

GENESIS Hands-on Part 2-1

Coarse-Grained Simulations in GENESIS

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2024/07/01

IUPAB2024 Hands-on Training Program | CHARMM-GUI/GENESIS MD Tutorial

Schedule of GENESIS Tutorials (6/30-7/2)

13:30 - 15:00



06/30 Day 1	
13:30 - 15:00	GENESIS basics and GENESIS on Fugaku (Kobayashi)
	Lecture
	Hands-on tutorial on Fugaku
07/01 Day 2	
14:30 – 15:30	Coarse-grained simulations in GENESIS (Tan)
15:30 – 16:30	High-performance computation with GENESIS (Jung)
07/02 Day 3	

Generalized-ensemble simulations using GENESIS (Ito)

Table of Contents



- Part 2-1-I: Coarse-Grained (CG) Models in GENESIS
 - Introduction to CG
 - Popular CG Models
 - CG MD in GENESIS

Part 2-1-II: GENESIS CG Simulations Hands-on

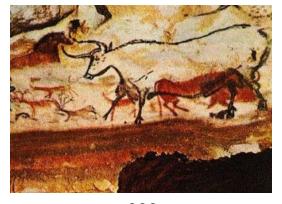
- GENESIS-cg-tool installation
- Protein folding simulation
- Simple CG MD data analysis

CG: Fewer Details in Picturing Molecules



ROM THE PREHISTORIC DRAWINGS ON THE CAVE WALLS OF Lascaux to T the paintings hanging in the Louvre in the late nineteenth century, the visual arts seemed to progress monotonically by adding ever more detail seemingly to give greater realism. A new movement then exploded near the turn of the twentieth century. Artists began to explore new ways of expressing truths about the visual world and the visual experience. These new ways involved generally more abstract and superficially less detailed techniques than those taught in the art schools. They required no less intellectual discipline, however. The new "modern art," while not to everyone's taste, nevertheless, in the opinion of many, brought into existence creations of great beauty and insight. Peter Wolynes, 2018

C-terminus C-terminus Coarse-graining N-terminus N-terminus

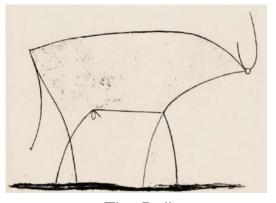


??? Anonymous, ~15000 BC



「Coarse-Grained Modeling of Biomolecules」

The Young Bull Paulus Potter, 1647



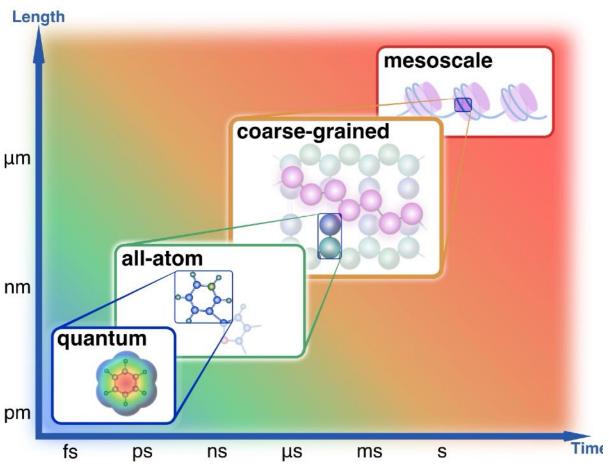
The Bull Pablo Picasso, 1945



coarse-grained model

Task of MD: numerically solving Newton's equations of motion





degree of freedom energy landscape computational speed

all-atom model

Classification of CG Models



Coarse-graining approaches:

- Bottom-up
- Top-down

Modeling strategies:

- Dynamics-based (physics-based)
 Systematic algorithms of AA -> CG mapping
 - Boltzmann Inversion
 - o Inverse Monte Carlo
- Knowledge-based
 Fitting experimental macroscopic properties
 - Miyazawa-Jernigan
 - Kim-Hummer
- Structure-based

Assumption of consistency principle

- Network models
- Gō models

Resolution

amino acid

per

beads

Rosetta: all-atom bb, C_{β} , sc. Baker PRIMO: 3-bead bb, 1~5-bead sc. Feig

OPEP: all-atom bb, one-bead sc. Derreumaux

MARTINI: 2~4-atom/bead. Marrink

CABS: C_{α} , C_{β} , sc, center-of-bond. Kolinski

AWSEM: C_{α} , C_{β} , O. Wolynes Papoian

UNRES: C_{α} , sc. Sheraga

Kim-Hummer: C_{α} . 1 HPS, Mpipi: C_{α} . Gō: C_{α} .

