# Deep Neural Networks for LArTPC Particle Image Detector at the HEP Intensity Frontier

Kazuhiro Terao SLAC National Accelerator Lab. 1st ML-at-SLAC Workshop (Feb. 19th 2019)





### Physics Interest (sorry): Neutrinos

# Small Cross-section (rare interaction)



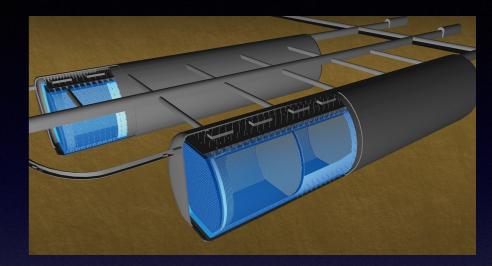
0.4 ton 100 PMTs (1956)



3000 ton 1000 PMTs (1956)



50,000 ton 11,000 PMTs (1996)



o(1M) ton ? PMTs (202X)

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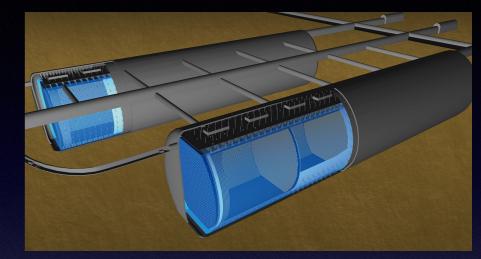
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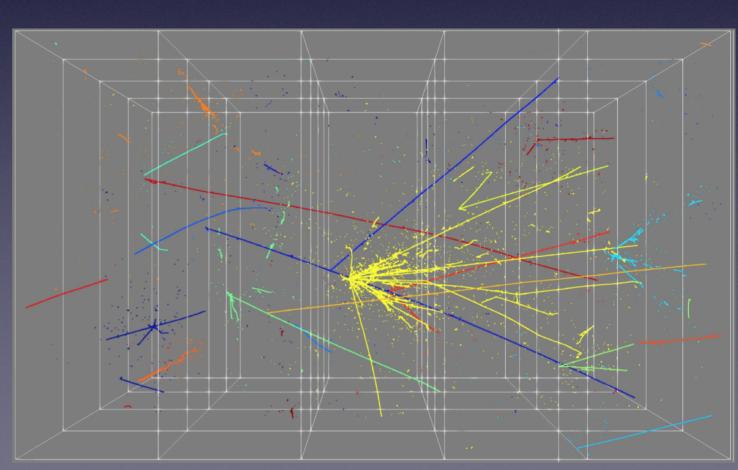


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# Study Secondaries





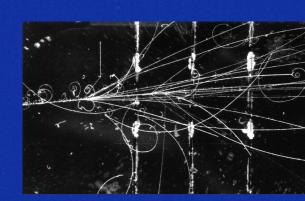




#### ~mm/pixel spatial resolution ~MeV level sensitivity

**MicroBooNE** 

~87 ton (school bus size)



**Bubble Chamber** 

 $\nu_{\mu}$ 

### **Liquid Argon Time Projection Chamber**

• Chamber-like images: digitized electronics readout

• Calorimetric measurement + scalability to a large mass

2015



#### **Topological shape** difference is a major

distinction for <u>"shower" particles</u>





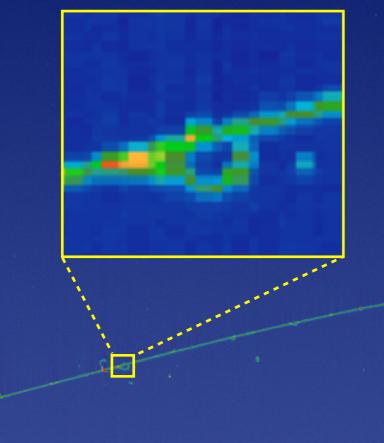
**Trajectory ends** are distinct, and useful for seeding particle clustering and trajectory fitting

75 cm



Many, local kinks caused by Multiple Coulomb Scattering process can be used for momentum estimation





