

# Transport Overview

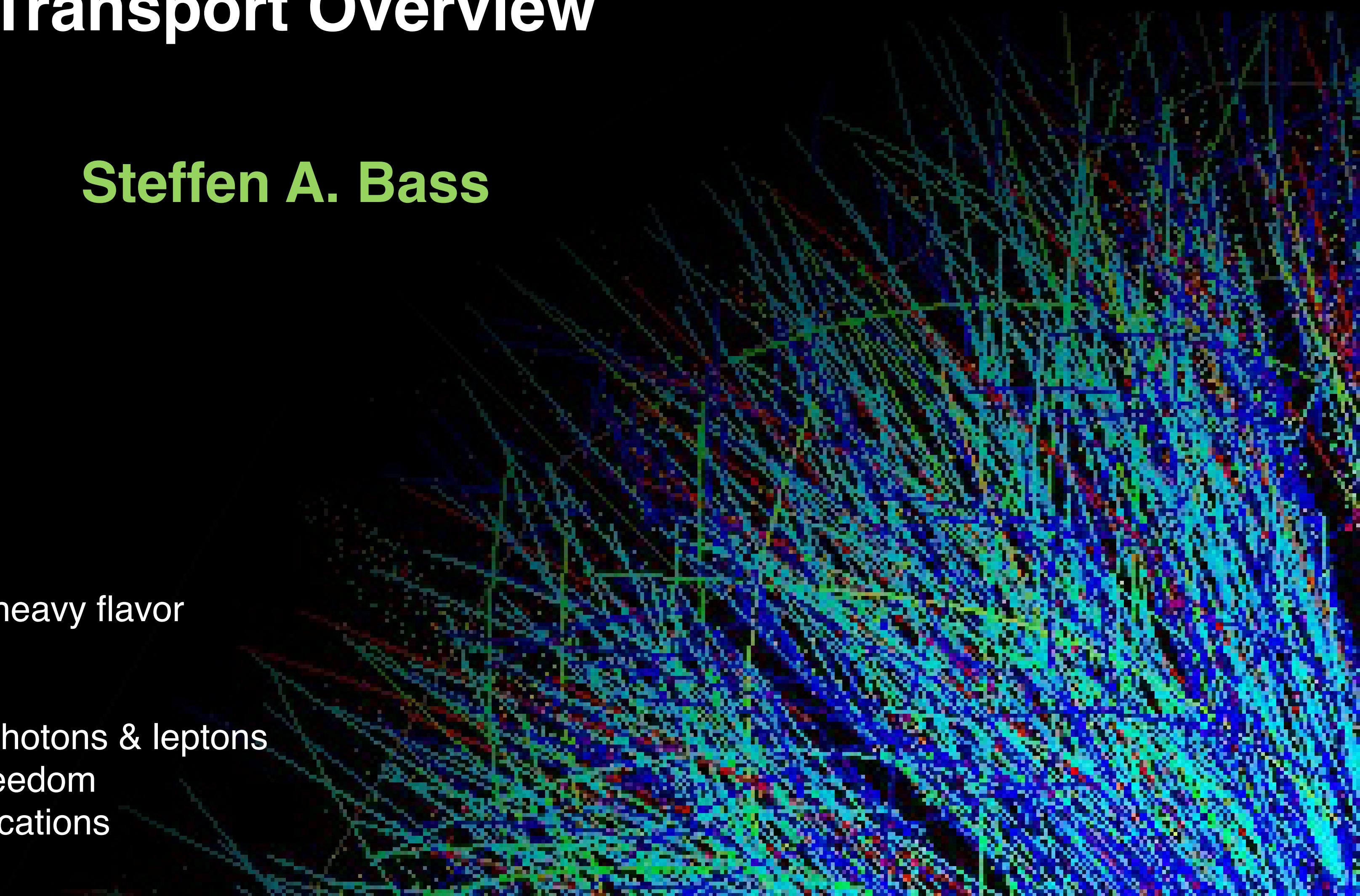
Steffen A. Bass

emphasis on:

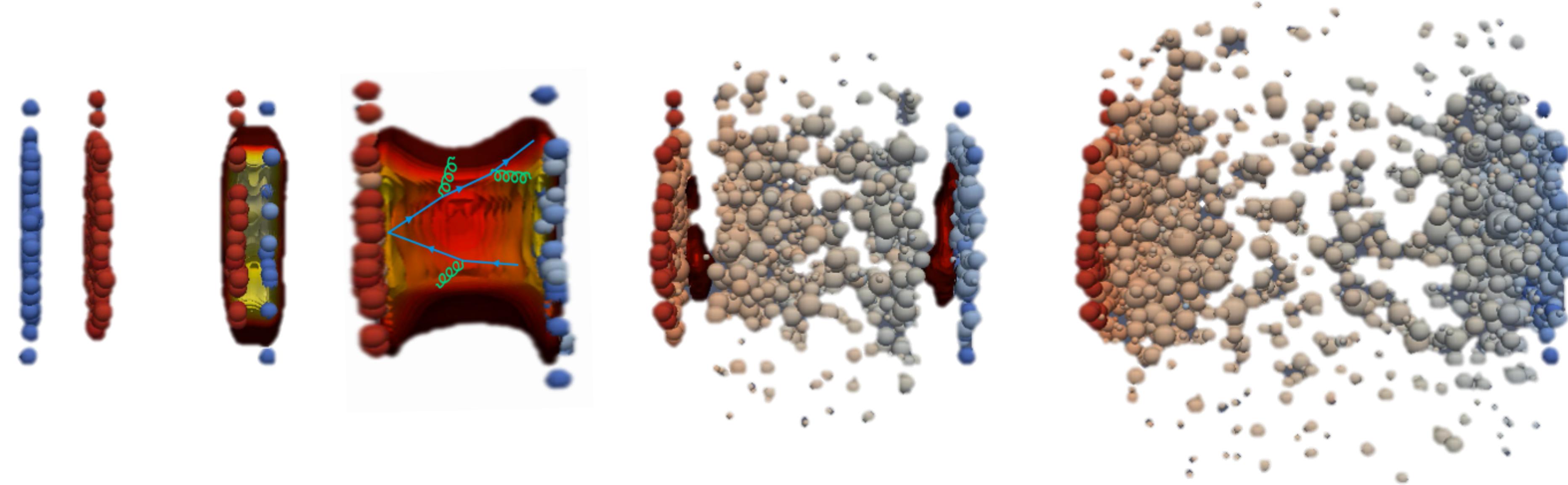
- transport relevant to strangeness and open heavy flavor

not covered:

- transport for (heavy flavor) jets, quarkonia, photons & leptons
- initial state phenomena & spin degrees of freedom
- general soft matter transport w/o flavor applications



# What is a transport model?



- A transport model describes the time-evolution of the collision, utilizing a set of physics processes that can be derived or approximated from some underlying theory.
- The model is capable of predicting quantities that can be measured in experiments, thus allowing for the testing of its underlying assumptions via a comparison to data and gaining physics insight from such a comparison
- Transport models are also utilized to gain understanding on processes not directly accessible to experimental observation

# Flavors of Transport Models:

## microscopic transport models based on the Boltzmann Equation:

- transport of a system of microscopic particles
- all interactions are based on **binary scattering**

$$\left[ \frac{\partial}{\partial t} + \frac{\vec{p}}{E} \times \frac{\partial}{\partial \vec{r}} \right] f_1(\vec{p}, \vec{r}, t) = \sum_{processes} C(\vec{p}, \vec{r}, t)$$

## diffusive transport models based on the Langevin Equation:

- transport of a system of microscopic particles in a thermal medium
- interactions contain a **drag term** related to the properties of the medium and a **noise term** representing random collisions

$$\vec{p}(t + \Delta t) = \vec{p}(t) - \frac{\kappa}{2T} \vec{v} \cdot \Delta t + \vec{\xi}(t) \Delta t$$

## (viscous) relativistic fluid dynamics:

- transport of macroscopic degrees of freedom
- based on conservation laws:

$$\begin{aligned} \partial_\mu T^{\mu\nu} &= 0 \\ T_{ik} &= \varepsilon u_i u_k + P (\delta_{ik} + u_i u_k) \\ &- \eta \left( \nabla_i u_k + \nabla_k u_i - \frac{2}{3} \delta_{ik} \nabla \cdot u \right) \\ &+ \varsigma \delta_{ik} \nabla \cdot u \end{aligned}$$

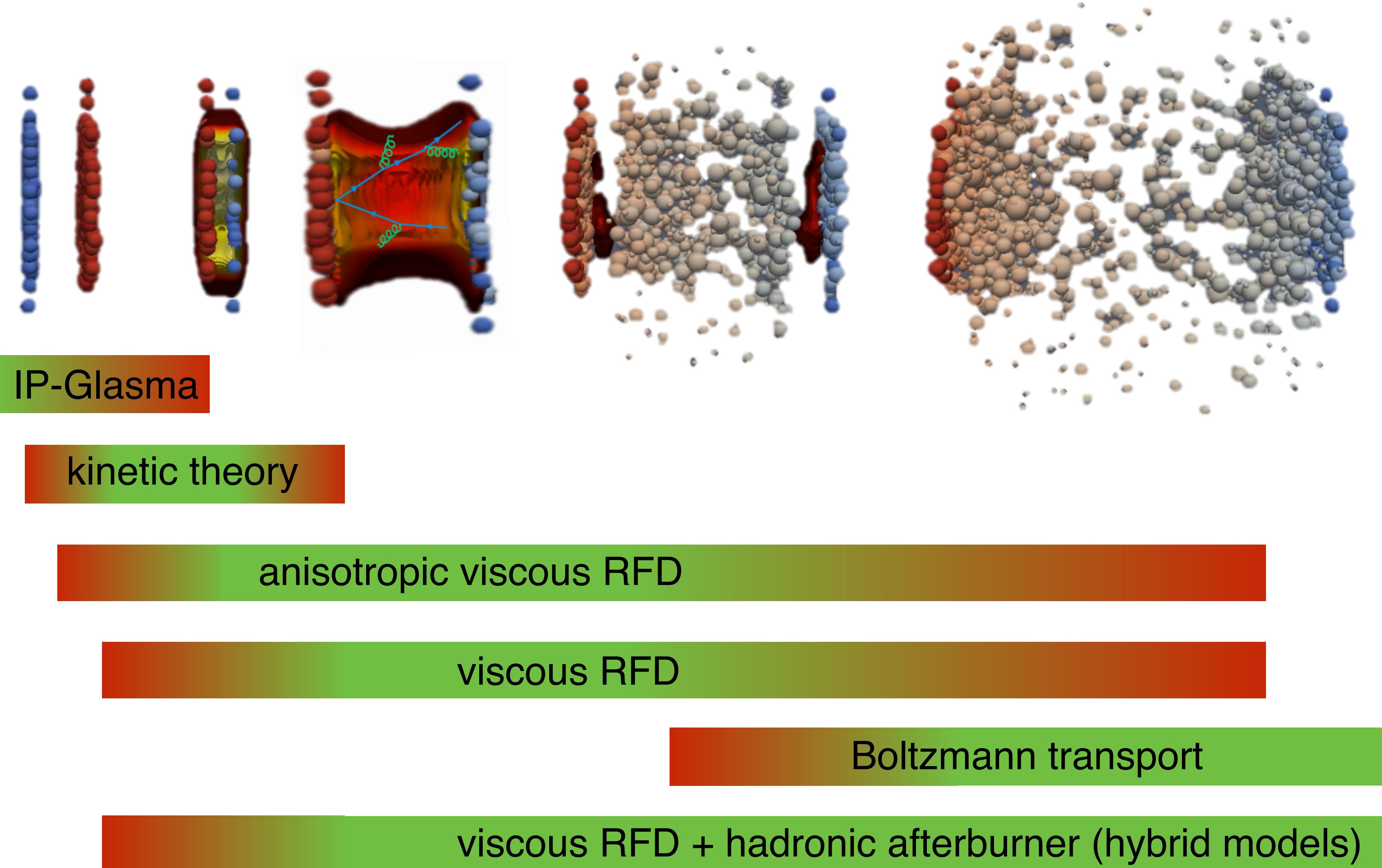
(plus an additional 9 eqns. for dissipative flows)

## hybrid transport models:

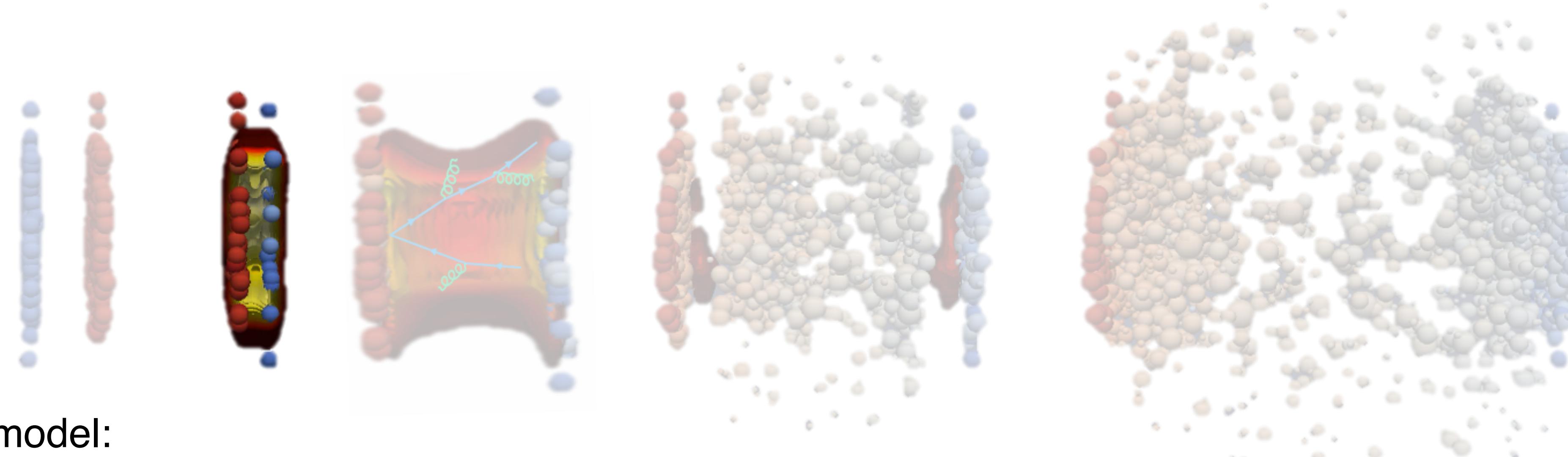
- combine microscopic & macroscopic degrees of freedom
- current state of the art for RHIC modeling

Each transport model relies on roughly a dozen physics parameters to describe the time-evolution of the collision and its final state. These physics parameters act as a representation of the information we wish to extract from experiment.

