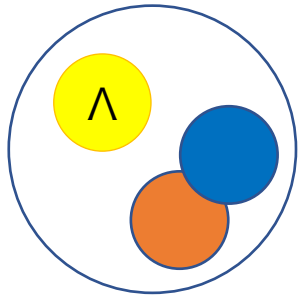




# Precise measurement of the binding energy of hypertriton - from the nuclear emulsion data using analysis with machine learning -

Ayumi Kasagi  
HENP RIKEN, Gifu Univ.

HADRON 2021



Hypertriton:  ${}^3_{\Lambda}\text{H}$   
Benchmark for hypernuclear physics

$\tau \doteq 263 \text{ ps??}$  (Free  $\Lambda$ :  $263 \pm 2 \text{ ps}$ )

Lifetime measurement with Heavy Ion beams

HypHI (GSI):  $183^{+42}_{-32} \text{ ps}$

STAR (RHIC):  $142^{+24}_{-21} \text{ ps}$

ALICE (LHC):  $237^{+33}_{-36} \text{ ps}$

C. Rappold et al., Nucl. Phys. A 913 (2013) 170

STAR Collaboration, Phys. Rev. C 97 (2018) 054909

ALICE Collaboration, Phys. Lett. B 797 (2019) 134905

GSI, J-PARC, ELPH, RHIC, LHC, FAIR and HIAF...

Will be presented in this session

Binding energy measurements

$B_{\Lambda} = 130 \pm 50(\text{stat.}) \pm ???(\text{syst.}) \text{ keV}$   
Nuclear emulsion ( $\sim 1973$ )

nature  
physics

LETTERS

<https://doi.org/10.1038/s41567-020-0799-7>

Check for updates

**Measurement of the mass difference and the binding energy of the hypertriton and antihypertriton**

The STAR Collaboration\*

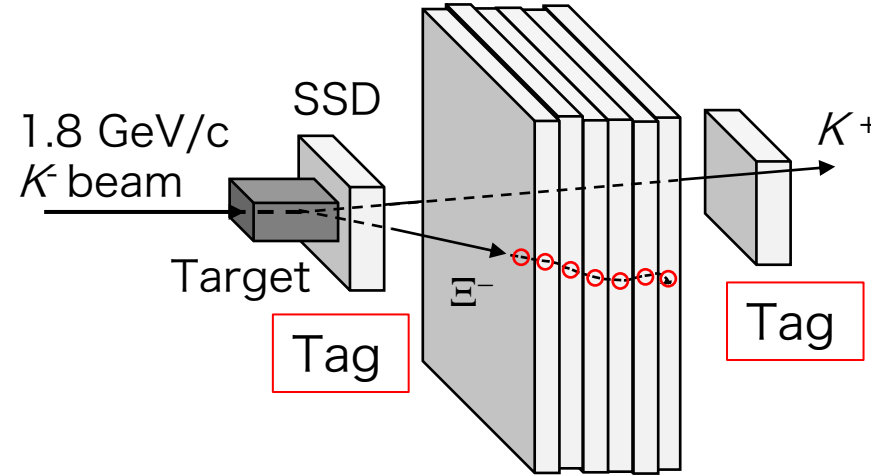
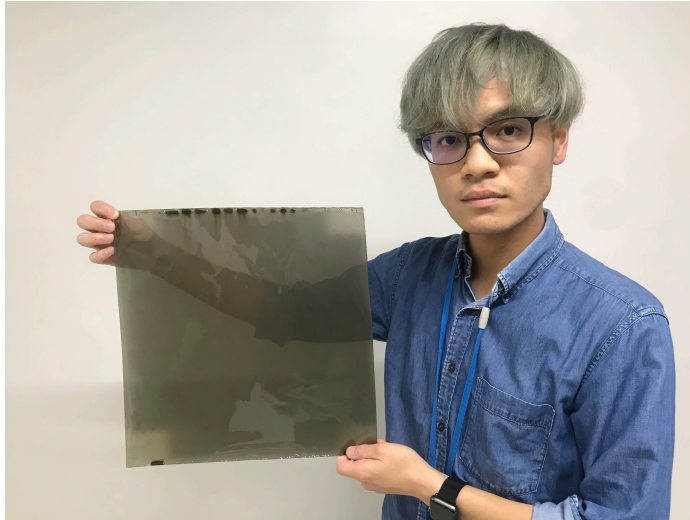
$B_{\Lambda} = 410 \pm 120(\text{stat.}) \pm 110(\text{syst.}) \text{ keV}$

The nature of Hypertriton is not clear...

→ New measurement of  $B_{\Lambda}$  with Nuclear Emulsion + modern technique

# Nuclear Emulsion data from J-PARC E07 experiment 2/10

Emulsion-Counter hybrid method for double-strangeness hypernuclear search

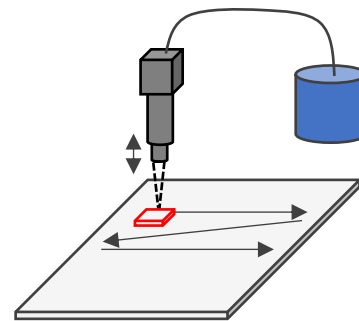
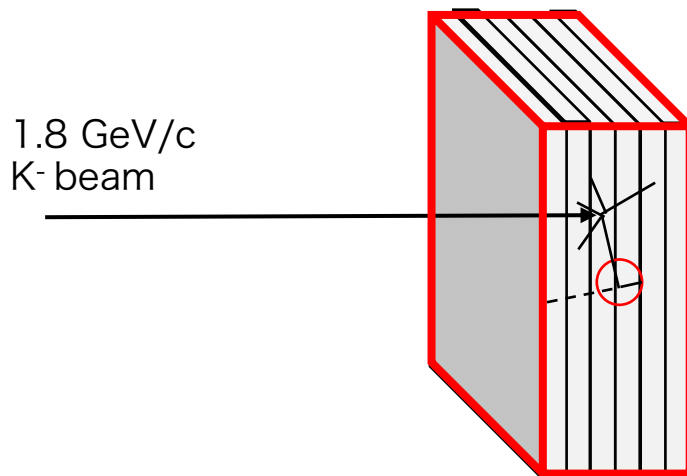


- Reduce the scanning volume and human load

H. Ekawa et al., PTEP, (2019)  
S. H. Hayakawa et al., PRL, (2021)  
M. Yoshimoto et al., PTEP, (2021)

Will be presented in this session

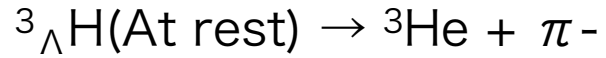
Overall scanning method



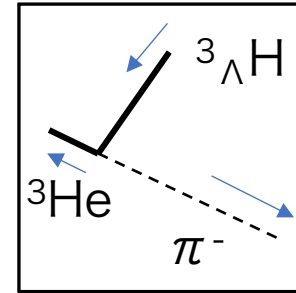
- 1000 double-strangeness hypernuclei
- Millions of Single-hypernuclei
- Requires a huge amount of visual inspections

# Hypertriton search from dataset

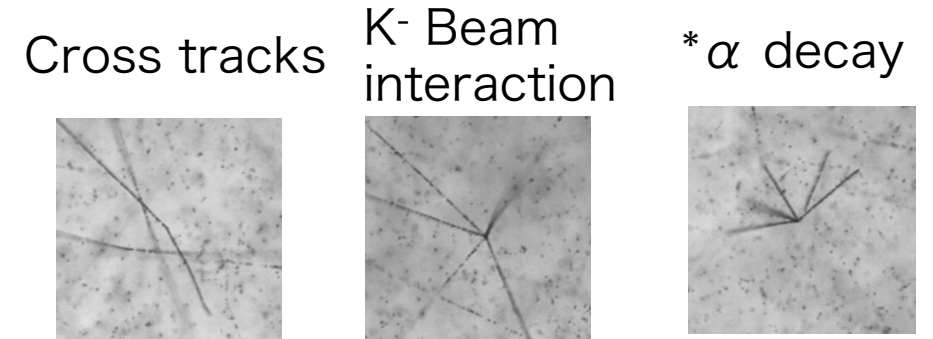
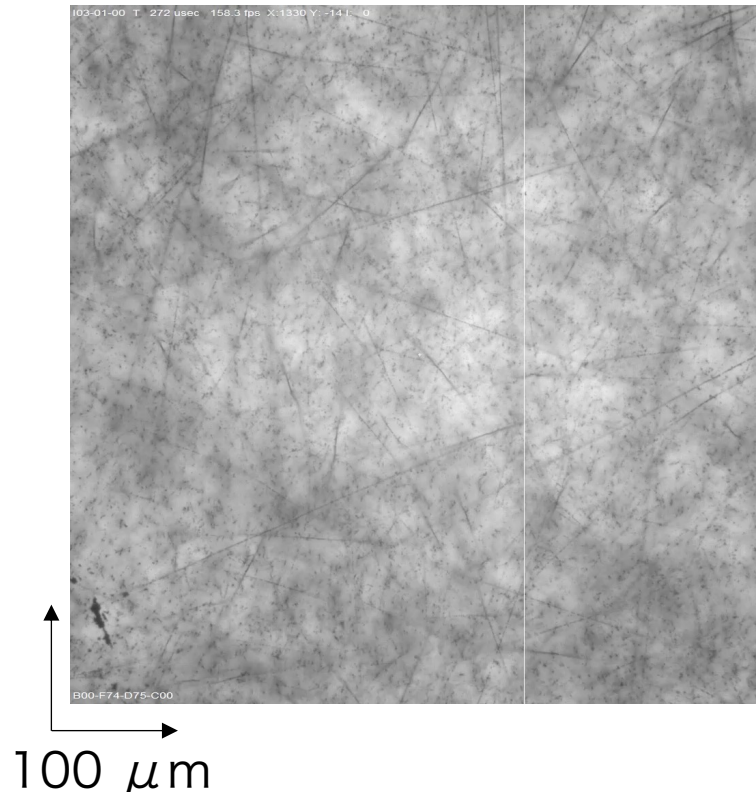
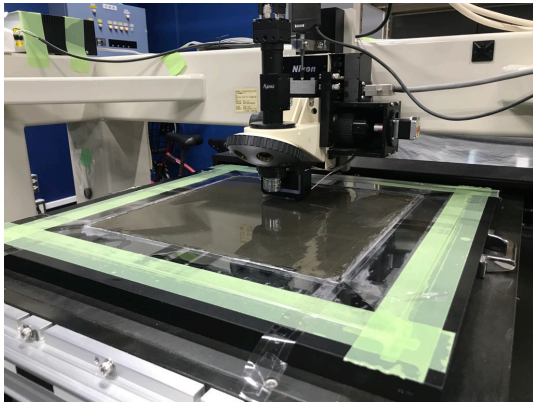
- Hypertriton can be visually observed.



- Measurement with sub- $\mu\text{m}$  spatial resolution
- Invariant mass measurement(event by event)



Scanning stage



\* Th chain  $\alpha$  decay with 5 tracks  
Radio isotope in emulsion

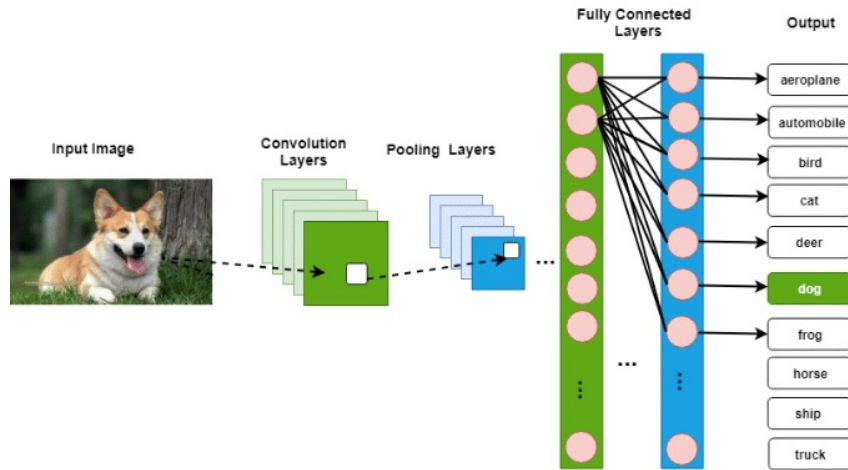
Data size : 140 PB (110 TB/sheet x 1300)  
 $10^9$  vertex like events  
Eye check : 560 years...



