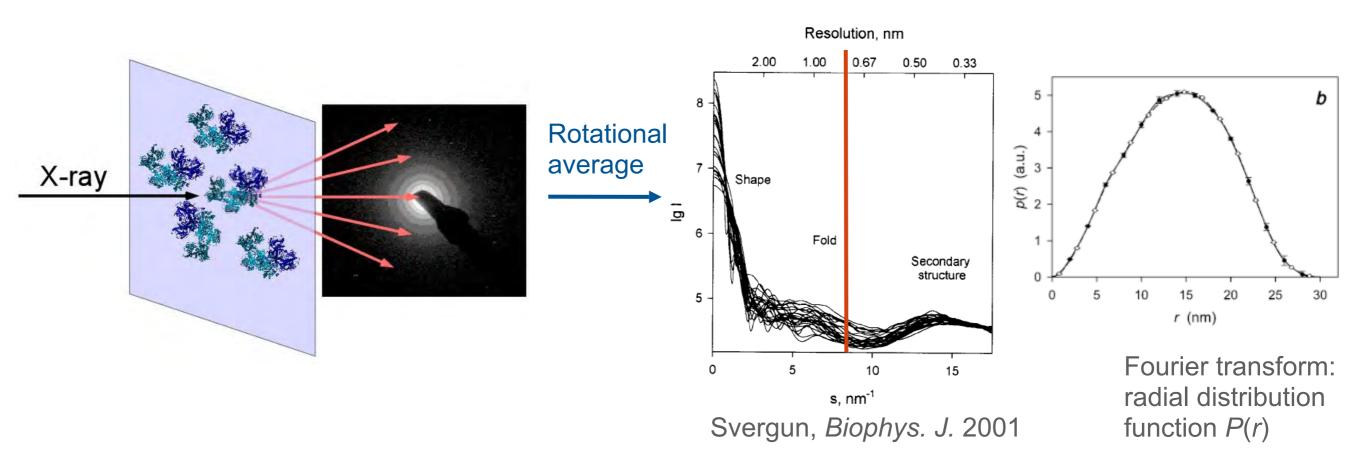
Summary

- MultiFit is method for simultaneous fitting of multiple components
- MultiFit can use near-native models of the components
- MultiFit provides a good starting model for higher resolution refinement methods
- Future work:
 - More robust discretization (anchor point computation)
 - More informative scoring function
 - Integration with flexible fitting methods

Topics

- 1. Introduction to integrative (hybrid) structure determination
- 2. Comparative model building
- 3. Predicting accuracy of atomic models
- 4. Iterative sequence-structure alignment and model building
- 5. Electron microscopy
- 6. Small angle x-ray scattering
- 7. Proteomics
- 8. Concluding Remarks

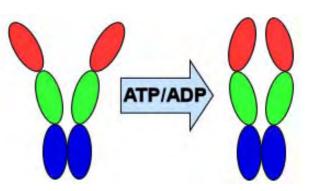
Small angle X-ray scattering (SAXS)



- Limited information content of a SAXS spectrum
- Integration with additional data
- Quaternary structure

