## Integrative modeling of macromolecular structures and networks



Andrej Sali http://salilab.org/



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

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- 1. Introduction to integrative (hybrid) structure determination
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#### MOLECULAR STRUCTURE OF NUCLEIC ACIDS

J. D. WATSON F. H. C. CRICK ND. 4356 April 25, 1953 NATURE

#### A Structure for Deoxyribose Nucleic Acid



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THYMINE

ADENINE

#### A Structure for Deoxyribose Nucleic Acid



Stoichiometry Chemical complementarity

#### MOLECULAR STRUCTURE OF NUCLEIC ACIDS

J. D. WATSON F. H. C. CRICK

#### NO. 4356 April 25, 1953 NATURE

#### A Structure for Deoxyribose Nucleic Acid



To understand and modulate cellular processes, we need their models. These models are best generated by considering all available information.

## Towards a spatial, temporal, and logical model of the cell?





## Towards a spatial, temporal, and logical model of the cell?





## **Structural biology:** Maximize accuracy, resolution, completeness, and efficiency of the structural coverage of macromolecular assemblies

Motivation: Models 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

characterized, involved in a few hundred core biological processes.

## **Integrative Structural Biology**

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.



					Vilality V	
X-ray	NMR	2D & single particle	electron	immuno-	chemical	affinity purification
crystallography	spectroscopy	electron microscopy	tomography	electron microscopy	cross-linking	mass spectroscopy
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 contact
subunit proximity	subunit proternity	subunit proximity	subunit proximity	subunit proximity	subunit proximity	subunit proximity
assambly symmetry	assembly symmetry	assembly symmetry	assembly symmetry	assembly symmetry		
assembly shape	assembly shape	assembly shape	assembly shape	and a state of a state		
assembly structure	assembly structure					
FRET	site-directed	veast two-hybrid	gene/protein	MGFLIKRGFGHGARWTG.	computational	bioinformatics
	mutagenesis	system	arrays	prediction subunit structure subunit shape	docking	
subunit-subunit contact subunit proximity	subunit-subunit contact	subunit-subunit contact subunit proximity	subunit-subunit contact subunit proximity		subunit-subunit contact	Subunit-subunit contact
				1		

Sali A, Earnest T, Glaeser R, Baumeister W. From words to literature in structural proteomics. *Nature* 422, 216-225, 2003. Ward A, Sali A, Wilson I. Integrative structural biology. *Science* 339, 913-915, 2013.

## A description of integrative structure determination



While it may be hard to live with generalization, it is inconceivable to live without it. Peter Gay, Schnitzler's Century (2002).

## **Integrative structure determination**

- Uses multiple types of information (experiments, physical theory, statistical inference).
- Maximizes accuracy, resolution, completeness, and efficiency of the structure determination.
- Finds all models whose computed data match the experimental data within an acceptable threshold.



Sali *et al. Nature* **422**, 216-225, 2003. Alber *et al. Nature* **450**, 683-694, 2007 Robinson *et al. Nature* **450**, 974-982, 2007 Alber *et al. Ann.Rev.Biochem.* **77**, 11.1–11.35, 2008 Russel *et al. PLoS Biology* **10**, 2012 Ward *et al. Science* **339**, 913-915, 2013 Schneidman *et al. Curr.Opin.Str.Biol.*, 96-104, 2014. Sali *et al. Structure* **23**, 1156-1167, 2015.



A model is built iteratively, contributes continuously.

