

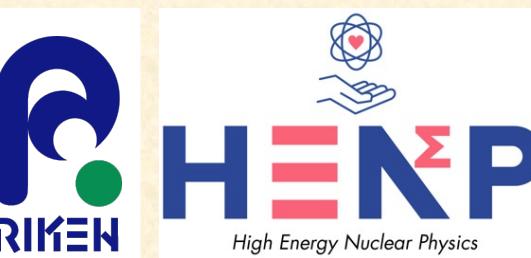
# Studies of Hypernuclei with HI-beams, Nuclear Emulsions and Machine Learning

Take R. Saito for the WASA-FRS HypHI collaboration, the Super-FRS Experiment Collaboration, and the Emulsion-ML collaboration

*High Energy Nuclear Physics Laboratory,  
Cluster for Pioneering Research,  
RIKEN,  
Japan*

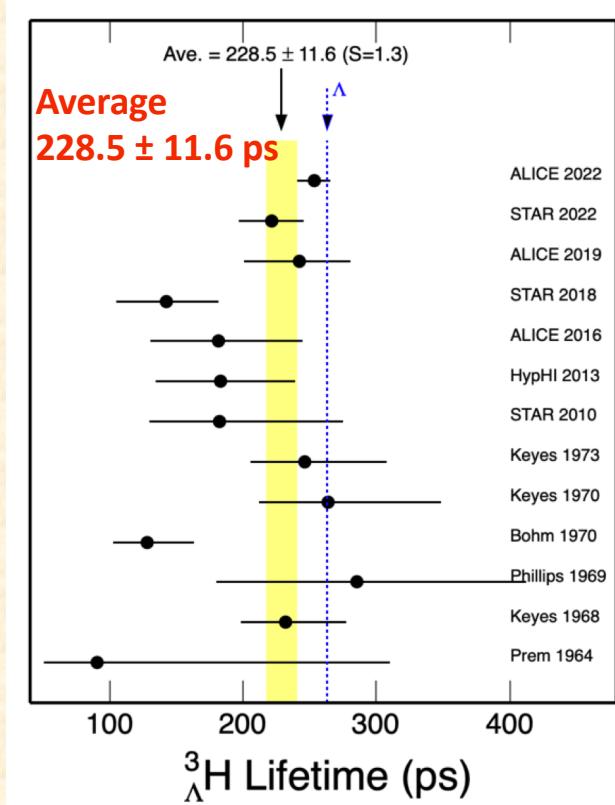


*HRS-HYS Research Group  
(High ReSolution - HYpernuclear Spectroscopy),  
FRS/NUSTAR department,  
GSI Helmholtz Center for Heavy Ion Research,  
Germany*



# Recent hot topics for few-body hypernuclei

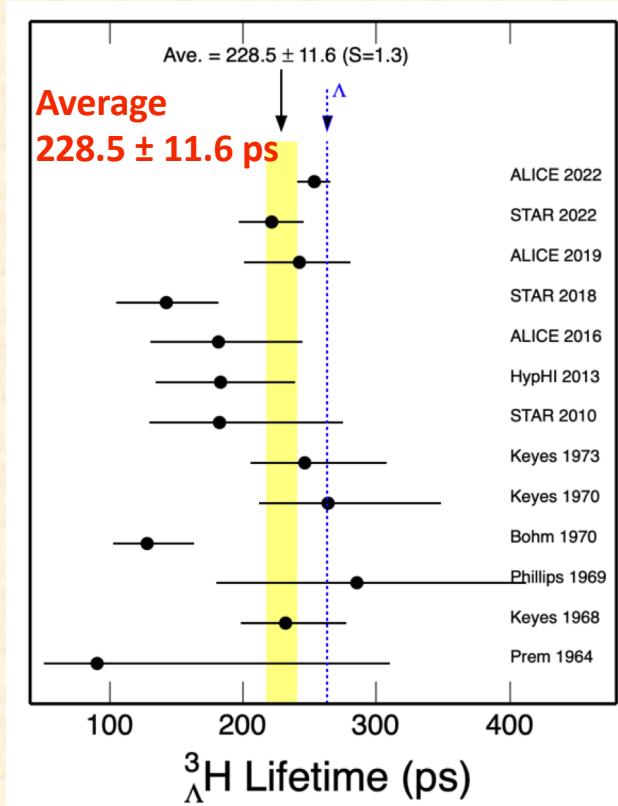
## On hypertriton



Talk by Zhangbu Xu on Monday

# Recent hot topics for few-body hypernuclei

## On hypertriton



### ${}^3\Lambda H$ Binding energy

$B\Lambda({}^3\Lambda H) : 0.13 \pm 0.05$  MeV

G. Bohm et al., NPB 4 (1968) 511

M. Juric et al., NPB 52 (1973) 1

### STAR (2020)

$0.41 \pm 0.12 \pm 0.11$  MeV

STAR Collaboration,  
Nat. Phys. 16 (2020) 409

### ALICE

$0.072 \pm 0.063 \pm 0.036$  MeV

arXiv.2209.07360 (2022)

talk by Chiara Pinto on Monday

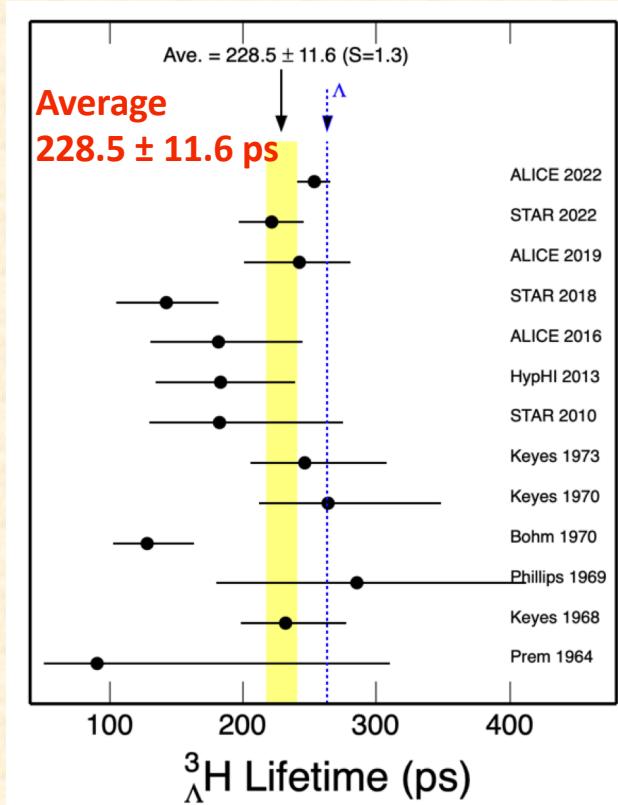
$0.102 \pm 0.063$  MeV

Talk by Stefania Bufalino  
in the Hadron2023 conf. (2023)

# Recent hot topics for few-body hypernuclei

on  $\Lambda nn$

## On hypertriton



Talk by Zhangbu Xu on Monday

## ${}^3\Lambda H$ Binding energy

$$B\Lambda({}^3\Lambda H) : 0.13 \pm 0.05 \text{ MeV}$$

G. Bohm et al., NPB 4 (1968) 511

M. Juric et al., NPB 52 (1973) 1

## STAR (2020)

$$0.41 \pm 0.12 \pm 0.11 \text{ MeV}$$

STAR Collaboration,  
Nat. Phys. 16 (2020) 409

## ALICE

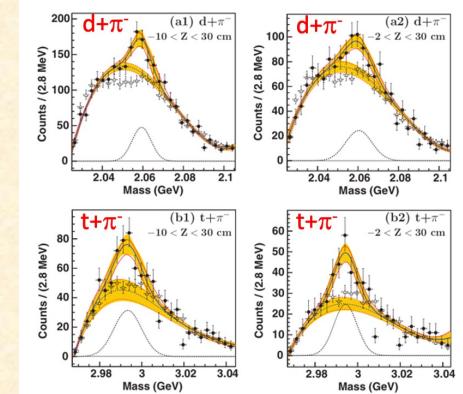
$$0.072 \pm 0.063 \pm 0.036 \text{ MeV}$$

arXiv.2209.07360 (2022)

talk by Chiara Pinto on Monday

$$0.102 \pm 0.063 \text{ MeV}$$

Talk by Stefania Bufalino  
in the Hadron2023 conf. (2023)



HypHI., PRC 88 (2013) 041001

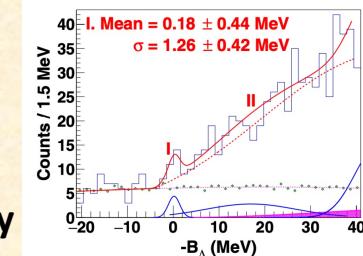


FIG. 5. The enlarged mass spectrum around the  $\Lambda nn$  threshold. Two additional Gaussians were fitted together with the known contributions (the accidentals, the  $\Lambda$  quasifree, the free  $\Lambda$ , and the  ${}^3\text{He}$  contamination). The one at the threshold is for the small peak, while the broad one is for the additional strength above the predicted quasifree distribution.

JLab E12-17-003., PRC 105 (2022) L051001

Talk by Liguang Tang on Wednesday

# Our projects

## With heavy-ion beams

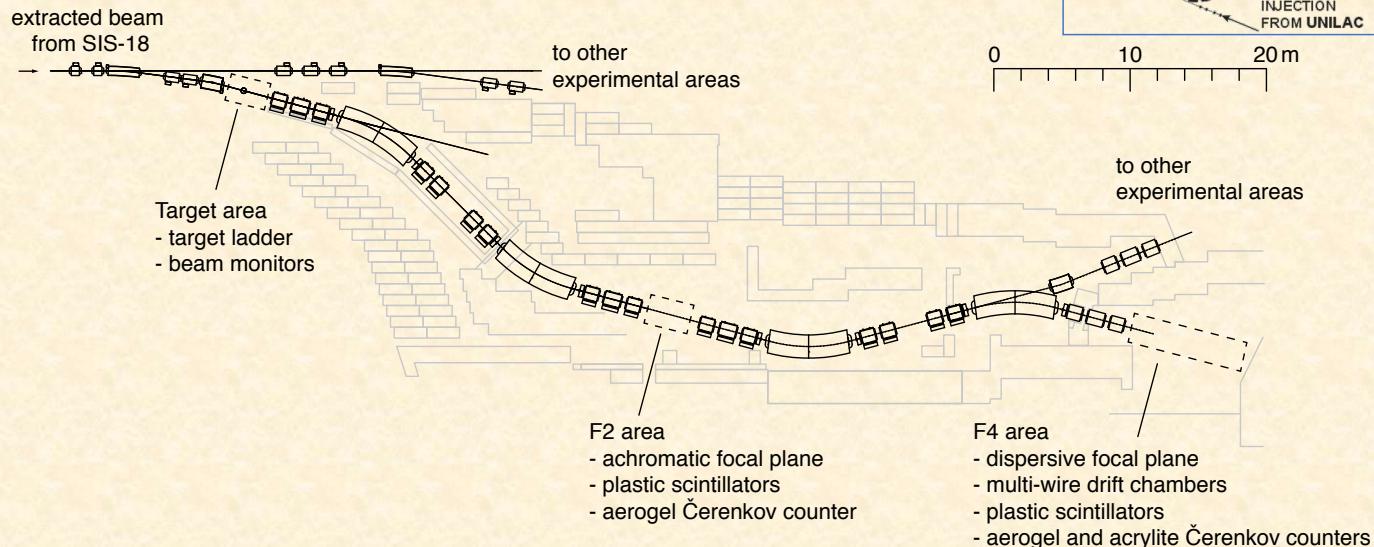
- Lifetime of light hypernuclei including hypertriton
- $\Lambda$ nn states

## With nuclear emulsions

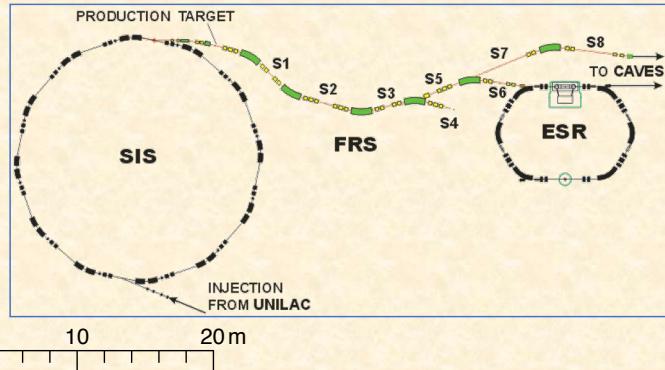
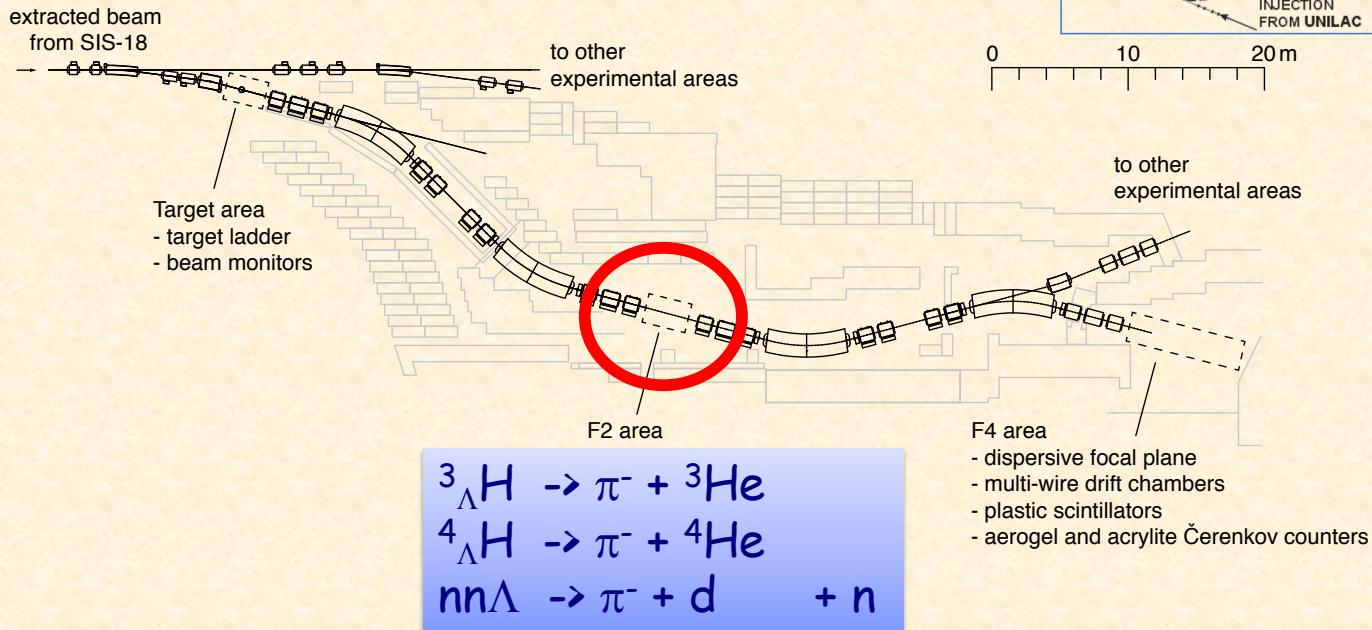
- Binding energy of hypernuclei
  - Single-strangeness hypernuclei
    - ✓ Stopped two-body decays
    - ✓ Multi-body decays
  - Double-strangeness hypernuclei

Machine learning

# The novel technique with FRS at GSI

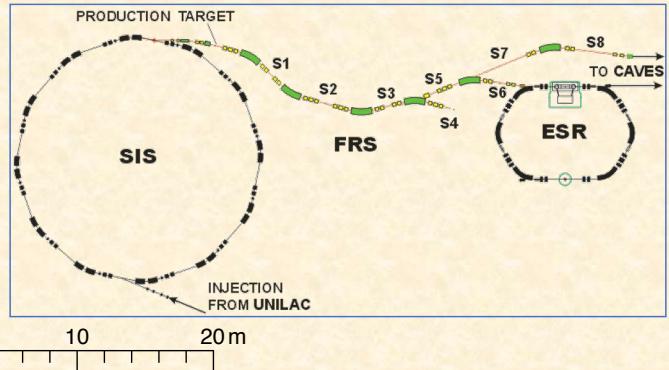
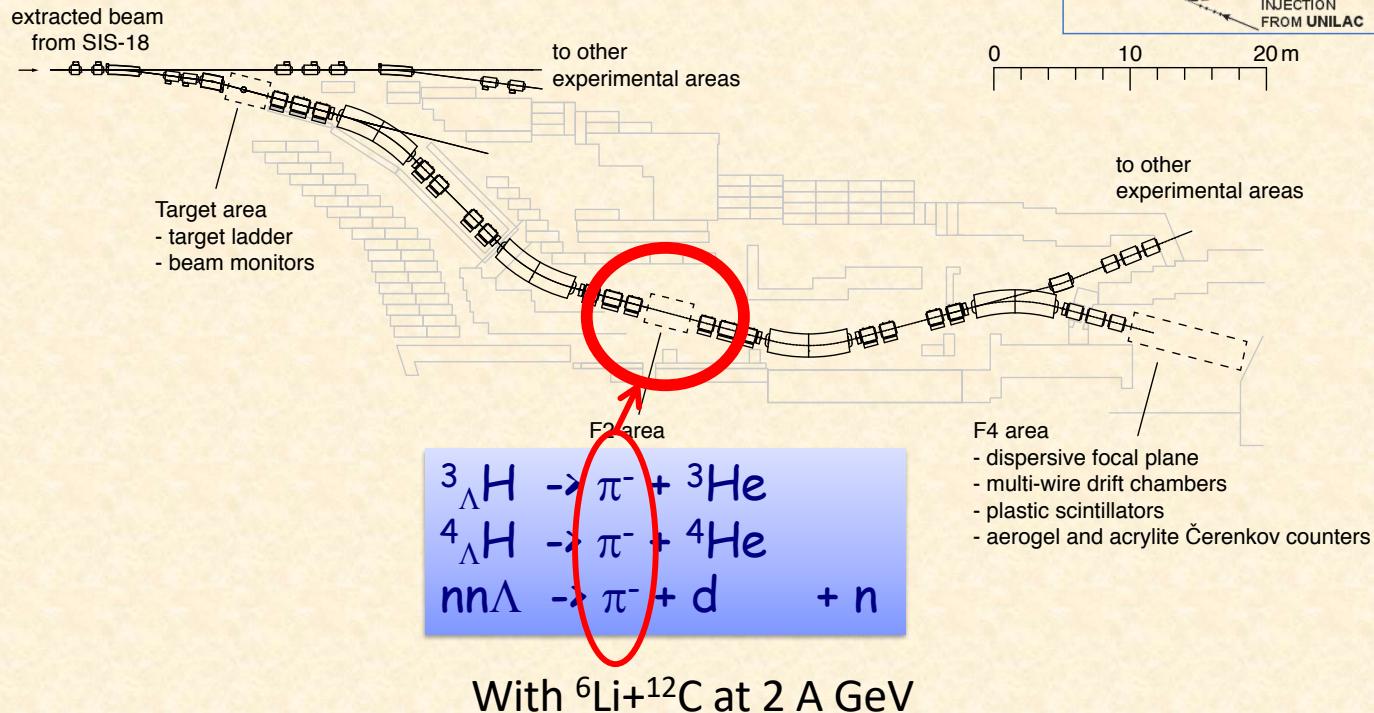


