

Kernel-Based Multi-channel PolyCovNet

AI Hackathon Challenge I



Team Borides

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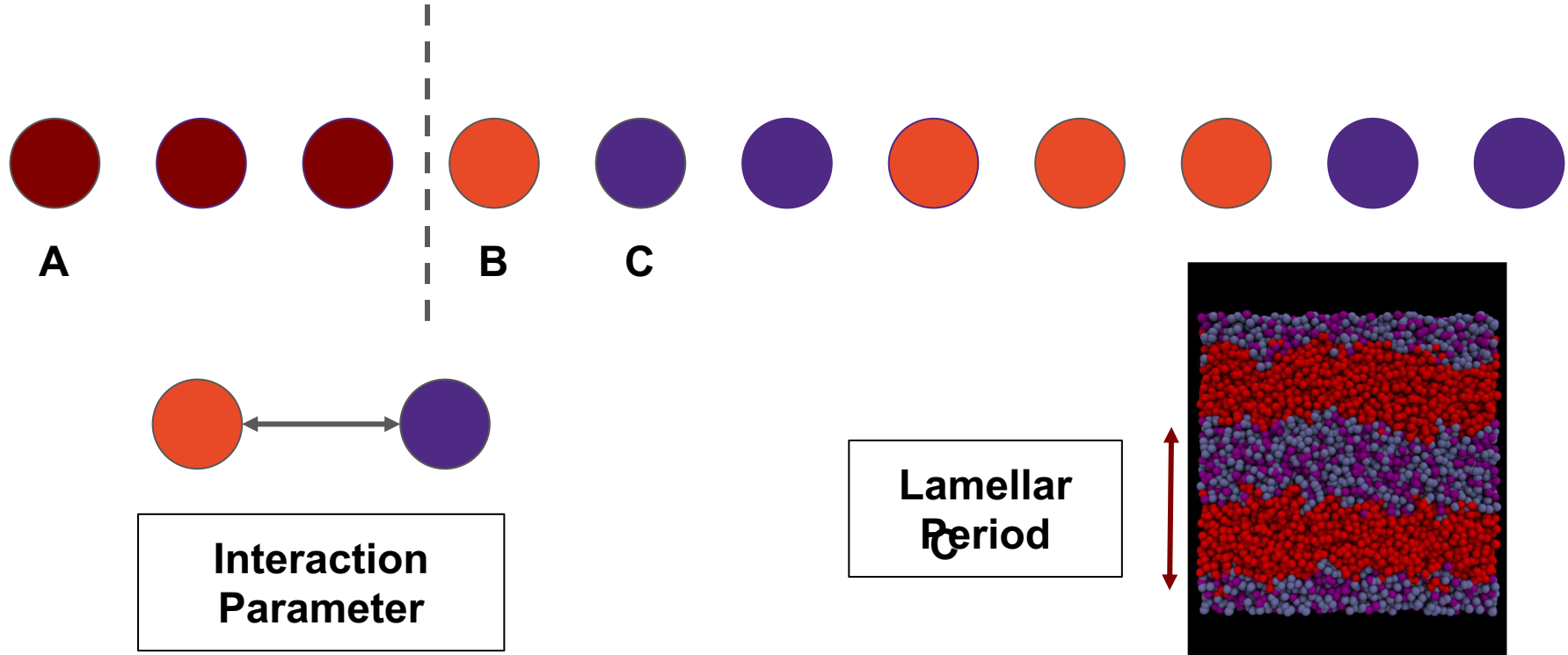
