

New Symmetry Energy Constraint from a Model-Independent Measurement of Isospin Diffusion with INDRA-FAZIA

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for the INDRA-FAZIA collaboration

GANIL seminar

June 24th, 2025

1 Introduction

- Nuclear equation of state
- Isospin transport phenomena

2 Experimental data

- The INDRA-FAZIA apparatus
- Model-independent impact parameter reconstruction
- Isospin analysis

3 Model predictions and comparison

- BUU@VECC-McGill simulations
- Comparison protocol
- Extraction of the sensitive density
- Constraint on the symmetry energy

Isospin transport phenomena

Probing the symmetry energy of the nEoS

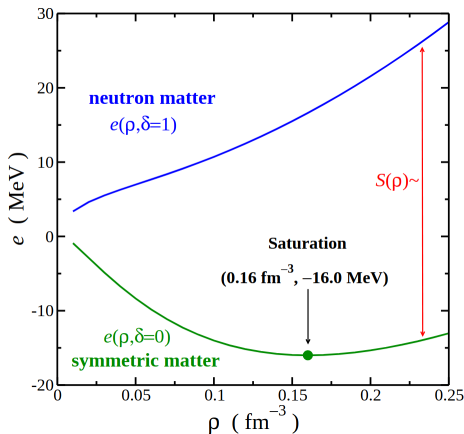
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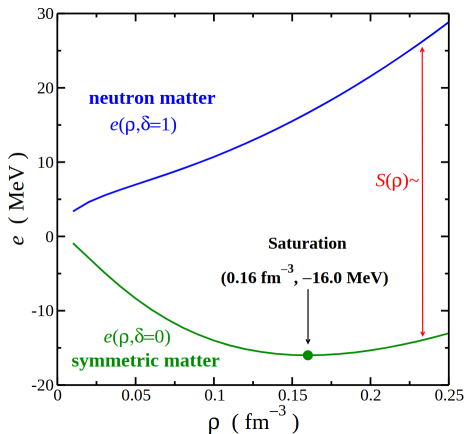


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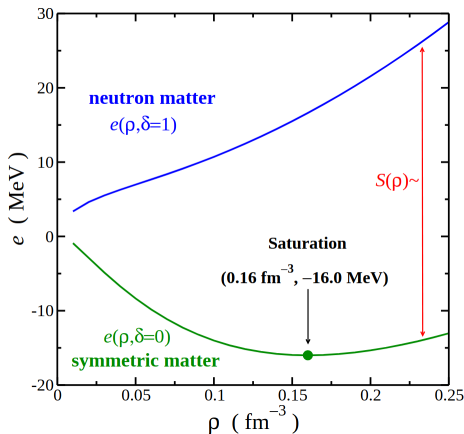
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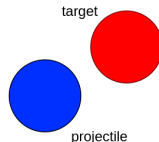
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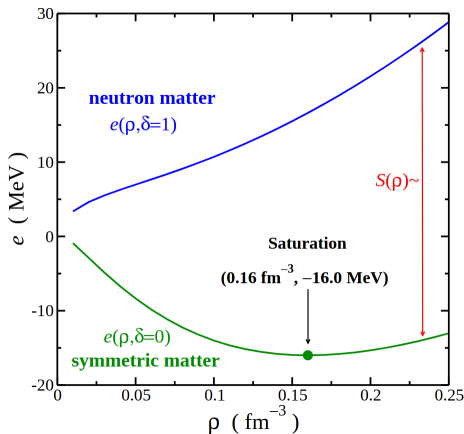


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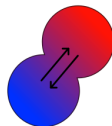
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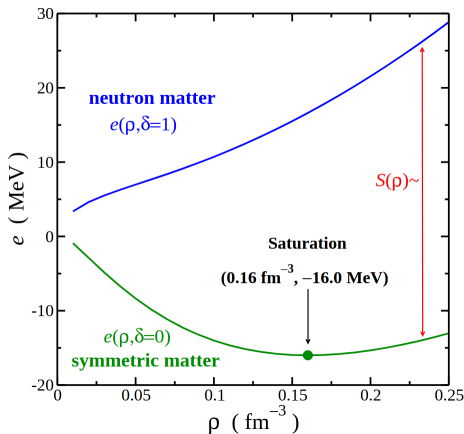


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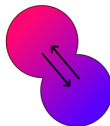
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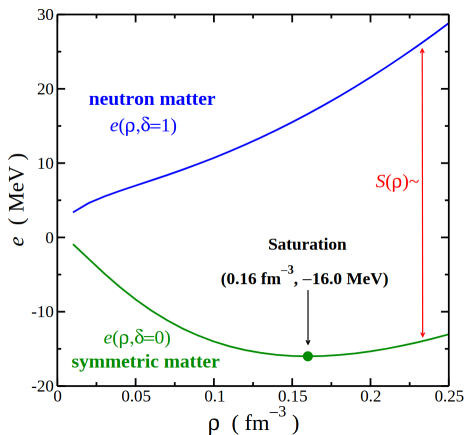


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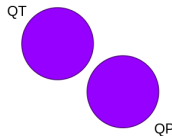
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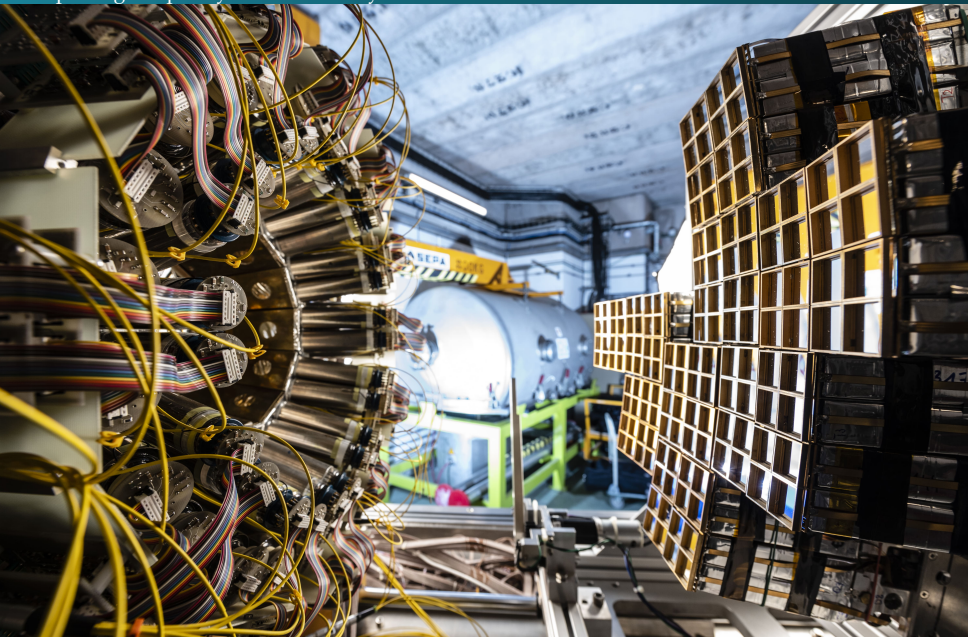
Experimental data

or

what do we need to “measure” isospin diffusion?

