

Scaling Distributed Training of Flood-Filling Networks on HPC Infrastructure for Brain Mapping

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Contents

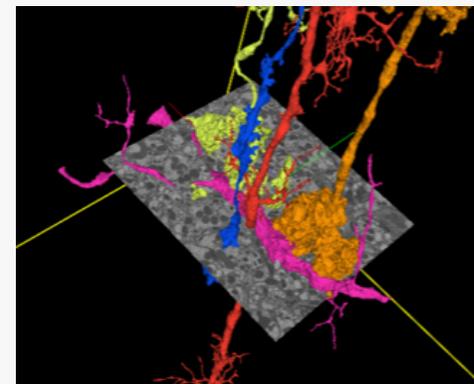
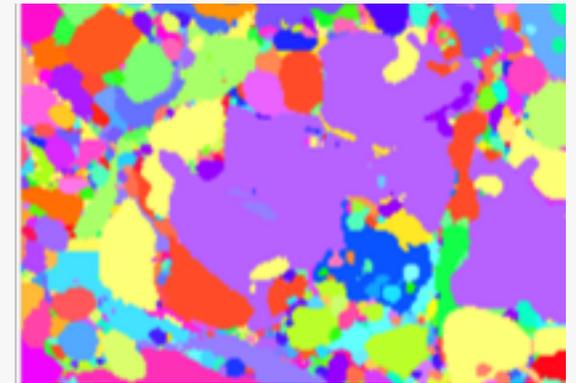
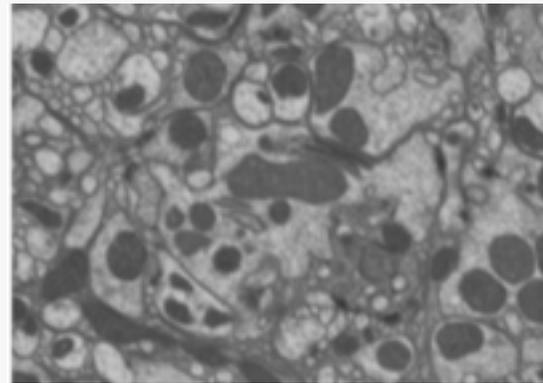
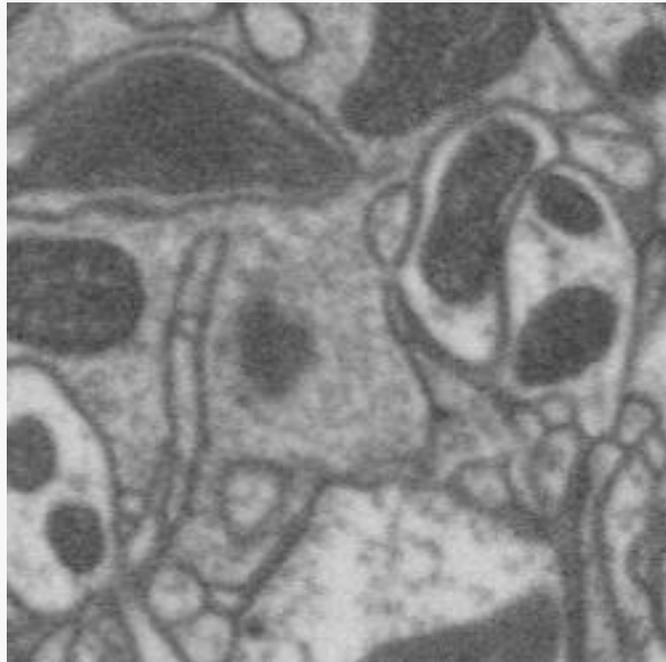
- **Background**
- **Methods**
- **Results**
- **Summary**