Small branchés on muonlike trajectories are knockedoff electrons, useful key for the direction



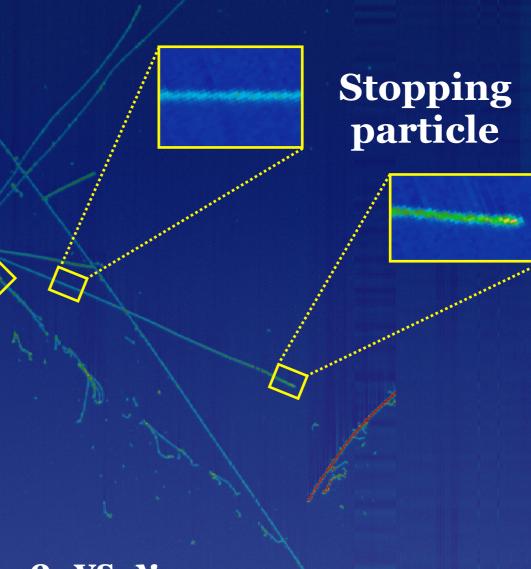
µBooNE

Energy deposition patterns (dE/dX) vary with particle mass & momentum, useful for analysis

Highly ionizing

µBooNE

Energy deposition patterns (dE/dX) vary with particle mass & momentum, useful for analysis



e- vs. γ using dE/dX

Run 3493 Event 41075, October 23<sup>rd</sup>, 2015

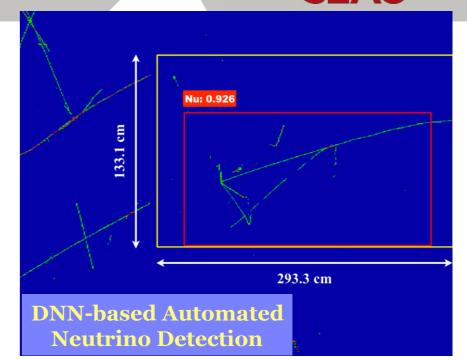
75 cm

Of course... we like **3D** better

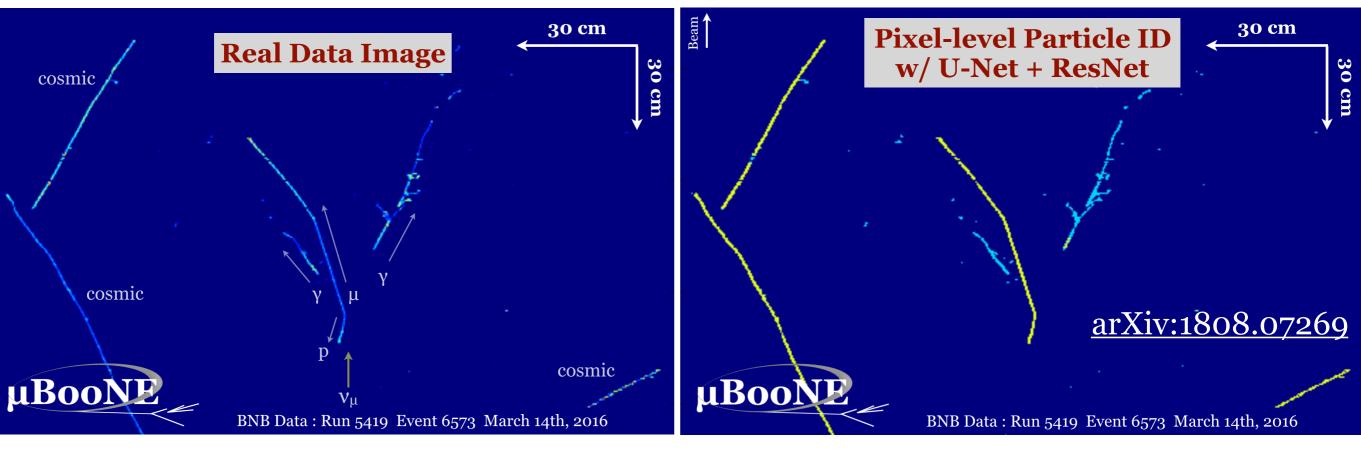
### ML @ HEP-Intensity Frontier Machine Learning for Particle Image Analysis

