Transport Model Evaluation Project (TMEP): Status and Perspectives

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ORIGINS Excellence Cluster

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On behalf of the Transport Model Evaluation Project (TMEP) Collaboration



Transport model comparison studies of intermediate-energy

Review

heavy-ion collisions

Outline:

- Motivation: Increase impact of HIC studies on determination of EOS
- TMEP strategy: Comparison of transport codes under controlled conditions, similar input -> similar output(?)
- box calculations: test individual ingredients, comparison to exact results,
- HIC: open systems, much less agreement, explanations, but no solid error estimates
- Intermediate conclusions: lessons learned and desirables for more robust conclusions
- Uncertainty quantification: Bayesian inference with many codes, BMA (Bayesian model averaging)
- Alternative: collaboration to construct modular common code

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Motivation: Increase impact of HIC in EOS studies

Importance of intermediate-energy heavy-ion collisions for the exloration of equation-of-state (EOS)

→ filling the gap between information from nuclear structure ($\rho \le \rho_0$) and neutron star observations ($\rho \ge 2.5 \rho_0$)

-> can make a contribution to constrain EOS

-> only astrophysics

-> xEFT, Astro and HICs (Huth, et al., Nature 602 (22)

-> structure, HICs and Astro (C.Y.Tsang, et al., Nat.Astro 8 (24)

model dependence of HIC results:

SπRIT data, Sn+Sn, 270 MeV/A, Jhang, et al., PLB 813 (21) predictions: best physics model of each code

large spread of results sensitivity to symmetry energy (size of boxes) relatively small



Transport theory: kinetic equation Boltzmann-Uehling-Uhlenbeck (BUU)

$$\frac{df}{dt} = \frac{\partial f}{\partial t} + \frac{\vec{p}}{m} \vec{\nabla}^{(r)} f - (\vec{\nabla}^{(r)} U(r, p) \vec{\nabla}^{(p)} + \vec{\nabla}^{(p)} U(r, p) \vec{\nabla}^{(r)}) f(\vec{r}, \vec{p}; t) = I_{coll} + \delta I_{fluc}$$

$$I_{coll} = \int d\vec{p}_2 d\vec{p}_1 d\vec{p}_2, v_2, \sigma_{12}^{in-med}(\Omega) (2\pi)^3 \delta(p_1 + p_2 - p_1 - p_2) \Big[f_1 f_2 f_1 f_2 - f_1 f_2 f_1 f_2 \Big]$$

$$\vec{f}_i := (1 - f_i) \text{ Pauli blocking factors,}$$

Physical model: mean field->EOS, in-medium xsec inelastic collisions

 $I_{coll} \rightarrow I_{el} (NN \leftrightarrow NN) + I_{inel} (NN \leftrightarrow N\Delta) + I_{decay} (\Delta \leftrightarrow N\pi)$

Two main reasons of model dependence: 1) fluctuations, 2) simulation strategies

1) Two families, depending on representation of phase space density $f(\vec{r}, \vec{p}; t)$ **BUU** phase space density represented by test particles (TP) $f(\vec{r}, \vec{p}; t) = \sum_{TP \ i} \delta(\vec{r} - \vec{r}_i(t)) \ \tilde{\delta}(\vec{p} - \vec{p}_i(t))$, deterministic and exact for #TP $\rightarrow \infty$; introduce fluctuations explicitely, Boltzmann-Langevin, add term δI_{fluc}

QMD product of wave packets in coordinate space $f(\vec{r}, \vec{p}; t) = \left(\frac{\hbar}{\sqrt{L}}\right)^3 \sum_i \exp\left[-\frac{(\vec{r} - \vec{R}_i(t))^2}{2L}\right] \delta(\vec{p} - \vec{P}_i(t))$ fluctuations on classical level by ansatz, fluctuations parametrized by width parameter L, "events"

-> Fluctuations influence many aspects of simulation

Transport theory: kinetic equation Boltzmann-Uehling-Uhlenbeck (BUU)

$$\begin{aligned} \frac{df}{dt} &= \frac{\partial f}{\partial t} + \frac{\vec{p}}{m} \vec{\nabla}^{(r)} f - (\vec{\nabla}^{(r)} U(r, p) \vec{\nabla}^{(p)} + \vec{\nabla}^{(p)} U(r, p) \vec{\nabla}^{(r)}) f(\vec{r}, \vec{p}; t) = I_{coll} \\ I_{coll} &= \int d\vec{p}_2 d\vec{p}_1 d\vec{p}_2, v_2 (\sigma_{12}^{in-med}(\Omega) (2\pi)^3 \delta(p_1 + p_2 - p_1 - p_2) \Big[f_1 f_2 (f_1 f_2 - f_1 f_2 (f_1 f_2)) \Big] \\ \bar{f}_i &:= (1 - f_i) \text{ Pauli blocking factors,} \end{aligned}$$