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Components

## **Configuration of 456 proteins** in the Nuclear Pore Complex



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## Integrative structure models from our lab





Ribosomes, Frank, Akey

PCSK9-Fab, Cheng, Agard, Pons

ab, Actin Agard, Chiu



TRiC/CCC Frydman, Chiu



RyR channel Serysheva, Chiu



Hsp90 landscape Agard

Nup133,

Rout, Chait



Substrate folding by Hsp90 Agard



Nuclear Pore Complex, Rout, Chait



Nuclear Pore Complex transport, Rout,Chait, Aitchison,Chook, Liphardt,Cowburn

Nup84 complex, Nup84 hub Rout, Chait Rout, Chait



Nup82 complex, Rout, Chait





SEA complex Rout, Chait, Dokudovskaya

PDE6 Chu



Spindle PoleBody Davis, Muller



Microtubule nucleation Agard



26 Proteasome Baumeister



PhoQ His kinase DeGrado



TFIIH Ranish





40S-elF1-elF3 Aebersold,Ban

Prion aggregation Prusiner

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### Integrative Modeling Platform (IMP) http://integrativemodeling.org



D. Russel, K. Lasker, B. Webb, J. Velazquez-Muriel, E. Tjioe, D. Schneidman, F. Alber, B. Peterson, A. Sali, PLoS Biol, 2012. R. Pellarin, M. Bonomi, B. Raveh, S. Calhoun, C. Greenberg, G.Dong, S.J. Kim, I. Chemmama, D. Saltzberg, S. Viswanath

Open source, versions, documentation, wiki, examples, mailing lists, unit testing, bug tracking, ...



Library of functional forms (ambiguity, ...)

## Integration across computational resources

**Experiment** 

Hypothesis Model

Goal: Maximize accuracy, resolution, completeness, and efficiency of the structural coverage of macromolecules





#### **Integrative Methods Task Force Workshop**





Andrej Sali, Helen M. Berman, Torsten Schwede, Jill Trewhella, Gerard Kleywegt, Stephen K. Burley, John Markley, Haruki Nakamura, Paul Adams, Alexandre Bonvin, Wah Chiu, Tom Ferrin, Kay Grünewald, Aleksandras Gutmanas, Richard Henderson, Gerhard Hummer, Kenji Iwasaki, Graham Johnson, Cathy Lawson, Frank di Maio, Jens Meiler, Marc Marti-Renom, Guy Montelione, Michael Nilges, Ruth Nussinov, Ardan Patwardhan, Matteo dal Peraro, Juri Rappsilber, Randy Read, Helen Saibil, Gunnar Schröder, Charles Schwieters, Claus Seidel, Dmitri Svergun, Maya Topf, Eldon Ulrich, Sameer Velankar, and John D. Westbrook. *Structure* **23**, 1156-1167, 2015.

First Integrative Methods Task Force Workshop was held at the European Bioinformatics Institute in Hinxton, UK, on October 6 and 7, 2014:

What should be archived?

How should integrative models be represented?

How should the data and integrative models be validated?

How should the data and models be archived?

What information should accompany the publication of integrative models?

Schneidman et al, COSB, 2014.

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  - 1. multi-scale models (atoms, unified atoms, secondary structure segments, protein domains described by points, spheres, ellipsoids, gaussians, ...).
  - 2. multi-state models (all states needed to explain the data)
  - 3. ordering of states in time (*eg*, a trajectory, functional cycle)
  - 4. ensemble of models (each model on its own explains the data; *eg*, NMR, SAXS)



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- 4. Uncertainty of the model coordinates should be explicitly considered.
- 5. Non-particle-based model representations (*eg*, continuum representations) need further consideration.



## Pushing the envelope of structural biology by integration of all available information

- Size
- Static systems in single and multiple states
- Dynamic systems
- Bulk and single molecule views
- Impure samples
- Overlapping with other domains such as systems biology















# Challenges in interpreting the data in terms of a structural model

- **1. Model representation**
- 2. Sampling
- 3. Scoring function:
  - Sparseness, due to incompleteness of measurements
  - Error, due to measurement and other imperfections
  - Ambiguity, due to, eg, multiple copies of a protein in a system
  - Incoherence (mixture), due to multiple states of a system in a heterogenous sample









## **Scoring function**

Rank models based on all available information:

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1. Least-squares scoring function:

$$S(M) = \sum_{i} w_i [D_i - f_i(M)]^2$$

- $M \mod$
- *D* measured data point
- f computed data point (forward model)
- w weight of data point

