Machine learning techniques applied to study light hypernuclei



High Energy Nuclear Physics Laboratory, RIKEN, Japan

Take R. Saito

and

HRS-HYS Research Group, GSI, Germany



for the Nuclear Emulsion + ML Collaboration

W. Dou, V. Drozd, H. Ekawa, S. Escrig, Y. Gao, <u>Y. He</u>, <u>A. Kasagi</u>, E. Liu, A. Muneem, <u>M. Nakagawa</u>, K. Nakazawa, C. Rappold, N. Saito, T.R. Saito, <u>S. Sugimoto</u>, M. Taki, Y.K. Tanaka, A. Yanai, J. Yoshida, M. Yoshimoto, and H. Wang

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S. Sugimoto Master student RIKEN & Saitama U.

HADRON2023, Genova, Italy, June 5th - 9th, 2023

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S. Sugimoto Master student **RIKEN & Saitama U.**

Technical staffs for nuclear emulsions





RIKEN



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M. Ando RIKEN

C. Harisaki R. Kobayashi

H. Kubota RIKEN

HADRON2023, Genova, Italy, June 5th – 9th, 2023

Quarks and sub-atomic nuclei







There are many identical quarks

Quarks and sub-atomic nuclei





STAR Collaboration, PRL 128 (2022) 202301



³ΛH Binding energy
 BΛ(³ΛH) : 0.13 ± 0.05 MeV
 G. Bohm et al., NPB 4 (1968) 511
 M. Juric et al., NPB 52 (1973) 1

STAR (2020)

0.41 ± 0.12 ± 0.11 MeV

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

STAR Collaboration, PRL 128 (2022) 202301



STAR Collaboration, PRL 128 (2022) 202301

On Λnn





```
0.41 ± 0.12 ± 0.11 MeV
```

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



HypHI., PRC 88 (2013) 041001



FIG. 5. The enlarged mass spectrum around the Λnn threshold. Two additional Gaussians were fitted together with the known contributions (the accidentals, the Λ quasifree, the free Λ , and the ³He 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



STAR Collaboration, PRL 128 (2022) 202301

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

Hiroyuki Ekawa

1.1

VAT

2412

Photos by Jan Hosan and GSI/FAIR



STAR Collaboration, PRL 128 (2022) 202301

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

Nuclear Emulsion:

Charged particle tracker with <u>the best spatial resolution</u> (easy to be < 1 μm, 11 nm at best)

Silver halide crystal Diameter: 200 nm Charged particle Medium: gelatin Development Silver clusters

(Latent image)

Getting bigger

20µm





J-PARC accelerator facility



J-PARC E07 experiment

K⁻ Beam (180cm above the floor)

Emulsion module

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

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J-PARC E07 experiment

K⁻ Beam (180cm above the floor)

Emulsion module



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

Results from J-PARC E07 (Hybrid method)



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

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

































Data size:

- 10⁷ images per emulsion (100 T Byte)
 10¹⁰ images per 1000 emulsions (100 P Byte)
 Number of background tracks:
 Beam tracks: 10⁴/mm²
- •Nuclear fragmentations: 10³/mm²

Current equipments/techniques with visual inspections

560 years







100µm

Data size:

- 10⁷ images per emulsion (100 T Byte)
 10¹⁰ images per 1000 emulsions (100 P Byte)
 Number of background tracks:
 Beam tracks: 10⁴/mm²
- •Nuclear fragmentations: 10³/mm²

Current equipments/techniques with visual inspections

560 years

3 vears

Machine Learning Sliced image



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

Monte Carlo simulations and GAN(Generative Adversarial Networks)



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

Edges to Photo

GAN: pix2pix

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

Monte Carlo simulations and GAN(Generative Adversarial Networks)



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



https://www.cis.upenn.edu/~jshi/ped_html/



Detection of each object

berson (

At large object density

car 0.92

car 0.860 car 0.931

Hypertriton search with Mask R-CNN

³He





Simulated image

50 µm

³∧H

 π^{-}



50 µm

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

Hypertriton search with Mask R-CNN



Status of the project for hypertriton and ${}^4_{\Lambda}H$



Discovery of the first hypertriton event in E07 emulsions

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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

Departure 2013 relation from 12 metallicity control regulary

nature reviews physics



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 π^{-} track length



Ayumi Kasagi. Ph.D. thesis (2023)

Current status

No. events: 174 (<u>0.4% of the entire E07 data</u>) • ³_^H: 36

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

Calibrated events: 143

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

Calibration of nuclear emulsions

(A)



Calibration of nuclear emulsions

(A)

S.F

Counts/0.02

1.6 1.65 1.75

1.8 1.85 1.9



Density Counts/0.02 Entries 74 Mean 3.633 Std Dev 0.03844 Underflow 10 Overflow 6 3.5 3.55 3.6 3.65 3.7 Density [g/cm³] Module dependence

Density



Ayumi Kasagi. Ph.D. thesis (2023)

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_AH



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_AH



→Weighted avarage: To be obtained





Byproduct 1:

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



Byproduct 2:



Byproduct 2:



Byproduct 2:

Hypernuclear scattering



Current machine learning developments

Improvements for the hypertriton binding energy

- Automated pion tracking
- Automated emulsion calibration

Detection of three- and multi-body single- Λ hypernuclear decay (from May 2022)

Three-body decay event



0.9422

Courtesy of Shohei Sugimoto and Manami Nakagawa

Shohei Sugimoto, Master thesis

Three-body decay event



Current machine learning developments

Improvements for the hypertriton binding energy

- Automated pion tracking
- Automated emulsion calibration

Detection of three- and multi-body single- Λ hypernuclear decay (from May 2022)

Search for double-strangeness hypernuclei (from June 2022)

MOD100_PL02_AREA00

V3451

E- 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



#3

#5

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^{a,b}, V. Drozd^{a,c,d}, H. Ekawa^a, S. Escrig^{a,e}, Y. Gao^{a,f,g}, Y. He^{a,h}, A. Kasagi^{a,i,j}, E. Liu^{a,f,g}, A. Muneem^{a,k}, M. Nakagawa^a, K. Nakazawa^{a,i,l}, C. Rappold^e, N. Saito^a, T.R. Saito^{a,d,h}, S. Sugimoto^{a,b}, M. Taki^j, Y.K. Tanaka^a, A. Yanai^{a,b}, J. Yoshida^{a,m}, M. Yoshimotoⁿ, and H. Wang^a

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Summary

Nuclear emulsion + Machine learning

• Binding energy of ${}^{3}_{\Lambda}$ H and ${}^{4}_{\Lambda}$ H

For ${}^4\Lambda H$

Currently 0.4 % of the entire data: $\Delta B_{\Lambda} \sim 180$ keV (only with He track length) With the entire data: $\Delta B_{\Lambda} \sim 10$ keV (but systematic uncertainty will be 20 keV) Analyzing 25 % data will be sufficient

Single-strangeness hypernuclei with multi-body decays

Heavier hypernuclei

>Increasing statistics for ${}^{3}{}_{\Lambda}$ H and ${}^{4}{}_{\Lambda}$ H

Double-strangeness hypernuclei