We need a Breakthrough!!



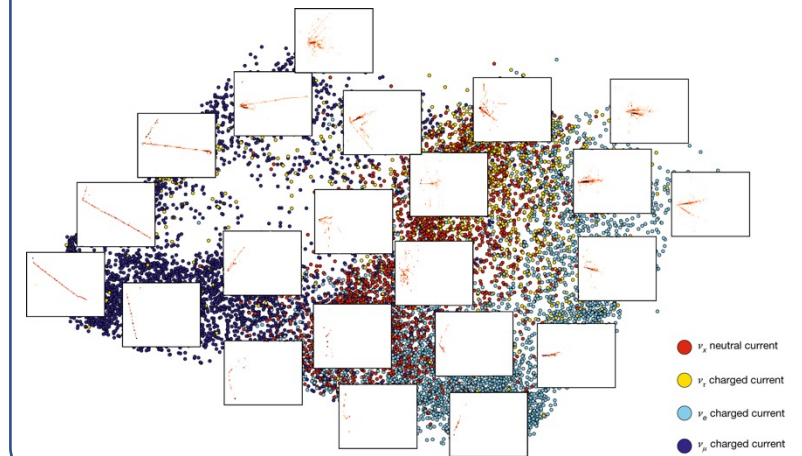
## Image classification



## Object detection

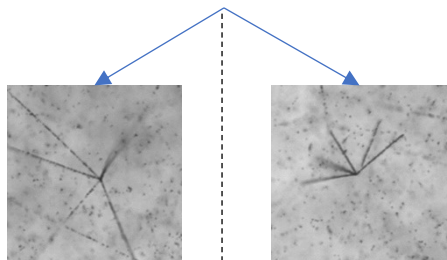


## Event-selection for neutrino interactions



## What is good for us?

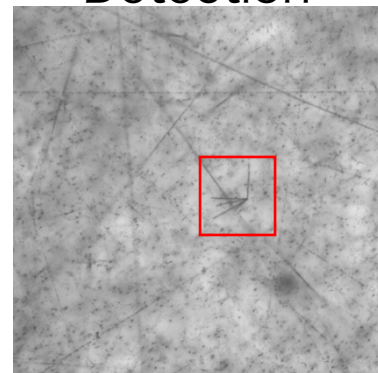
### Classification



A

B

### Detection



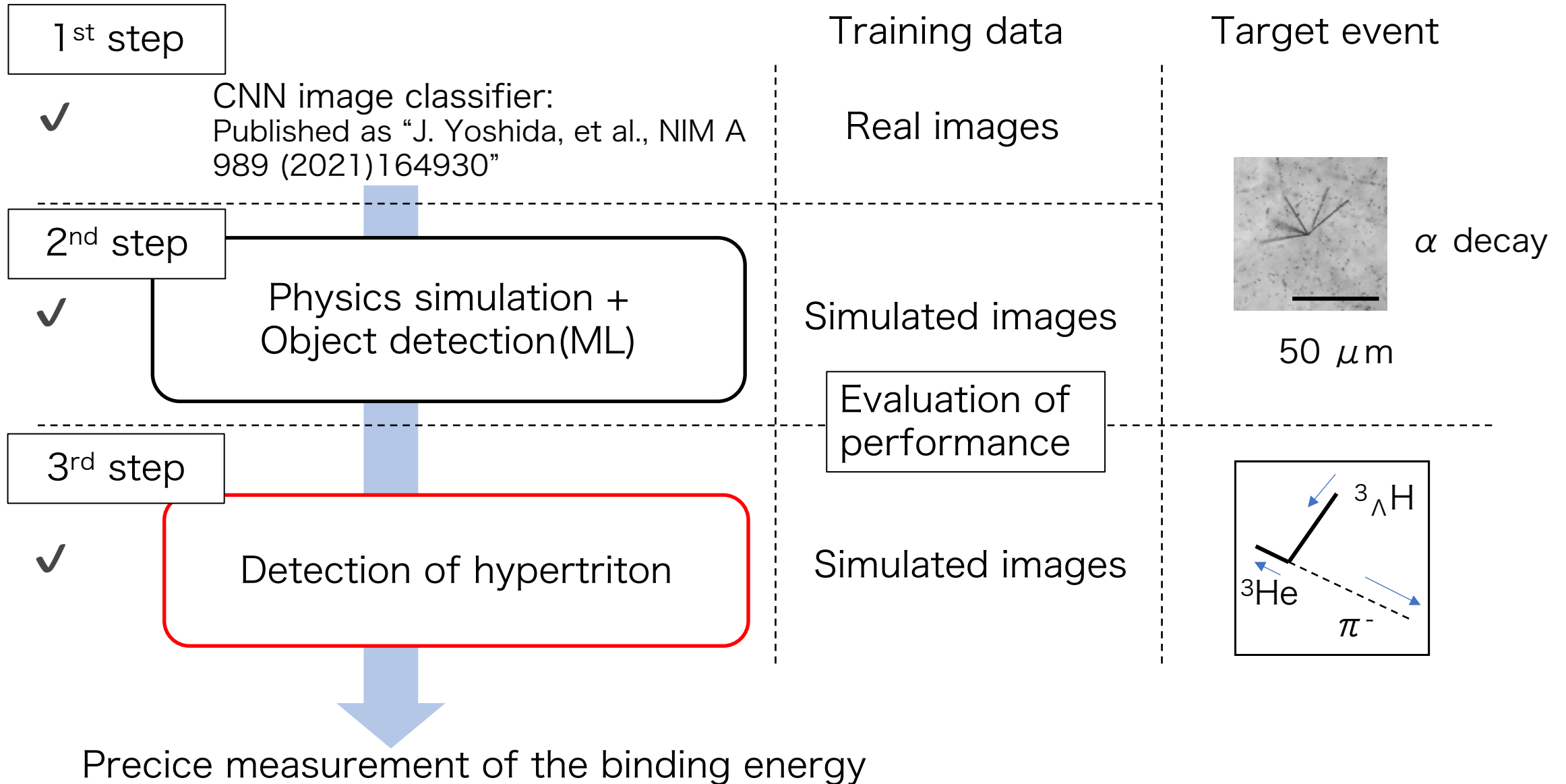
<https://transcranial.github.io/keras-js/#/mnist-cnn>, <http://bdm.change-jp.com/Nature> 560, 41–48 (2018). <https://doi.org/10.1038/s41586-018-0361-2>

- Training dataset:  
At least a few thousand images is necessary.  
“How to train the model for rare events?”

Physics simulation + ML

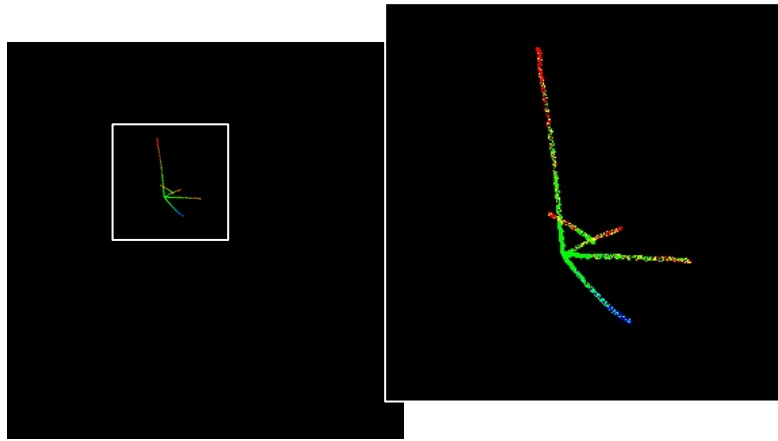
# Strategy of ML + Nuclear Emulsion

5/10

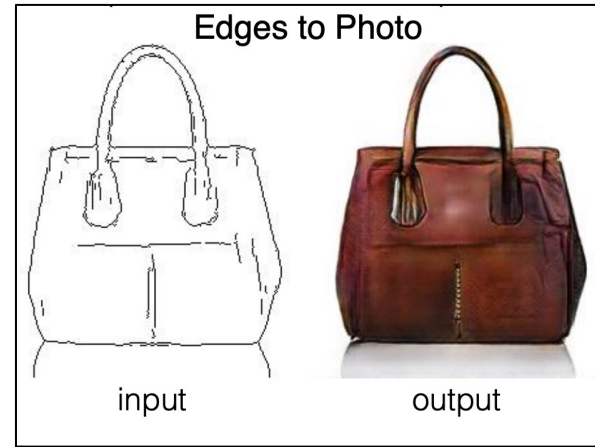


# Training data with Geant4 simulation + ML

$\alpha$  decay event

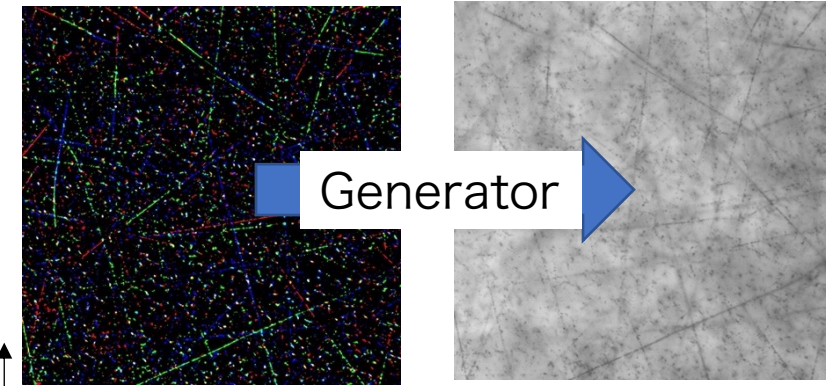


Generative Adversarial Networks (GAN)  
Image transformation by ML.



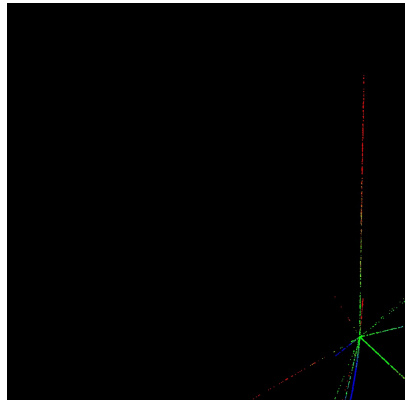
<https://arxiv.org/abs/1611.07004>

Training data (Real images)

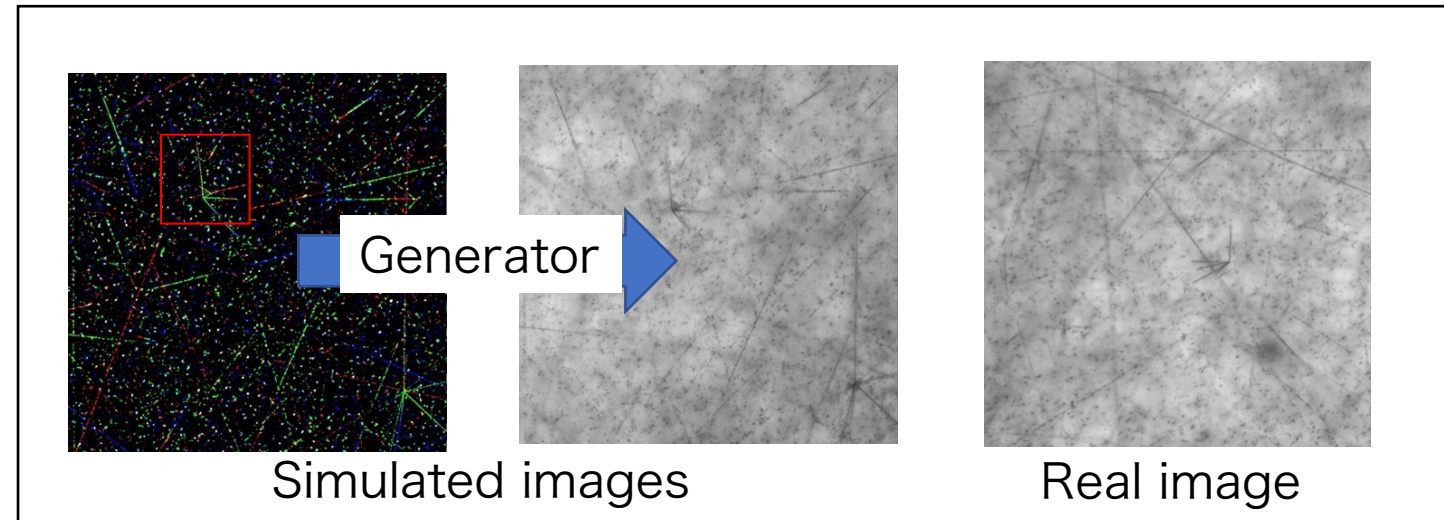
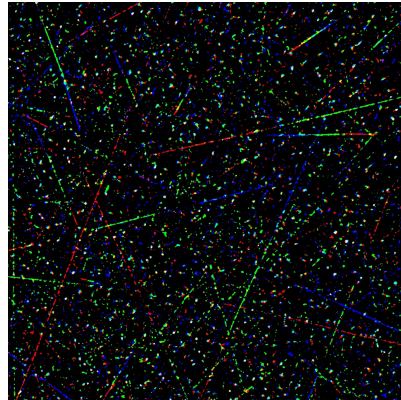


