

Machine-learned reaction coordinates for energy-entropy disentanglement

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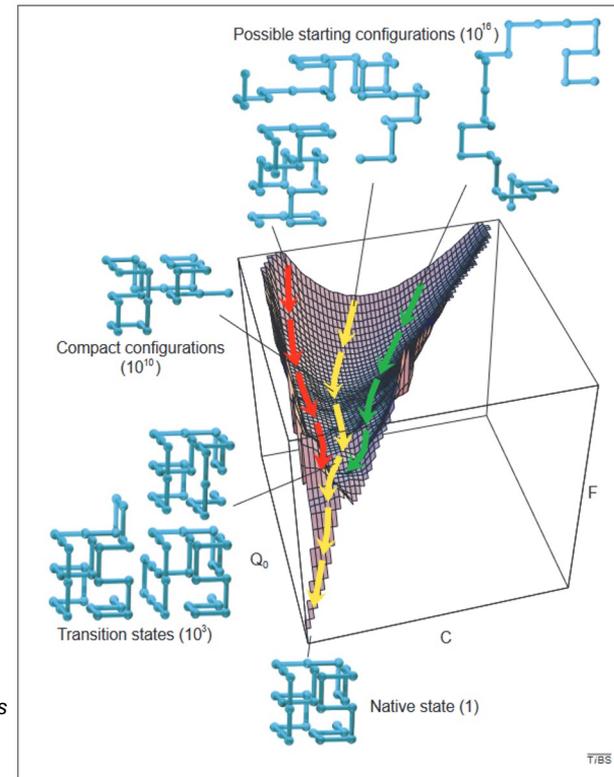
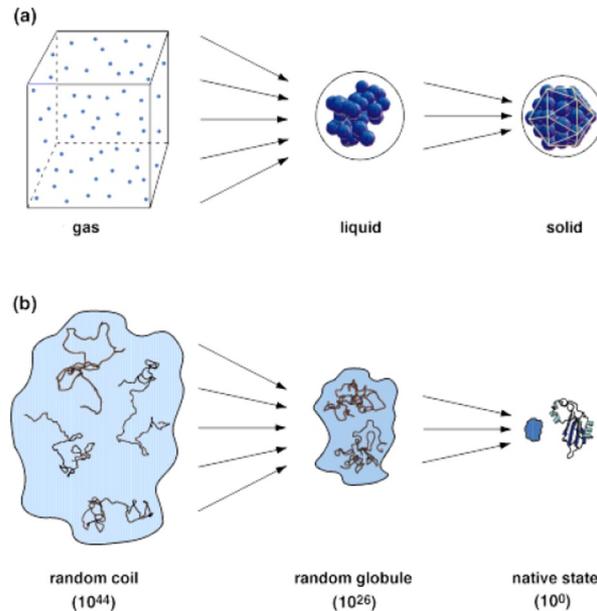
Outline

1. Introduction – free energy in physical systems
2. Discovering a relevant latent space – SPIB
3. Energy and entropy in the SPIB latent space
4. Application to simple systems
5. Augmenting the SPIB loss function to learn energetic and entropic pathways
6. Conclusions

Introduction

- All physical processes occurring in a thermostatted ensemble are subject to free-energy barriers, which are a sum of energy and entropy:

$$\Delta A = \Delta U - T\Delta S$$
$$\Delta G = \Delta H - T\Delta S$$
$$= \Delta A + P\Delta V$$



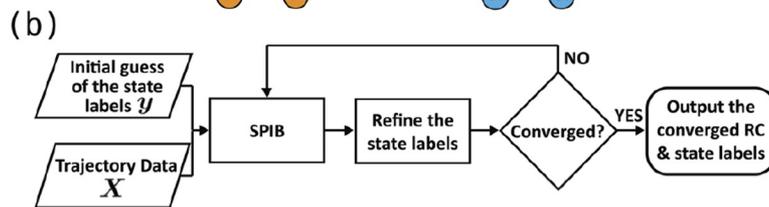
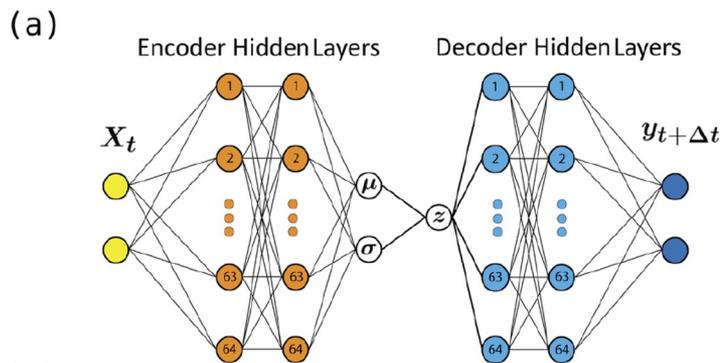
Dinner, Sali, Smith, Dobson, and Karplus, "Understanding protein folding via free-energy surfaces from theory and experiment," *Trends in Biochemical Sciences*, **25**, 331-339, 2000