Contents

- **Background**
- Methods
- Results
- Summary

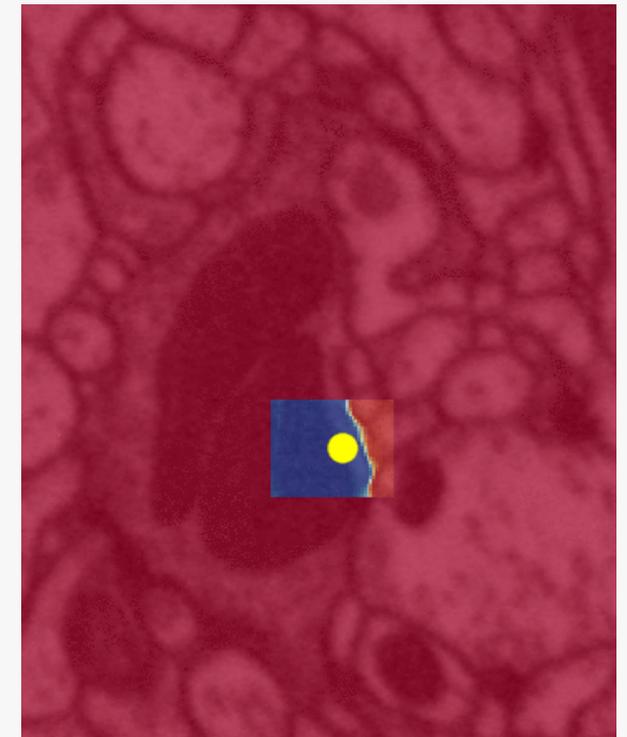
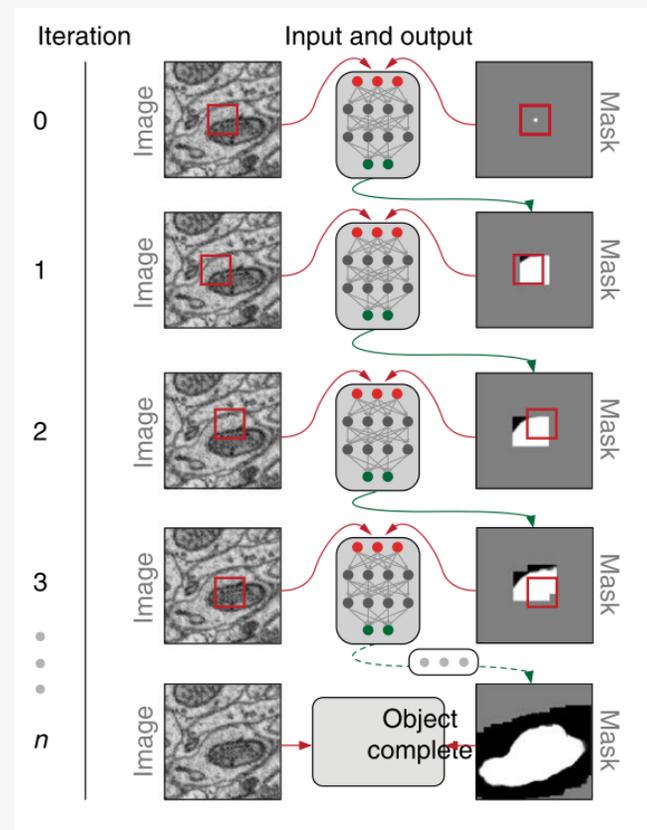
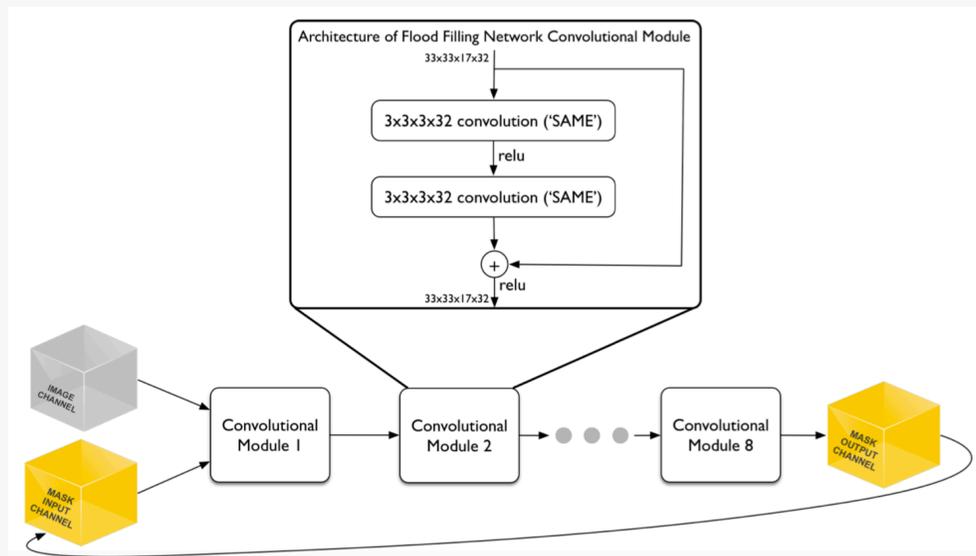
Background

- Brain mapping



Background

- Brain mapping
- Flood-Filling Networks (FFN)