The INDRA-FAZIA setup

Exploring isospin dynamics in heavy ion collisions



The INDRA-FAZIA setup

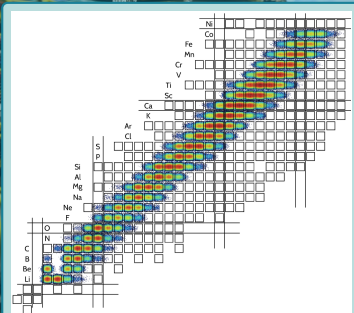
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FAZIA

Forward-angle A and Z Identification Array

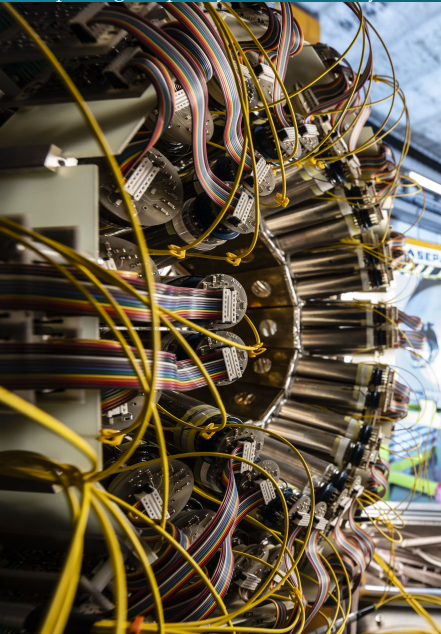
State of the art of ion identification
in the Fermi energy domain

Covers the most forward polar angles
for (Z, A) identification of QP-like fragments



The INDRA-FAZIA setup

Exploring isospin dynamics in heavy ion collisions



INDRA

*Identification de Noyaux et Détection
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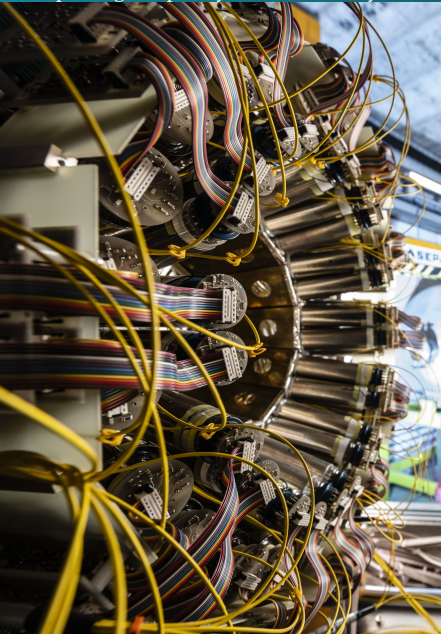
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Provides a good
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Impact parameter reconstruction

Basic structure of the method

Centrality-related observable $X \longleftrightarrow$ deduce the correspondence with b
(see J. D. Frankland et al., PRC104, 034609 (2021), R. Rogly et al., PRC98, 024902 (2018))
 $\Rightarrow$ Need to model the conditional probability distribution: $P(\mathbf{X}|\mathbf{b})$

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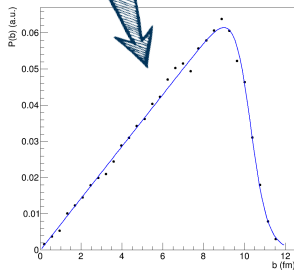
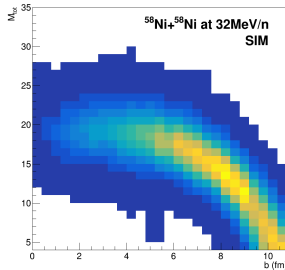
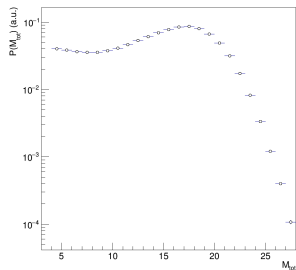
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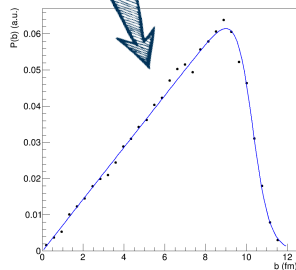
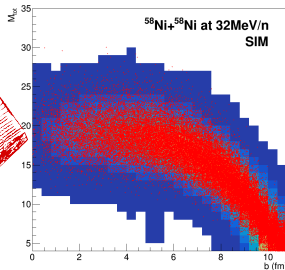
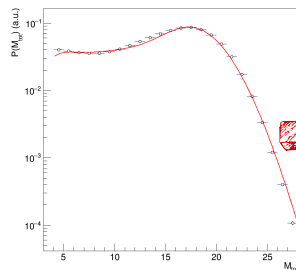
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Step 3

Having the $\mathbf{P}(\mathbf{X}|\mathbf{b})$, for each X selection we can evaluate:

$$P(b|x_1 < X < x_2) = \frac{\int_{x_1}^{x_2} P(b, X) dX}{\int_{x_1}^{x_2} P(X) dX} = \frac{\int_{x_1}^{x_2} P(X) P(b|X) dX}{\int_{x_1}^{x_2} P(X) dX} = \frac{\int_{x_1}^{x_2} P(b) \mathbf{P}(\mathbf{X}|\mathbf{b}) dX}{\int_{x_1}^{x_2} P(X) dX}$$



To obtain the impact parameter distribution, it is necessary to perform the fit on the most inclusive $P(X)$ distribution, for which the $P(b)$ above can be assumed.

Impact parameter reconstruction

Applying the method: overview of the INDRA and FAZIA datasets

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INDRA-FAZIA dataset $\rightarrow$ $^{58,64}\text{Ni} + ^{58,64}\text{Ni}$ at 32 MeV/nucl.

Includes the information on the *isospin content of the QP remnant*.

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Some events are discarded in a non-trivial way, especially for semiperipheral collisions. The triangular $P(b)$ distribution does not well represent the experimental one.

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INDRA dataset $\rightarrow {}^{58}\text{Ni} + {}^{58}\text{Ni}$ at 32 MeV/nucl.

- Trigger condition: $M_{\text{tot}} \geq 4$

Minimum bias, the $P(b)$ can be well approximated as shown before (with $\Delta b \approx 0.4\text{fm}$).
(see J. D. Frankland et al., Phys. Rev. C 104, 034609 (2021), E. Vient et al., Phys. Rev. C 98, 044612 (2018))

Suitable for the application of the *impact parameter reconstruction method*.

Impact parameter reconstruction

Implementation of the impact parameter reconstruction

Procedure for the reaction in common $^{58}\text{Ni}+^{58}\text{Ni}$ at 32 MeV/nucleon:

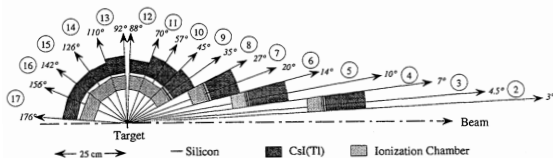
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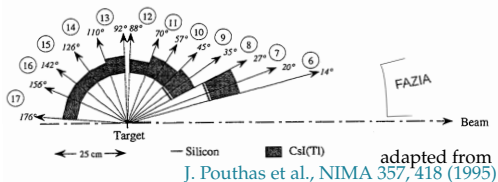


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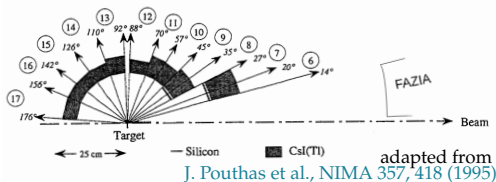


