BIOL 555: Simulations of Macromolecular Structure

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Everything that living things do can be understood in terms of the jigglings and wigglings of atoms - Richard Feynman.

Role of molecular simulations

Combining molecular-simulation techniques with experiment accelerates fundamental as well as applied research, as molecular simulations

(a) provide mechanistic insights that complement experiment, and(b) serve as an inexpensive route to scan the role of multiple physical parameters and environmental variables.

Computational techniques have become an integral part of biological research and discovery:

- (a) Predicting bio-molecular (protein/RNA/lipid) structure
- (b) Refining X-ray, NMR and neutron diffraction data
- (c) Post-processing structural information
 - Interpret and analyze experimental data in terms of interactions at the atomic level
 - Understand fast time scale dynamics, such as during enzyme catalysis
- (d) Structure-based drug design

Hierarchy of molecular simulation techniques



Access timescales: integrate classical equations of motion

Molecular Dynamics: Integrate Newton's/Hamilton's equation of motion

$$m_i \frac{\partial^2 \boldsymbol{r}_i}{\partial t^2} = \boldsymbol{F}_i, \ i = 1 \dots N. \qquad \boldsymbol{F}_i = -\frac{\partial V}{\partial \boldsymbol{r}_i}$$

Langevin Dynamics: The 'unimportant' degrees of freedom are averagedout in such a way that the thermodynamic and long time-scale properties are preserved.The reduction of degrees of freedom depends on the problem one wishes to solve. The interactions change into potentials of mean force, and the omitted degrees of freedom are replaced by noise and friction.

$$\boldsymbol{F}_{i}(t) = -\frac{\partial V^{\mathrm{mf}}}{\partial \boldsymbol{r}'_{i}} + \boldsymbol{F}_{i}^{\mathrm{friction}} + \boldsymbol{F}_{i}(t)^{\mathrm{noise}}$$

Brownian Dynamics: Langevin dynamics at high friction.

Monte Carlo: are a class of computational algorithms that rely on repeated random sampling to compute their results

Energy Minimization: No time step. Move atoms so as to reduce the net forces (the gradients of potential energy) on the atoms until they become negligible.

Comparing molecular simulations against experiment





Pairwise additive assumption

$$\mathbf{v}(\mathbf{r}_1,\ldots\mathbf{r}_N) = \sum_{i< j} V_{ij}(\mathbf{r}_{ij});$$

$$\mathbf{F}_i = -\sum_j \frac{dV_{ij}(r_{ij})}{dr_{ij}} \frac{\mathbf{r}_{ij}}{r_{ij}} = -\mathbf{F}_j$$

Non-bonded interactions: Eg. Lennard-Jones interaction





$$V_c(r_{ij}) = f \frac{q_i q_j}{\varepsilon_r r_{ij}}$$

Bond stretching: Eg. Harmonic potential

$$V_b(r_{ij}) = \frac{1}{2}k^b_{ij}(r_{ij} - b_{ij})^2$$



Define the potential energy V in atomistic simulations

Angle stretching: Eg. Harmonic potential

$$V_a(\theta_{ijk}) = \frac{1}{2}k^{\theta}_{ijk}(\theta_{ijk} - \theta^0_{ijk})^2$$

Dihedral potential: Eg. Ryckaert-Bellemans function



Polarization effects: Eg. Thole formulation

$$V_{thole} = \frac{q_i q_j}{r_{ij}} \left[1 - \left(1 + \frac{\bar{r}_{ij}}{2} \right) \exp^{-\bar{r}_{ij}} \right]$$

...and a whole bunch of restraints can be added to address a given problem.

mole⁻¹) 8



270



Protein folding landscape



Critical Assessment of protein Structure Prediction (CASP)

http://www.predictioncenter.org

This protein folding challenge aims at establishing the current state of the art in protein structure prediction, identifying what progress has been made, and highlighting where future effort may be most productively focused.