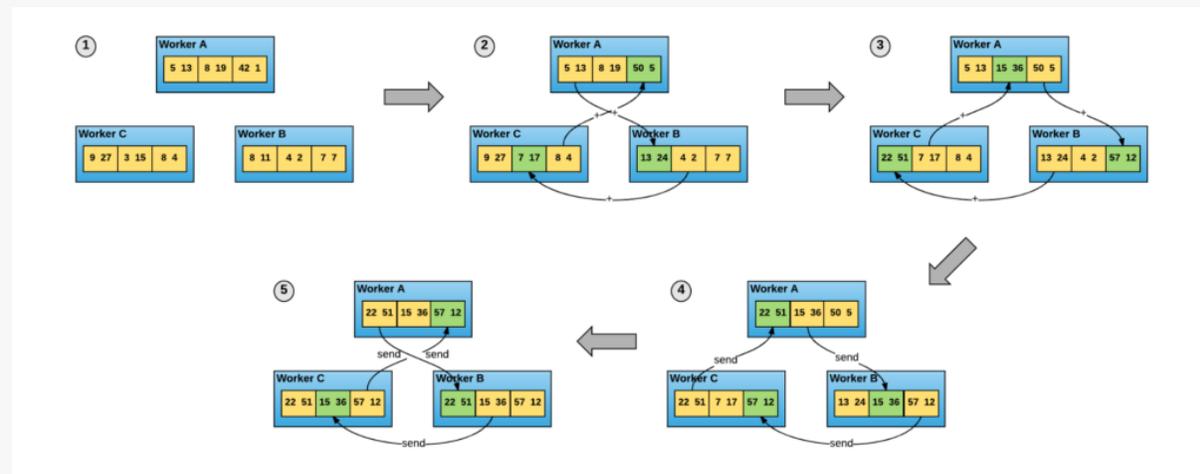
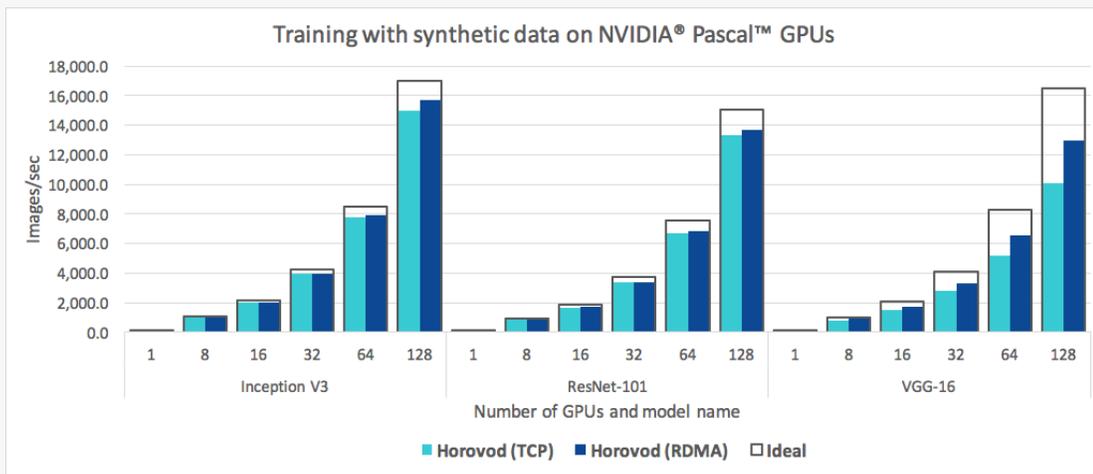