Dobson, Sali, and Karplus, "Protein Folding: A Perspective from Theory and Experiment," *Angew. Chem. Int. Ed.* **37**, 868 - 893, 1998

Introduction

- Generally, there is not a simple, low-dimensional coordinate system describing the entropy- and energy-dominated pathways.
 - Important for determining enhanced sampling method utilized
- Goal of the research described here is to embed the system dynamics in a low-dimensional latent space where energy and entropy pathways are disentangled.
- Our embedding method of choice is to use a modified version of the reweighted autoencoder variational Bayes (RAVE) method called SPIB.

SPIB



$$\mathcal{L}_{\text{SPIB}} = \frac{1}{M \cdot L} \sum_{n=1}^N \sum_{l=1}^L \left[\underbrace{\log q_{\theta}(\mathbf{y}^{n+s} | \mathbf{z}^{(n,l)})}_{\text{reconstruction error}} - \beta_{\text{SPIB}} \underbrace{\log \frac{p_{\theta}(\mathbf{z}^{(n,l)} | \mathbf{X}^n)}{q_{\theta}(\mathbf{z}^{(n,l)})}}_{\text{KL divergence}} \right]$$

$$\mathcal{L}_{\text{SPIB}} \leq \mathcal{L}_{\text{IB}} = \underbrace{I(\mathbf{y}, \mathbf{z})}_{\sim \text{accuracy}} - \beta \underbrace{I(\mathbf{X}, \mathbf{z})}_{\sim \text{compression}}$$

Energy and Entropy in the Embedded SPIB Latent Space

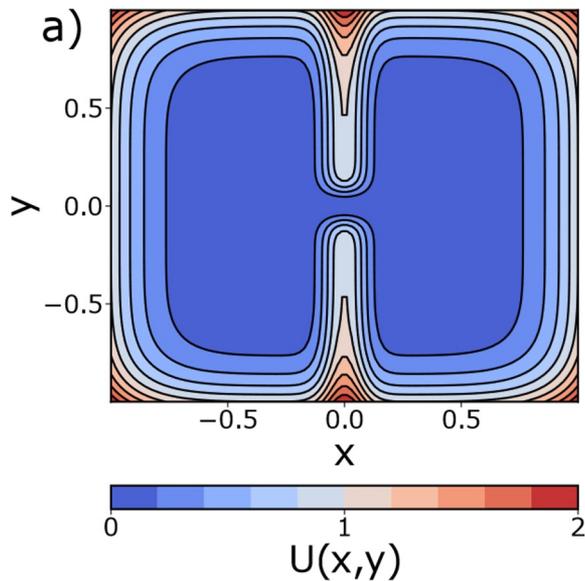
- Use the formulation provided by Hartmann et al.¹ to determine the free energy and energy along the discovered RCs:

$$G(\mathbf{z}) = -\beta^{-1} \ln \underbrace{\int_{\mathbb{R}^n} d\mathbf{x} \exp(-\beta V(\mathbf{x}))}_{\text{Boltzmann weight of feature } \mathbf{x}} \underbrace{\det(\tilde{G})^{\frac{1}{2}}}_{\text{Grammian}} \underbrace{\delta(\Phi(\mathbf{x}) - \mathbf{z})}_{\text{Count points on the level set/histogram bin of interest}}$$

$$\begin{aligned} \langle U(z) \rangle_{\Sigma(z)} &:= U(z) \\ &= \frac{1}{N_z} \int U(\mathbf{x}) e^{-U(\mathbf{x})/k_B T} \times \delta(\Phi(\mathbf{x}) - z) \det(\tilde{G})^{1/2} d\mathbf{x} \end{aligned}$$

