

Machine Learning in Computational Biology

ViDa: Visualizing DNA reactions with biophysics-informed deep graph embeddings

December 1st, 2023

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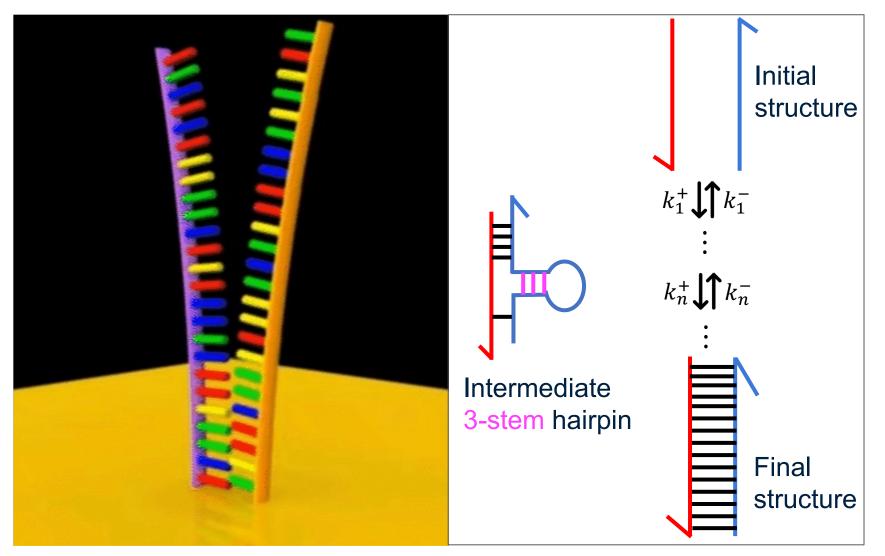


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MOTIVATION

Substantial number of intermediate states, making the visualization problem challenging!