# The novel technique with FRS at GSI

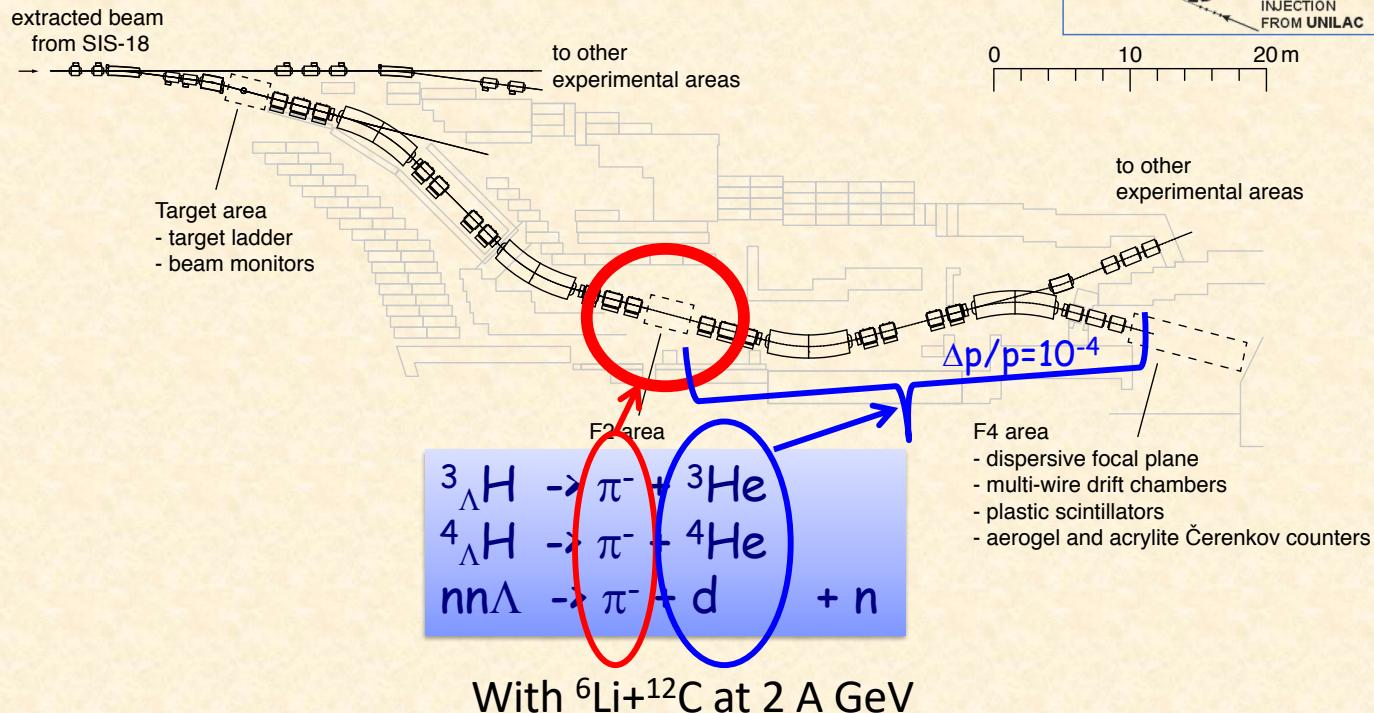


With  ${}^6 Li + {}^{12} C$  at 2 A GeV

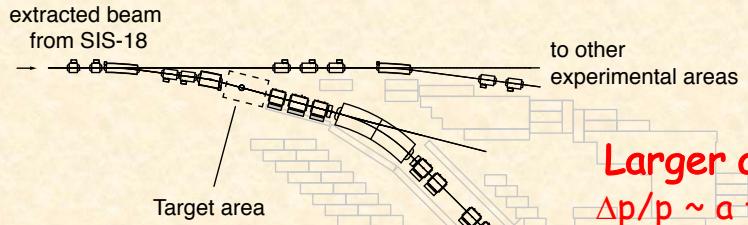
# The novel technique with FRS at GSI



# The novel technique with FRS at GSI



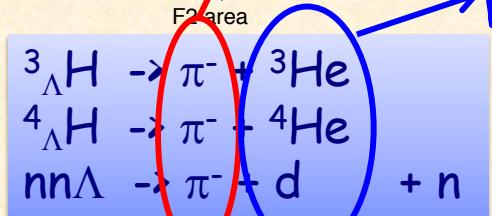
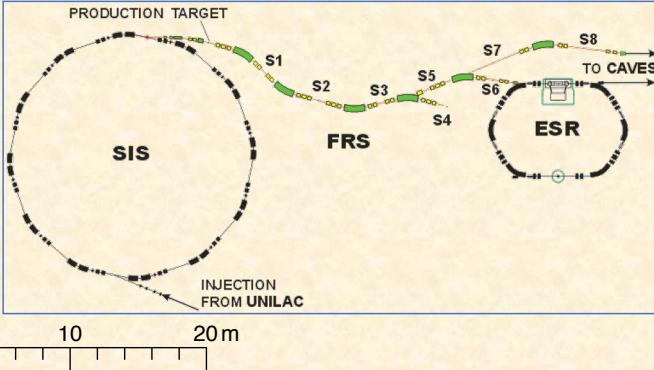
# The novel technique with FRS at GSI



Target area  
- target ladder  
- beam monitors

Larger acceptance for  $\pi^-$   
 $\Delta p/p \sim \text{a few \%}$

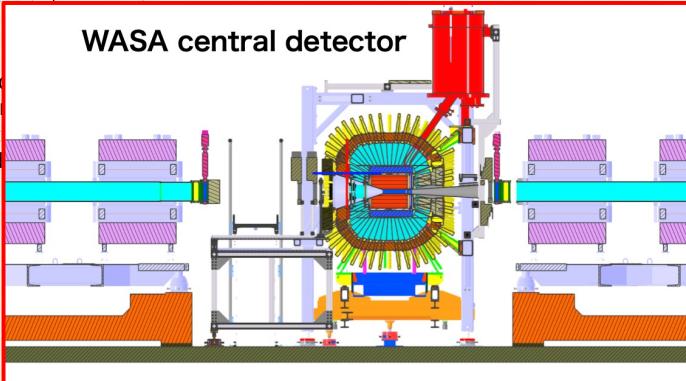
0 10 20 m



With  $^{6}\text{Li}+^{12}\text{C}$  at 2 A GeV

F4 area  
- dispersive fo  
- multi-wire d  
- plastic scint  
- aerogel and

WASA central detector



# The WASA-FRS setup



Talk by Christophe Rappold  
just before this presentation

# The WASA-FRS collaboration

T.R. Saito<sup>a,b,c,l</sup>, P. Achenbach<sup>d,e</sup>, H. Alibrahim Alfaki<sup>b</sup>, F. Amjad<sup>b</sup>, M. Armstrong<sup>b,f</sup>, K.-H. Behr<sup>b</sup>, J. Benlliure<sup>g</sup>, Z. Brencic<sup>h,i</sup>, T. Dicke<sup>b,j</sup>, V. Drozd<sup>b,k</sup>, S. Dubey<sup>b</sup>, H. Ekawa<sup>a</sup>, S. Escrig<sup>i,a</sup>, M. Feijoo-Fontán<sup>g</sup>, H. Fujioka<sup>m</sup>, Y. Gao<sup>a,n,o</sup>, H. Geissel<sup>b,j</sup>, F. Goldenbaum<sup>p</sup>, A. Graña González<sup>s</sup>, E. Haettner<sup>b</sup>, M.N. Harakeh<sup>k</sup>, Y. He<sup>a,c</sup>, H. Heggen<sup>b</sup>, C. Hornung<sup>b</sup>, N. Hubbard<sup>b,q</sup>, K. Itahashi<sup>t,s,2</sup>, M. Iwasaki<sup>t,s</sup>, N. Kalantar-Nayestanaki<sup>k</sup>, A. Kasagi<sup>a,t</sup>, M. Kavatsyuk<sup>k</sup>, E. Kazantseva<sup>b</sup>, A. Khreptak<sup>u,v</sup>, B. Kindler<sup>b</sup>, R. Knoebel<sup>b</sup>, H. Kollmus<sup>b</sup>, D. Kostyleva<sup>b</sup>, S. Kraft-Bermuth<sup>w</sup>, N. Kurz<sup>b</sup>, E. Liu<sup>a,n,o</sup>, B. Lommel<sup>b</sup>, V. Metag<sup>i</sup>, S. Minami<sup>b</sup>, D.J. Morrissey<sup>x</sup>, P. Moskal<sup>v,y</sup>, I. Mukha<sup>b</sup>, A. Muneeam<sup>a,z</sup>, M. Nakagawa<sup>a</sup>, K. Nakazawa<sup>t</sup>, C. Nociforo<sup>b</sup>, H.J. Ong<sup>n,aa,ab</sup>, S. Pietri<sup>b</sup>, J. Pochodzalla<sup>d,e</sup>, S. Purushothaman<sup>b</sup>, C. Rappold<sup>i</sup>, E. Rocco<sup>b</sup>, J.L. Rodríguez-Sánchez<sup>g</sup>, P. Roy<sup>b</sup>, R. Ruber<sup>a,c</sup>, S. Schadmand<sup>b</sup>, C. Scheidenberger<sup>b,j</sup>, P. Schwarz<sup>b</sup>, R. Sekiya<sup>ad,r,s</sup>, V. Serdyuk<sup>p</sup>, M. Skurzok<sup>v,y</sup>, B. Streicher<sup>b</sup>, K. Suzuki<sup>b,ac</sup>, B. Szczepanczyk<sup>b</sup>, Y.K. Tanaka<sup>a,3</sup>, X. Tang<sup>n</sup>, N. Tortorelli<sup>b</sup>, M. Venczel<sup>j,h</sup>, H. Wang<sup>a</sup>, T. Weber<sup>b</sup>, H. Weick<sup>b</sup>, M. Will<sup>b</sup>, K. Wimmer<sup>b</sup>, A. Yamamoto<sup>af</sup>, A. Yanai<sup>ag,a</sup>, J. Yoshida<sup>a,ah</sup>, J. Zhao<sup>b,ai</sup>, (WASA-FRS/Super-FRS Experiment Collaboration)

<sup>a</sup>High Energy Nuclear Physics Laboratory, RIKEN Cluster for Pioneering Research, RIKEN, 351-0198 Wako, Saitama, Japan,

<sup>b</sup>GSI Helmholtzzentrum für Schwerionenforschung GmbH, 64291 Darmstadt, Germany,

<sup>c</sup>School of Nuclear Science and Technology, Lanzhou University, 730000 Lanzhou, China,

<sup>d</sup>Institute for Nuclear Physics, Johannes Gutenberg University, 55099 Mainz, Germany,

<sup>e</sup>Helmholtz Institute Mainz, Johannes Gutenberg University, 55099 Mainz, Germany,

<sup>f</sup>Institut für Kernphysik, Universität Köln, 50923 Köln, Germany,

<sup>g</sup>Universidade de Santiago de Compostela, 15782 Santiago de Compostela, Spain,

<sup>h</sup>Jozef Stefan Institute, 1000 Ljubljana, Slovenia,

<sup>i</sup>University of Ljubljana, 1000 Ljubljana, Slovenia,

<sup>j</sup>Universität Gießen, 35392 Gießen, Germany,

<sup>k</sup>University of Groningen, 9747 AA Groningen, The Netherlands,

<sup>l</sup>Instituto de Estructura de la Materia - CSIC, 28006 Madrid, Spain,

<sup>m</sup>Tokyo Institute of Technology, 152-8550 Tokyo, Japan,

<sup>n</sup>Institute of Modern Physics, Chinese Academy of Sciences, 730000 Lanzhou, China,

<sup>o</sup>School of Nuclear Science and Technology, University of Chinese Academy of Sciences, 100049 Beijing, China,

<sup>p</sup>Institut für Kernphysik, Forschungszentrum Jülich, 52425 Jülich, Germany,

<sup>q</sup>Institut für Kernphysik, Technische Universität Darmstadt, 64289 Darmstadt, Germany,

<sup>r</sup>Meson Science Laboratory, Cluster for Pioneering Research, RIKEN, 2-1 Hirosawa, 351-0198 Wako, Saitama, Japan,

<sup>s</sup>Nishina Center for Accelerator-Based Science, RIKEN, 2-1 Hirosawa, 351-0198 Wako, Saitama, Japan,

<sup>t</sup>Graduate School of Engineering, Gifu University, 501-1193 Gifu, Japan ,

<sup>u</sup>INFN, Laboratori Nazionali di Frascati, Frascati, 00044 Roma, Italy,

<sup>v</sup>Institute of Physics, Jagiellonian University, 30-348 Kraków, Poland,

<sup>w</sup>TH Mittelhessen University of Applied Sciences, 35390 Gießen, Germany,

<sup>x</sup>National Superconducting Cyclotron Laboratory, Michigan State University, MI 48824 East Lansing, USA,

<sup>y</sup>Center for Theranostics, Jagiellonian University, 30-348 Krakow, Poland,

<sup>z</sup>Faculty of Engineering Sciences, Ghulam Ishaq Khan Institute of Engineering Sciences and Technology, 23640 Topi, Pakistan,

<sup>aa</sup>Joint Department for Nuclear Physics, Lanzhou University and Institute of Modern Physics, Chinese Academy of Sciences, 730000 Lanzhou, China,

<sup>ab</sup>Research Center for Nuclear Physics, Osaka University, 567-0047 Osaka, Japan,

<sup>ac</sup>Uppsala University, 75220 Uppsala, Sweden,

<sup>ad</sup>Kyoto University, 606-8502 Kyoto, Japan,

<sup>ae</sup>Ruhr-Universität Bochum, Institut für Experimentalphysik I, 44780 Bochum, Germany,

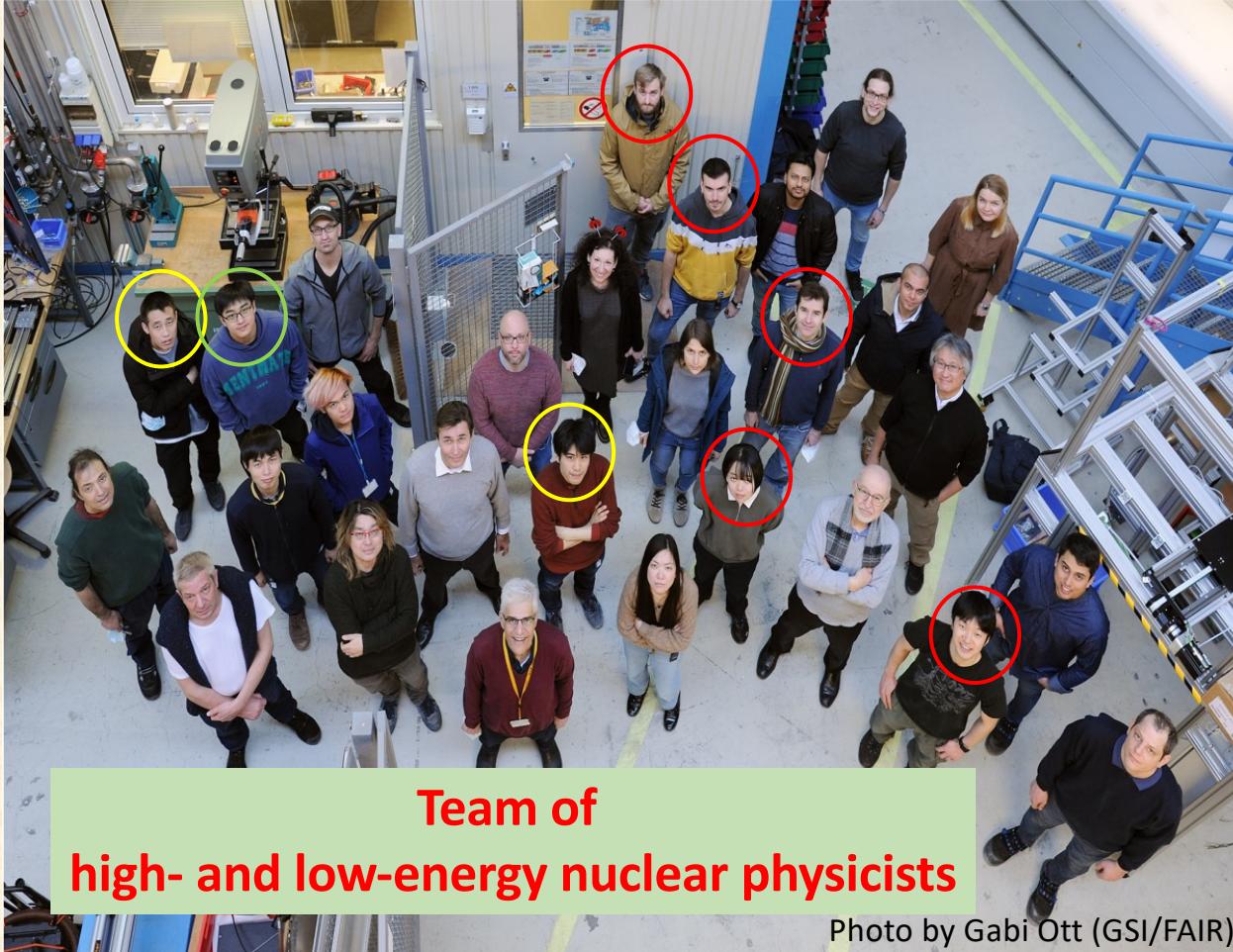
<sup>af</sup>KEK, 305-0801 Tsukuba, Ibaraki, Japan,

<sup>ag</sup>Saitama University, Sakura-ku, 338-8570 Saitama, Japan,

<sup>ah</sup>Tohoku University, 980-8578 Sendai, Japan,

<sup>ai</sup>Peking University, 100871 Beijing, China,

# Part of the WASA-FRS collaboration



# Data taking (January – March 2022)

Run	Period	Data size
Commissioning run	28th Jan. - 7th Feb.	7 TB
Physics run for $\eta'$ nuclei	22nd Feb. - 28th Feb.	40 TB
Physics run for HypHI	10th Mar. - 19th Mar.	48 TB

92 % of the prop.

