

Semi-Automated Bacterial Protein Dynamics Analysis

High throughput PlzC oscillation tracking in *V. Cholerae*



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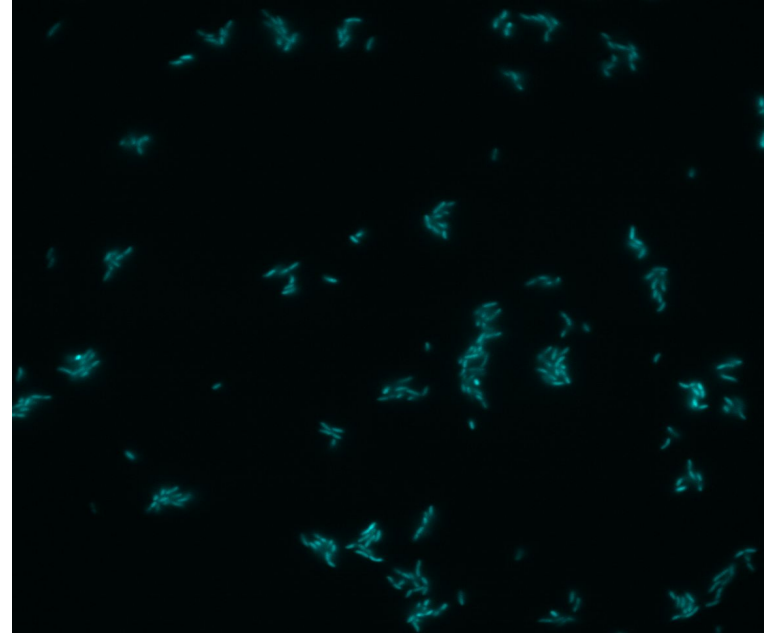
UCLA
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School of Engineering

Biofilms are colonies of aggregating bacteria that adhere to a surface

Biofilms have a widespread impact in medicine and industry

- Allow pathogens to survive in chronic diseases
- Allow formation of resistant enclaves of bacteria on and around implants
- Affects water transport and food industries



100 μ m

The economic significance of biofilms

- In 2017 the global expenditure on wounds in healthcare was \$7.8T, of which it is estimated that \$281B corresponds to biofilms (*Nature*)
 - Annual cost of revision surgery due to biofilm-mediated infections: \$7.8B globally
 - Annual market estimate of surfaces resistant to microbial contamination: \$7.1B
 - Foodborne pathogens and biofilms are responsible for \$78.6B lost every year
- The market of biofilms in wastewater treatment is an est. \$313B globally
- Increased hospitalization and cleaning needs as a result of the COVID-19 pandemic have increased the market by an additional \$5T

We developed a machine learning model to classify time-position plots of intracellular protein dynamics

Introducing a unique solution to a data bottleneck

- Understanding *why* biofilms form is crucial to preventing formation
- Observing the movement of protein within bacteria allowed us to create a machine learning model
- This model can be generalized to other bacterial species and other single-protein tracking applications



