**ROBUST SCIENCE WORKSHOP FEBRUARY 17, 2021** 

### **TOWARDS AN HPC AUTOMATED PIPELINE** FOR CONNECTOMICS

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## SERIAL ELECTRON MICROSCOPY ACQUISITION AND ANALYSIS STEPS FOR CONNECTOMICS



### Scalable, trustworthy, and reproducible Our experience: EM PIPELINE @ ARGONNE

- **Optimizing codes** for montage, alignment, and segmentation to run on Argonne supercomputers with natural concurrency (tens to thousands of compute nodes)
- Assembling codes into a pipeline
- Respect (and exploit) concurrency of individual applications
- Reduce human-intensive aspects of large-scale computing
- Schedule jobs in a more optimal/efficient fashion
- Support user interaction with data throughout pipeline (orange boxes)

Argonne National Laboratory is a U.S. Department of Energy laboratory managed by UChicago Argonne, LLC. Toward an Automated HPC Pipeline for Processing Large Scale Electron Microscopy Data, Vescovi, Li, et al (XLOOP workshop, Supercomputing, Nov. 2020)





## LARGE-SCALE FLOOD FILLING NETWORKS (FFN)

### Scaled training of Google's FFN to thousands of nodes on ALCF's Theta Large-scale training achieves greater accuracy faster Theta nodes $6 \times 10^{-1}$ Faster training opens opportunity to make many runs to reach optimal model 64 cores/node 32 $4 \times 10^{-1}$ (hyperparameter optimization) 64 128 ⊑ 3×10<sup>-1</sup> 256 **FFN Segmentation** rate on Theta: ~350M voxels/nodehour 512 1024 $2 \times 10^{-1}$ Distributed inference on many nodes allows us to reconstruct larger volumes faster (weak scaling) $10^{-1}$ 100 Training time [hour] 0.0010 1.00 — — linear scaling – – sqrt scaling 0.90 0.0008 9000.0 rate 0.98 -earning 0.97 0.0004 0.96 0.0002 0.95 0.0000 <del>|-</del> 10º 0.94 103 101 $10^{2}$ Batch Size [FOVs]



Wushi Dong, Murat Keceli, Rafael Vescovi, Hanyu Li, Corey Adams, Tom Uram, Venkatram Vishwanath, Bobby Kasthuri, Nicola Ferrier, Peter Littlewood, "Scaling Distributed Training of Flood-Filling Networks on HPC Infrastructure for Brain Mapping", 2019 IEEE/ACM Third Workshop on Deep Learning on Supercomputers (DLS) at SC19



## **WORKSHOP QUESTION**

Can publishing/dissemination/sharing standard practices and procedures contribute to achieving or encouraging scalability, trust, and reproducibility in the applications results?

- Neuroscience computational community shares procedures, practices, data
- Scalability being addressed by some
- Trust from human evaluation bottleneck!

Main challenges:

- Reducing human effort (proof reading)
  - Develop better metrics for intermediate steps





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