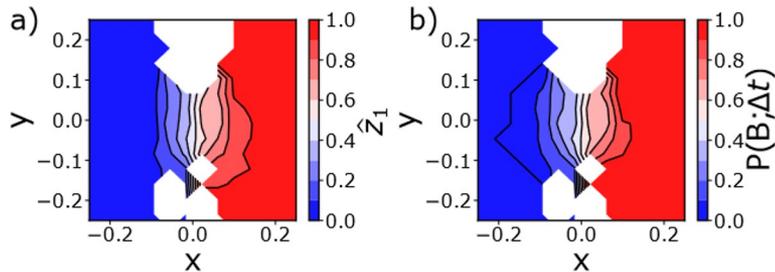
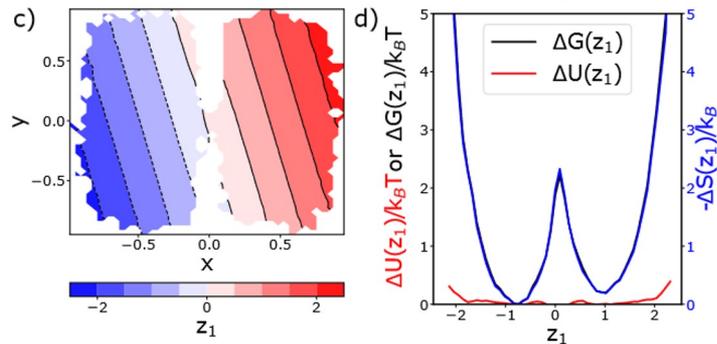
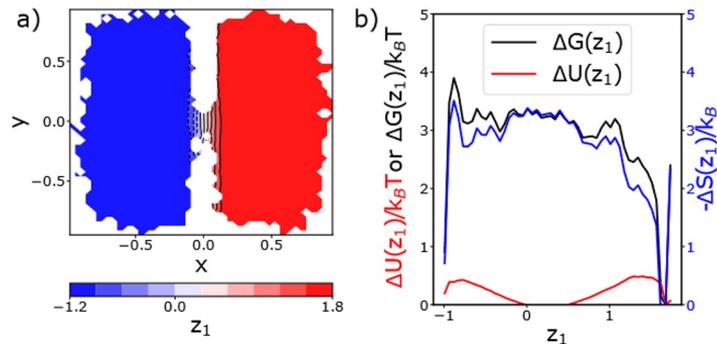
$$\begin{aligned} \Delta G(z) &= \Delta U(z) - T \Delta S(z) \\ &\Rightarrow \Delta S(z) \\ &= \frac{1}{T} (\Delta U(z) - \Delta G(z)) \end{aligned}$$

Energy-Entropy for Some Simple Systems

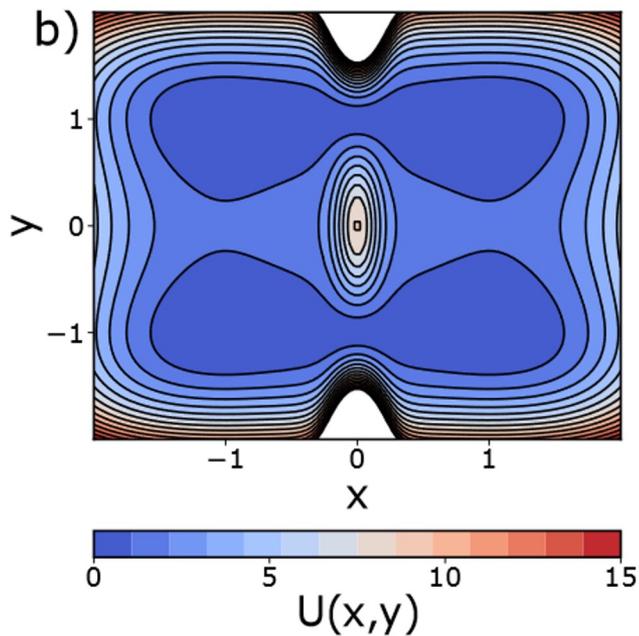


$$U(x,y) = x^6 + y^6 + \exp[-y^2/\sigma_y^2] (1 - \exp[x^2/\sigma_x^2])$$

“Elber potential”