DNA hybridization

1. Background

- Reaction trajectory
- Multistrand simulator
- Coarse-grained visualization

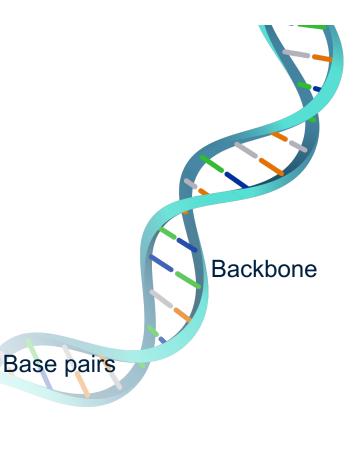
2. Method: ViDa

- Model architecture
- Loss functions

3. Results

- Secondary structure and trajectory embeddings
- Comparison with other methods

4. Future work



1. Background

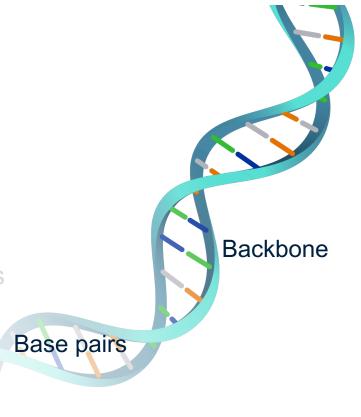
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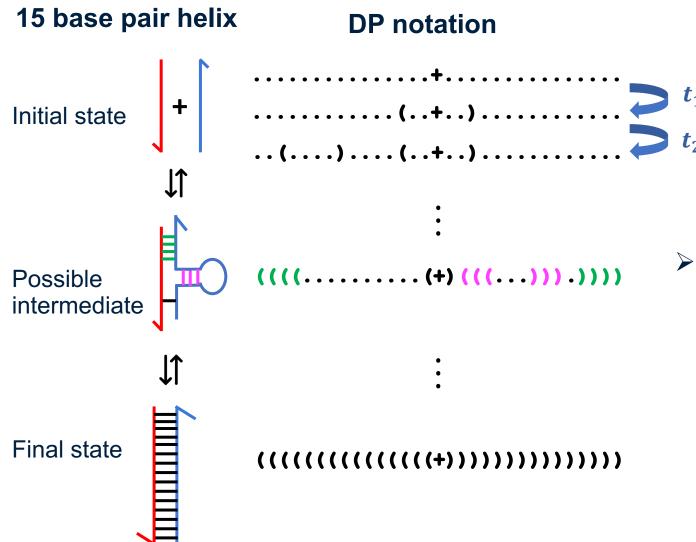
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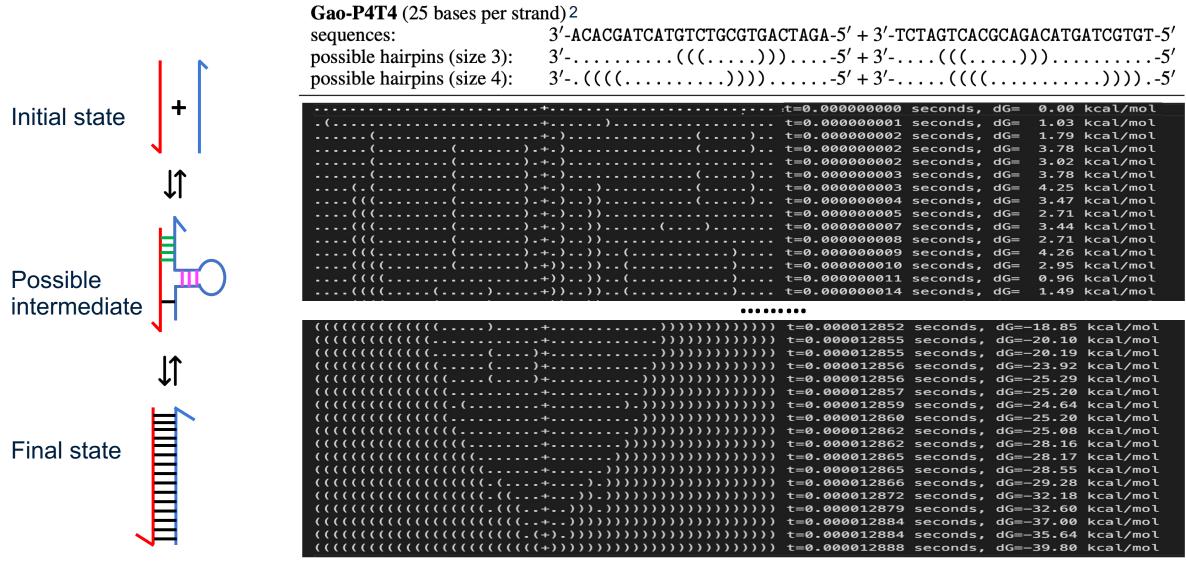
BACKGROUND – REACTION TRAJECTORY



- Dot-parenthesis (DP) notation:
 - Dots: unpaired bases;
 - Parentheses: paired bases;
 - "+" sign: separates two strands
- Reaction trajectory:
 - The sequence of secondary
 structures, from the reactants to the
 products of a DNA reaction, along with
 the time to transition from one state to
 the next.

BACKGROUND – MULTISTRAND SIMULATOR

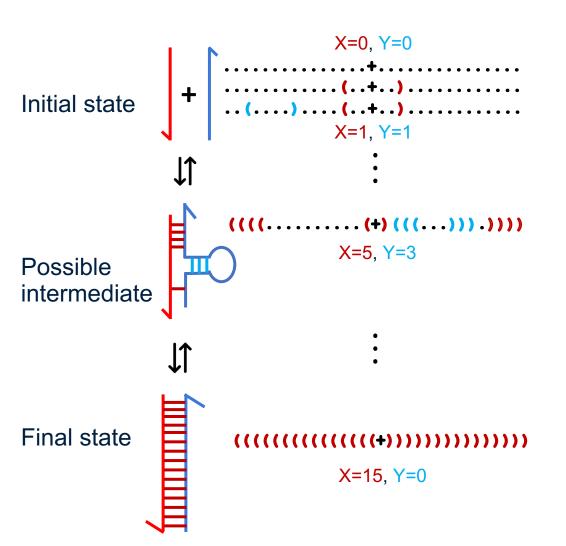
> Multistrand: DNA elementary step kinetics simulator¹

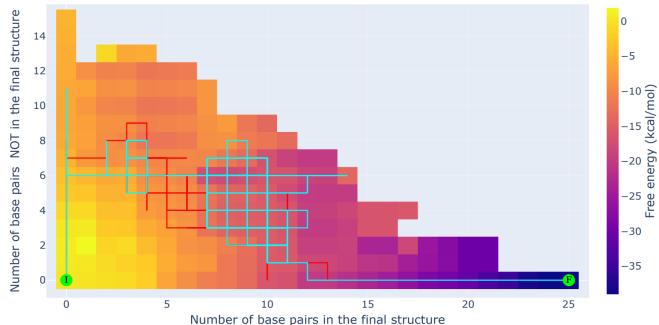


¹J.M. Schaeffer. PhD thesis, California Institute of Technology, 2013.

