

Chamaileon: Cross-Context Binder Design with Contextualized Modeling and Mixed Sampling

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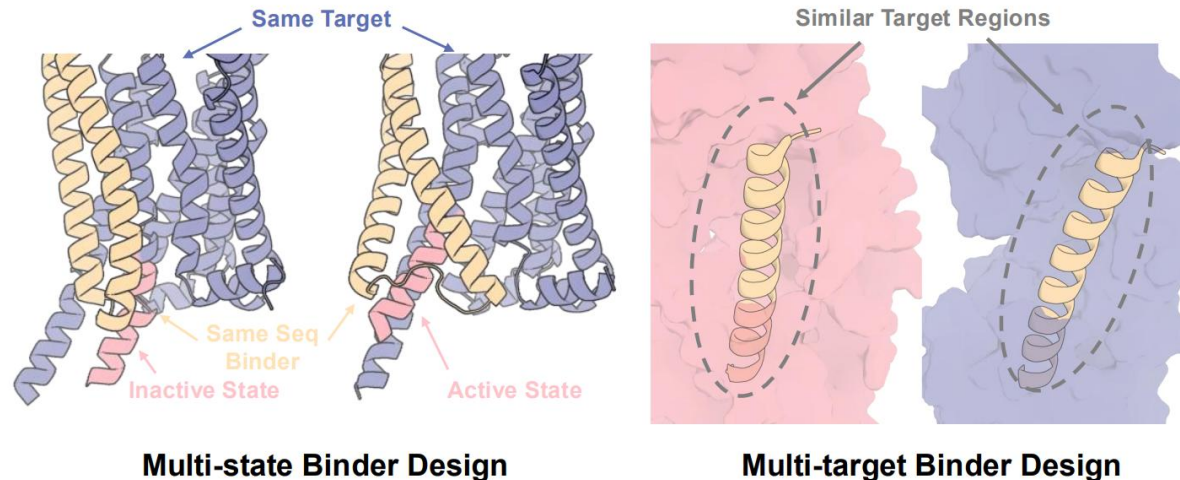
Introduction ► Motivation

- **The Problem with Current Methods**

Despite remarkable progress in *de novo* protein binder design, most existing methods still **treat binder design as a one-to-one problem**: optimizing a binder against a single target structure under a single binding objective.

- **However, many real-world scenarios demand designs that go beyond this paradigm:**

- **Multi-State binder design** — Two targets share an identical sequence but adopt distinct conformations; the binder must accommodate these conformational changes accordingly.
- **Multi-Target binder design** — Two targets have different sequences; the binder must engage their shared (homologous) epitopes.





Introduction ▶ A Unified View

- **A Unified View of Multi-state and Multi-target binder design**

Multi-state and multi-target design share the same computational structure: **one sequence must satisfy multiple binding constraints with explicit trade-offs.**

- **Cross-Context Binder Design**

- We unify multi-state and multi-target binder design under what we call **Cross-Context Binder Design.**
- Here, we define contexts as **the protein interfaces to specifically bind**, regardless of whether they come from different conformational states of the same protein (*Multi-State, MS*) or from distinct targets (*Multi-Target, MT*).