dataset.jl



tune.jl



crop.jl



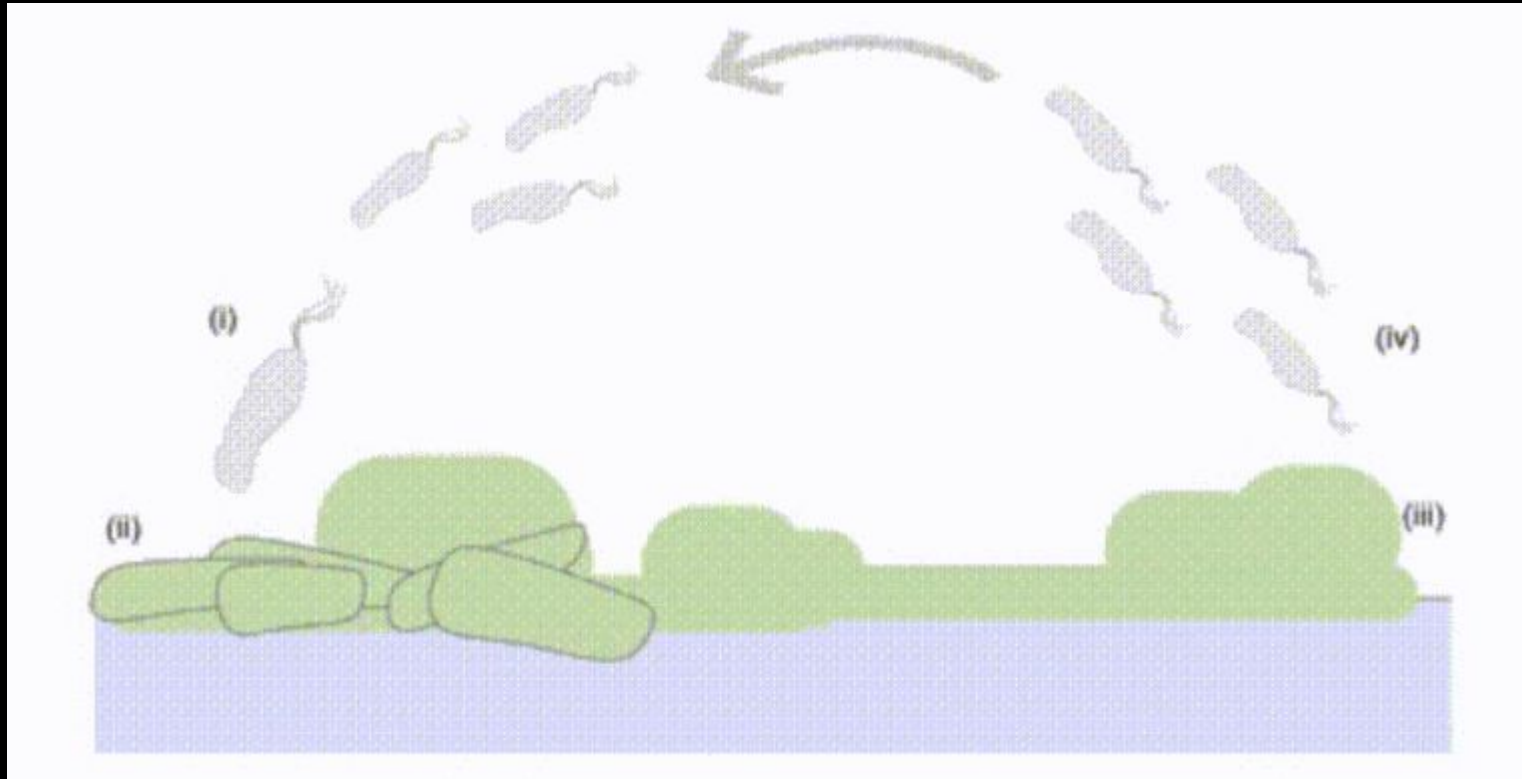
pick.jl



trace.jl

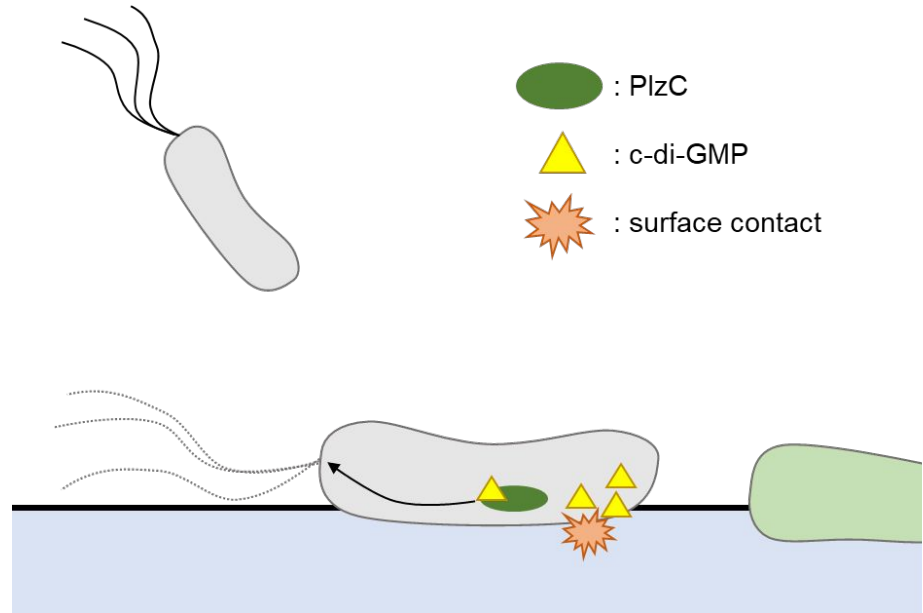


analysis.jl

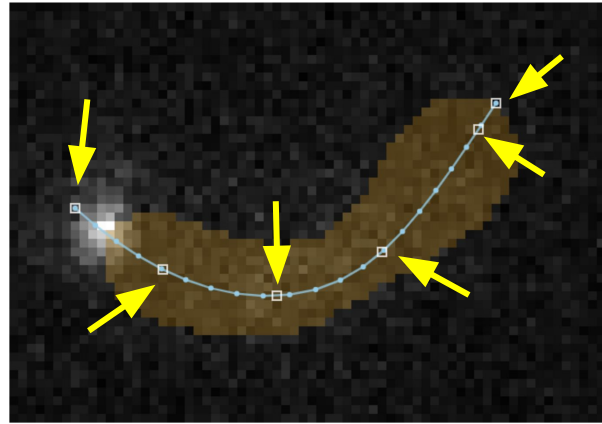
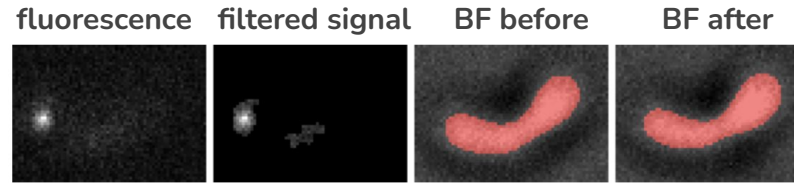


PlzC-protein has an unknown role in biofilm formation

- **Known:** c-di-GMP concentration level is related with biofilm formation
- **Hypothesis:** PlzC is a carrier protein for c-di-GMP, which is correlated with biofilm formation
- Tracking PlzC **movement** using fluorescence microscopy is essential to understanding its relationship with c-di-GMP and biofilm formation



Tracking PlzC-protein with fluorescence microscopy

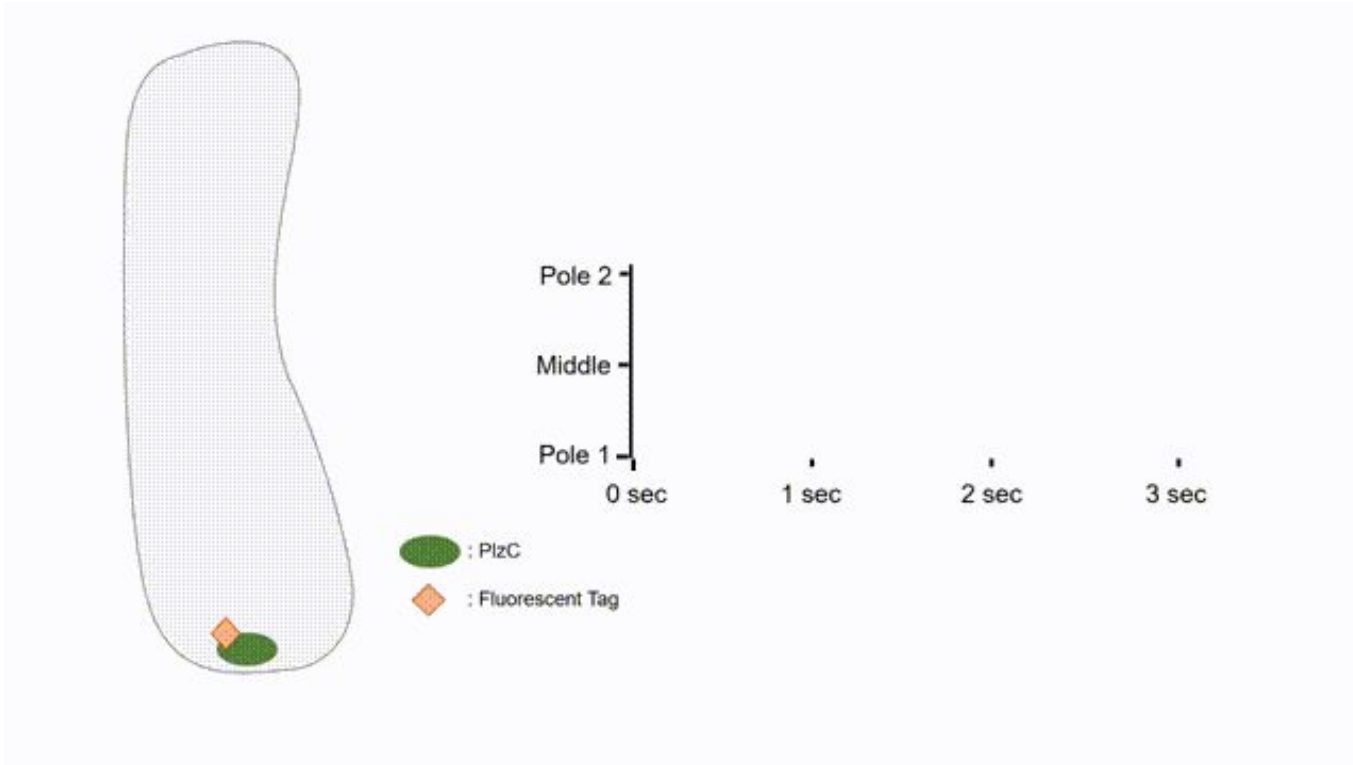


Frame #1

0.25 μm

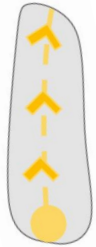
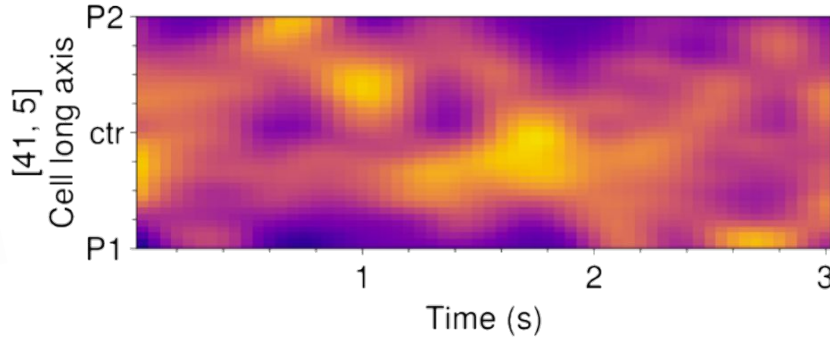
WT, PlzC-tYFP