# Applicability ranges for transport @ high energy



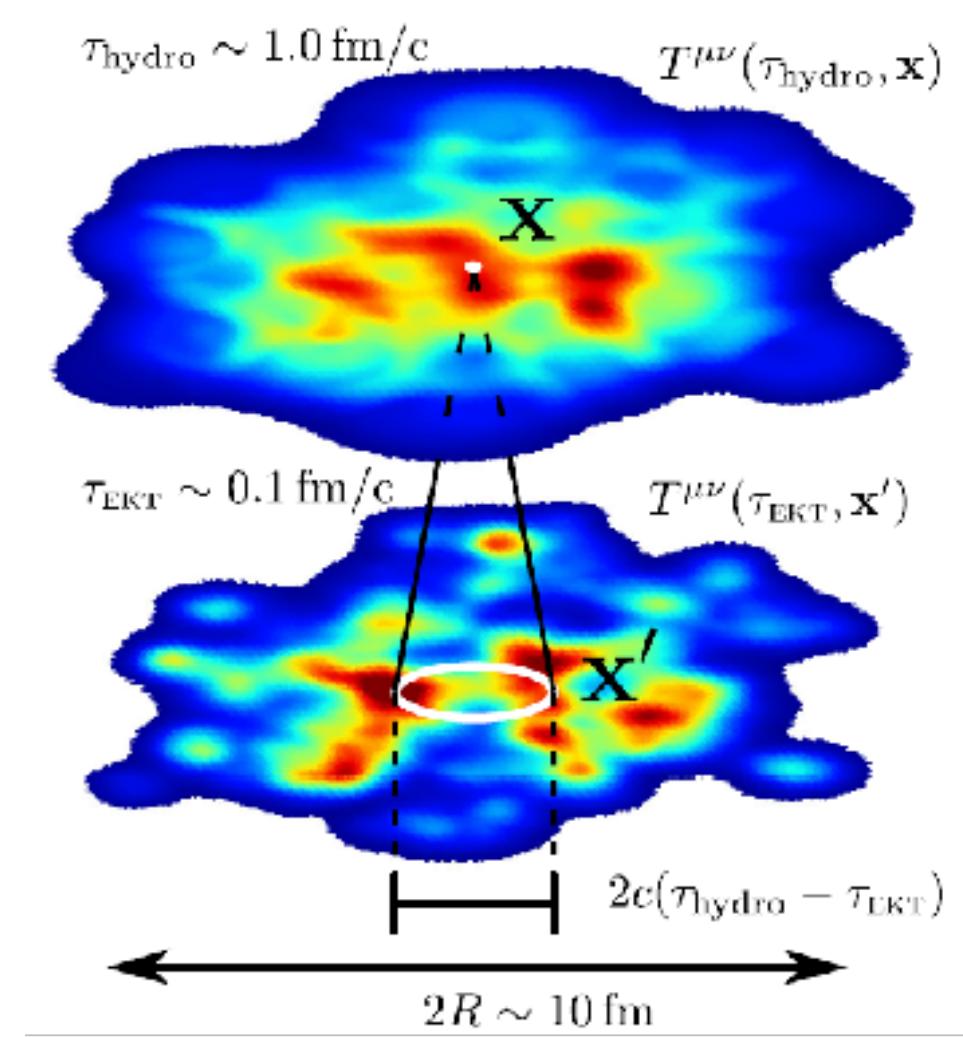
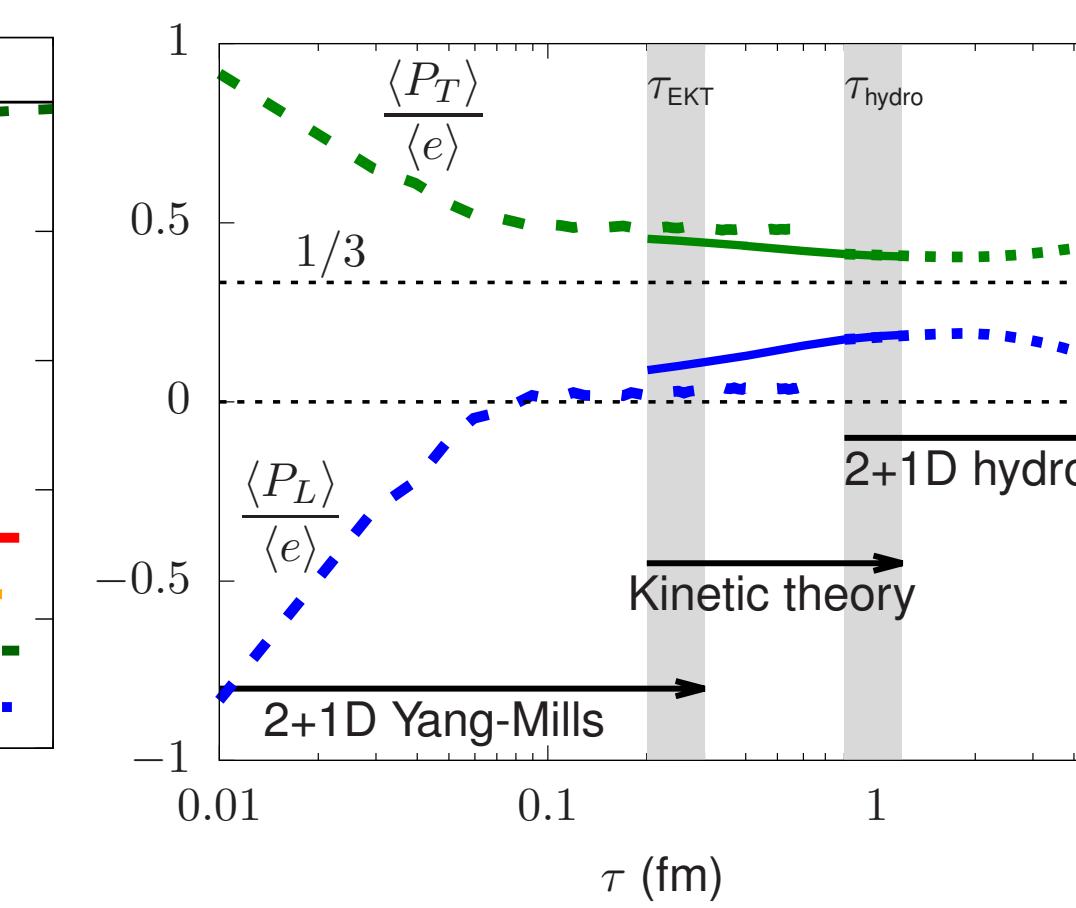
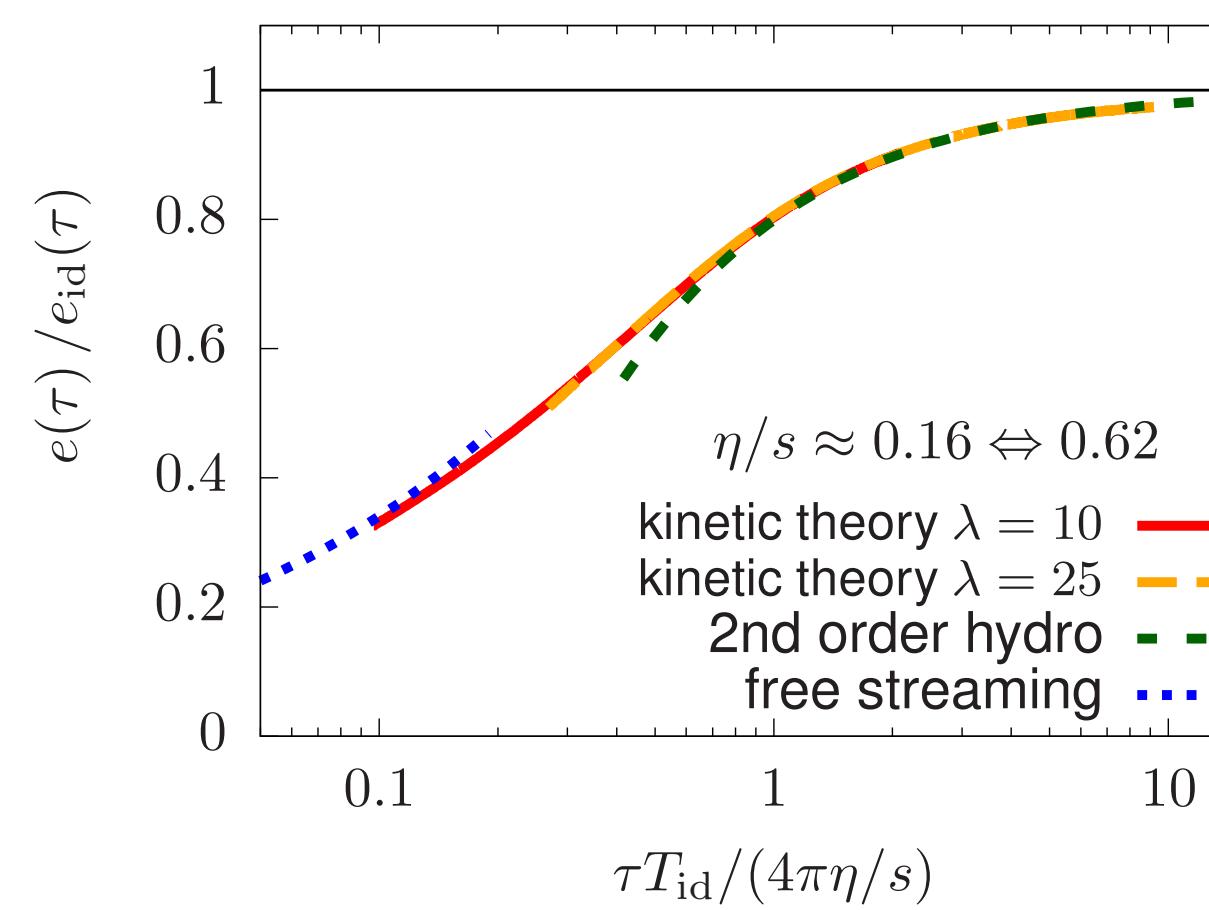
# Kinetic theory: from initial state to hydrodynamic evolution



KøMPøST model:

$$T^{\mu\nu}(\tau_{\text{hydro}}, \mathbf{x}) = \bar{T}_{\mathbf{x}}^{\mu\nu}(\tau_{\text{hydro}}) + \frac{\bar{T}_{\mathbf{x}}^{\tau\tau}(\tau_{\text{hydro}})}{\bar{T}_{\mathbf{x}}^{\tau\tau}(\tau_{\text{EKT}})} \int d^2\mathbf{x}' G_{\alpha\beta}^{\mu\nu}(\mathbf{x}, \mathbf{x}', \tau_{\text{hydro}}, \tau_{\text{EKT}}) \delta T_{\mathbf{x}}^{\alpha\beta}(\tau_{\text{EKT}}, \mathbf{x}')$$

- non-equilibrium linear response formalism:  
evolve energy-momentum tensor from its  
non-equilibrium form at  $\tau_{\text{EKT}}$  up to  $\tau_{\text{hydro}}$
- decompose energy-momentum tensor into  
a local average and linearized  
perturbations  $\delta T$
- response functions  $G$  describe evolution of  
perturbations calculated via kinetic theory

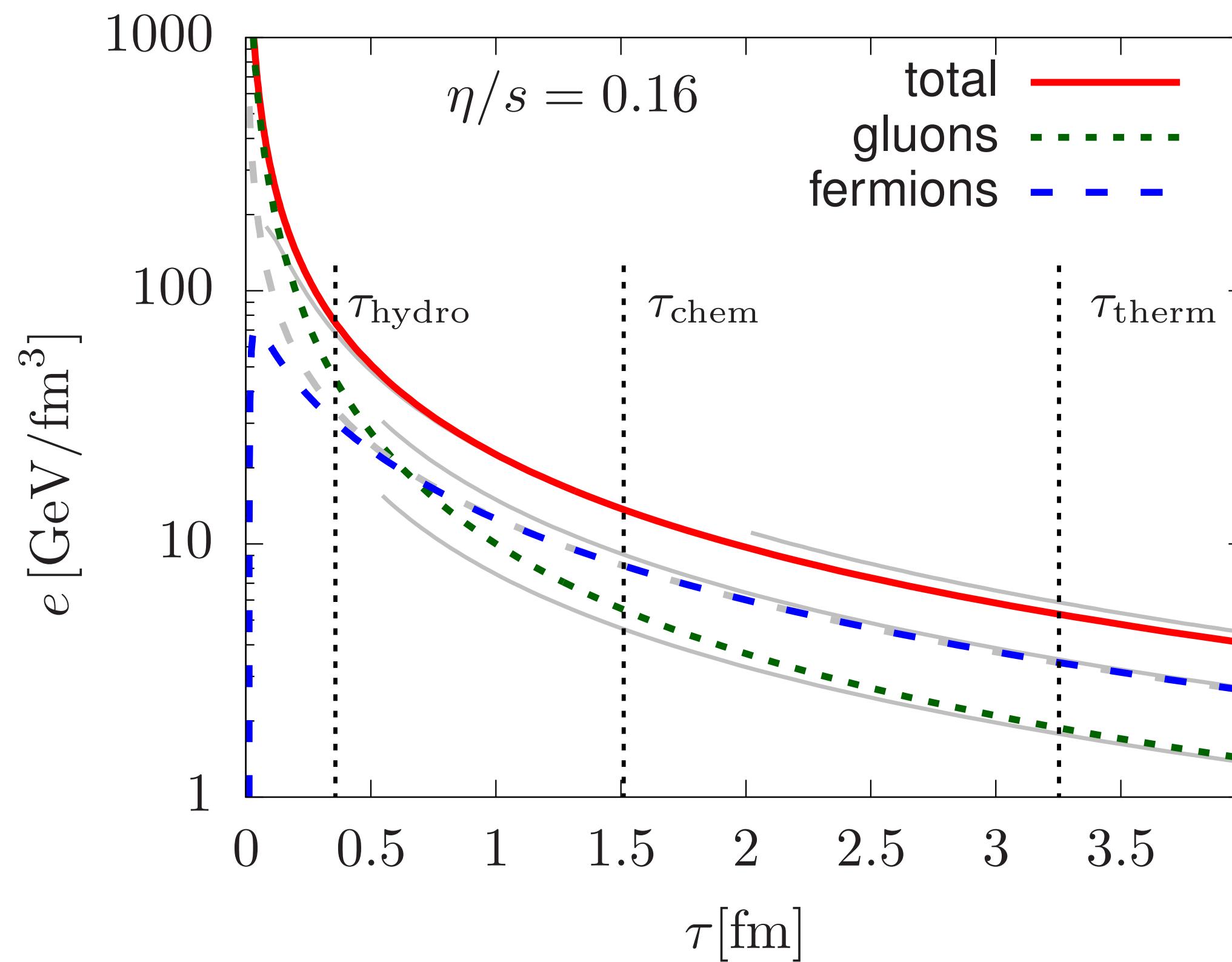


# Kinetic theory: chemical equilibration

Numerically solve Boltzmann equation for homogenous boost invariant q and g distribution functions:

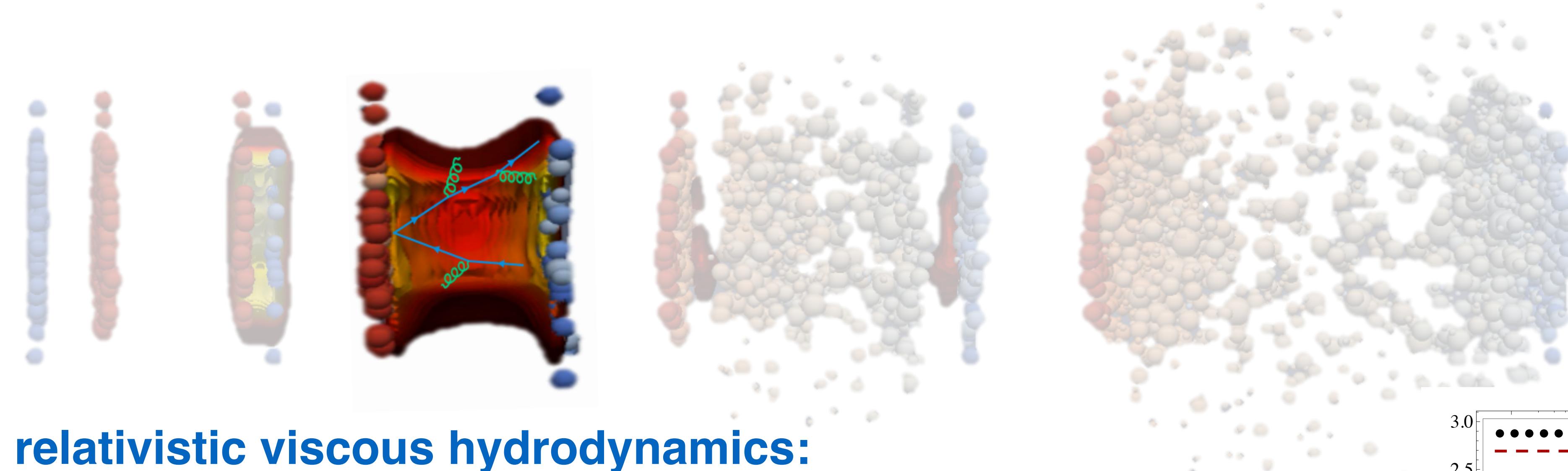
$$\partial_\tau f_s(\mathbf{p}, \tau) - \frac{p^z}{\tau} \partial_{p^z} f_s(\mathbf{p}, \tau) = -\mathcal{C}_{2 \leftrightarrow 2}^s[f] - \mathcal{C}_{1 \leftrightarrow 2}^s[f]$$

with scattering rates for  $gg \leftrightarrow gg$ ,  $gq \leftrightarrow gq$ ,  
 $qq \leftrightarrow qq$ ,  $gg \leftrightarrow qq\bar{q}$ ,  $g \leftrightarrow gg$ ,  $q \leftrightarrow qg$ ,  $g \leftrightarrow q\bar{q}$



