



# Ab initio characterization of protein molecular dynamics with **AI<sup>2</sup>BMD**

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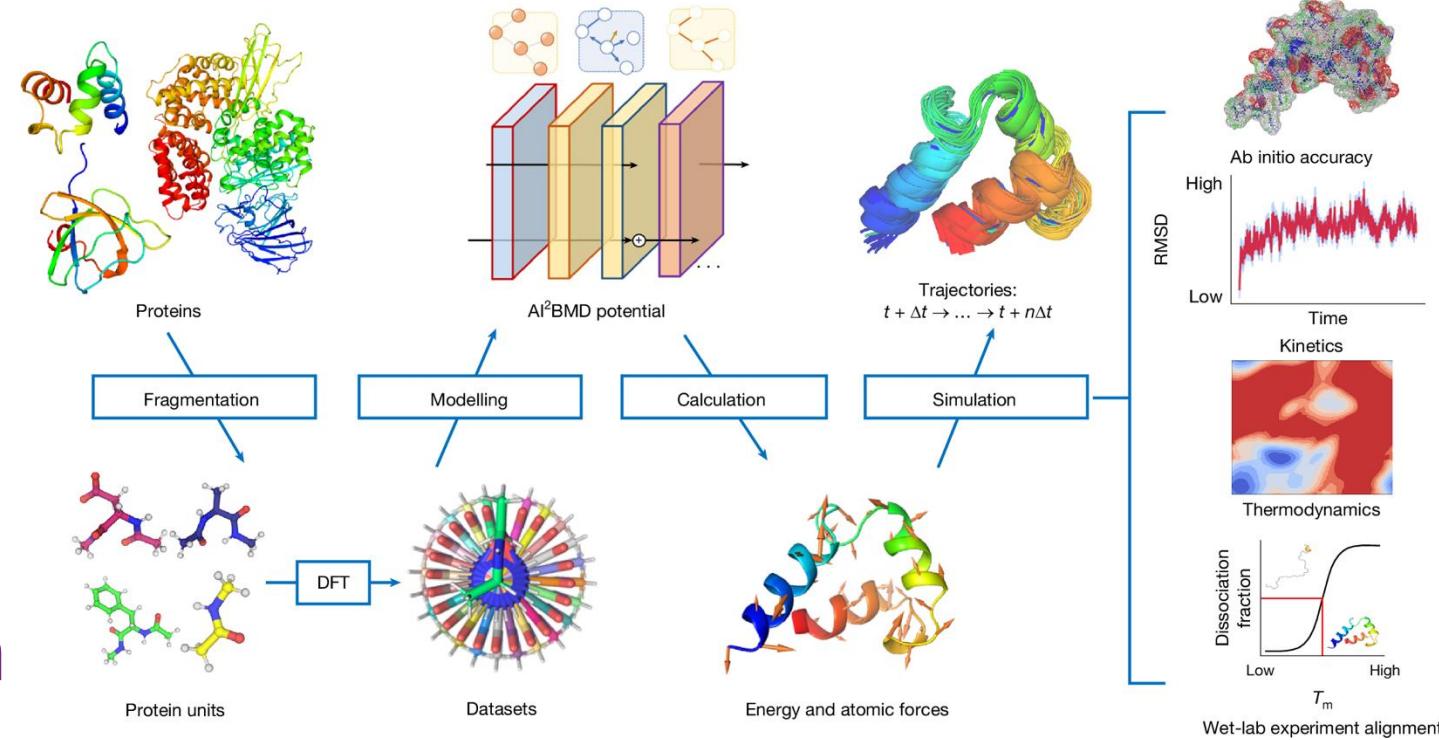
# Outline

## 1. Background

## 2. AI<sup>2</sup>BMD workflow

## 3. Results and discussion

## 4. Conclusion and perspectives





# Research Background

- As computational simulations now achieve near-experimental accuracy, they are transitioning from auxiliary interpretative tools to a central paradigm for discovery in the life sciences.
- Molecular dynamics (MD) plays an essential role:
  - classical MD
  - ab initio MD (AIMD)
  - machine learning force field (MLFFs)
- Key challenges for biomolecular simulations:
  - Diversity of conformational space
  - Data scarcity due to time and cost of dataset generalization

Accurate but face scalability challenges

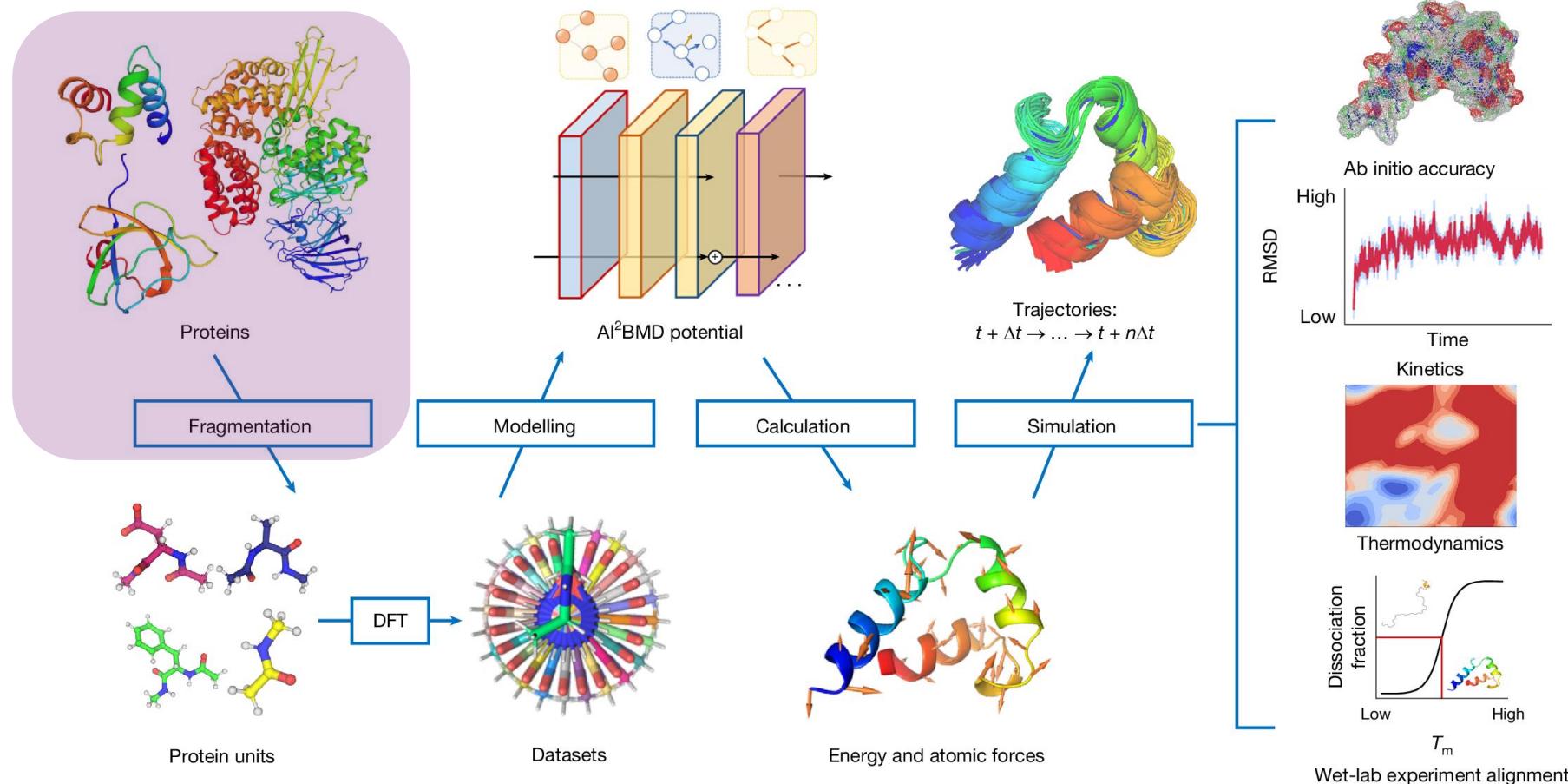
## AI<sup>2</sup>BMD Artificial intelligence-based *ab initio* biomolecular dynamics system