Visualizing PlzC movement in *V. Cholerae*

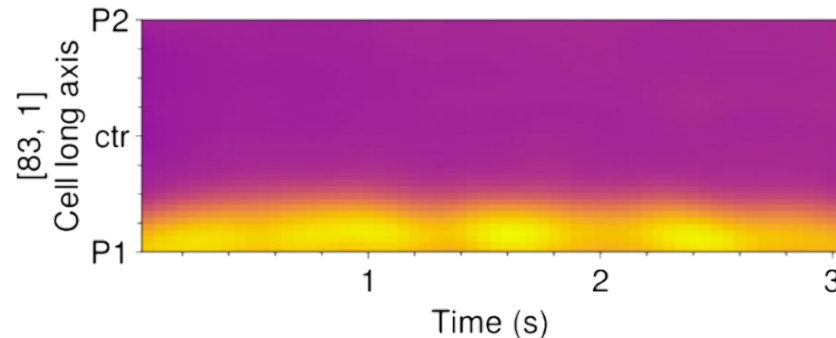


Kymographs visualize PlzC movement along splines

PlzC directed motion early/before surface sensing



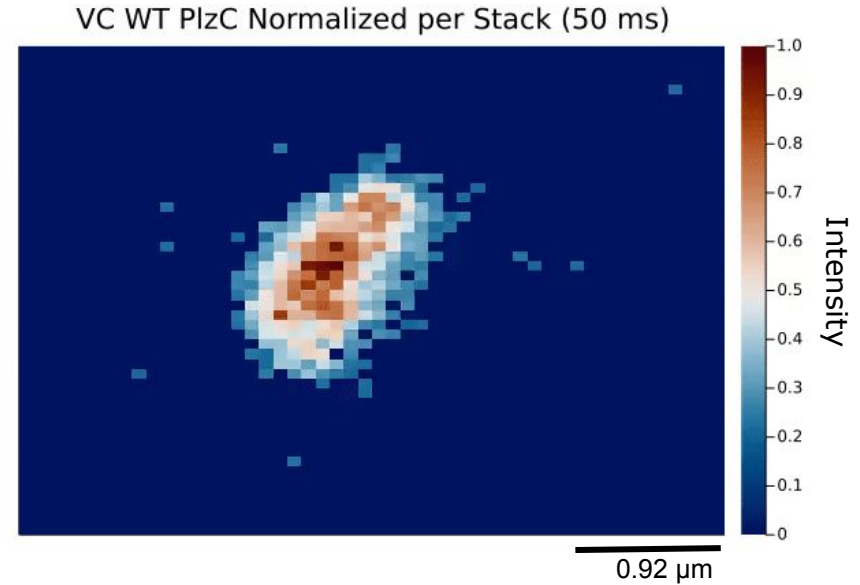
PlzC polar localization in biofilms



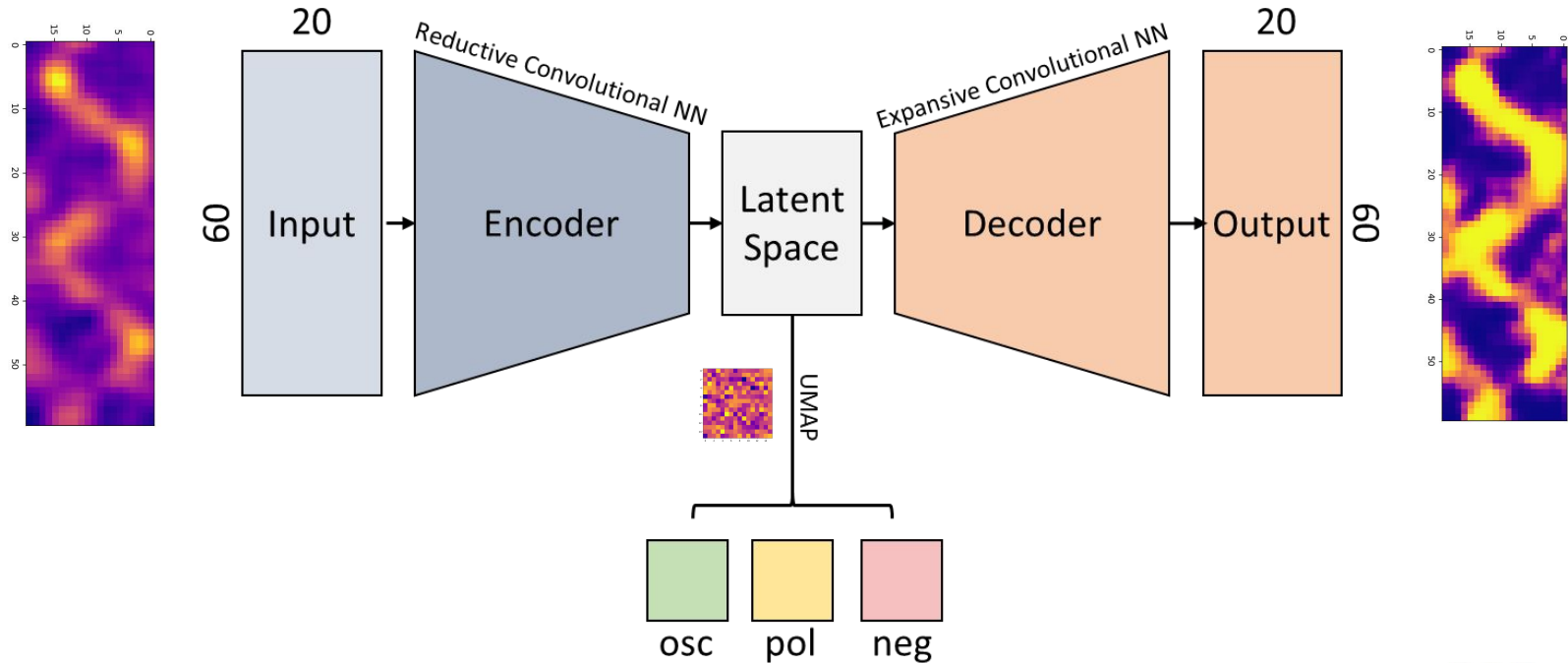
- Each cell crop has a kymograph of PlzC movement
- Classify distinctions in intracellular PlzC movement across varying strains
- **Scale of data too large and complex** datasets to find patterns manually