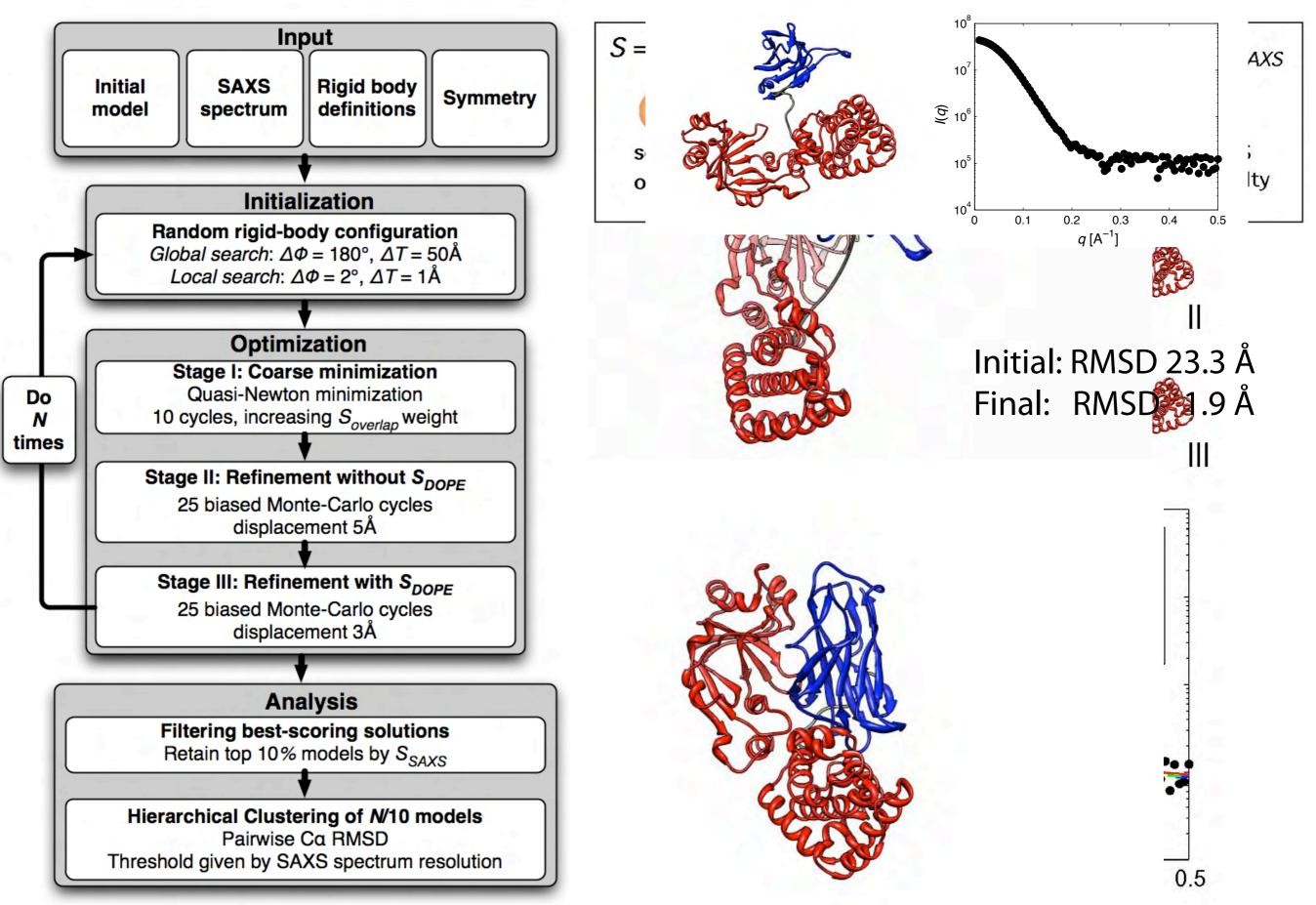


Changes in quaternary structure

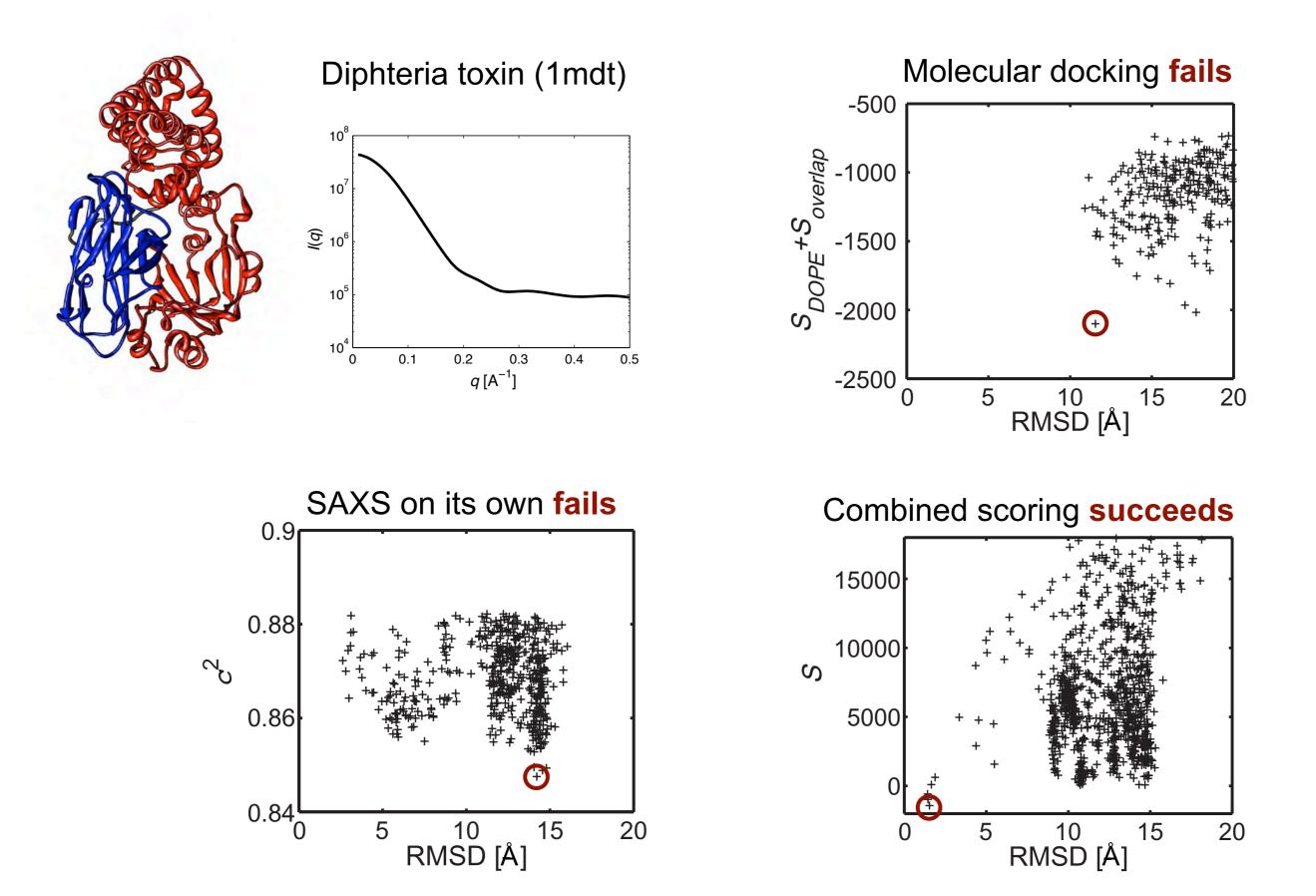


Protocol

F. Förster, B. Webb, K.A. Krukenberg, H. Tsuruta, D.A. Agard, A. Sali. J. Mol. Biol. 382, 1089-1106, 2008.



Benefit of integration of SAXS with modeling

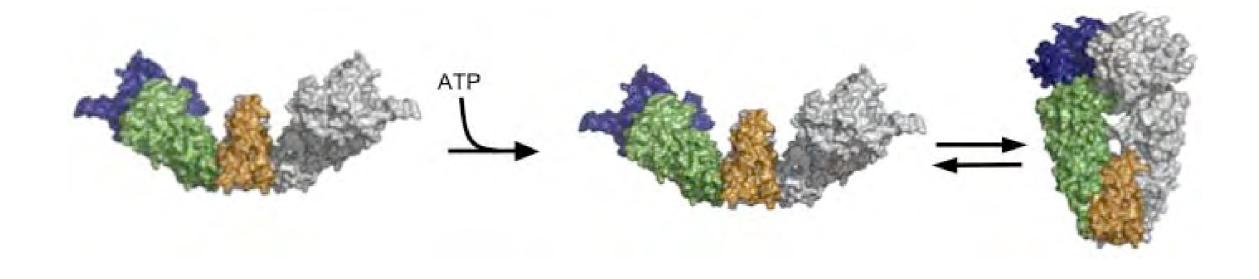


Summary of SAXS method

- Atomic models can be determined that are consistent with given SAXS data and additional restraints.
- Integration of information increases accuracy.
- Configurations can be sampled "exhaustively" for up to 4 domains.