Acquired data for S447 (hypernuclei)

Beam	Fragment at S4	Amount	Time	Accepted trigger rate
<b><math>^6\text{Li}</math> beam</b>	$^3\text{He}$	$3.3 \times 10^8$	40.9 hours	2600 Hz
	$^4\text{He}$	$0.9 \times 10^8$	43.9 hours	1800 Hz
	deuteron	$1.8 \times 10^8$		
	proton (mid-rapidity)	$5.3 \times 10^6$	3.2 hours	680 Hz
<b><math>^{12}\text{C}</math> beam</b>	$^3\text{He}$	$1.0 \times 10^8$	13.5 hours	2400 Hz
	$^9\text{C}$	$2.4 \times 10^5$		

$^3\Lambda\text{H}$

$^4\Lambda\text{H}$

$\text{nn}\Lambda$

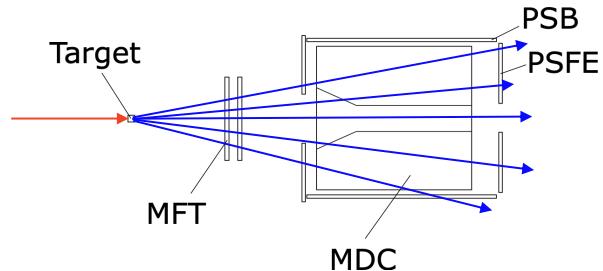
$\Lambda$

$^3\Lambda\text{B}$

$^9\Lambda\text{B}$

# Graph Neural Network (GNN) for WASA

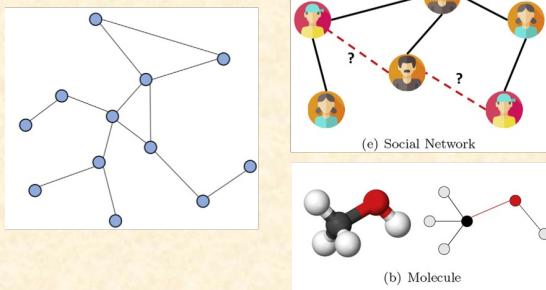
## Track Finding



- Multi particles in HI reaction
- Combinatorial background

→ Track Finding with  
Graph Neural Network  
(GNN)

## Graph



- Node : Data point
- Edge : Connection

Eur. Phys. J. A (2023) 59:103  
<https://doi.org/10.1140/epja/s10050-023-01016-5>

THE EUROPEAN  
PHYSICAL JOURNAL A



Special Article - New Tools and Techniques

### Development of machine learning analyses with graph neural network for the WASA-FRS experiment

H. Ekawa<sup>1,2</sup>, W. Dou<sup>1,2</sup>, Y. Gao<sup>1,3,4</sup>, Y. He<sup>1,5</sup>, A. Kasagi<sup>1,6</sup>, E. Liu<sup>1,3,4</sup>, A. Munem<sup>1,7</sup>, M. Nakagawa<sup>1</sup>, C. Rappold<sup>1</sup>, N. Saito<sup>1</sup>, T. R. Saito<sup>1,9,5</sup>, M. Taki<sup>10</sup>, Y. K. Tanaka<sup>1,6</sup>, H. Wang<sup>1</sup>, J. Yoshida<sup>1,11</sup>

<sup>1</sup> High Energy Nuclear Physics Laboratory, Cluster for Pioneering Research, RIKEN, Wako, Japan

<sup>2</sup> Department of Physics, Saitama University, Saitama, Japan

<sup>3</sup> Institute of Modern Physics, Chinese Academy of Sciences, Lanzhou, China

<sup>4</sup> University of Chinese Academy of Sciences, Beijing, China

<sup>5</sup> School of Nuclear Science and Technology, Lanzhou University, Lanzhou, China

<sup>6</sup> Graduate School of Engineering, Gifu University, Gifu, Japan

<sup>7</sup> Faculty of Engineering Sciences, Ghulam Ishaq Khan Institute of Engineering Sciences and Technology, Topi, Pakistan

<sup>8</sup> Instituto de Física de la Materia Condensada y de Investigaciones Científicas (CSIC), Madrid, Spain

<sup>9</sup> GSI Helmholtz Center for Heavy Ion Research, Darmstadt, Germany

<sup>10</sup> Graduate School of Artificial Intelligence and Science, Rikkyo University, Tokyo, Japan

<sup>11</sup> Department of Physics, Tohoku University, Sendai, Japan

Received: 29 July 2022 / Accepted: 24 April 2023

© The Author(s), under exclusive licence to Società Italiana di Fisica and Springer-Verlag GmbH Germany, part of Springer Nature 2023

Communicated by Takashi Nakamura

**Abstract** The WASA-FRS experiment aims to reveal the nature of light A hypernuclei with heavy-ion beams. The lifetimes of hypernuclei are measured precisely from their decay lengths and kinematics. To reconstruct a  $\pi^-$  track emitted from hypernuclear decay, track finding is an important issue. In this study, a machine learning analysis method with a graph neural network (GNN), which is a powerful tool for deducing the connection between data nodes, was developed to obtain track associations from numerous combinations of hit information provided in detectors based on a Monte Carlo simulation. An efficiency of 98% was achieved for tracking  $\pi^-$  mesons using the developed GNN model. The GNN model can also estimate the charge and momentum of the particles of interest. More than 99.9% of the negative charged particles were correctly identified with a momentum accuracy of 6.3%.

stand it for the middle- and long-range interactions based on a variety of nuclear experiments. To reveal the unknown features of the nuclear force, such as short-range interaction, considering a more detailed structure inside the baryons is essential. All baryons consist of three quarks, and nucleons such as neutrons and protons consist of up and down quarks. By introducing other types of quarks into ordinary nuclear systems, one can study the nuclear force in a more general picture. In particular, because the mass of the strange quark is close to that of the up and down quarks, interactions among these three quarks are described under flavoured SU(3) symmetry. Therefore, a hyperon, which is a type of baryon that contains strange quark(s), plays an important role in investigating baryon-baryon interactions. As the lifetime of hyperon is short ( $\sim 10^{-10}$  s), using them as projectiles or targets is difficult. Therefore, hyperon-nucleon interactions have been studied via hypernuclei, which contain at least

Published in EPJA (May 2023)

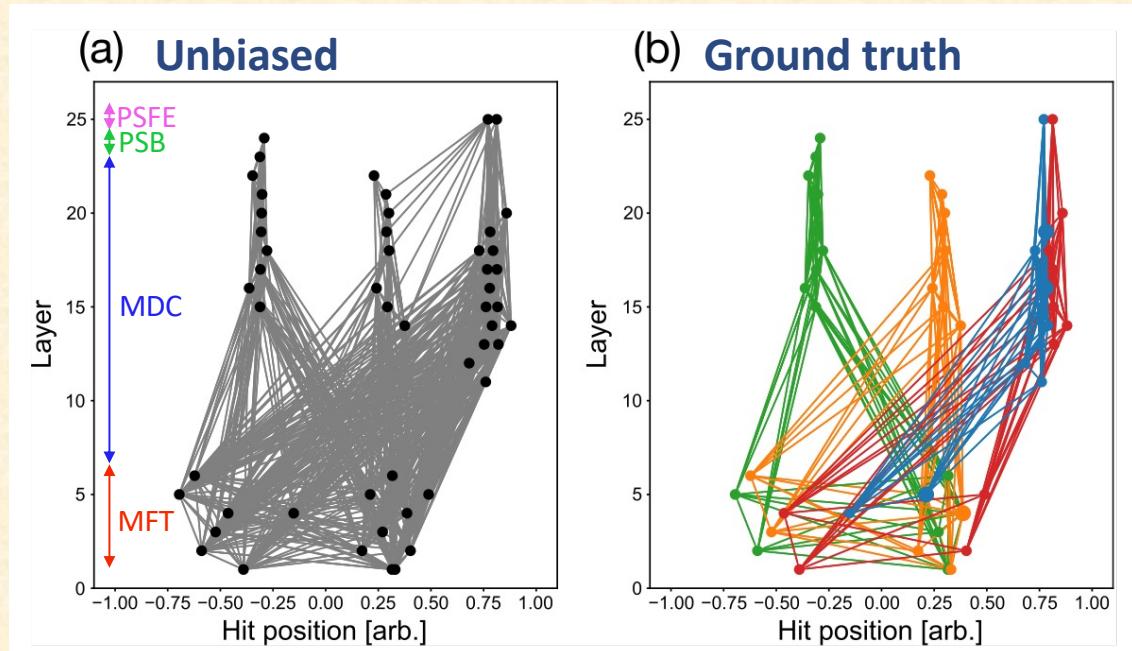
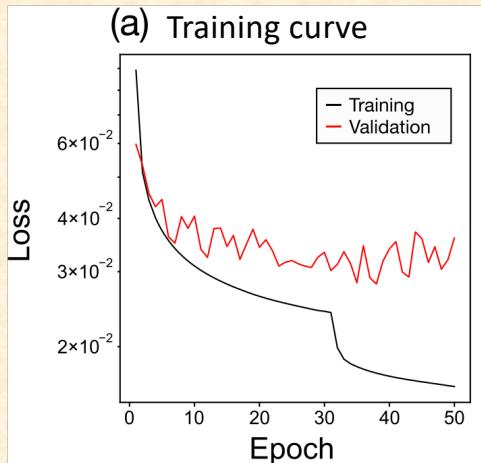
H. Ekawa et al., Eur. Phys. J. A (2023) 59, 103

DOI : [10.1140/epja/s10050-023-01016-5](https://doi.org/10.1140/epja/s10050-023-01016-5)

# GNN: Node clustering

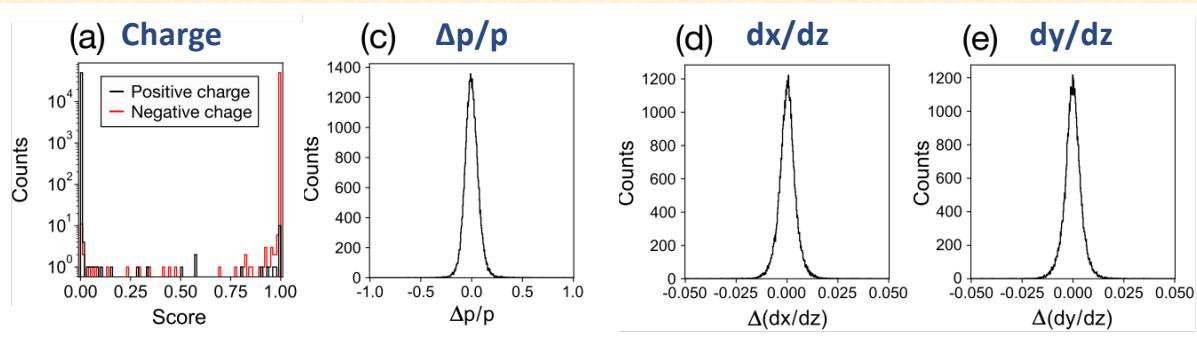
## Machine Learning

- PyTorch + PyTorch Geometric
- Monte Carlo (MC) simulation
  - Training, Validation, Test
- Learning object
  - Edge : ON / OFF
  - Node : Shared / Not-Shared



# GNN: Performance with MC

Data size	Clustering efficiency for $\pi^-$	Clustering efficiency for others	Training time [h/epoch]
100k	96.3%	95.1%	0.6
300k	97.4%	96.2%	2.0
<b>1M</b>	<b>98.1%</b>	<b>97.1%</b>	<b>7.5</b>



Data size	Eff.	False-positive	$\frac{\Delta p}{p}$	$\Delta \frac{dx}{dz}$ [mrad]	$\Delta \frac{dy}{dz}$ [mrad]	Training time[h/epoch]
500k	99.70%	0.07%	9.9%	6.1	6.0	0.2
1M	99.82%	0.04%	8.5%	5.3	5.3	0.4
2M	99.92%	0.03%	6.8%	4.1	4.1	0.8
<b>4M</b>	<b>99.95%</b>	<b>0.04%</b>	<b>6.3%</b>	<b>3.7</b>	<b>3.7</b>	<b>1.5</b>

cf. Kalman filter

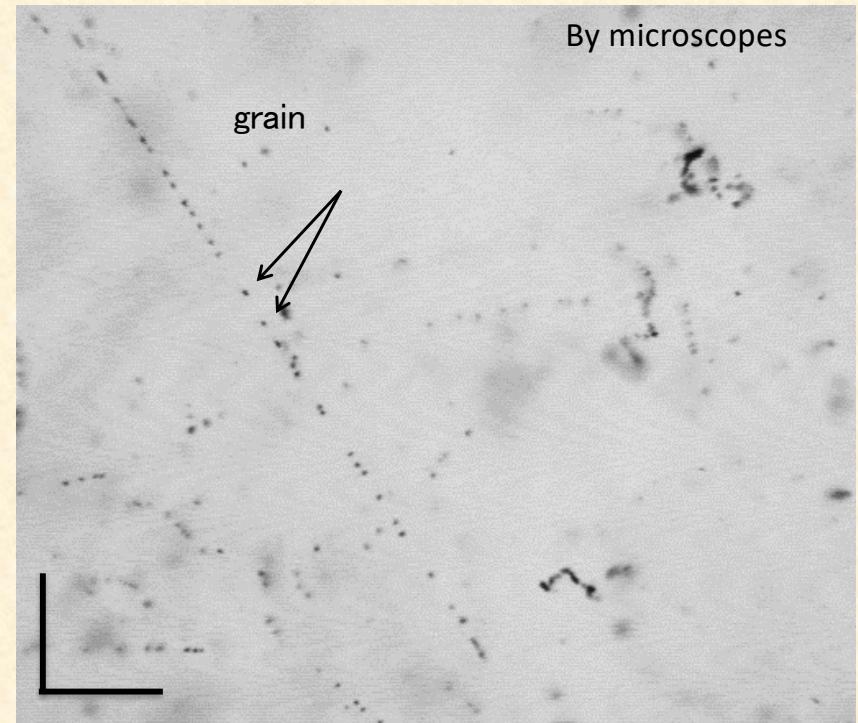
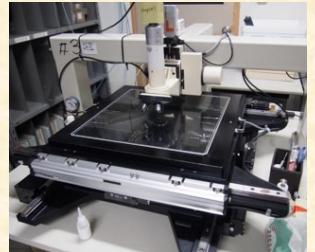
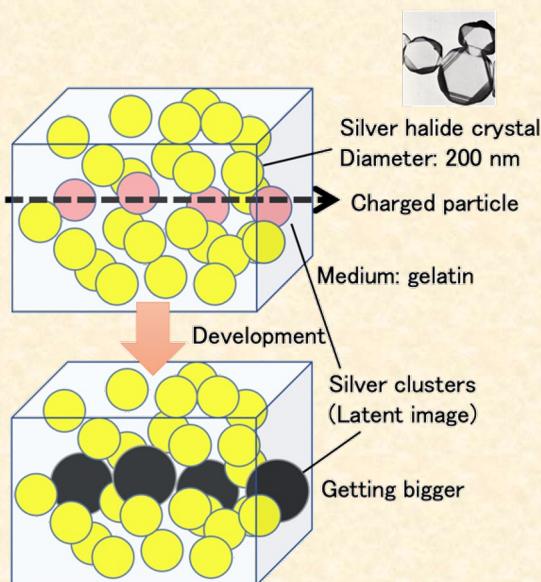
$\Delta p/p : 9.9\%$

$\Delta(dx/dz) : 3.0\text{ mrad}$   
 $\Delta(dy/dz)$

How about  
the hypernuclear binding energy?

# Nuclear Emulsion:

Charged particle tracker with  
the best spatial resolution  
(easy to be < 1  $\mu\text{m}$ , 11 nm at best)