An automated and general workflow for large scale protein movement analysis

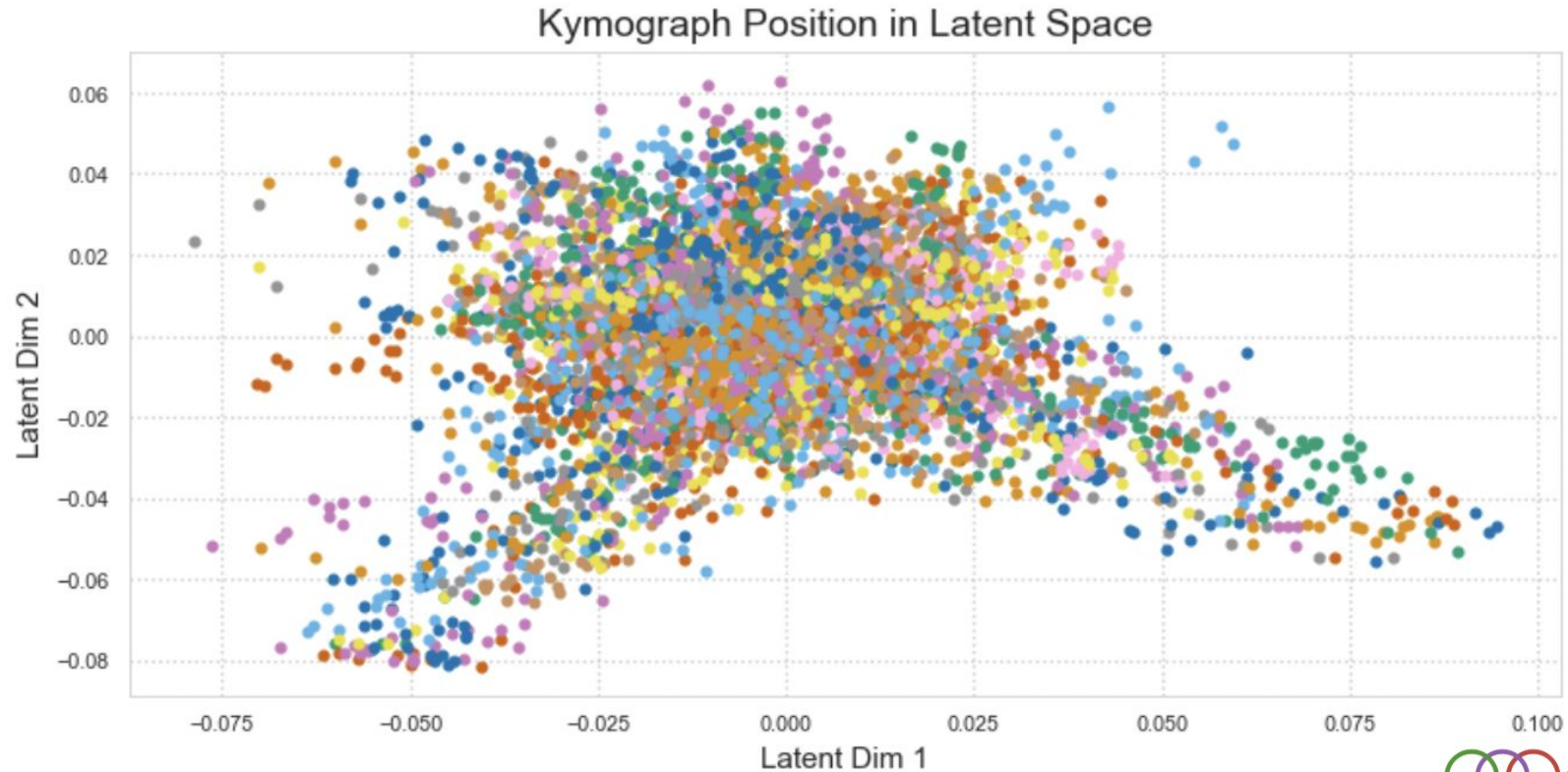
- Tracking intracellular movement of PlzC protein
- Machine learning allows for analysis of large datasets
- Observation of patterns to draw hypotheses about role in biofilm formation



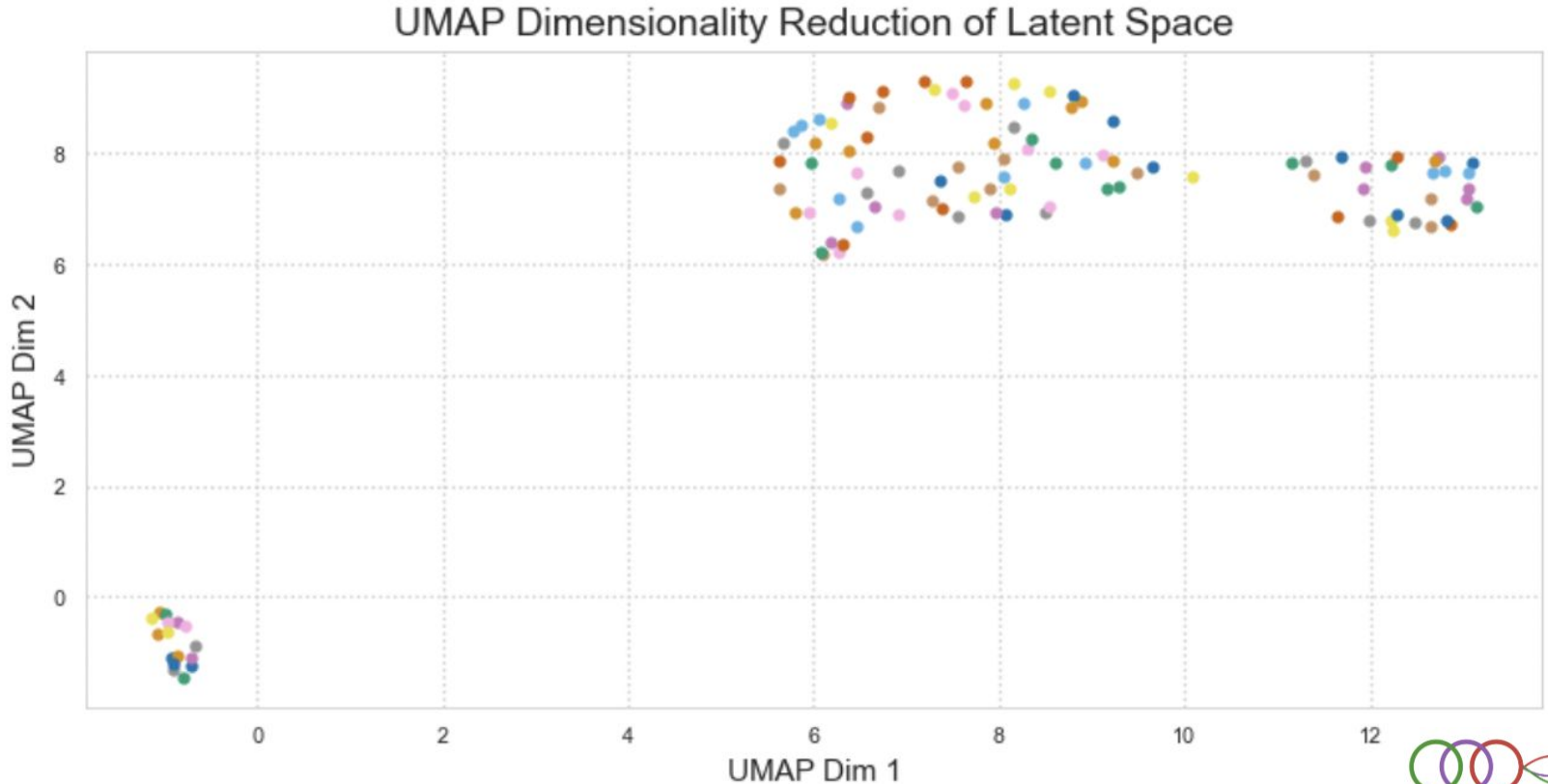
Our Variational Autoencoder (VAE) allows us to classify kymographs