- ordering of equilibration time scales:  
 $\tau_{\text{hydro}} < \tau_{\text{chem}} < \tau_{\text{therm}}$
- for reasonable values of the coupling (in terms of  $\eta/s$ ) one obtains the following time scale estimates
  - hydrodynamization:  $\sim 0.5$  fm/c
  - chemical equilibration:  $\sim 1.5$  fm/c
  - thermalization:  $\sim 3.3$  fm/c
- validates application of hydrodynamics at times  $\tau_0 \sim 0.6$  fm/c
- use of Lattice EoS in full chemical equilibrium at early times is questionable

# State of the Art: viscous relativistic fluid dynamics



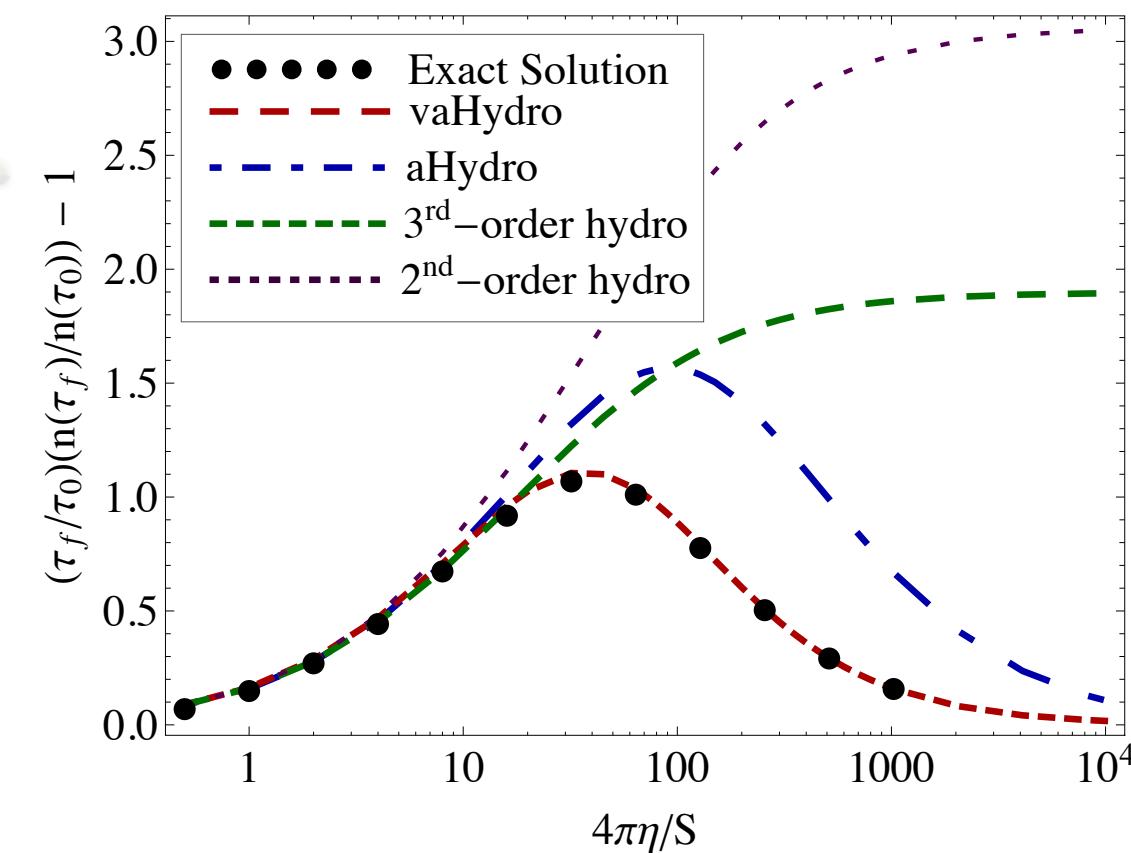
## standard relativistic viscous hydrodynamics:

- based on 2nd order Israel-Stewart theory
- (Lattice) Equation of State & (temperature dependent) shear & bulk viscosities as input
- several extensively validated codes are publicly available: MUSIC, VISHNU, vHLL

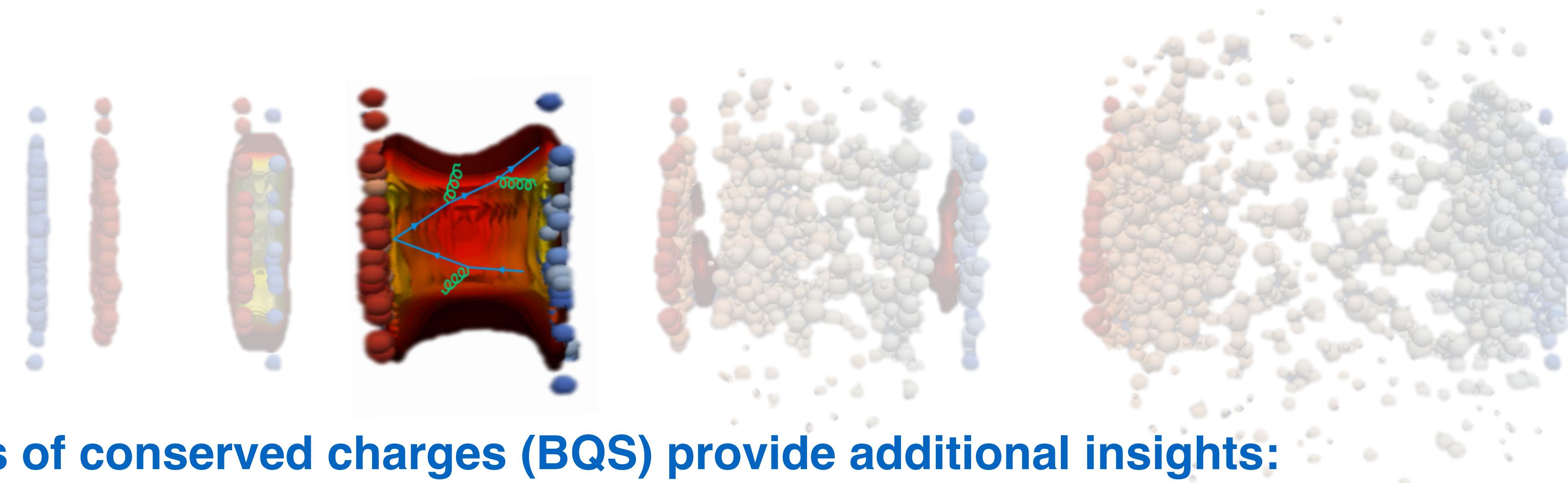
## anisotropic hydro:

- allows to describe systems far from isotropy (equilibrium)
- aHydroQP: QCD EoS via a temperature-dependent quasi-particle mass, includes shear & bulk viscosities

- VISH: Song & Heinz; Phys. Rev. C77 (2008) 064901
- MUSIC: Schenke, Jeon & Gale; Phys. Rev. C82 (2010) 014903
- vHLL: Karpenko, Huovinen & Bleicher; Computer Phys. Comm. 185 (2014) 3016-3027
- VISHNU: Chen, Song, Bernhard, Bass & Heinz; Computer Phys. Comm. 199 (2016) 61-65
- aHydroQP: M. Alqahtani, M. Nopoush & M. Strickland: Phys. Rev. C92 (2015) 054910

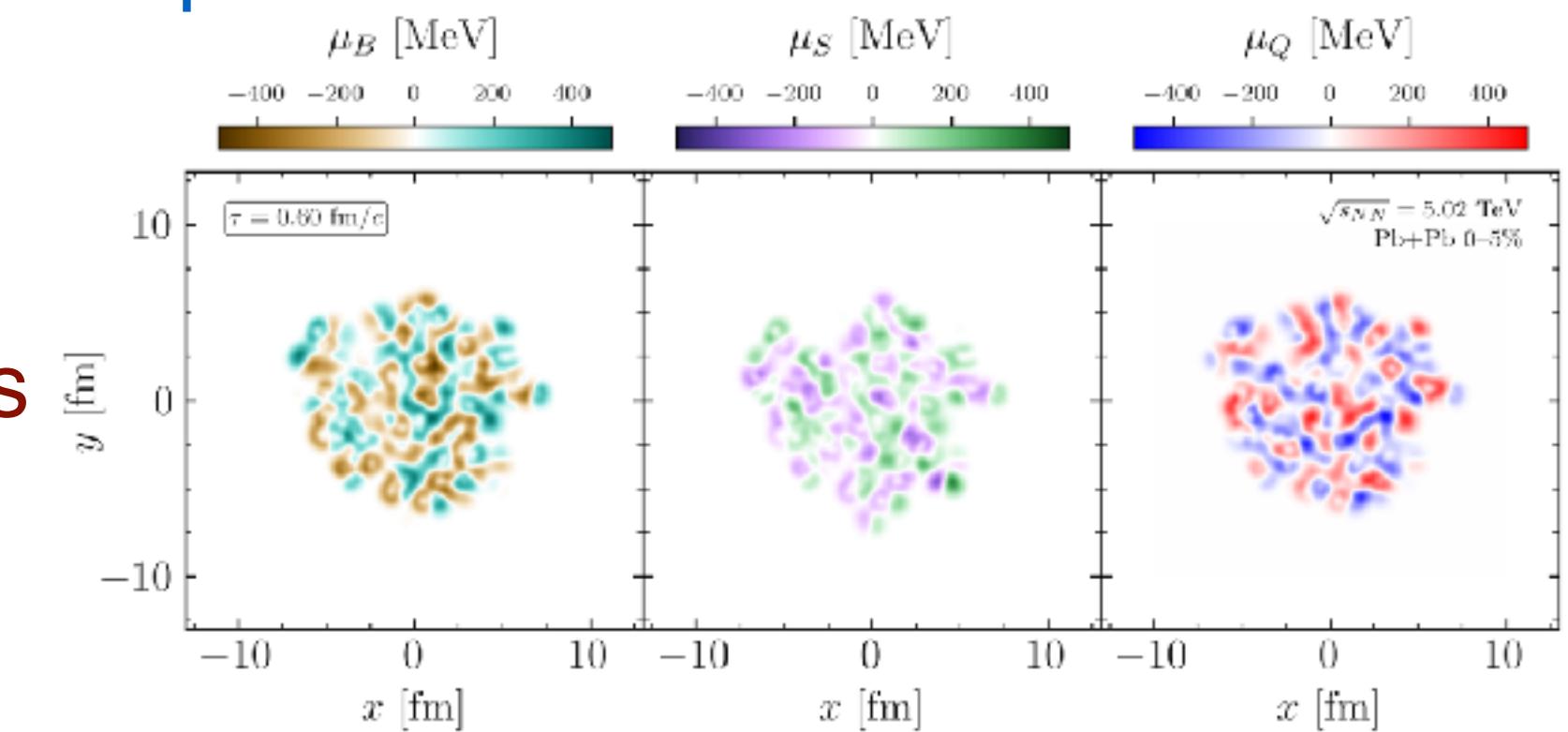


# New developments: hydro with conserved charges



## Dynamics of conserved charges (BQS) provide additional insights:

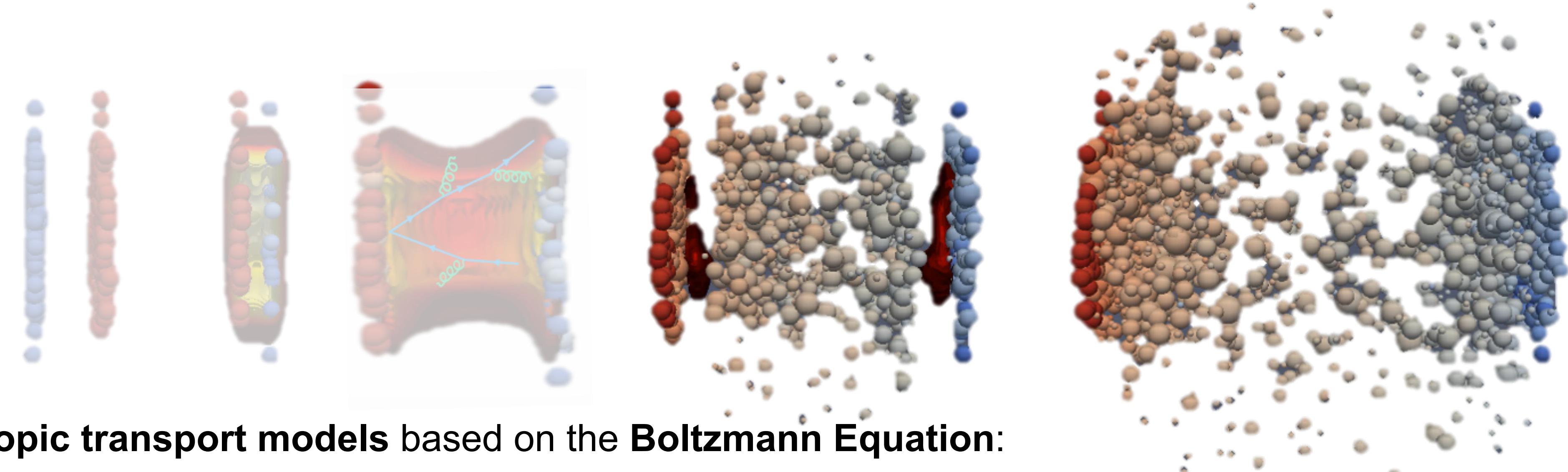
- impact of gluon splitting (charge creation) in the initial state
- formation and dynamics of multi-strange hadrons
- proper modeling of conserved charge fluctuations as probes of the QCD critical point



## Hydro with BSQ conserved charges:

- conservation of Baryon-, Strangeness and Charge (vHLLE)
- 4D Equation of State, EoM with a 3x3 diffusion matrix for conserved charges
- coupling terms between shear & bulk viscosities and BSQ currents
- available codes: CCAKE w/ ICCING initial condition (SPH), MUSIC

# (Mostly) hadronic transport: UrQMD and SMASH



**microscopic transport models based on the Boltzmann Equation:**

$$\left[ \frac{\partial}{\partial t} + \frac{\vec{p}}{E} \times \frac{\partial}{\partial \vec{r}} \right] f_1(\vec{p}, \vec{r}, t) = \sum_{processes} C(\vec{p}, \vec{r}, t)$$