²Gao Y., et al. Nucleic Acids Res. 2006.

BACKGROUND – COARSE-GRAINED VISUALIZATION





Limited visual interpretation of physical process:

i. A single "cell" (macrostate) may contain with very

different structures;

ii. Overlapping trajectories may represent very distinct pathways.

CHALLENGES

What is a good state embedding for domain experts?

- > High resolution:
 - Distinguishability of individual secondary structure embeddings
- Well-preserved global structure (energy landscape)
 - $\circ~$ State embedding follows the trend of the course of a reaction
- Well-preserved local structure (state topology)
 - Closeness of embeddings for structurally adjacent states
 - Separation of embeddings for structurally distinct states
- Distinguishable and smooth trajectories laid over the embedded states
 - Separated trajectories and no large jumps along the trajectories

CONTRIBUTIONS

Criterion	Coarse-grained	ViDa (ours)
High resolution	×	
Well-preserved global structure		
Well-preserved local structure	×	
Distinguishable and smooth trajectories	×	

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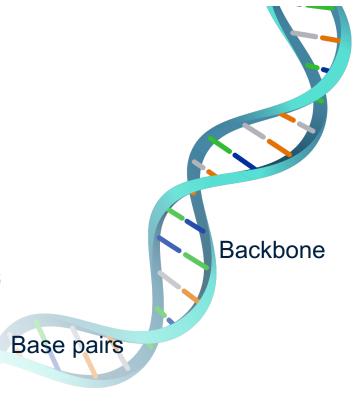
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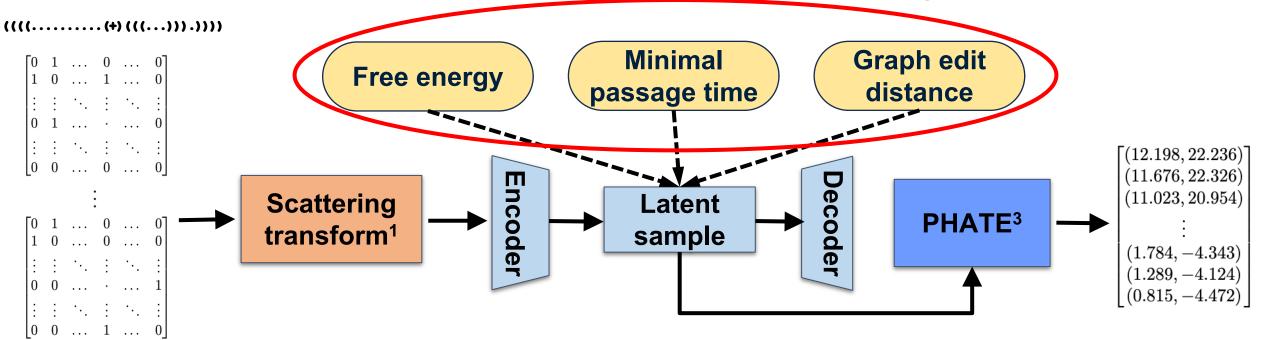
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METHOD – MODEL ARCHITECTURE