Contents

- Background
- **Methods**
- Results
- Summary

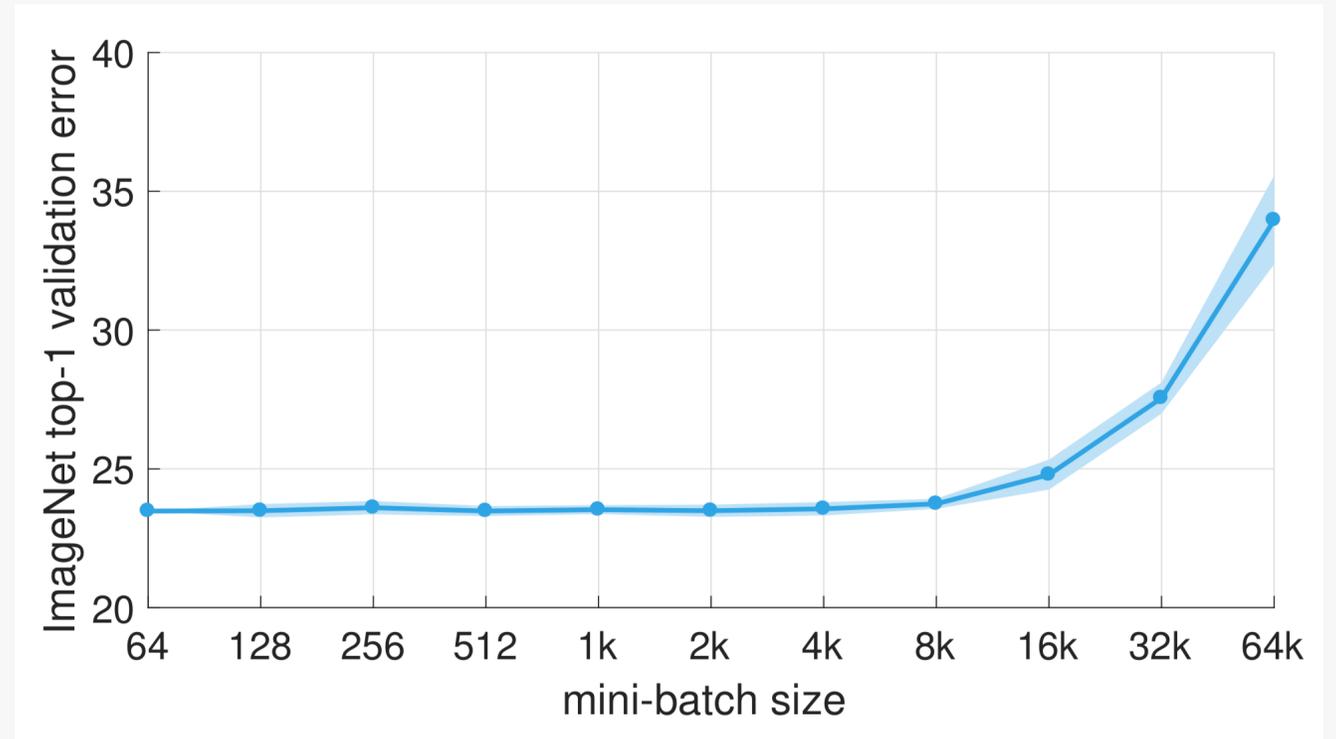
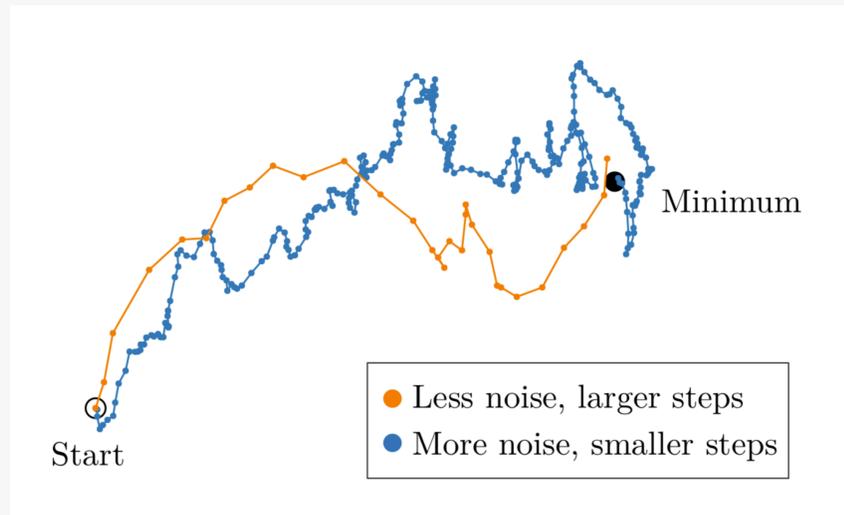
Methods

