

Applications of machine learning to ARM/ASR science: The Ice Particle and Aggregate Simulator

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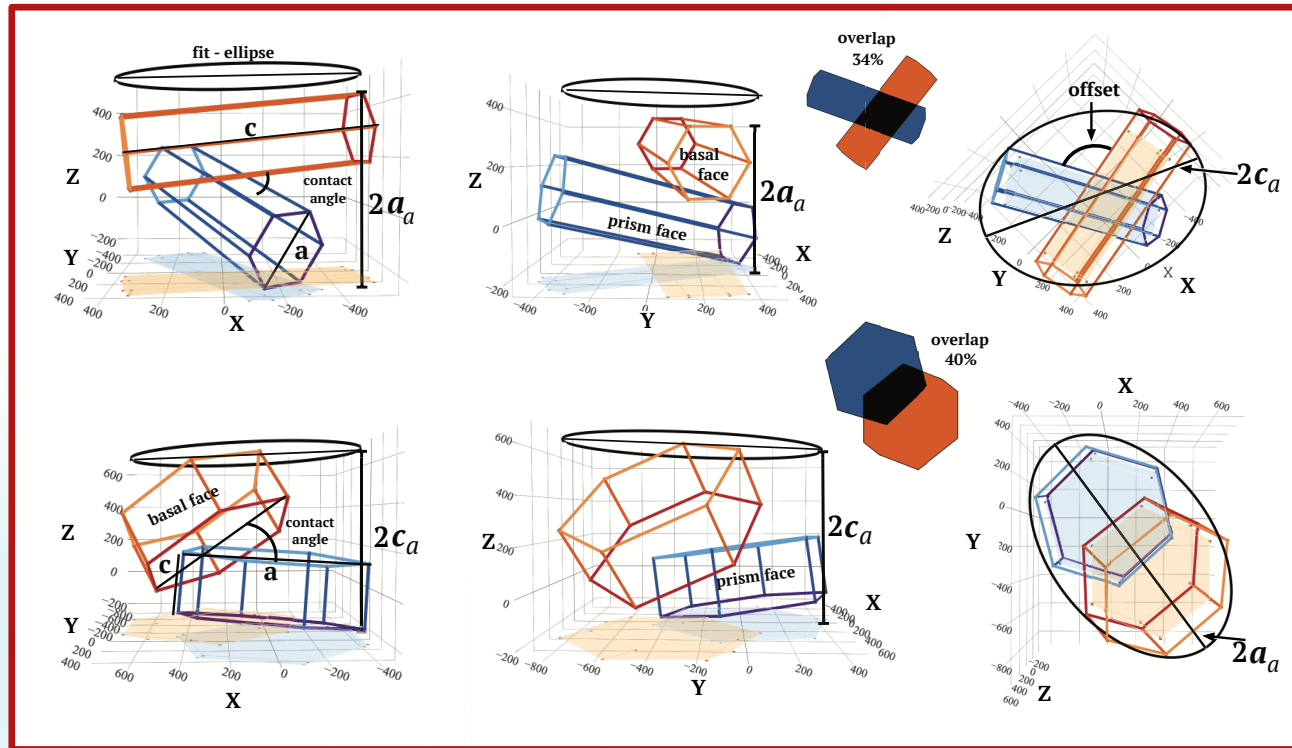
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What is IPAS?

The Ice Particle and Aggregate Simulator

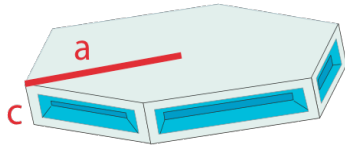
Przybylo et al. 2019



To emulate laboratory data, IPAS acts as a 'theoretical laboratory' and provides likely aggregate properties given an input number of monomers and shape.

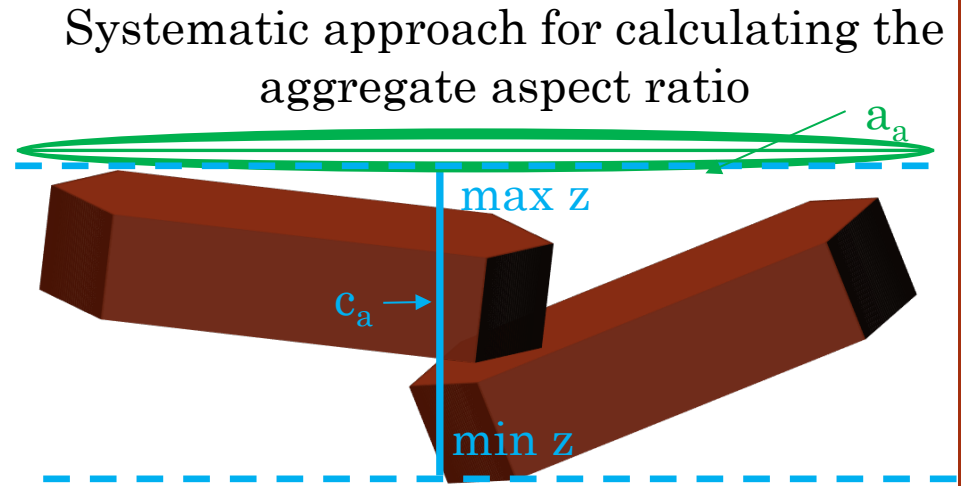
The expected falling orientations, overlap of each monomer, and any contact angle that may form through so-called constrained randomization can be recorded.