Biophysics-informed features



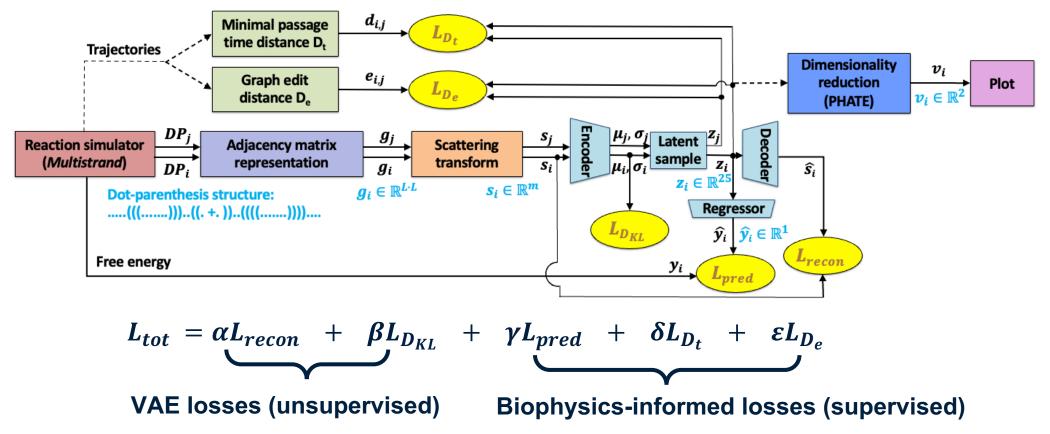
Nodes: basesMap graph intoEmbed input vectors with a VAE²Further reducePlotEdges: base pairsa vector spacetrained by incorporatingdimension to 2D& backbones(scatteringbiophysics-informed features intoteatures intocoefficients)the loss functionteatures into

1. Gao, F. et al. ICML, pp. 2122-2131, 2019.

2. Kingma, D. P. et al. Foundations and Trends® in Machine Learning, 12 (4): 307-392, 2019.

3. Moon, K.R., et al. Nat. Biotechnol., 37 (12): 1482-92, 2019.

METHOD – LOSS FUNCTION



≻ Free energy (ΔG) loss: $L_{pred} = \frac{1}{n} \sum_{i} (\hat{y}_i - y_i)^2$, where \hat{y}_i is the predicted energy; y_i is Multistrand energy.

▶ Graph edit distance (GED) loss: $L_{D_e} = \sum_{i,j} (||z_i - z_j|| - e_{i,j})^2$, where $e_{i,j}$ is the graph edit distance between state *i* and *j*.

➤ Minimal passage time (MPT) loss: $L_{D_t} = \sum_{i,j} w_{i,j} (||z_i - z_j|| - d_{i,j})^2$, where $w_{i,j}$ is an importance weight; $d_{i,j}$ is a normalized estimate of minimal passage time from state *i* to *j* or from *j* to *i*.

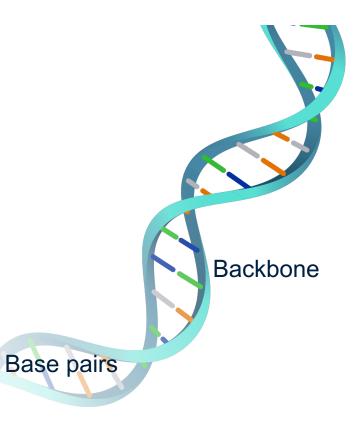
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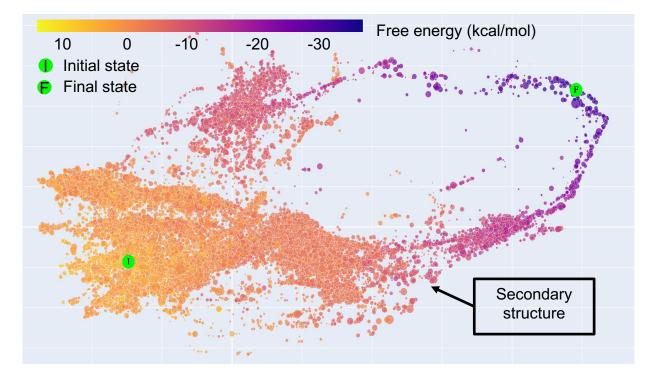
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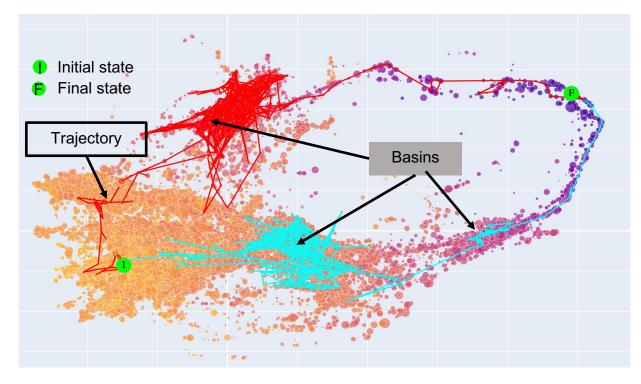
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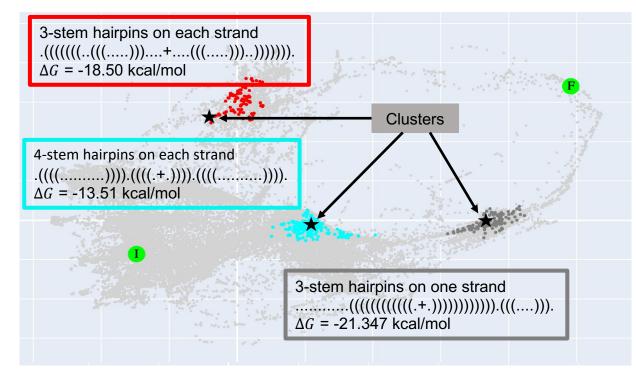