A generalizable solution for efficiently simulating a wide range of full-atom proteins with *ab initio* accuracy, surrounded by an explicit solvent modelled by a polarizable force field.

- A generalizable protein fragmentation approach splits proteins into overlapped protein units.
- Based on ViSNet<sup>1</sup>, calculates the energy and atomic forces for the protein with *ab initio* accuracy.
- Exhibits good alignment with wet-lab experimental data, such as the melting temperature of fast-folding proteins.
- Detects different phenomena than molecular mechanics (MM)

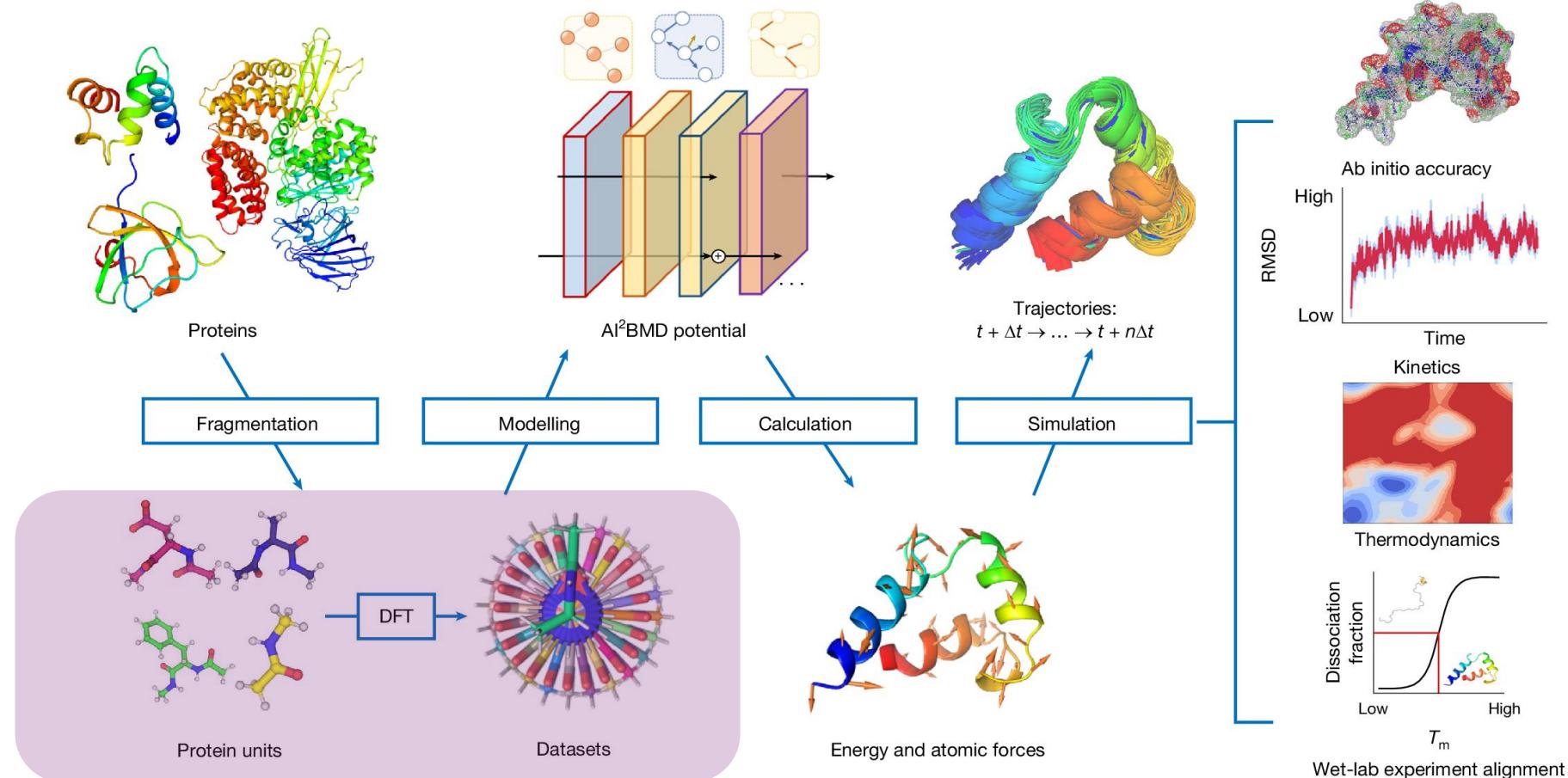
# AI<sup>2</sup>BMD workflow

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1. Fragmenting proteins into smaller units, specifically dipeptides, calculate intra- and inter-unit interactions.
2. Assemble them to determine the protein energy and forces acting on the atoms