### • DNN for image data analysis

- Computer Vision: use DNNs to extract interesting features for physics from image data (JINST 12, P03011) down to pixellevel (arXiv:1808.07269)
- **Crucial**: avoid & study discrepancies in simulation (train) and data (test) domains



#### **Detection Network**

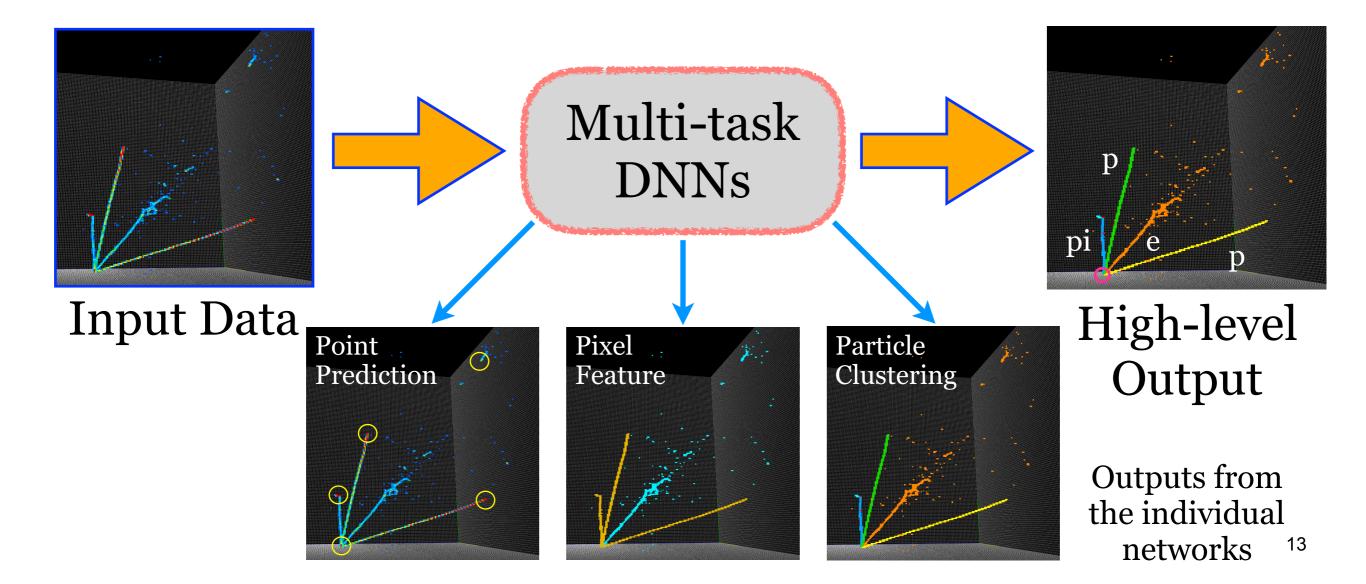


### ML @ HEP-Intensity Frontier Machine Learning for Particle Image Analysis

#### **Multi-task DNN for Physics Reconstruction**

Introduce physical feature extraction tasks (auxiliary targets) to bias the data transformation path to support producing a logical conclusion. Optimize the whole reconstruction chain.

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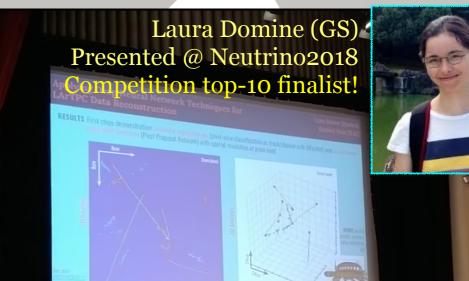