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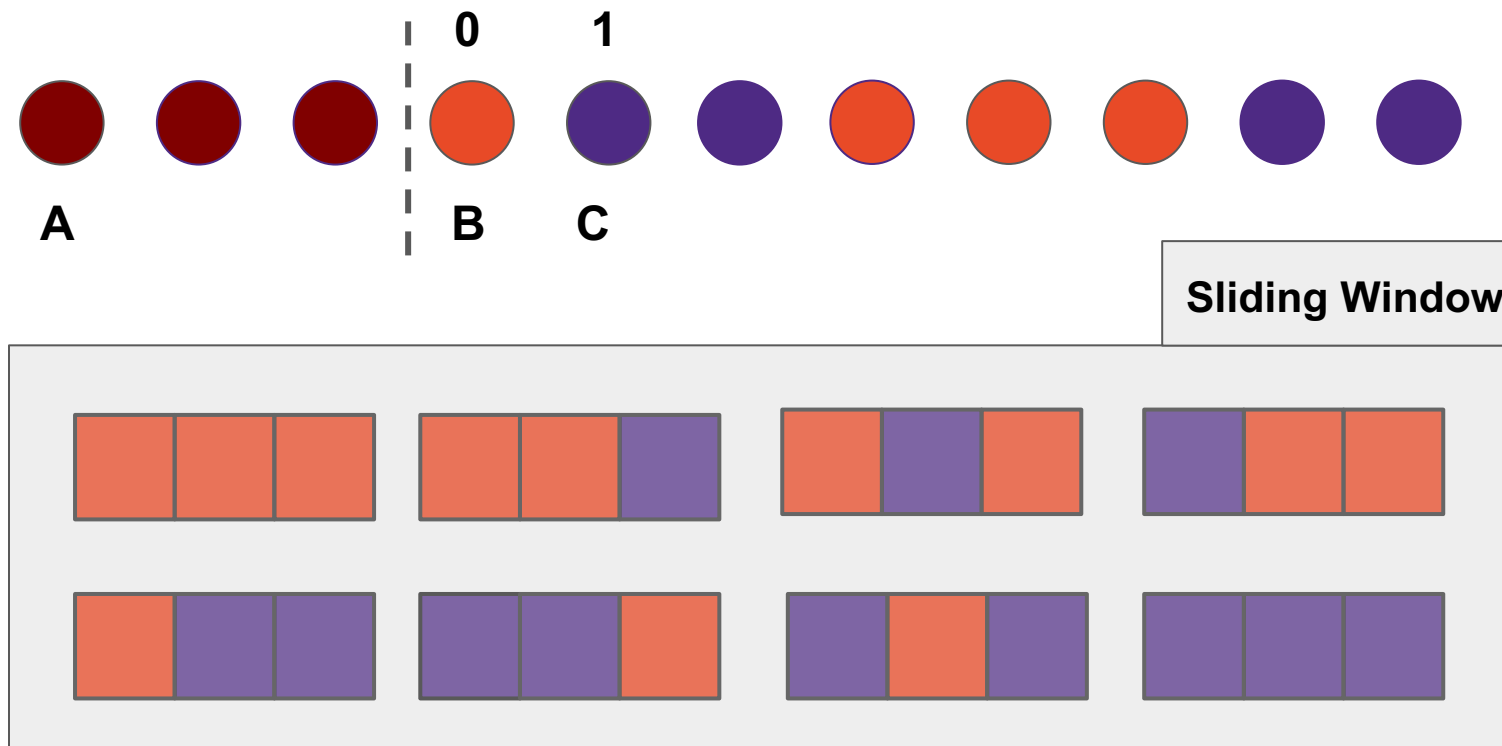
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Problem Restatement