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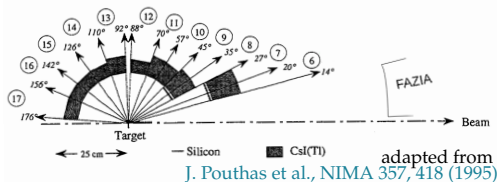
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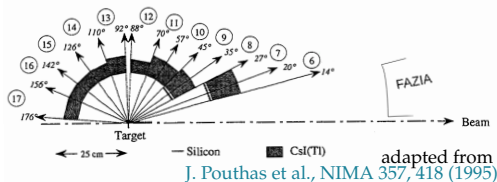
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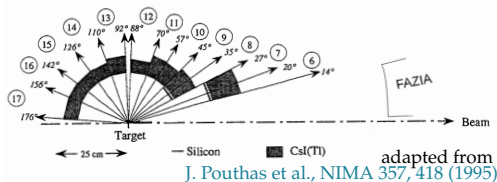
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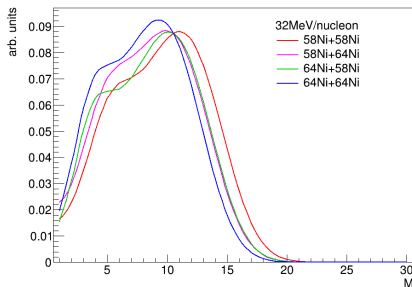
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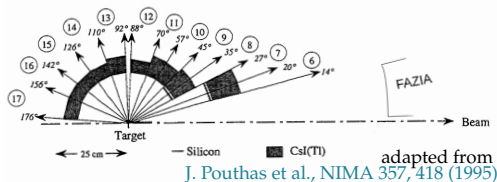


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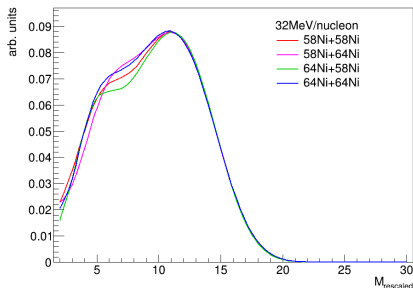
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Procedure for the other systems:

→ rescale the detected multiplicity M_{sys} into a corresponding M_{resc} value for $^{58}\text{Ni}+^{58}\text{Ni}$.

$$M_{\text{resc}} = \lfloor \alpha \cdot (M_{\text{sys}} + r) + \beta \rfloor$$

where $r \sim U([0, 1])$ is a uniformly distributed random variable taking values in $[0, 1]$.



Impact parameter reconstruction

Setting the parameters for $P(b)$ model independently

To set the parameters of $P(b)$ in a model independent way:

$$P(b) = \frac{2\pi b}{1 + \exp[(b - b_0)/\Delta b]}$$

$\Delta b \approx 0.4 \text{ fm}$ as verified in PRC 104, 034609 (2021)

b_0 by inverting $\sigma_R = -2\pi(\Delta b)^2 \text{Li}_2\left[-\exp\left(\frac{b_0}{\Delta b}\right)\right]$

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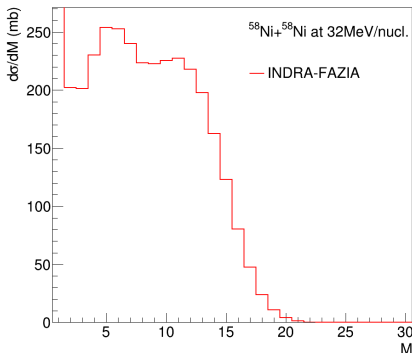
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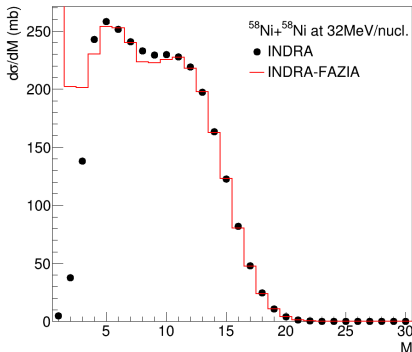
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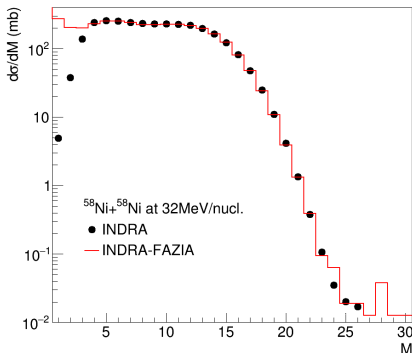
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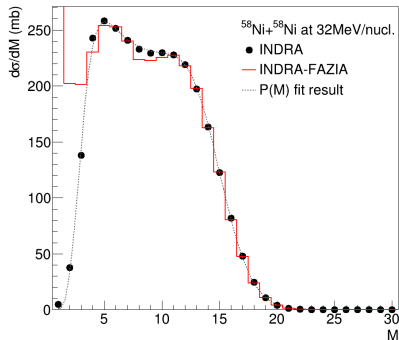


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- Transfer the normalization to the INDRA dataset using the high multiplicity tail, after correcting for small trigger effect
- From the total reaction cross section σ_R for INDRA dataset $\Rightarrow b_0 = (9.8 \pm 0.7) \text{ fm}$

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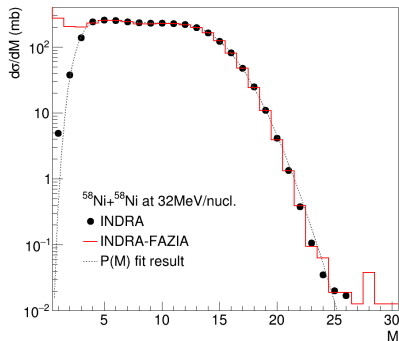
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Fit result on multiplicity M of identified and unidentified particles in INDRA rings 6-17 for $^{58}\text{Ni}+^{58}\text{Ni}$ at 32 MeV/nucleon on INDRA dataset

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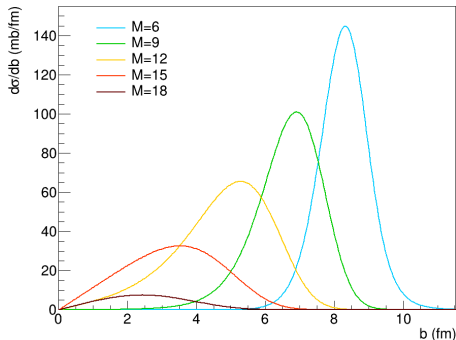
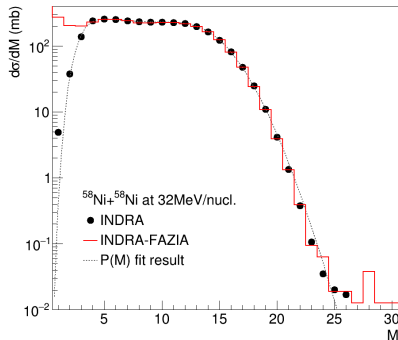
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Fit result on multiplicity M of identified and unidentified particles in INDRA rings 6-17 for $^{58}\text{Ni}+^{58}\text{Ni}$ at 32 MeV/nucleon on INDRA dataset

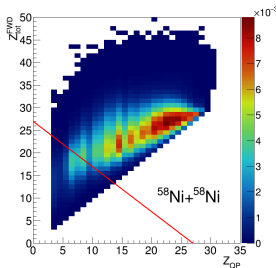
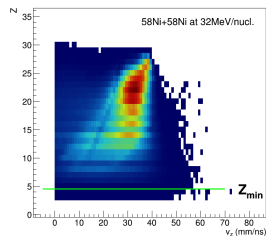
→ important role of **intrinsic fluctuations**: relatively different M selections populate partly (or entirely) superimposed b intervals

Isospin analysis

Selection of the events

In view of producing the most general result, easily comparable with any theoretical prediction, we avoid a strictly exclusive analysis.