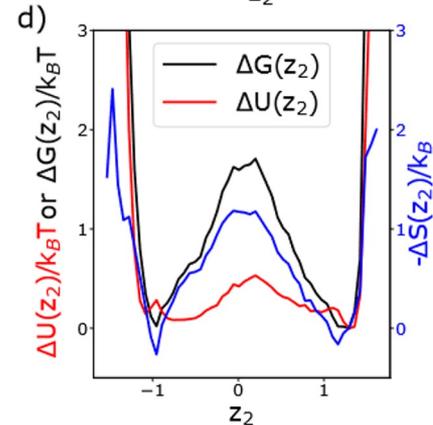
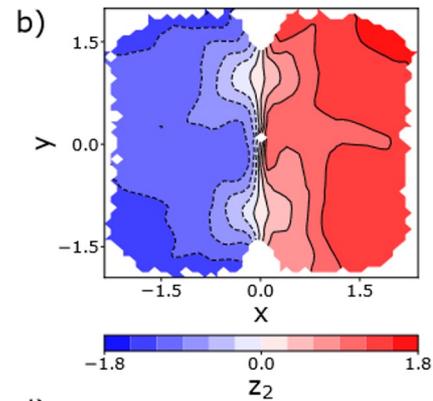
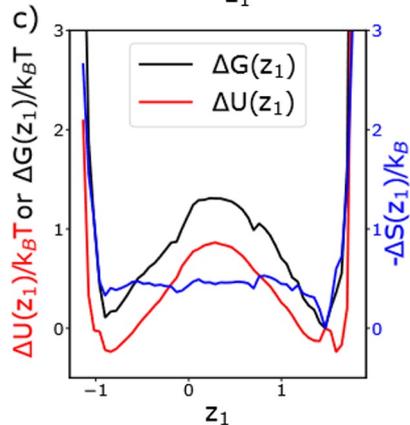
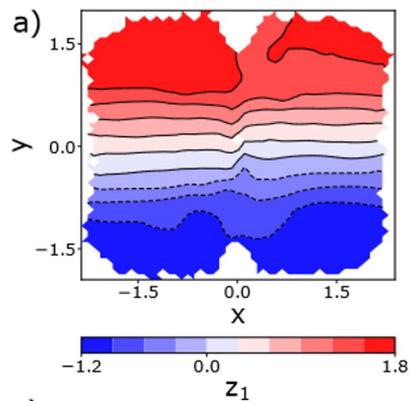


Energy-Entropy for Some Simple Systems

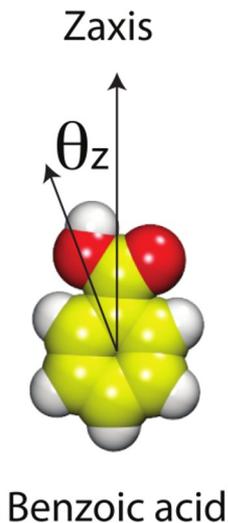
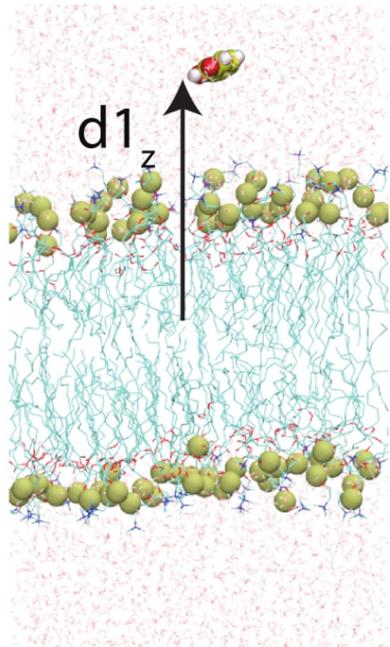