- Configuration accuracy depends on rigid body accuracy (~3 Å Cα RMSD necessary).
- Integration of further information is possible.

SAXS maps Hsp90 states

K.A. Krukenberg, F. Förster, L. Rice, A. Sali, D.A. Agard, Structure 16, 755-765, 2008.

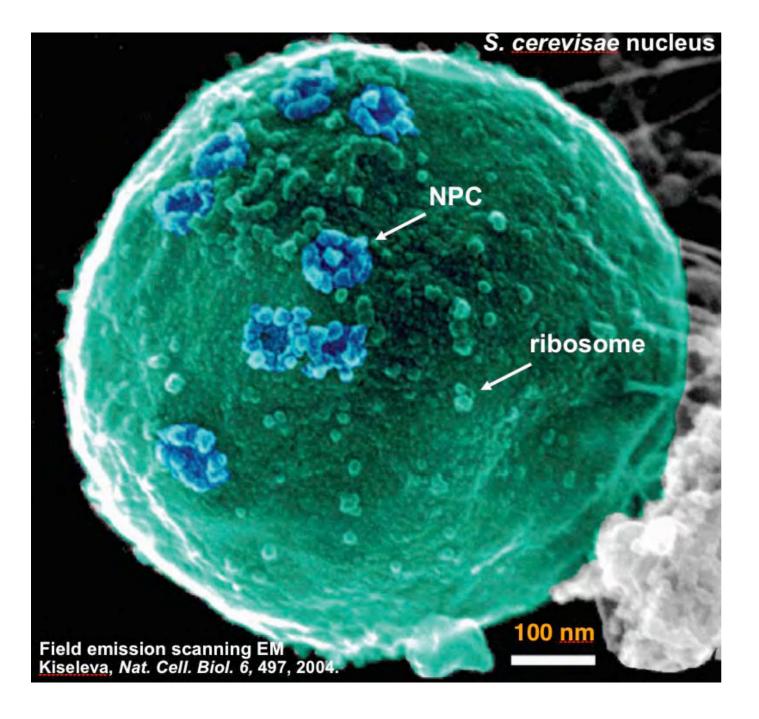


- Crystallographic structures of opened and closed states are probably inaccurate representations of solution states.
- The apo structure of *E. coli* Hsp90 is wide open.
- *E. coli* ATP-Hsp90 is in equilibrium between the wide-opened and closed states.