- No distinction among different output channels
- QP remnant selected as:
 - 1 fragment with largest Z in forward hemisphere
 - 2 if more than one with same Z , select largest $v_z^{\text{c.m.}}$
- Minimum size to consider a QP remnant: $Z_{\text{QP}} \geq 5$
→ include light products from very dissipative events



Carefully check the completeness of the event:

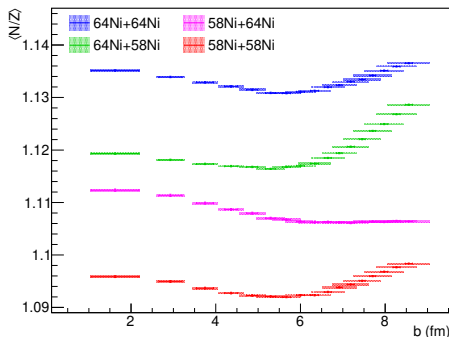
- The total undetected charge in the forward hemisphere should not exceed Z_{QP}
- Accept event if $Z_{\text{QP}} \geq 28 - Z_{\text{tot}}^{\text{FWD}}$
- We verified that by removing $< 13\%$ of events, the final result becomes stable against reasonable variations of $Z_{\text{QP}}^{\text{min}}$

Isospin analysis

Evolution of isospin equilibration with centrality

Each event is assigned an impact parameter value randomly drawn from the b distribution associated with its corresponding M_{resc} .

⇒ Take into account the fluctuations



Model-independent $\langle N/Z \rangle$ for the QP remnant as a function of b for the four systems in the INDRA-FAZIA dataset

Clear effect of isospin equilibration down to the most central collisions:

- *peripheral*: similar result for reactions with same projectile
- *central*: $\langle N/Z \rangle$ depends on target, mixed systems tend to each other

The horizontal error bars are associated with the uncertainty on the estimation of b_0 in the $P(b)$ assumed for the impact parameter reconstruction method, affecting less central collisions to a greater extent.

Isospin analysis

Model independent isospin transport ratio

Isospin transport ratio: can highlight the isospin diffusion effect, bypassing the effects acting similarly on the four systems (F. Rami et al., *Phys. Rev. Lett.* 84, 1120 (2000))

$$R(x) = \frac{2x_i - x_{AA} - x_{BB}}{x_{AA} - x_{BB}}$$

where x is an isospin sensitive observable,
 $A = {}^{64}\text{Ni}$, $B = {}^{58}\text{Ni}$ and $i = AA, AB, BA, BB$.

Isospin analysis

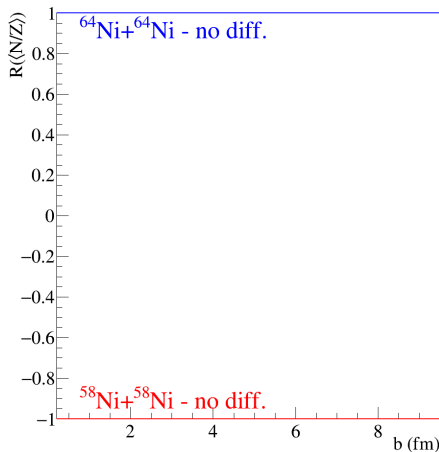
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$R(x) = \pm 1 \rightarrow$ non equilibrated

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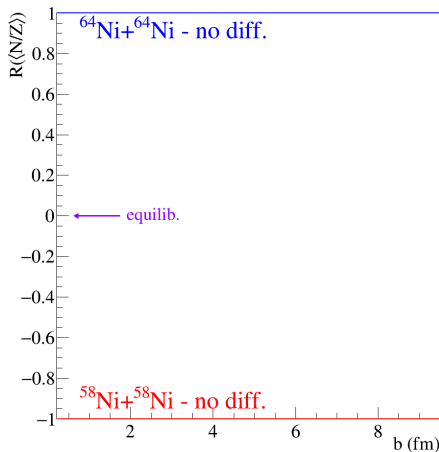
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$R(x_{AB}) = R(x_{BA}) \rightarrow$ full equilibration

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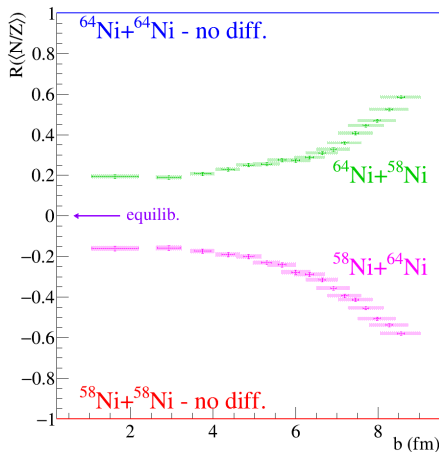
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Model-independent isospin transport ratio
 $R(\langle N/Z \rangle)$ for the QP remnant as a function
of the impact parameter b



Isospin analysis

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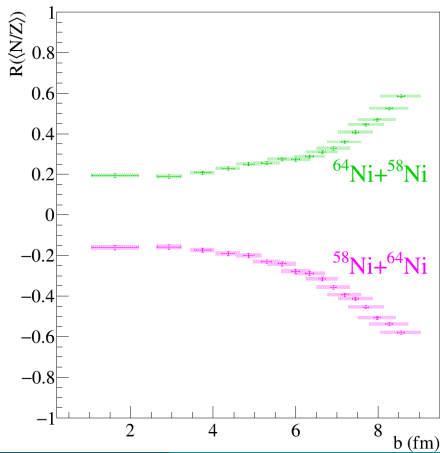
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Model-independent isospin transport ratio
 $R(\langle N/Z \rangle)$ for the QP remnant as a function
of the impact parameter b

Experimental result providing a reference for
comparison with theoretical predictions
from transport models.

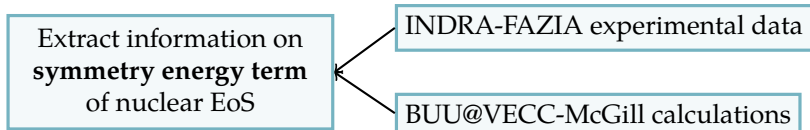
C. Ciampi *et al.*, *Phys. Rev. C* 111, 044601 (2025)



Model predictions

or

how do we extract a constraint on the symmetry energy?



Compare the experimental isospin transport ratio with the one predicted for *primary fragments* by the transport code:

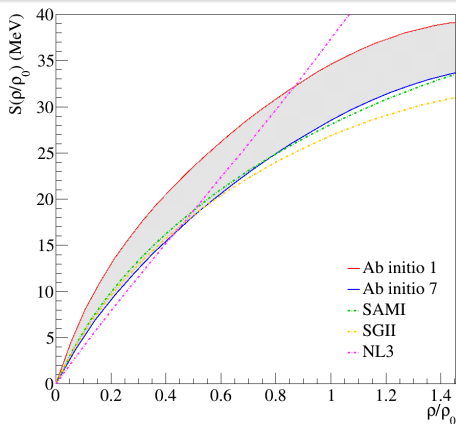
- Avoid spurious effect from afterburner coupling
- Exploit i.t.r. property of largely suppressing the effects of statistical deexcitation
A. Camaiani et al., Phys. Rev. C 102, 044607 (2020), S. Mallik et al., J. Phys. G 49, 015102 (2021)

Simulations carried out with the BUU@VECC-McGill transport code:

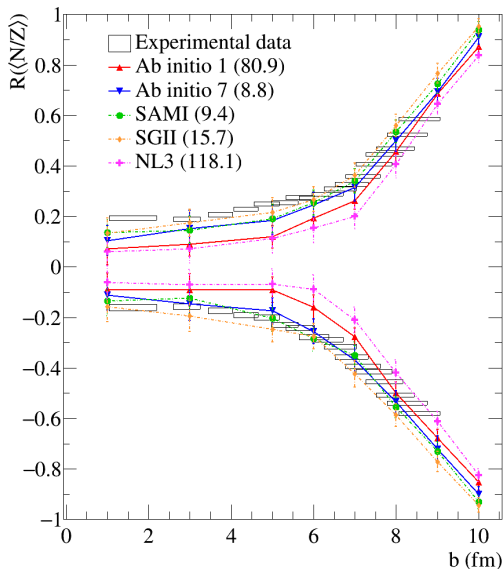
- BUU code, setting 100 test particles/nucleon
S. Mallik et al., Phys. Rev. C 91, 034616 (2015)
- 8 impact parameters for each reaction/parametrization
- 200 events for each impact parameter/reaction/parametrization
- Simulation run until 300fm/c (ITR convergence)
- Possibility to plug in and test different nEoS via metamodeling technique
J. Margueron et al., Phys. Rev. C 97, 025805 (2018)