$$U(x,y) = h_x (x^2 - 1)^2 + (h_y + a((x, \delta))) (y^2 - 1)^2$$

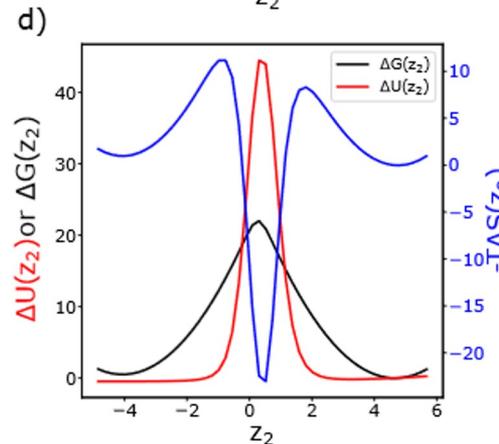
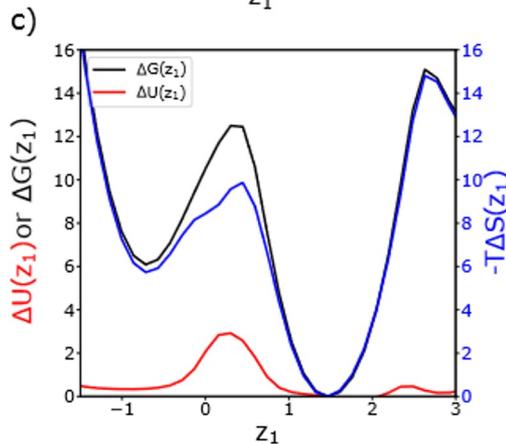
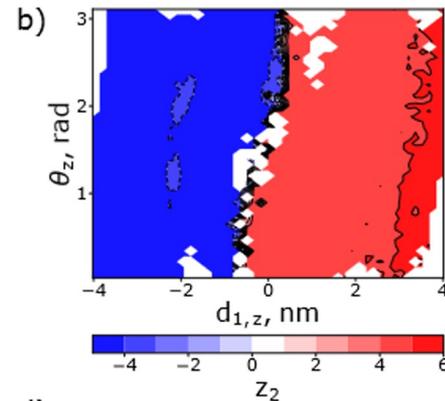
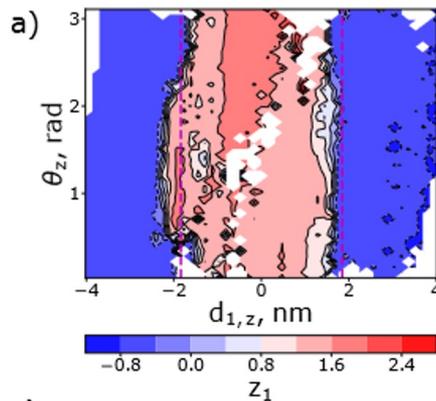
“Temperature switch”



Energy-Entropy for Some Simple Systems



Benzoic acid



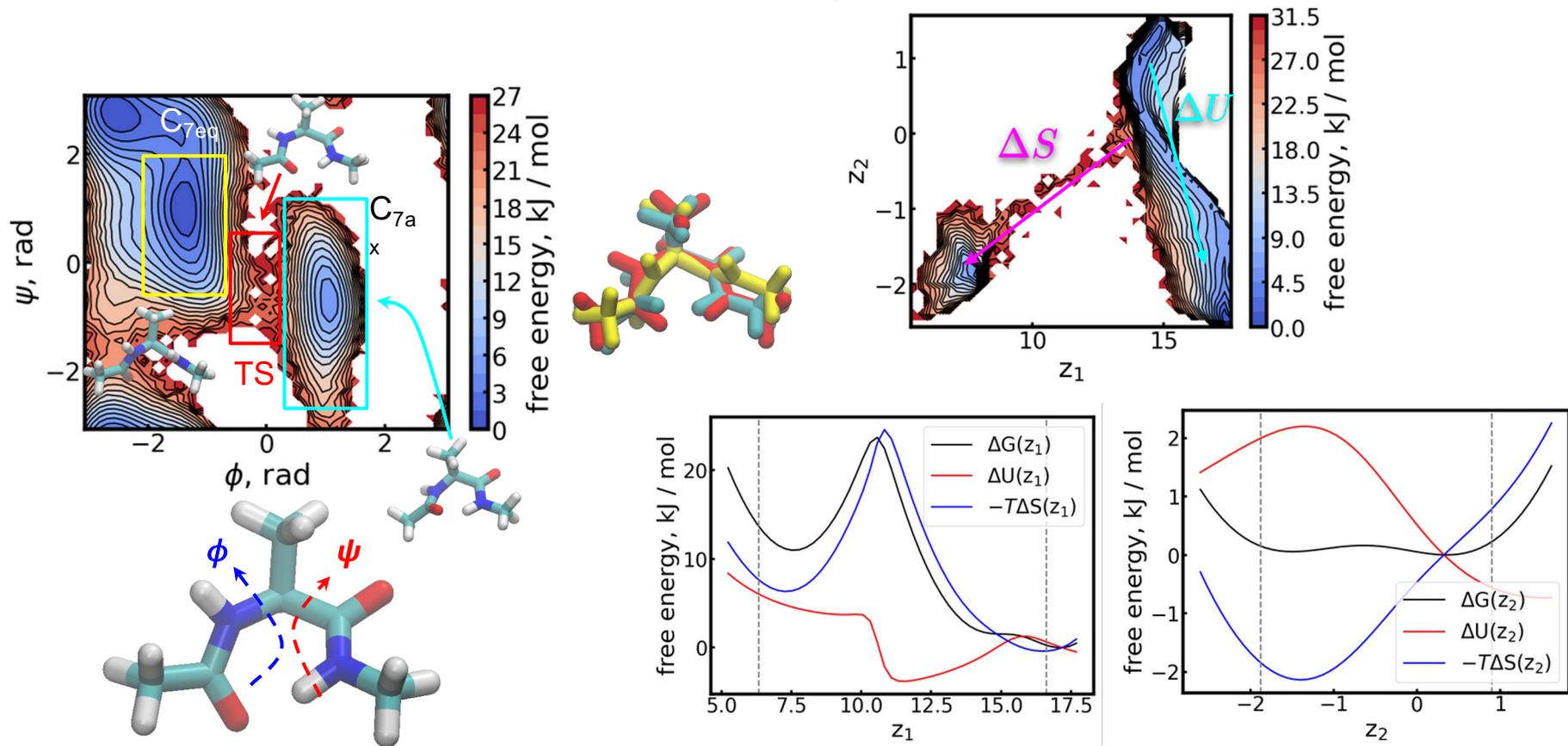
Forcing the Latent Space to Learn Thermodynamics