- Synchronous VS Asynchronous training



Methods

- Synchronous VS Asynchronous training
- Large-batch training

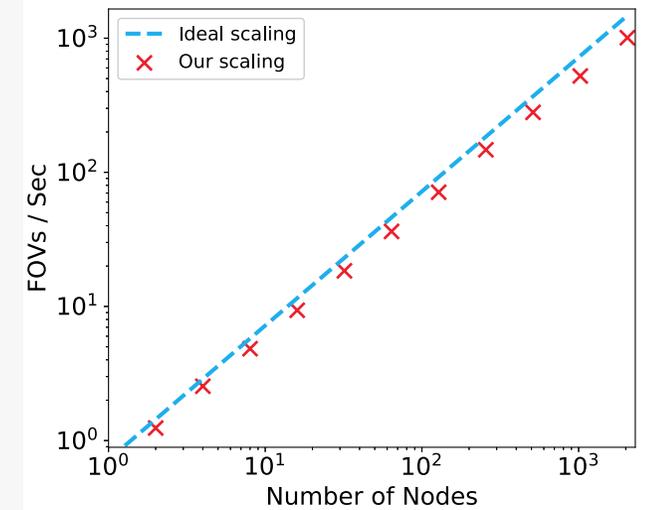
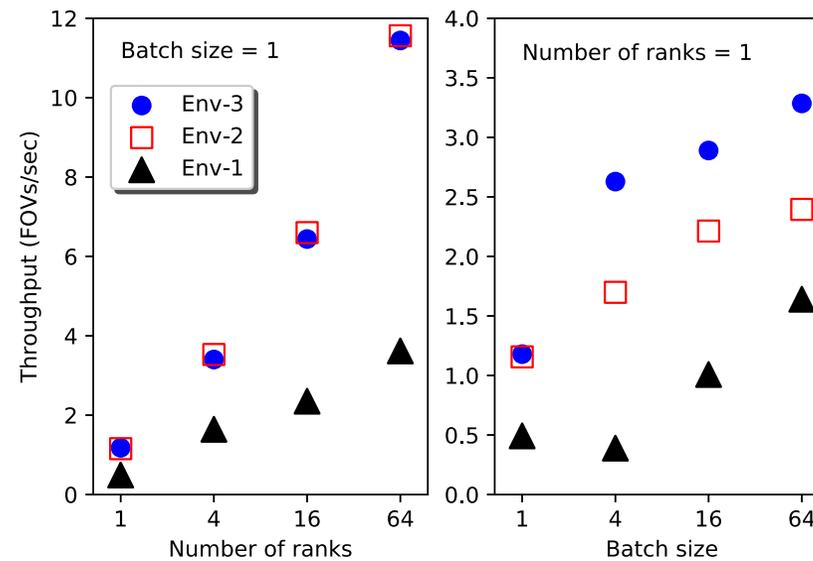
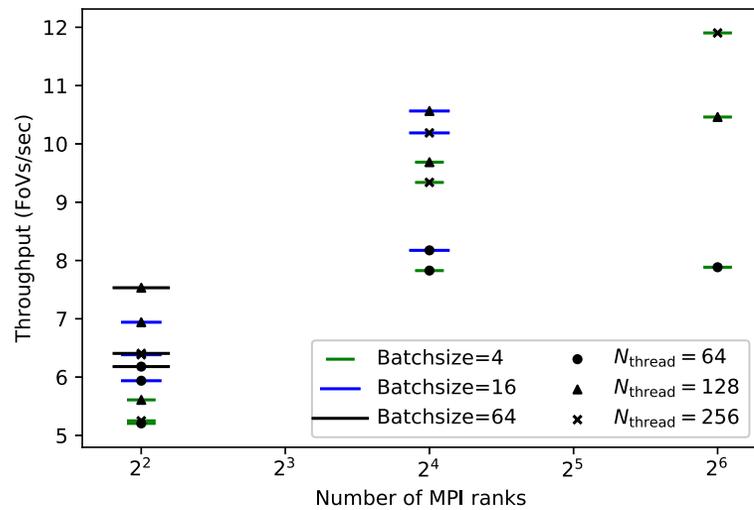