NEoS parametrizations from *ab initio* and phenomenological approaches:

- *Ab initio*: two extreme χ -EFT interactions from C. Drischler *et al.*, Phys. Rev. C 93, 054314 (2016)
- Phenomenological approaches:
 - SAMI (Skyrme), X. Roca-Maza *et al.*, Phys. Rev. C 86, 031306(R) (2012)
 - SGII (Skyrme), Nguyen Van Giai *et al.*, Phys. Lett. B106, 379 (1981)
 - NL3 (RMF), G. A. Lalazissis *et al.*, Phys. Rev. C 55, 540 (1997)



	n_{sat} (fm $^{-3}$)	E_{sat} (MeV)	E_{sym} (MeV)	L_{sym} (MeV)	K_{sat} (MeV)	K_{sym} (MeV)
<i>Ab initio</i> 1	0.189	-16.92	34.57	48.5	241	224
<i>Ab initio</i> 7	0.140	-13.23	28.53	43.9	43.9	-144
SAMI	0.1587	-15.93	28.16	43.7	245	-120
SGII	0.1583	-15.59	26.83	37.6	215	-146
NL3	0.1480	-16.24	37.35	118.3	271	101



A few observations:

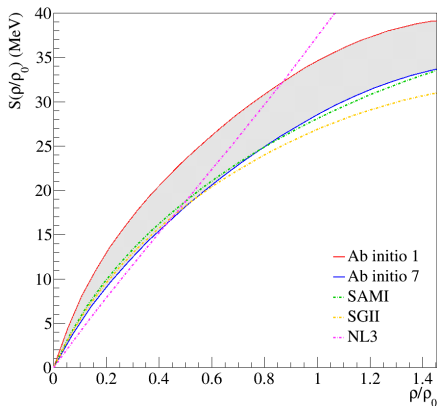
- SAMI vs *ab initio* 7:
similar symmetry energy, different isoscalar behavior. Similar ITR indicates we are probing E_{sym} .
- NL3, *ab initio* 1 $\rightarrow$ bad match:
largest E_{sym} values above $0.5\rho_0$ can be excluded.
- SAMI, SGII, *ab initio* 7 $\rightarrow$ good match:
they feature similar E_{sym} values above $0.5\rho_0$

χ^2 values calculated by interpolating between the simulated data points:

	AI1	AI7	SAMI	SGII	NL3
χ^2	80.9	8.8	9.4	15.7	118.1

Experimental-simulated data comparison

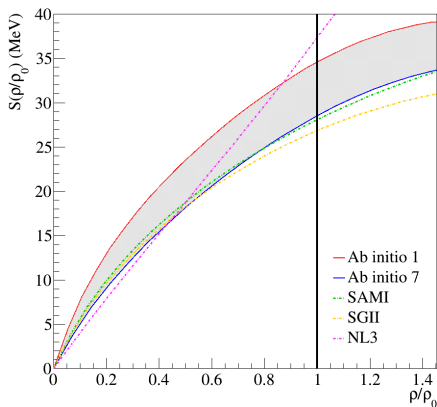
Extraction of approximate confidence regions in the $S - \rho$ plane



Experimental-simulated data comparison

Extraction of approximate confidence regions in the $S - \rho$ plane

At each value of ρ/ρ_0 :

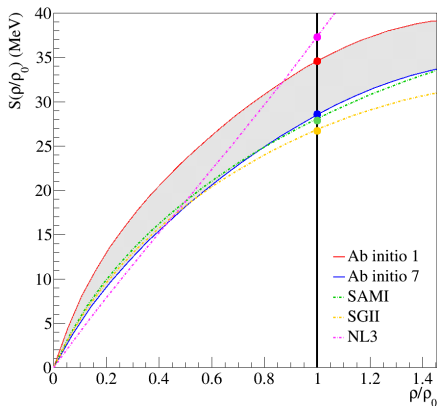


Experimental-simulated data comparison

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At each value of ρ/ρ_0 :

- Pick $S(\rho/\rho_0)$ values corresponding to the 5 nEoS parametrizations

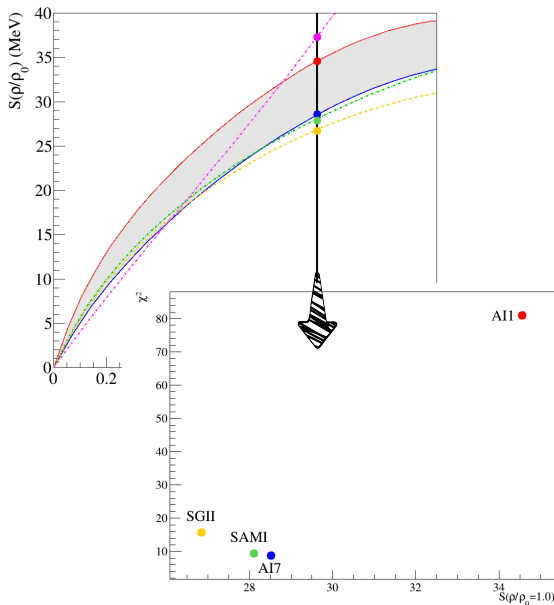


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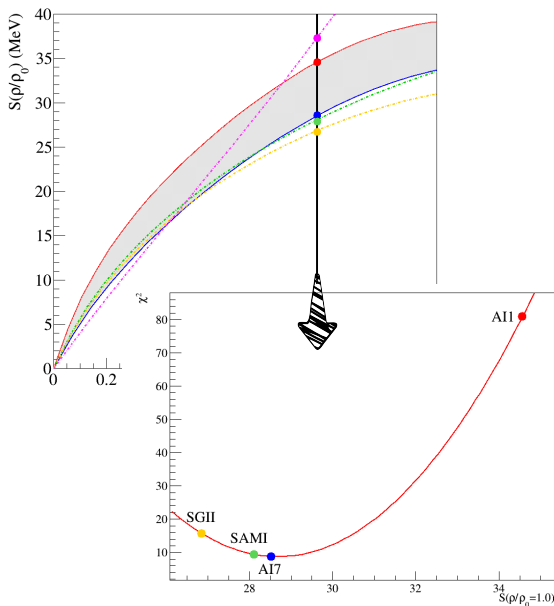


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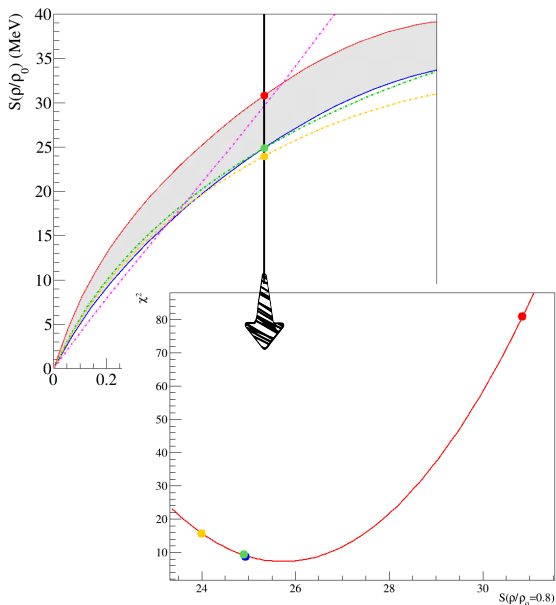
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Repeat along the ρ/ρ_0 axis.