Residue-level CG

Hummer Mittal

Collepard-Guevara

Clementi Brooks Takada

AJ Pak & GA Voth, Curr. Opin. Struct. Biol. 2018.

Residue-Level CG Models



Residue-level coarse-graining: ~10 atoms / CG particle

Protein: AICG2+

W. Li et al. 2014, PNAS.

DNA: 3SPN.2C

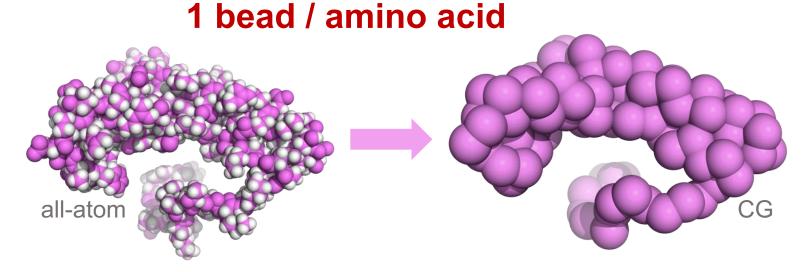
G. Freeman et al. 2014, JCP.

RNA: Gō-like

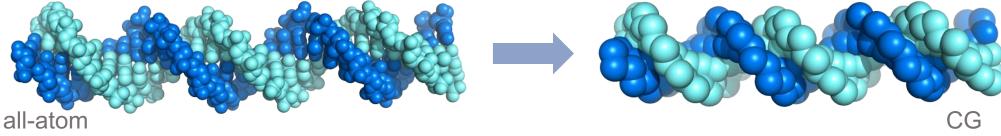
N. Hori et al. 2012, JCTC.

IDP: HPS

Dignon et al. 2018 PLoS Comput. Biol.







• Protein-DNA: PWMcos C.Tan, S. Takada, 2018, JCTC.

AICG2+ Protein Model



Atomic Interaction-based Coarse-Grained (AICG) model

$$\begin{split} V_{AICG2+}(\mathbf{R}|\mathbf{R}_{0}) &= \sum_{ibd} K_{b,ibd} (b_{ibd} - b_{ibd,0})^{2} + V_{loc}^{flp} \\ &+ \sum_{j=i+2} \varepsilon_{loc,ij} \exp\left(-\frac{(r_{ij} - r_{ij0})^{2}}{2W_{ij}^{2}}\right) + \sum_{j=i+3} \varepsilon_{loc,ij} \exp\left(-\frac{(\phi_{ij} - \phi_{ij0})^{2}}{2W_{\phi,ij}^{2}}\right) \\ &+ \sum_{i < j-3}^{nat \ contact} \varepsilon_{go,ij} \left[5\left(\frac{r_{ij0}}{r_{ij}}\right)^{12} - 6\left(\frac{r_{ij0}}{r_{ij}}\right)^{10}\right] \\ &+ \sum_{i < j-3}^{non-native} \varepsilon_{exv} \left(\frac{\sigma_{ij}}{r_{ij}}\right)^{12} \ . \end{split}$$

Bonded interactions

- Bond length (harmonic, structure based)
- 1-3 distance (Gaussian, physics + structure based)
- Dihedral angle (Gaussian, physics + structure based)
- Flexible local potentials (Boltzmann inv., statistical)
 - Angle
 - Dihedral

Nonbonded interactions:

- Native contact (12-10)
 - r_0 : native-structure based
 - ε_{qo} : physics + structure based
- Nonnative contact (12-repulsion)
 - σ_{ij} : knowledge based

 ε_{loc} for 1-3 distance and dihedral angle, and ε_{go} :

- Energy decomposition from atomistic simulations
- Count hydrogen-bond numbers

3SPN.2C DNA Model



3-Site-Per-Nucleotide (3SPN) model for double-stranded DNA (dsDNA)

$$U_b = U_{bond} + U_{ang} + U_{dih}.$$

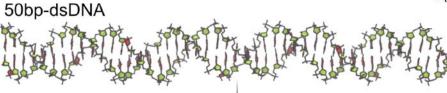
Bonded

$$U_{bond} = \sum_{i}^{bonds} k_b (r_i - r_{i,0})^2 + 100 k_b (r_i - r_{i,0})^4.$$