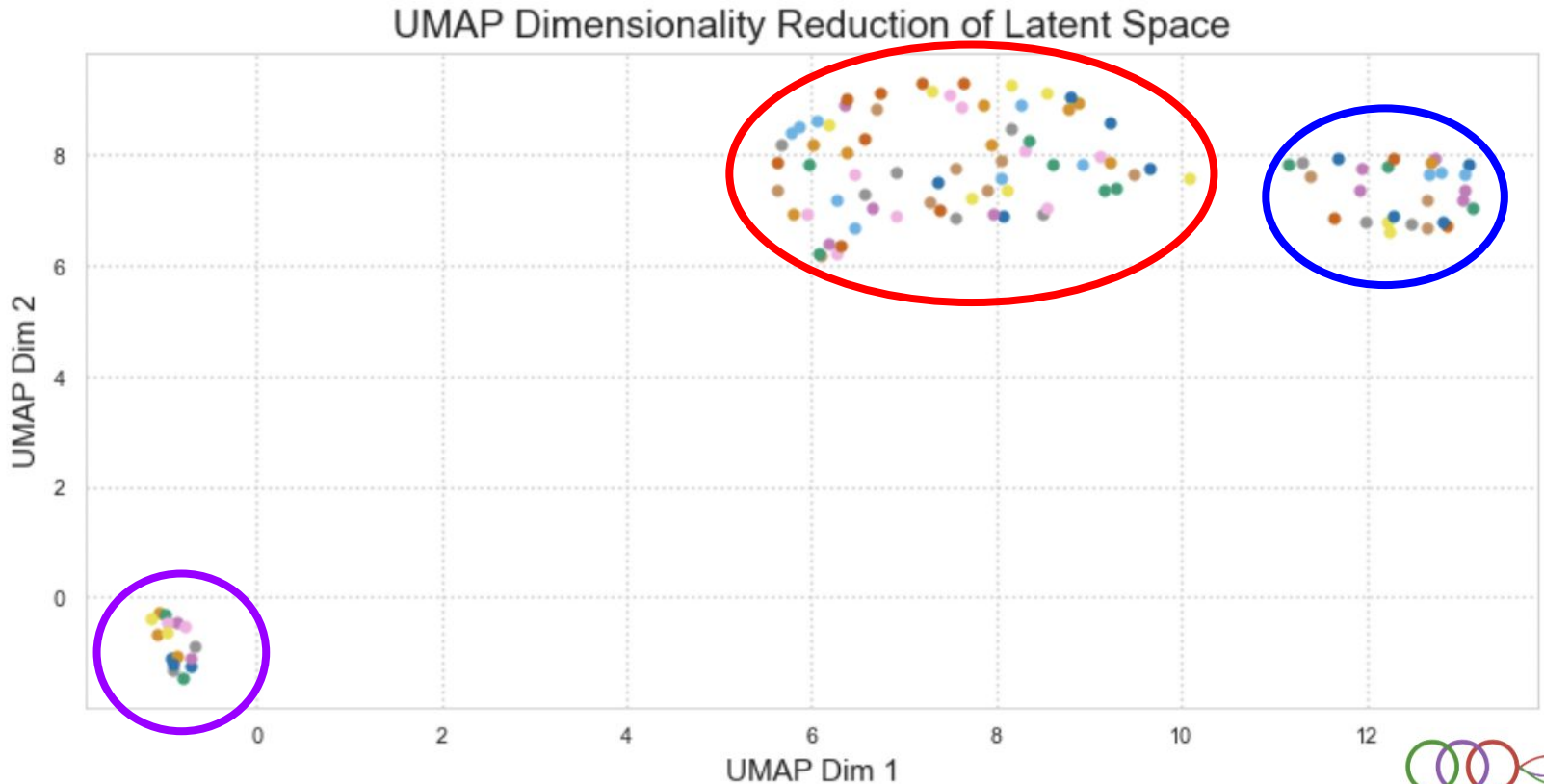
Visualizing the raw latent space is too busy



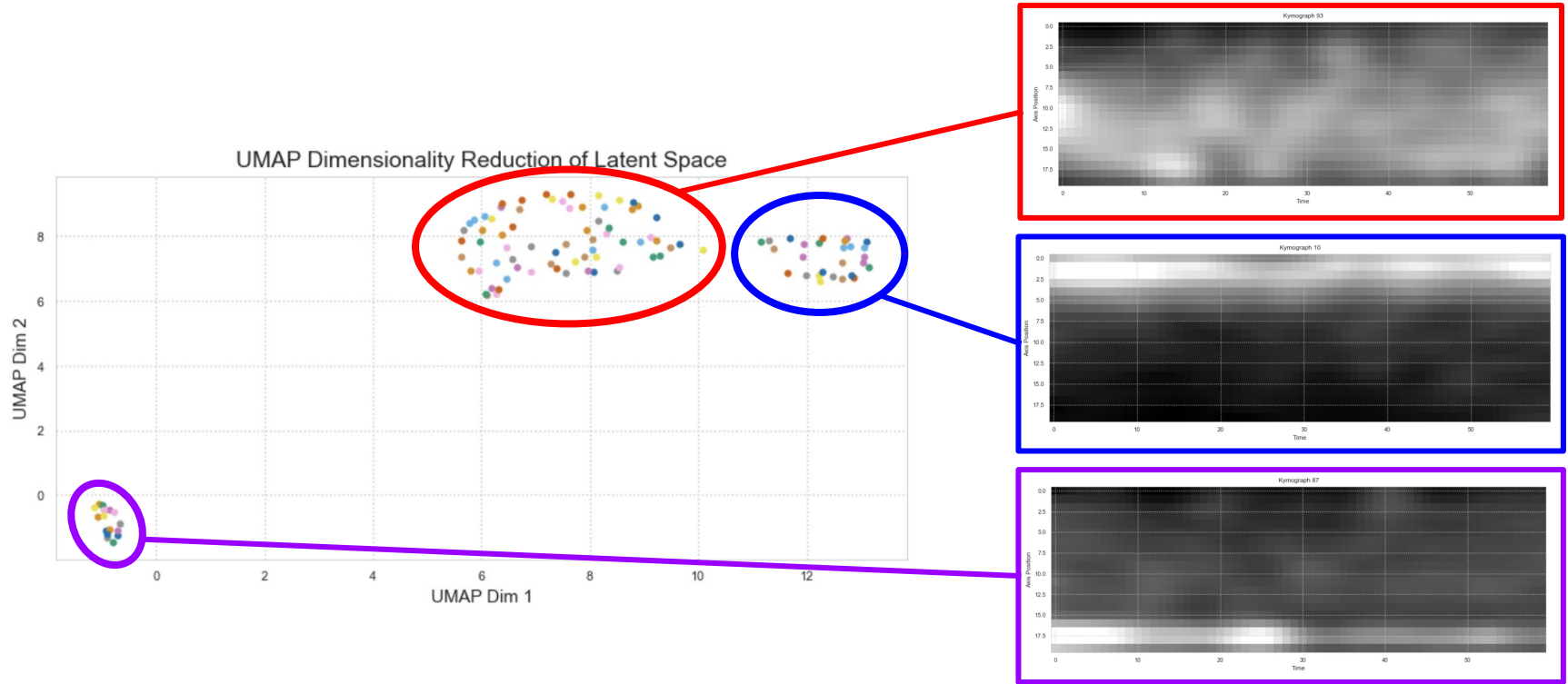
UMAP dimensionality reduction reveals grouping