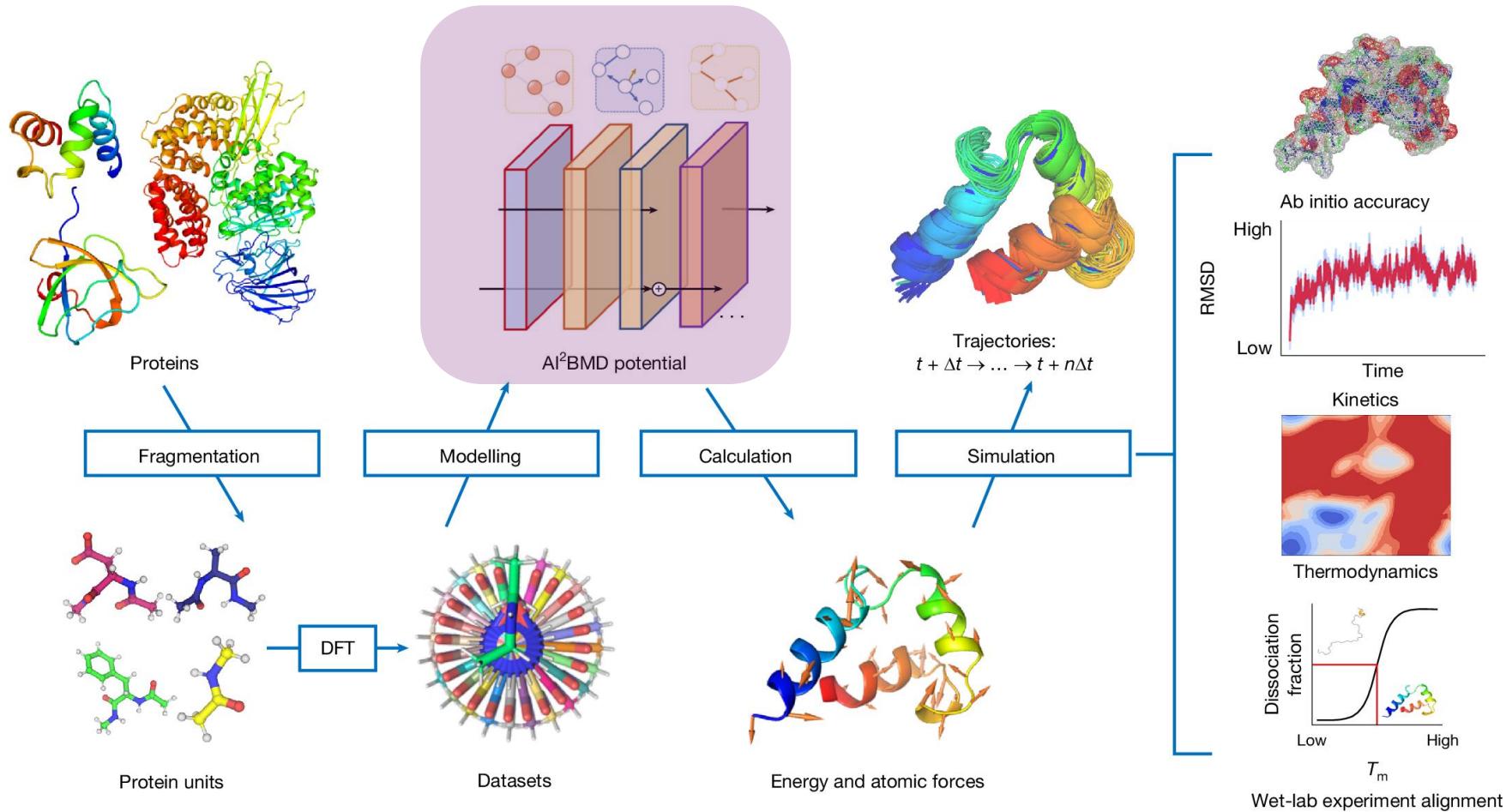
# AI<sup>2</sup>BMD workflow



1. 21 protein units, with atomic number ranging from 12-36
2. Comprehensive AIMD conformation sampling by scanning main-chain dihedrals