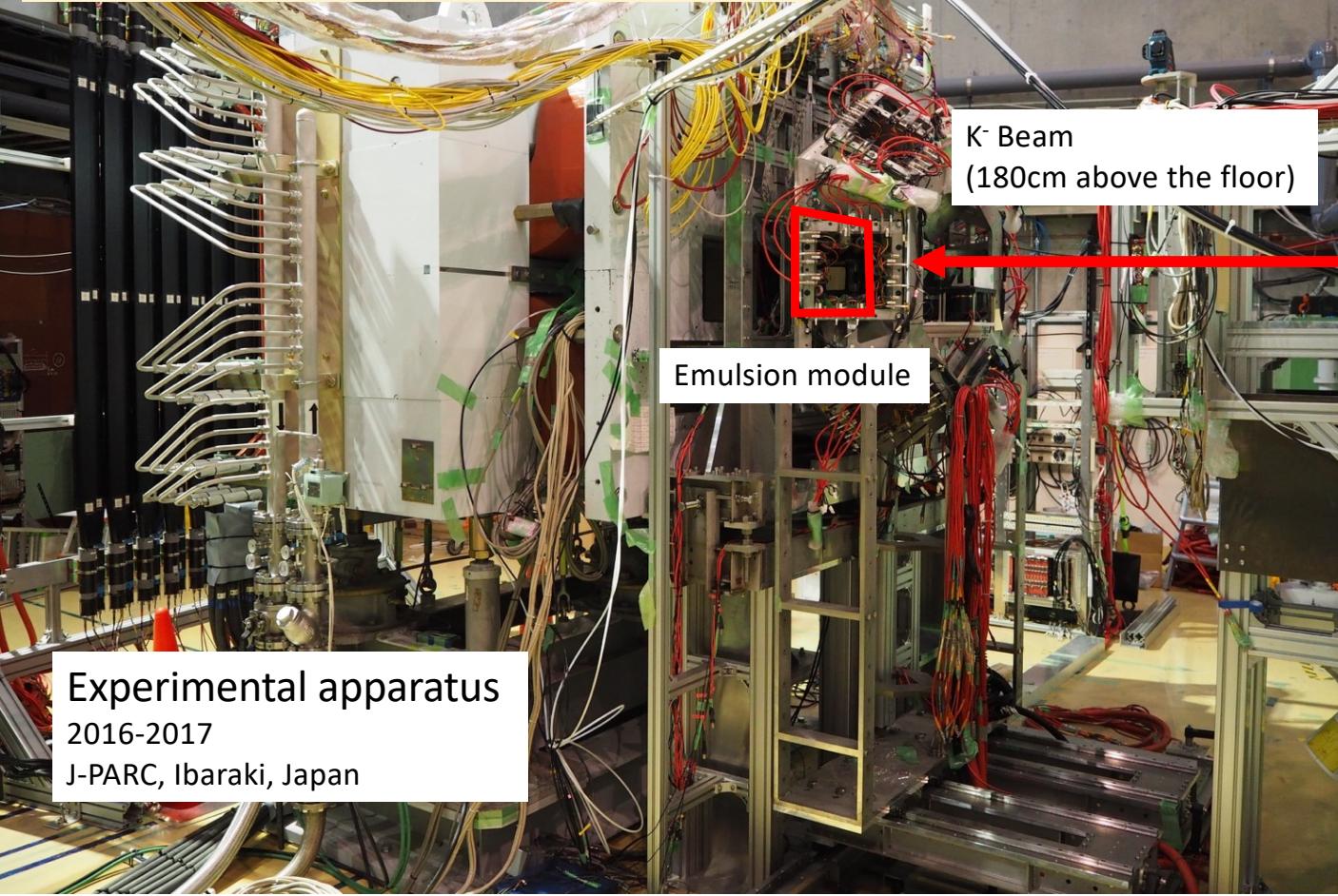


By microscopes

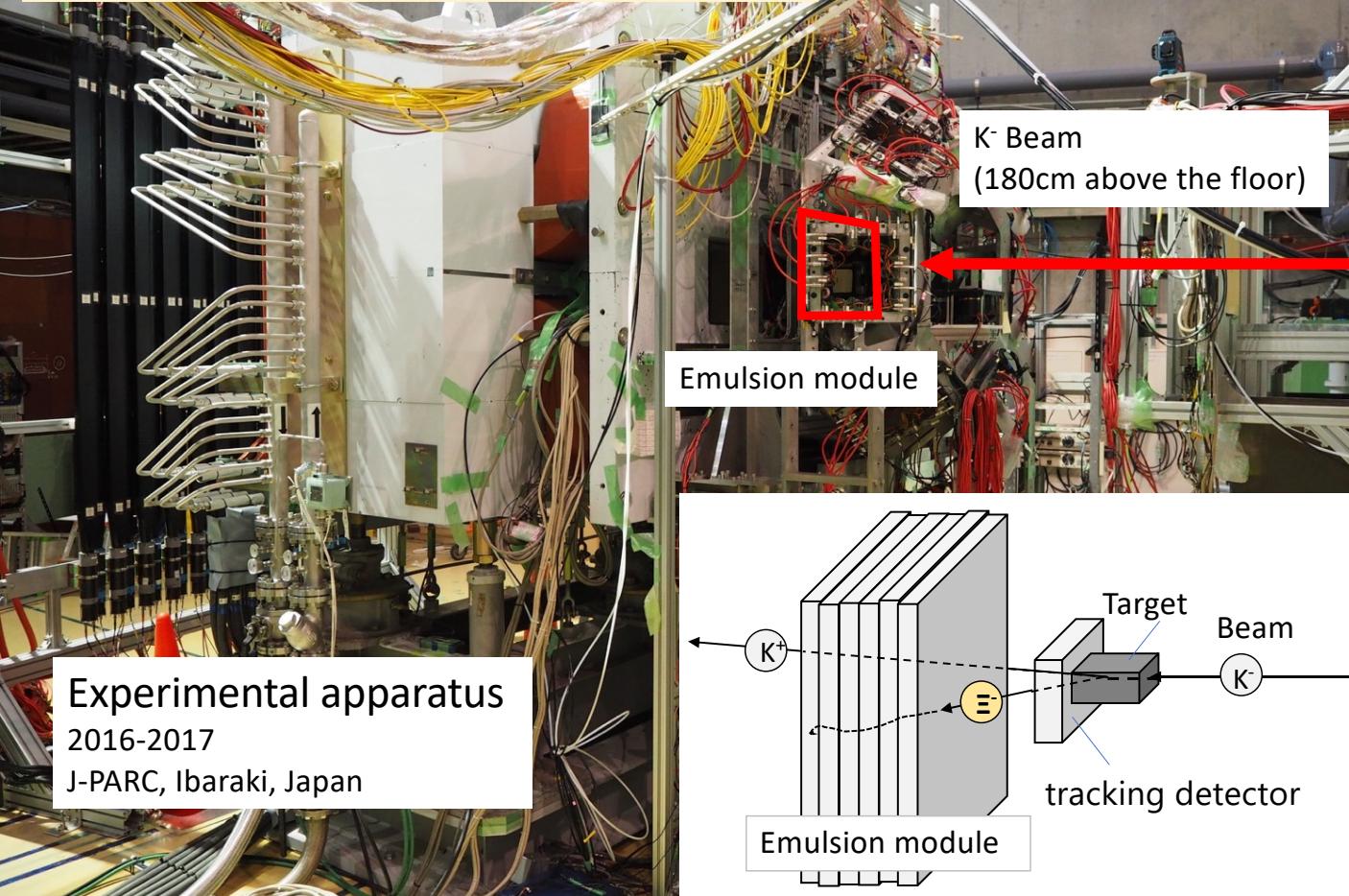
# J-PARC accelerator facility



# J-PARC E07 experiment



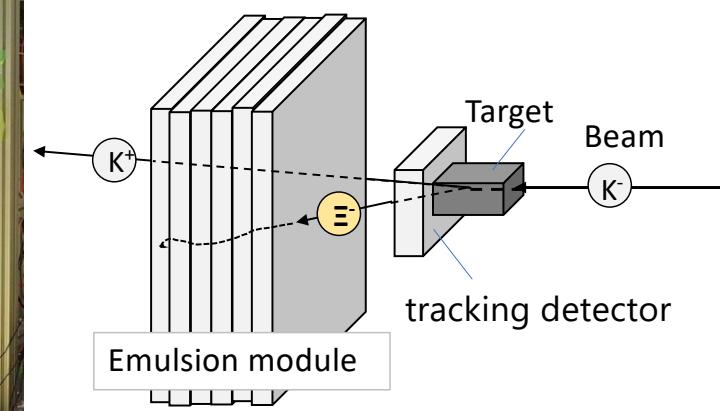
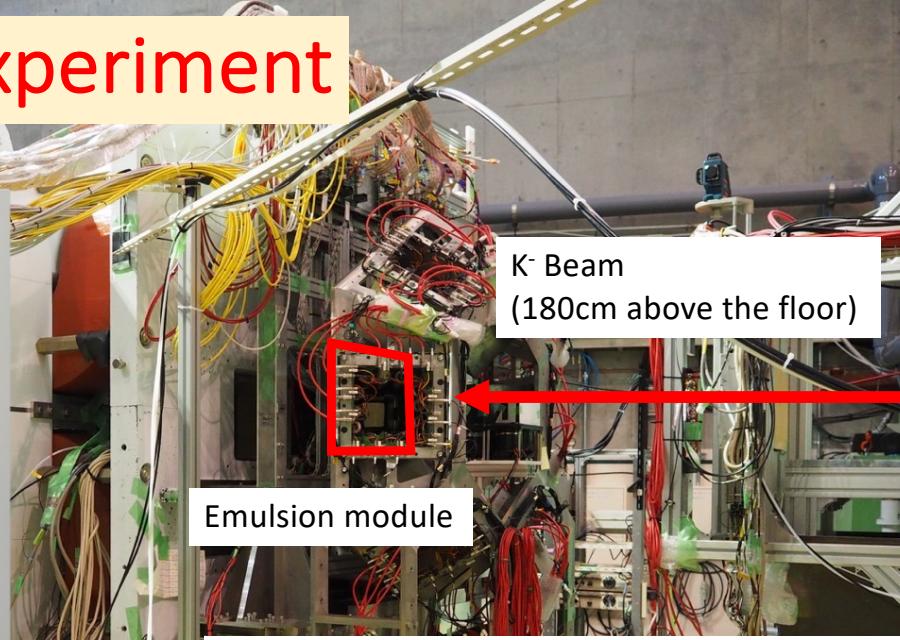
# J-PARC E07 experiment



# J-PARC E07 experiment

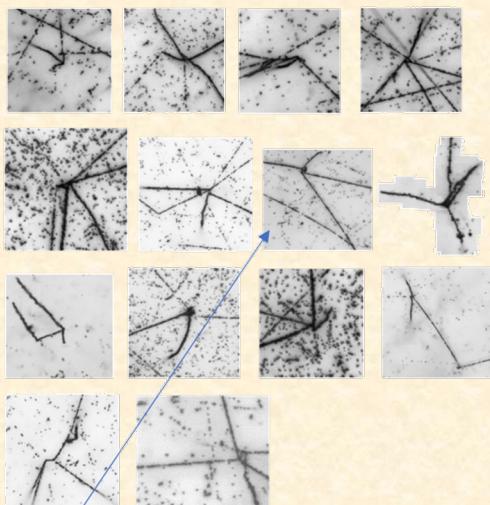


Experimental apparatus  
2016-2017  
J-PARC, Ibaraki, Japan

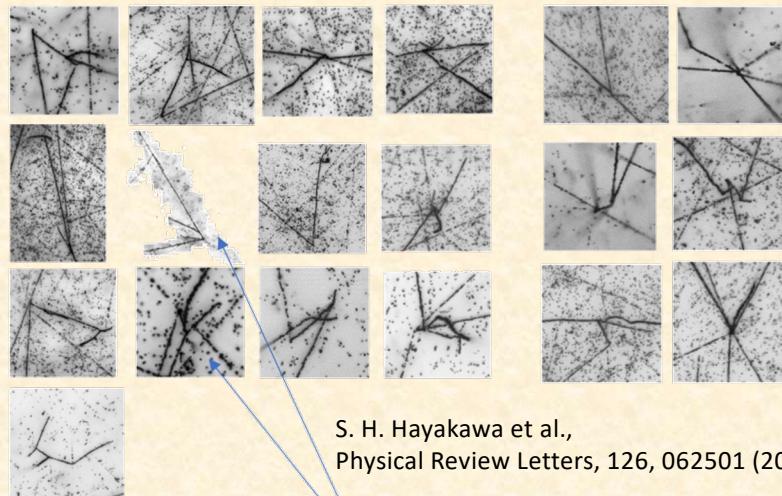


# Results from J-PARC E07 (Hybrid method)

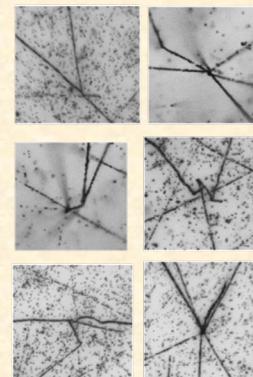
$\Lambda\Lambda$  candidates: 14



Twin  $\Lambda$  events: 13



Others: 6



$\Lambda\Lambda$ Be

H. Ekawa et al., Prog. Theor. Exp. Phys. 2019, 021D02

M. Yoshimoto et al.,

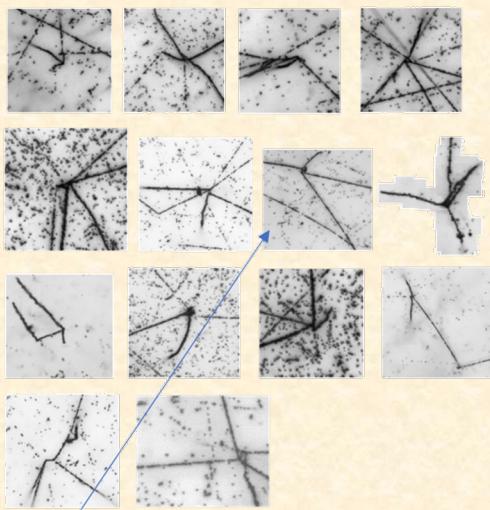
Prog. Theor. Exp. Phys. 2021, 073D02

$^{15}\Xi$ C

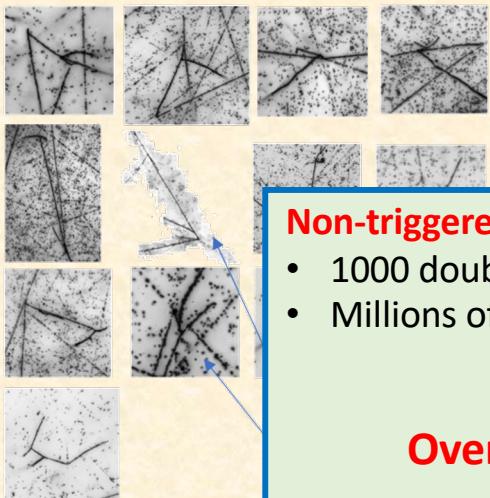
S. H. Hayakawa et al.,  
Physical Review Letters, 126, 062501 (2021)

# Results from J-PARC E07 (Hybrid method)

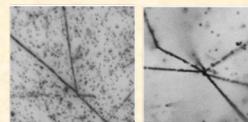
$\Lambda\Lambda$  candidates: 14



Twin  $\Lambda$  events: 13



Others: 6



**Non-triggered events recorded in 1000 emulsions sheets**

- 1000 double-strangeness ( $\Lambda\Lambda$ - and  $\Xi$ -) hypernuclear events
- Millions of single-strangeness hypernuclear events



**Overall scanning of all emulsion sheets  
( $35 \times 35 \text{ cm}^2 \times 1000$ )**

M. Yoshimoto

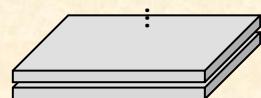
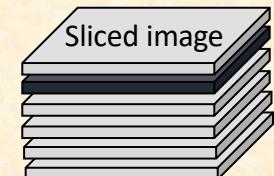
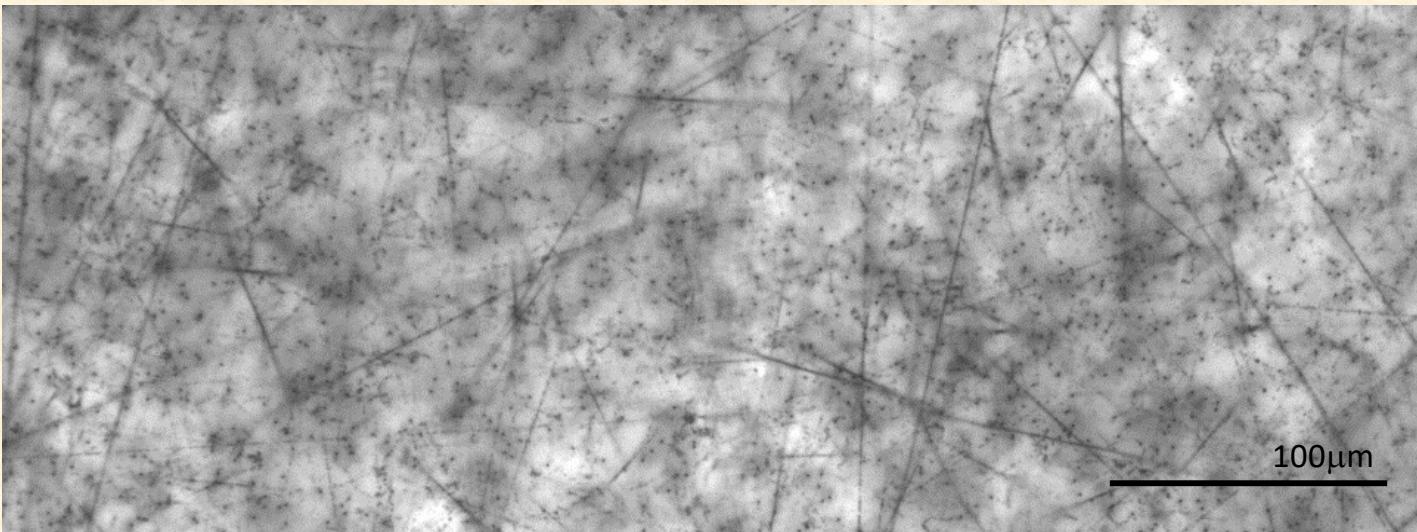
Prog. Theor. Exp. Phys. 2021, 073D02

$\Lambda\Lambda$  Be

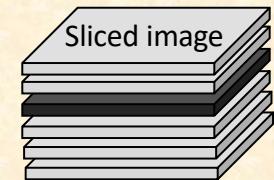
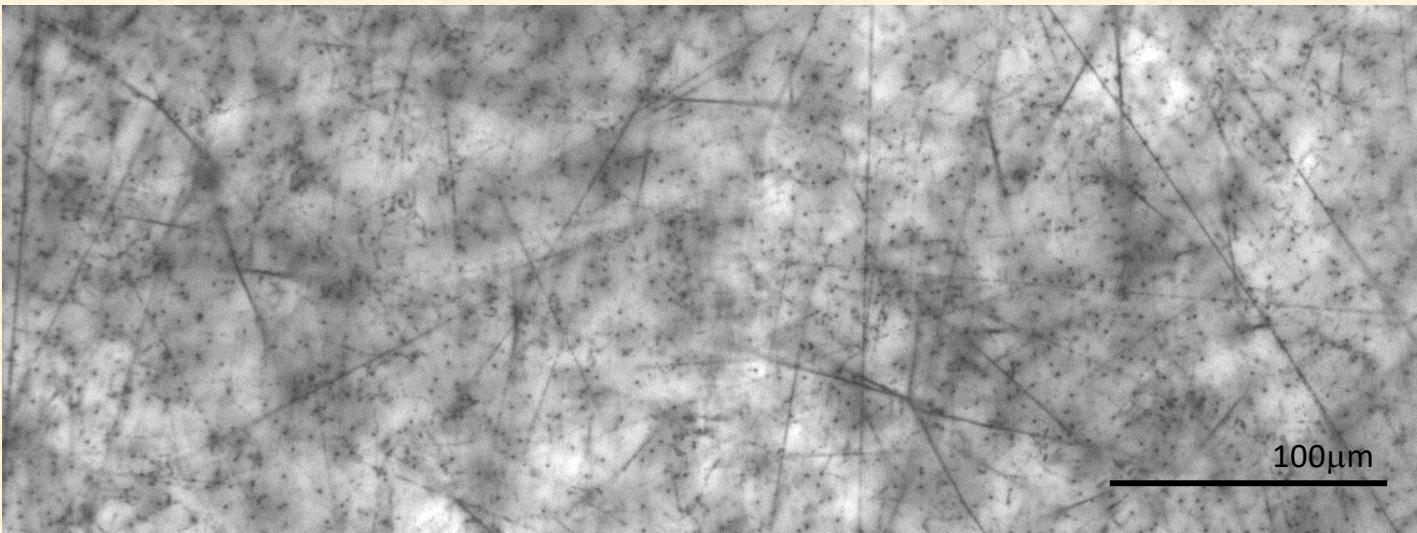
H. Ekawa et al., Prog. Theor. Exp. Phys. 2019, 021D02

$^{15}\Xi$ C

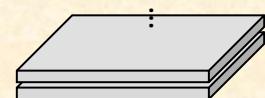
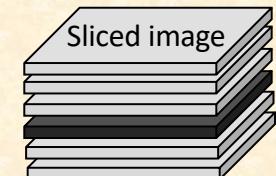
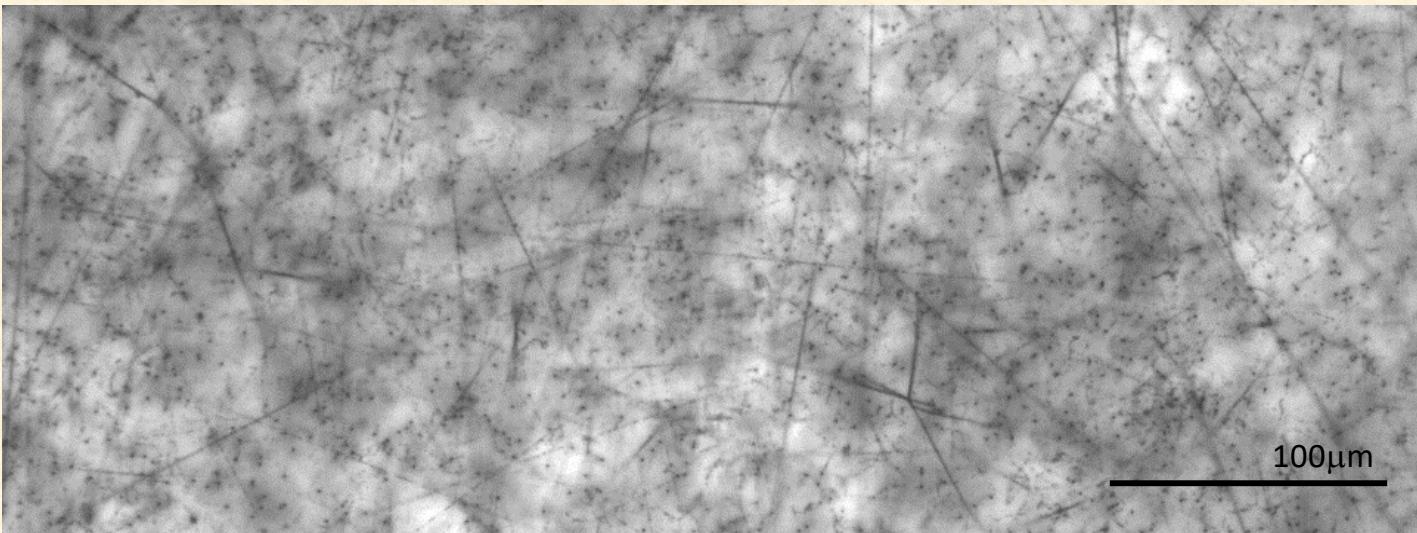
# Overall scanning for E07 emulsions