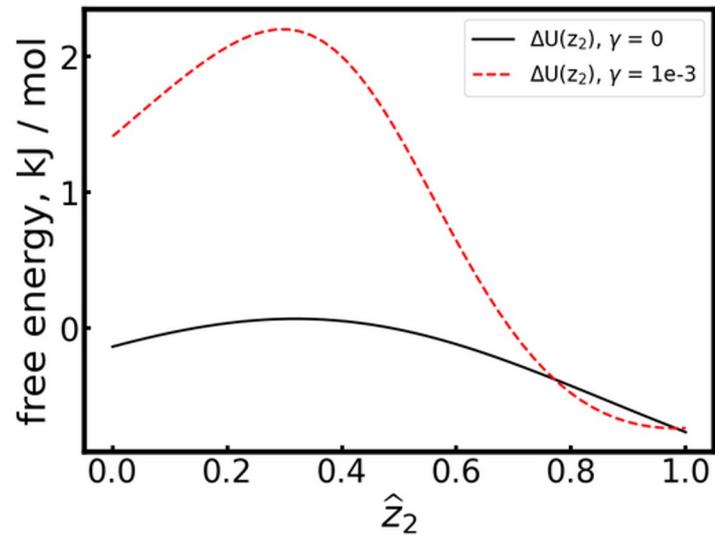
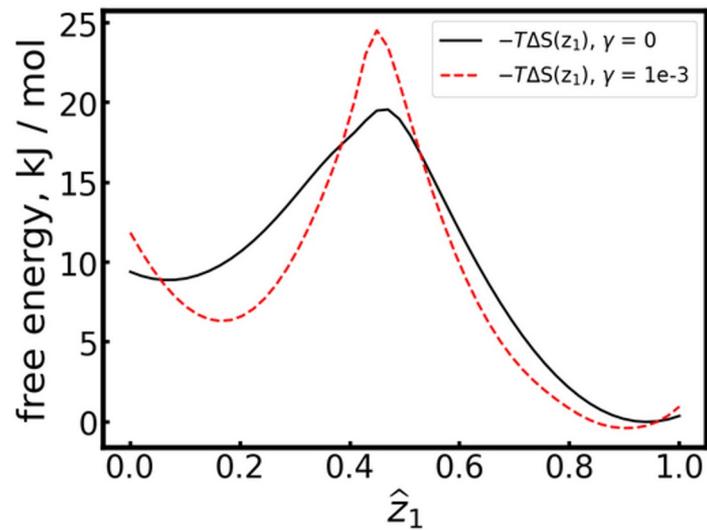
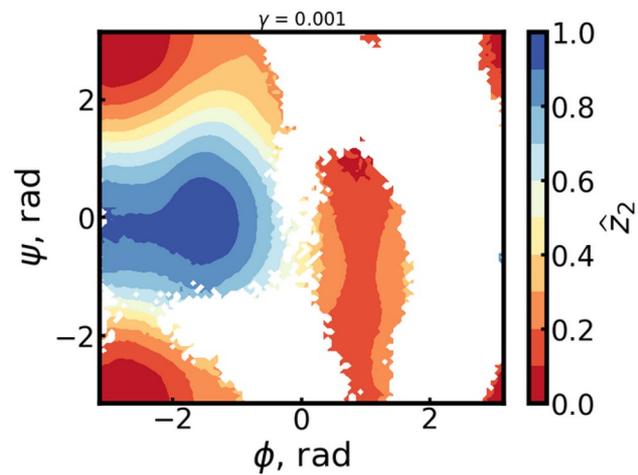
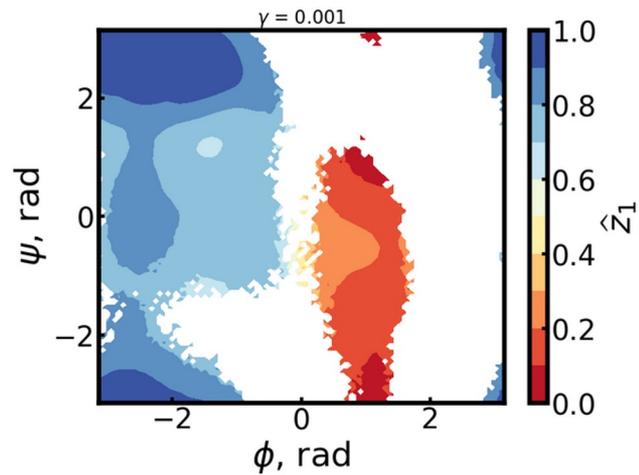
- Add an extra regularization term to the loss function to force the latent space to learn the energy and energy pathways for an arbitrary system.