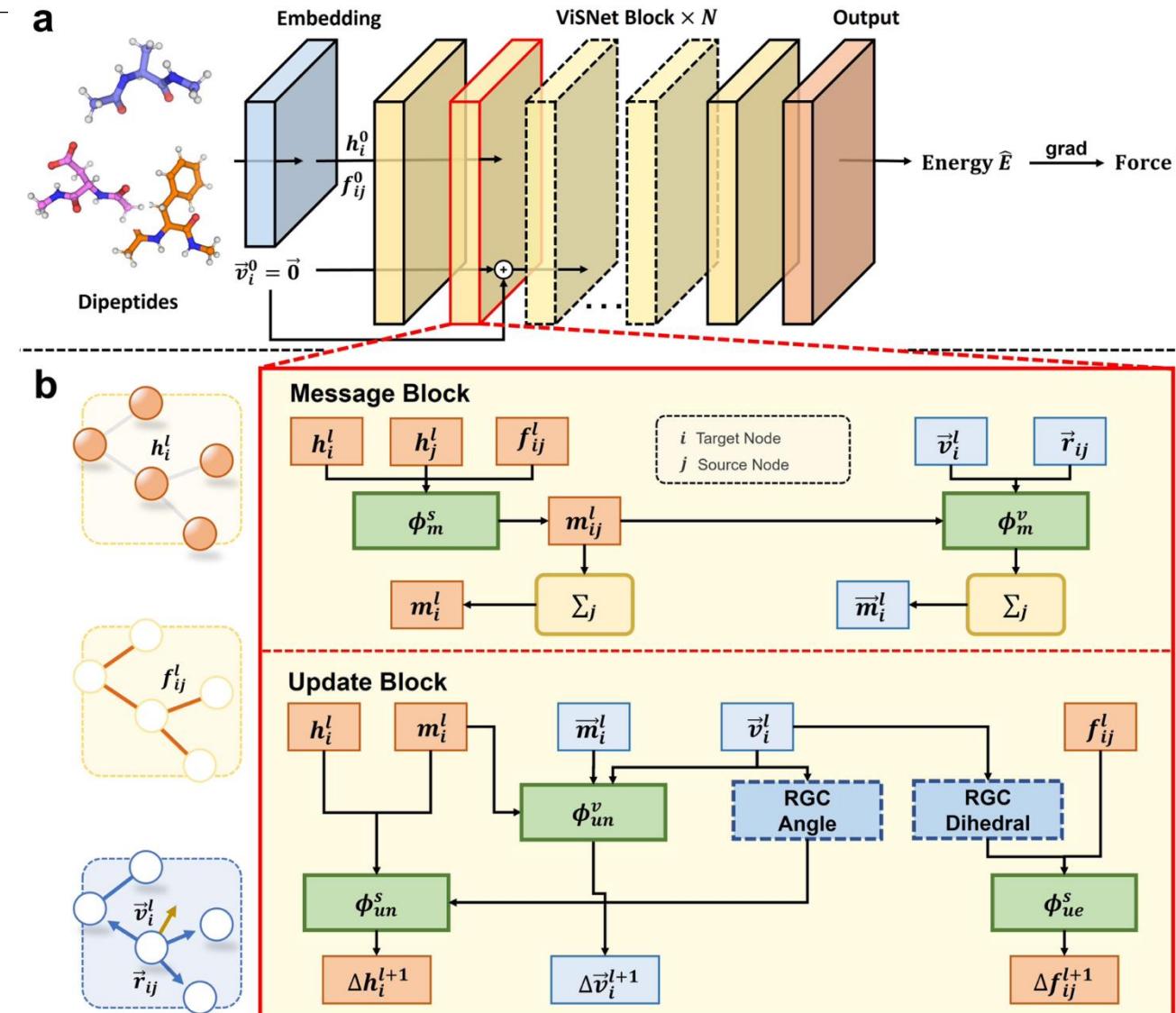
**20.88 million samples**

# AI<sup>2</sup>BMD workflow



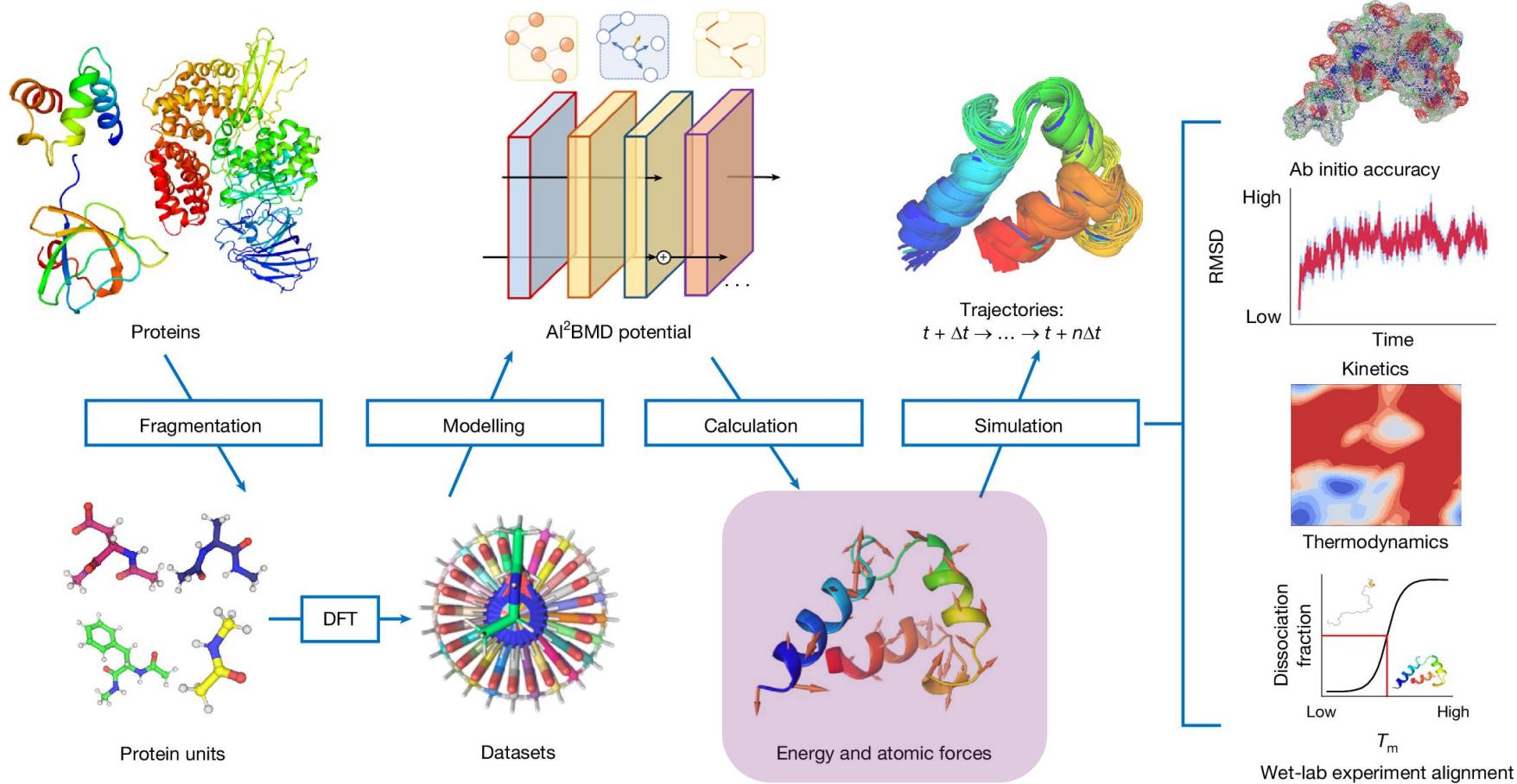
ViSNet models are trained, encoding physics-informed molecular representations and calculates four-body interactions with linear time complexity.

# AI<sup>2</sup>BMD workflow



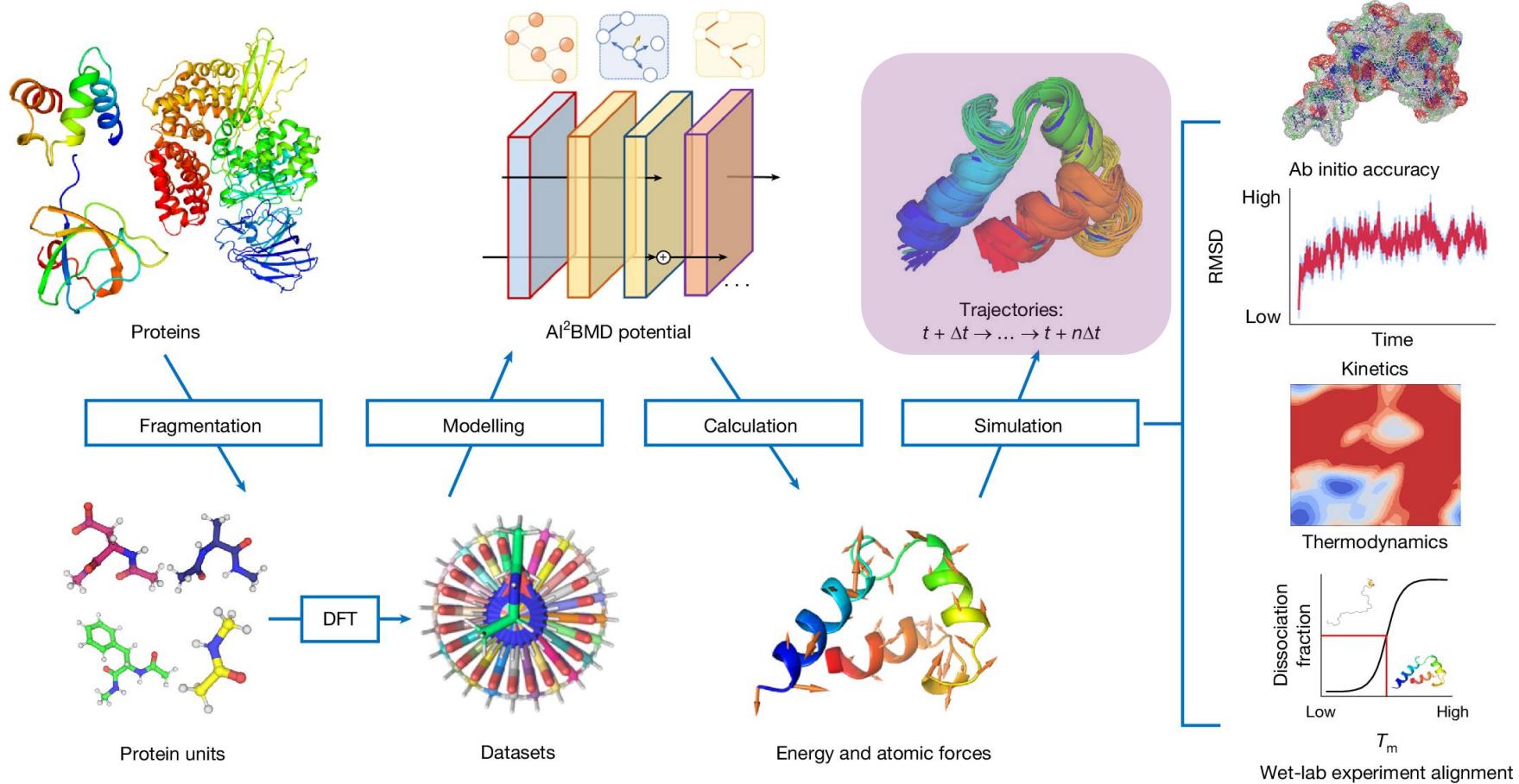
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# AI<sup>2</sup>BMD workflow