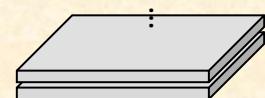
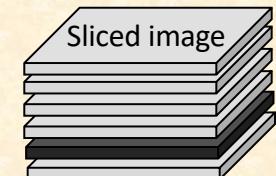
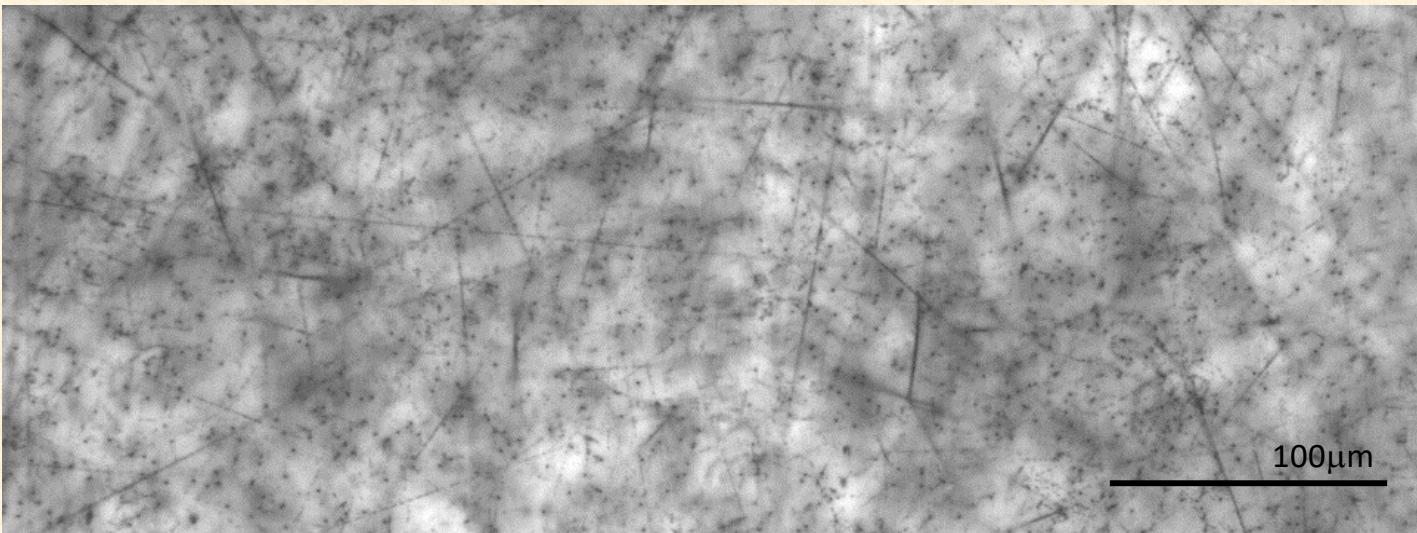
# Overall scanning for E07 emulsions



# Overall scanning for E07 emulsions



# Overall scanning for E07 emulsions



# Overall scanning for E07 emulsions

## Data size:

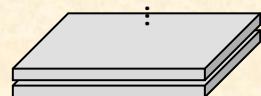
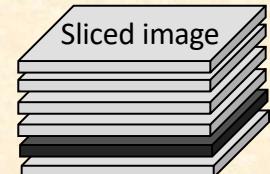
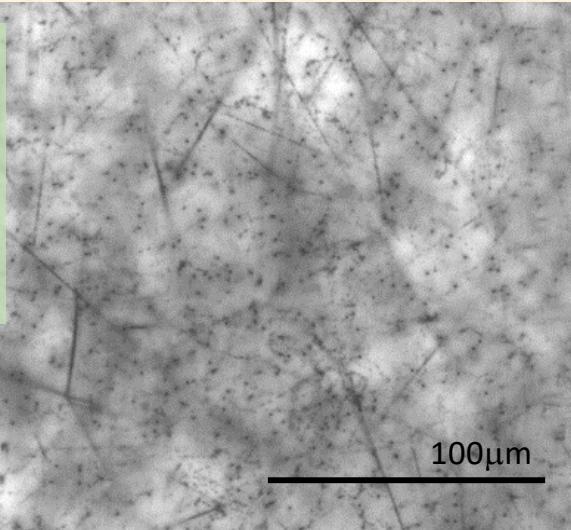
- $10^7$  images per emulsion (100 T Byte)
- $10^{10}$  images per 1000 emulsions (100 P Byte)

## Number of background tracks:

- Beam tracks:  $10^4/\text{mm}^2$
- Nuclear fragmentations:  $10^3/\text{mm}^2$

Current equipments/techniques  
with visual inspections

**560 years**



# Overall scanning for E07 emulsions

## Data size:

- $10^7$  images per emulsion (100 T Byte)
- $10^{10}$  images per 1000 emulsions (100 P Byte)

## Number of background tracks:

- Beam tracks:  $10^4/\text{mm}^2$
- Nuclear fragmentations:  $10^3/\text{mm}^2$

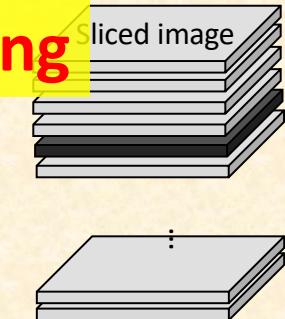
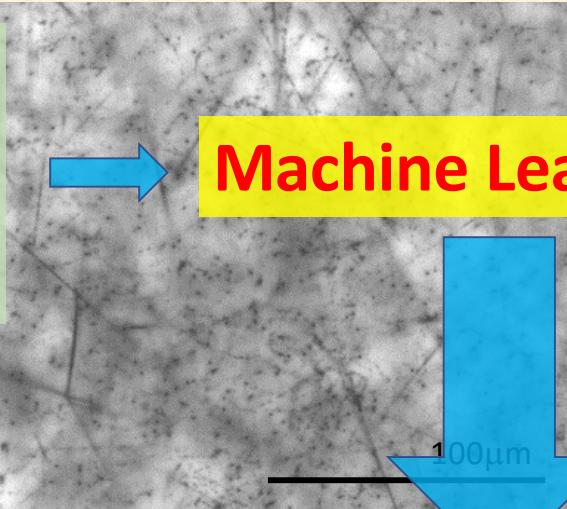
Current equipments/techniques  
with visual inspections

560 years



3 years

Millions of single-strangeness hypernuclei  
1000 double strangeness hypernuclei (formerly only 5)



# Setup for analyzing emulsions at the High Energy Nuclear Physics Laboratory in RIKEN

- Hypernuclear physics
- Neutron imaging



# Challenges for Machine Learning Development

MOST IMPORTANT:

- **Quantity and quality of training data**

However,

**No existing data for hypertriton with emulsions for training**

Our approaches:

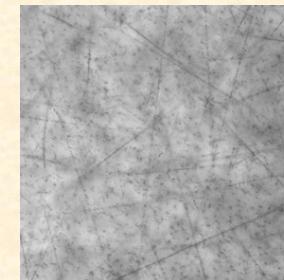
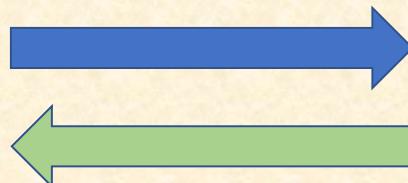
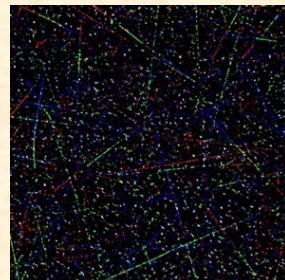
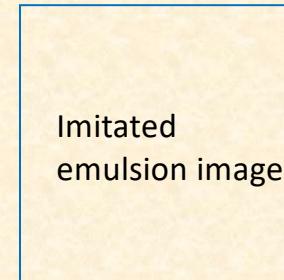
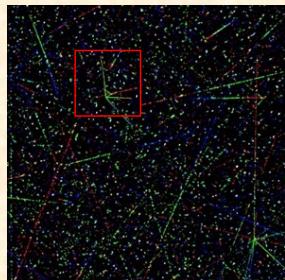
**Producing training data with**

- Monte Carlo simulations
- Image transfer techniques

# Production of training data

Monte Carlo simulations and GAN(Generative Adversarial Networks)

Binarized tracks from MC simulations  
+ background from the real data



Binarized (like for simulations)

Real emulsion image

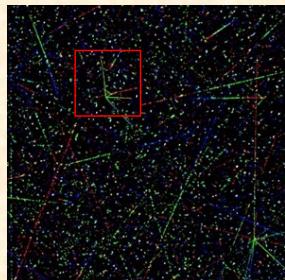
GAN: pix2pix  
Edges to Photo



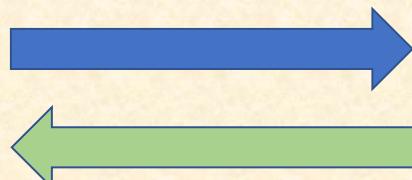
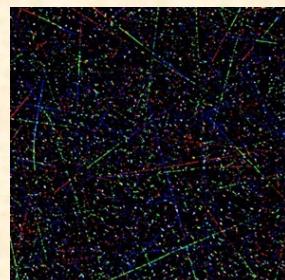
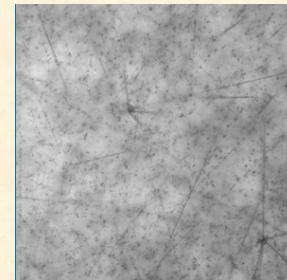
# Production of training data

Monte Carlo simulations and GAN(Generative Adversarial Networks)

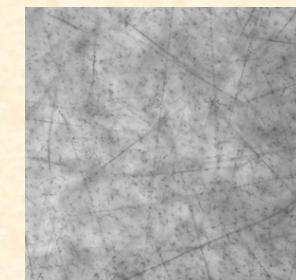
Binarized tracks from MC simulations  
+ background from the real data



Produced training data



Binarized (like for simulations)



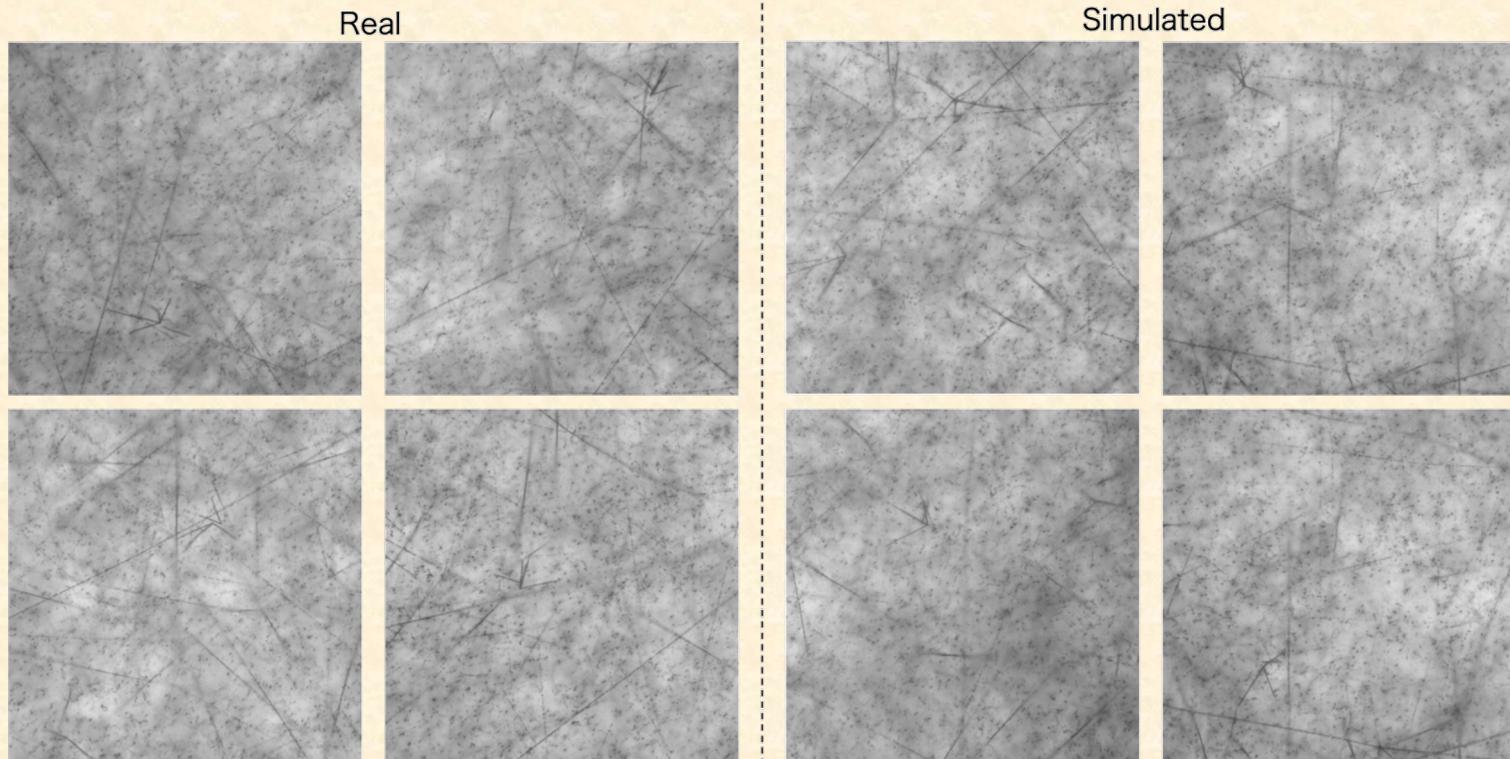
Real emulsion image

GAN: pix2pix  
Edges to Photo



# Production of training data

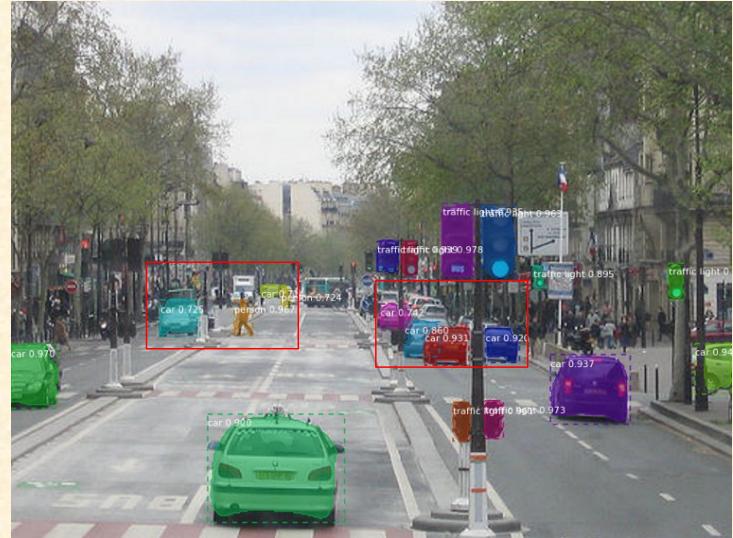
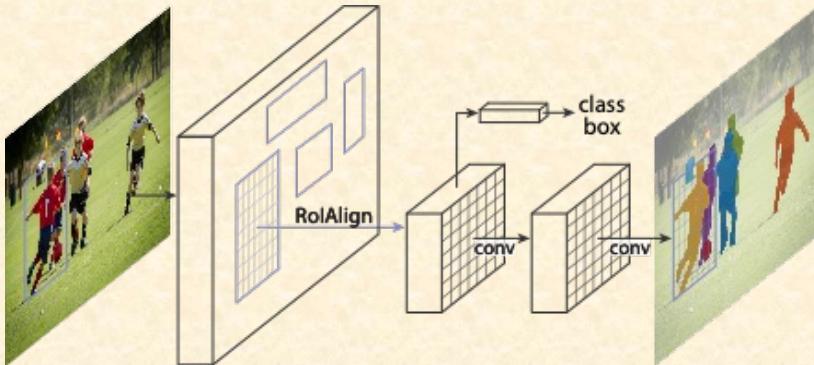
Monte Carlo simulations and GAN(Generative Adversarial Networks)



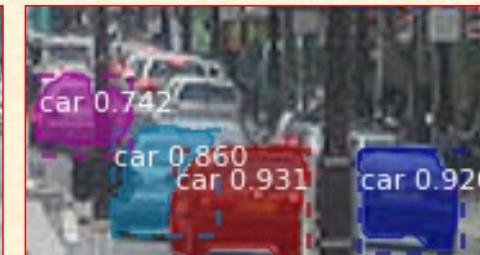
# Detection of hypertriton events

## With Mask R-CNN model

K. He, et al., arXiv <https://arxiv.org/abs/1703.06870> (2017).



Detection of each object

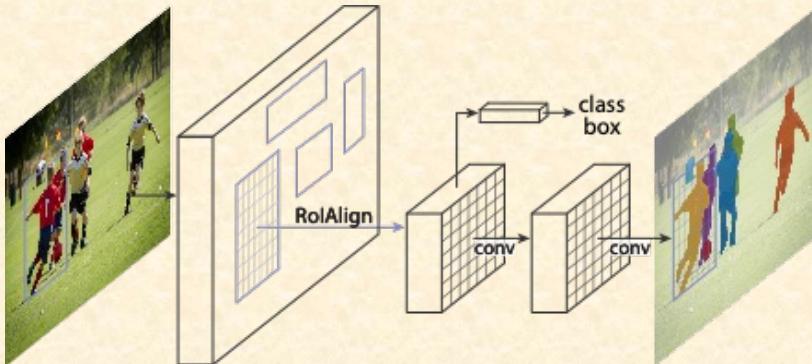


At large object density

# Detection of hypertriton events

## With Mask R-CNN model

K. He, et al., arXiv <https://arxiv.org/abs/1703.06870> (2017).

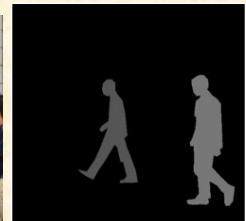


Example of training dataset

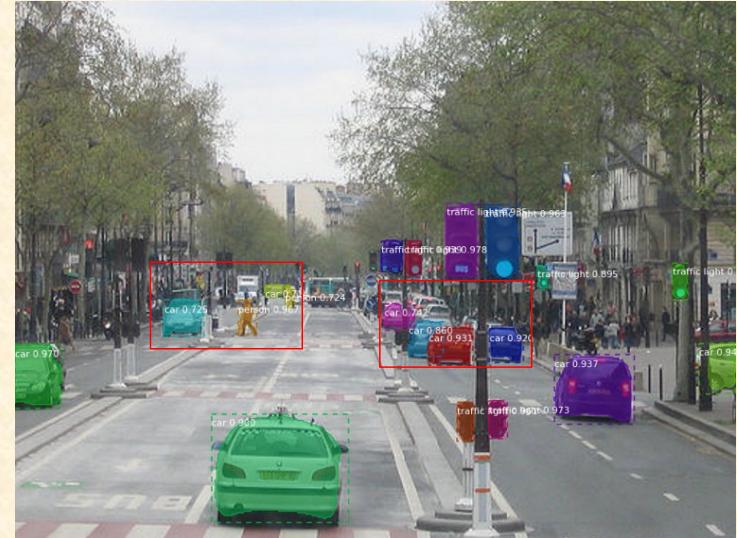
Image



Mask



A Pedestrian dataset



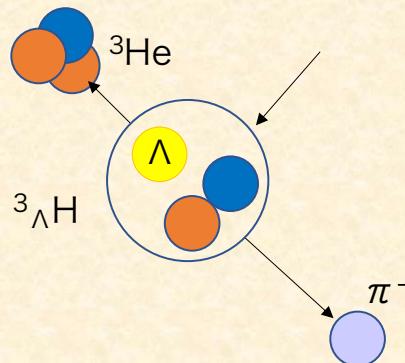
Detection of each object



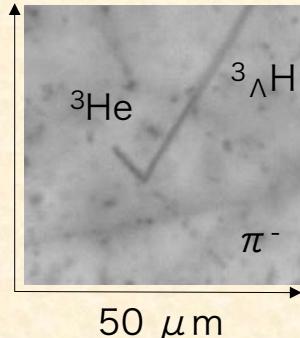
At large object density

# Hypertriton search with Mask R-CNN

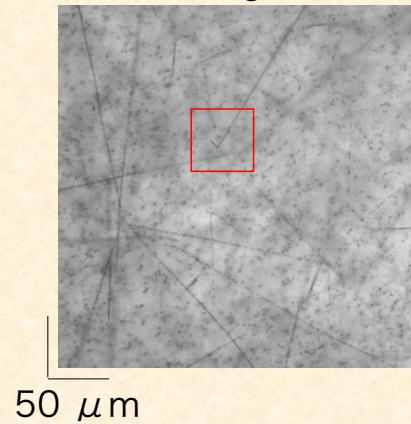
Two body decay of  ${}^3\Lambda\text{H}$