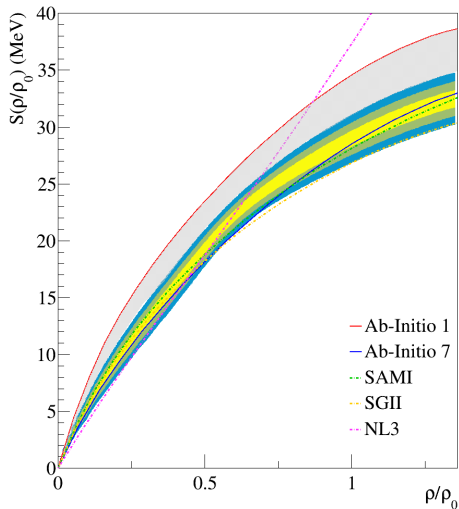
Experimental-simulated data comparison

Unweighted confidence regions

At each ρ/ρ_0 we thus define a likelihood along $S(\rho/\rho_0)$ as:

$$\mathcal{L}(S(\rho/\rho_0)) \propto e^{-\chi^2(S(\rho/\rho_0))/2}$$

→ extract the $N\sigma$ confidence intervals for $S(\rho/\rho_0)$
→ build approximate confidence region in the $S - \rho/\rho_0$ plane.



Experimental-simulated data comparison

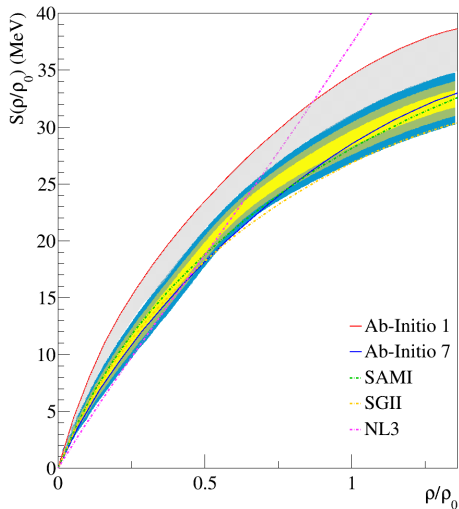
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For a meaningful constraint we take into account the density region probed by the isospin diffusion phenomenon
→ define a *weight function*

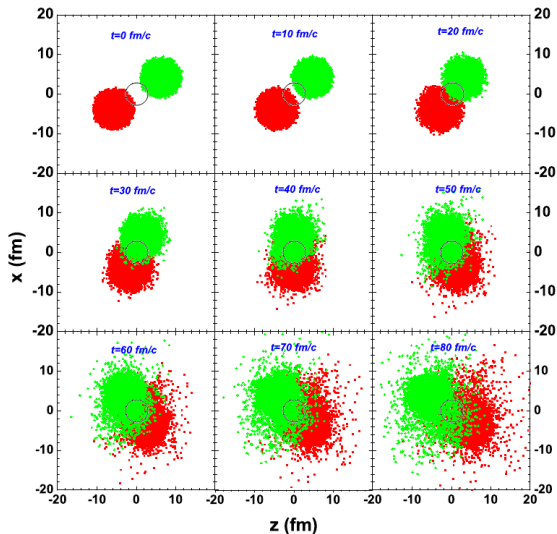


Sensitive density region

Isospin diffusion dynamics

Study of the evolution of **baryonic density** and **isospin current density** extracted from BUU@VECC-McGill

- *Ab initio* 7 nEoS
- 200 events for 4 selected impact parameters ($b = 3, 5, 7, 9$ fm)
- Quantities evaluated over spherical volume with $r = 3$ fm



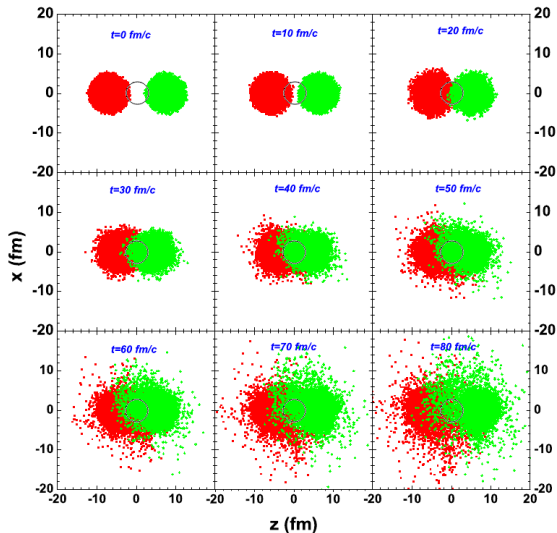
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Principal axis extracted at each timestep by diagonalizing the momentum of inertia tensor. Current densities evaluated in its reference frame.



Sensitive density region

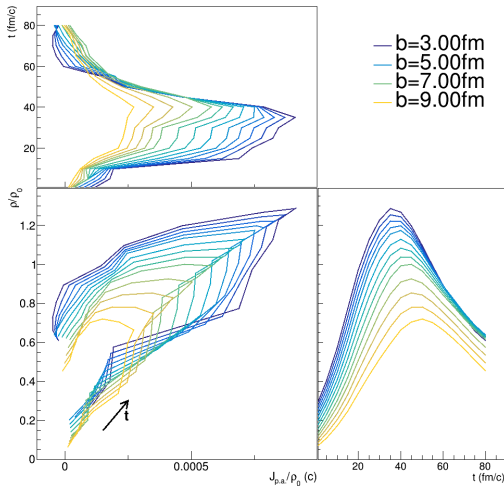
Isospin current and baryonic density

n/p current densities evaluated as:

$$\vec{j}_q^{(X)} = \frac{1}{V} \int_V d^3r \rho_q^{(X)}(\vec{r}) \vec{v}_q^{(X)}(\vec{r}).$$

Nucleon exchange determined by $\vec{j}_q^{(X)}$ components along the principal axis.

Net isospin current density obtained as the difference between neutron and proton current densities.



Sensitive density region

Isospin current and baryonic density

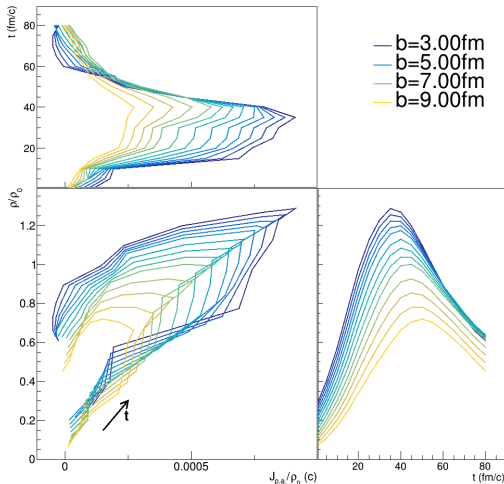
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Net isospin current density obtained as the difference between neutron and proton current densities.

N.B. maximum exchange found in the compression phase, when the highest densities are reached.