- **The Fundamental Gap in Current Architectures**

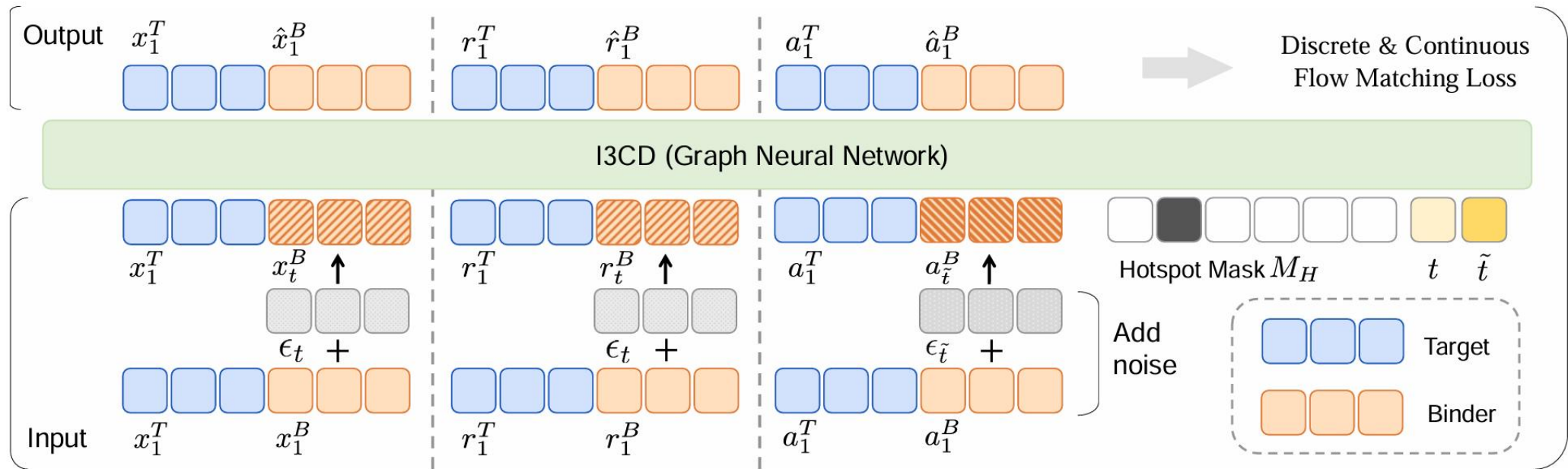
- **Decoupled noise schedules for sequence and structure** to maintain sequence consistency across conformational states during joint target-binder modeling.
- **Inference-time optimization** to navigate the trade-offs between conflicting structural constraints.
- **Comprehensive evaluation** that measures success across a contextual ensemble rather than on isolated snapshots.



Method ▶ In-Context Complex Co-Design (I3CD)

• Training Pipeline of I3CD

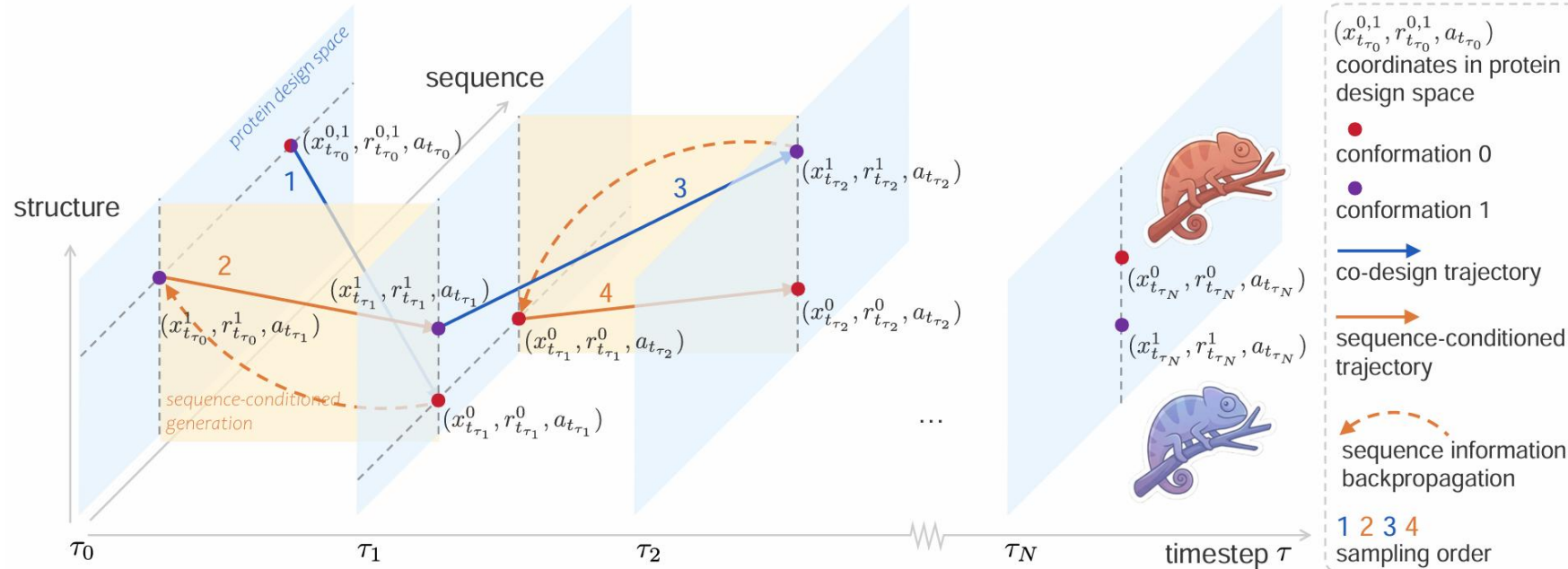
- In-context generation
- Decoupled noise schedules for sequence and structure
- Single-state binder codesign framework