Predicting the lamellar period from monomer sequences and interaction parameters

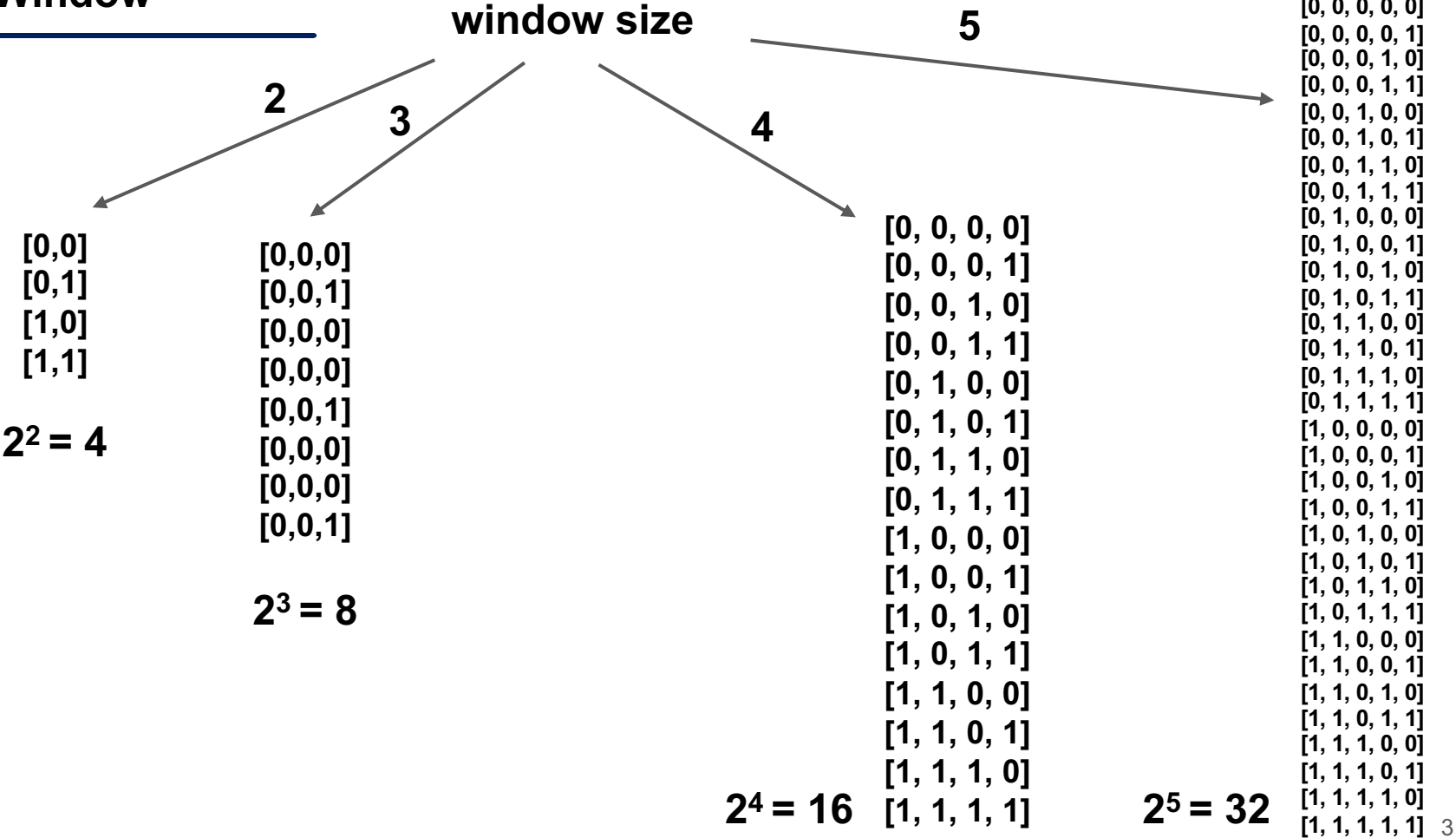


Sliding window - extract monomer sequence features



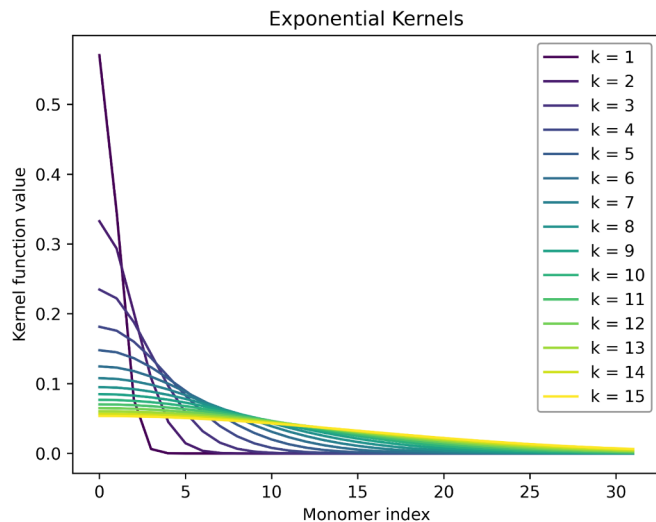
Sliding Window

Algorithm