Processed

Other vertex event  
(Negative sample)



Back ground  
From real image



Simulated images

Real image



# Training of Mask R-CNN with Simulated image

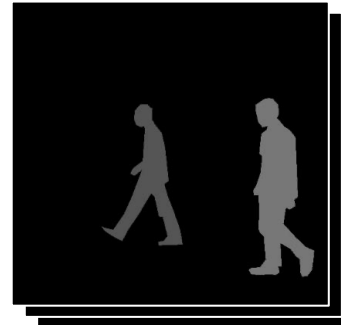
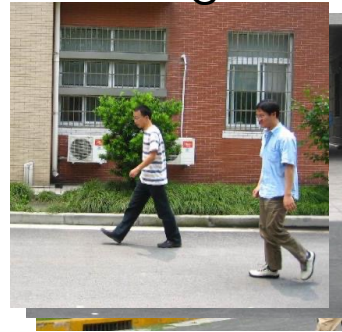
## Mask R-CNN



### Example of training dataset

Image

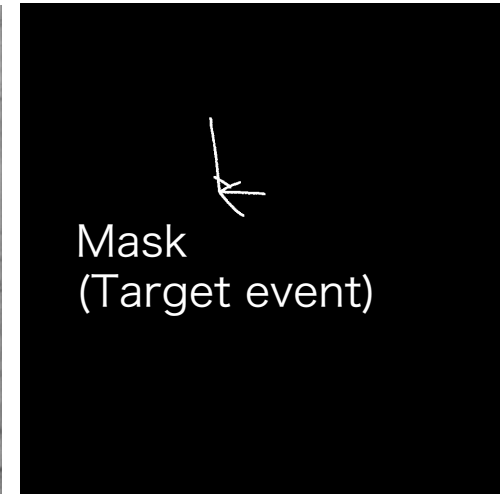
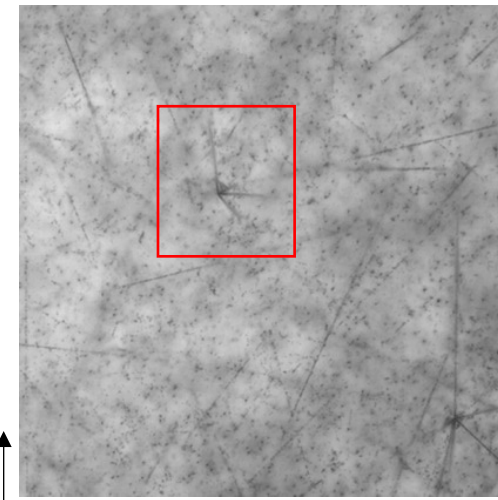
Mask



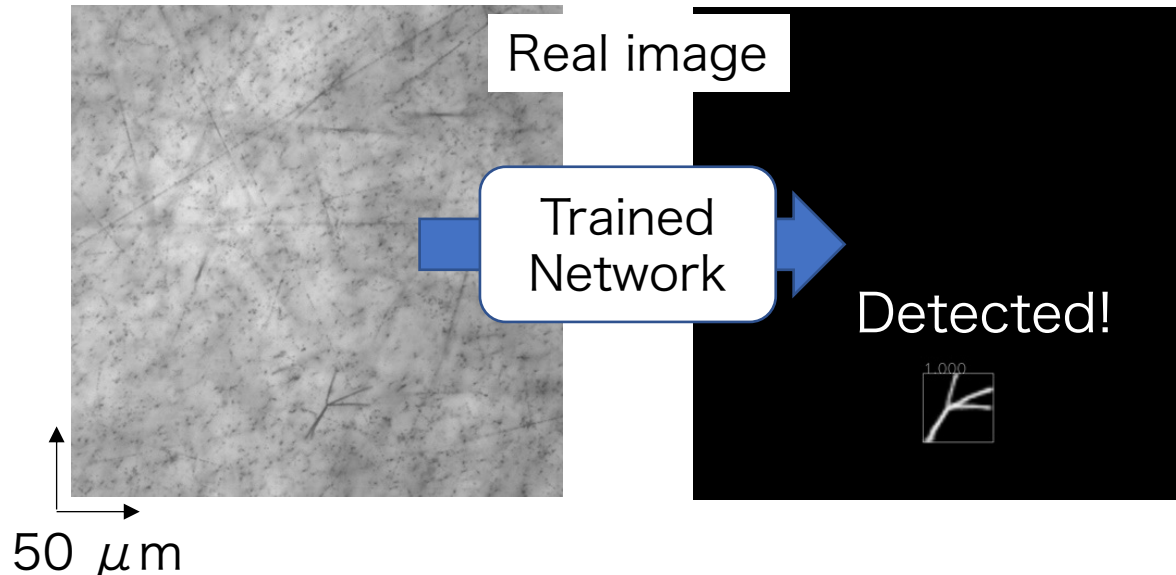
A Pedestrian dataset

[https://www.cis.upenn.edu/~jshi/ped\\_html/](https://www.cis.upenn.edu/~jshi/ped_html/)

### Training data (Simulated image)



50  $\mu\text{m}$



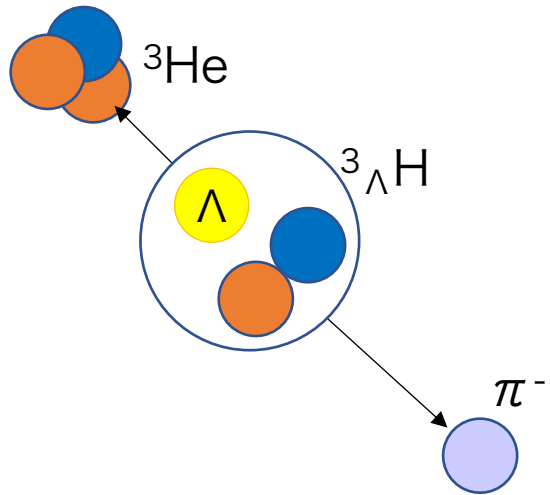
- Training dataset: Simulated images.
- Network can detect events in real images.
- High detection efficiency(80%~90%)

A.Kasagi et.al,  
To be submitted to Computer Physics Communications

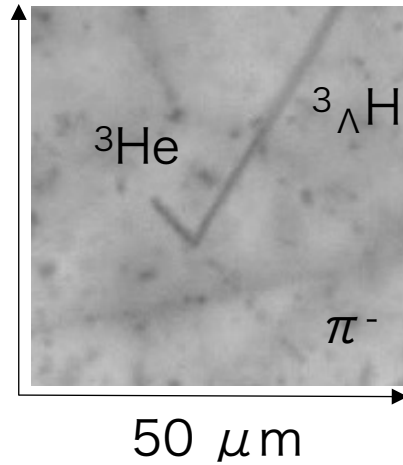
# Hypertriton search by Mask R-CNN

## Rare event detection

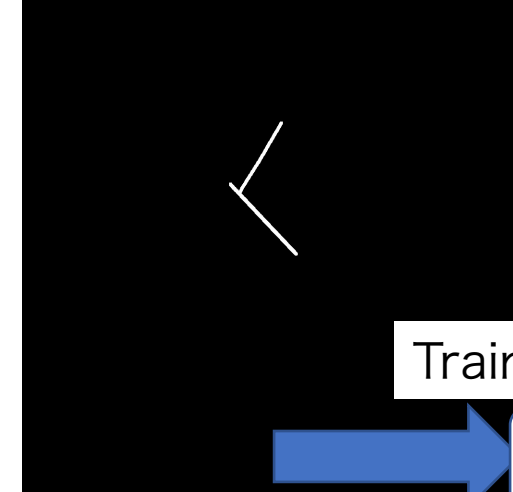
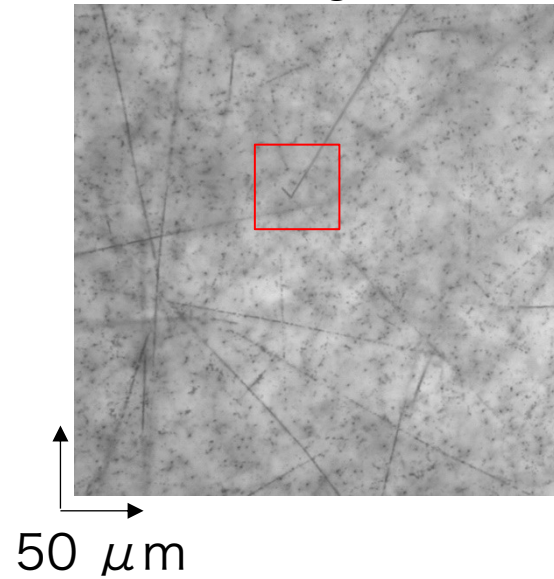
- Two body decay of  ${}^3_{\Lambda}\text{H}$



Simulated image

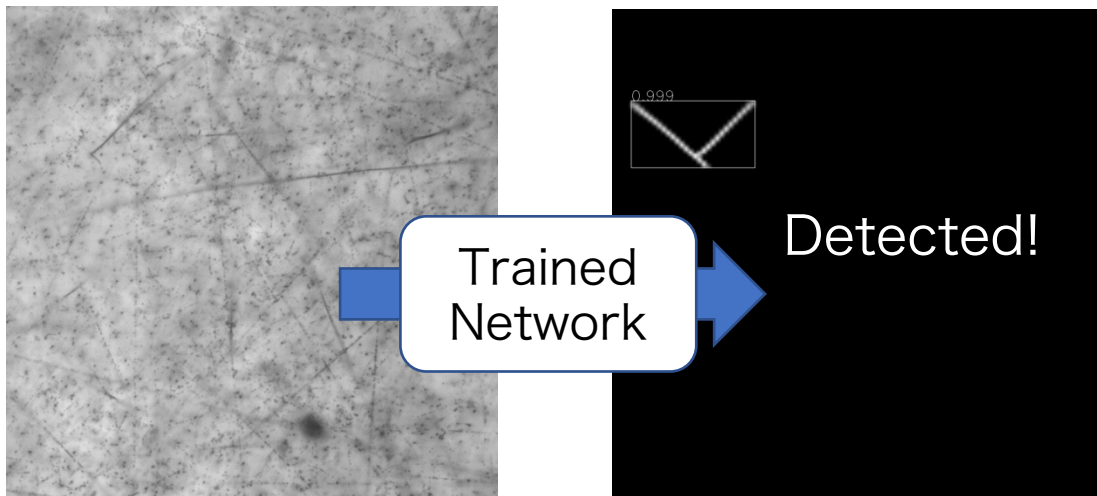


Training dataset (Simulated images)