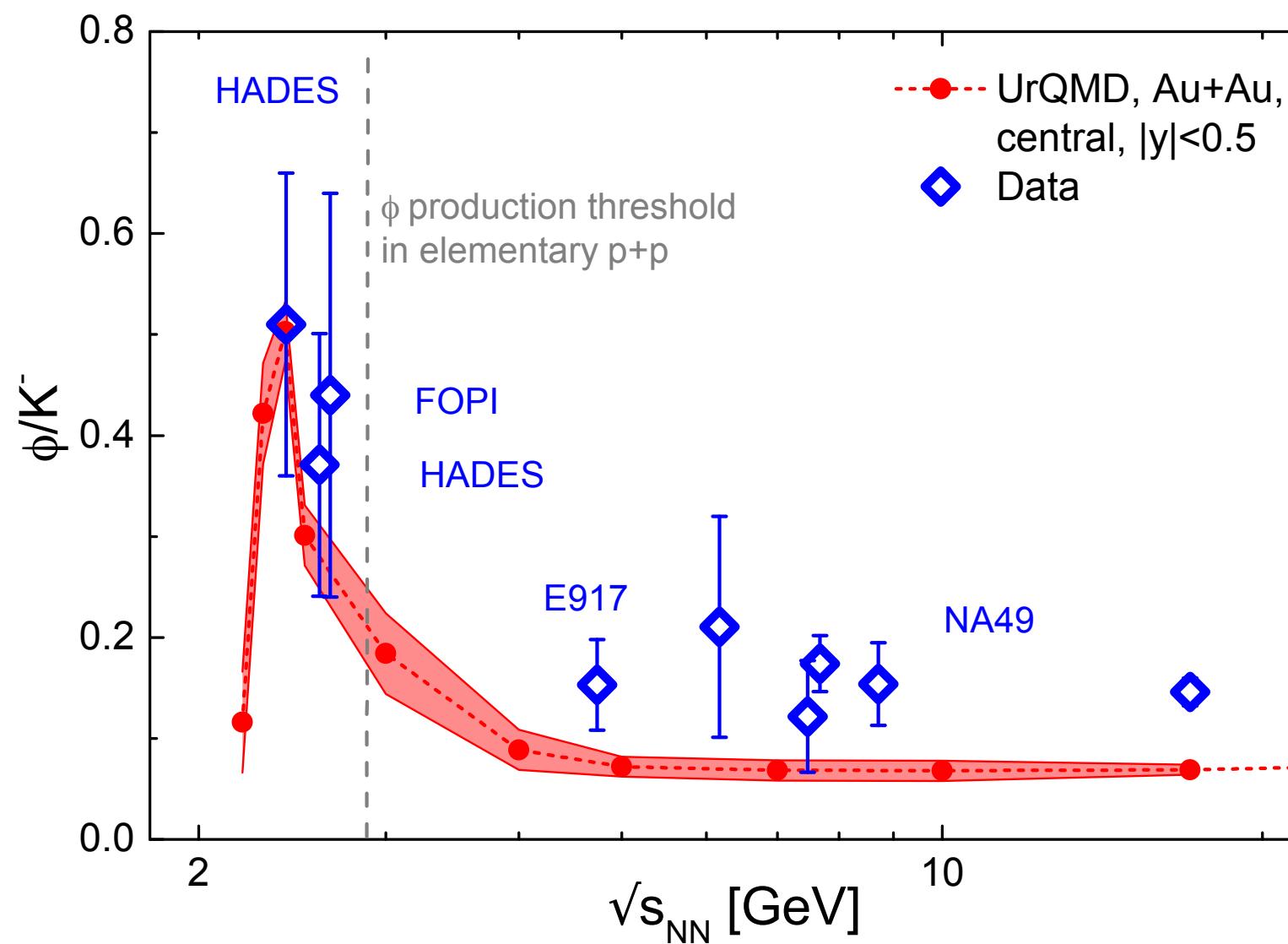
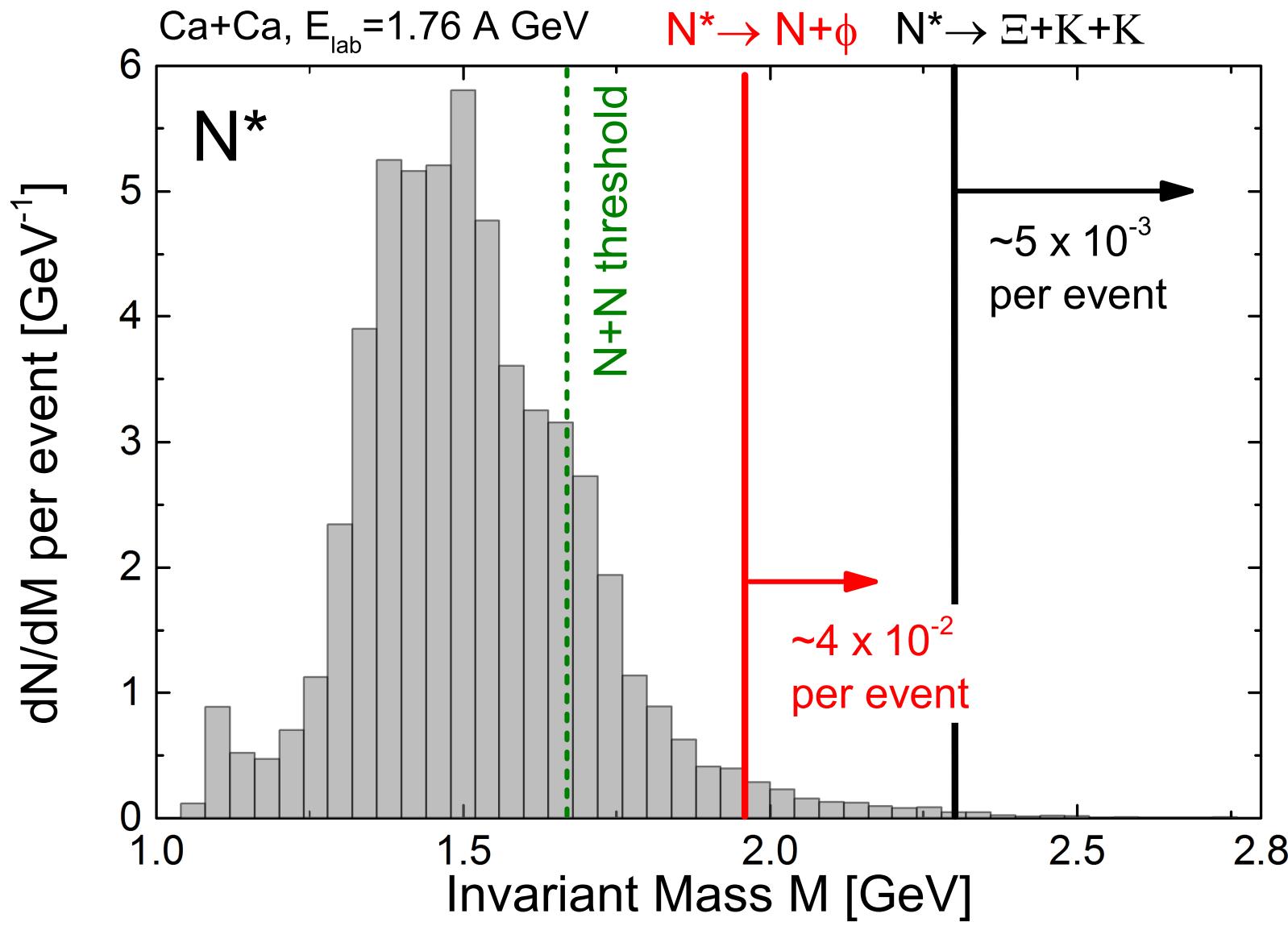
- transport of a system of hadrons (all well-established hadrons & resonances listed by PDG up to mass of  $M \approx 2.4$  GeV)
- interactions are based on **binary scattering** using a **geometric collision criterion**:  $d_{min} \leq \sqrt{\frac{\sigma_{tot}}{\pi}}$
- multi-particle decays via formation of intermediate resonance states
- string fragmentation via PYTHIA for initial particle production in high-energy collisions
- Potentials modeled via QMD type sum over two-particle interactions (UrQMD) or test particle method (SMASH)

- S. A. Bass et al.: Prog.Part.Nucl.Phys. 41 (1998) 255-369
- M. Bleicher et al.: J.Phys.G 25 (1999) 1859-1896
- J. Weil et al.: Phys. Rev. C 94 (2016) 054905

# Phi meson production: resonances & rescattering

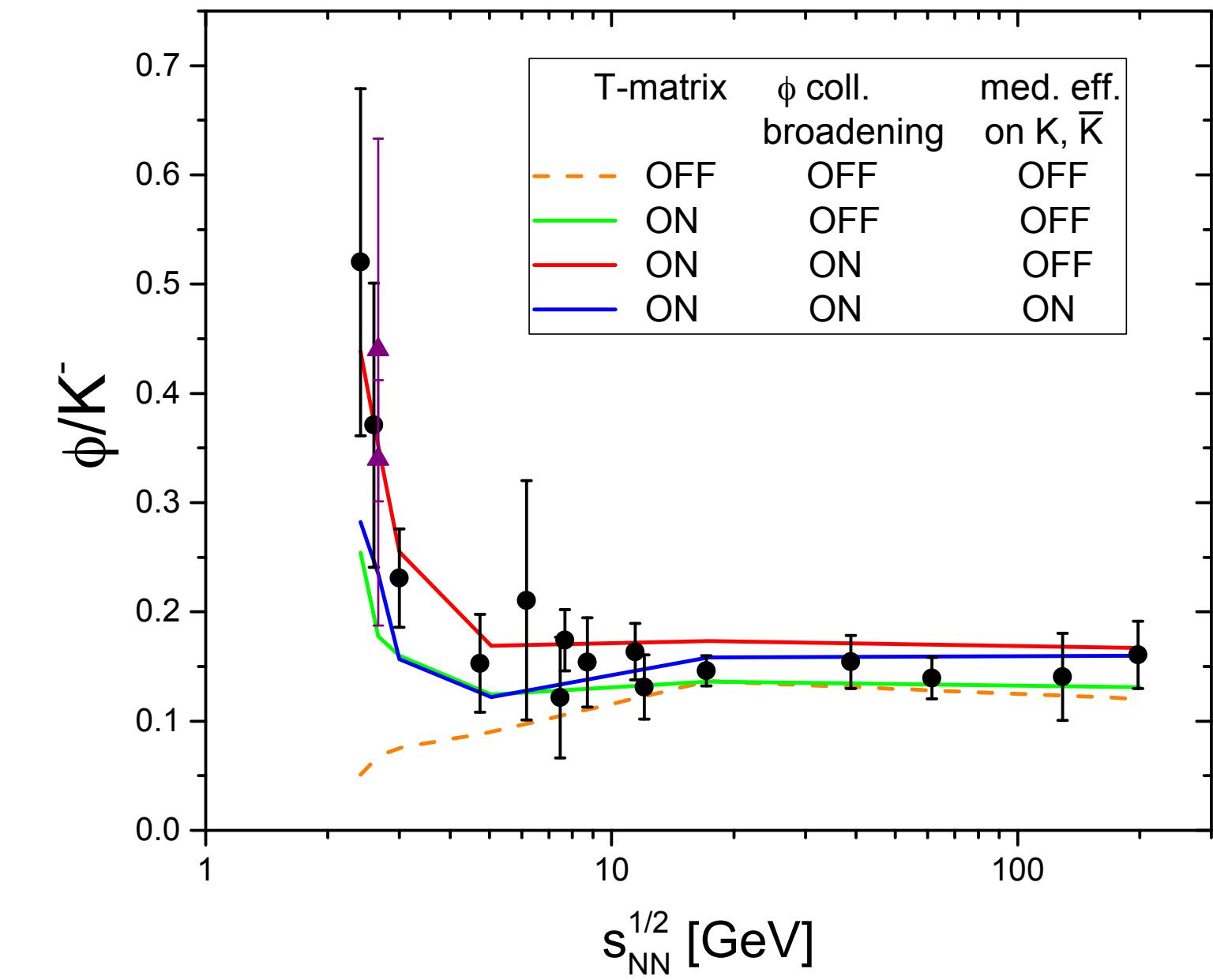
## UrQMD:

- at SIS energies,  $\phi$  meson production is strongly suppressed since only the highest mass tails of the spectral function of high mass  $N^*$  resonances allow for an on-shell decay into a  $\phi$  meson final state
- BB scattering with subsequent BB and mB rescattering allows for the excitation of heavy resonances w/ a branching ratio to  $\phi$  mesons  $N^* \leftrightarrow \phi B$  at beam energies below the pp threshold



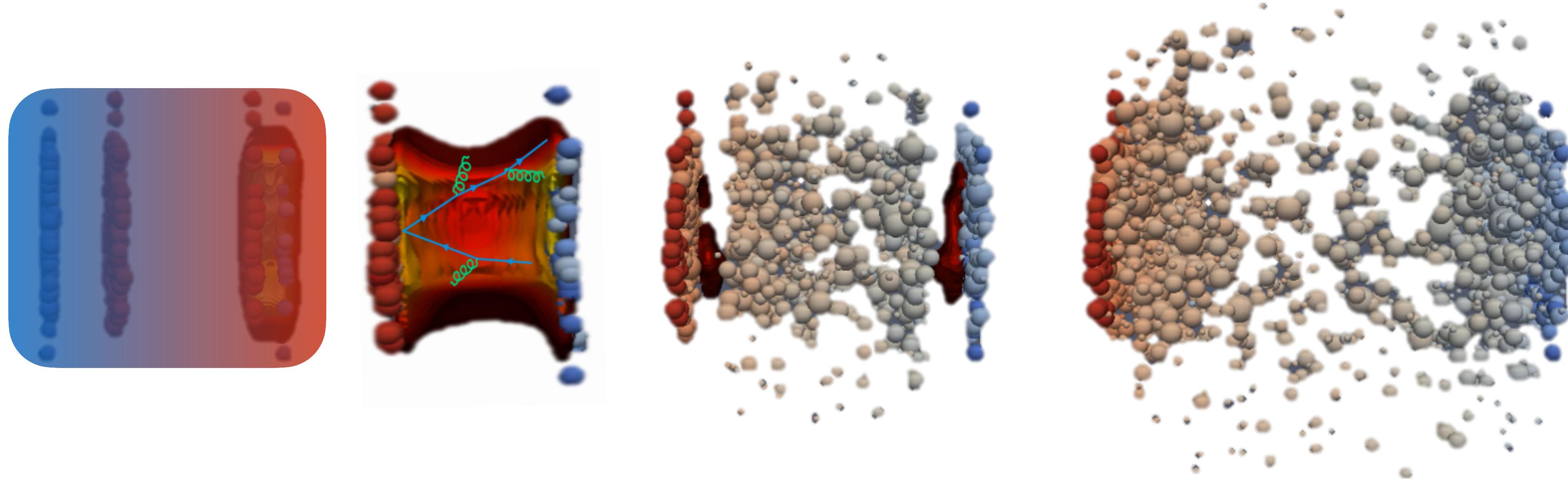
## PHSD:

- introduce  $mB \leftrightarrow \phi B$  channels via SU(6) based T-matrix approach
- include potentials that modify kaon mass as function of baryon-density: enhances anti-K yield and suppresses K yield near threshold



- J. Steinheimer & M. Bleicher: J. Phys. G 43 (2016) 015104
- T. Song, J. Aichelin & E. Bratkovskaya: Phys. Rev. C 106 (2022) 024903

# State of the Art @ High Energy: Macro + Micro Hybrid



## Initial condition:

- calculated on the basis of gluon saturation physics or pomeron exchange
- alternatively: phenomenological scheme for entropy deposition & constrained by global model to data fit
- examples: IP-Glasma, Trento

## viscous hydrodynamics

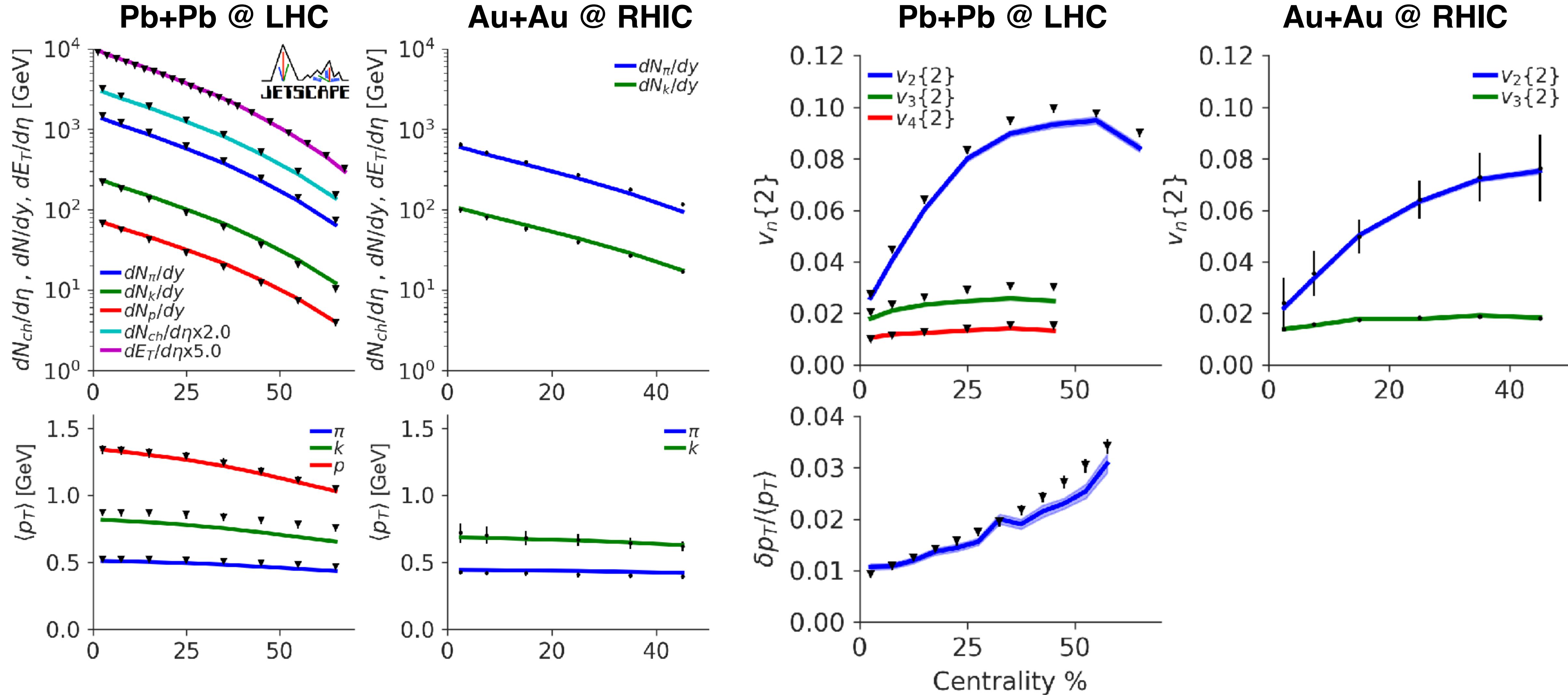
- EbE 3+1D viscous RFD
- describes QGP dynamics & hadronization
- Lattice QCD EoS
- examples: MUSIC, VISHNU, vHLLE