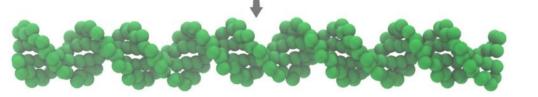
$$U_{ang} = \sum_{i}^{angles} k_{ heta} (heta_i - heta_{i,0})^2.$$

$$U_{dih,Gaussian} = \sum_{i}^{dihedrals} -k_{\phi,Gaussian} \exp\left(\frac{-(\phi_i - \phi_{i,0})^2}{2\sigma_{\phi}^2}\right). \quad U_{cstk} = \sum_{i}^{n_{cstk}} f(K_{BP}, \Delta\theta_3) f(K_{CS}, \Delta\theta_{CS}) U_m^{attr}(\epsilon_{CS}, \alpha_{CS}, r_{kl}).$$

$$U_{dih,periodic} = \sum_{i}^{dihedrals} k_{\phi,periodic} igl[1 + \cos(\phi_i - \phi_{i,0}) igr].$$



Coarse-graining



$$U_{nb} = U_{bstk} + U_{bp} + U_{cstk} + U_{exv} + U_{ele}$$
. Nonbonded

$$U_{bstk} = \sum_{m}^{n_{bstk}} U_m^{rep}(\epsilon_{BS}, \alpha_{BS}, r_{ij}) + f(K_{BS}, \Delta\theta_{BS}) U_m^{attr}(\epsilon_{BS}, \alpha_{BS}, r_{ij}).$$

$$U_{bp} = \sum_{m}^{n_{bp}} U_{m}^{rep}(\epsilon_{BP}, \alpha_{BP}, r_{ij}) + \frac{1}{2} (1 + \cos(\Delta\phi_{1})) f(K_{BP}, \Delta\theta_{1}) f(K_{BP}, \Delta\theta_{2}) U_{m}^{attr}(\epsilon_{BP}, \alpha_{BP}, r_{ij}).$$
5'

$$U_{cstk} = \sum_{m}^{n_{cstk}} f(K_{BP}, \Delta\theta_3) f(K_{CS}, \Delta\theta_{CS}) U_m^{attr}(\epsilon_{CS}, \alpha_{CS}, r_{kl})$$

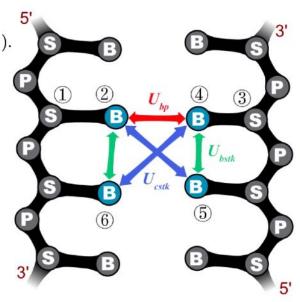
$$U_{exv} = \sum_{i < j} \begin{cases} \epsilon_r \left[\left(\frac{\sigma_{ij}}{r_{ij}} \right)^{12} - 2 \left(\frac{\sigma_{ij}}{r_{ij}} \right)^6 \right] + \epsilon_r & r_{ij} < \sigma_{ij}, \\ 0 & r_{ij} > \sigma_{ij}. \end{cases}$$

$$U_{ele} = \sum_{i < j}^{n_{ele}} \frac{q_i q_j e^{-r_{ij}/\lambda_D}}{4\pi\epsilon_0 \epsilon(T, C) r_{ij}}.$$

Fitted experimental properties:

- Persistence length
- Melting temperature
- Hybridization rate constant
- **Sequence-dependent curvature**
- Local base-step flexibility

DM Hinckley et al., JCP 2013 JS Freeman et al., JCP 2014



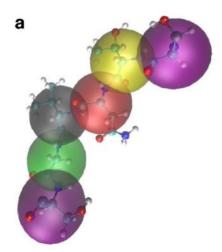
Weak points:

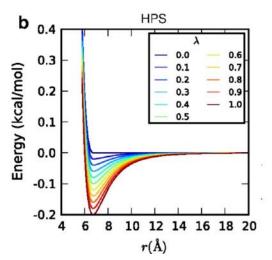
- High ionic concentration
- Single-stranded DNA

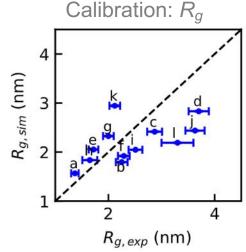
HPS/KH IDP Model



Hydrophobicity-scale (HPS) and Kim-Hummer (KH) model for Intrinsically Disordered Protein (IDP)







Dignon et al. 2018 PLoS Comput. Biol.