Physical model: mean field->EOS, in-medium xsec inelastic collisions

Two main reasons of model dependence: 1) fluctuations, 2) simulation strategies

2) Transport equation solved by simulations, involves strategies mf dynamics: - Hamiltonan equations-of-motion (for TP or nucleons) finite-size TP, use of lattice Hamiltonian, non-linear density functionals $\propto \rho^{\sigma}$, often approximated in QMD

collision term: - stochastic two-body collisions geometric coll Kriterien $d < \sqrt{\sigma(\sqrt{s})/\pi} \leftrightarrow$ local thermal equilibrium blocking: need for averaging, coarse graining, surface **TMEP**: compare different transport models und controlled conditions, same physical model and similar calculational parameters -> similar results??

two modes of comparisons:

- box calculations, periodic boundary conditions, simulates infinite nuclear matter, study individually different part of physical model, compare to exact limits
- 2, full heavy-ion collisions (HICs), compare codes between each other,

convergence or amount of disagreement







QMD event

Box 1: Mean field evolution (M. Colonna, et al., PRC104 (2021))





Box2: Collision intergral (only nucleons, with Pauli blocking, initialize at T=5 MeV) (YX. Zhang, et al., PRC 97 (2018))

Collision rates, compared to exact result:

Systematic difference between BUU and QMD results





Reason: Fluctuations in Pauli blocking factor (1-f) exact: red

average: blue

effective (enforce f≤1): black

generally underblocking (black \leftrightarrow red)



Understanding differences:

correlations between collisions (non-Markovian behavior) geometric criterion not optimal, statistical criteria better strategies in handling sequence of elastic and inelastic collisions Cancel rather well in ratios



Box4: Collision integral with momentum-dependent interactions (D. Cozma, et al., in preparation)

threshold shift in inelastic collisions with momentum dependent mean fields

4) full_mdi_th: - mean field (K₀=230 MeV; m*=0.70; Δm*_m/m_N=-0.33δ; S₀=32 MeV; L=60 MeV); threshold effects included - initialization uses effective masses (Boltzmann T=60 MeV) - results: dcQMD, RVUU, sJAM 20 12 π^0 π^+ π 10 16 8 12Z 6 8 4 dcOMD – RVUU CT33222227437-8741 - sJAM 4 2 dcQMD RVUU SJAM 0 100 0 50 100 0 500 50 100 0 50 100 0 50 100 150 $t \, [fm/c]$ $t \, [fm/c]$ $t \, [fm/c]$ $t \, [fm/c]$ $t \,[\mathrm{fm/c}]$

dcQMD RVUU sJAM

solid lines: with threshold effect dashed lines: without

thin dashed line: exact result (thermal model)

rather good agreement between codes, but some deviations (being investigated) demonstrates importance of considering threshold shift

learn much about simulations,

- comparison to exact results: absolute measure of quality, but strategies optimized for box, may not be equally good for HIC
- largely understand differences between codes
- recommend optimal strategies, e.g.
 - geometric criterion to determine next collisior $d < \sqrt{\sigma(\sqrt{s})/\pi}$ may induce non-Markovian effects choose next collisions by statistical criteria (only implemented in some codes)



- affect clustering (not yet tested)

Now look at full heavy-ion collisions (HIC)



HIC: Sn+Sn@270 MeV/A (J. Xu, et al., PRC 109, 044609 (2024)) similar to Au+Au@100,400 MeV/A + pion observables controlled input: common initializ., simple mom.-indep. EOS, σ_{el} =const, NN \leftrightarrow N Δ , $\Delta \leftrightarrow$ N π

with PB w/o cou
with PB with cou
w/o PB w/o cou
w/o PB with cou



Conclusion: differences in the evolution of the system (caused here by approx. in averaging of force) leads to difference in pion observables

HIC are open systems: small differences can propagate and lead to larger final results, ingrediens interact (unlike in box) no exact results, comparison between codes, not to experiment, since (here) models are simplistic

But differences can be understood

Look closer at charged pion ratio, assumed to be a good probe for the symmetry energy



Lessons and intermediate conclusions from Code comparisons

- Code comparisons are very interesting and teach us a lot about simulation physics. in most cases the differences can be explained, but the codes often cannot be made to converge sufficiently to determine physical parameters, like J, L, K_{sym}, etc
- 2. Difficult to assign an error to transport results as a whole, i.e. to determine the uncertainty due to the model dependence, The mean and variance in code comparisons do not represent a reasonable error estimate.
- 3. There are no exact results for HIC and code-to-code comparisons do not decide about the reliability of a code. But there is experiment. One can estimate the reliability of a code on it ability to describe a relevant set of experimental data (relevant to the physical question asked).