Method ▶ Mixture-of-Paths Sampling (MoPS)

- **Why not map multiple targets to a single binder sequence with multiple conformations?**
 - Scarcity of multi-state training data.
 - Limited flexibility in handling variable conformations
- **MoPS for Cross-Context Binder Design**
 - Sequence and structure co-design
 - Sequence-conditioned generation





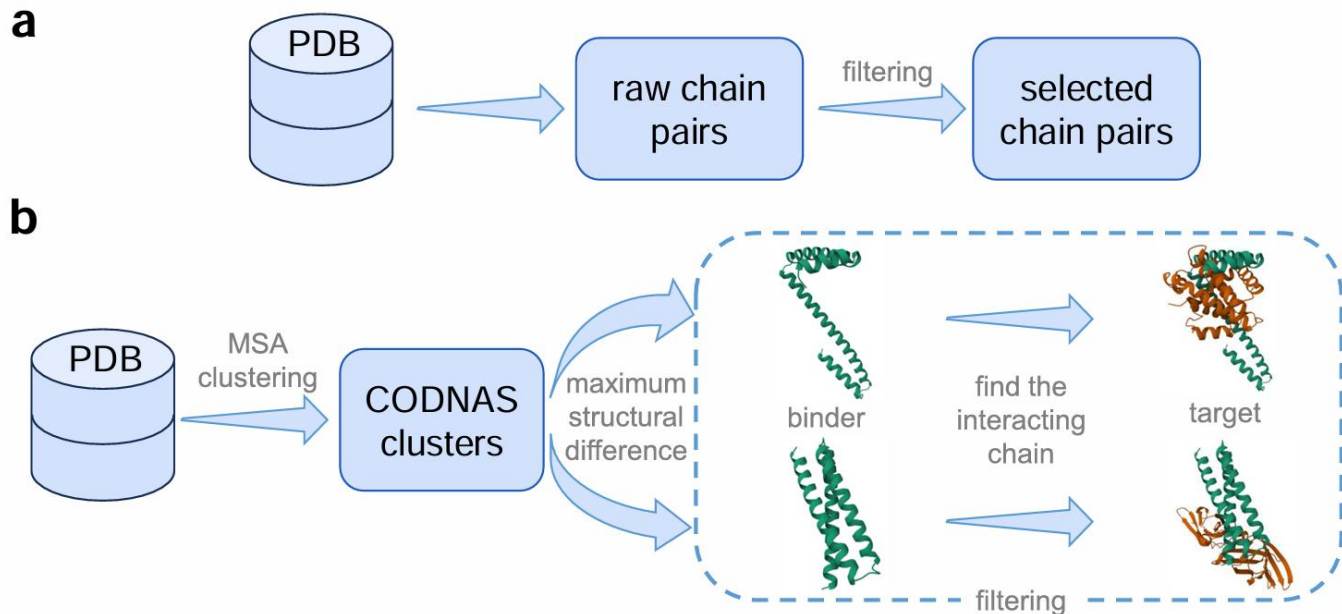
Method ▶ (Multi-State) Complex Data Collection