Network

Real image

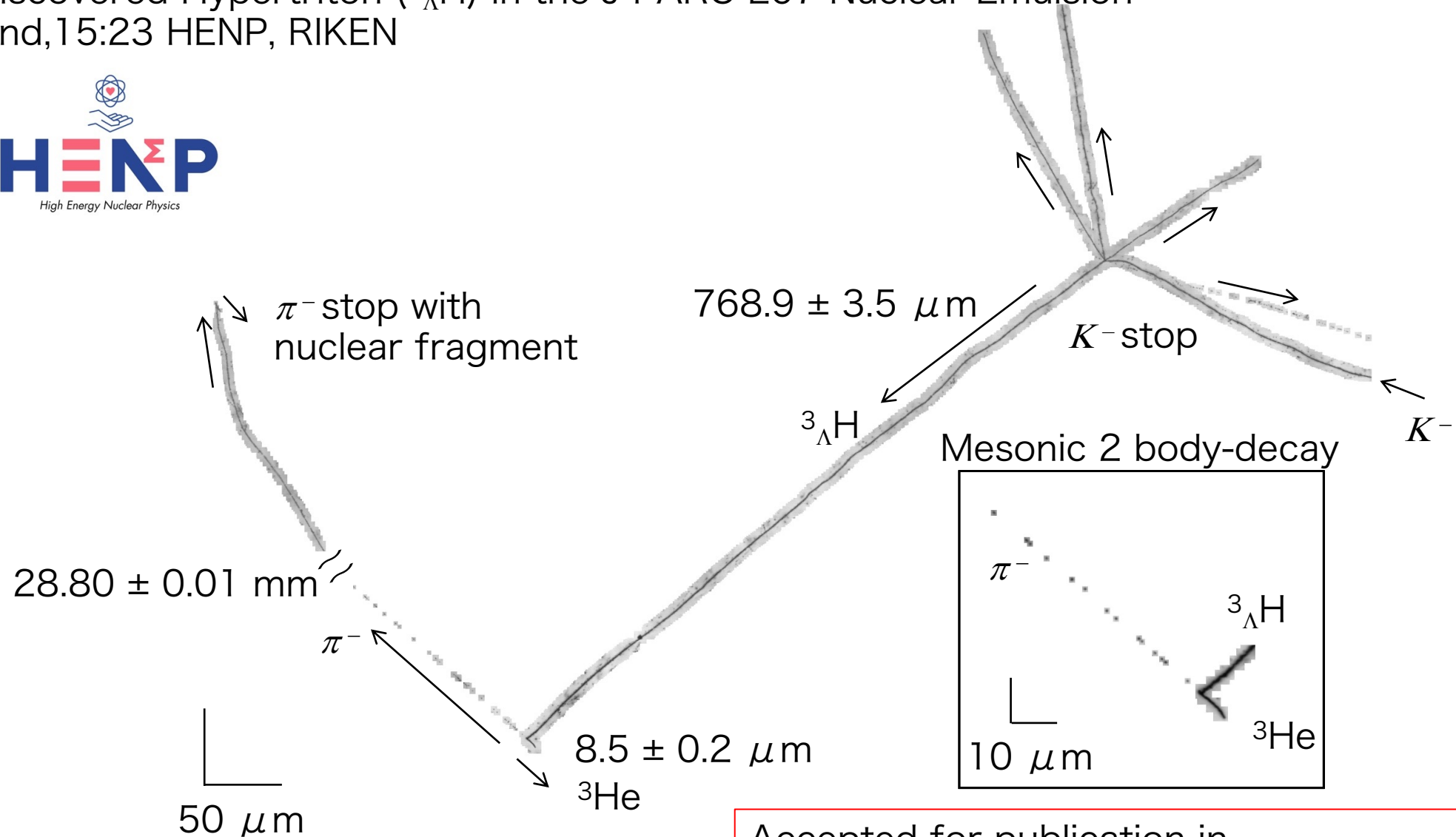
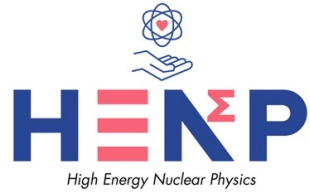


- Training dataset:  $10^4$  simulated  ${}^3_{\Lambda}\text{H}$
- Rare events are detected!



# The First Discovered Hypertriton

The First Discovered Hypertriton ( ${}^3_{\Lambda}\text{H}$ ) in the J-PARC E07 Nuclear Emulsion  
February 2nd, 15:23 HENP, RIKEN



Accepted for publication in  
Nature Reviews Physics as a Perspective article

- Hypertriton puzzle on
  - Lifetime
  - Binding energy
- Measurement of Hypertriton binding energy very precisely
  - J-PARC E07 nuclear emulsion
  - Machine learning
    - Result of binding energy will come soon.
- Precise measurement of binding energy of various hypernuclei
  - ${}^4_{\Lambda}\text{He}$ ,  ${}^5_{\Lambda}\text{He}$ , (3 body decay) ...  ${}^4_{\Lambda\Lambda}\text{H}$ ,  ${}^5_{\Lambda\Lambda}\text{H}$  ... (Double-strangeness hypernuclei)

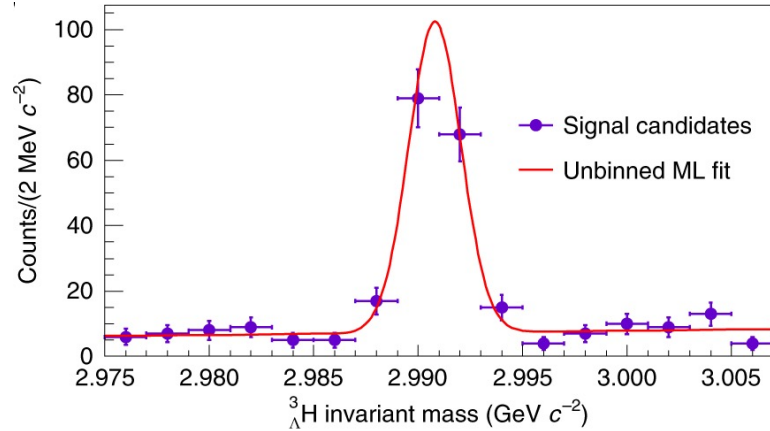
\*A. Kasagi<sup>1,2</sup>, E. Liu<sup>1,3,4</sup>, M. Nakagawa<sup>1</sup>, H. Ekawa<sup>1</sup>, J. Yoshida<sup>1,5</sup>, W. Dou<sup>1,6</sup>, A. Muneem<sup>1,7</sup>, K. Nakazawa<sup>2</sup>, C. Rappold<sup>8</sup>, N. Saito<sup>1</sup>, T R. Saito<sup>1,9,10</sup>, M. Taki<sup>11</sup>, Y K. Tanaka<sup>1</sup>, H. Wang<sup>1</sup>, M. Yoshimoto<sup>12</sup>

1. High Energy Nuclear Physics Laboratory,, RIKEN, 2. Gifu University  
3. Institute of Modern Physics, CAS, 4. University of Chinese Academy of Sciences,  
5. Tohoku University, 6. Saitama University,  
7. Ghulam Ishaq Khan Institute of Engineering Sciences and Technology,  
8. Instituto de Estructura de la Materia, CSIC-Madrid, 9. GSI, 10. Lanzhou University,  
11. Rikkyo University 12. Nishina Center, RIKEN

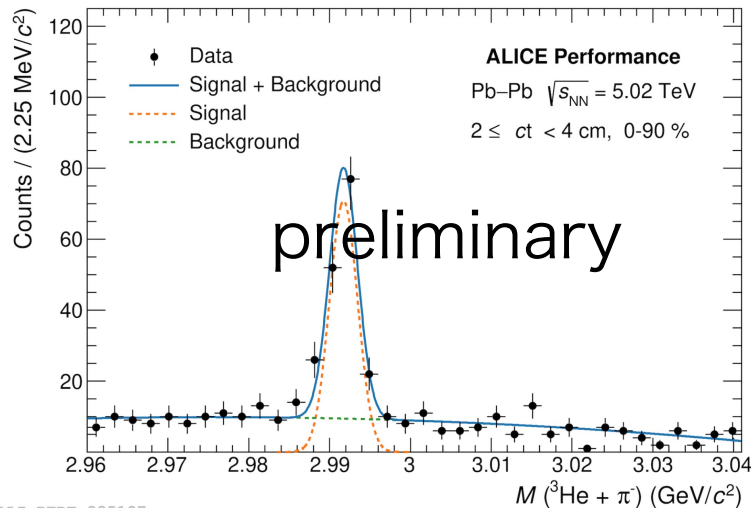
backup

# Binding energy of Hypertriton

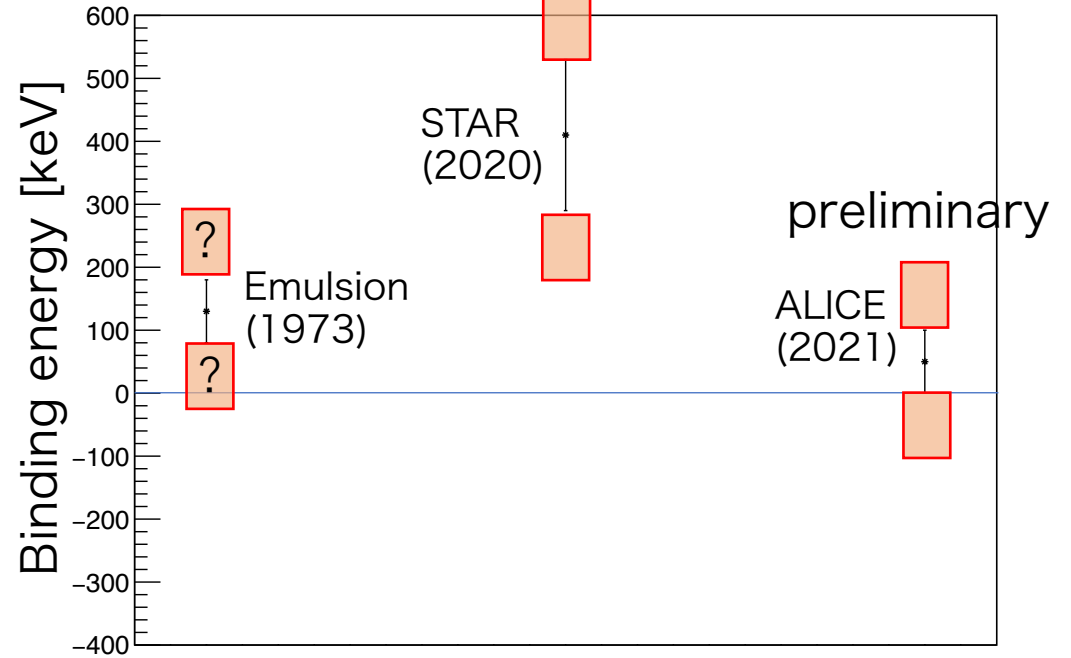
STAR (2020):  $410 \pm 120(\text{stat.}) \pm 110(\text{syst.})$  keV



ALICE (2021):  $\sim 50 \pm 50(\text{stat.}) \pm 90(\text{syst.})$  keV



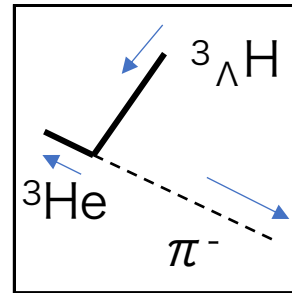
Nuclear emulsion ( $\sim 1973$ )  
 $130 \pm 50(\text{stat.}) \pm ???(\text{syst.})$  keV



- New results on hypertriton binding energy  
 → Importance of evaluating systematic errors
- To understand hypertriton nature  
 → Precise measurement of lifetime and binding energy  
 → New measurement with nuclear emulsion

# Precise measurement with Nuclear Emulsion

- Background-free measurement  
Hypertriton decay can be visually observed.  
Unique topology in two-body decay  
( ${}^3_{\Lambda}\text{H}$ :  $\sim 28$  mm,  ${}^4_{\Lambda}\text{H}$ :  $\sim 42$  mm)



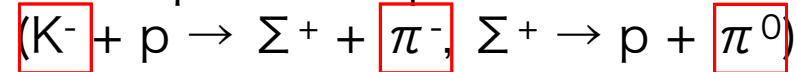
	${}^3_{\Lambda}\text{H}$	${}^4_{\Lambda}\text{H}$	${}^6_{\Lambda}\text{He}$	${}^7_{\Lambda}\text{He}$
He, Li etc... [ $\mu\text{m}$ ]	$\sim 8$	$\sim 8$	$\sim 2$	$\sim 2$
$\pi^{-}$ [mm]	$\sim 28$	$\sim 42$	$\sim 24$	$\sim 28$

$\sim$   
 $\sim$

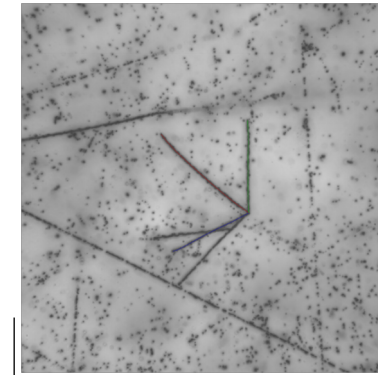
- Calibration

	$\sim 1970\text{s}$	In our analysis
Calibration source	Proton ( $\Sigma^+ \rightarrow p + \pi^0$ )	${}^4\text{He}$ (RI in emulsion)
Volume	$6.5 \times 10^3 \text{ cm}^3$	$5 \text{ cm}^3$

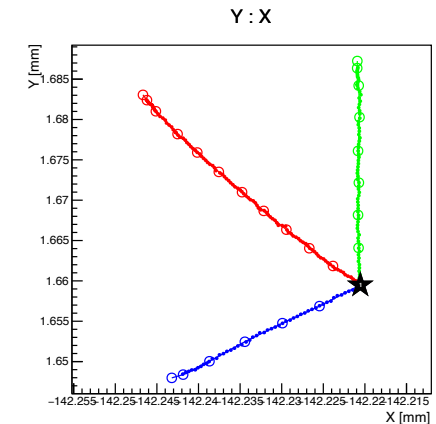
K.E of proton depends on Mass of  $\Sigma^+$ .