Topics

- 1. Introduction to integrative (hybrid) structure determination
- 2. Comparative model building
- 3. Predicting accuracy of atomic models
- 4. Iterative sequence-structure alignment and model building
- 5. Electron microscopy
- 6. Small angle x-ray scattering
- 7. Proteomics
- 8. Concluding Remarks

Nuclear Pore Complex (NPC)





Alber *et al.* Nature 450, 683-694, 2007 Alber *et al.* Nature 450, 695-701, 2007 Devos *et al.* PNAS 14, 2172-2177, 2006 Devos *et al.* PLoS Biology 12, 1-9, 2004



Consists of broadly conserved nucleoporins (nups).

50 MDa complex: ~480 proteins of 30 different types.

Mediates all known nuclear transport, via cognate transport factors.

Mike Rout

Svetlana Dokudovskaya, Liesbeth Veenhoff Orit Karni-Schmidt, Julia Kipper, Tari Suprapto, Julia Kipper

Brian Chait Wenzhu Zhang, Rosemary Williams

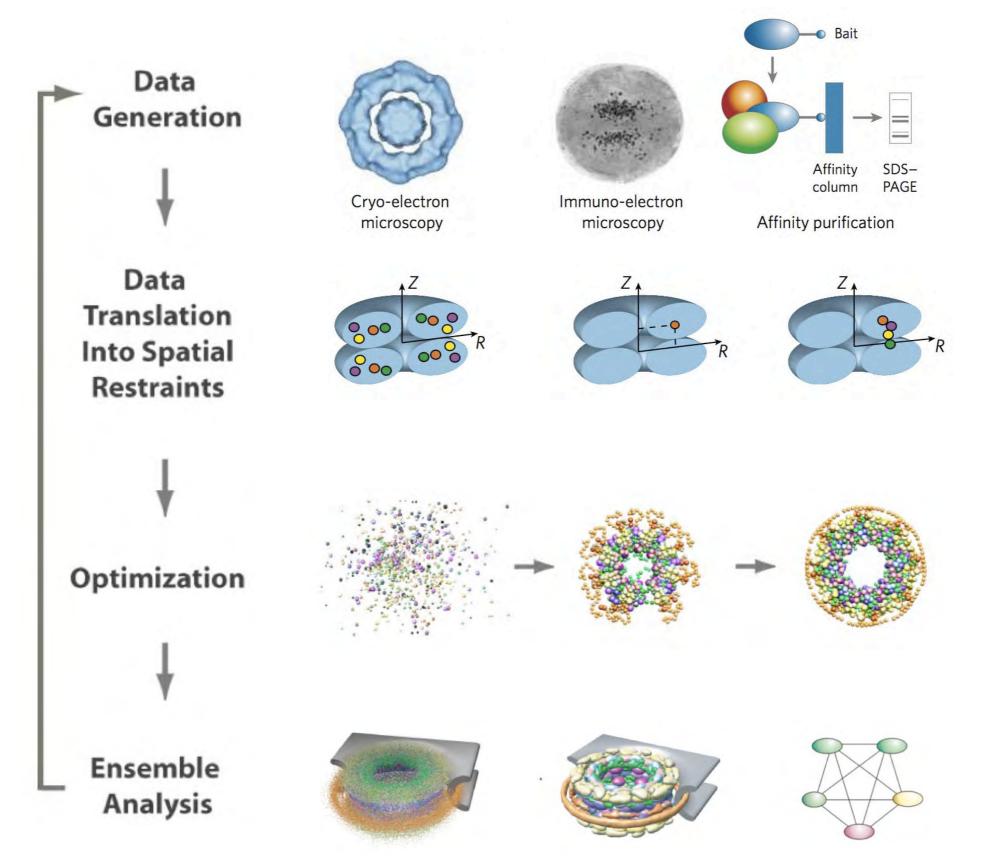
Rockefeller University

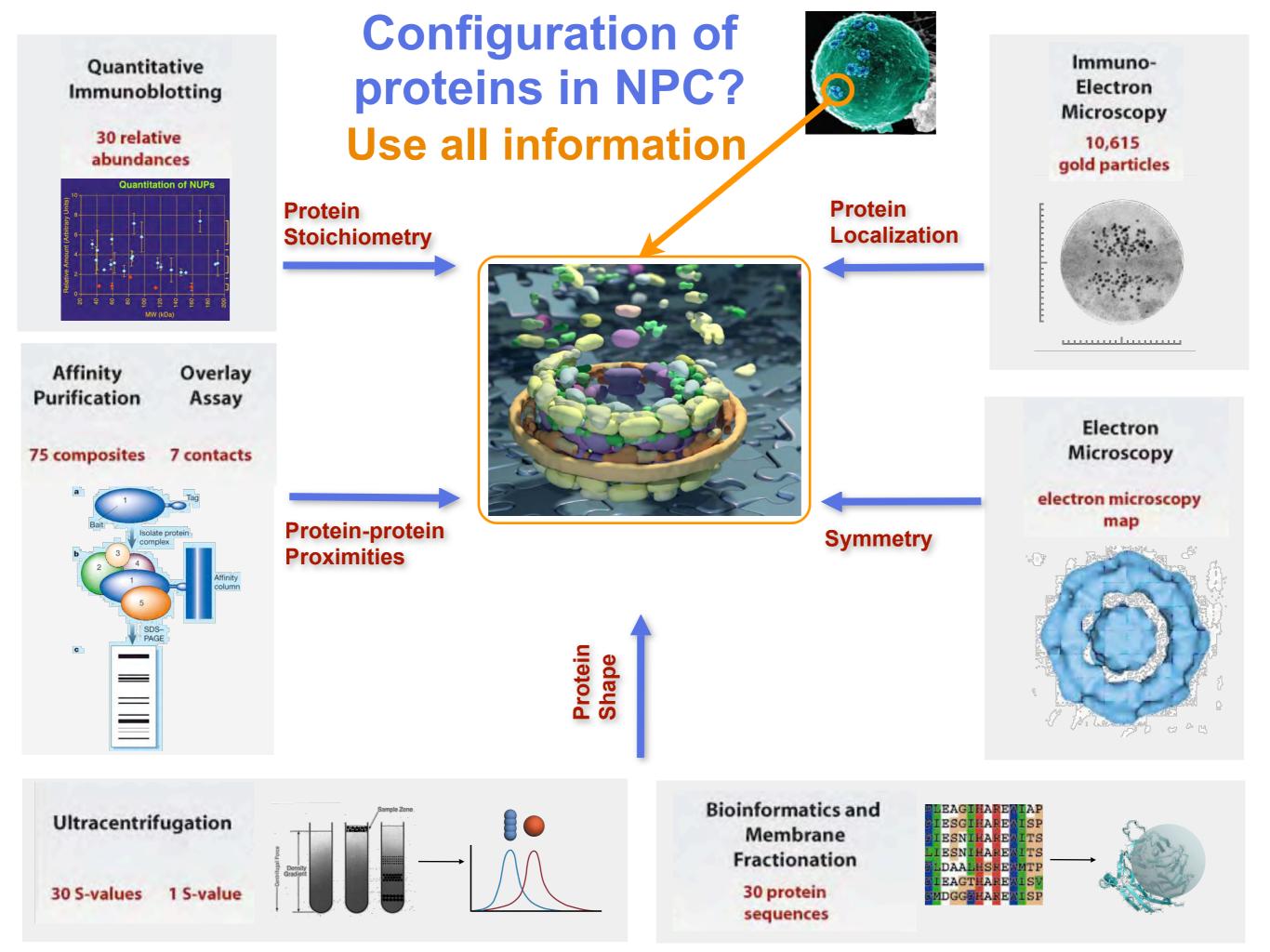
Andrej Sali Frank Alber, Damien Devos Narayanan Eswar, Marc Marti-Renom

UCSF

Using All Spatial Information

Alber *et al. Nature* 450, 683-694, 2007. Robinson, Sali, Baumeister. *Nature* 450, 974-982, 2007. Alber, Foerster, Korkin, Topf, Sali. *Annual Reviews in Biochemistry* 77, 11.1–11.35, 2008.





Optimization

- Start with a random configuration of protein centers.
- Minimize violations of input restraints by conjugate gradients and molecular dynamics with simulated annealing.
- Obtain an "ensemble" of many independently calculated models (~200,000).