- **Training Set Construction**

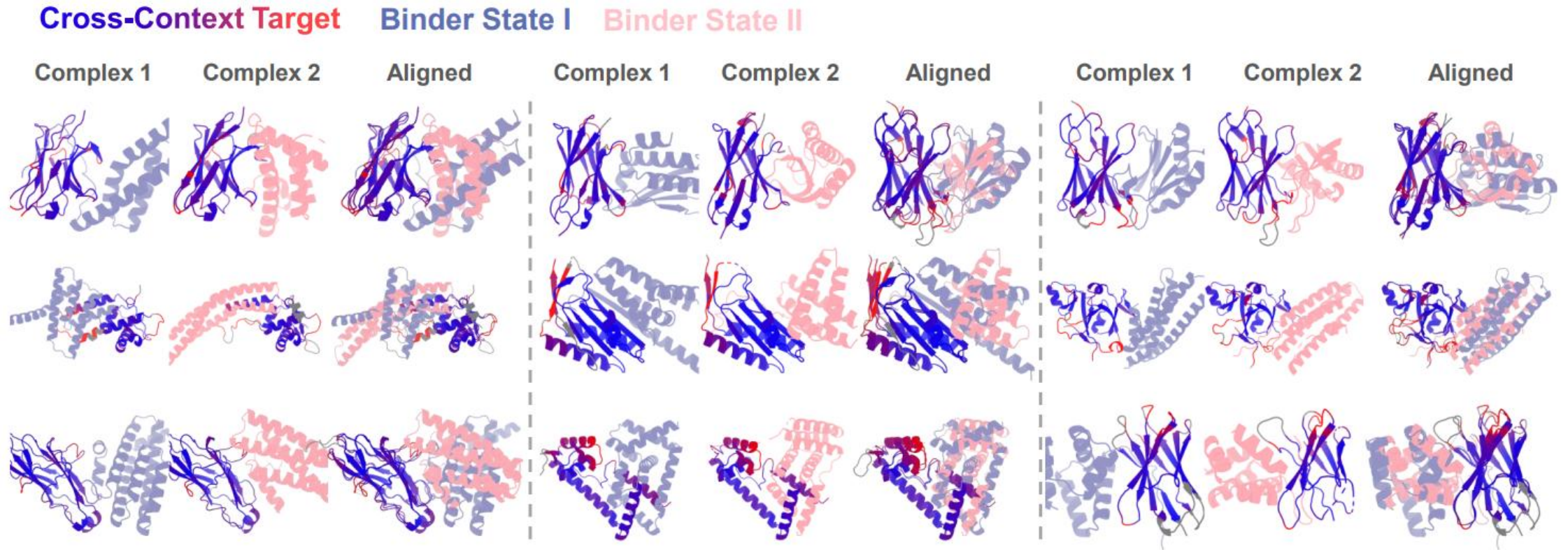
- Find raw chain pairs.
- Filtering.

- **Cross-Context Binder Design Benchmark**

- CROSS (Comprehensive Recognition of Specific Surfaces)
- Find multi-state binders.
- Find corresponding targets.
- Filtering.



Results ▶ Qualitative results



Results ► Quantitative results

• Ablation Studies

		Conformation 0					Conformation 1					Both Success
		ipAE	binder pLDDT	binder scRMSD	Unique Success	novelty	ipAE	binder pLDDT	binder scRMSD	Unique Success	Novelty	
Module Ablation	w/o MoPS	6.26	82.9	1.83	17	0.500	6.27	82.9	2.49	2	0.863	2
	w/o beam search	5.04	88.2	1.64	5	0.363	6.18	83.9	1.76	8	0.304	5
	full version	4.32	90.6	1.96	8	0.437	4.23	88.7	2.07	10	0.588	7
Beam Search Candidate Number	1	5.27	88.7	2.39	7	0.385	4.78	89.5	1.77	6	0.297	5
	2	4.15	88.8	1.10	7	0.381	4.36	89.8	1.59	7	0.461	5
	4	4.32	90.6	1.96	8	0.437	4.23	88.7	2.07	10	0.588	7
Beam Search Frequency	250	4.23	89.8	1.45	9	0.494	4.97	87.1	1.78	11	0.447	8
	100	5.02	89.2	1.90	10	0.431	4.94	89.6	1.70	9	0.266	8
	50	4.32	90.6	1.96	8	0.437	4.23	88.7	2.07	10	0.588	7
MoPS Frequency	250	-	-	-	0	-	5.30	84.8	1.70	19	0.310	0
	100	5.64	86.4	1.88	17	0.473	2.21	91.7	1.11	1	0.928	1
	50	2.98	92.3	2.19	1	0.903	4.93	87.5	2.12	15	0.546	1
	20	5.14	88.1	1.81	13	0.398	4.10	90.1	2.07	9	0.510	7
	10	4.32	90.6	1.96	8	0.437	4.23	88.7	2.07	10	0.588	7

• Comparison with Baselines

Method	Conformation 0				Conformation 1				Both Success
	ipAE	binder pLDDT	binder scRMSD	Unique Success	ipAE	binder pLDDT	binder scRMSD	Unique Success	
Baseline 1 (best)	4.85	84.2	15.5	0	4.73	85.3	13.5	0	0
Baseline 2 (best)	19.1	67.7	1.37	0	18.2	68.1	3.20	0	0
Chamaileon (mean)	4.32	90.6	1.96	8	4.23	88.7	2.07	10	7



Conclusion ▶

- **Chamaileon is a unified framework containing I3CD, MoPS, and CROSS.**
- **Cross-context modeling represents a pivotal step toward the “programmable” design of advanced modulatory effects and multi-specific therapeutics.**

Thanks for listening

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