

Classification of Micro Array Protein Crystal Images With Laplacian Pyramidal Filters and Neural Networks

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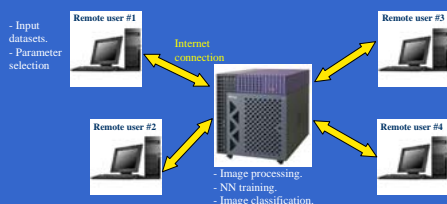
INTRODUCTION

- This project is part of the Northeast Structural Genomics Consortium (NESG). The goal of this consortium is to develop efficient and integrated technologies for high-throughput (HTP) protein production and 3D structure determination.
- This project focuses on the design of an image analysis system to classify protein crystal structures in a production oriented environment.
- The method presented performs classification of microscopic images as protein precipitates or crystals versus clear solutions.
- Preliminary experiments are presented, which provided high classification accuracy on large datasets of protein crystallization experiments using expert classification for ground truth.

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METHODOLOGY

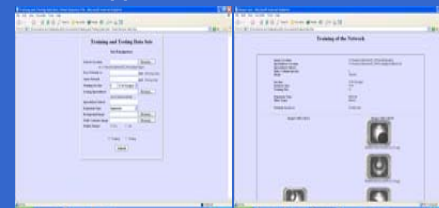
Remote Application Server: Web page interface managed by a Matlab® Server.



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RESULTS

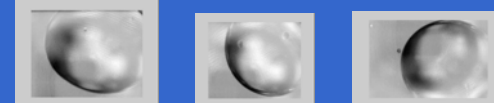
1. Remote Application Web-server Interface



2. Classification with a New Image Database

- **Image Features for Neural Network**
 - Quantitative shape descriptions of the first-order histogram of Laplacian pyramid coefficients.
 - 8 statistics for each coefficient subset.
- **Training data set**
 - 500 images
 - 1/2 clear drops, 1/2 drops with precipitates/crystals
- **Testing experiments**
 - 100 images with precipitates/crystals: FP = 7%.
 - 100 images with clear drops: FN = 3%.

3. Examples of Misclassified Images



Motion blurring, complex boundaries, light reflection, etc.

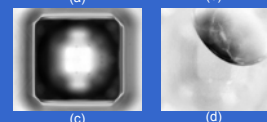
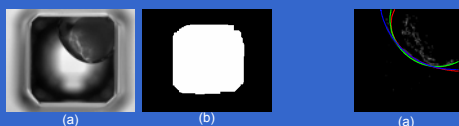


Small bubble size, no sharp gradient, background influence, etc.

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METHODOLOGY

1. Preprocessing of Microscopic Images



(a) Original image with an oil drop containing a precipitate. (b) Mask computed from the contours of the wells. (c) Background image acquired without drops. (d) Pre-filtered image.

(a) Ellipsoidal Hough transform [1] to detect the three most probable ellipses, plotted over the edge map of the pre-filtered image. (b) Pre-filtered image cropped with the minimal rectangular area encompassing the three ellipses.

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DATABASES

1. Microscopic images were acquired with a CCD camera under robotic control.
2. Gray scale 8-bits images saved in tiff/jpeg format.
3. Arrays of 256 × 256 wells screened at three time instants after experimental setup: 1 day, 1 week and 2 weeks.
4. Database of 13,490 scored images:
 - 509 drops containing precipitates and/or crystals.
 - 3,349 clear drops.
 - 9,632 'others' combining drops with organic matter, or dye.

INITIAL STUDY

- **Training data set**
 - 100 images
 - 1/3 clear drops, 2/3 drops with precipitates/crystals
- **Binary classification**
 - 0 = clear drop, 1 = not a clear drop.
- **Testing experiments**
 - Classification errors: false positive (FP) and false negative (FN).
 - 100 images with precipitates/crystals: FP = 24%.
 - 100 images with clear drops: FN = 28%.
 - For our application, false positive errors are more critical than false negative since it would lead the operator to discard images with precipitate/crystals.

⇒ The goal of the optimization of the neural network design was therefore to minimize FP errors.

■ Performance of the classification system:

- Accuracy = $(TP+TN)/(TP+TN+FP+FN)$
- Precision = $TP/(TP+FP)$
- Recall = $TP/(TP+FN)$

[Accuracy Precision Recall] = [74% 0.73 0.76] obtained in this study

[Accuracy Precision Recall] = [88% 0.83 0.76] [3] single plate

[Accuracy Precision Recall] = [85% 0.24 0.66] [3] multiple plates

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REFERENCES

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- [2] P. J. Burt and E. H. Adelson, "The Laplacian pyramid as a compact image code," *IEEE Transactions on Communications*, vol. 31, pp. 532-540, 1983.
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DISCUSSION

- Introduction of robotic manipulation of the crystals for HTP protein production requires the automation of image analysis of crystallization experiments for classification of solution content.
- The proposed feed forward neural network showed promising results in classifying microscopic images.
- Features of representation were computed from Laplacian pyramid expansion histograms. The histogram made the features invariant to orientation which was a desirable feature in order to be able to characterize the diversity and complexity of precipitate appearances.
- The Laplacian expansion provided a representation of the image edge and texture patterns at different scales with extremely fast implementation.

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CONCLUSION

- Future applications of the project include further classification of the microscopic image database to separate crystals from precipitates and drops with organic matter.
- A parallel task of this project includes the creation of a web-based infrastructure for testing and development. An experimental test-bed for crystallization screening is currently being developed.

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