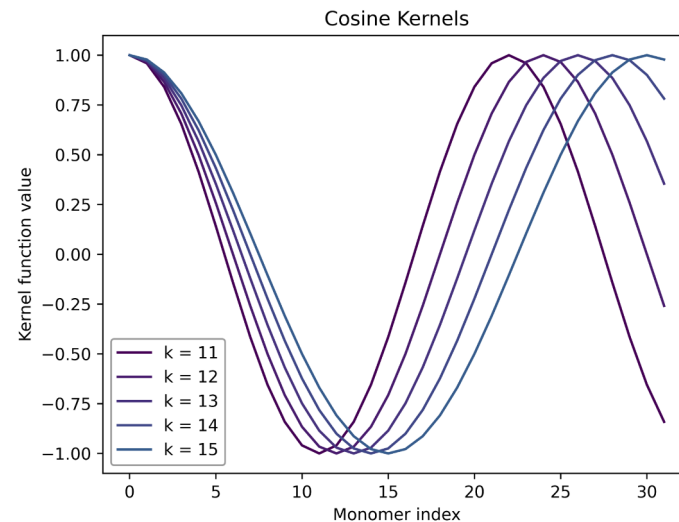


Non-linear kernel functions - preprocess monomer sequences

1. Exponential kernels $\exp(\frac{x^2}{2k^2})$

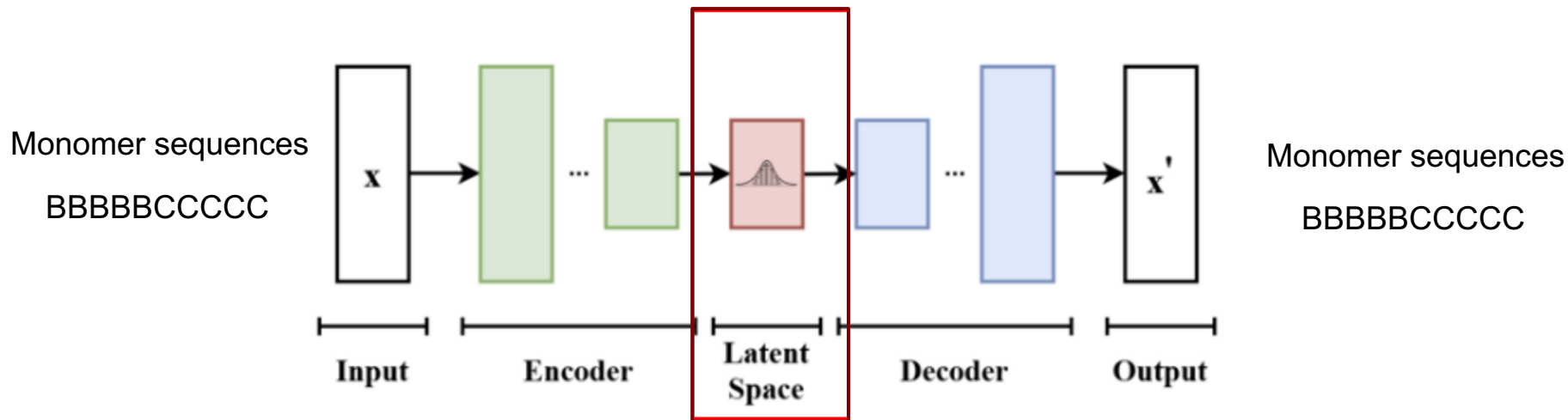


2. Cosine kernels $\cos(\frac{\pi x}{k})$