## Hadronic afterburner:

- non-equilibrium evolution of an interacting hadron gas
- separation of chemical and kinetic freeze-out
- hadron gas shear & bulk viscosities are implicitly contained in calculation
- Examples: UrQMD, SMASH

Similar Macro + Micro Hybrids: EPOS 4 & Trajectum

# Macro + Micro Hybrid: comprehensive description of data



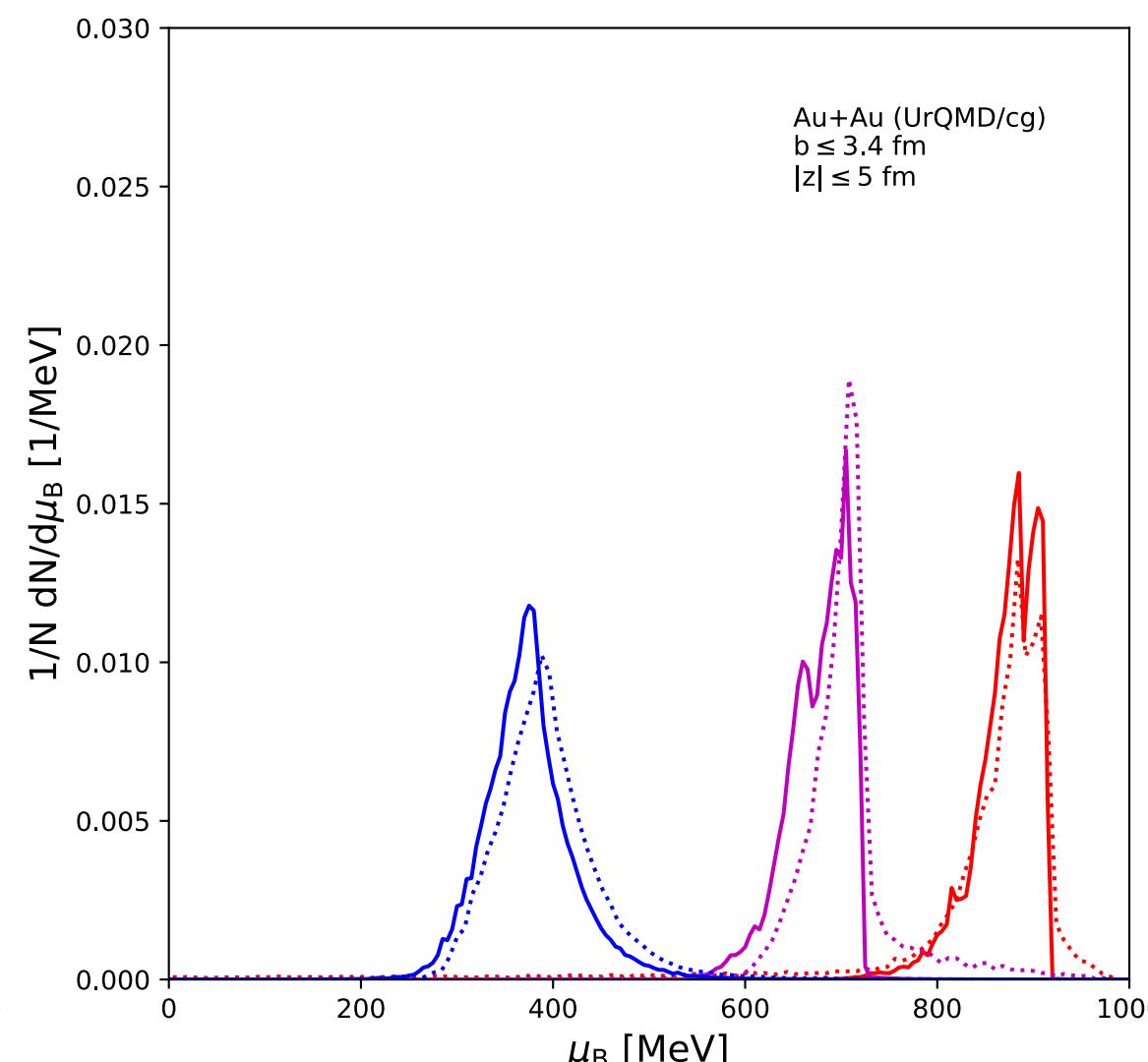
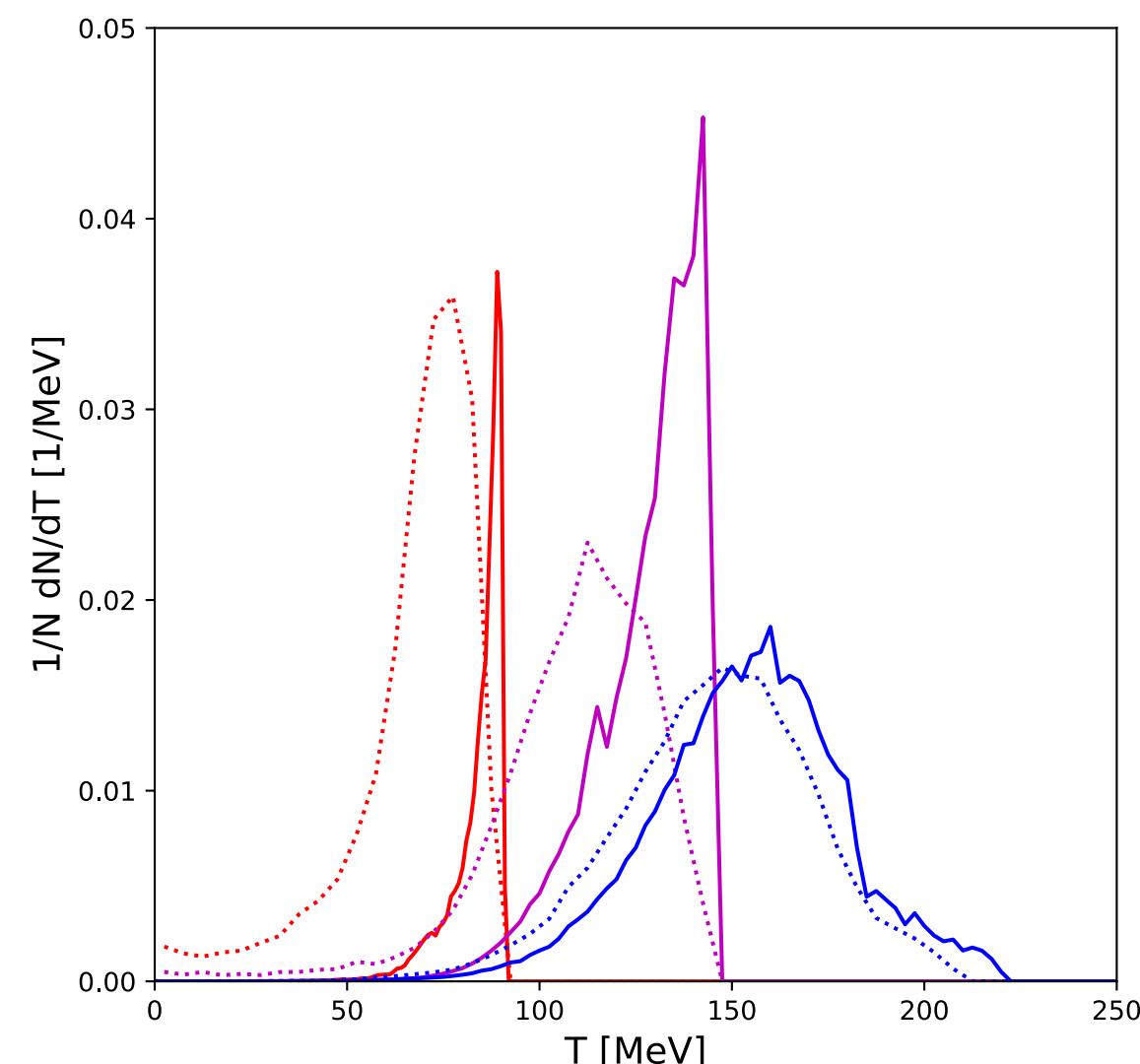
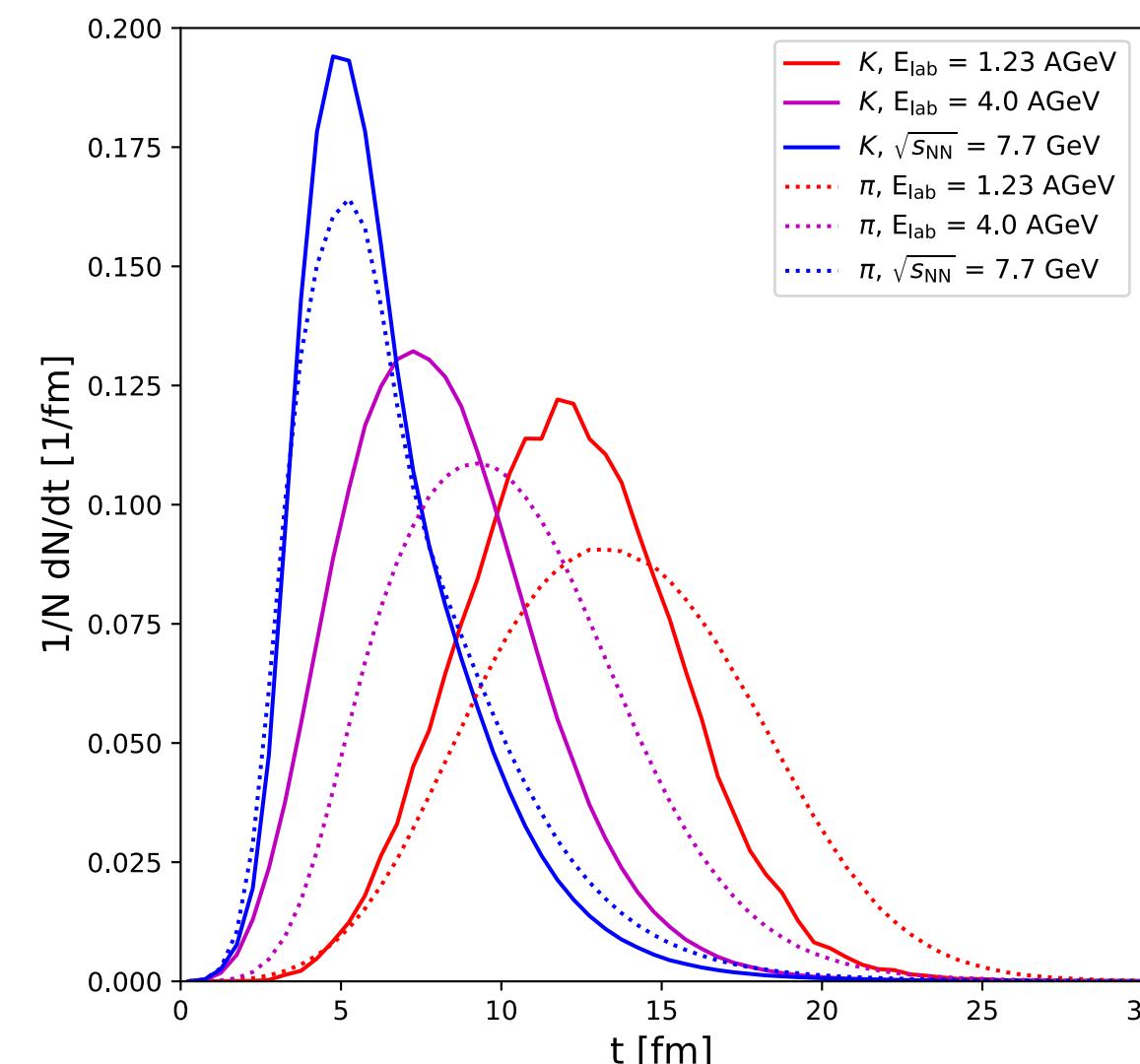
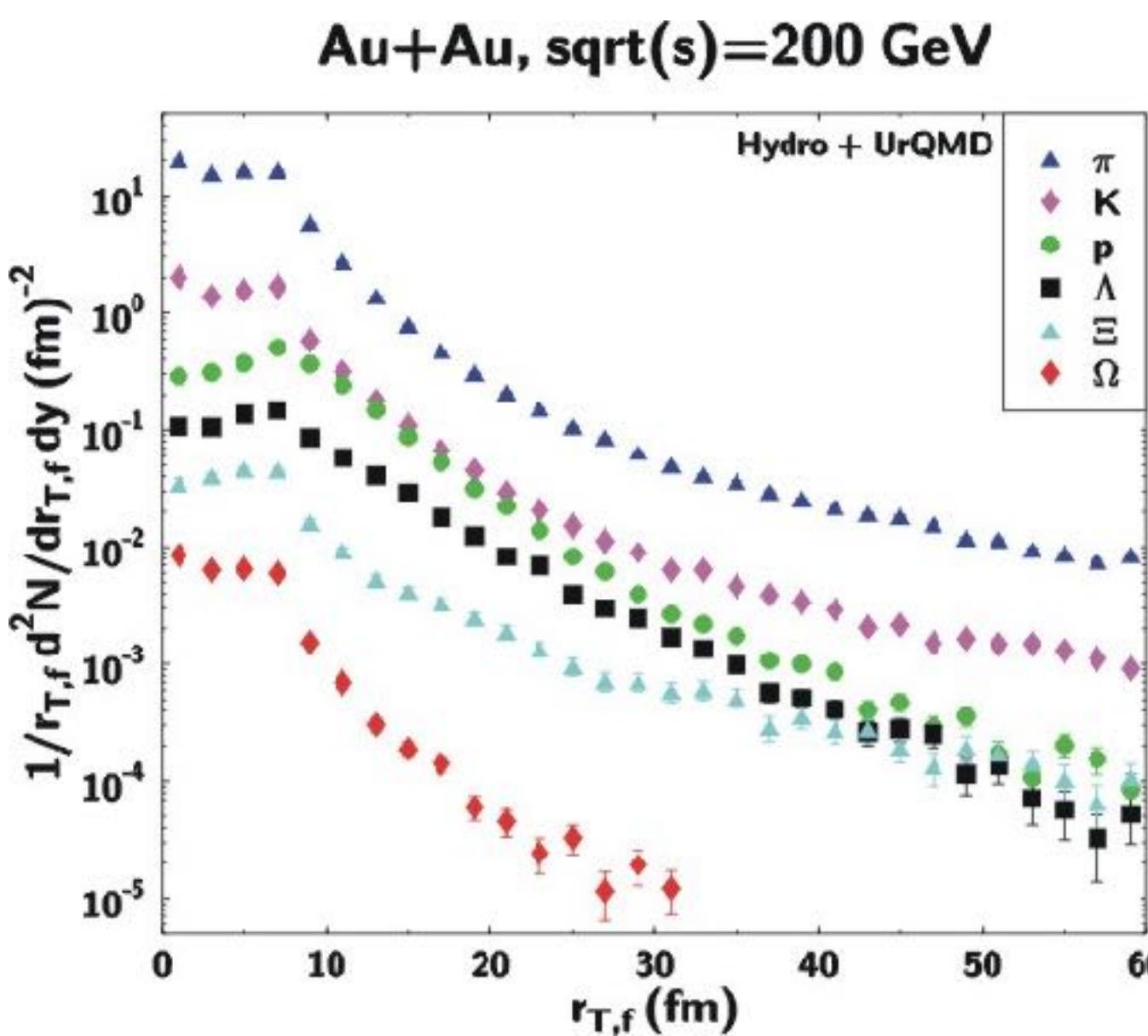
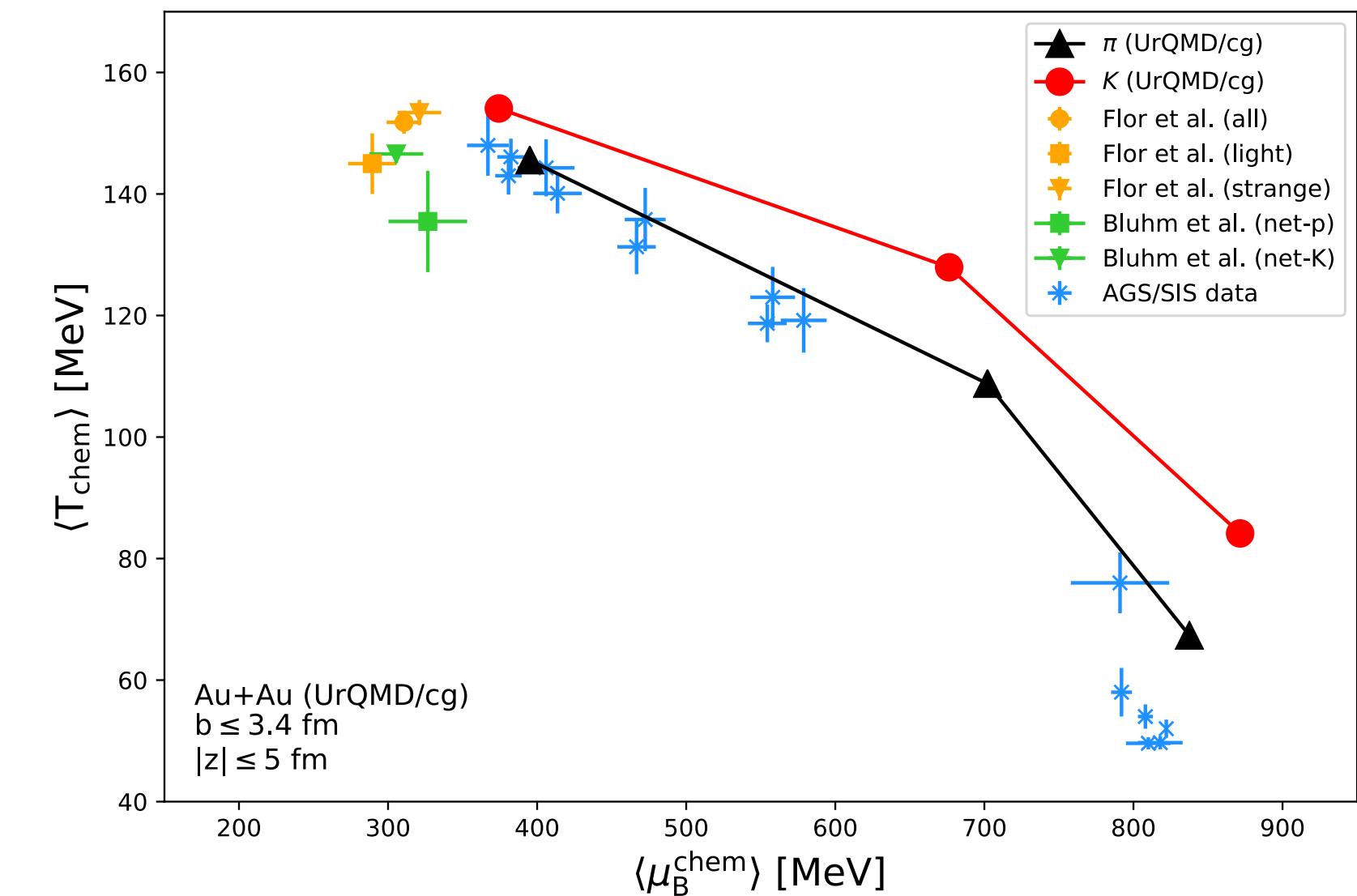
- Macro + Micro hybrid models are capable of describing a comprehensive set of bulk observables across RHIC & LHC