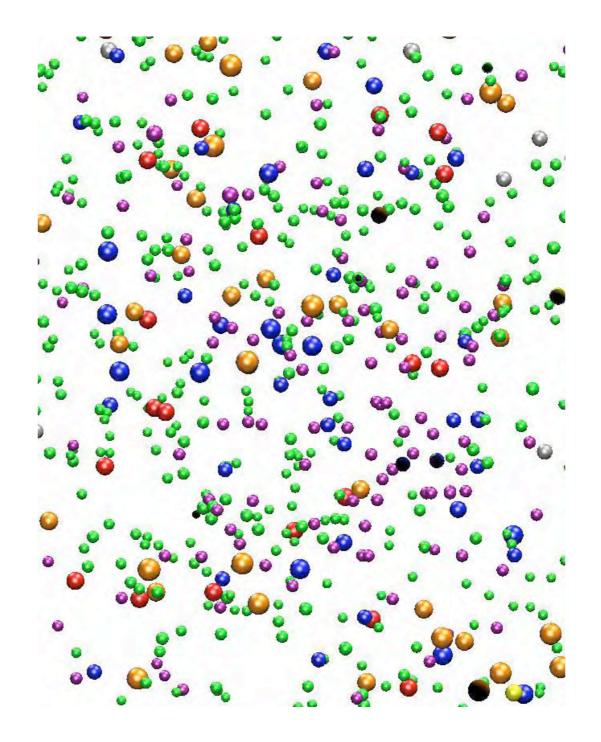
Membrane spanning proteins: Pom152 Pom34 Ndc1

FG repeat proteins:Nup159Nup60Nsp1Nup59Nup1Nup57Nup100Nup53Nup116Nup49Nup145NNup42

Nup84 complex: Nup84 Seh1 Nup85 Sec13 Nup120 Nup145C Nup133

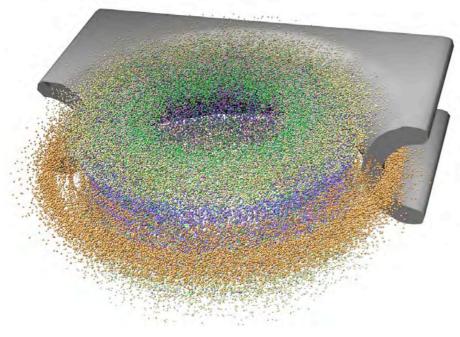
Large Core proteins: Nup192 Nup170 Nup188 Nup157

Nup82 Nic96

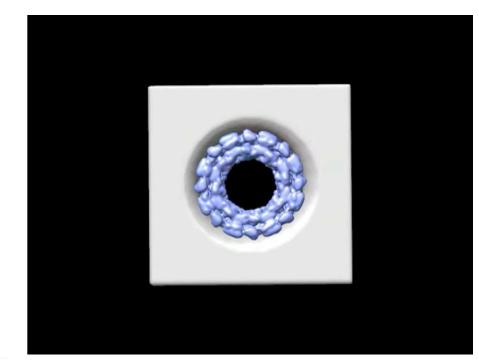


Protein Localization Probability and Volume

Calculated from the structural superposition of the ensemble of models that satisfy all input restraints

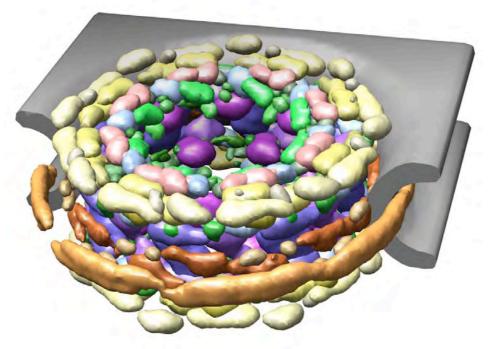


Ensemble of solutions



Animation

can see position of every NPC protein



Protein localization

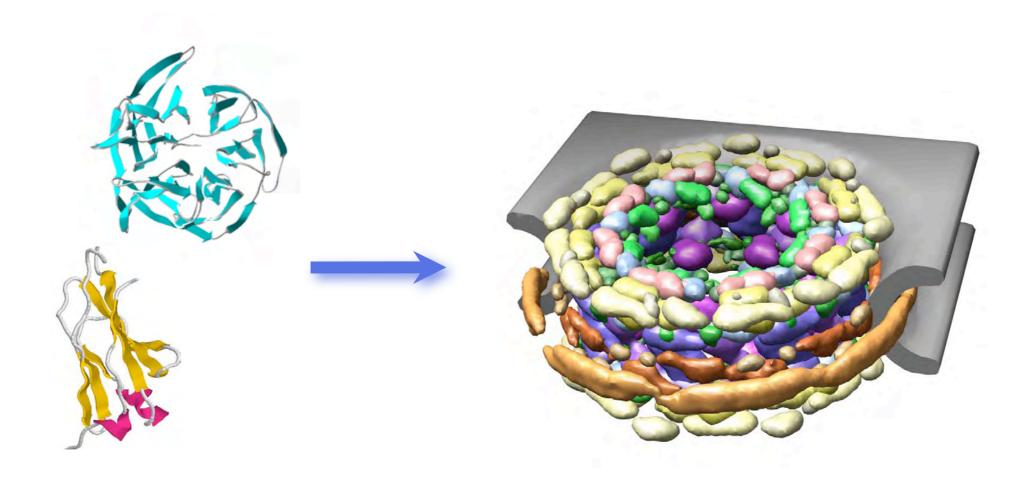
How accurate is the structure of the NPC? Assessing the well-scoring models