Contents

- Background
- Methods
- **Results**
- Summary

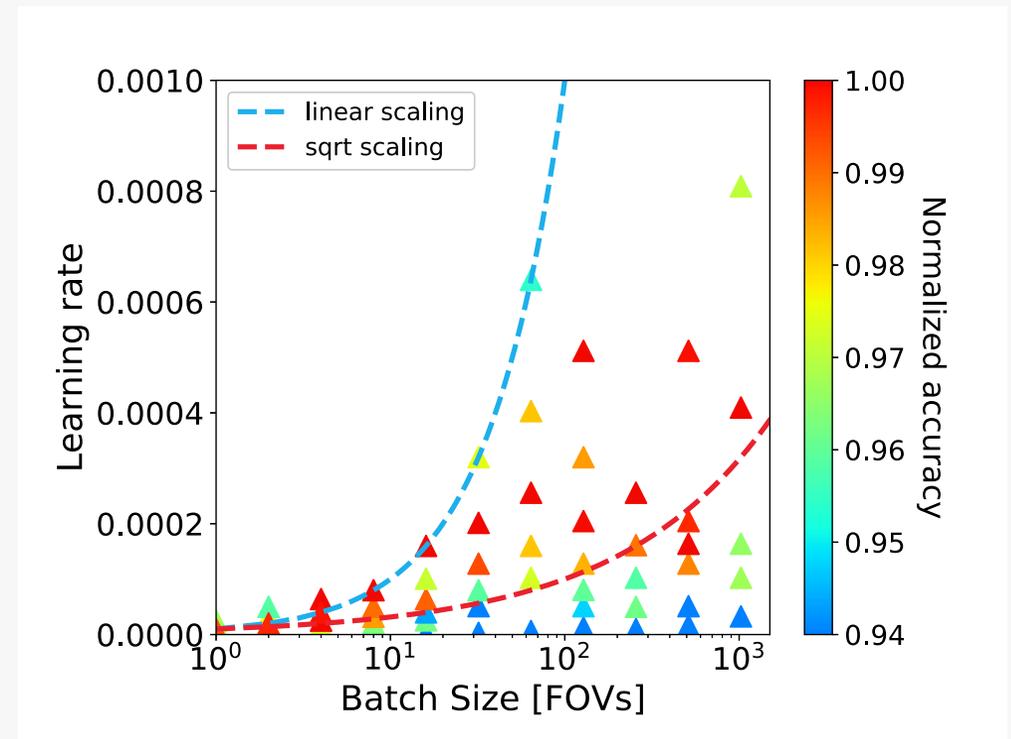
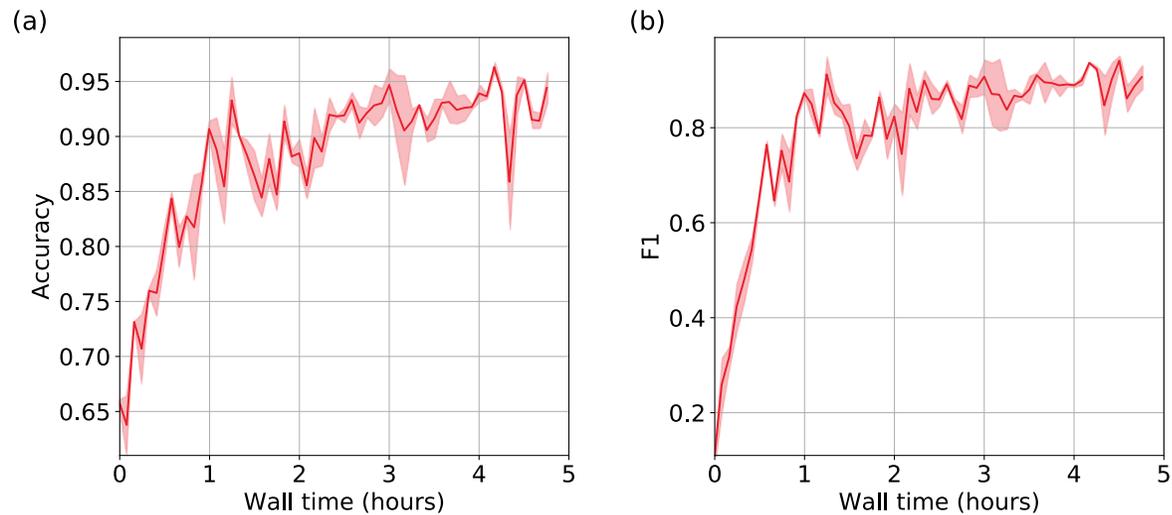
Results