UMAP dimensionality reduction reveals grouping

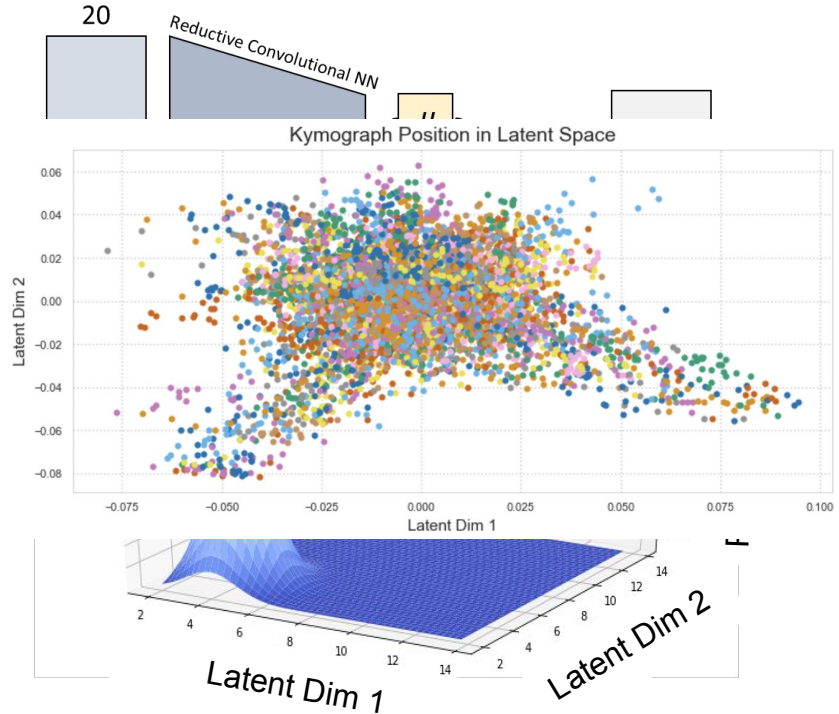


UMAP dimensionality reduction reveals grouping

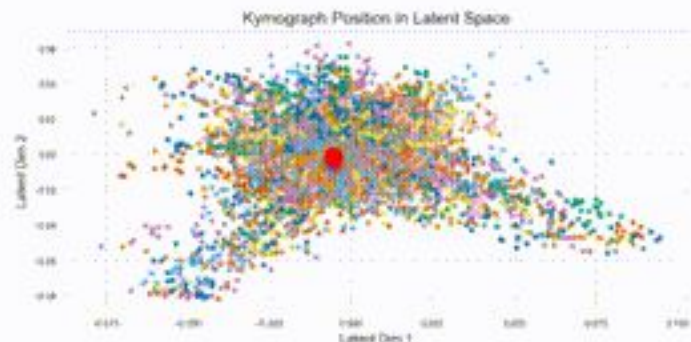
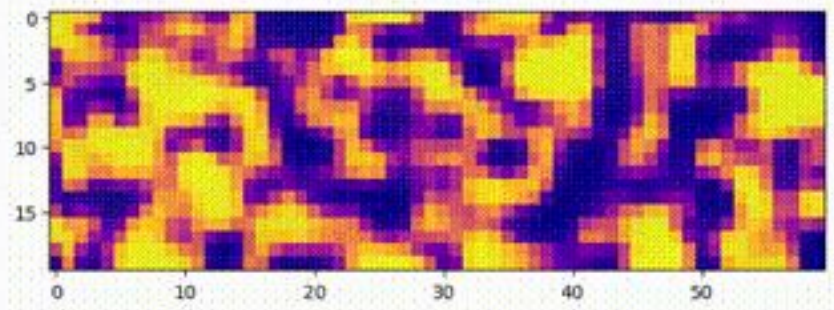
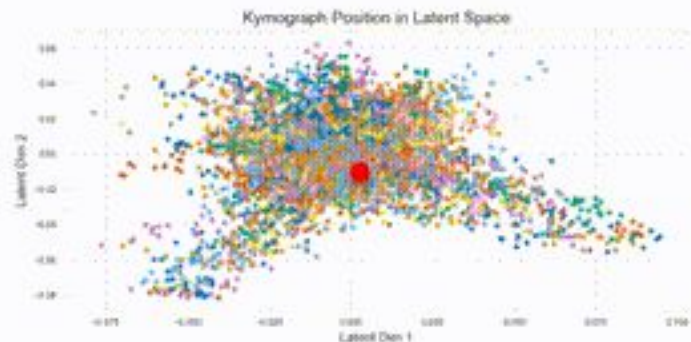
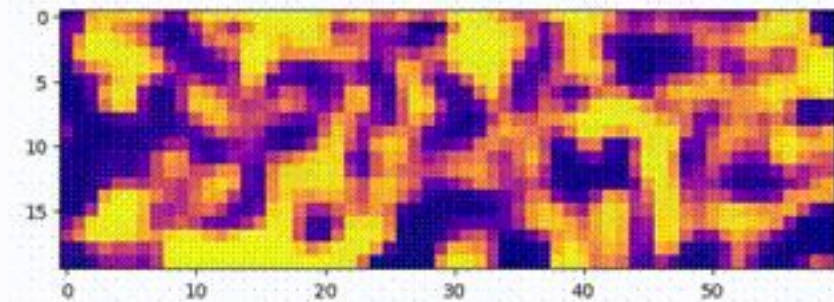


VAEs preserve latent relationships

- *Probabilistic* representation in lower dimensional space
- Extract **essential spatial characteristics** and map to latent space



Reconstructing the latent space



Summary

- Created a tool for high-resolution protein dynamics visualization and interpretation
 - Developed to investigate molecular machinery of biofilm formation in *V. cholerae*
 - Quantified and classified differences in fluorescence kymographs of super-diffusive PlzC movement
- **Generalizable** to any form of single-molecule fluorescence tracking on the microscopic scale

Acknowledgements

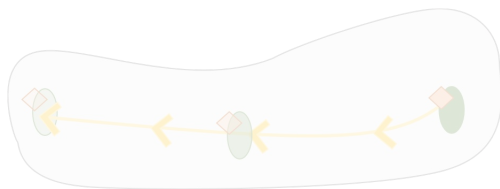
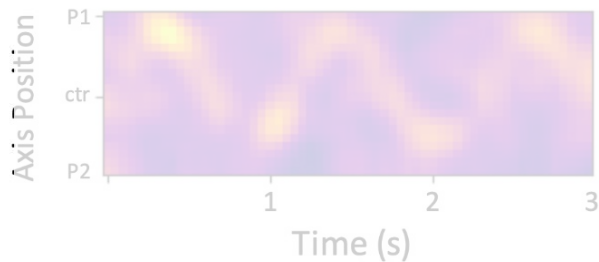
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And thank you to Dr. Gao, Salil, and Xiaoxi for their
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References

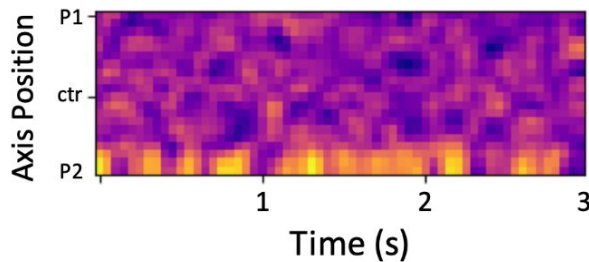
- Cámara, M., Green, W., MacPhee, C. E., Rakowska, P. D., Raval, R., Richardson, M. C., Slater-Jefferies, J., Steventon, K., & Webb, J. S. (2022). Economic significance of biofilms: A multidisciplinary and cross-sectoral challenge. *Npj Biofilms and Microbiomes*, 8(1), 1–8. <https://doi.org/10.1038/s41522-022-00306-y>
- Mark B. Abelson, M. D. (2012, September 6). *Of biomes, biofilm and the ocular surface*. Review of Ophthalmology. Retrieved July 4, 2022, from <https://www.reviewofophthalmology.com/article/of-biomes-biofilm-and-the-ocular-surface>
- Lopez, L. (2021, April 19). *Pseudomonas aeruginosa biofilms and their partners in Crime*. Institute for Bioengineering of Catalonia. Retrieved July 6, 2022, from <https://ibecbarcelona.eu/pseudomonas-aeruginosa-biofilms-and-their-partners-in-crime/>
- Pratt, Jason T., Rita Tamayo, Anna D. Tischler, and Andrew Camilli. "PilZ Domain Proteins Bind Cyclic Diguanylate and Regulate Diverse Processes in *Vibrio Cholerae*." *The Journal of Biological Chemistry* 282, no. 17 (April 27, 2007): 12860–70. <https://doi.org/10.1074/jbc.M611593200>.
- Berk, Veysel, Jiunn C. N. Fong, Graham T. Dempsey, Omer N. Develioglu, Xiaowei Zhuang, Jan Liphardt, Fitnat H. Yildiz, and Steven Chu. "Molecular Architecture and Assembly Principles of *Vibrio Cholerae* Biofilms." *Science* 337, no. 6091 (July 13, 2012): 236–39. <https://doi.org/10.1126/science.1222981>.
- Quan Xue and M. C. Leake, "A novel multiple particle tracking algorithm for noisy in vivo data by minimal path optimization within the spatio-temporal volume," 2009 IEEE International Symposium on Biomedical Imaging: From Nano to Macro, 2009, pp. 1158–1161, doi: 10.1109/ISBI.2009.5193263.