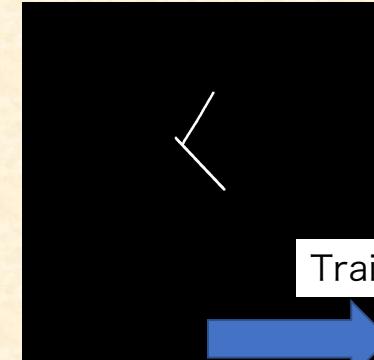
Simulated image



Training dataset (Simulated images)  
Image

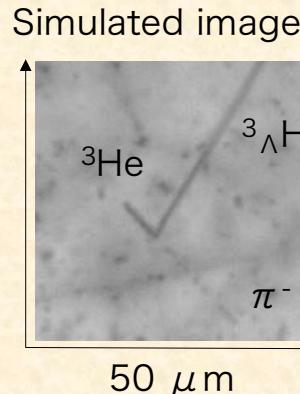
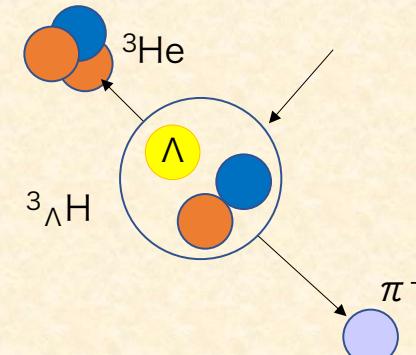


Mask

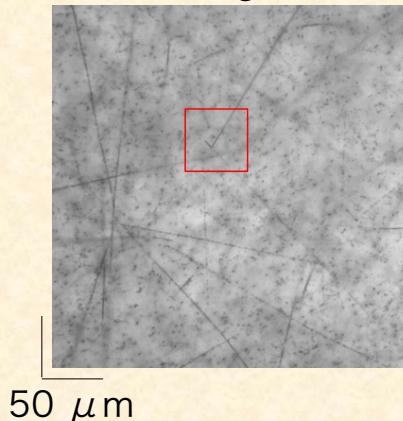


# Hypertriton search with Mask R-CNN

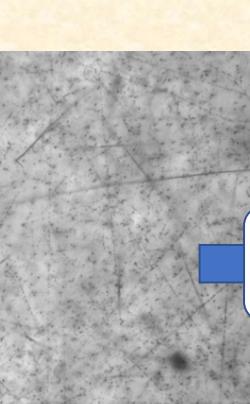
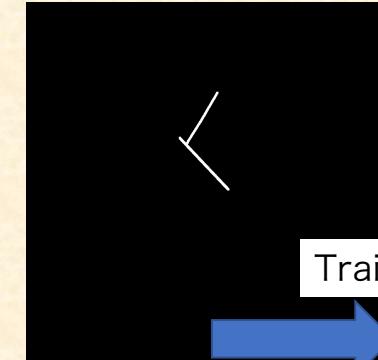
Two body decay of  ${}^3\Lambda\text{H}$



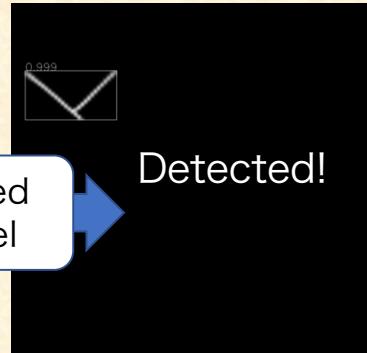
Training dataset (Simulated images)  
Image



Mask



Trained  
model



# Discovery of the first hypertriton event in E07 emulsions

nature reviews physics

Explore content ▾ About the journal ▾ Publish with us ▾

nature > nature reviews physics > perspectives > article

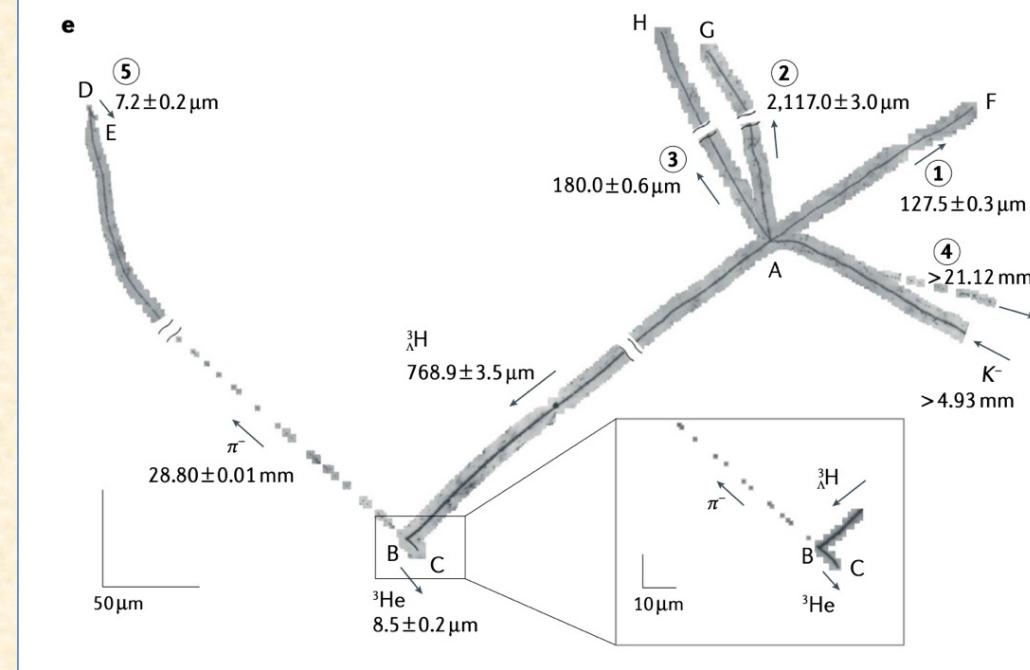
Perspective | Published: 14 September 2021

## New directions in hypernuclear physics

Takehiko R. Saito , Wenbou Dou, Vasyl Drozd, Hiroyuki Ekawa, Samuel Escrig, Yan He, Nasser Kalantar-Nayestanaki, Ayumi Kasagi, Myroslav Kavatsyuk, Enqiang Liu, Yue Ma, Shizu Minami, Abdul Muneem, Manami Nakagawa, Kazuma Nakazawa, Christophe Rappold, Nami Saito, Christoph Scheidenberger, Masato Taki, Yoshiki K. Tanaka, Junya Yoshida, Masahiro Yoshimoto, He Wang & Xiaohong Zhou

Nature Reviews Physics (2021) | Cite this article

TRS et al., Nature Reviews Physics, 803-813 (2021)  
Cover of December 2021 issue



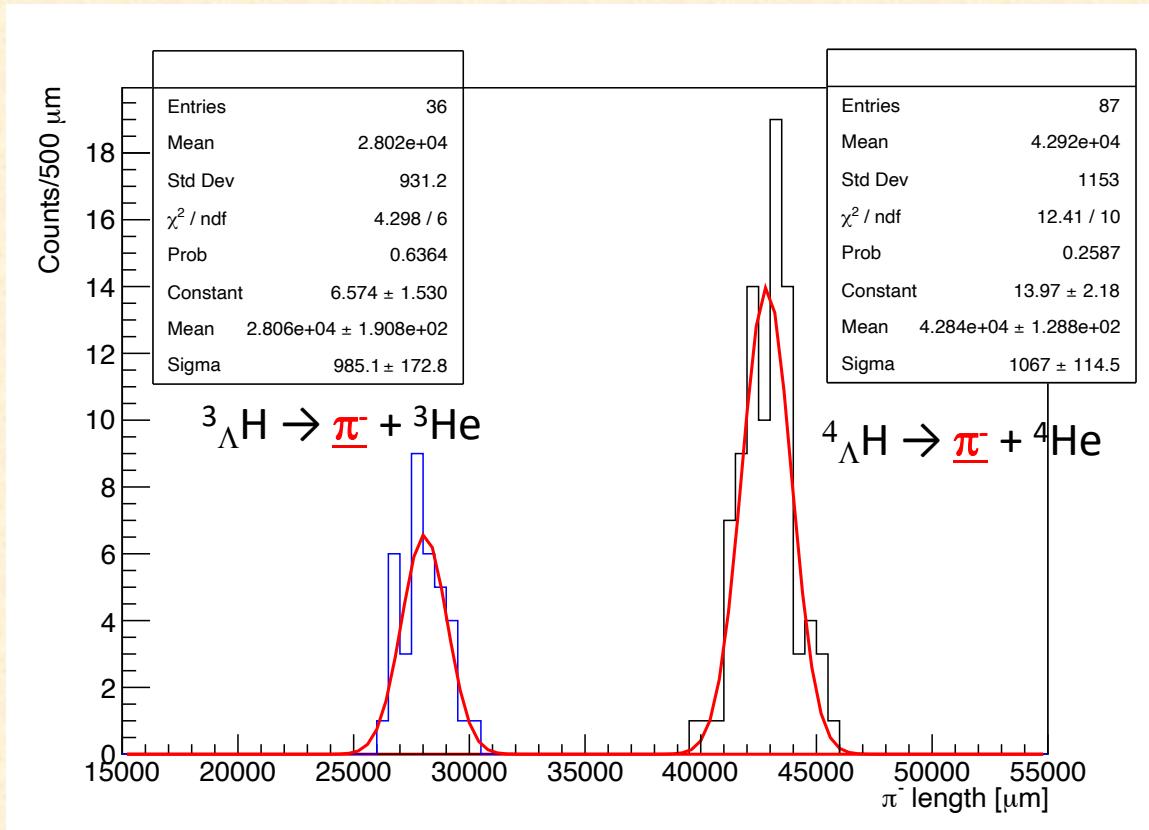
Guaranteeing the determination of  
the hypertriton binding energy SOON

Precision: 28 keV

E. Liu et al., EPJ A57 (2021) 327

Ayumi Kasagi. Ph.D. thesis (2023)  
A. Kasagi et al., submitted to NIM A

# Identification of hypertriton and ${}^4\Lambda H$ by $\pi^-$ track length



# Current status

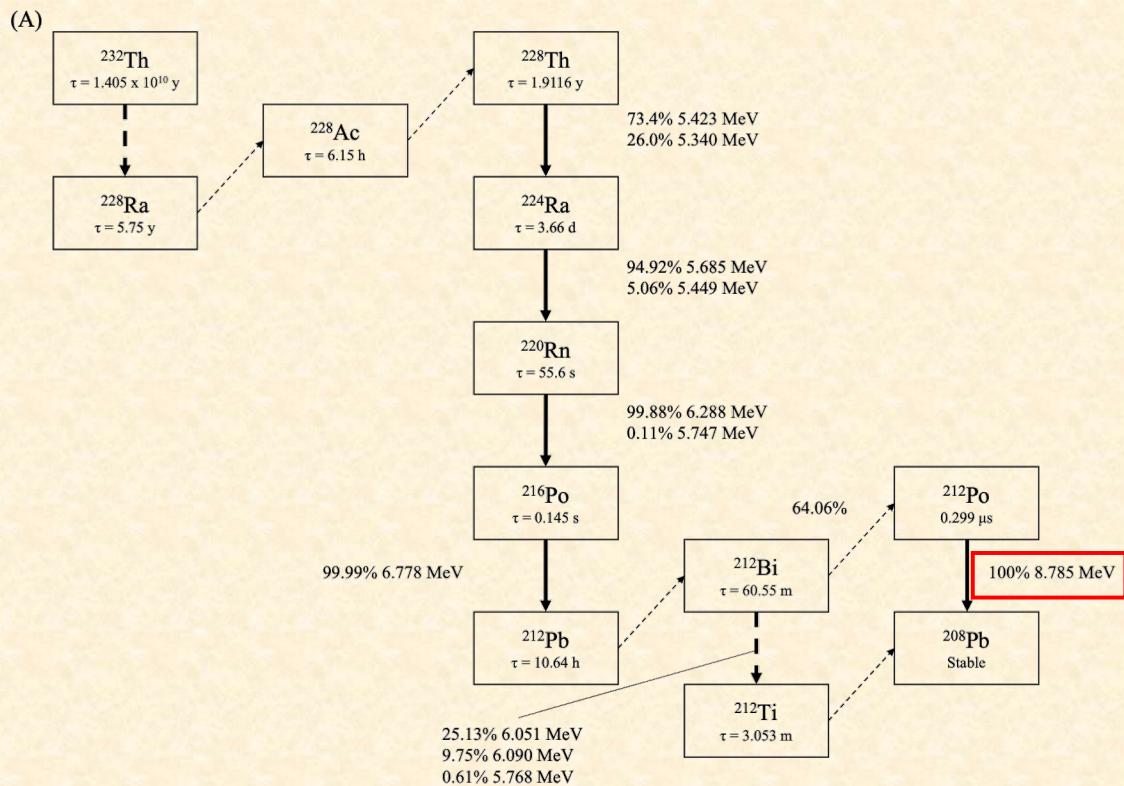
No. events: 174 (0.4% of the entire E07 data)

- ${}^3_{\Lambda}H$ : 36
- ${}^4_{\Lambda}H$ : 138 (Identified: 87 + Penetrated: 51)

Calibrated events: 143

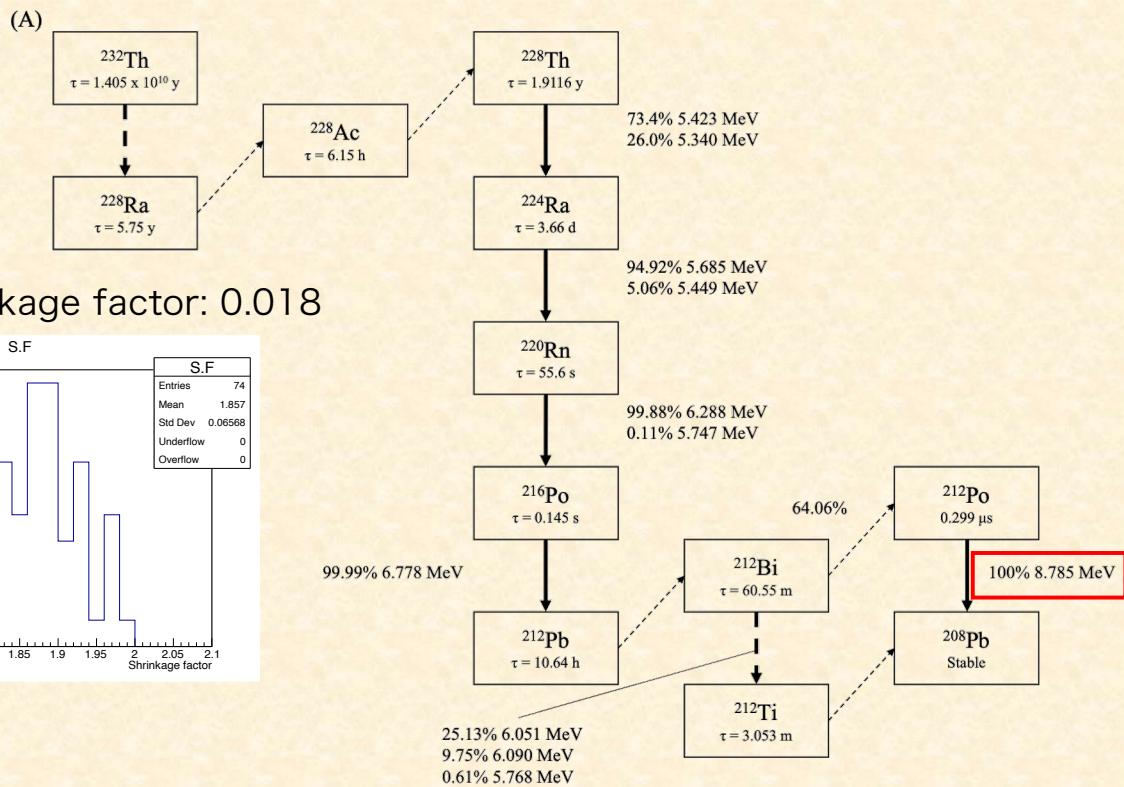
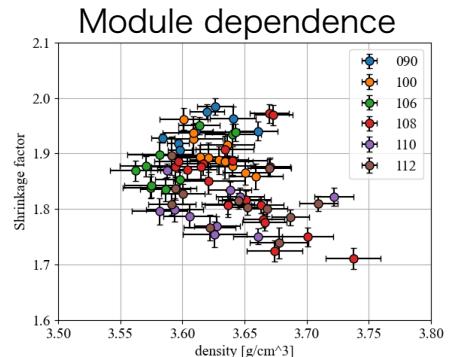
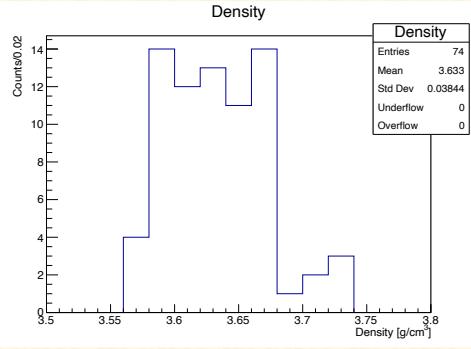
- ${}^3_{\Lambda}H$ : 36
- ${}^4_{\Lambda}H$ : 107 (Identified: 72 + Penetrated: 35)