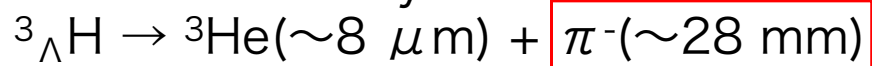
Range measurement for  $\alpha$  tracks by fitting



20  $\mu\text{m}$



- Kinematical analysis



$\sim 1970\text{s}$ : Measurement of Mass of  $\Lambda$   
( $\Lambda \rightarrow p + \pi^-$ )  $\pi^-$ : 10 $\sim$ 20 mm

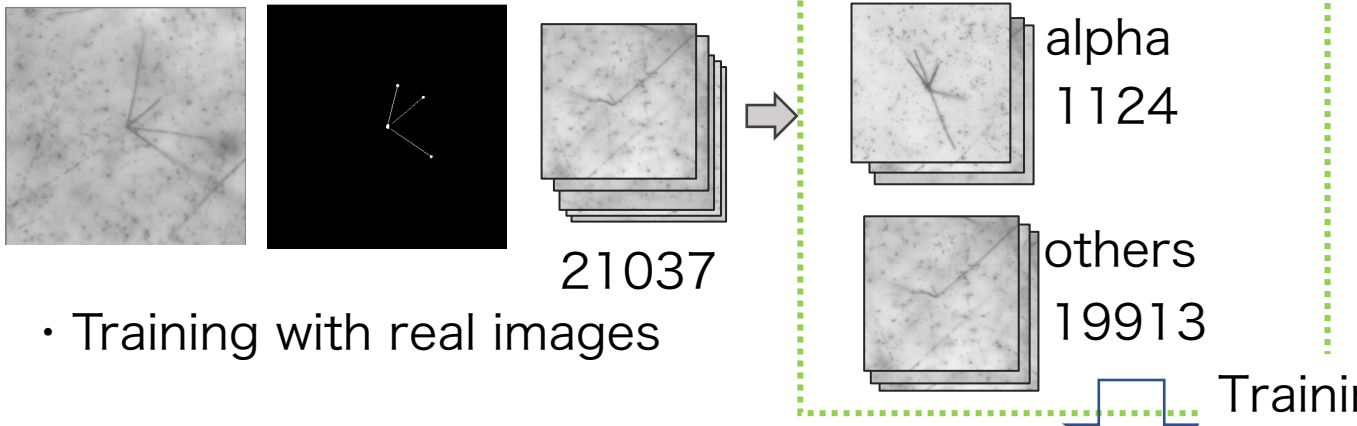
- ${}^3\text{He}$ ,  ${}^4\text{He}$  calibrated by  ${}^4\text{He}$
- Calibration in each small volume  
→ We can achieve  $\sim 30$  keV syst. error.
- Analysis of  ${}^4_{\Lambda}\text{H}$  at the same systematic



# Image classification by CNN filter

Feasibility study of machine learning training with real images

Detected by vertex picker      Training data (real images)

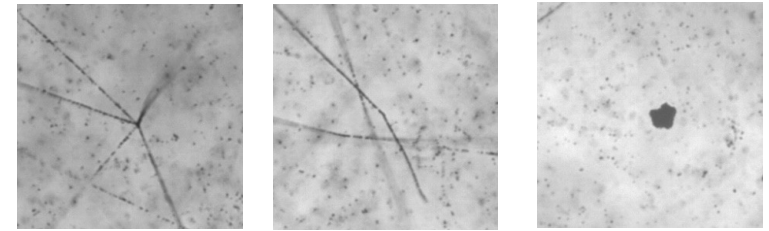


• Training with real images

Test data(not used in training)

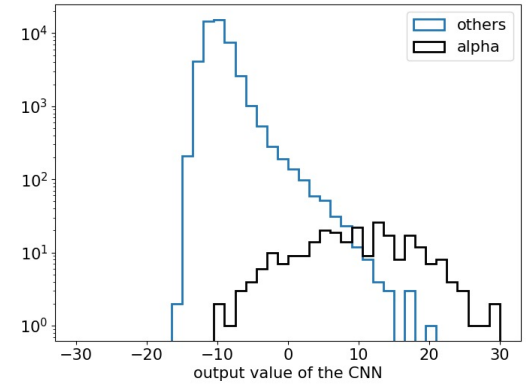


Noise: others

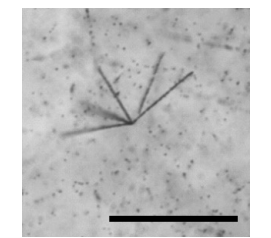


Other interaction      Cross      Dust

Scalar value



Target



50  $\mu$ m  
 $\alpha$  decay

	Purity	Efficiency	No. images
Image processing	$0.081 \pm 0.006$	$0.788 \pm 0.056$	2489
CNN filter	$0.547 \pm 0.025$	0.788	$366 \pm 18$

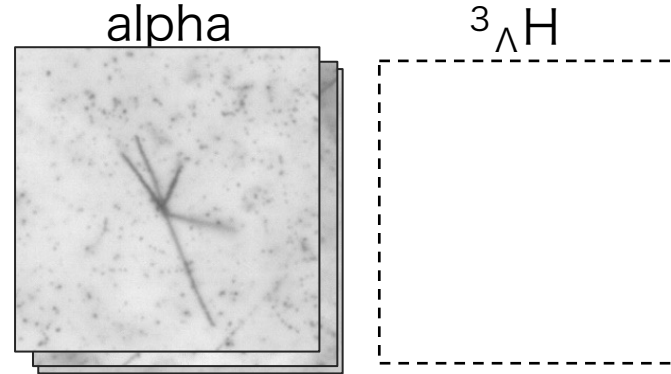
Classification performance was improved by 7 times.

Published as “J. Yoshida, et al., Nuclear Instrument and Method A, 989 (2021) 164930”

# New method for detecting rare events

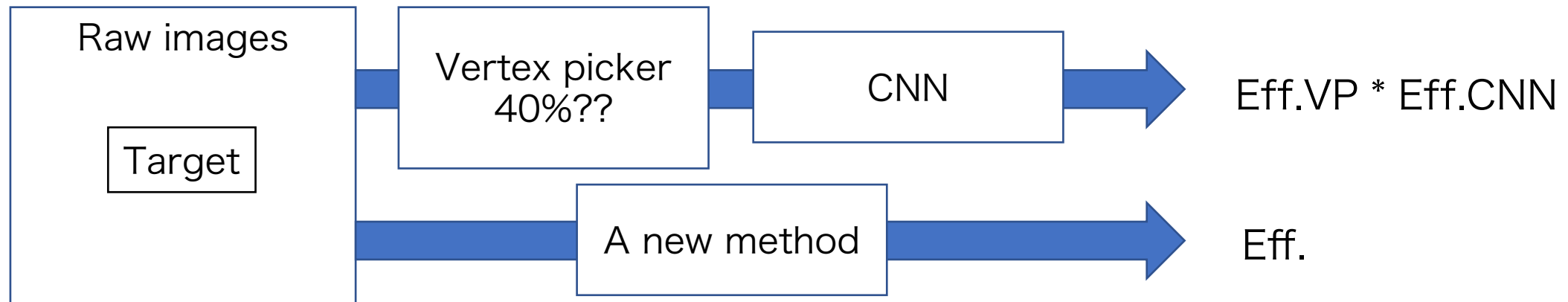
2 issues

- Images for training  
(At least a few thousand~)



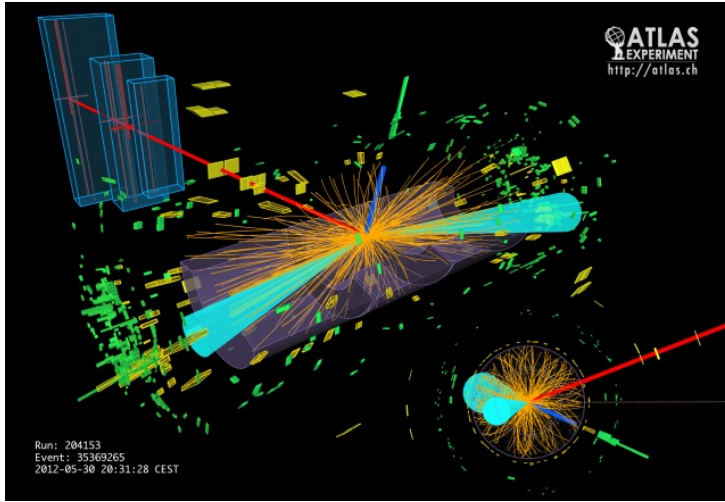
Generating training data : Physics simulation + Image conversion

- Development of a direct detection method  
ex) Vertex picker: Efficiency for alpha decay: ~40%



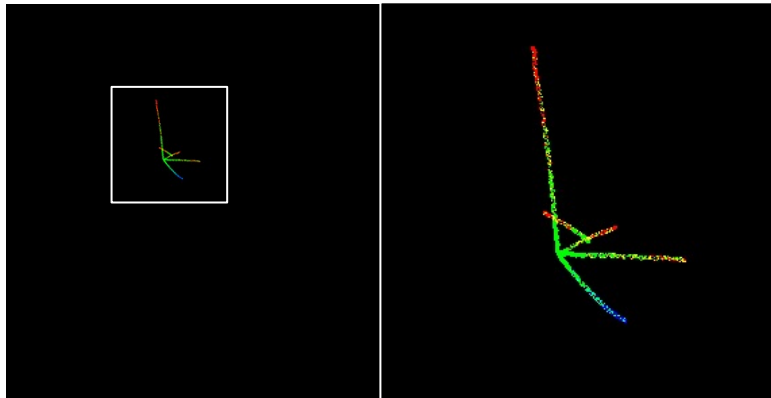
Direct detection method : Object detection by ML

## Geant4 simulation

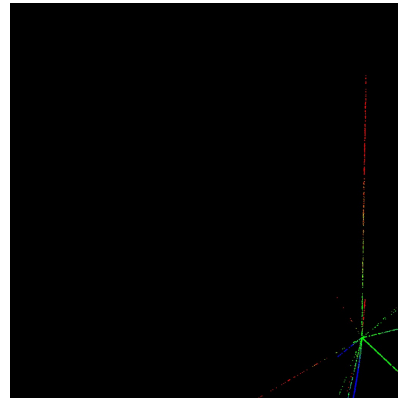


<http://cds.cern.ch/record/1631395>

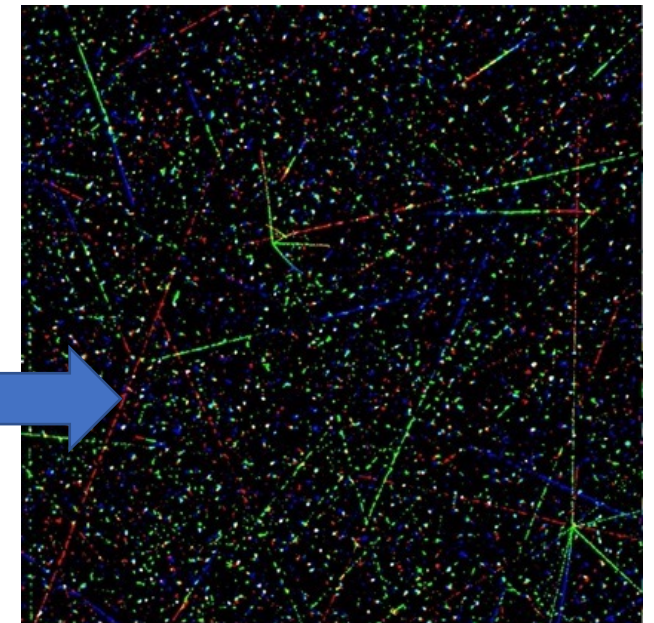
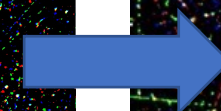
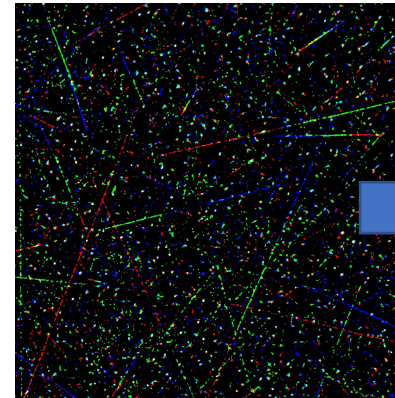
$\alpha$  decay generated by Geant4