Well-preserved global and local structure



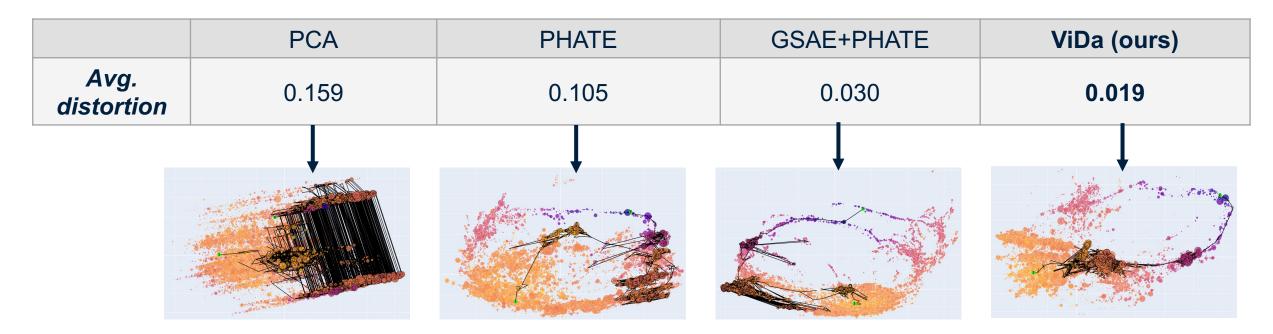
- Well-preserved global and local structure
- Embedded trajectories are distinguishable and smooth



- Well-preserved global and local structure
- Embedded trajectories are distinguishable and smooth
- Two major energy basins with different hairpins
 - 4-stem hairpins on each strand (consistent with analysis by Schreck et al.¹)
 - 3-stem hairpins on each strand (new finding)



Average distortion: the frequency-weighted mean Euclidean distance between consecutive secondary structure pairs in the sampled trajectory dataset



Integrating biophysics-informed features improves embeddings and reduces trajectory distortion!

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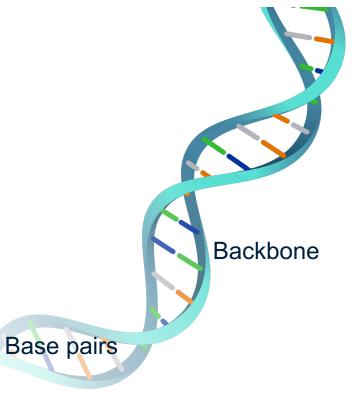
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FUTURE WORK

Generalize ViDa for more complicated DNA reactions (e.g. three-way strand displacement), and for other types of datasets such as RNA reactions, molecular dynamics trajectories, etc.

Improve ViDa to enable partitioning of secondary structure microstates into clusters corresponding to different strand-level complexes.

ACKNOWLEDGEMENT

 Co-supervisors Anne Condon(UBC)



• Co-workers Jordan Lovrod (UBC)



Khanh Dao Duc (UBC)



Boyan Beronov (UBC)



Collaborators
 Erik Winfree (Caltech)





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Thank you for listening! Q&A



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https://chwzhang.com