$$\phi = c/a = \text{aspect ratio}$$



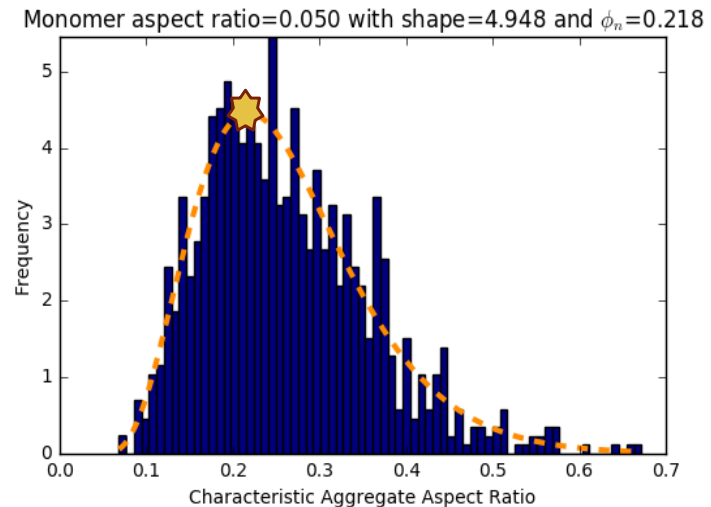
Original monomer aspect ratio

Form aggregate



x 300

Use the peak of the distribution in bulk models



Machine Learning

Objective:

- To compare and analyze the aggregate properties computed in IPAS with properties of observed aggregates

Methodology:

- Apply machine learning (ML) algorithms to classify images from a cloud particle imager (CPI) aircraft probe based on ice particle type
- Once the images are sorted and labeled, another algorithm (or statistical analysis through PDFs) will be used to determine particle dimensional characteristics such as perimeter, area, 2D length and width, etc.
 - Compare these characteristic values to IPAS to further refine simulations for the AHM
- Potentially determine trends or patterns in the particle type/size/shape as a function of height, temperature, relative humidity, etc. (e.g., by plugging in an image of a given particle, the model would be able to “predict” what environment it was likely in).

Concept:



High-Res CPI dataset
after preprocessing

unsupervised learning to classify particle type

~~Successful!~~

Unsuccessful

create training dataset/test
dataset with target concept

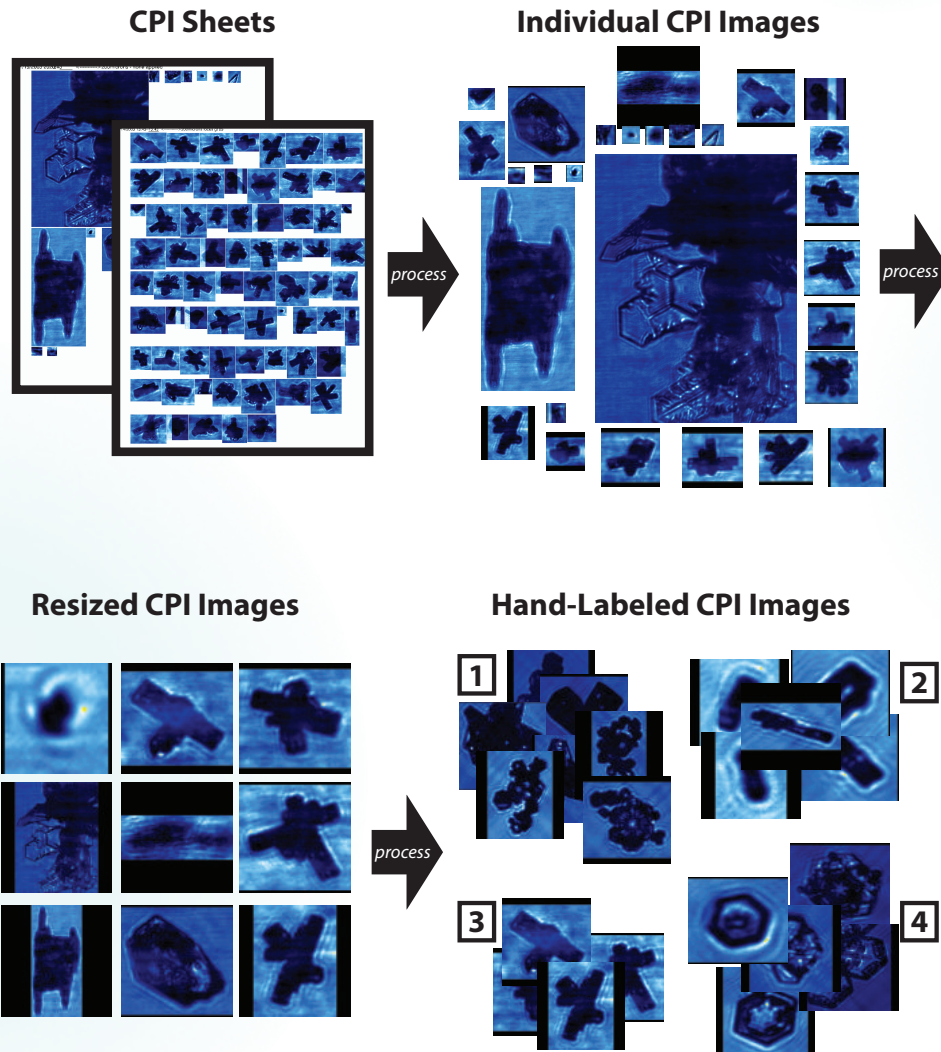
CNN training and object detection

parameter optimization

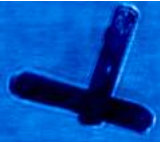
CNN model prediction

new image dataset with labels
final detection results

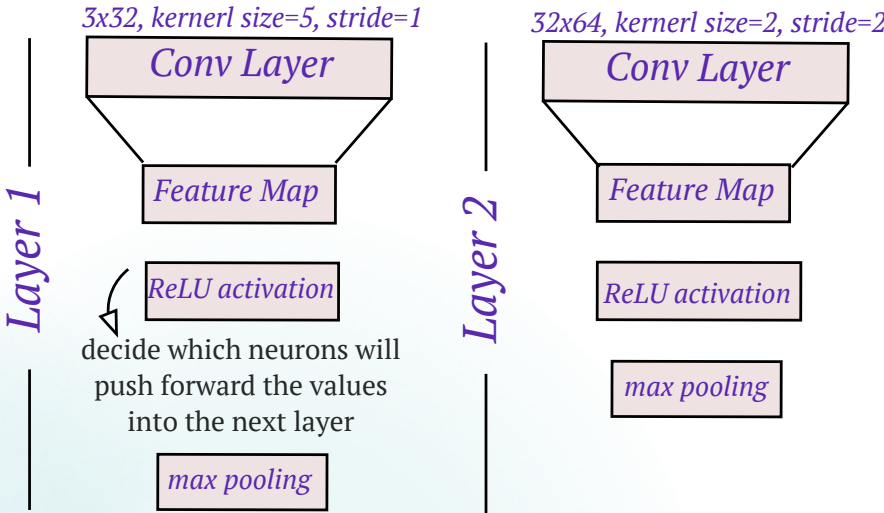
Pre-processing



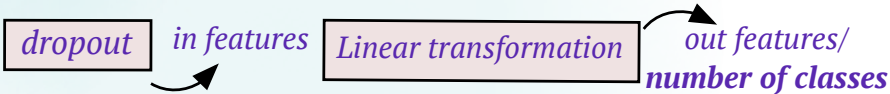
Input Layer



Normalized pixel dimensions ranging from 0.0-1.0



to avoid neuron co-dependency (overfitting)



training a classification problem with multiple classes

- Cross Entropy Loss** a mathematical way of measuring how wrong your predictions are
- lr = .0002**
- Adams optimizer** typically requires little tuning and is computationally efficient

ML hyperoptimization packages:

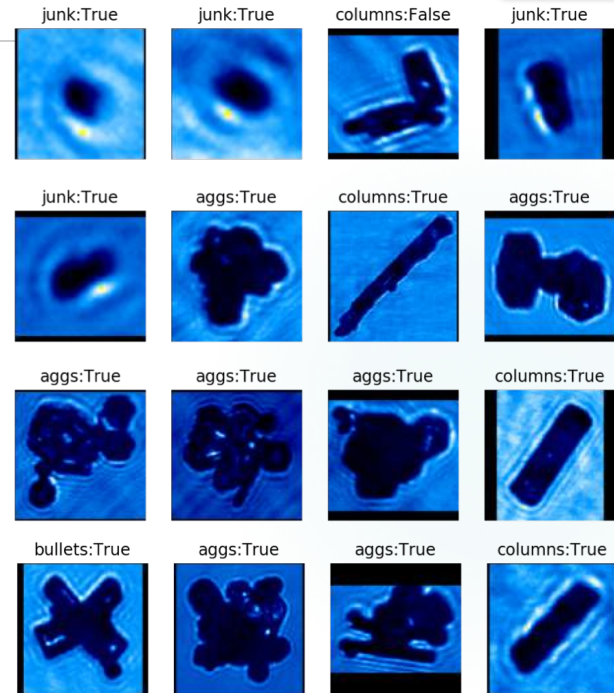
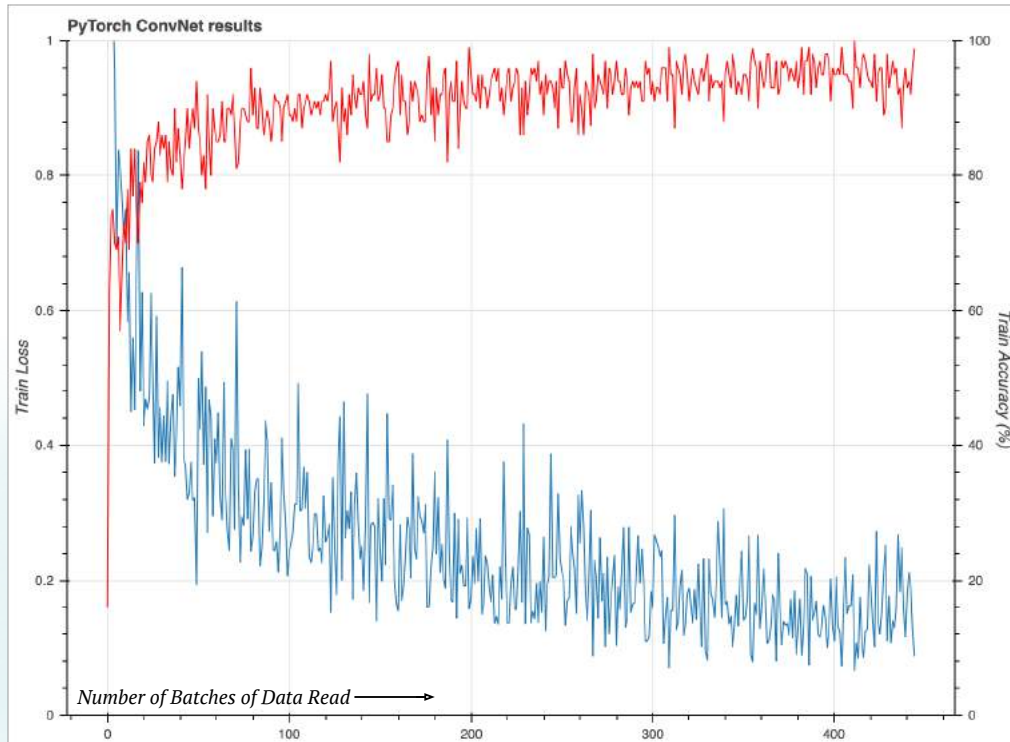
Pytorch

- Randomized Search CV
- **Comet optimizer**

Keras

- Tensorflow backend
- Hyperas
- Comet Optimizer

Final Model:



Accuracy of aggs : 94.444444 %
Accuracy of bullets : 54.545455 %
Accuracy of columns : 85.365854 %
Accuracy of junk : 96.000000 %

Conclusions

- Automatized classification through use of a CNN is generating realistic hydrometeor categories, which is useful to other scientists by reducing image pre-processing time
- A generous, automated, and classified dataset is useful toward better prediction of aggregate properties and segregation of the dynamical processes taking place based on environmental conditions (e.g., aggregation, riming, etc.)
- The results of the proposed work will guide future improvements of both IPAS and bulk models while identifying regions of inadequate predictability
- Next Steps:
 - Build the database of hand-labeled images
 - Apply nested classification
 - Potential for primary and secondary habits
 - Extend results to more probes
 - Perform simulations in IPAS given observed particle properties to compare and validate theoretical aggregates
 - Statistically quantify any similarities and differences

Thank You!
Questions?