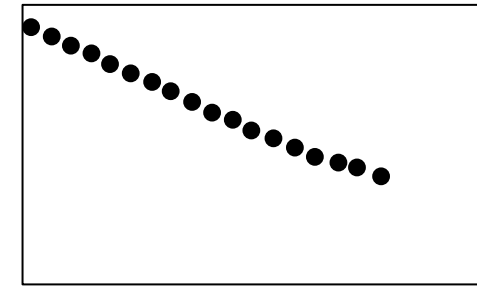
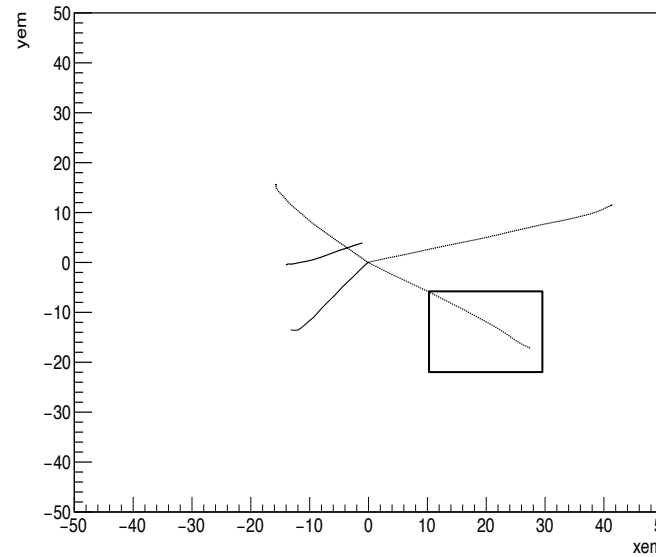
Other vertex event  
(Negative sample)



Back ground  
From real image



yem:xem {ev==1}



Step information of particles



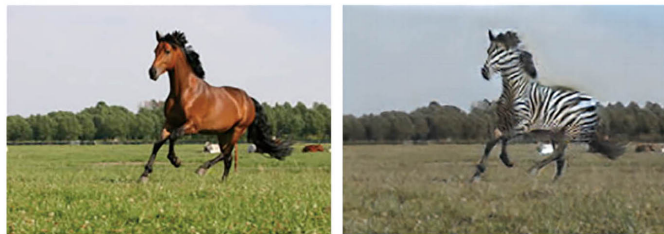
# Generating training data with GAN

pix2pix: Image transformation by ML.

example

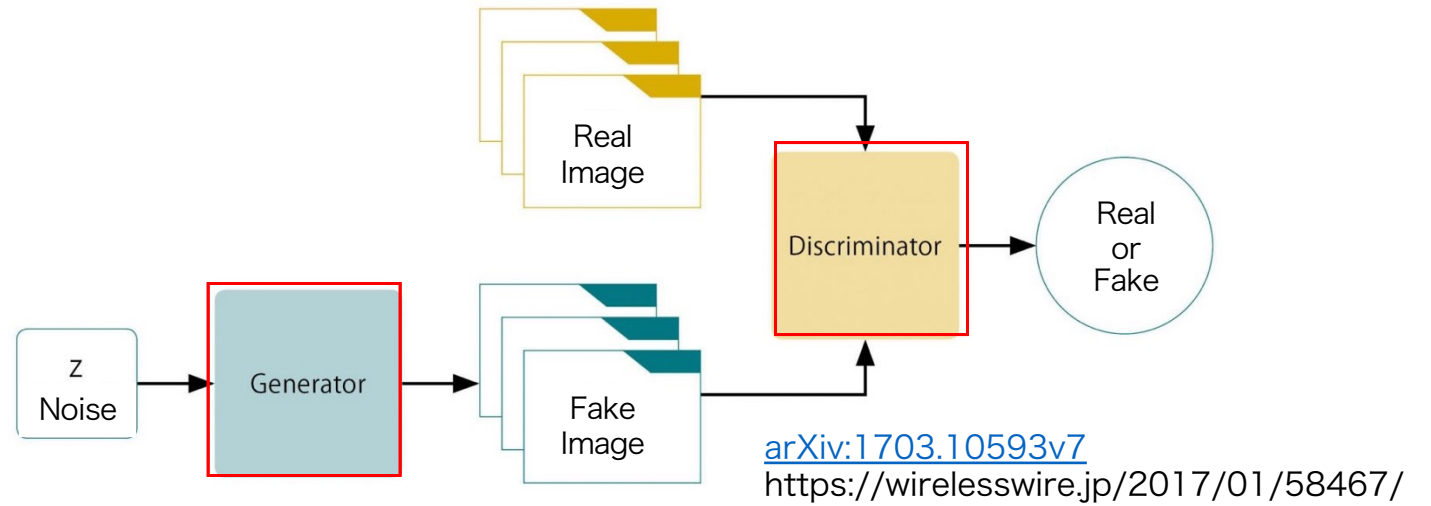


Zebra -> Horse

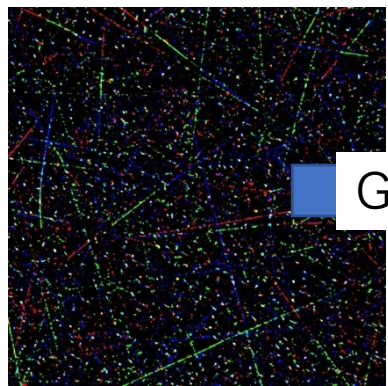


Horse -> Zebra

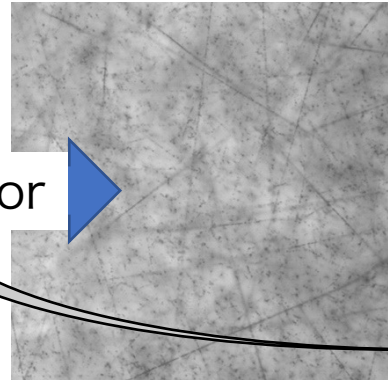
GAN(Generative Adversarial Networks)



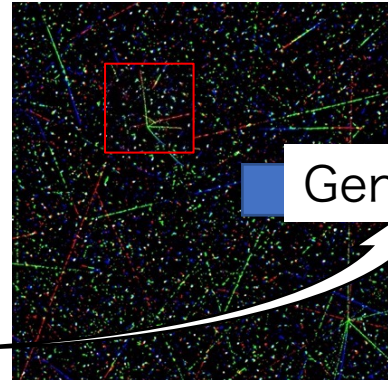
Training data (Real images)



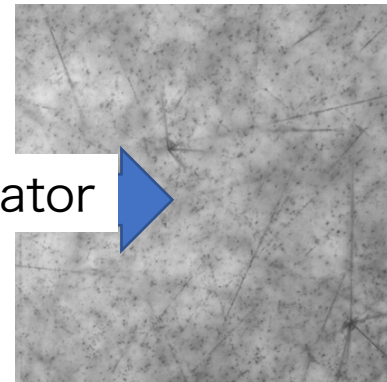
Generator



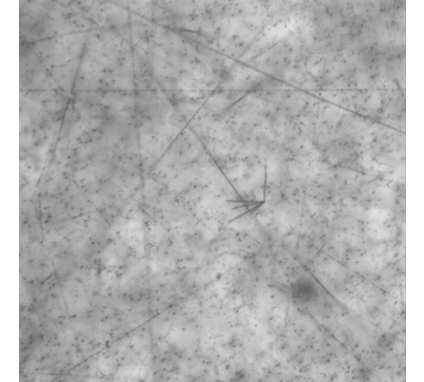
Original



Generator



Simulated

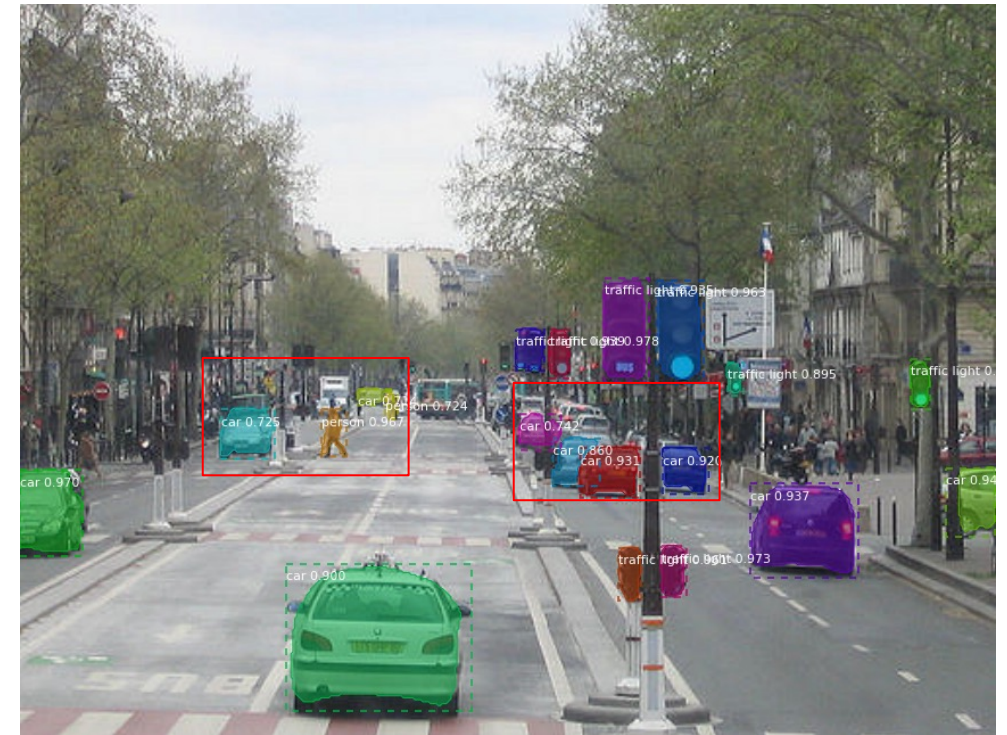
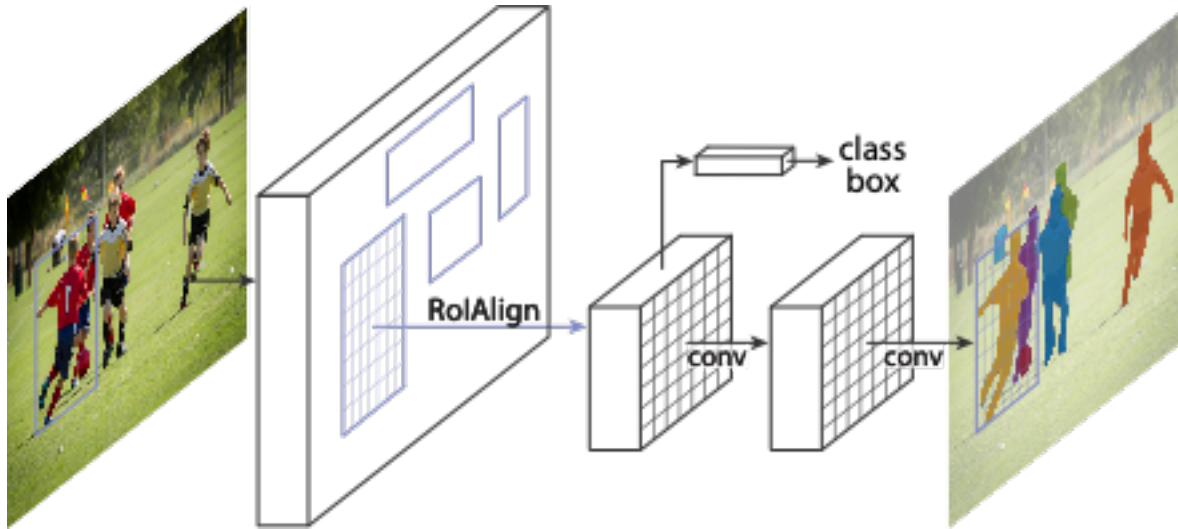