4. chain of observables



5. Note, that agreement with experiment is not sufficient that code is "correct", but agreement with a relevant and sufficiently wide set of data does make this more probable

What is necessary for this program?

- a realistic model, containing all necessary ingredients for the physical question, (i.e. momentum dependence of isoscalar and isovector forces, threshold shift effects, energy conservation),
- 2. robust computational strategies, such as converged propagation of particles, correct treatment of coarse graining, etc.
- 3. realistic fluctuations (strong dependence of simulation results on the amount of fluctuations),

-> make the BUU and QMD approaches compatible:

Boltzmann-Vlasov (BUU) + fluctuation term \rightarrow Boltzmann-Langevin (approximations SMF, BLOB) Molecular dynamics (QMD) -> heuristic fluctuations and correlations, depending on a parameter (Δx) need to agree !

4. clustering and correlation

influences directly emission of light clusters (LC) and as seeds intermediate mass fragments (IMF) but indirectly may influence the evoluiton of the system, and observables as pion production



Method to implement this program: Bayesian Model Averaging (BMA).

Bayesian model inference is standard for one model, where the likelyhod is taken from this model. What about obtaining the uncertainty for analyses with several models?

-> Bayesian model mixing with weights, which are determined from the ability of each model to describe a set data.

example: from determination of Nuclear Symmetry Energy from nuclear structure: see talk of Mengying Qiu later

Could be applied to transport analyses of HIC to constrain the Nuclear Symmetry Energy (NSE).

- 1. Inferenz of NSE from Bayesian model analysis for several codes from isospin-sensitive observable.
- 2, Use as weights for model averaging ability to describe a relevant and sufficiently large of data on nucleon observables, e.g., stopping, flow, nucleon emission, to make sure, that reaction evolution is sufficiently well described.
- 3. Only then inferences from secondary observables, like pion observables, can be reasonably believed, and the averaged probability distribution gives the uncertainty of transport analyses as a whole.
- 4. Bayesian model mixing (P. Giuliani): essentially th same, except one wants to make predictions; Uses PCA (principal component analysis) to use only essentially different models. Also useful here.
- 5. Possibility to also use different sets of data from structure, HIC and astrophysics

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Alternatives:

- 1. Many codes, BMM
- 2. Construct commom code in a modular approach



EOS of dense matter from HIC: extract a rather simple quantity (the bulk EOS of infinite matter) from a complicated non-equilibrium process.

However, it is the only way to obtain this information in an intermediate density regions above saturation.

bonus: obtain information not only on bulk EOS, but detailed information, like the composition, response, phase transitions, etc.

mandatory, to make use of the impressive (and costly) development of facilities and detectors

TMEP: Estimate and reduce the systematical theoretical error (the model dependence) of conclusions of transport model simulations to extract information on the EOS from heavy-ion collisions.

method: code comparisons of HIC under controlled conditions.

difficult to achieve sufficient convergence of codes to make deductions on small effects , like the nuclear symmetry energy.

Possible ways to make progress:

uncertainty quantification of transport analyses using Bayesian Model averaing

* code development constructing a common code in a modular approach in a dedicated collaboration probably funded collaboration needed π

Thank you for your attention

backup



Bayesian model averaging: Example from nuclear structure.

"Bayesian model averaging for nuclear symmetry energy from effective proton-neutron chemical potential difference of neutron-rich nuclei", Mengying Qiu, Bao-Jun Cai Lie-Wen Chen, Cen-Xi Yuan, Zhen Zhang, PLB 849 (24) 183435 "to extract the symmetry energy around $2\rho_0/3$ from the measured $\Delta\mu^*_{pn}$ of 5 doubly magic nuclei 48Ca, 68Ni, 88Sr, 132Sn and 208Pb"

using two models: Skyrme and RMF, each with a number of parameters.

Correlation coefficient between $E_{sym}(\rho)$ calculated for a sampling of the model space for each model and the $\Delta \mu^*_{pn}$:

-> strongest correlation with E_{sym} at $\frac{2}{3}\rho_0$

Perform Bayesian inference for each model and obtain posterior probabity distribution for Esym(2/3 ρ_0): different for the two models: mean and width. Now average the two model with weights given by the evidence of fitting the chemical potential differences by each model Mi

 $p(\mathbf{y} | \mathcal{M}_i) = \int p(\mathbf{y} | \theta_i, \sigma_i, \mathcal{M}_i) \pi(\theta, \sigma | \mathcal{M}_i) d\theta_i d\sigma_i.$ likelyhood prior

(Skyrme has a higher evidence, because the correlation is tighter, faxtor 23.)

→Obtain probability distribution of for anylsis of two models (solid line) is Skyrme the better model? No, because evidence based on very small data set. but if data set is enlarged, e.g. BE, radii, sp energies, the weight would be more meaningful