CASP 9, 2010 results: 2 representative examples













Membrane structure prediction - atomistic simulations



Kanchanawong et al. Nature 2010

Structure/dynamics of ions in water: quantum mechanical simulations

AIMD trajectory generated using

PW91 functional implemented in VASP 4.2, NVE ensemble, PAW method, PME with background charge, Cut-off_{KE}= 500 eV

X-ray and NMR structure refinement

Vol 450 8 November 2007 doi:10.1038/nature06249

ARTICLES

nature

High-resolution structure prediction and the crystallographic phase problem

Bin Qian¹*, Srivatsan Raman¹*, Rhiju Das¹*, Philip Bradley¹, Airlie J. McCoy², Randy J. Read² & David Baker¹

The energy-based refinement of low-resolution protein structure models to atomic-level accuracy is a major challenge for computational structural biology. Here we describe a new approach to refining protein structure models that focuses sampling in regions most likely to contain errors while allowing the whole structure to relax in a physically realistic all-atom force field. In applications to models produced using nuclear magnetic resonance data and to comparative models based on distant structural homologues, the method can significantly improve the accuracy of the structures in terms of both the backbone conformations and the placement of core side chains. Furthermore, the resulting models satisfy a particularly stringent test: they provide significantly better solutions to the X-ray crystallographic phase problem in molecular replacement trials. Finally, we show that all-atom refinement can produce *de novo* protein structure predictions that reach the high accuracy required for molecular replacement without any experimental phase information and in the absence of templates suitable for molecular replacement from the Protein Data Bank. These results suggest that the combination of high-resolution structure prediction with state-of-the-art phasing tools may be unexpectedly powerful in phasing crystallographic data for which molecular replacement is hindered by the absence of sufficiently accurate previous models.

X-ray and NMR structure refinement contd.

a

Acyl CoA binding protein

Superposed crystal structure (blue), NMR model (red) and the lowest energy all-atom refined model (green)

Solving the X-ray crystallographic phase problem via molecular replacement: Improvement in electron density using model from rebuilding and refinement in molecular replacement searches.

Structure refinement

How does cPLA₂ propagate the Inflammation Signaling Pathway? Ligand Ca C2a Domain of cPLA2 Cytoplasmic Membrane Channel Receptor Intracellular Membrane PLC Ca Y B Ga cPLA2 C2 Lipase DAG IP3 ററ ER Arachadonic Acid COX Aspirin, Ibuproten cPLA2 Ca MAPK PKC C2 Lipase Prostaglandins Inflammation

X-Ray Vs NMR

Used a combined experimental and computational approach to understand the

differences in secondary structure

Membrane docking region

X-ray Structure

O. Perisic et al., J Biol. Chem. 1998

G. Xu et al., J Mol. Biol. 1998

Malmberg, Varma, Jakobsson & Falke, Biochemistry. 2004 Varma & Jakobsson, Biophys. J. 2007

Post processing structural information

Collagen

Collagen cont.

Collagen is the main protein in our connective tissue What gives collagen its strength?

Cartoon: adapted from Gauitieri et al. 2010

Collagen cont.

Fluctuations of isolated collagen molecule in 0.2 M NaCl solution

Selectively regulate K⁺ concentration gradients across cell membranes to enable numerous physiological tasks, including nerve conduction and muscle contraction What is the basis for K⁺/Na⁺ selectivity?

Cartoon taken from: http://nobel.se

Problems with the conventional picture

Thermal fluctuations obscure sub-angstrom size differences between Na⁺ and K⁺ ions

New picture of K⁺ in water

Quantum mechanical simulation

The probability to find an 8-fold coordination is negligible in liquid water

The binding sites in K-channels

(1) DO NOT mimic the structure of K⁺ ions in bulk water and, in fact,

(2) over-coordinate the K⁺ ion.

Varma & Rempe, Biophys. Chem., 2006

Noskov et al., Nature 2004

Problems with the conventional picture

K-channel filter adopts multiple configurations during ion binding

NaK channels

share an overall architecture with Kchannels but are non-selective

> X-ray structure: Jiang and coworkers

Valinomycin

does not share an overall architecture with K-channels, but is as selective as Kchannels

Different mechanisms of K+/Na+ selectivity

Different mechanisms of K+/Na+ selectivity

Computer-aided drug design

2. Selecting representative compounds for in vivo test