# Calibration of nuclear emulsions



# Calibration of nuclear emulsions

Typical error: Density: 0.019 g/cm<sup>3</sup>, Shrinkage factor: 0.018

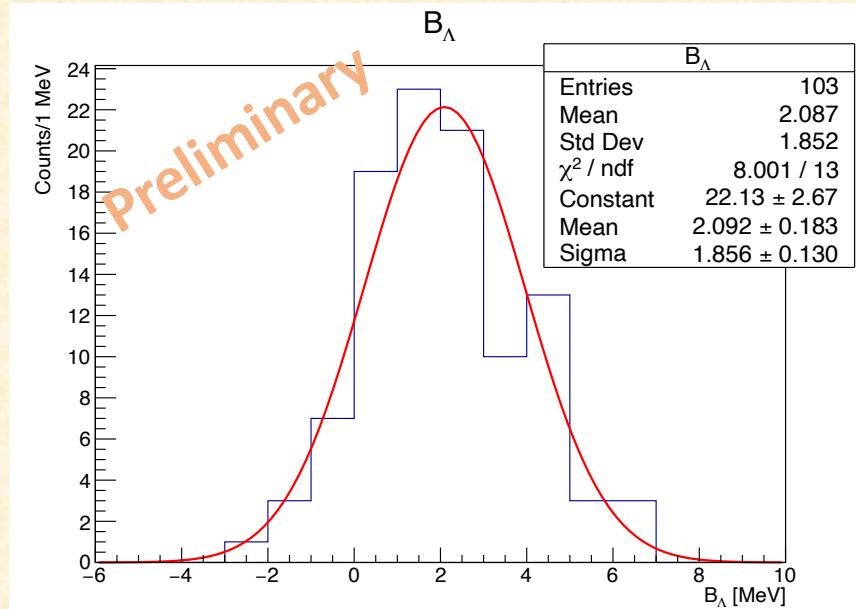
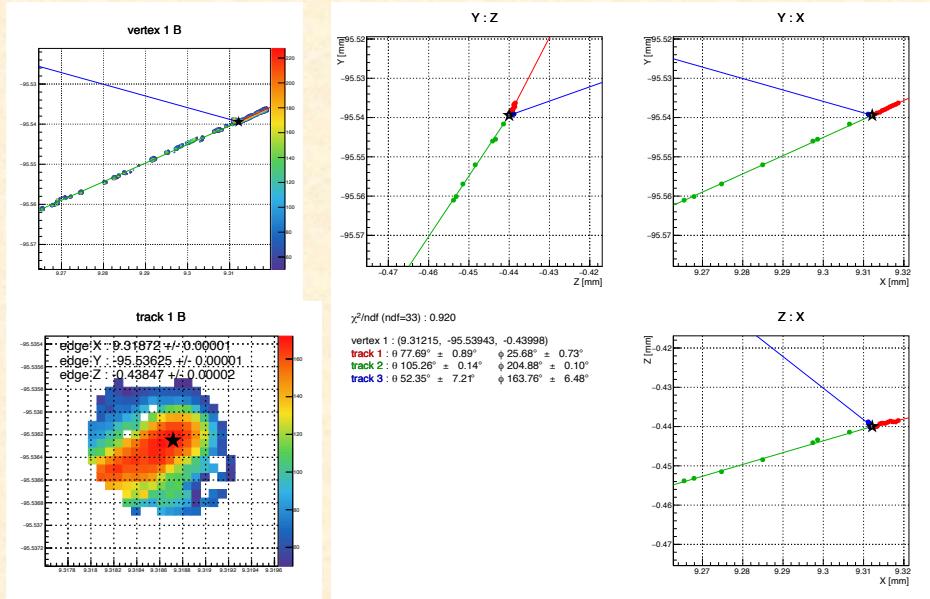


# Analysis for ${}^4\Lambda H$ binding energy

- With measured Helium momentum (Back-to-back)
- Cut conditions: Inner product = -1 ( $\pm 3\sigma$ )

Binding energy on  ${}^4\Lambda H$

Fitting results (can be checked one by one)

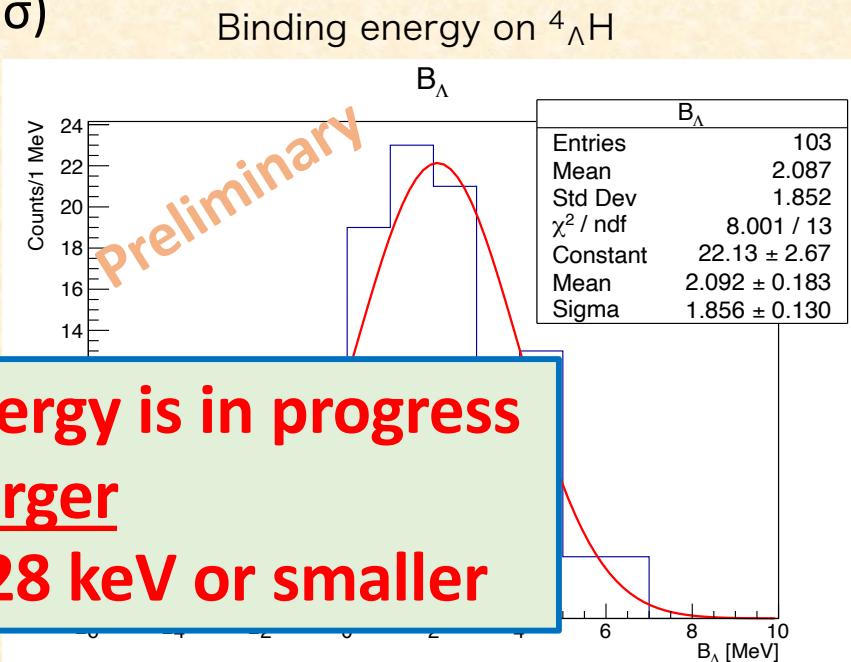
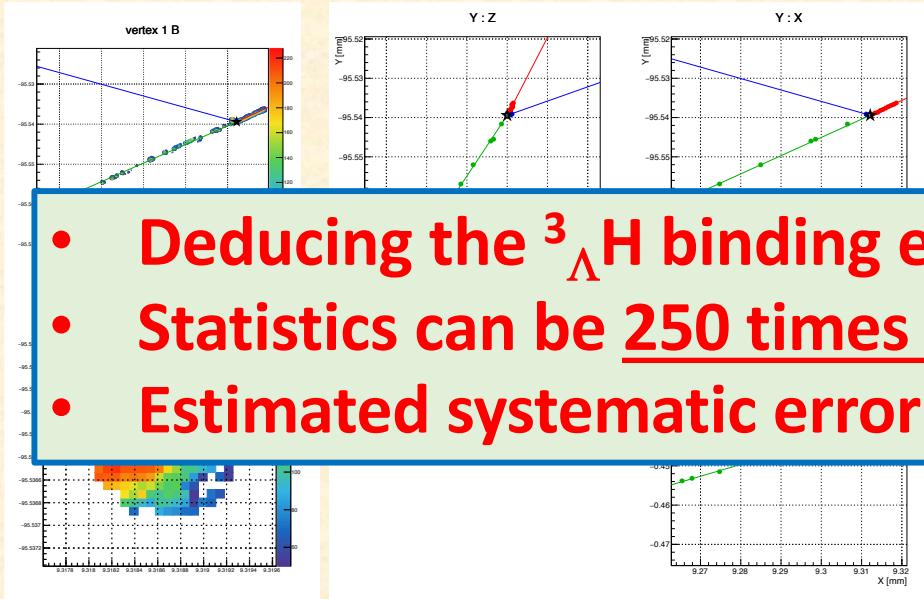


- Analyses for  $\pi^-$  will be included
- Error analysis for each event
- Weighted average: To be obtained

# Analysis for ${}^4\Lambda$ H binding energy

- With measured Helium momentum (Back-to-back)
- Cut conditions: Inner product = -1 ( $\pm 3\sigma$ )

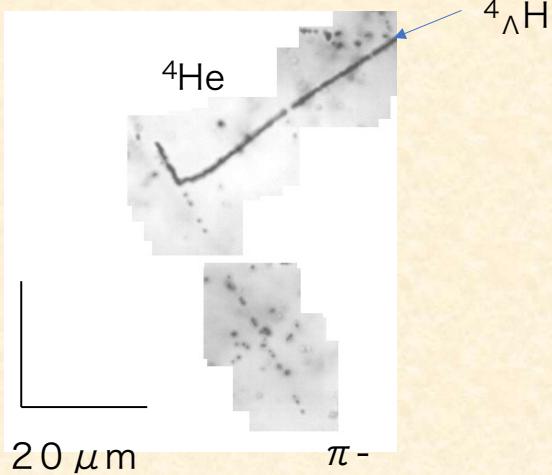
Fitting results (can be checked one by one)



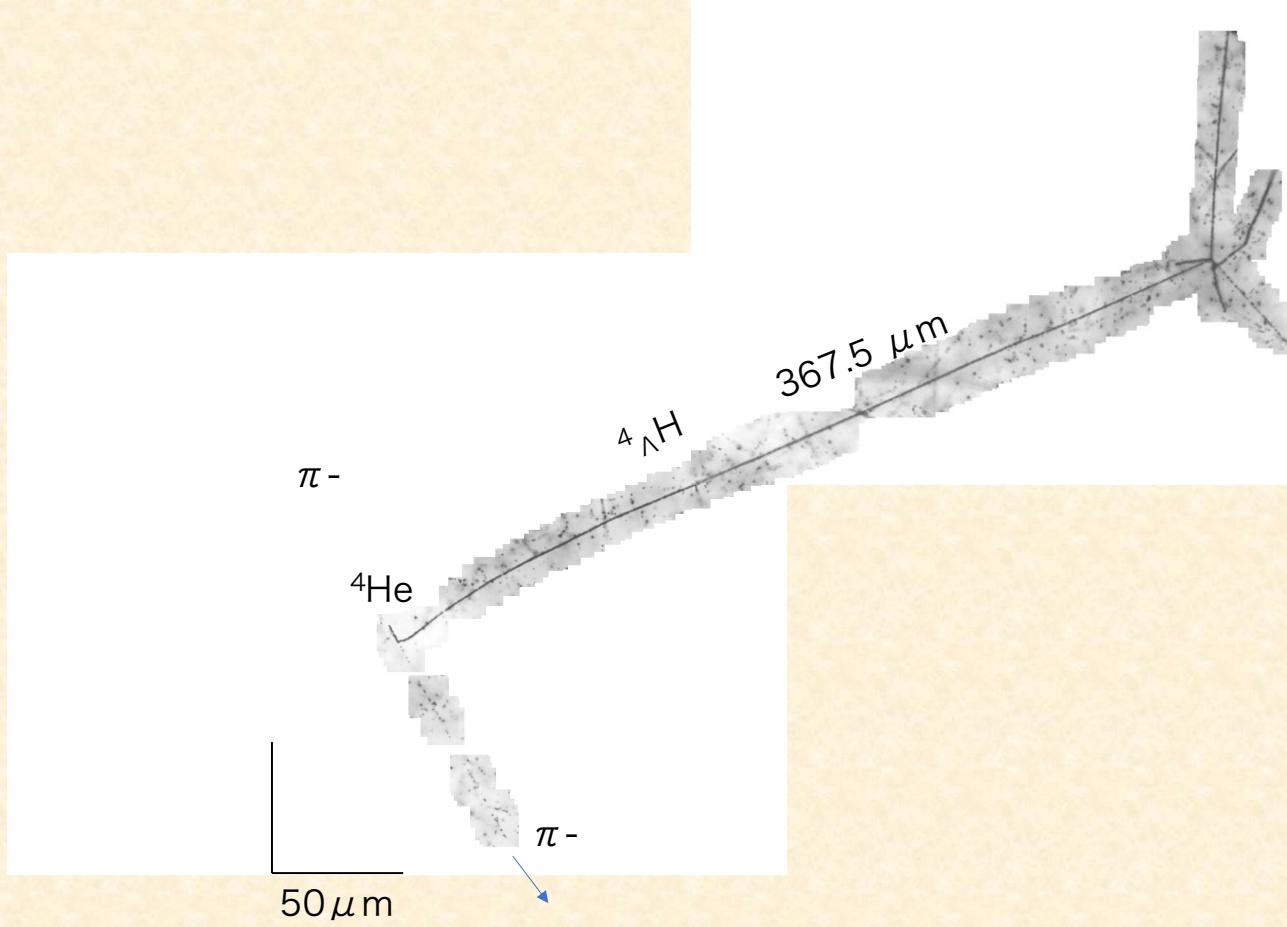
- Deducing the  ${}^3\Lambda$ H binding energy is in progress
- Statistics can be 250 times larger
- Estimated systematic error: 28 keV or smaller

- Analyses for  $\pi^-$  will be included
  - Error analysis for each event
- Weighted average: To be obtained

# Byproduct 1:

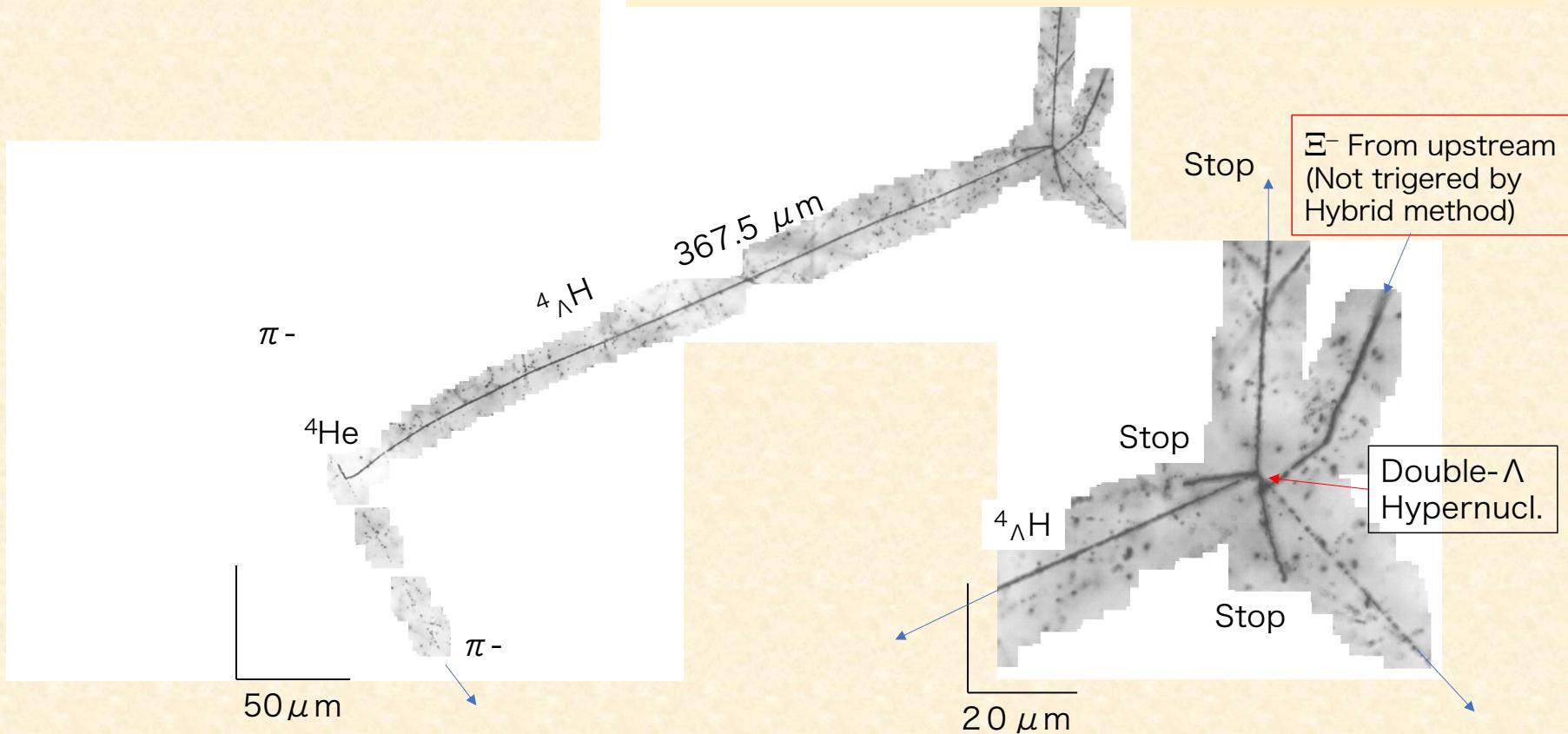


# Byproduct 1:

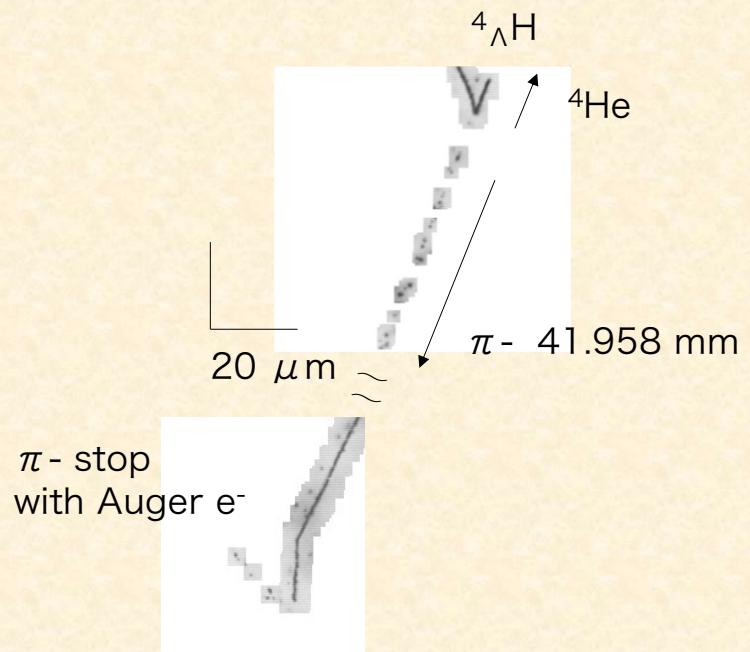


# Byproduct 1:

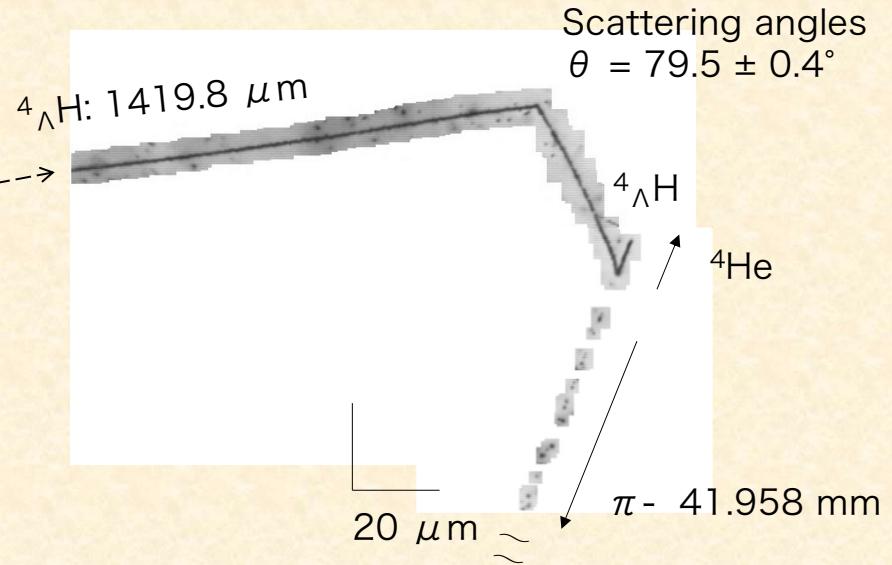
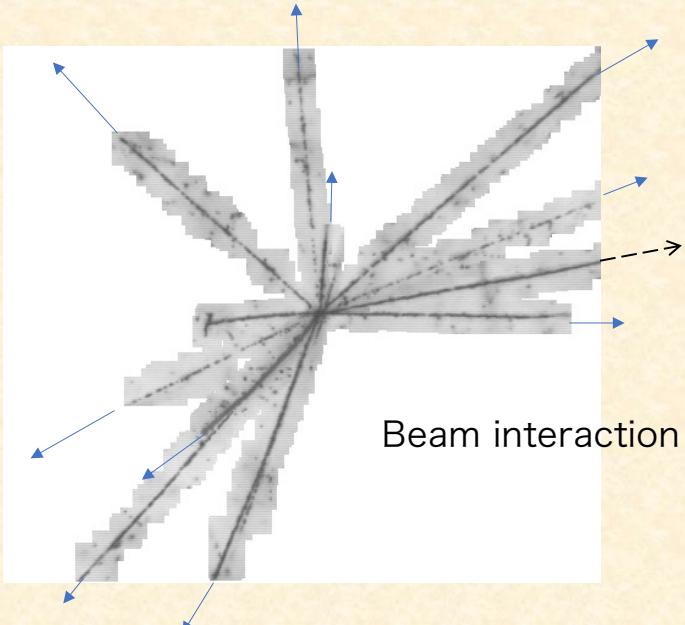
Discovery of double- $\Lambda$  hypernucleus  
as a biproduct of  ${}^3_{\Lambda}\text{H}$  search