- 1. Self-consistency of independent experimental data.
- 2. Structural similarity among the configurations in the ensemble that satisfy the input restraints.
- 3. Simulations where a native structure is assumed, corresponding restraints simulated from it, and the resulting calculated structure compared with the assumed native structure.
- 4. Patterns emerging from a mapping of independent and unused data on the structure that are unlikely to occur by chance.
- 5. Experimental spatial data that were not used in the calculation of the structure.

Towards a higher resolution structure of NPC?

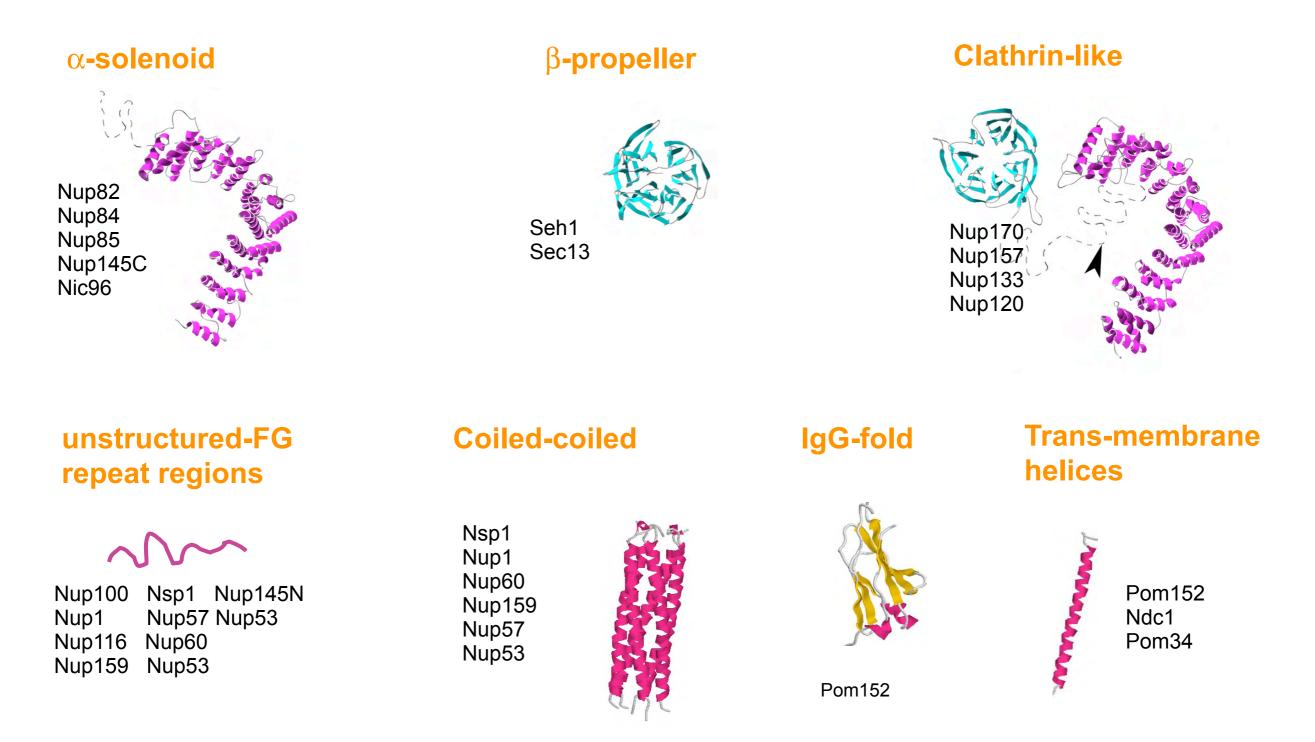
Characterize structures of the individual subunits, then fit them into the current low-resolution structure

(aided by cross-linking information and cryoEM maps of subcomplexes).



Fold Prediction

Devos, Dokudavskaya, Alber, Williams, Chait, Sali, Rout. *PLoS Biology* **12**, *1*, 2004 Devos, Dokudavskaya, Williams, Alber, Eswar, Chait, Rout, Sali, *PNAS* **14**, 2172, 2006.



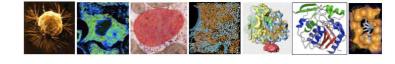
These fold assignments cover all 44 domains and 95% of the NPC residues.