# Transport and strangeness: mapping freeze-out

**Chemical Freeze-Out:** flavor composition fixed

**Kinetic Freeze-Out:** cessation of elastic scattering

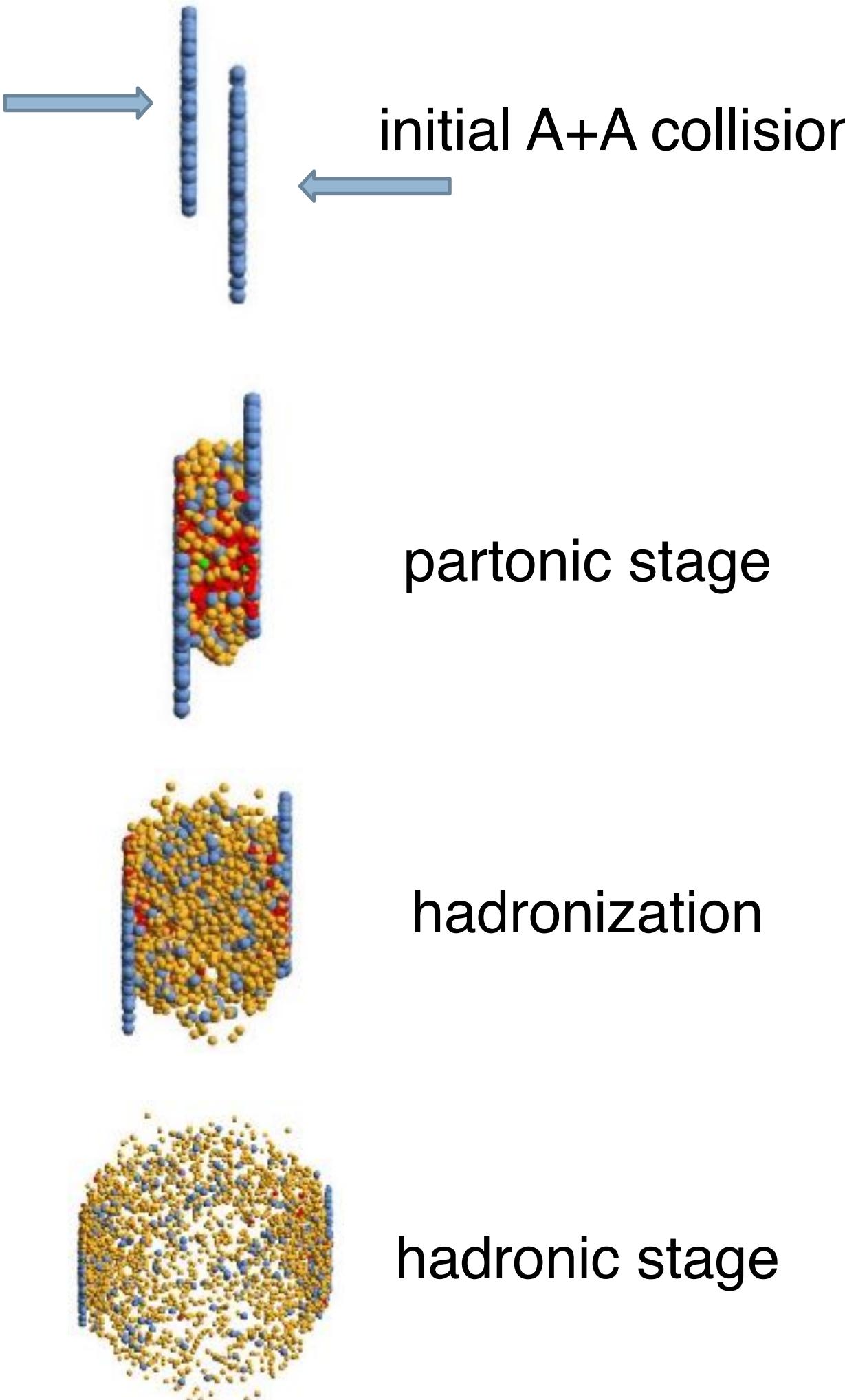
- freeze-out times/radii/temperatures are broad distributions with long tails
- freeze-out is species-dependent: cross sections drive the process
- strangeness suppresses magnitude of cross section: hierarchy of freeze-out times, temperatures and radii
- multi-strange baryons are emitted closest to the phase-boundary



• S.A. Bass & A. Dumitru: Phys. Rev. C 61 (2000) 064909

• T. Reichert, G. Inghirami & M. Bleicher: European Physics Journal Web Conf. 259 (2022) 10005 [SQM 2021]

# Microscopic transport: PHSD



- String formation in primary NN collisions  
→ decays to pre-hadrons (baryons and mesons)
- Formation of a QGP state if  $\varepsilon > \varepsilon_{critical}$  :  
Dissolution of pre-hadrons → DQPM  
→ massive quarks/gluons and mean-field energy
- (quasi-)elastic collisions :      inelastic collisions :  

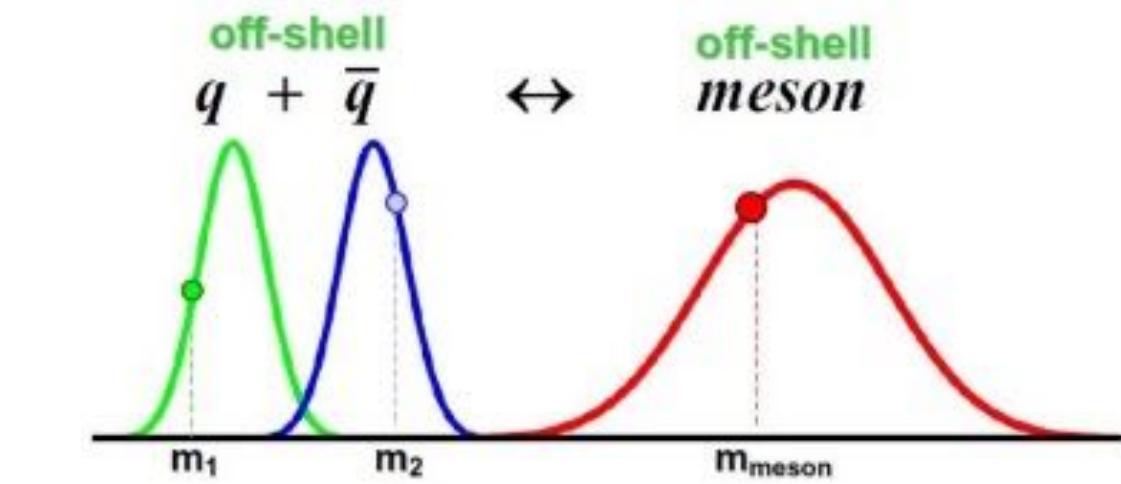
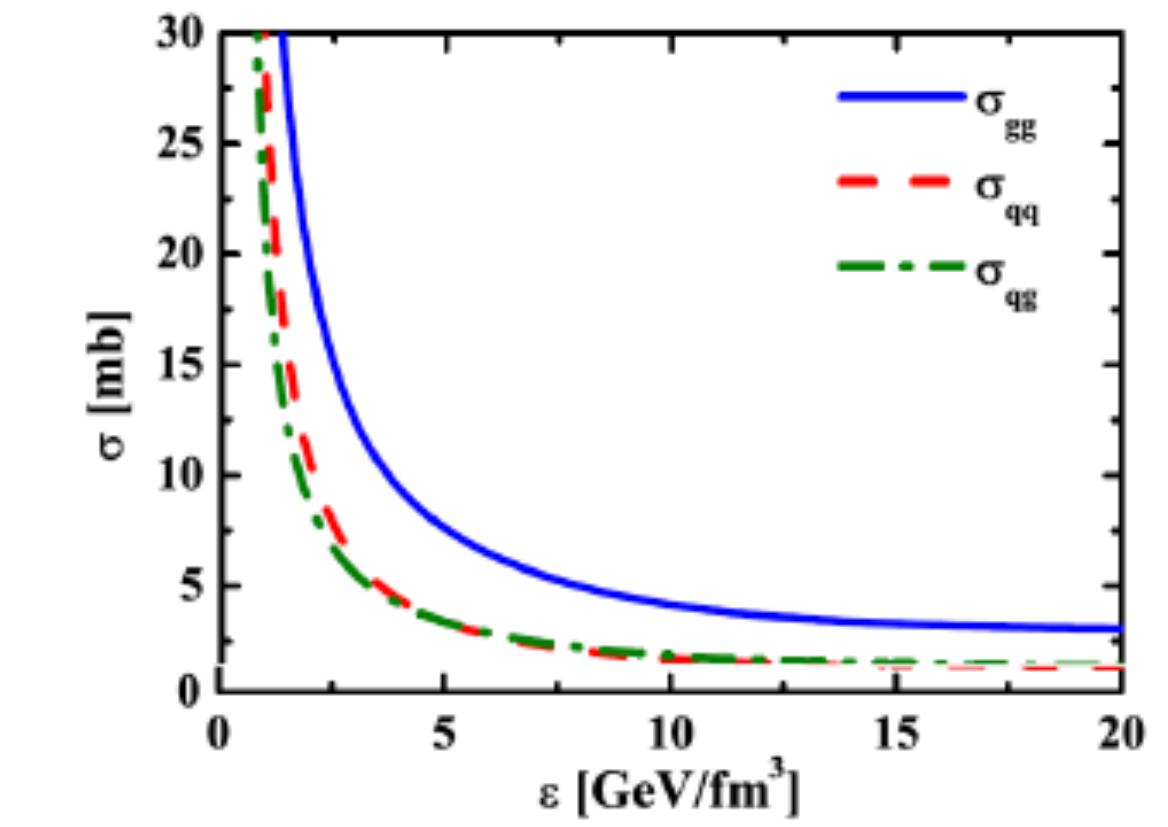
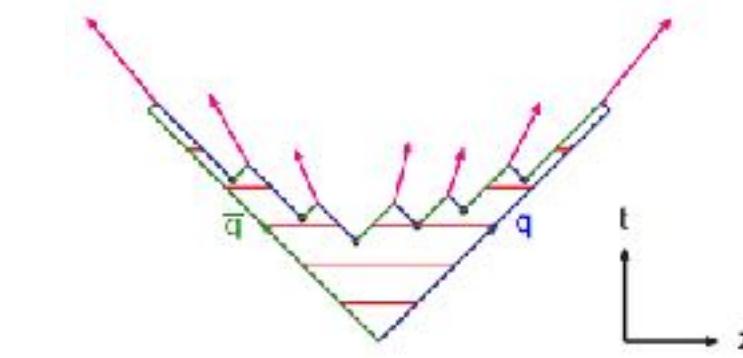
$$\begin{array}{ll} g + q \rightarrow g + q & g + q \rightarrow g + q \\ g + \bar{q} \rightarrow g + \bar{q} & g + \bar{q} \rightarrow g + \bar{q} \\ g + g \rightarrow g + g & g \rightarrow g + g \end{array}$$

$$\begin{array}{ll} q + \bar{q} \rightarrow g + g & q + \bar{q} \rightarrow g + g \\ g \rightarrow g + g & g \rightarrow g + g \end{array}$$
- Hadronization to colorless off-shell mesons and baryons  

$$g \rightarrow q + \bar{q}, \quad q + \bar{q} \leftrightarrow \text{meson ('string')} \quad \text{Strict 4-momentum and quantum number conservation}$$