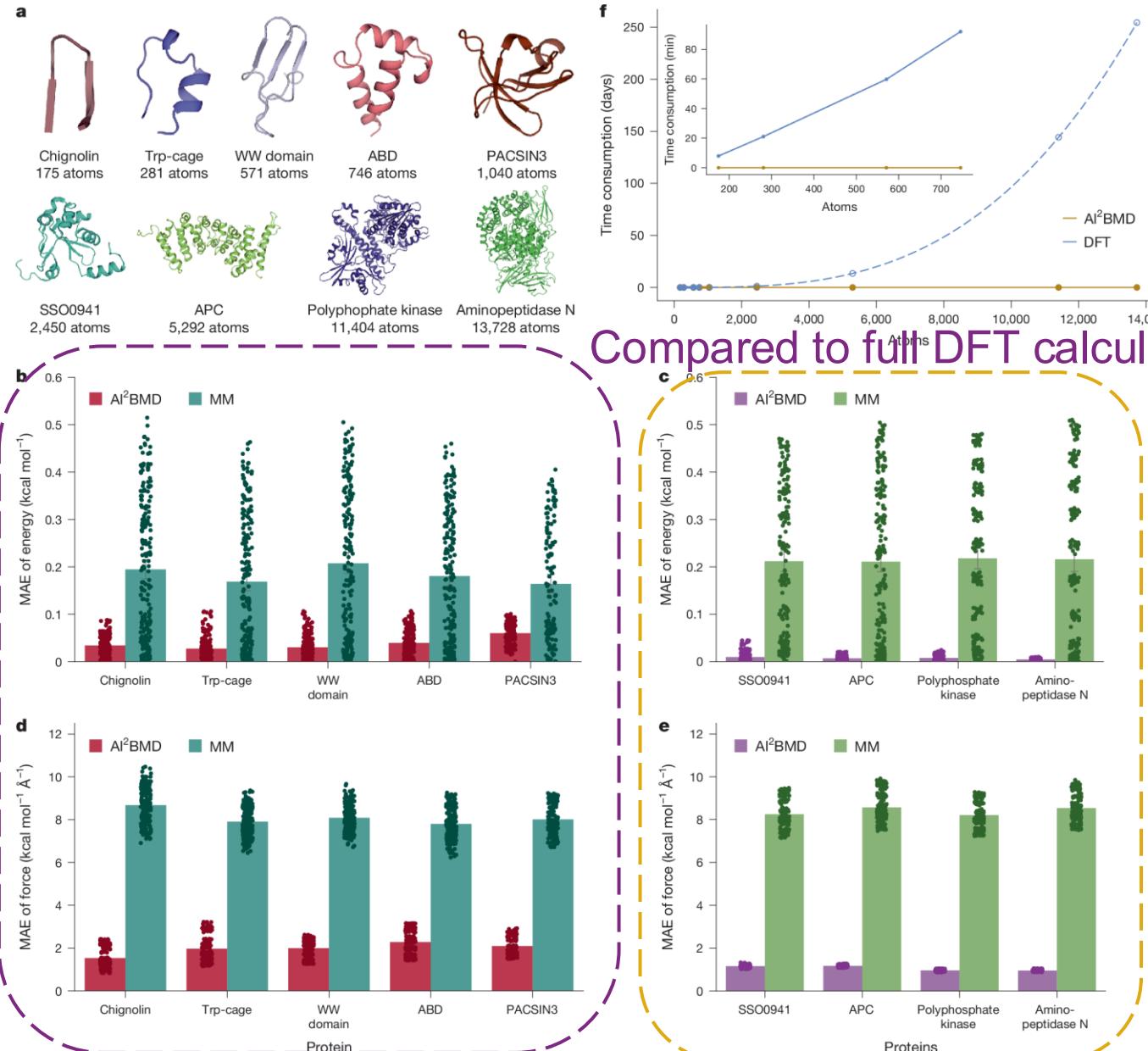
Model generates precise energy and atomic forces based on atom types and coordinates

# AI<sup>2</sup>BMD workflow



MD simulation system with polarized solvent were developed and applied for 9 proteins with number of atoms ranging from 175-13728.

# AI<sup>2</sup>BMD workflow



AI<sup>2</sup>BMD outperformed the MM force field by approximately two orders of magnitude in energy and force

(AI<sup>2</sup>BMD: 0.045 kcal mol<sup>-1</sup>, 0.078 kcal mol<sup>-1</sup> Å<sup>-1</sup>;  
MM: 3.198 kcal mol<sup>-1</sup>, 8.125 kcal mol<sup>-1</sup> Å<sup>-1</sup>).

## Comparison on running time per simulation step for proteins solvated with a 10 Å water box

Protein	Atom number of protein	Atom number of system	AI <sup>2</sup> BMD (s)	DPMD (s)	Allegro (s)	Tinker Amoeba (s)	Amber FF19S B (s)
Chignolin	175	4,715	0.047	0.040	0.238	0.117	0.004
Trp-cage	281	6,067	0.052	0.055	0.322	0.136	0.005
WW domain	571	10,678	0.070	0.095	0.626	0.196	0.008
ABD	746	11,793	0.085	0.106	0.712	0.208	0.008
PAC SIN 3	1,040	17,923	0.106	0.162	-	0.292	0.011
SSO0941	2,450	44,401	0.213	0.414	-	0.699	0.027
APC	5,292	54,999	0.449	0.580	-	0.938	0.033
Polyphosphat e Kinase	11,404	97,657	0.966	-	-	1.487	0.058