Sensitive density region

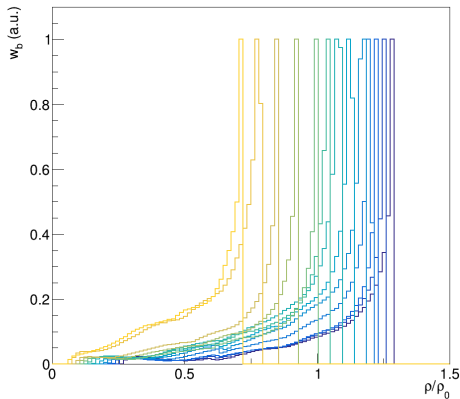
Weight function

Weight function for a given impact parameter $w_b(\rho/\rho_0)$:

built as baryonic density cumulated over time, weighted by the corresponding current density

$$w_b(\rho/\rho_0) = \int_{t_{\text{start}}}^{t_{\text{stop}}} j_{\text{p.a.}}(t) \delta(\rho/\rho_0 - \rho(t)/\rho_0) dt$$

→ evaluated between 0 and 80 fm/c



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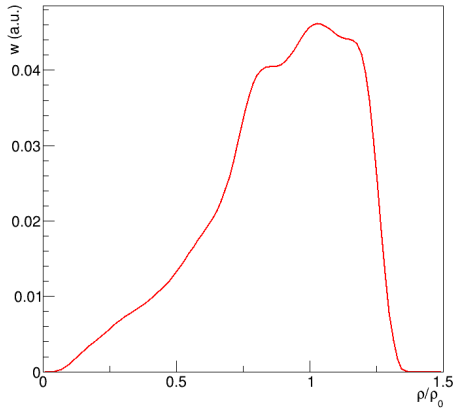
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→ evaluated between 0 and 80 fm/c

Global weight function $w(\rho/\rho_0)$:

$$w(\rho/\rho_0) = \int_{b_{\text{min}}}^{b_{\text{max}}} \tilde{w}_b(\rho/\rho_0) db$$

→ b interval between 3 and 9 fm

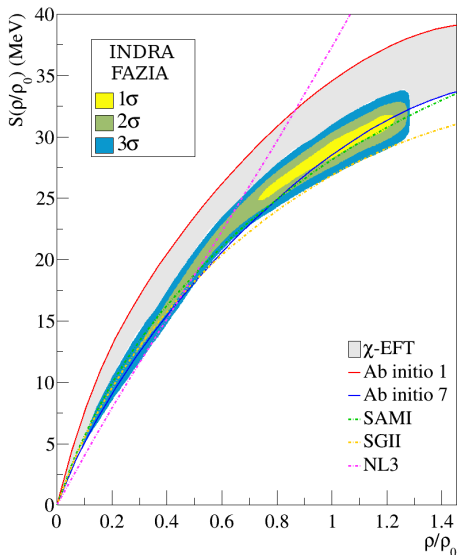


Constraint on the nuclear EoS

Weighted confidence regions

Confidence regions 1σ , 2σ , 3σ defined based on weighted $\mathcal{L}(S(\rho/\rho_0))$:

- Sensitivity close to saturation - anticipated by NL3 rejection, different from SAMI, SGII, AI7 for $\rho > 0.5\rho_0$



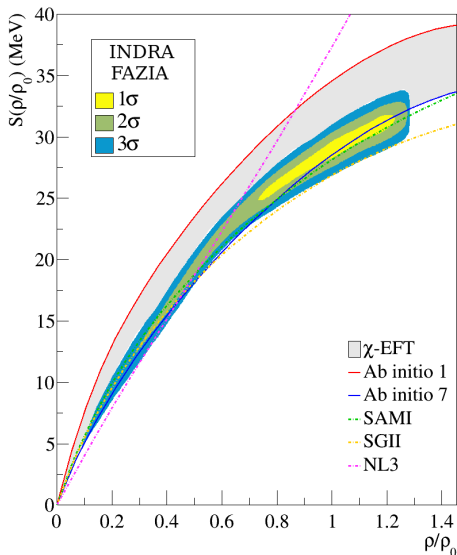
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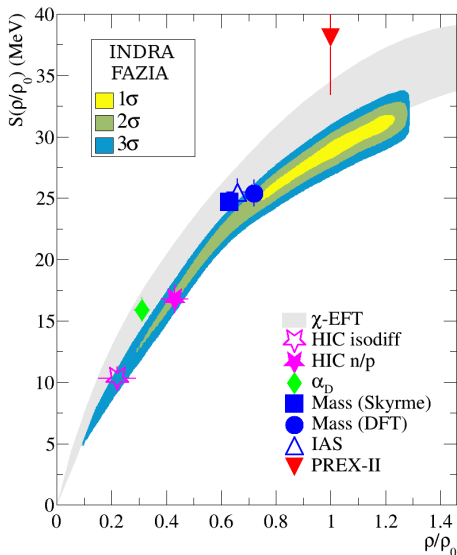
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- Good agreement with χ -EFT - towards the softer side of its uncertainty band
- Good agreement with isospin diffusion data in Sn+Sn collisions - (but we declare a different ρ sensitivity)

W. G. Lynch *et al.*, Phys. Lett. B 830, 137098 (2022).
M. B. Tsang *et al.*, Phys. Rev. Lett. 102, 122701 (2009).
P. Morfouace *et al.*, Phys. Lett. B 799, 135045 (2019).
Z. Zhang *et al.*, Phys. Rev. C 92, 031301R (2015).
M. Kortelainen *et al.*, Phys. Rev. C 85, 024304 (2012).
P. Danielewicz *et al.*, Nucl. Phys. A 958, 147 (2017).
B. T. Reed *et al.*, Phys. Rev. Lett. 126, 172503 (2021).



C. Ciampi, S. Mallik *et al.*, submitted (2025).

Summary and conclusions

A new nuclear EoS constraint from isospin diffusion data:

- Comparison of model-independent experimentally-measured isospin transport ratio in $^{58,64}\text{Ni}+^{58,64}\text{Ni}$ collisions at 32 MeV/nucleon with the predictions of BUU@VECC-McGill transport model.
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Thank you!

C. Ciampi, S. Mallik, F. Gulminelli, D. Gruyer, J. D. Frankland, N. Le Neindre, R. Bougault, A. Chbihi, L. Baldesi, S. Barlini, B. Borderie, A. Camaiani, G. Casini, I. Dekhissi, J. A. Dueñas, Q. Fable, F. Gramegna, M. Henri, B. Hong, S. Kim, A. Kordyasz, T. Kozik, I. Lombardo, O. Lopez, T. Marchi, S. H. Nam, J. Park, M. Pârlog, G. Pasquali, S. Piantelli, G. Poggi, S. Valdré, G. Verde, E. Vient



Backup slides

Impact parameter reconstruction

Detailed structure of the method

Given a centrality observable X , its inclusive distribution $P(X)$ can be expressed as:

$$P(X) = \int_0^\infty P(X, b) db = \int_0^\infty P(b) P(X|b) db = \int_0^1 P(X|c_b) dc_b$$

where a change of variables is applied, introducing the centrality $c_b \equiv \int_0^b P(b') db'$ and exploiting that $P(c_b) = 1$.

Key step: model the $P(X|c_b)$ and extract its parameters by fitting the experimental $P(X)$
 X assumes positive values $\rightarrow$ non-negative gamma distribution as fluctuation kernel:

$$P(X|c_b) = \frac{1}{\Gamma(k)\theta^k} X^{k-1} e^{-X/\theta} \quad \text{where } \bar{X} = k\theta \text{ and } \sigma_X = \sqrt{k}\theta$$

where k and θ generally evolve with centrality. For them we assume:

- $k(c_b) = k_{\max}[1 - c_b^\alpha]^\gamma + k_{\min}$, where α , γ , $k_{\min}$ and $k_{\max}$ are parameters of the fit
- θ independent of centrality (problem is underconstrained) $\rightarrow \theta$ is a fit parameter

Once the $P(X|c_b)$ is determined, one obtains:

$$P(c_b|x_1 \leq X \leq x_2) = \frac{\int_{x_1}^{x_2} P(c_b, X) dX}{\int_{x_1}^{x_2} P(X) dX} = \frac{\int_{x_1}^{x_2} P(X|c_b) dX}{\int_{x_1}^{x_2} P(X) dX}$$

and by changing back the variable: $P(b|x_1 \leq X \leq x_2) = P(b) P(c_b(b)|x_1 \leq X \leq x_2)$

Reaction centrality

Assessing the impact parameter in experimental data

Physical information is obtained by comparing experimental results and transport model simulations *assuming the same conditions*.

Impact parameter: experimentally, it can be only deduced from observables such as multiplicities, transverse energies, flow angle...