Variations:

Regi et al. 2021 Proteins
Dannenhoffer-Lafage et al. 2021 JPCB
Tesei et al. 2021 PNAS
Regi et al. 2020 NAR (protein and RNA)

Electrostatic interaction

$$E_{ij}(r) = rac{q_i q_j}{4\pi D r} \exp(-r/\kappa),$$

HPS potential

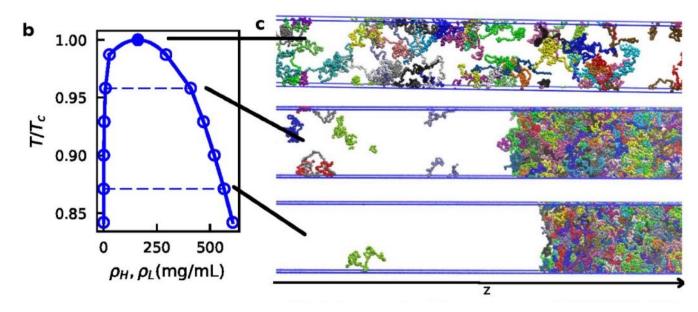
Ashbaugh-Hatch functional form

$$\Phi(r) = egin{cases} \Phi_{{\scriptscriptstyle L}{\scriptscriptstyle J}} + (1-\lambda)\epsilon, & ext{if} \ \ r \leq 2^{1/6}\sigma \ \ \lambda \Phi_{{\scriptscriptstyle L}{\scriptscriptstyle J}}, & ext{otherwise} \end{cases}$$

Lennard-Jones potential

$$\Phi_{{\scriptscriptstyle L}{\scriptscriptstyle J}} = 4\epsilonigg[\Big(rac{\sigma}{r}\Big)^{{\scriptscriptstyle 12}} - \Big(rac{\sigma}{r}\Big)^{{\scriptscriptstyle 6}}igg].$$

Slab method: simulating multi-chain systems

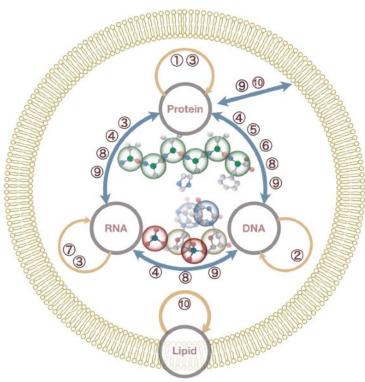


CG Models and MD in GENESIS

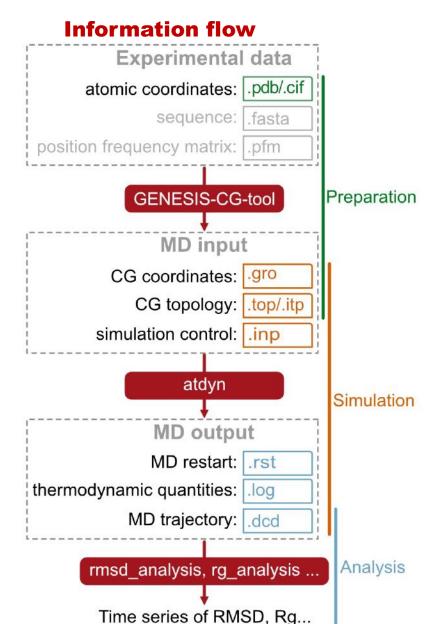


C.Tan et al. 2022, PLoS Comput. Biol.

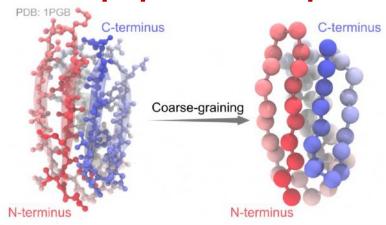
Available Models



- 1. W. Li et al. (2014). Proc. Natl. Acad. Sci.
- 2. G.S. Freeman et al. (2014). J. Chem. Phys.
- 3. G.L. Dignon et al. (2018). PLoS Comput. Biol.
- 4. C. Clementi et al. (2000). J. Mol. Biol.
- 5. C. Tan & S. Takada (2018). J. Chem. Theory Comput.
- 6. G.B. Brandani et al. (2018) Nucl. Acids Res.
- 7. N. Hori. & S. Takada (2012). J. Chem. Theory Comput.
- 8. C. Tan & S. Takada (2016). J. Am. Chem. Soc.
- 9. P. Debye & E. Hückel (1923). Physikalische Zeitschrift
- 10. D Ugarte La Torre et al. (2023). J. Chem. Phys.