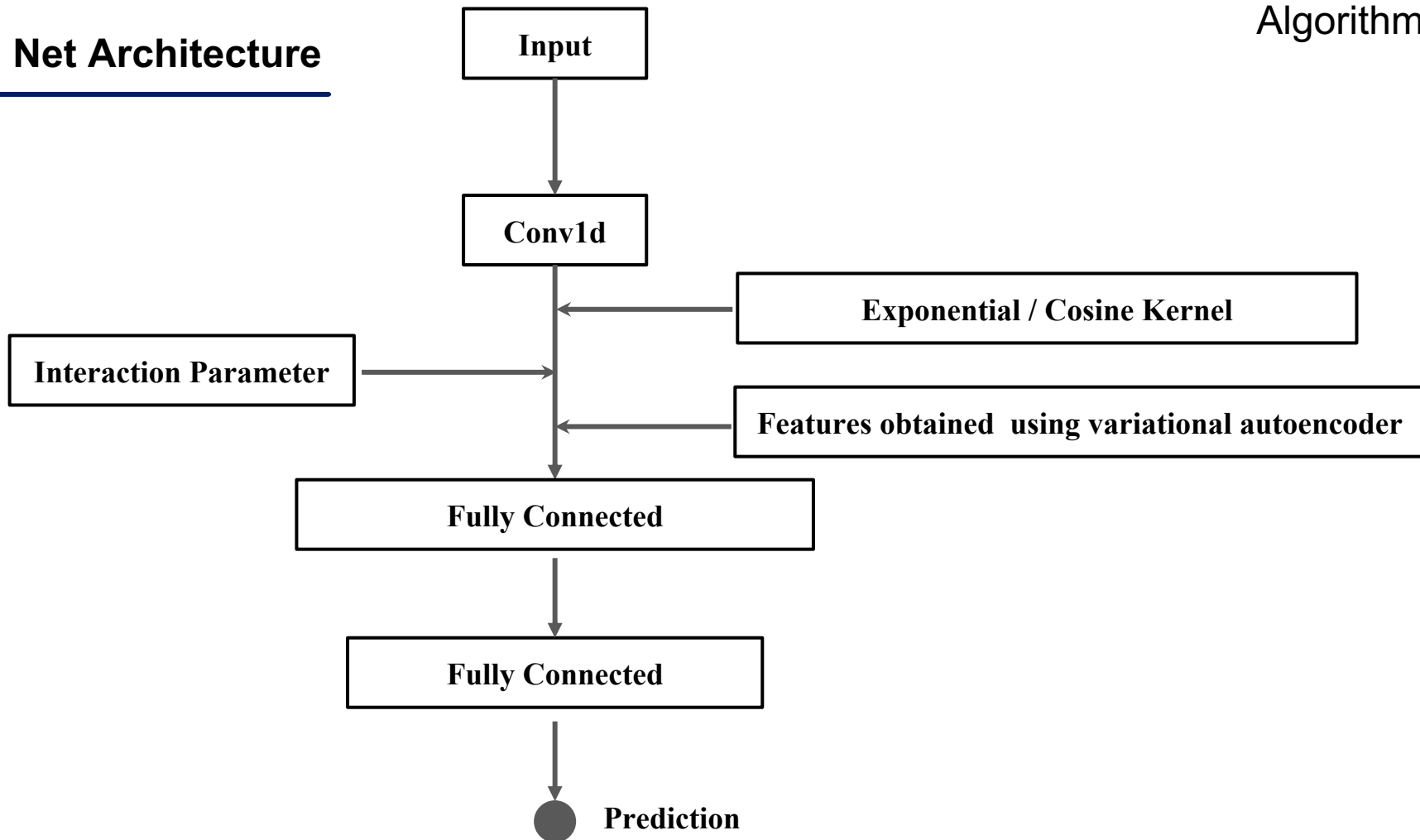


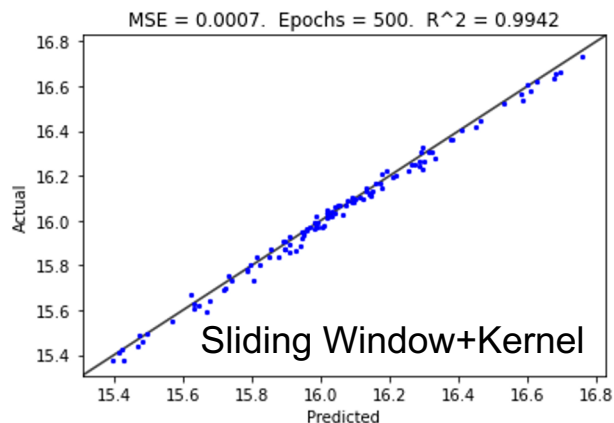
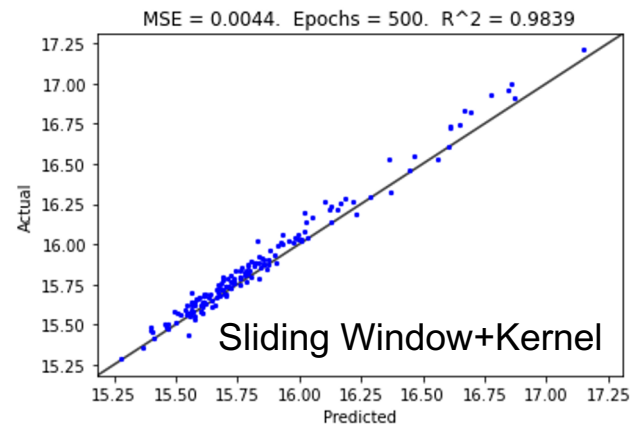
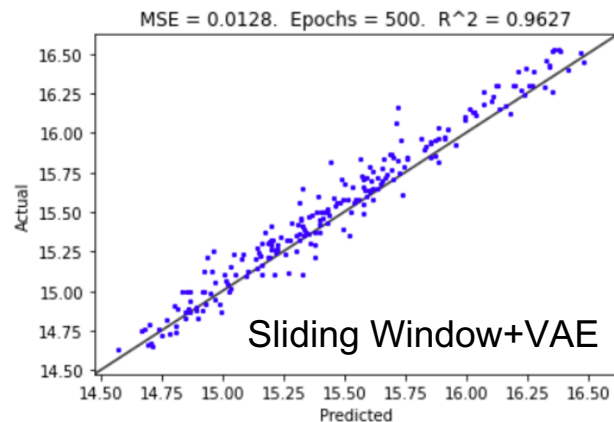
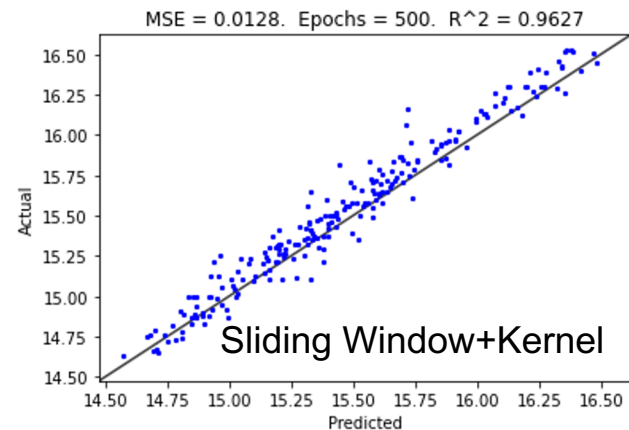
Applying the above kernels on monomer sequences ➡ Non-linearity