- Throughput optimization



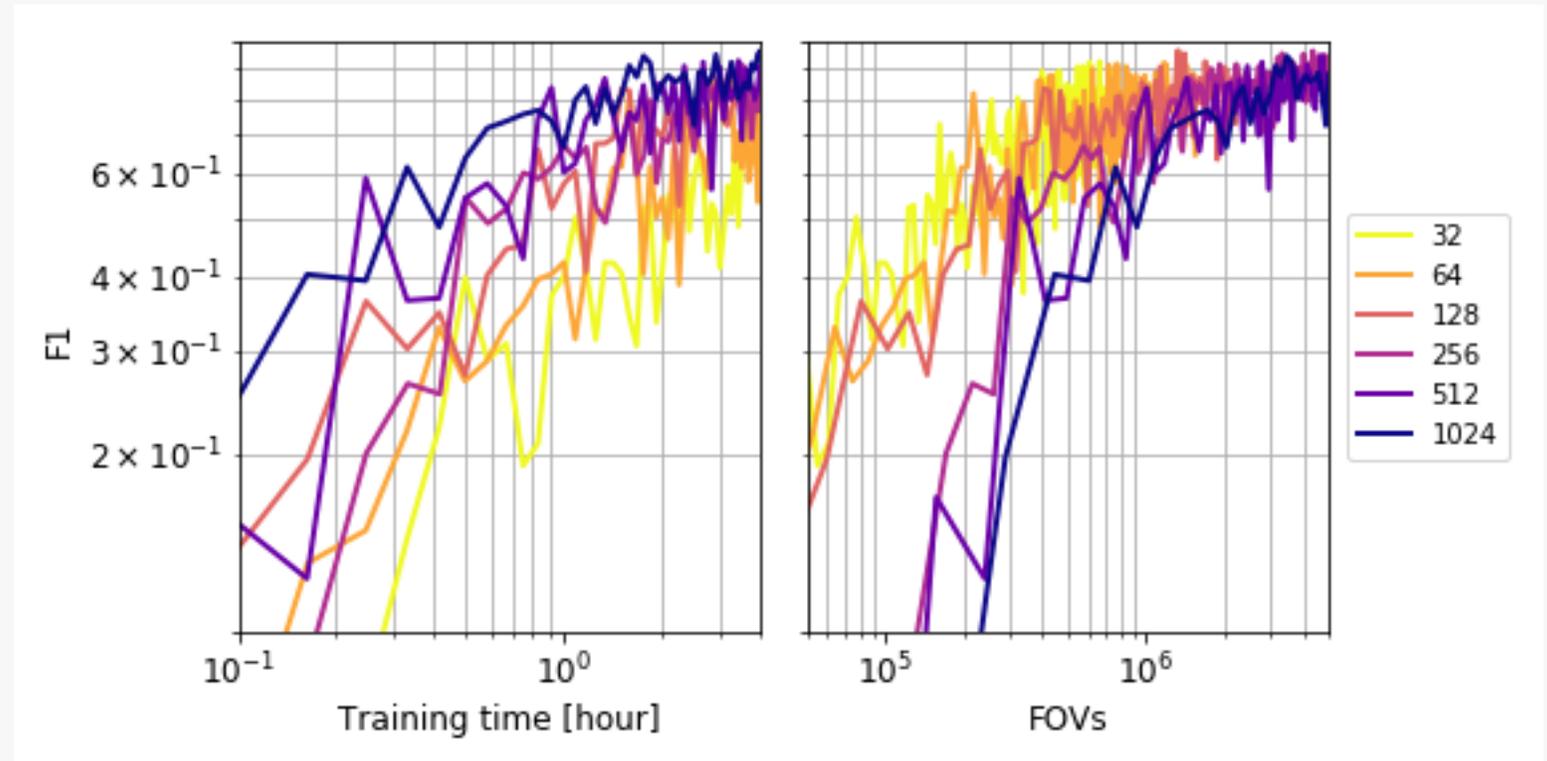
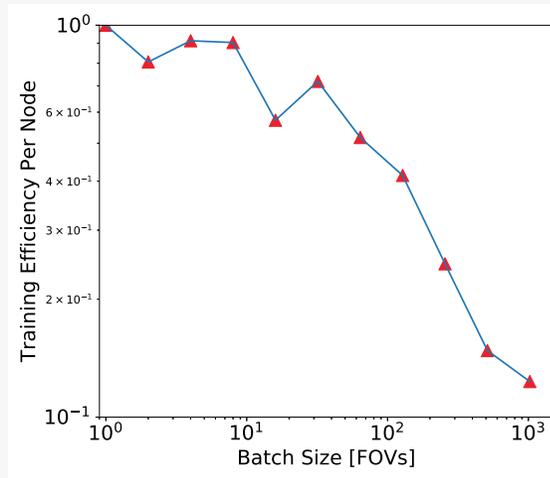
Results

- Throughput optimization
- Optimal learning rate scaling



Results

- Throughput optimization
- Optimal learning rate scaling
- Effect of batch sizes



Results

- Throughput optimization
- Optimal learning rate scaling
- Effect of batch sizes
- Training evaluation

True Positive (TP)

False Positive (FP)

True Negative (TN)

False Negative (FN)

Adapted Rand Error (ARE)

$$\text{accuracy} = \frac{TP + TN}{TP + TN + FP + FN}$$

$$\text{precision} = \frac{TP}{TP + FP}$$

$$\text{recall} = \frac{TP}{TP + FN}$$

$$\text{F1} = \frac{2 \times \text{recall}}{\text{precision} + \text{recall}}$$

$$\text{ARE} = 1 - \text{F1}$$

Variation of Information (VOI)