$$\mathcal{L} = \sum_{n=1}^N \left[\log(p(\mathbf{y}^{n+s} | \mathbf{z}^n)) - \beta \log\left(\frac{p(\mathbf{z}^n | \mathbf{X}^n)}{p(\mathbf{z}_\theta)}\right) \right] + \gamma f(\mathbf{z})$$

$$f(\mathbf{z}) = \max_{\text{int}(z_1)} (-T\Delta S(z_1)) + \max_{\text{int}(z_2)} (\Delta U(z_2))$$

Case Study: Alanine Dipeptide in Vacuum





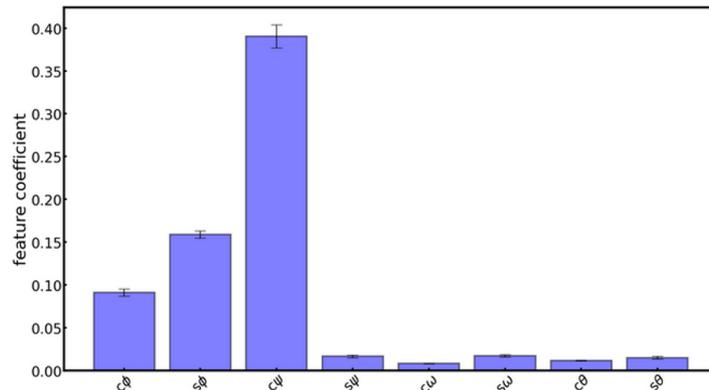
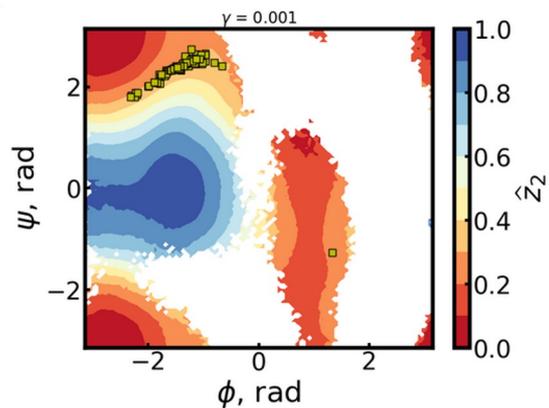
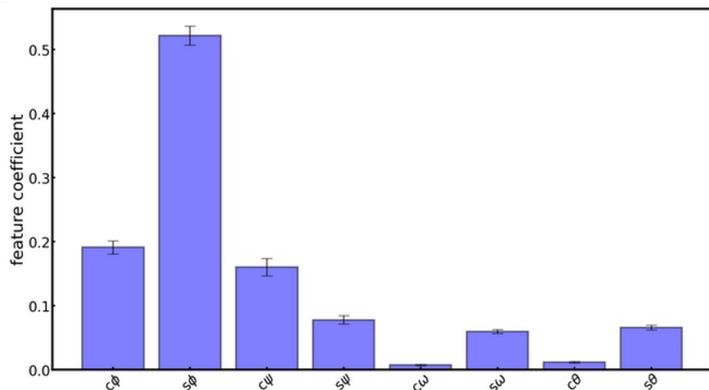
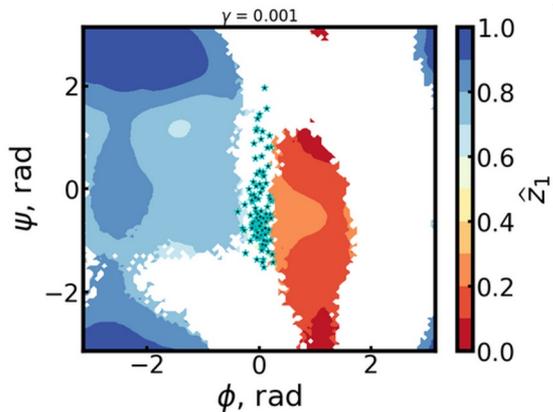
Interpretation