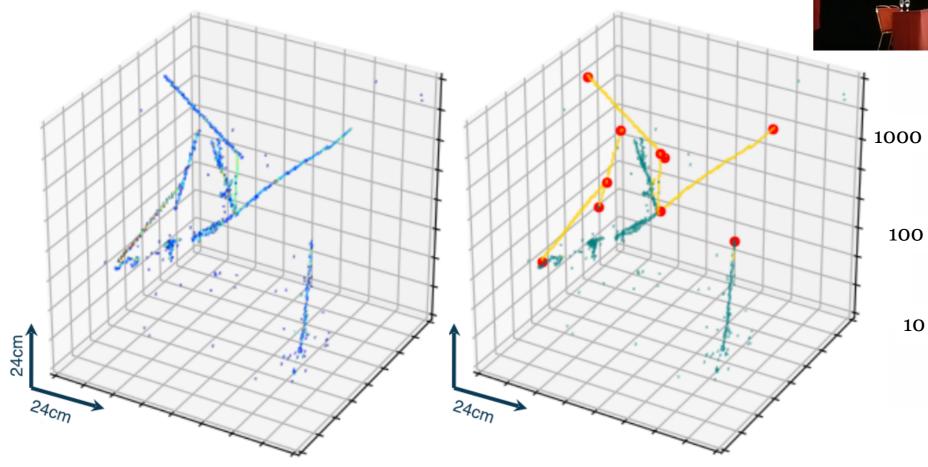
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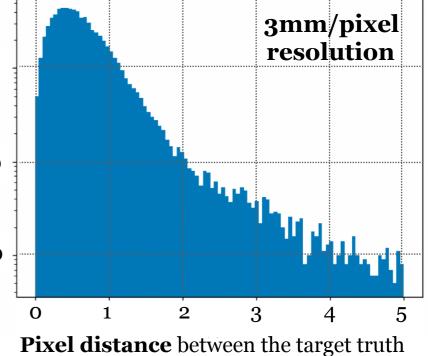
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#### **Multi-Task Network Cascade**

- Chain of Segmentation + Detection
  - Feature points: "shower start" and "track edges"
  - Classify each pixel into "shower" vs. " track"
- Extension to 3D data
  - Change in tensor dimensions, identical algorithms





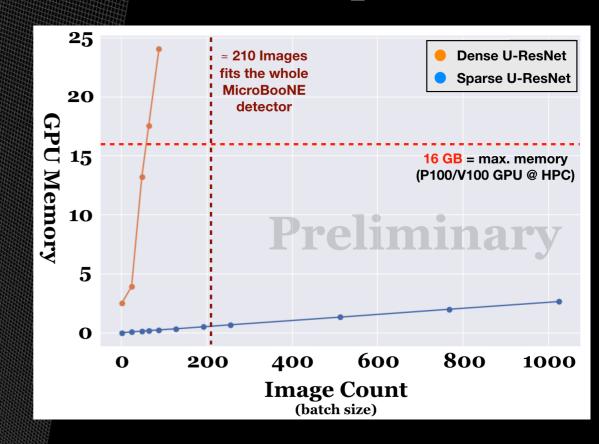


point to the closest proposed point  $14_{14}$ 

#### ML @ HEP-Intensity Frontier Scalability for 3D and larger detectors

Data feature: generally sparse, locally dense image, and very large volume (1e10 to 1e12 pixels)

si ac



Dedicated DNN libraries for sparse data are essential

### ML @ HEP-Intensity Frontier Research collaborations and outreach



### ML connects research frontiers!

- HEP-ML cross-frontier with Michael K. & Phil M.
- Computer Vision DNN for Cryo-EM with Yee T. & Wah C.
- ... and other misc. consultations with random visitors to my office
- Workshop/Tutorials (e.g. ML-for-HyperK @ TRIUMF in April)
- DeepLearnPhysics: cross-experimental organization for R&D

## Sharing of R&D data & software

- <u>Open data sample</u>
- Software distribution w/ containers (<u>docker</u>, <u>singularity</u>)
- ... some R&D papers using those soon hopefully :)

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### ... and I won't have a summary ... Please find me with questions during breaks! 17