Real

50  $\mu$ m

# Object detection (Mask R-CNN)

- Convolutional operation on the region of interest (ROI)
- Determine the category and bounding box for each object



<https://arxiv.org/abs/1703.06870>  
[https://github.com/matterport/Mask\\_RCNN](https://github.com/matterport/Mask_RCNN)

• Detecting objects directly in images



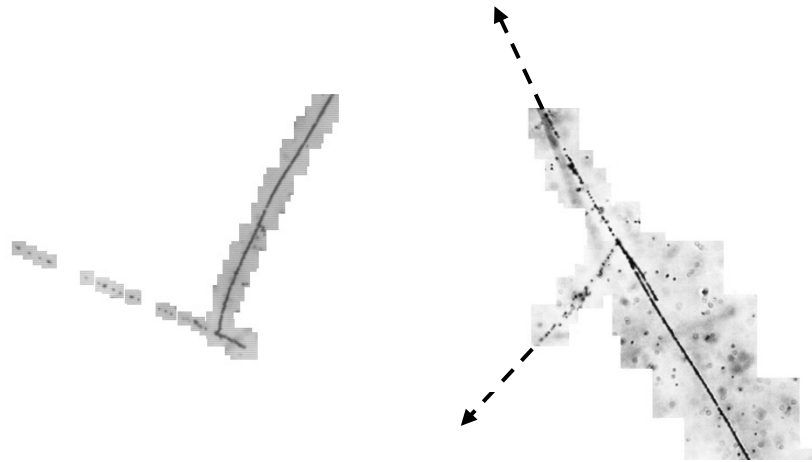
Classification of each object

For crowded region.



# Precise measurement of Hypernuclei

- $^3_{\Lambda}\text{H}$ ,  $^4_{\Lambda}\text{H}$ ,  $^4_{\Lambda}\text{He}$ ,  $^5_{\Lambda}\text{He}$ ...  
(2body & 3 body decay)



Precise measurement  $B_{\Lambda}$

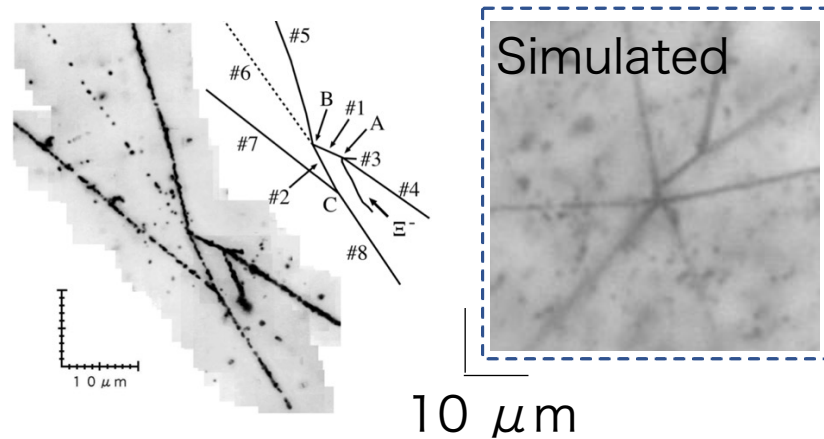
A lot of Single hypernuclei

- $^3_{\Lambda}\text{H}$ :  $\pm 50$  keV (100 events)
- $^4_{\Lambda}\text{H}$ :  $\pm 40$  keV (150 events)
- $^4_{\Lambda}\text{He}$ :  $\pm 30$  keV (200 events)
- $^5_{\Lambda}\text{He}$ :  $\pm 20$  keV ( $\sim 800$  events)
- 
- 
- 

Lifetime measurements  
at FAIR  $\sim$  a few ps accuracy



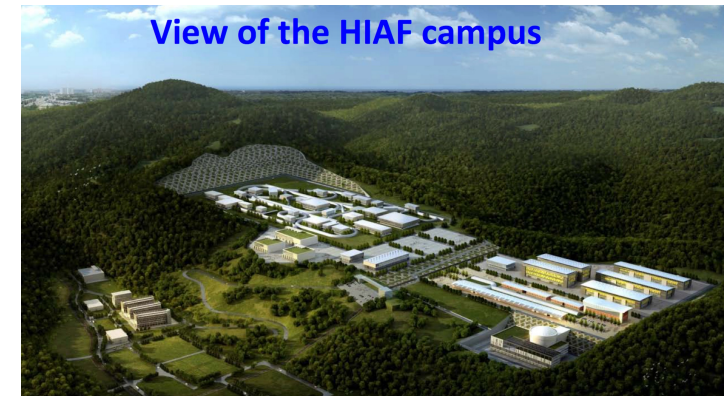
- $^5_{\Lambda\Lambda}\text{H}$ ,  $^6_{\Lambda\Lambda}\text{He}$ ,  $^{15}_{\Xi}\text{C}$



Precise measurement  
 $B_{\Lambda\Lambda}$  and  $B_{\Xi^-}$

- $^6_{\Lambda\Lambda}\text{He}$ :  $\pm 20$  keV (10 $\sim$  events)
- $^{15}_{\Xi}\text{C}$ :  $\pm 30$  keV (10 $\sim$  events)

at HIAF  $\sim 10$  ps accuracy



Precise measurements of Binding energy & Lifetime

## Vertex picker: Line segment analysis by image processing

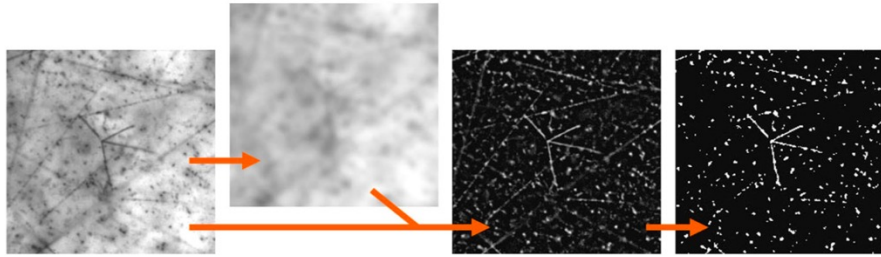


Fig. 4. The first step of the image processing for vertex detection, i.e., raw image, Gaussian-blur, difference of Gaussians and binary thresholding from left to right. The area of the FOV is  $120 \times 120 \mu\text{m}^2$ , which are cropped from the original FOV having  $1140 \times 200 \mu\text{m}^2$ .

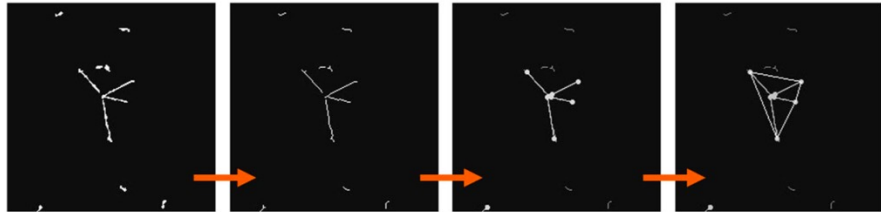
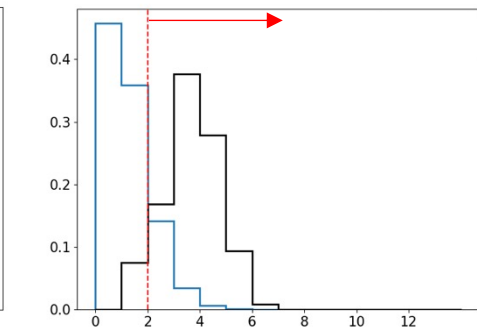
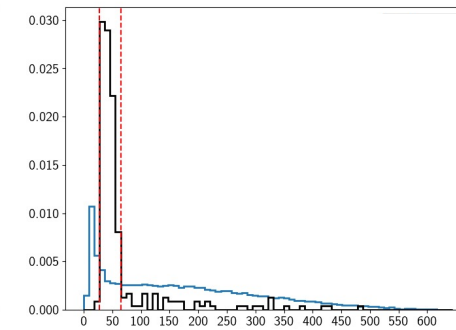
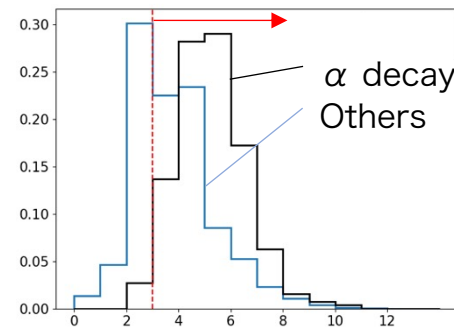
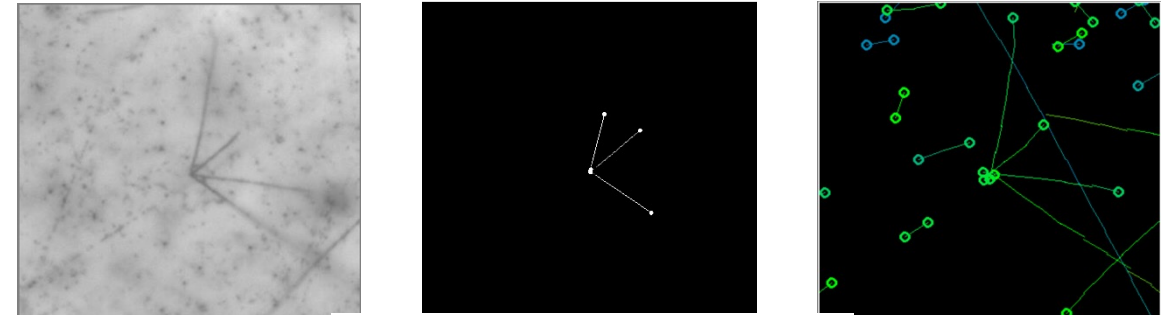
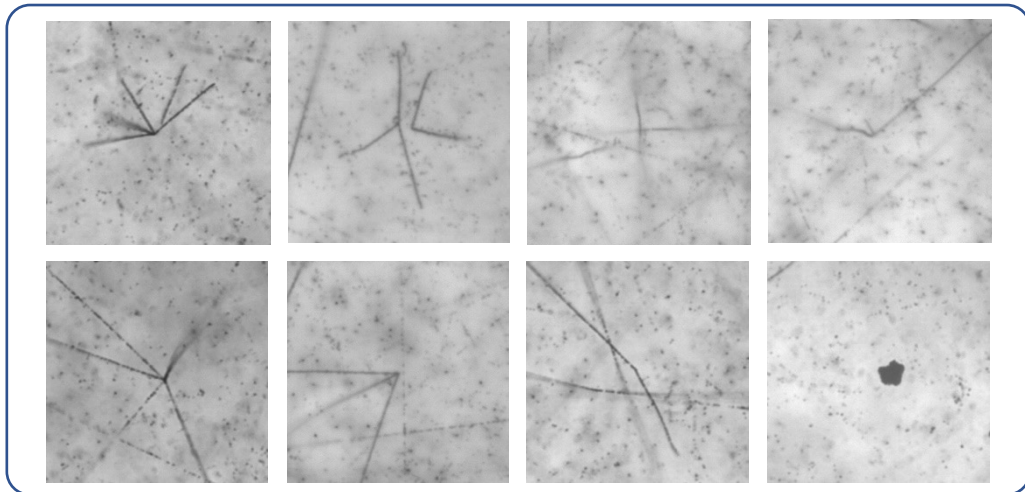
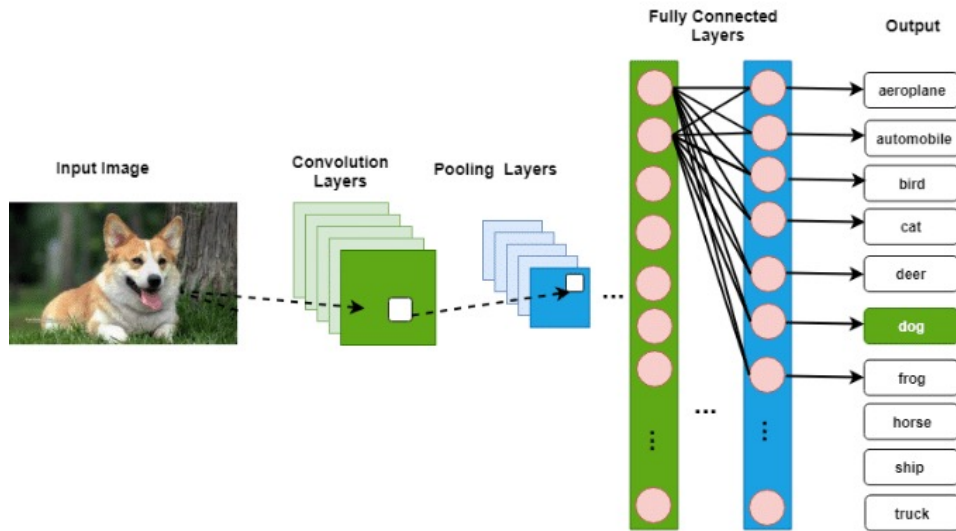


Fig. 5. The second and third step of the image processing: noise reduction, thinning, line detection and vertex detection.