Variational autoencoder - extract features from monomer sequences



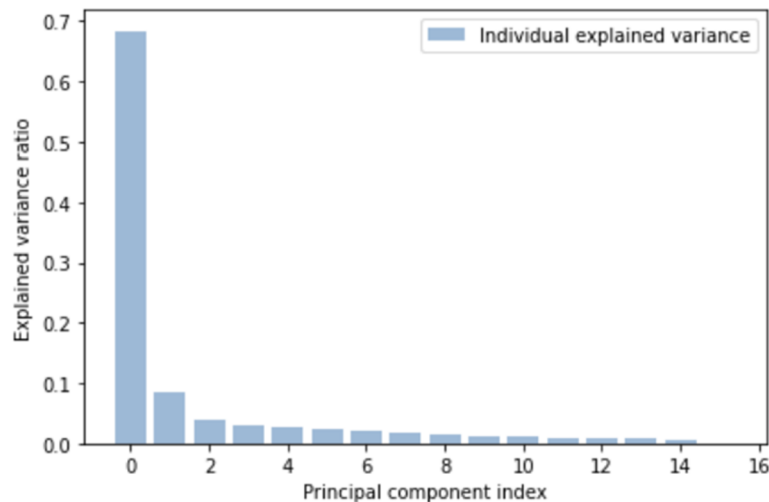
Reduced feature vectors



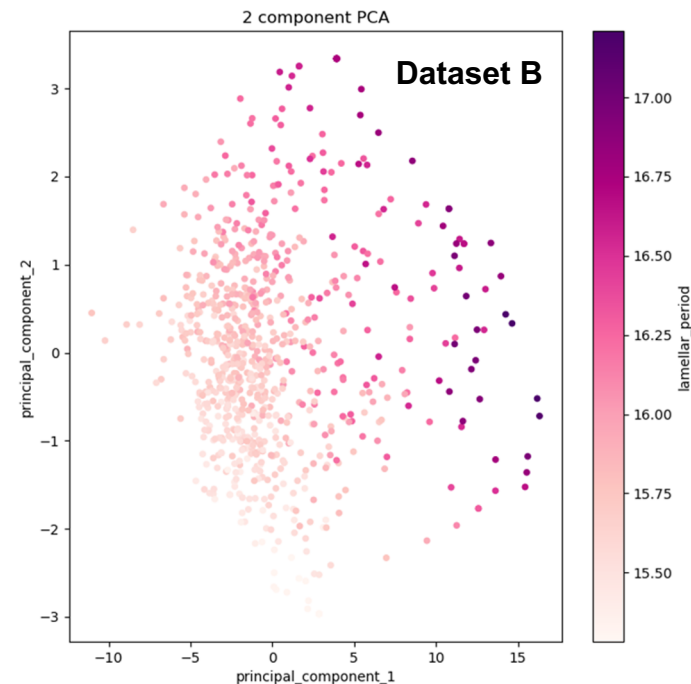
**DataSetA****DataSetB****DataSetC****DataSetD**

