

Data-Driven Approaches for the Classification of Crystallisation Outcomes from In-line Images





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DataFactory Autonomous **Crystalline platform** robotic system (images, temp, shape, Measure **Transfer** growth and agglom) **API Solvent library** [2° Analysis Dispense Dispose] e.g. Raman/XRPD **Zinsser Chrissy automated solvent** and solid dispensing Plan Model Crystallisation **Parameter Database**

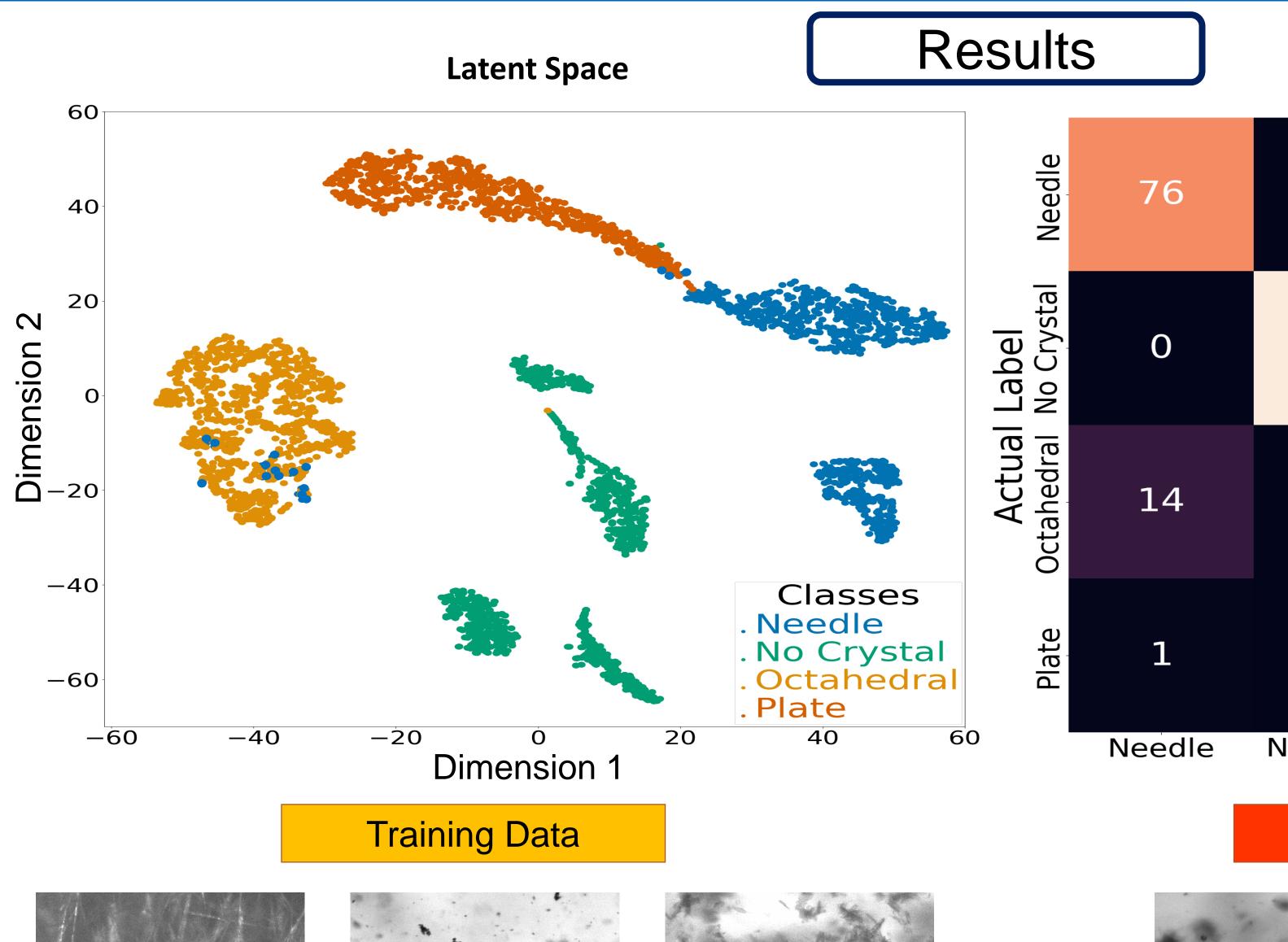
Motivation

- Implementing a data-driven approach that can automate the classification of different phases of high-throughput crystallisation processes can be a powerful tool for [1][2]:
 - ☐ Predict the quality and consistency of the final pharmaceutical product
 - ☐ Optimise manufacturing processes
 - ☐ Reduce waste
- Advanced analytical techniques and machine learning algorithms can be leveraged to extract relevant features on crystallisation outcomes from in-line images.

Method

- Model comprised of pre-trained ResNet-18 feature extractor and one-layer linear classifier.
- 3164 images were used as training data (791 images per class), 788 images as validation data (197 images per class) and 400 images were used as test data (100 images per class).
- Augmentations (edge crop, image rotation and padding) applied to the training data to increase data diversity allowing improved model generalisation.

Input Images Deep Learning Model Output Resnet18 Feature Classifier Plate Needle No Crystal Octahedral



Confusion Matrix 76 0 1 23 0 100 0 0 100 0 0 14 0 86 0 Needle No Crystal Octahedral Plate Predicted Toot Date

True: Octahedral Predicted: Needle Predicted: Plate

Conclusion

- The model achieves an average precision and recall greater than 85%.
- In the latent space graph, we can see features from the needle class overlap with the plate and octahedral cluster space.
- A more varied dataset and more augmentations (brightness and contrast) are necessary to further improve the classification of crystallisation outcomes.

References

[1] Borsos, A., Szilágyi, B., Agachi, P. S. & Nagy, Z. K. Real-Time Image Processing Based Online Feedback Control System for Cooling Batch Crystallization. *Org Process Res Dev* 21, 511–519 (2017).

[2] Silva, A. F. T. et al. Particle sizing measurements in pharmaceutical applications: Comparison of in-process methods versus off-line methods. Eur J Pharm Biopharm 85, 1006–1018 (2013).



True: Needle



True: Plate

True: Octahedral