One-line preparation example



```
$ ls
PR01.pdb
 genesis_cg_tool/src/aa_2_cg.jl PR01.pdb
$ ls
```

GENESIS-cg-tool: (in Julia)

https://github.com/genesis-release-r-ccs/genesis_cg_tool

Online wiki:

https://github.com/genesis-release-r-ccs/genesis_cg_tool/wiki

GENESIS source code + manual:

• https://www.r-ccs.riken.jp/labs/cbrt/genesis-version-2-1/

Tutorials:

https://r-ccs.riken.jp/labs/cbrt/tutorials2022/

GENESIS Online Tutorials Sec. 11



https://www.r-ccs.riken.jp/labs/cbrt/tutorials2022/

C.Tan et al. 2022, PLoS Comput. Biol.

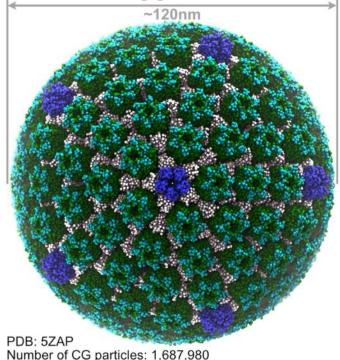
- 11. Advanced MD simulations with the coarse-grained model
 - 11.1 Coarse-grained simulation of protein with AICG2+ model
 - 11.2 Coarse-grained simulation of double-stranded DNA with 3PSC.N model
 - 11.3 Coarse-grained simulation of protein-DNA interactions with PWMcos model

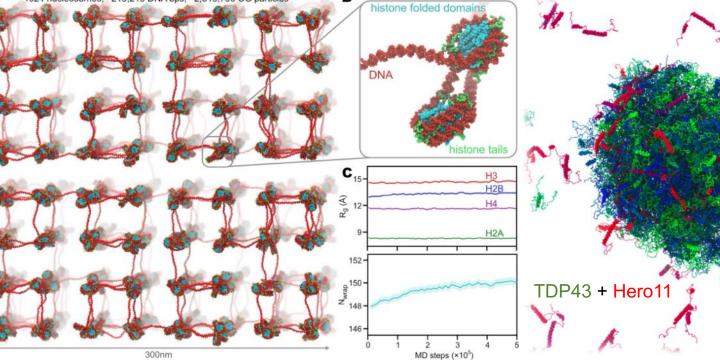
1024 nucleosomes: 219.213 DNA bps: 2.318.796 CG particles

11.4 Coarse-grained simulation of FUS condensation with HPS model

- Protein folding
- DNA dynamics
- Protein-DNA recognition
- Phase behaviors of IDP

Possible applications





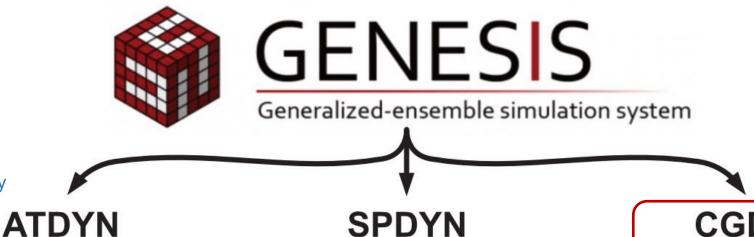
Other possible applications: C.Tan et al. JACS 2016; PNAS 2020.

C.Tan, A. Niitsu, Y. Sugita, JACS Au, 2023.

Extra: High-Performance CG MD

t=9×106steps





Atomic decomposition dynamics

SPDYN

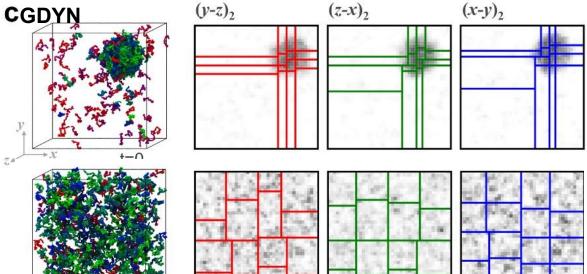
Spatial decomposition dynamics Load imbalance

CGDYN beta version

Memory limit

Dynamic domain decomposition scheme to address the nonuniform distribution

This CG tutorial only applies to ATDYN!





CG molecular **dyn**amics

Dr. Jaewoon Jung

Jung, Tan, Sugita, Nat. Commun. 2024