# Interpolation of Molecular Dynamics with Bi-Directional Neural Networks

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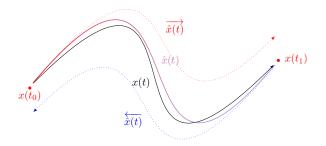
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- Can we predict MD trajectories directly with a neural network (NN)?
- Use NN to predict solver for MD equations
- NN are trained on predicting change in position and momentum
- But we still want the accuracy of real MD simulations?
- Use coarse grained MD simulation to predict initial and final conditions of differential equation governing MD trajectories
- Reconstruct high dimensional components to interpolate smartly with NN between coarse grained solution

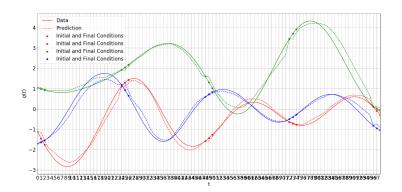
- ullet Given true dynamics f learn approximate dynamics  $f_{ heta}$  with NN
- Train  $f_{\theta}$  to predict the true solutions x(t)
- o Integrate approximate dynamics  $f_{\theta}$  to obtain approximate solution  $\hat{x}(t)$
- Use coarse grained MD simulation to predict initial and final conditions of differential equation governing MD trajectories
- Reconstruct high dimensional components to interpolate smartly with NN between coarse grained solution

- Predict forward solution  $\hat{x}(t)$  and backward solution  $\hat{x}(t)$  with the same dynamics  $f_{\theta}$
- Use adiabatic connection to interpolate  $\overrightarrow{\hat{x}(t)}$  and  $\overrightarrow{\hat{x}(t)}$  to  $\hat{x}(t)$

$$\hat{x}(t) = (1 - \lambda(t)) \ \overrightarrow{\hat{x}(t)} + \lambda(t) \ \overleftarrow{\hat{x}(t)}$$
 (1)



 Unidirectional LSTM architecture for Benzene MD trajectory interpolating over 20 time steps



- Bidirectional LSTM architecture for Benzene MD trajectory interpolating over 20 time steps
- Final condition and additional bidirectional training smooth trajectories significantely