AI tools

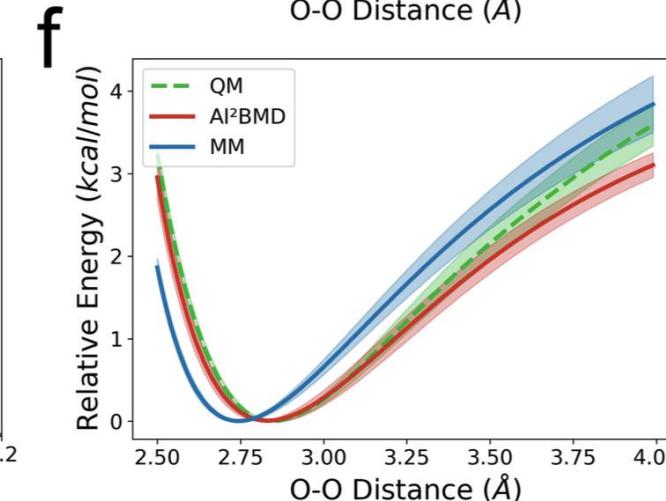
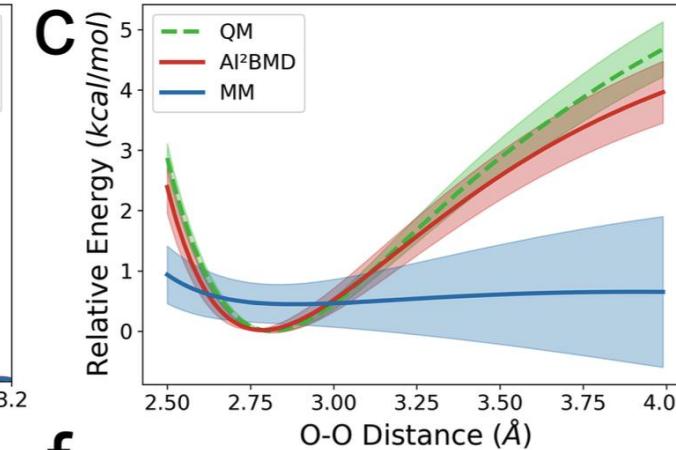
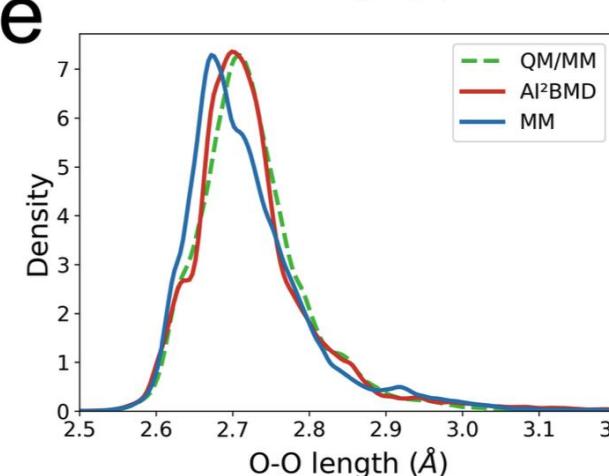
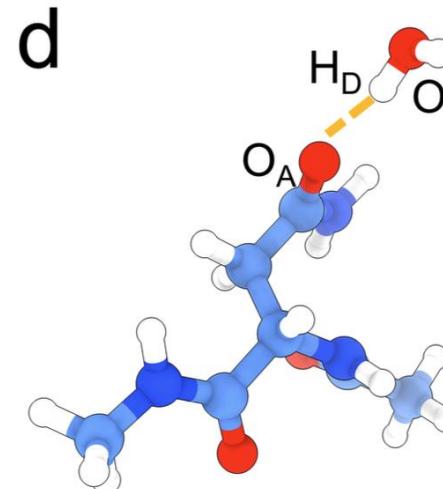
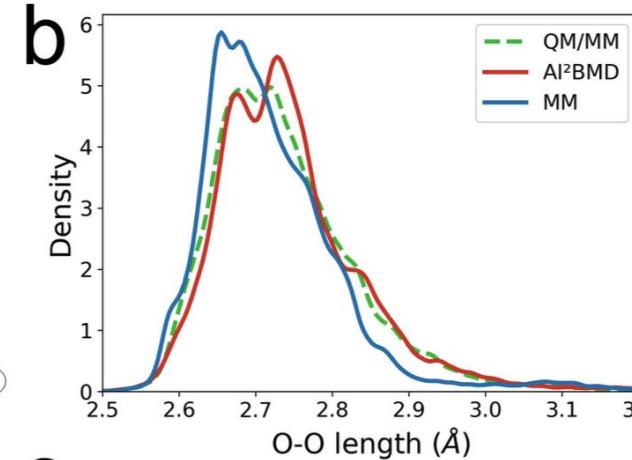
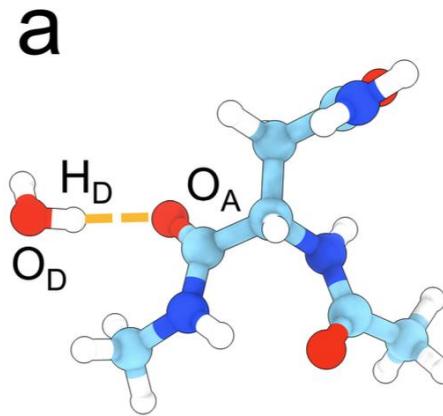
Force-field like tools

AI<sup>2</sup>BMD outperformed other AI-driven simulation tools such as DPMD and Allegro.