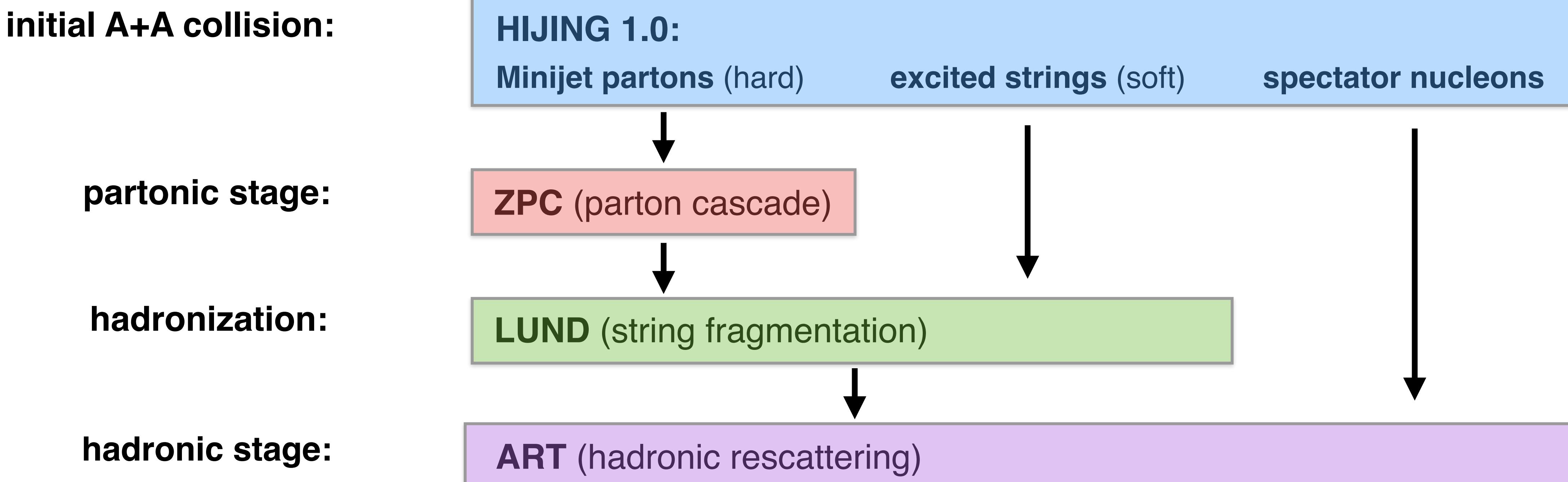
$$q + q + q \leftrightarrow \text{baryon ('string')}$$
- Hadron-string interactions – off-shell HSD

LUND string model



# Microscopic transport: AMPT 1.X

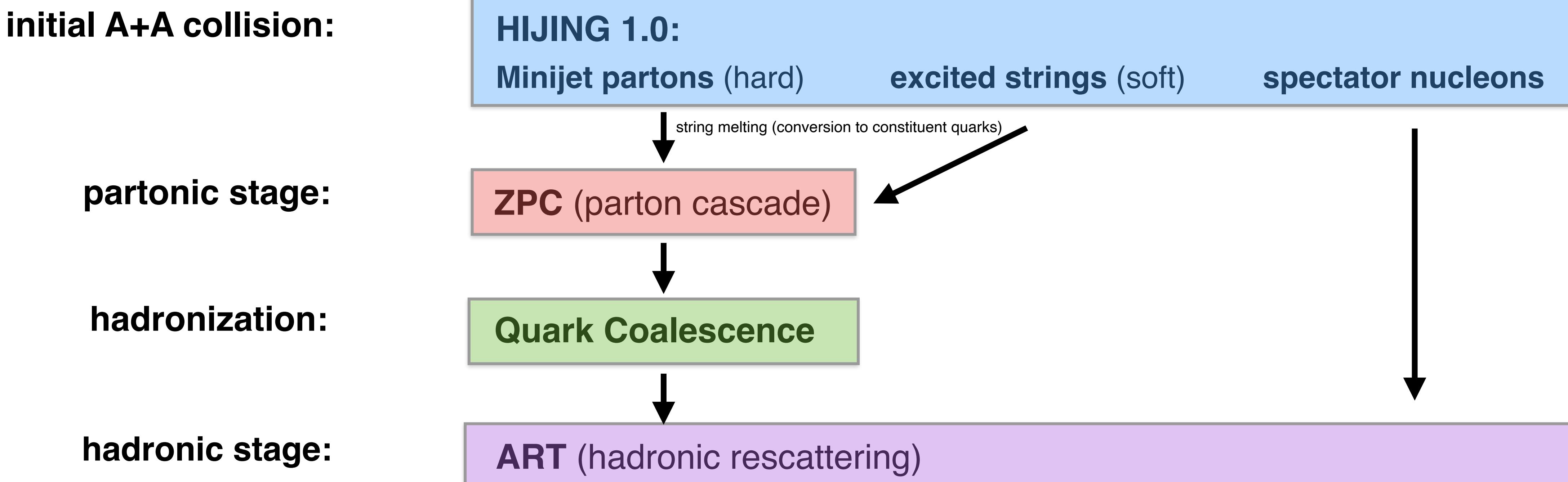
- A multi-phase transport (AMPT) model: constructed as a self-contained kinetic description of heavy ion collisions based on microscopic transport (Boltzmann eqn)
- evolves the system from initial condition to final observables via a changing set of microscopic degrees of freedom
- includes productions of all flavours, 3D, conserved charges
- non-equilibrium initial condition & dynamics/evolution



- Z. Lin, C. Ko, B. Zhang & S. Pal: Phys. Rev. C72 (2005) 064901
- Z. Lin & L. Zheng: Nucl.Sci.Tech. 32 (2021) 113
- <https://myweb.ecu.edu/linz/ampt/>

# Microscopic transport: AMPT 2.X

- A multi-phase transport (AMPT) model: constructed as a self-contained kinetic description of heavy ion collisions based on microscopic transport (Boltzmann eqn)
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# Low/Medium Energy: Quantum Molecular Dynamics

The Boltzmann Equation does not contain many-body correlations needed to describe cluster formation: use a many-body approach instead: **Quantum Molecular Dynamics (QMD)**

generalized Ritz variational principle:

$$\delta \int_{t_1}^{t_2} \left\langle \Phi \left| i\hbar \frac{d}{dt} - H \right| \Phi \right\rangle dt = 0 \quad \text{with} \quad \Phi = \prod_i \phi_i(\vec{x}, \vec{q}_i, \vec{p}_i, t) \quad (\text{N-body wave function w/o anti-symmetrization})$$

and  $\phi_i(\vec{x}; \vec{q}_i, \vec{p}_i, t) = \left( \frac{2}{L\pi} \right)^{3/4} \exp \left\{ -\frac{2}{L} (\vec{x} - \vec{q}_i(t))^2 + \frac{1}{\hbar} i \vec{p}_i(t) \cdot \vec{x} \right\}$

for a Hamiltonian of the form:

$$H = \sum_i T_i + \frac{1}{2} \sum_{ij} V_{ij}$$

the variational principle yields EoM for the centers of the Gaussians:

$$\dot{\vec{p}}_i = -\frac{\partial \langle H \rangle}{\partial \vec{q}_i} \quad \text{and} \quad \dot{\vec{q}}_i = \frac{\partial \langle H \rangle}{\partial \vec{p}_i}$$

- inclusion of potentials / a realistic equation of state is crucial for cluster formation
- current models include: UrQMD, PHQMD, IQMD

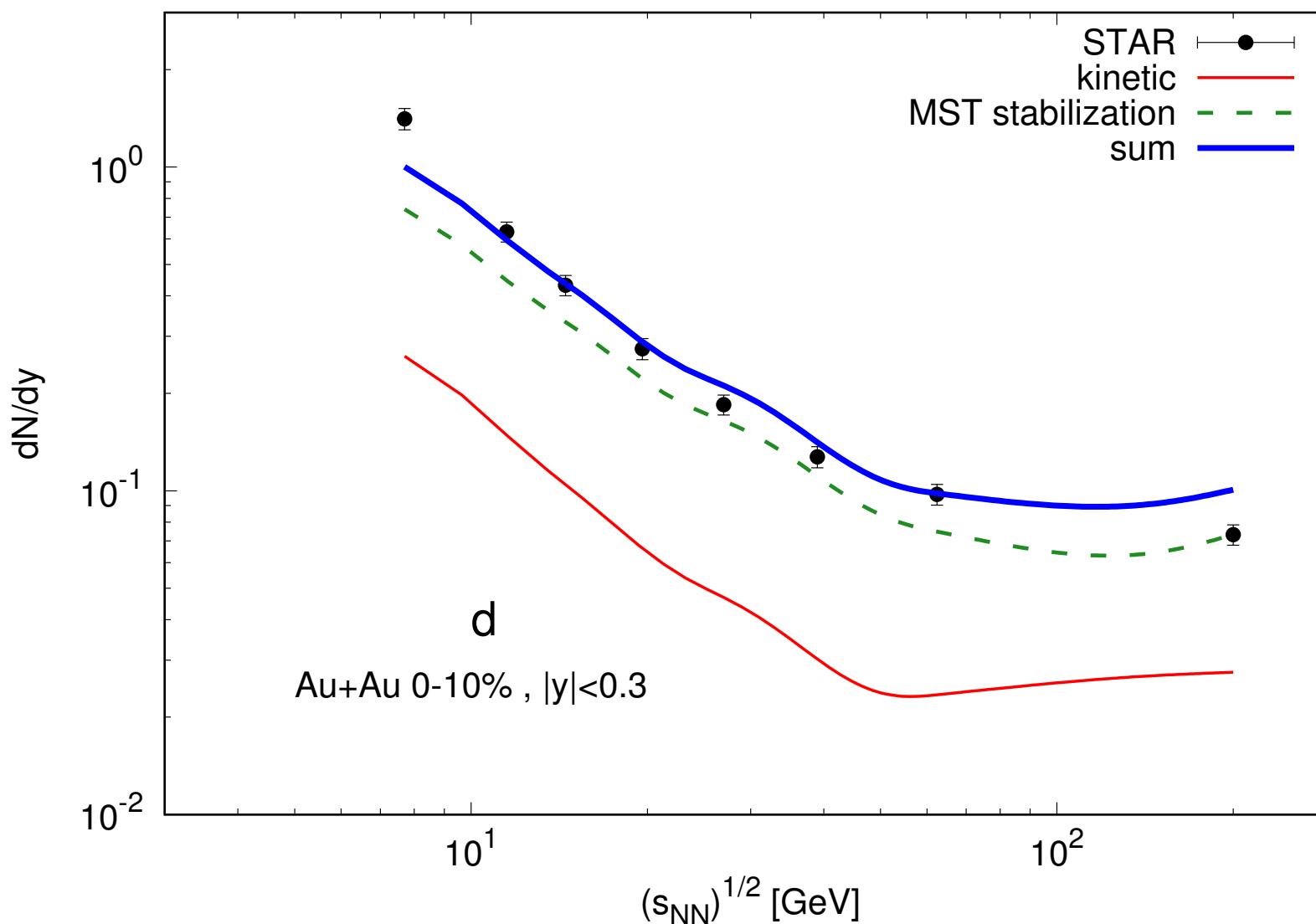
- J. Aichelin: Phys. Rept. 202 (1991) 233-360
- S. A. Bass et al.: Prog. Part. Nucl. Phys. 41 (1998) 255-369
- J. Aichelin, E. Bratkovskaya, A. Le Fevre, V. Kireyev & V. Kolesnikov: Phys. Rev. C101 (2020) 044905

# Cluster formation at low/medium energies:

Transport is a powerful tool for the study of the non-equilibrium dynamics of cluster formation:

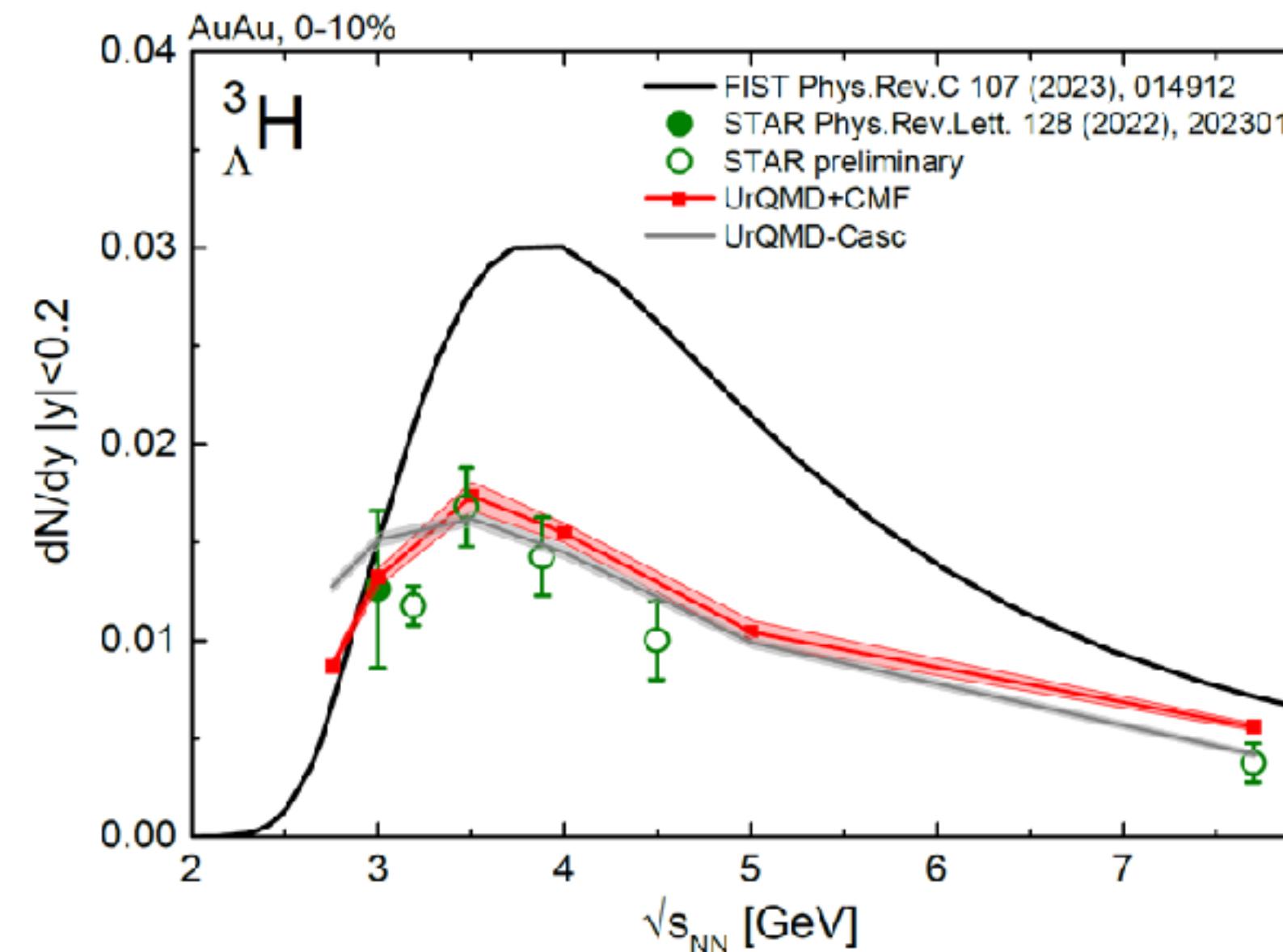
## Deuteron formation (RHIC-BES):

- PHQMD w/ realistic EoS & MST cluster algorithm
- Calculation shows the need to include correlations and binding energy effects