$$\text{VOI}_{\text{split}} = H(X|Y)$$

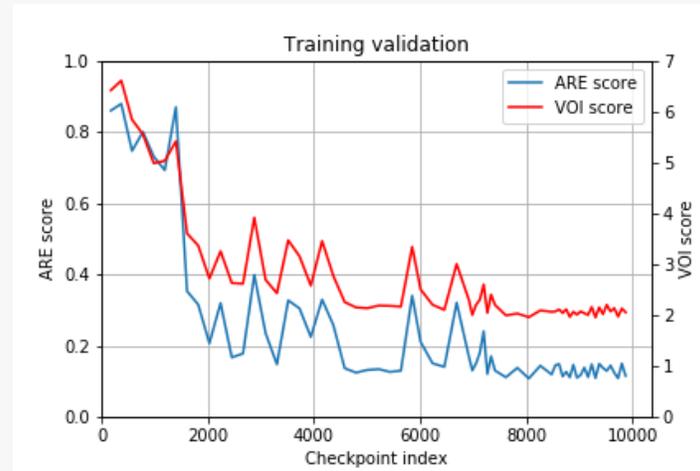
$$\text{VOI}_{\text{merge}} = H(Y|X)$$

$$\text{VOI} = \text{VOI}_{\text{split}} + \text{VOI}_{\text{merge}}$$

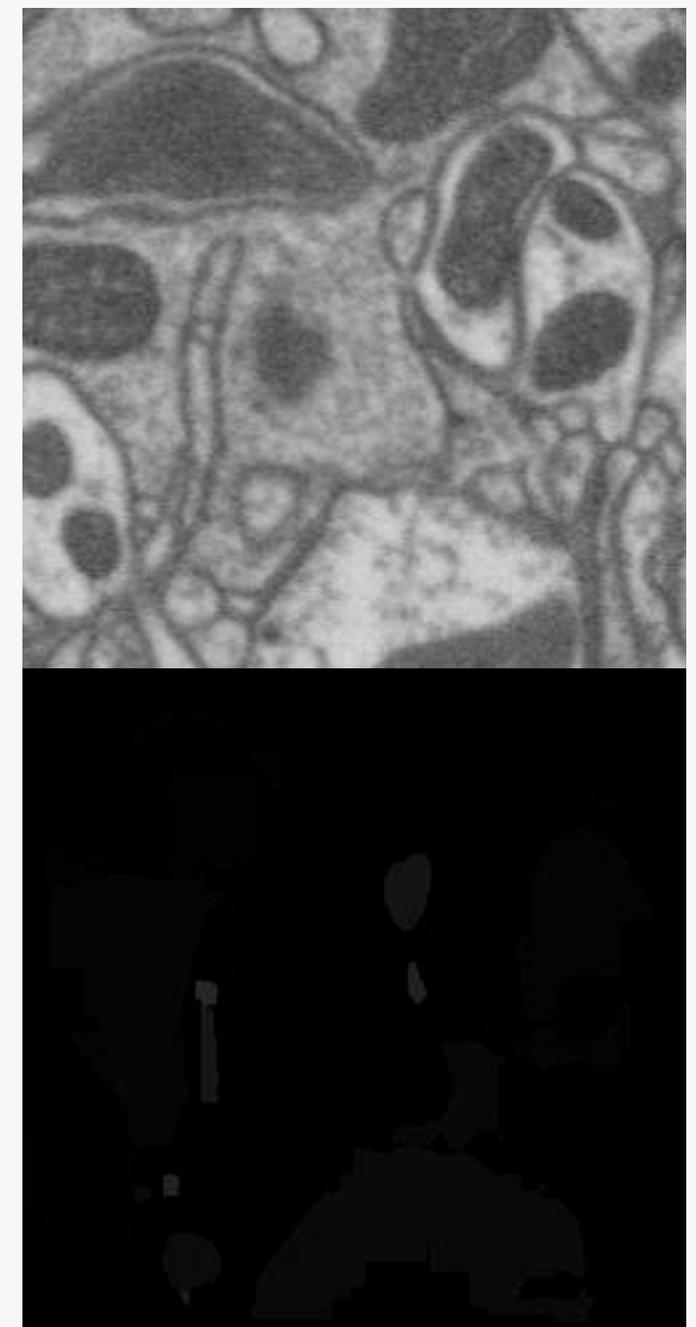
H(X) denotes the entropy of X

Results

- Throughput optimization
- Optimal learning rate scaling
- Effect of batch sizes
- Training evaluation



Segmentation	ARE	VOI
MALA	N/A	1.1470
CELIS-MC	N/A	1.1208
Original FFN paper	0.0973	3.2085
Ours	0.1074	1.9518



Contents

- Background
- Methods
- Results
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Summary

- Implemented data-parallel synchronous training of FFN and scaled it up to 2048 KNL nodes on Theta
- Reduced FFN training time needed to reach similar levels of evaluation quality
- Showed the tradeoff between compute-efficiency and time-efficiency
- First step towards a complete computational pipeline to produce large-scale brain maps
- Take-home message: Efficient training on HPC requires efficient usage of large training batches.

Acknowledgement

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Thank You!

For more information: [arXiv 1905.06236](https://arxiv.org/abs/1905.06236)