Use an in-house method called TERP¹:

“Unfaithfulness”

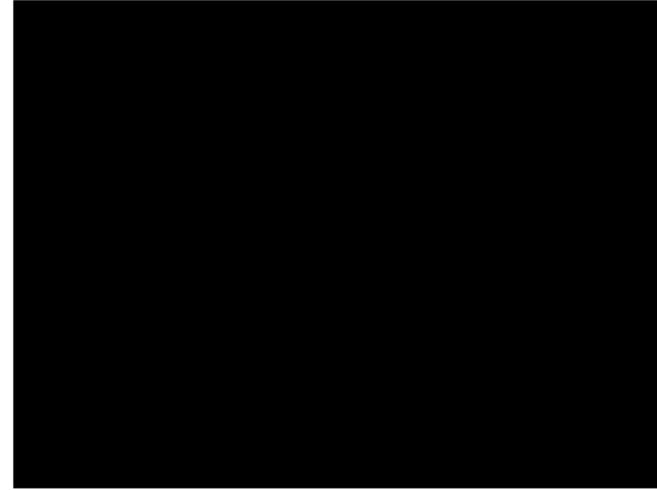
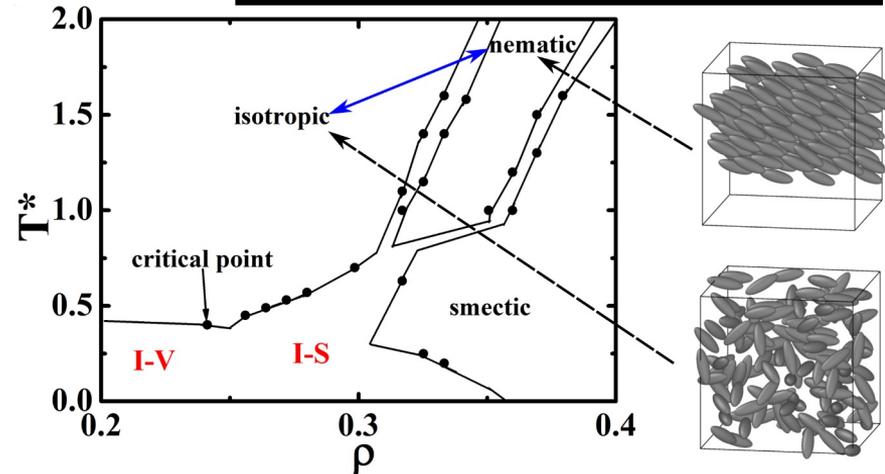
“Simplicity”

$$\zeta[f, d] = U[f, d] - \theta S[f]$$



Conclusions and Coming Attractions

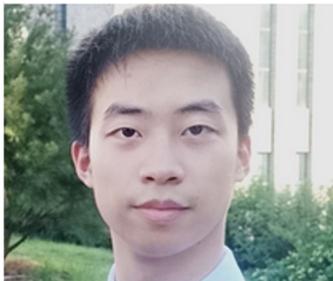
- SPIB can discover an effective low-dimensional latent space for crossing energetic and entropic barriers.
- For alanine dipeptide in vacuum, energy and entropy barriers can be enhanced by adding a regularization term to the loss function.
- We look to apply this method to more complicated soft matter systems (hydrophobic ligand binding, liquid crystal models) in the near future.



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Pakora



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XSEDE

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