## **Scoring function**

#### Rank models based on all available information:

1. Least-squares scoring function:

$$S(M) = \sum_{i} w_i [D_i - f_i(M)]^2$$

2. Bayesian scoring function:

$$p(M|D,I) \propto p(D|M,I) \cdot p(M|I)$$
posterior
likelihood
prior
$$f$$

$$D - f(M)$$

- M model
- *D* measured data point
- f computed data point (forward model)
- w weight of data point
- *I* prior information

**Posterior** is the probability density of model *M*, given data *D* and information *I*.

Model *M* can include coordinates of one or more structures as well as additional parameters (noise levels, weights, calibration parameters, ...).

**Likelihood** is the probability density of observing data *D*, given model *M* and prior information *I* (by relying on a model of noise and a forward model, which computes data *D* given model *M*).

**Prior** is the probability density of model *M*, given prior information *I*.

 $p(AB) = p(BA) = p(A) \cdot p(B/A) = p(B) \cdot p(A/B)$ 

Rieping, Habeck, Nilges. Science, 2005

**Topics** 

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# Very low-resolution modeling of large assemblies

Many times the structures of some subunits are not available.

In such cases, we can only model the **configuration** of the subunits in the complex.



# **Nuclear Pore Complex (NPC)**



Consists of broadly conserved nucleoporins (nups). 50 MDa complex: ~480 proteins of 30 different types. Mediates all known nuclear transport, *via* cognate transport factors (karyoferins or kaps)

- 1. Structure
- 2. Evolution
- 3. Mechanism of transport
- 4. Mechanism of assembly
- 5. Interactions with other systems
- 6. Modulation and therapy

A large collaborative effort with Mike Rout and Brian Chait at Rockefeller University, also involving many other collaborators (Acknowledgments).

NCDIR National Center for Dynamic Interactome Research



## What was known about the NPC structure?





M. Beck, V. Lucic, F. Forster, W. Baumeister, O. Medalia Nature 449, 611–615 (2007).

R. Milligan, W. Baumeister, O. Medalia, G. Blobel,E. Hurt, U. Aebi, T. Schwartz, M. Stewart,C. Akey, B. Chait, M. Rout, ...

## An approach to integrative structural biology

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 Russel *et al. PLoS Biology* **10**, 2012



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## **Determination by experiment versus prediction by modeling**









# **Symmetry Restraints**



Yang, Rout, Akey, *Mol. Cell.* 1, 223, 1998.

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Yang, Rout, Akey, Mol. Cell. 1, 223, 1998.

half-spoke contains ~30 nucleoporin proteins (NUPs).

~480 NUPs in NPC.

# **Symmetry Restraints**



Yang, Rout, Akey, Mol. Cell. 1, 223, 1998.

half-spoke contains ~30 nucleoporin proteins (NUPs).

~480 NUPs in NPC.

Configurations in spokes and rings are restrained to be similar to each other *via* a DRMS-type restraint.

The same handedness of the half-spokes and rings is achieved *via* dihedral angle restraints on subsets of nucleoporins.

## Axial and Radial Localization Restraints on C-terminal Protein Beads



radial position



axial posiiton

## Axial and Radial Localization Restraints on C-terminal Protein Beads



axial posiition



axial posiiton

## Axial and Radial Localization Restraints on C-terminal Protein Beads



## Tagging, Affinity Purification and Analysis of Nucleoporin "Composites"



## Tagging, Affinity Purification and Analysis of Nucleoporin "Composites"



- several hundred "composites"
  - ~1,300 protein bands identified by MS





Sec13 Sec13 Sec13 Sec13 Sec13 Sec13

# Composites are informative structurally, but subject to assignment ambiguity



Alber *et al.* Nature 450, 683-694, 2007 Alber *et al.* Structure 13, 435-445, 2005

# Composites are informative structurally, but subject to assignment ambiguity



- A composite implies at least three direct protein interactions that connect all four protein types.
- But there is assignment ambiguity:
  - Which protein copies interact?
  - What domains interact?
- Many possible alternative restraint assignments are consistent with the composite data.