# AI<sup>2</sup>BMD capabilities



## 1. Conformational space exploration

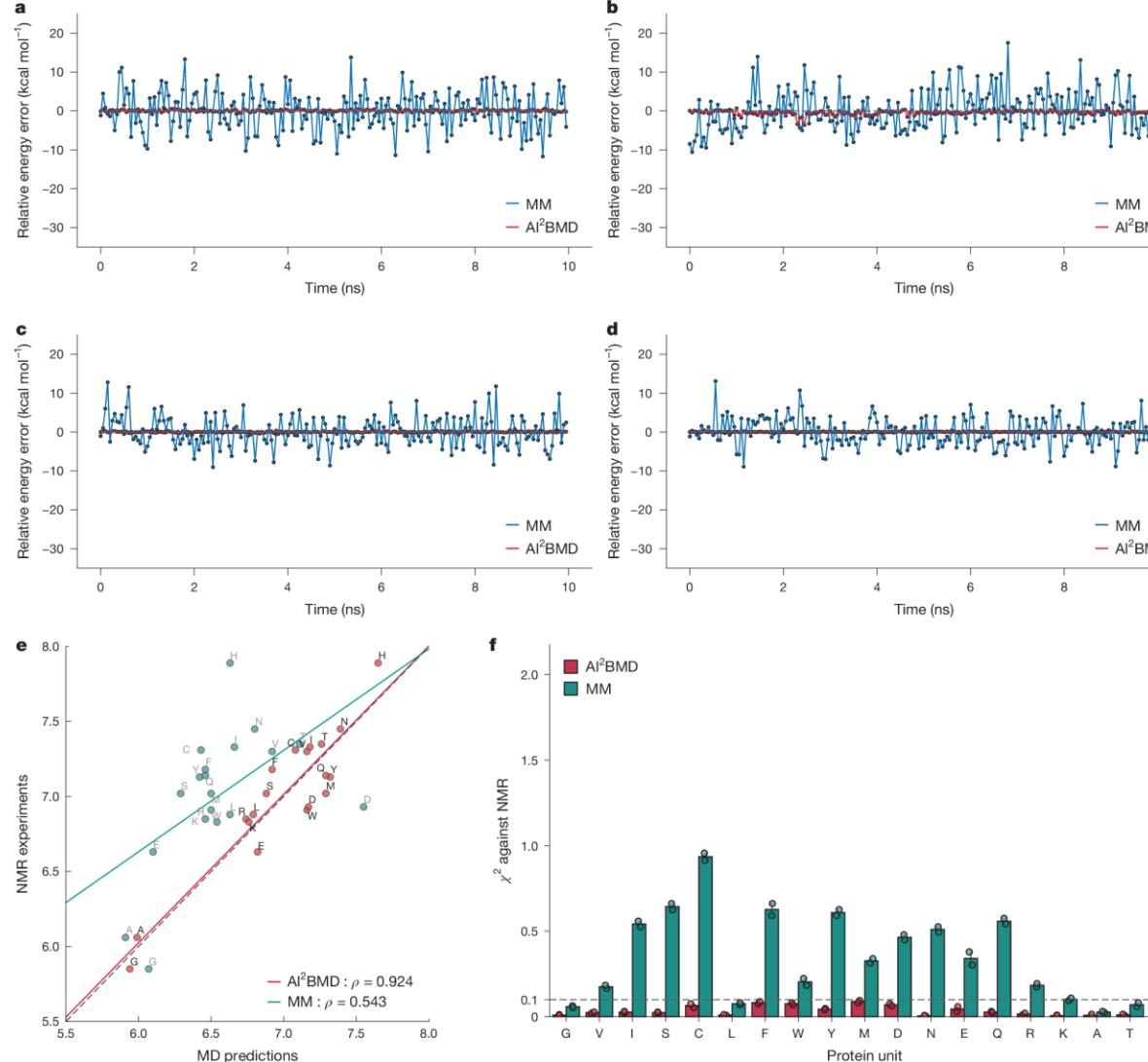


Examination of hydrogen bonding between water and the asparagine dipeptide (Ace-N-Nme) dipeptide.

AI<sup>2</sup>BMD demonstrated an energy distribution much more consistent with QM–MM than MM in bond scanning!

## 1. Conformational space exploration

AI<sup>2</sup>BMD simulations for protein units and comparisons with NMR experiments



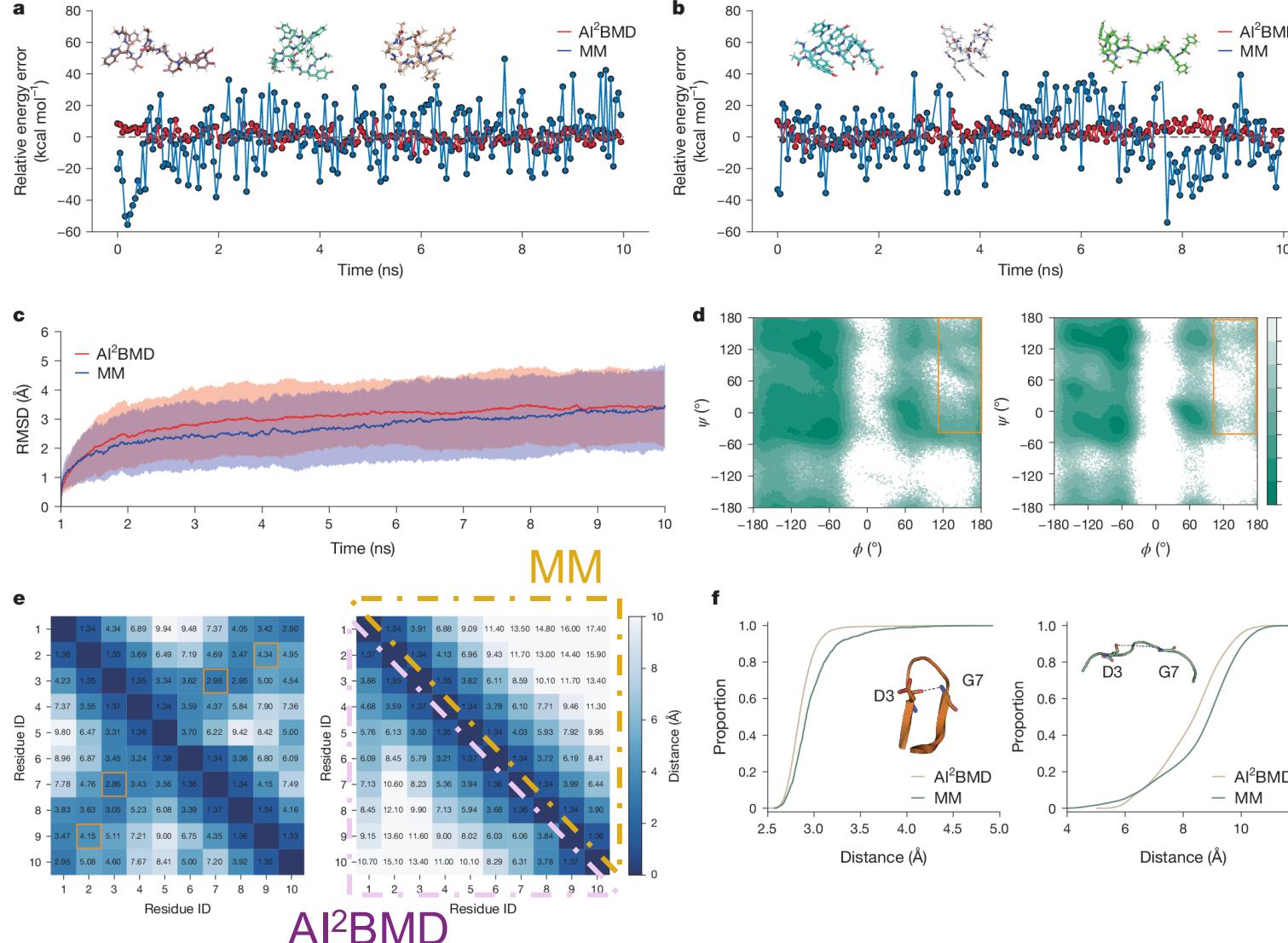
AI<sup>2</sup>BMD is much more consistent with QM–MM than MM