500 epoch
0.01 learning rate

Principal component analysis on the latent space



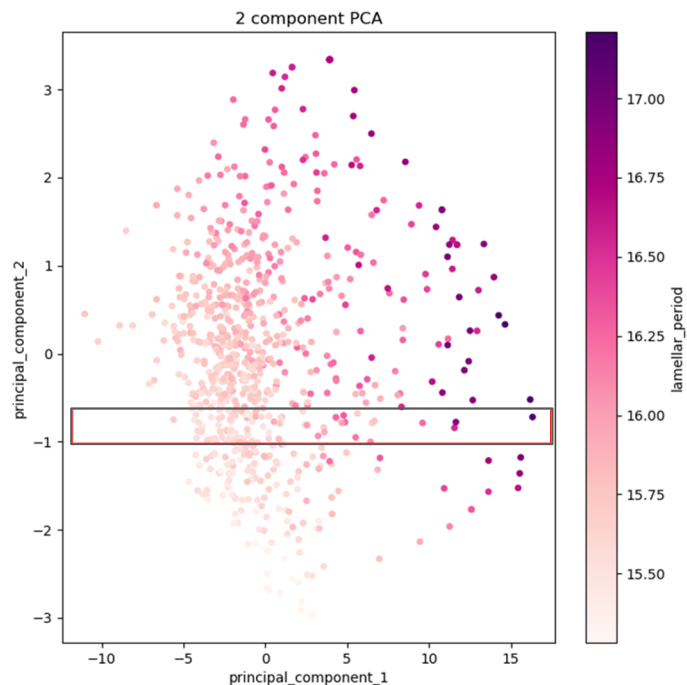
Principal component 1 dominates



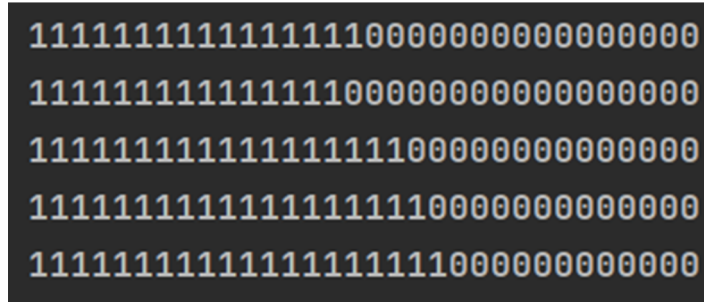
Larger principal component 1

Larger lamellar period

Blockiness is important to result in a high lamellar period



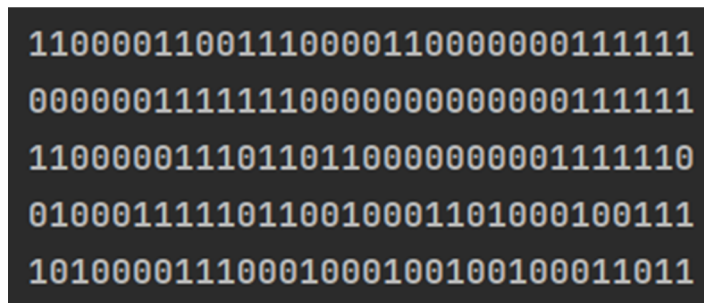
Higher principal component 1



Blockiness

0.875000
0.874510
0.873016
0.870445
0.866667

Lower principal component 1



Blockiness

0.498039
0.740891
0.372549
0.000000
-0.036437

Computational Efficiency (500 epochs)

Feature Generation	Time (min)
2-channel Sliding Window Features	0.5
Exponential / Cosine Kernel Features	0.08
VAE Features	30

Model Training / Validation	Time (min)
Training	1
Validation	0.02

* All runtimes are reported using ThetaGPU

Thank you!