Alber *et al.* Nature 450, 683-694, 2007 Alber *et al.* Structure 13, 435-445, 2005

# 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



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## **Protein Localization Probability and Volume**

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



Ensemble of solutions

# **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.

## Assessment 3/5:

## Validation of the structure by a "simulated" model

- 1. Define a structure of the NPC as the native structure.
- 2. Simulate the restraints, given the native structure.
- 3. Calculate the structure based on the restraints.
- 4. Compare the calculated structure with the native one.

## Assessment 3/5:

## Validation of the structure by a "simulated" model

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## Assessment 4/5:

### Patterns that are unlikely to occur by chance

X. Zhou (USC): clustering of nucleoporin expression profiles



## Assessment 4/5:

## Patterns that are unlikely to occur by chance

# X. Zhou (USC): clustering of nucleoporin expression profiles





## **Assessment 5/5:**

# Experimental spatial data about the modeled structure that were not used in the calculation of the model



M. Lutzmann, R. Kunze, A. Buerer, U. Aebi & E. Hurt, EMBO J. 21, 387, 2002.

## Assessment 5/5:

# Experimental spatial data about the modeled structure that were not used in the calculation of the model



M. Lutzmann, R. Kunze, A. Buerer, U. Aebi & E. Hurt, EMBO J. 21, 387, 2002.



## Towards a higher resolution structure of the NPC

Characterize structures of the individual subunits, then fit them into the current low-resolution structure, aided by additional experimental information.





## Integrative structure determination of the Nup82 complex

#### Rout et al. Cell 2016, in press

#### **Experimental data**

## Statistical inference and physical principles



## 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 *formal* integration of **experiment**, **physics**, and **statistical inference**, spanning all relevant size and time scales.







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.

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#### Acknowledgements Integrative structural biology

#### Our group (current)

Seung Joong Kim Peter Cimermancic Barak Raveh Ben Webb Charles Greenberg Sara Calhoun Daniel Saltzberg Shruthi Viswanath Ilan Chemmama Seth Axen

#### **Our group (former)**

Dina Schneidman Elina Tjioe Daniel Russel GQ Dong Max Bonomi Riccardo Pellarin Frank Alber Bret Peterson Friedrich Foerster Mike Kim Maya Topf

#### Collaborators

Tom Ferrin (UCSF) Jeff Ranish (ISB) Chris Akey (BU) Steven Ludtke (Baylor) David Cowburn (AECOM) Steven Almo (AECOM) **David Stokes (NYU)** David Agard (UCSF) Nevan Krogan (UCSF) **Robert Stroud (UCSF)** Bill Degrado (UCSF) James Hurley (Berkeley) John Tainer (Berkeley) Trisha Davis (Univ of Wash) Mark Winey (U Colorado) Ivan Rayment (U Wisconsin) Wolgang Baumeister (MPI) Carol Robinson (Oxford) **Ruedi Aebersold (ETH)** Juri Rappsilber (Berlin) Al Burlingame (UCSF) Wah Chiu (Baylor) Joachim Frank (Columbia U) Ian Wilson (Scripps) Andrew Ward (Scripps) **Roger Kornberg (Stanford) Stanley Prusiner (UCSF)** Haim Wolfson (TAU) Michael Nilges (Pasteur) Nenad Ban (ETH) Jan Ellenberg (EMBL) Lan Huang (UCI) **Jonathan Baker (AECOM)** Ray Stevens (USC, iHuman) Pat Griffin (Scripps Florida) **Christine Humblet (Lilly)** Josh Salafsky (Biodesy) **Jaume Pons (Rinat)** 

...

#### wwPDB Hybrid/Integrative Methods Task Force

Helen Berman Jill Trewhella Stephen Burley Gerard Kleywegt

Torsten Schwede Frank Dimao Jens Meiler Gerhard Hummer



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