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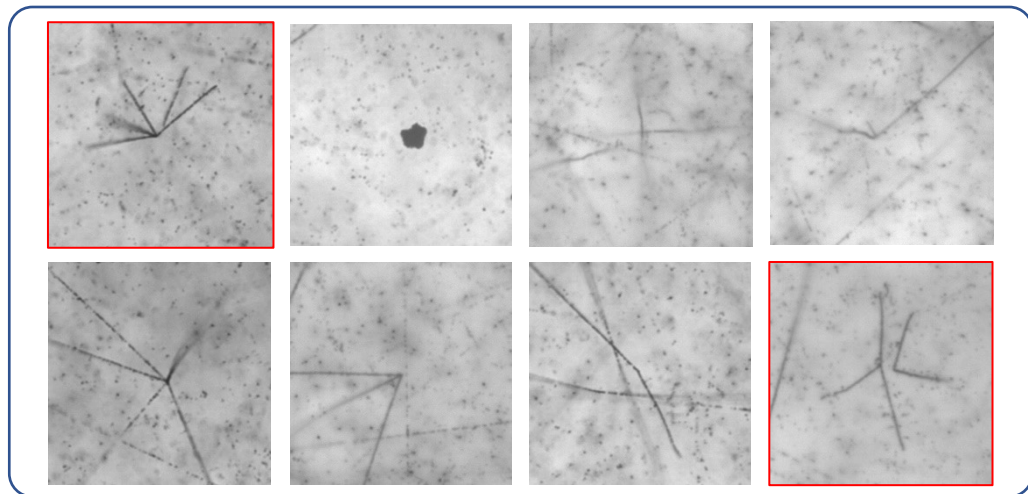


## Convolutional Neural Network

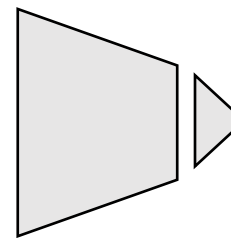


- images and correct answer labels (training data)
- Quantify features by convolutional operations
- Iteratively updating the calculation weights (parameters) in each layer

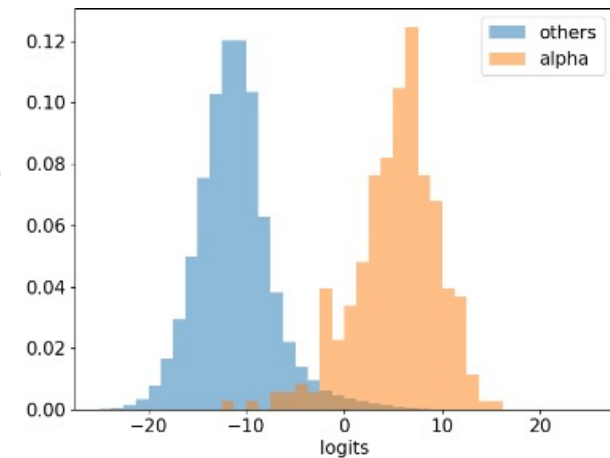
→ Making the best model



Model



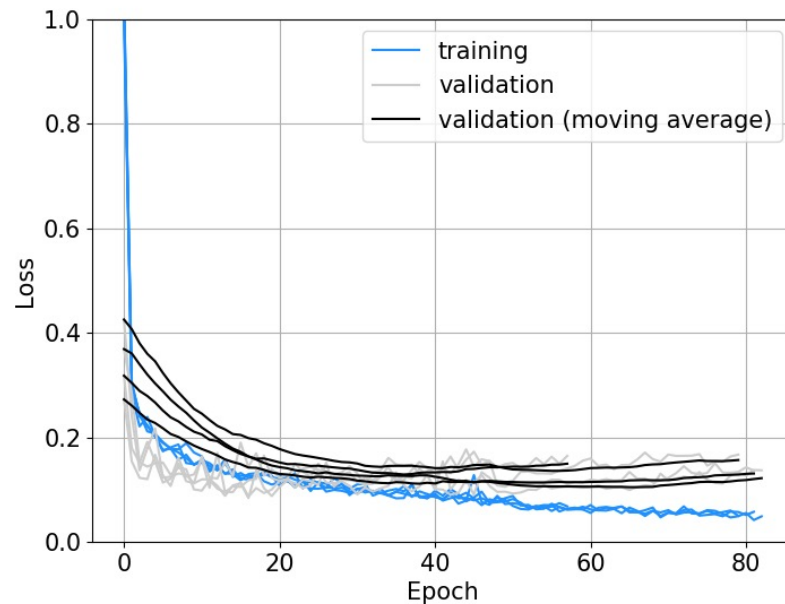
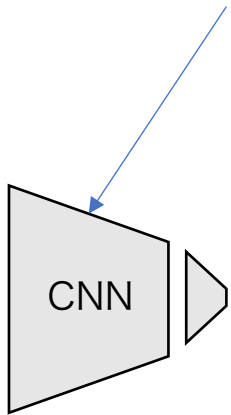
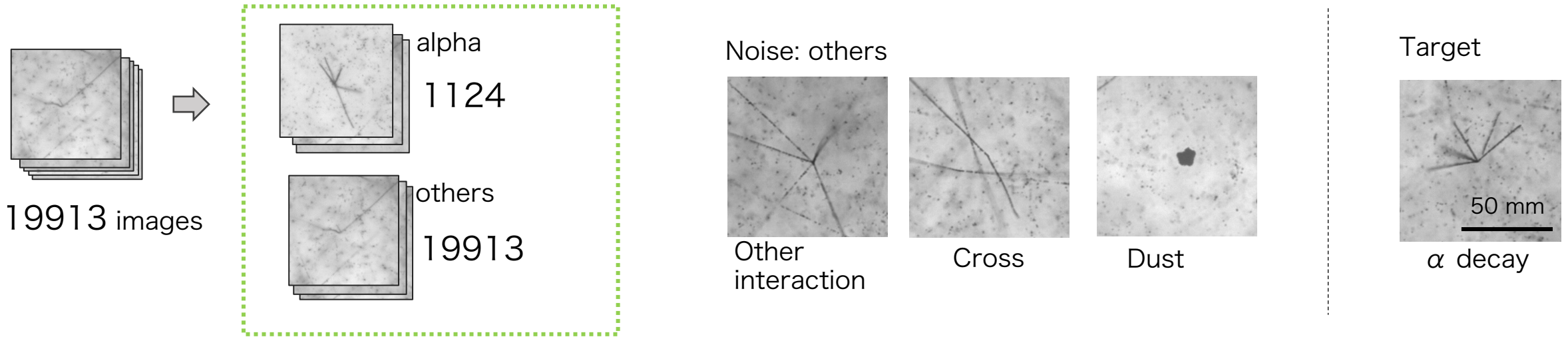
Noise



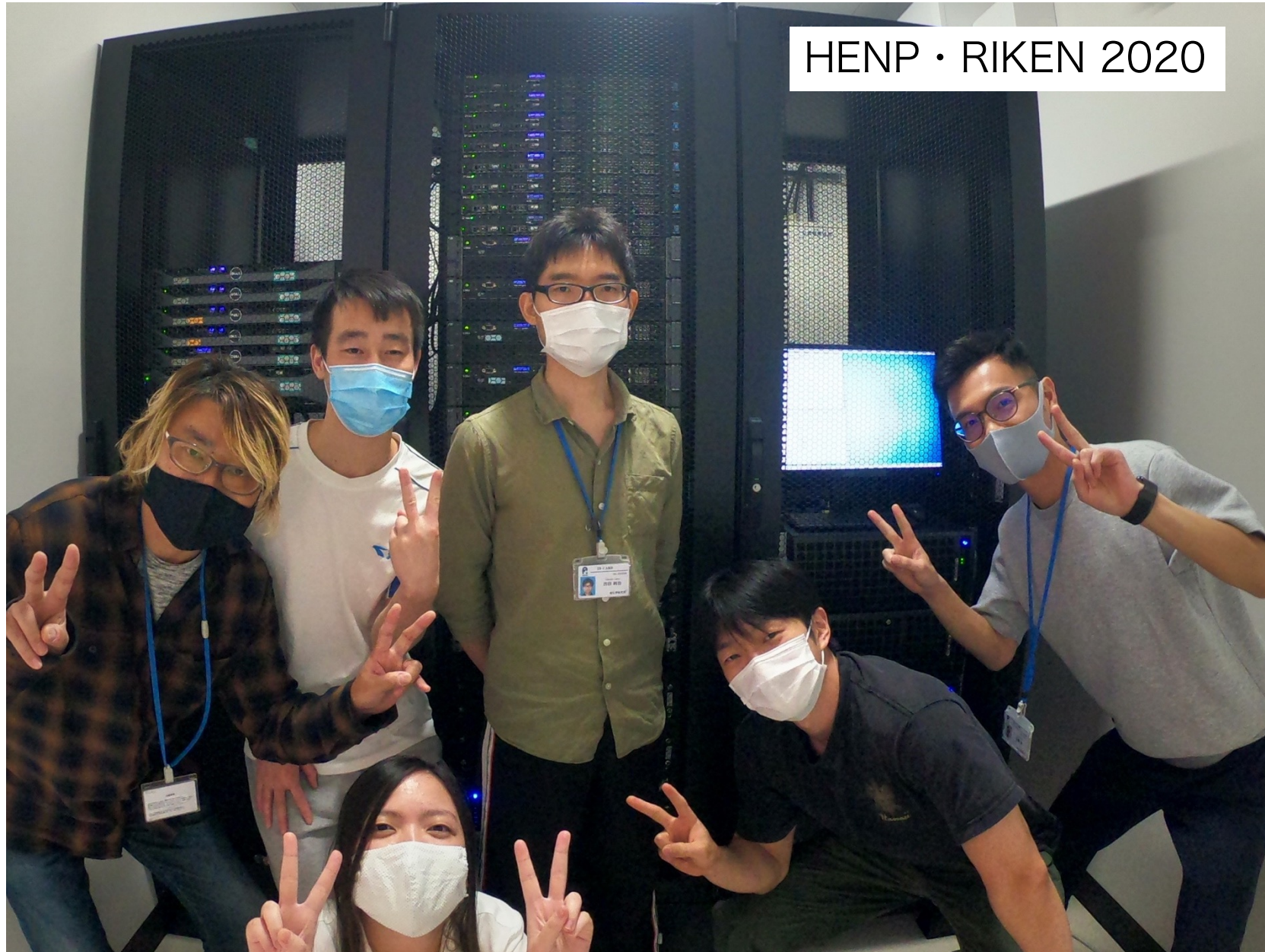
Signal

# Training of CNN model

We classified 20k images to make train







CPU: 1400 cores  
GPU: 36 GPU boards  
Storage: 1000 TB