# AI<sup>2</sup>BMD capabilities



## 1. Conformational space exploration

Analysis of chignolin dynamics by AI<sup>2</sup>BMD simulations



Simulations performed by AI<sup>2</sup>BMD exhibit similar structure fluctuations with MM

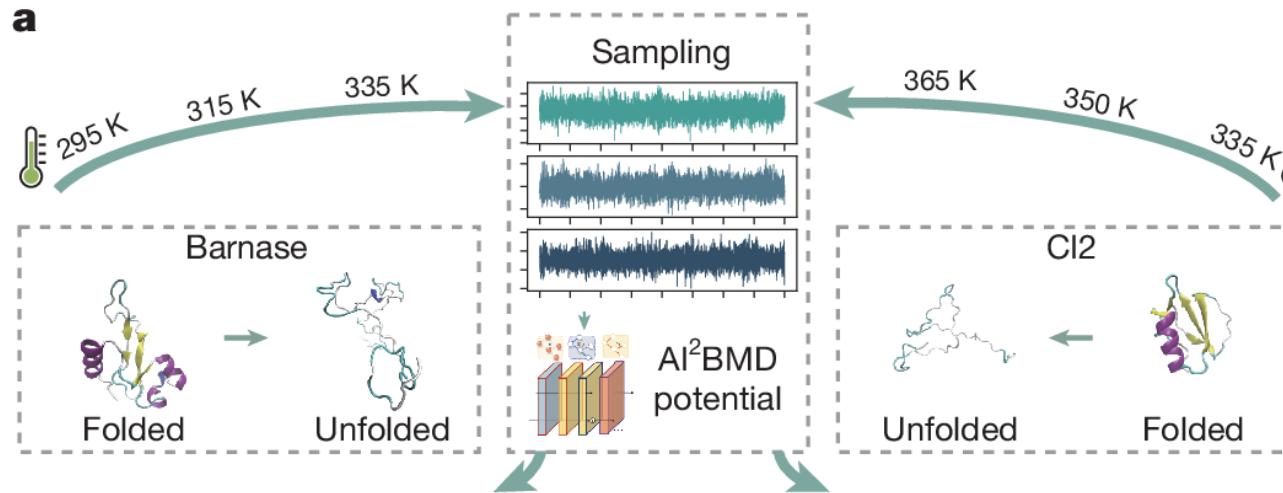
AI<sup>2</sup>BMD can detect both meaningful conformational changes and detailed interatomic interactions to study protein dynamics.

# AI<sup>2</sup>BMD capabilities

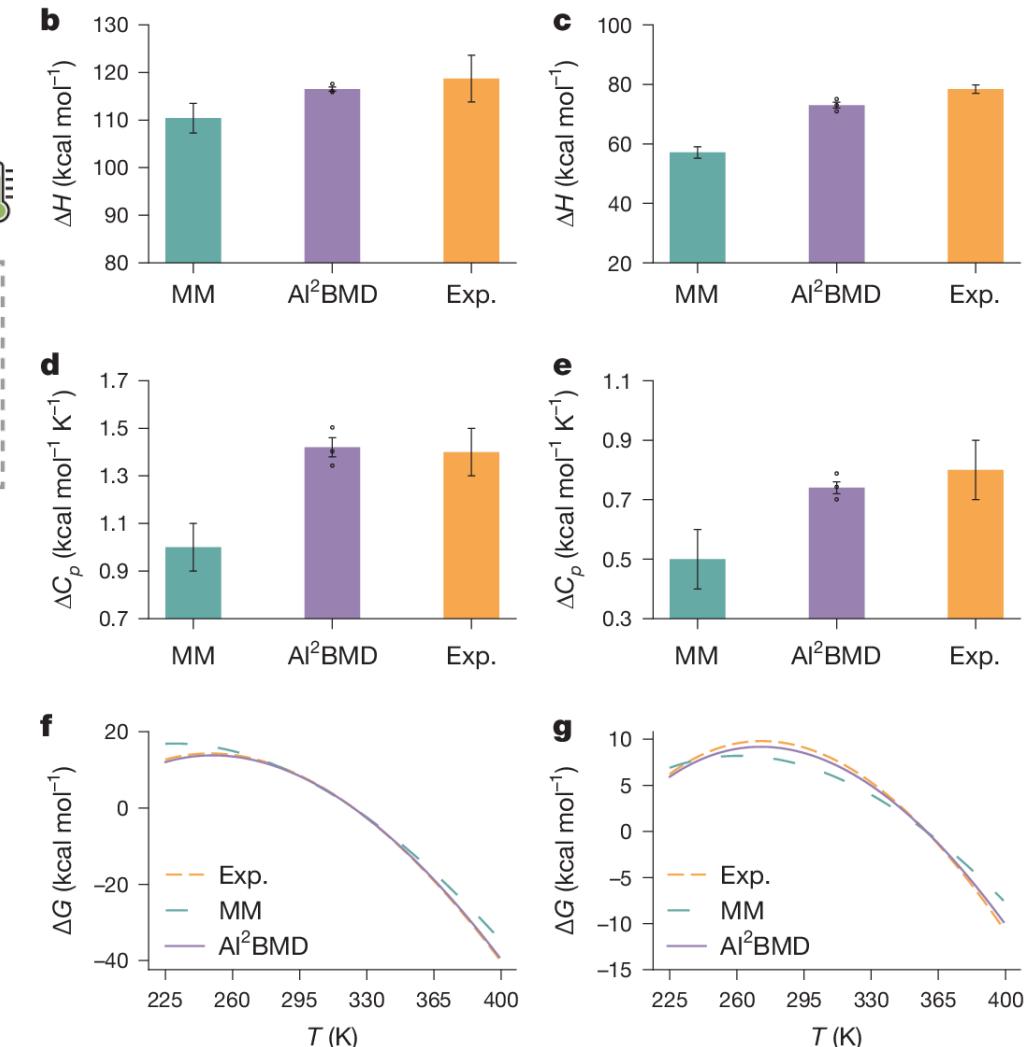


## 2. Protein property estimation

Comparison of the change of enthalpy, heat capacity and free energy of two-state proteins, barnase and CI2.



Protein property evaluations on diverse proteins reveals AI<sup>2</sup>BMD's capability of calculating conformational ensembles accurately, leads to reasonable estimations of protein folding thermodynamic.



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# Conclusion and perspectives

## ➤ AI2BMD

- ❖ **Expands ab initio calculation from a small preset QM region to the whole full-atom protein without any prior knowledge.**
  - Eliminates the potential incompatibility of QM and MM mechanics on the boundary for proteins and accelerates QM region calculation.
  - Offer opportunities with new perspectives for complex biomolecular dynamics that QM-MM cannot deal with.
  
- ❖ **Exhibits generalization ability based on fundamental assembling principles that most proteins are composed of common kinds of amino acid**, which can be expanded to other biochemical systems such **as** lipids, nucleotides, nanomaterials and solute-solvent interfaces.

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# Questions? Comments?

# Thank You