In Conclusion

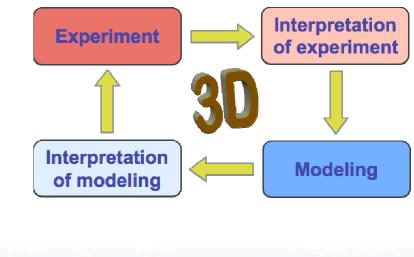
The goal is a comprehensive description of the multitude of interactions between molecular entities, which in turn is a prerequisite for the discovery of general structural principles that underlie all cellular processes.

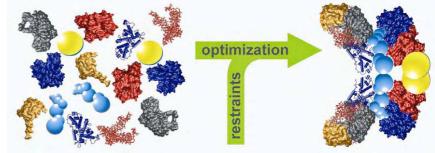


This goal will be achieved by a *tight* integration of **experiment**, **physics**, and **statistical inference**, spanning all relevant size and time scales.



		Ö		A REAL		
K-ray	NMR	2D & single particle	electron	immuno-	chemical	affinity purificat
rystallography	spectroscopy	electron microscopy	tomography	electron microscopy	cross-linking	mass spectrosco
subunit structure	subunit structure				subunit structure	
subunit shape	subunit shape	subunit shape	subunit shape			
subunit-subunit contact	subunit-subunit contact	subunit-subunit contact	subunit-subunit contact		subunit-subunit contact	subunit-subunit conta
subunit proximity	subunit proximity	subunit proximity	subunit proximity	subunit proximity	subunit proximity	subunit proximity
subunit stoichiometry	subunit stoichiometry					
assembly symmetry	assembly symmetry	assembly symmetry	assembly symmetry	assembly symmetry		
assembly shape assembly structure	assembly shape assembly structure	assembly shape	assembly shape			
Distance		DBD HIS3		MGFLIKRGFGHGARWTG		A ESCIA STRAAT
FRET	site-directed mutagenesis	yeast two-hybrid system	gene/protein arrays	protein structure prediction	computational docking	bioinformatics
				subunit structure		
autorell autorell anatorel				subunit shape		
subunit-subunit contact subunit proximity	subunit-subunit contact	subunit-subunit contact	subunit-subunit contact		subunit-subunit contact	subunit-subunit contac
Suburit proximity		subunit proximity	subunit proximity			





Sali, Earnest, Glaeser, Baumeister. From words to literature in structural proteomics. *Nature* 422, 216-225, 2003. Robinson, Sali, Baumeister. The molecular sociology of the cell. *Nature* 450, 974-982, 2007. Alber, Foerster, Korkin, Topf, Sali. *Annual Reviews in Biochemistry* 77, 11.1–11.35, 2008.

Acknowledgments http://salilab.org

Former Group Members

Roberto Sanchez (MSSM) Eric Feyfant (Wyeth) Francisco Melo (Catholic Univ.) Ilya Vakser (Kansas Univ.) Ash Stuart (Rampallo Coll.) Bino John (Pittsburgh Univ.) **Bozidar Yerkovich (Merck)** Valentin Ilyin (NE Univ.) Andras Fiser (AECOM) Nebojsa Mirkovic (Columbia U.) Maya Topf (Birkbeck College) Damien Devos (EMBL) Dmitry Korkin (UM,Columbia) Rachel Karchin (Johns Hopkins) Andrea Rossi (Pfizer) Bret Peterson (Google) M.S. Madhusudhan (Singapore) Fred Davis (HHMI) Frank Alber (USC) Marc Marti-Renom (Valencia) Friederich Foerster (MPI) Min-yi Shen (Appl.Biosystems) Ranyee Chiang (New York U.) Libusha Kelly (MIT) Mike Kim (The Mechanical Zoo) Mark Peterson (Boston Consulting) Narayanan Eswar (E. Lilly) David Eramian (UCSF)

Current Group Members

Ursula Pieper Ben Webb Keren Lasker Daniel Russell Hao Fan Javier Velazquez David Barkan Jeremy Phillips Adam Marko Dina Schneidman Avner Schlessinger Seung Joong Kim Mark Maupin



Collaborators

Tom Blundell (Cambridge U.) Brian Shoichet (UCSF) Robert Stroud (UCSF) Kathy Giacomini (UCSF) Tom Ferrin (UCSF) Patsy Babbitt (UCSF) Matt Jacobson (UCSF) David Agard (UCSF) Baldo Oliva (UB) Mike Rout (RU) Brian Chait (RU) David Stokes (NYSBC) Wolfgang Baumeister (MIPS) Wah Chiu (Baylor) Joachim Frank (Wadsworth) Chris Akey (BU) John Aitchison (ISB) Haim Wolfson (TAU)

NIH NSF The Sandler Family Foundation Human Frontiers Science Program IBM Intel Hewlett-Packard NetApps Structural Genomix Pharmaceuticals Mike Homer Ron Conway