Going back to the biology

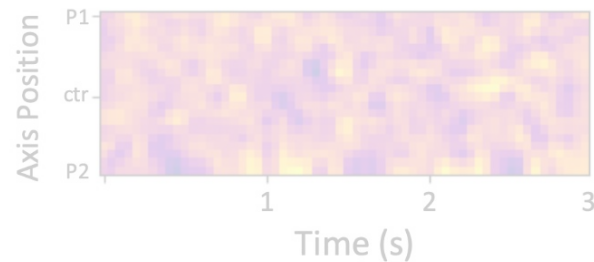
Oscillatory



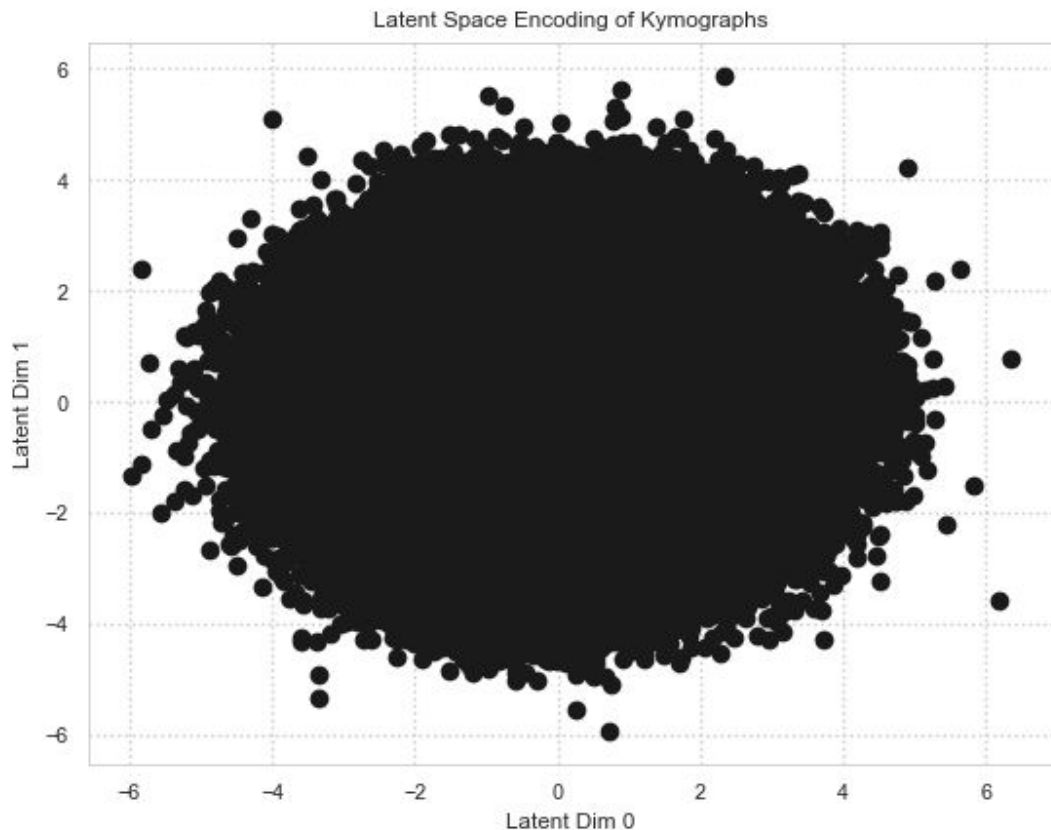
Polar



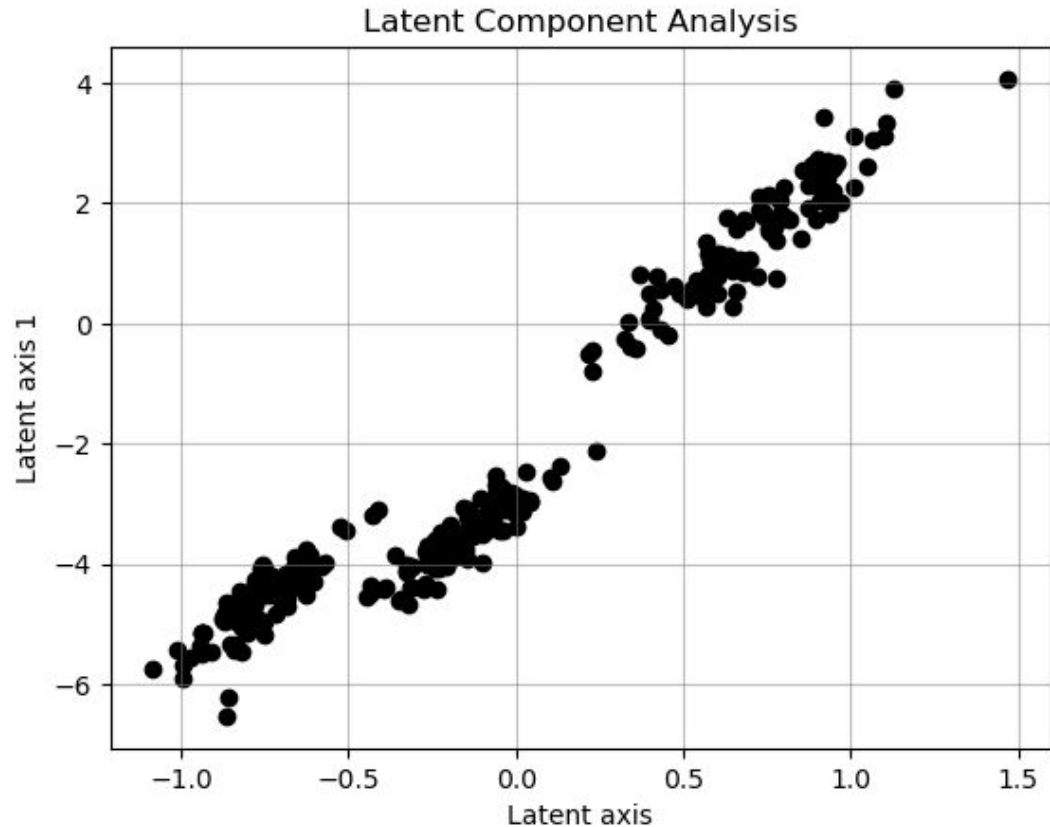
Negative



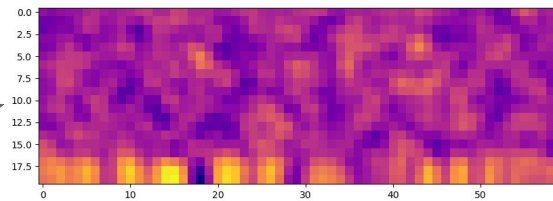
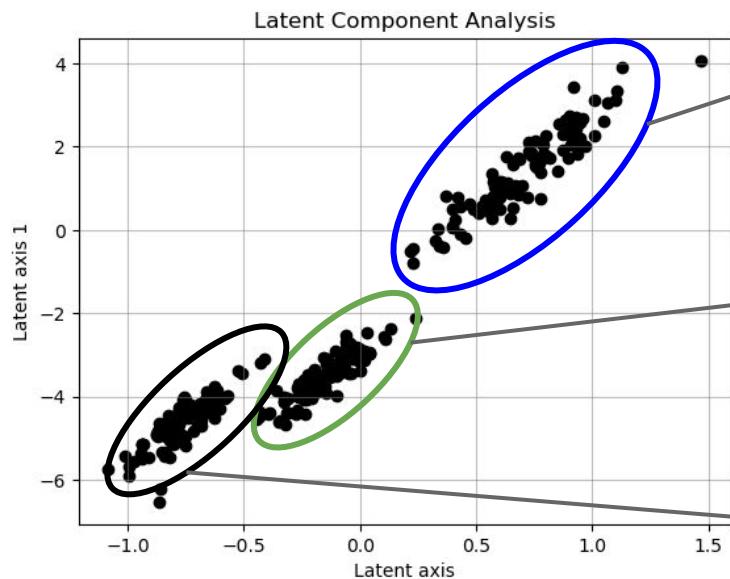
Raw VAE latent space still has too many dimensions