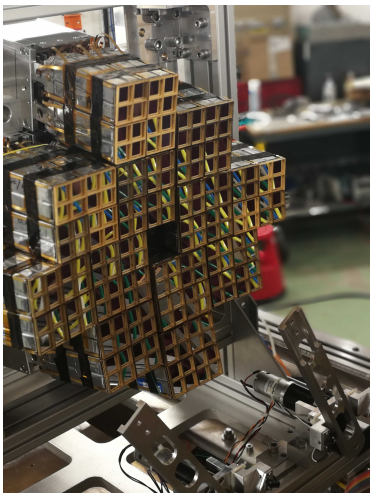
All physics observables are affected by **intrinsic fluctuations** associated with the underlying processes.

Such fluctuations can limit the accuracy in treating centrality and bias the comparisons with simulated data.

L. Li et al., PRC 97,044606 (2018), G. Q. Zhang et al., PRC 84, 034612 (2011)

Different approaches for reaction centrality characterization:

- Sharp cutoff approximation C. Cavata et al., Phys. Rev. C 42, 1760 (1990)
- Machine learning algorithms trained on simulations
F. Li et al., PRC 104,034608 (2021), F. Haddad et al., PRC 55, 1371 (1997)
- **Model-independent** method to reconstruct impact parameter distributions → includes **fluctuations**
J. D. Frankland et al., PRC104, 034609 (2021), R. Rogly et al., PRC98, 024902 (2018)



FAZIA (*Forward-angle A and Z Identification Array*): optimal ion identification in the Fermi energy domain.

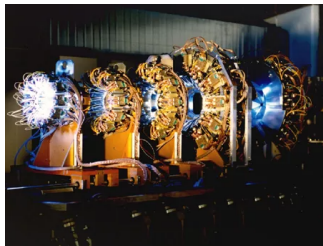
- Result of R&D activities to refine:
 - detector performance
 - digital treatment of signals
- Basic module: **block**, consisting of 16 three stage **telescopes** ($2 \times 2 \text{ cm}^2$ active area):
 - Si1 300 μm thick
 - Si2 500 μm thick
 - CsI(Tl) 10cm thick
- + read-out electronics for all telescopes.
- Identification techniques: ΔE -E / PSA
 - Charge discrimination tested up to $Z \sim 55$
 - Mass discrimination up to $Z \sim 25$ / $Z \sim 22$

R. Bougault et al., *Eur. Phys. J. A* 50, 47 (2014)

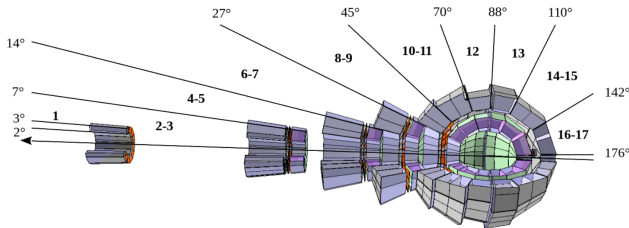
S. Valdré et al., *NIMA* 930, 27 (2019)

INDRA (*Identification de Noyaux et Détection avec Résolutions Accrues*): highly segmented array for detection and identification of charged products of heavy ion collisions at intermediate energies ($10 < E < 100$ AMeV).

- Original configuration of 17 rings:
 - 1: Phoswich detectors
 - 2-9: Ionisation ch. + Si + CsI(Tl)
 - 10-17: Ionisation ch. + CsI(Tl)
- Charge discrimination up to uranium, mass discrimination up to $Z \sim 4$
 → Electronics upgrade (2020): now up to $Z \sim 10$
[J. D. Frankland et al., Nuovo Cim. C 45, 43 \(2022\)](#)



- Large solid angle coverage (90%) with high granularity (336 modules)



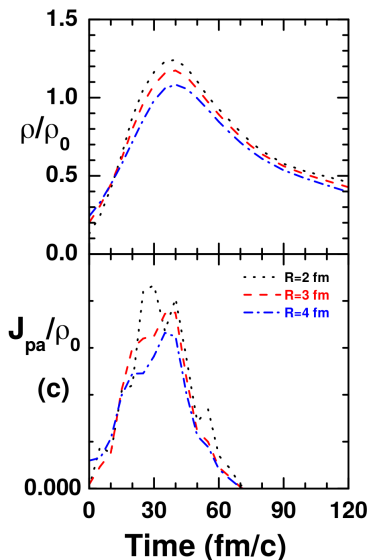
Sensitive density region

Radius of the examination sphere

The values of baryonic density and isospin current density depend on the radius of the spherical volume used for their evaluation.

- For $r > 3$ fm the sphere extends beyond the neck region (including empty areas), which reduces both the density and the current density
- For $r < 3$ fm, the finite number of test particles within the sphere introduces strong statistical fluctuations that affect the current evaluation

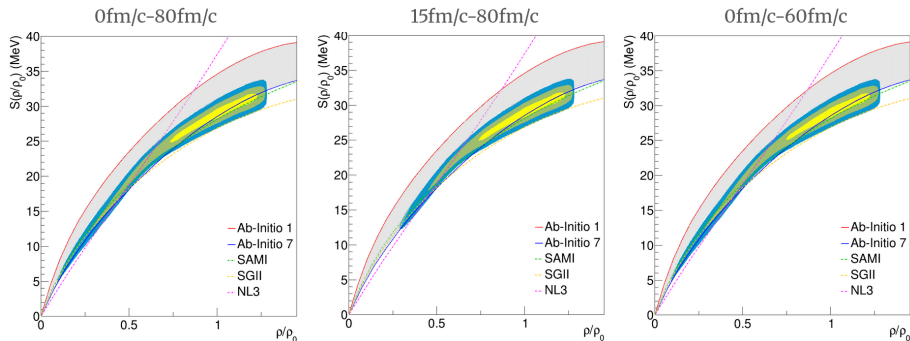
The radius equal to 3 fm has been chosen as the optimal value, balancing these competing effects.



Constraint on the nuclear EoS

Weighted confidence regions - dependence on t intervals

The confidence regions slightly depend on the choice of b and t intervals for the weight function, but our observations are stable against these arbitrary choices.



e.g., varying the time interval for the weight function evaluation essentially changes the contribution from the low density tail.

Constraint on the nuclear EoS

Bayesian framework

The proposed procedure corresponds to a **Bayesian parameter estimation**

- of the value of the symmetry energy at different densities $S_i = S(\rho_i)$
- **not** of the nuclear matter parameters $S_i \propto \frac{d^i E_{\text{sym}}}{d\rho^i}(\rho_0)$ (derivatives at saturation)

The posterior marginalized distribution of S_i is given by:

$$p(S_i) = \frac{1}{N} \sum_{m=1}^N p(m|\vec{d}) \delta \left(S_i - S_i^{(m)} \right) \quad \text{where} \quad p(m|\vec{d}) = \mathcal{N} p^{\text{prior}}(m) p(\vec{d}|m)$$

- the set of data $\vec{d} = \{d_1, \dots, d_K\}$ is the set of *ITR* data points $d_j = R(b_j)$
- we consider a gaussian likelihood $p(\vec{d}|m) \equiv \mathcal{L}_{\vec{d}}(m) = \exp(-\chi_m^2/2)$ with $\chi_m^2 = \sum_{j=1}^K (d_j - d_j^{(m)})^2 / \sigma_{jm}^2$ and $\sigma_{jm}^2 = \sigma_{\text{exp},j}^2 + \sigma_{\text{th},jm}^2$
- we take a flat constant prior $p^{\text{prior}}(S(\rho_i))$ for each density ρ_i , to which we apply the weight function $p_i^{\text{prior}} = w(\rho_i/\rho_0)$, as a confidence factor on the density axis.
 $\Rightarrow$ globally $p_i(m|\text{ITR}) = w(\rho_i/\rho_0) \exp(-\chi_m^2/2)$