# Byproduct 2:



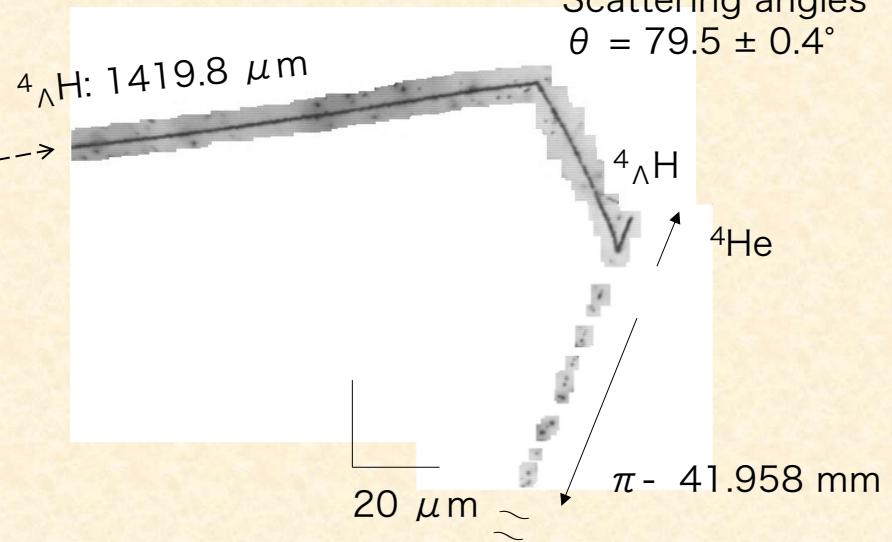
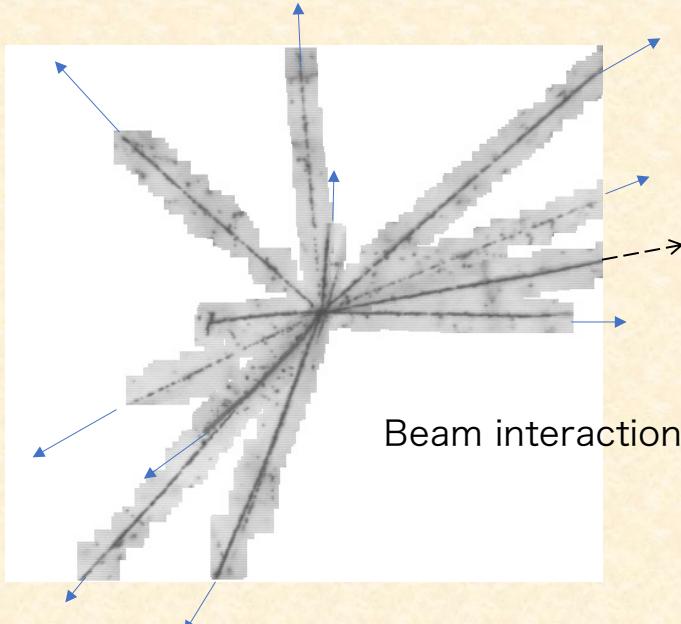
# Byproduct 2:



$\pi^-$  stop  
with Auger  $e^-$

# Byproduct 2:

## Hypernuclear scattering



$\pi^-$  - stop  
with Auger  $e^-$

# Current machine learning developments

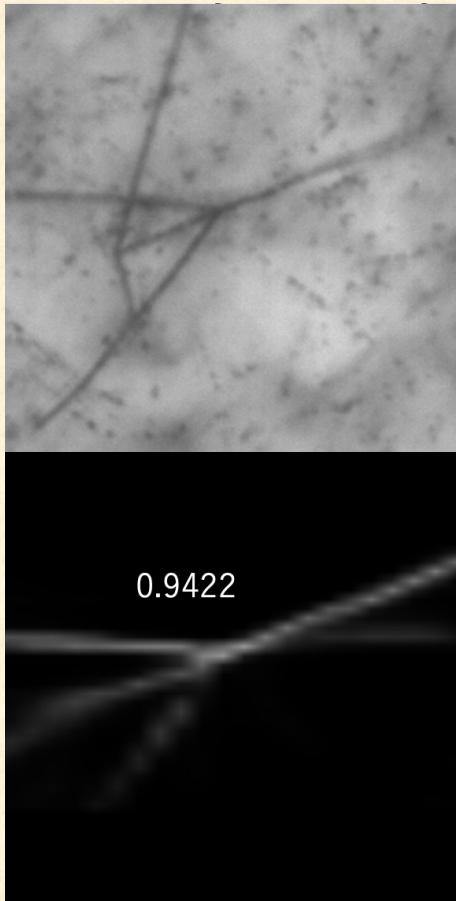
## Improvements for the hypertriton binding energy

- Automated pion tracking
- Automated emulsion calibration

## Detection of three- and multi-body single- $\Lambda$ hypernuclear decay

(from May 2022)

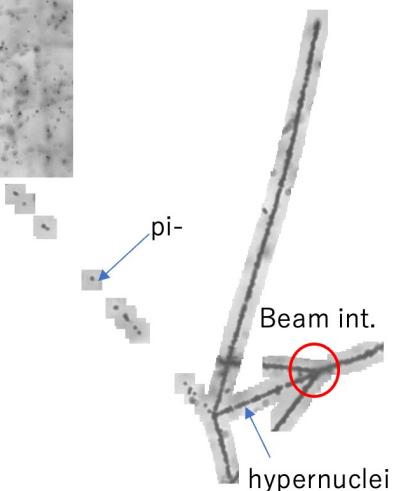
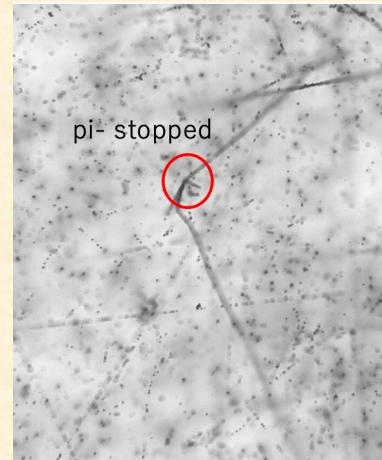
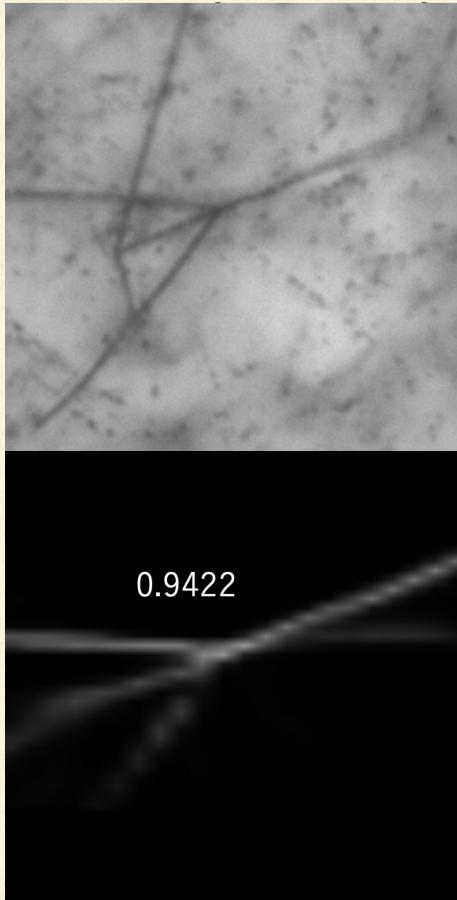
# Three-body decay event



Courtesy of Shohei Sugimoto and Manami Nakagawa

Shohei Sugimoto, Master thesis

# Three-body decay event



Courtesy of Shohei Sugimoto and Manami Nakagawa

Shohei Sugimoto, Master thesis

# Current machine learning developments

## Improvements for the hypertriton binding energy

- Automated pion tracking
- Automated emulsion calibration

## Detection of three- and multi-body single- $\Lambda$ hypernuclear decay

(from May 2022)

## Search for double-strangeness hypernuclei

(from June 2022)

MOD100\_PL02\_AREA00

V3451

$\Xi^-$  capture:

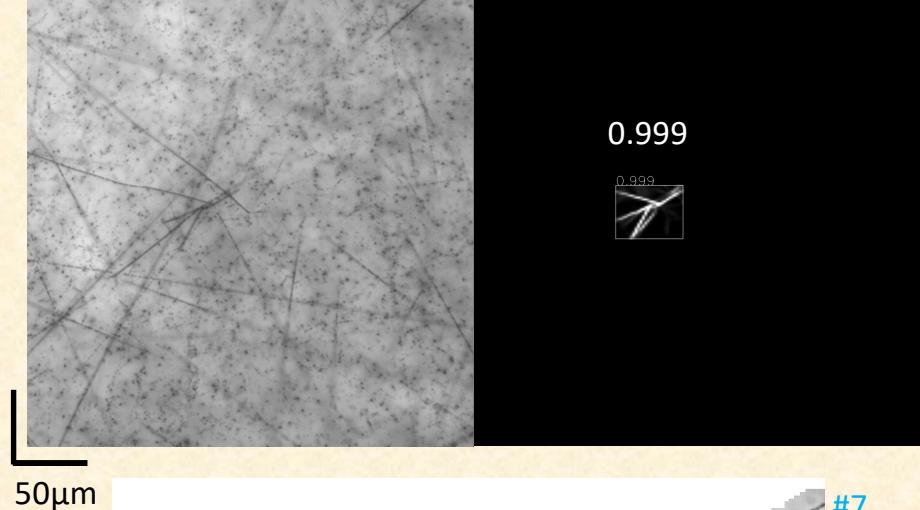
- #1: penetrate
- #2: stop
- #3: stop
- #4: decay

second vertex:

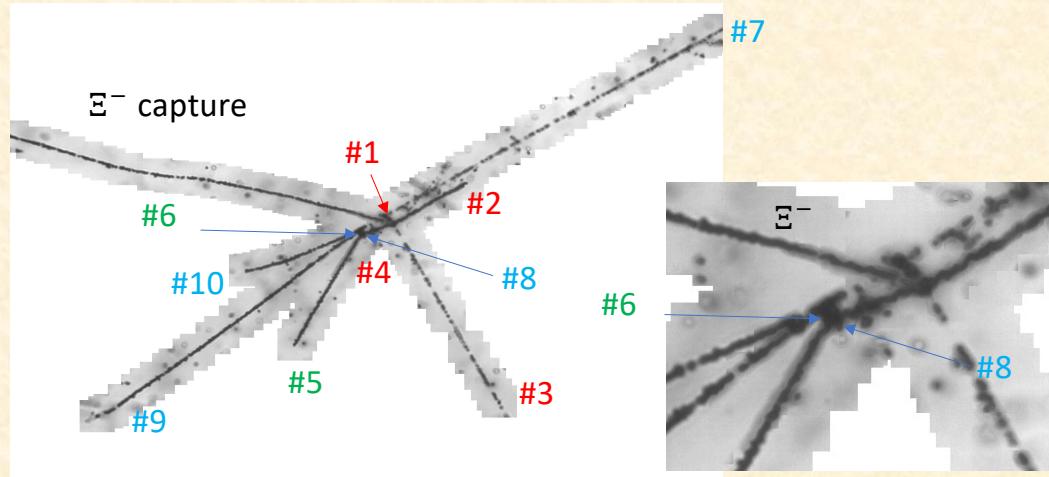
- #5: stop
- #6: decay

third vertex:

- #7: stop
- #8: stop
- #9: stop
- #9: stop



50 $\mu\text{m}$



Courtesy of Yan He and Manami Nakagawa

Only  $\sim 0.03\%$  of the entire data analyzed

Yan He, Ph.D. thesis

# Nuclear Emulsion + Machine Learning Collaboration

W. Dou<sup>a,b</sup>, V. Drozda<sup>a,c,d</sup>, H. Ekawa<sup>a</sup>, S. Escrig<sup>a,e</sup>, Y. Gao<sup>a,f,g</sup>, Y. He<sup>a,h</sup>, A. Kasagi<sup>a,i,j</sup>, E. Liu<sup>a,f,g</sup>, A. Muneem<sup>a,k</sup>, M. Nakagawa<sup>a</sup>, K. Nakazawa<sup>a,i,l</sup>, C. Rappold<sup>e</sup>, N. Saito<sup>a</sup>, T.R. Saito<sup>a,d,h</sup>, S. Sugimoto<sup>a,b</sup>, M. Taki<sup>j</sup>, Y.K. Tanaka<sup>a</sup>, A. Yanai<sup>a,b</sup>, J. Yoshida<sup>a,m</sup>, M. Yoshimoto<sup>n</sup>, and H. Wang<sup>a</sup>

<sup>a</sup> High Energy Nuclear Physics Laboratory, RIKEN, Japan

<sup>b</sup> Department of Physics, Saitama University, Japan

<sup>c</sup> Energy and Sustainability Research Institute Groningen, University of Groningen, Netherlands

<sup>d</sup> GSI Helmholtz Centre for Heavy Ion Research, Germany

<sup>e</sup> Instituto de Estructura de la Materia, Spain

<sup>f</sup> Institute of Modern Physics, Chinese Academy of Sciences, China

<sup>g</sup> University of Chinese Academy of Sciences, China

<sup>h</sup> School of Nuclear Science and Technology, Lanzhou University, China

<sup>i</sup> Graduate School of Engineering, Gifu University, Japan

<sup>j</sup> Graduate School of Artificial Intelligence and Science, Rikkyo University, Japan

<sup>k</sup> Faculty of Engineering Sciences, Ghulam Ishaq Khan Institute of Engineering Sciences and Technology, Pakistan

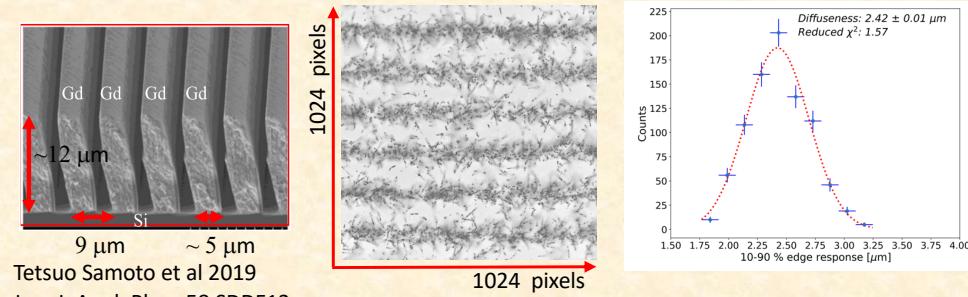
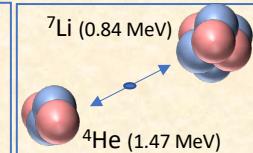
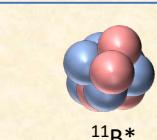
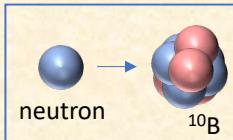
<sup>l</sup> Faculty of Education, Gifu University, Japan

<sup>m</sup> Department of physics, Tohoku University, Japan

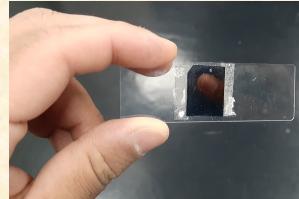
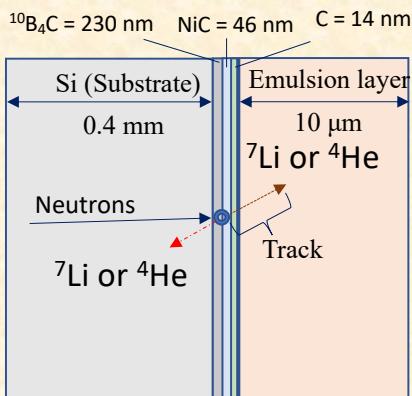
<sup>n</sup> RIKEN Nishina Center, RIKEN, Japan

# Neutron imaging with nuclear emulsions

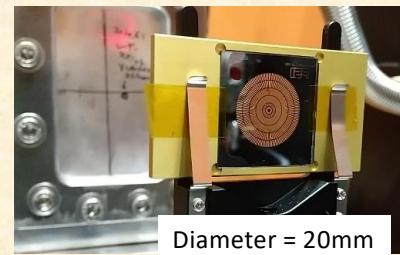
## Physics process



## Neutron detector



## Siemens star pattern



A. Muneem et al., Journal of Applied Physics, 133, 054902 (2023)

Precise 2D imaging  
Submicron resolution

Image reconstruction using beams  
with large angular dispersion and ML

Neutron CT

# Summary

## The WASA-FRS experiment at GSI (2022) with HI beams

- Hypertriton lifetime
- nn $\Lambda$  state
- Proton-rich hypernuclei
- Development with **Machine Learning (GNN)**
- Data analyses in progress

## Analyses of the J-PARC E07 nuclear emulsion with Machine Learning

- Binding energy of  ${}^3_{\Lambda}\text{H}$  and  ${}^4_{\Lambda}\text{H}$
- Single- $\Lambda$  hypernuclei with multi-body decay channels
- Double-strangeness hypernuclei