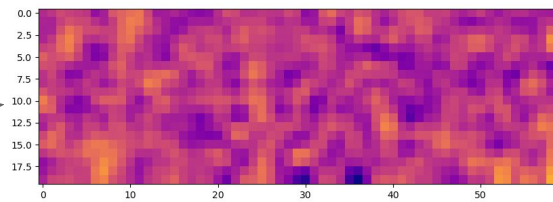
UMAP dimensionality reduction to reveal data groups



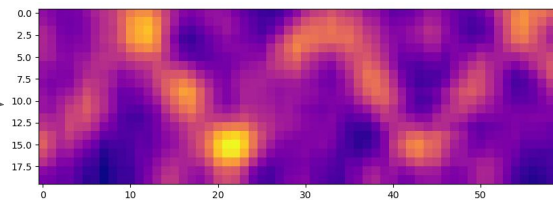
Sample points from each group using kymograph IDs



Polar

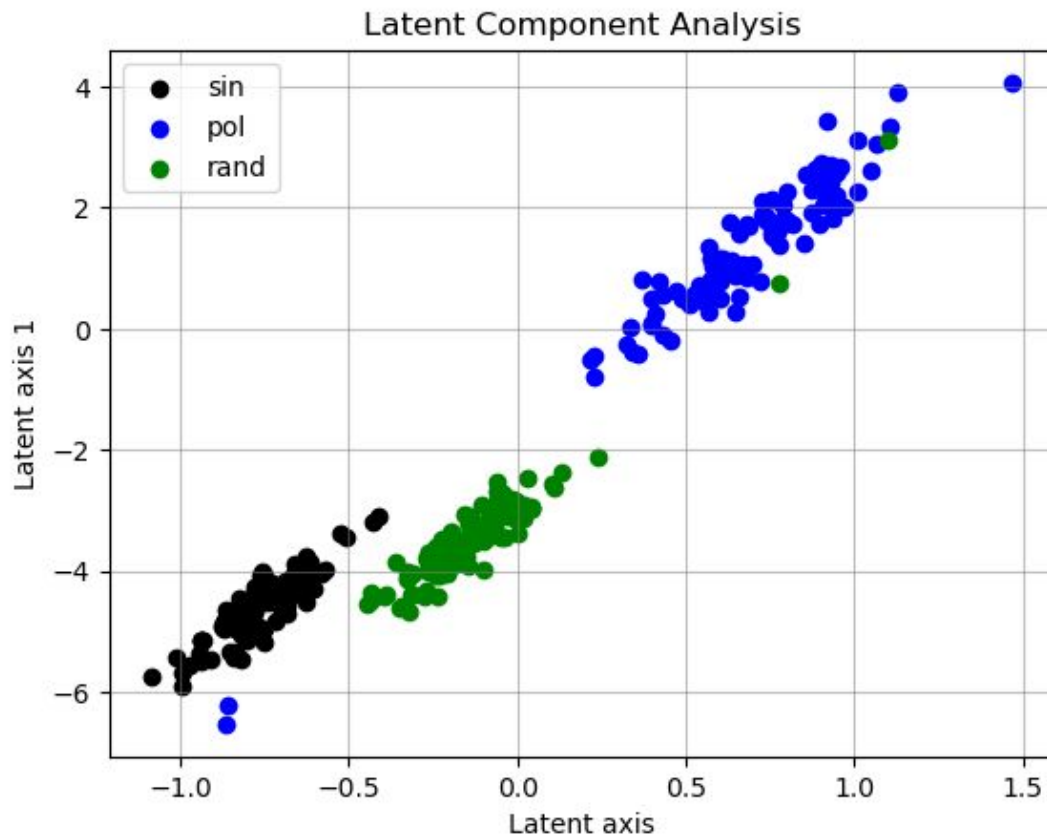


Random

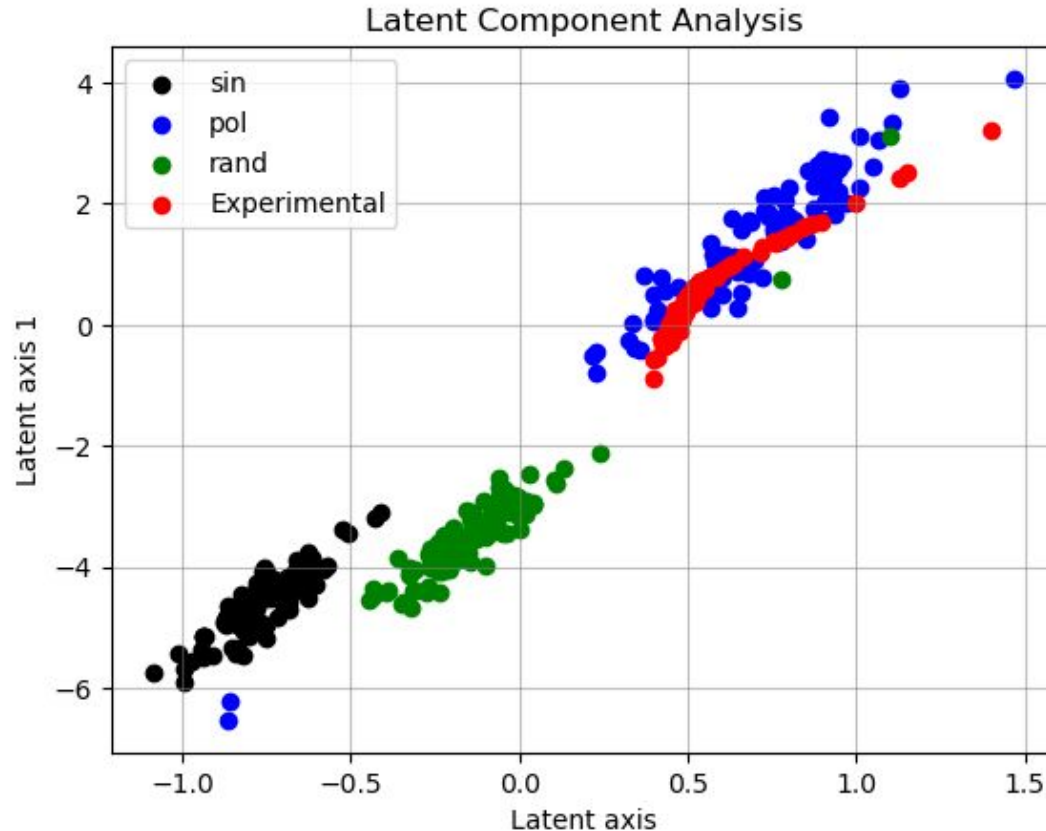


Oscillatory

Label groups accordingly



Encode experimental data and look for similarities



Experimental data occupies a similar region of latent space as the **Polar Kymographs**

Wild Type PlzC is likely Polar