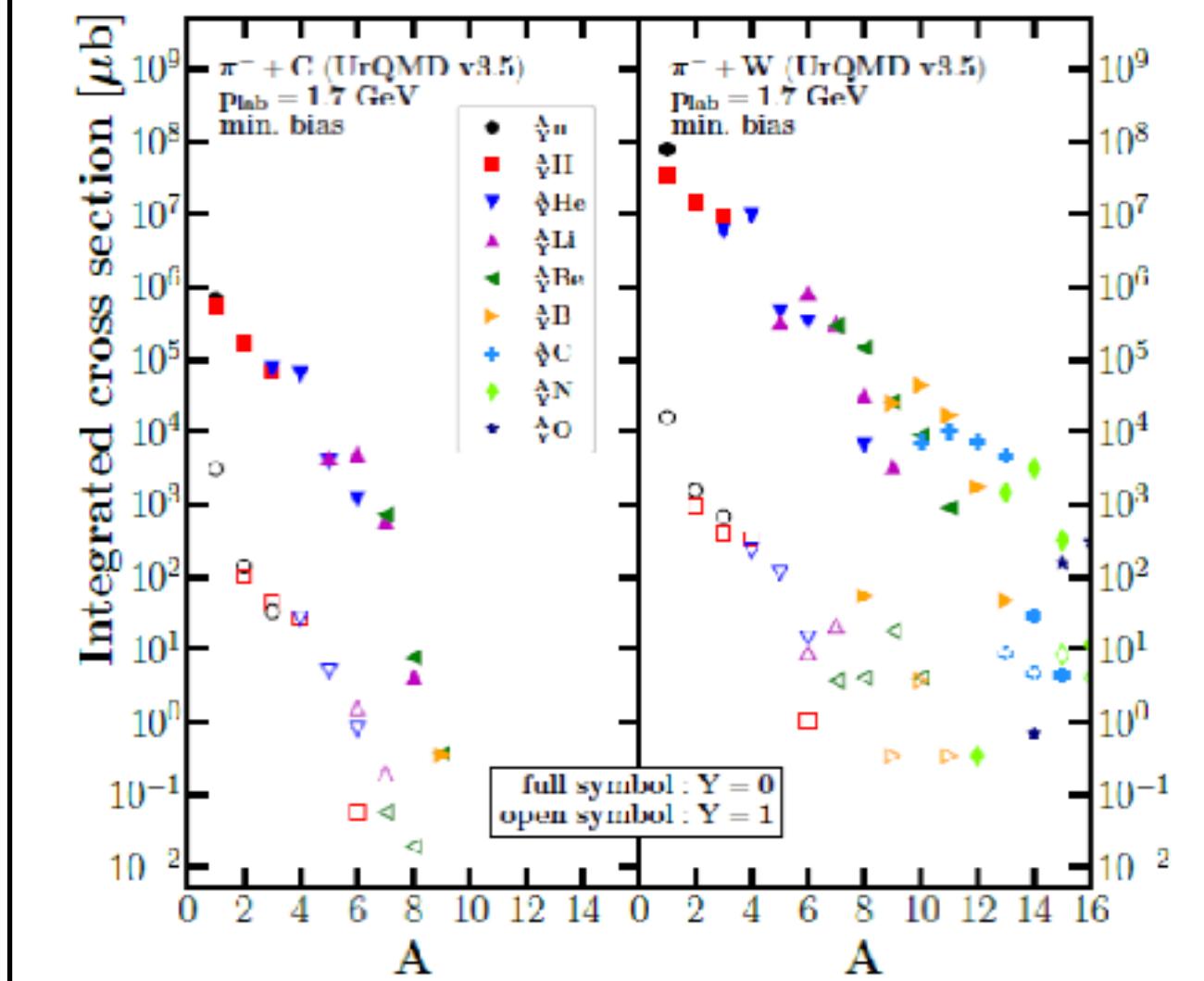
## Hyper-triton formation (RHIC-BES):

- UrQMD w/ coalescence
- Transport dynamics provide more robust description of data than a thermal model approach



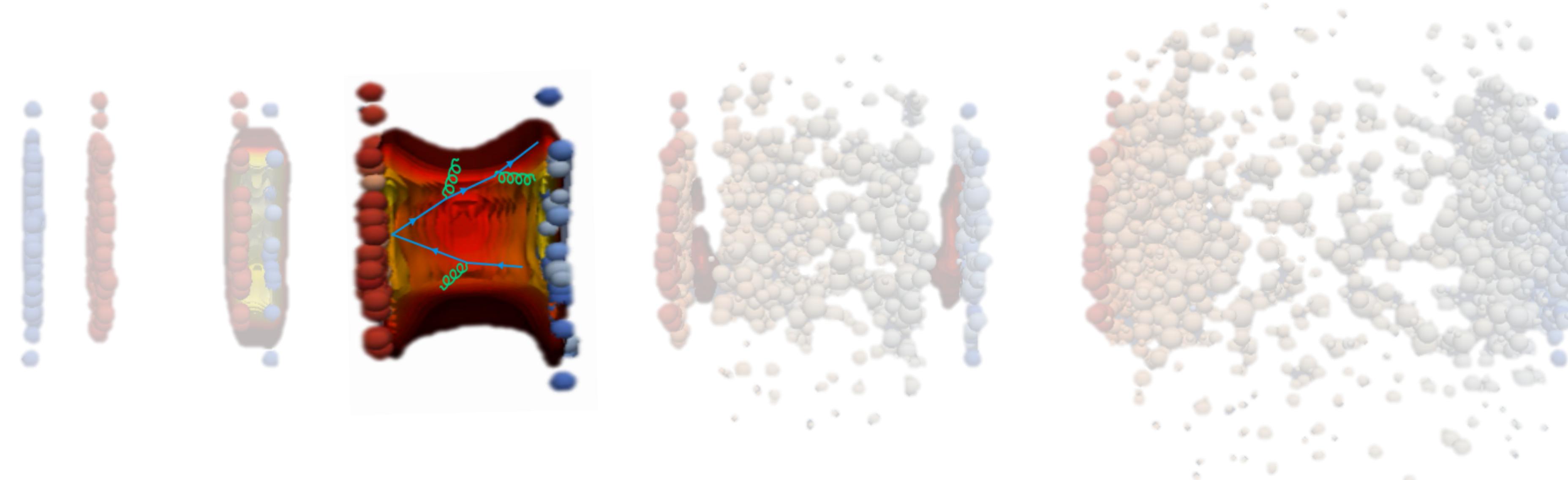
## Hyper-nuclei in $\pi+A$ at SIS:

- UrQMD predicts  $\pi+A$  collisions may produce a large variety of hyper-nuclei at SIS
- allows for study of the strange matter EoS



- G. Coci, S. Glaessel, V. Kireyeu, J. Aichelin, C. Blume, E. Bratkovskaya, V. Kolesnikov & V. Voronyuk: Phys. Rev. C108 (2023) 014902
- T. Reichert, J. Steinheimer, V. Vovchenko, B. Doenigus & M. Bleicher: Phys. Rev. C 107 (2023) 014912
- A. Kittiratpattana, T. Reichert, N. Buyukcizmeci, A. Botvina, A. Limphirat, C. Herold, J. Steinheimer & M. Bleicher: Phys. Rev. C 109 (2024) 044913

# Heavy Quark Transport



## RFD with conserved charm current:

- fluid-dynamical description of charm evolution
- heavy quarks are treated as a conserved current after initial production in hard processes

## Boltzmann dynamics:

- medium constituents: thermal light partons
- heavy quarks scatter with medium partons & radiate gluons based on pQCD matrix elements

## Langevin dynamics:

- no assumptions on medium constituents
- heavy quarks get frequent kicks from the medium → transport coefficients

## Hybrid models

- Improved Langevin (with radiative energy loss)
- Lido - Linearized Boltzmann with diffusion model
- MATTER/LBT: multi-scale energy-loss

# Langevin with Radiative Processes

modify Langevin Eqn. with force term due to gluon radiation:

$$\frac{d\vec{p}}{dt} = -\eta_D(p)\vec{p} + \vec{\xi} + \vec{f}_g \quad \left\{ \begin{array}{l} \text{radiation force defined through rate of} \\ \text{radiated gluon momenta:} \\ \vec{f}_g = \frac{d\vec{p}}{dt} \end{array} \right.$$

- same noise correlator and fluctuation-dissipation relation still hold:

$$\eta_D(p) = \frac{\kappa}{2TE} \quad \text{and} \quad \langle \xi^i(t) \xi^j(t') \rangle = \kappa \delta^{ij} \delta(t - t')$$

- gluon radiation calculated in Higher Twist formalism:

$$\frac{dN_g}{dx dk_\perp^2 dt} = \frac{2\alpha_s(k_\perp)}{\pi} P(x) \frac{\hat{q}}{k_\perp^4} \sin^2 \left( \frac{t - t_i}{2\tau_f} \right) \left( \frac{k_\perp^2}{k_\perp^2 + x^2 M^2} \right)^4$$

Guo & Wang: *PRL* 85, 3591  
Majumder: *PRD* 85, 014023  
Zhang, Wang & Wang:  
*PRL* 93, 072301

- relevant transport coefficients are now:

$$D = \frac{t}{M\eta_D(0)} = \frac{2T^2}{\kappa} \quad \text{and} \quad \hat{q} = 2\kappa C_A/C_F$$

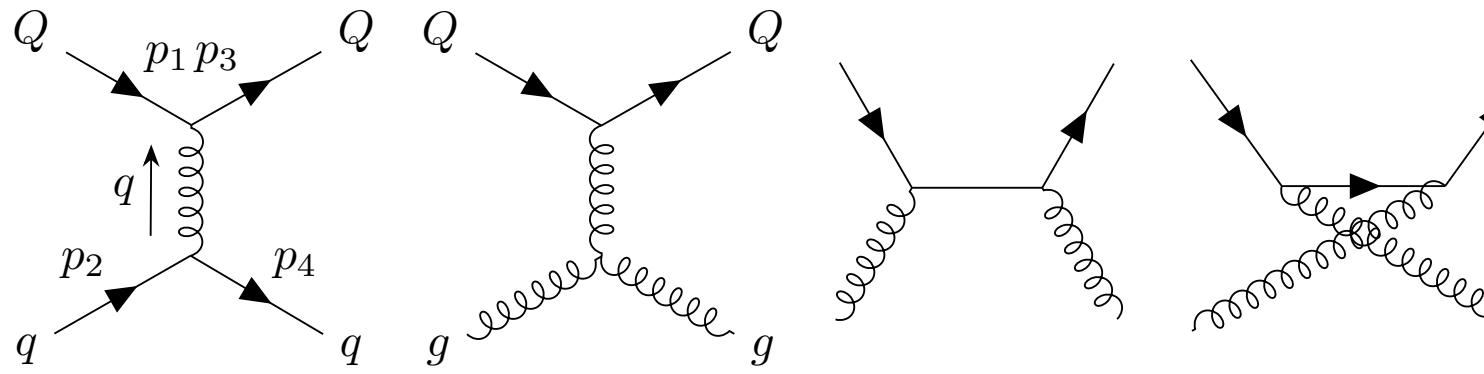
# Lido: Boltzmann + Langevin Hybrid

Combine the strength of the linearized-Boltzmann and Langevin approaches:

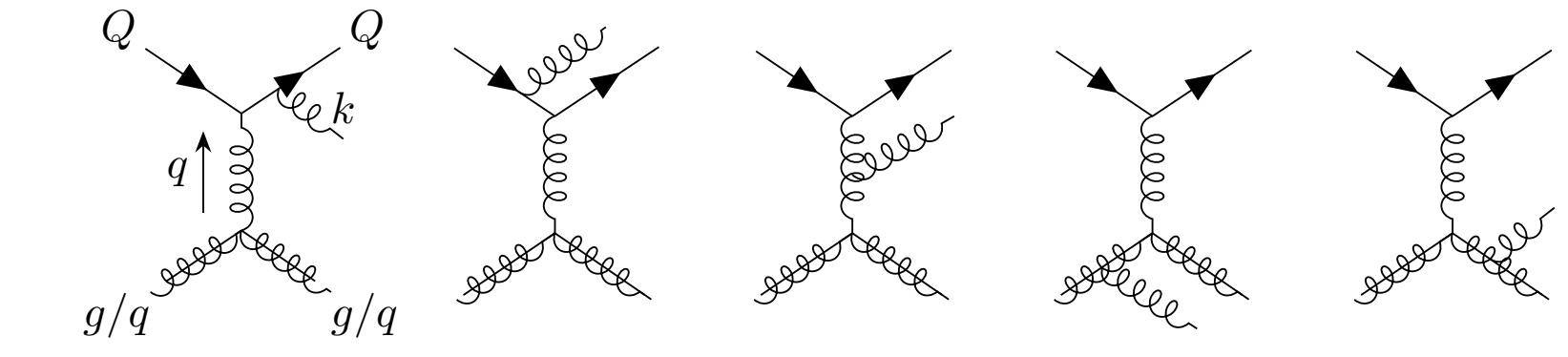
$$\frac{p \cdot \partial f_Q}{E} = \mathcal{C}[f_Q] - \frac{\partial}{\partial p_i} \left( A_i - \frac{1}{2} \frac{\partial}{\partial p_j} B_{ij} \right) f_Q = (\hat{\mathcal{C}} + \hat{\mathcal{D}}) f_Q$$

## perturbative processes:

- elastic scattering:



- inelastic: improved Gunion Bertsch



Fochler et al. PRD88 014018

- gluon radiation and absorption implemented to conserve detailed balance

## non-perturbative processes:

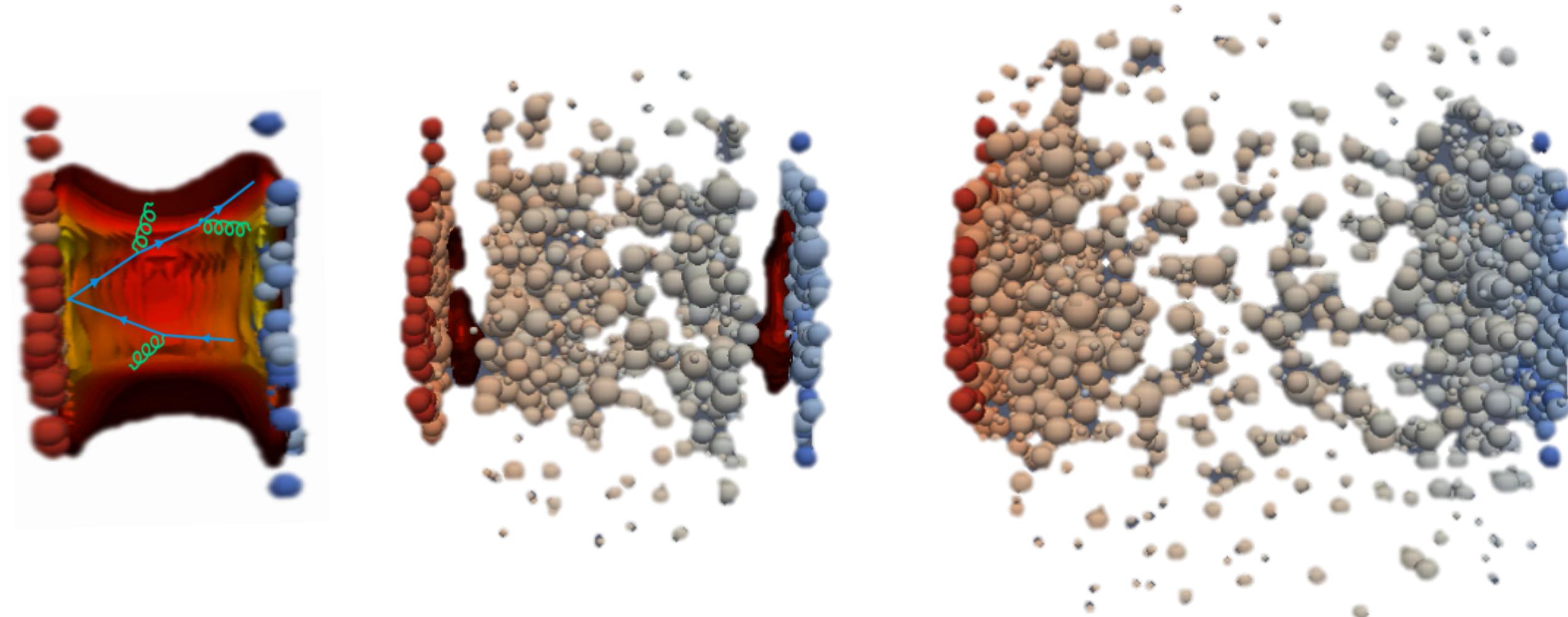
- treated in a Langevin equation with isotropic random force

$$\Delta \vec{x}_i = \frac{\vec{p}_i}{E} \Delta t \quad \Delta \vec{p}_i = -\eta_D \vec{p}_i \Delta t + \Delta t \vec{\xi}_i(t)$$

- Einstein relation connects random force to drag coefficient to ensure proper equilibrium

# Heavy Quark Transport: full collision dynamics

parameterized initial  
QGP state  
+  
initial HQ production



## Trento:

- based on simple phenomenological ideas for entropy deposition
- constrained by global model to data fit

## Heavy Quarks:

- PYTHIA to generate initial HQ ensemble

## viscous hydrodynamics

- EbE 2+1D viscous RFD
- describes QGP dynamics & hadronization
- Lattice QCD EoS

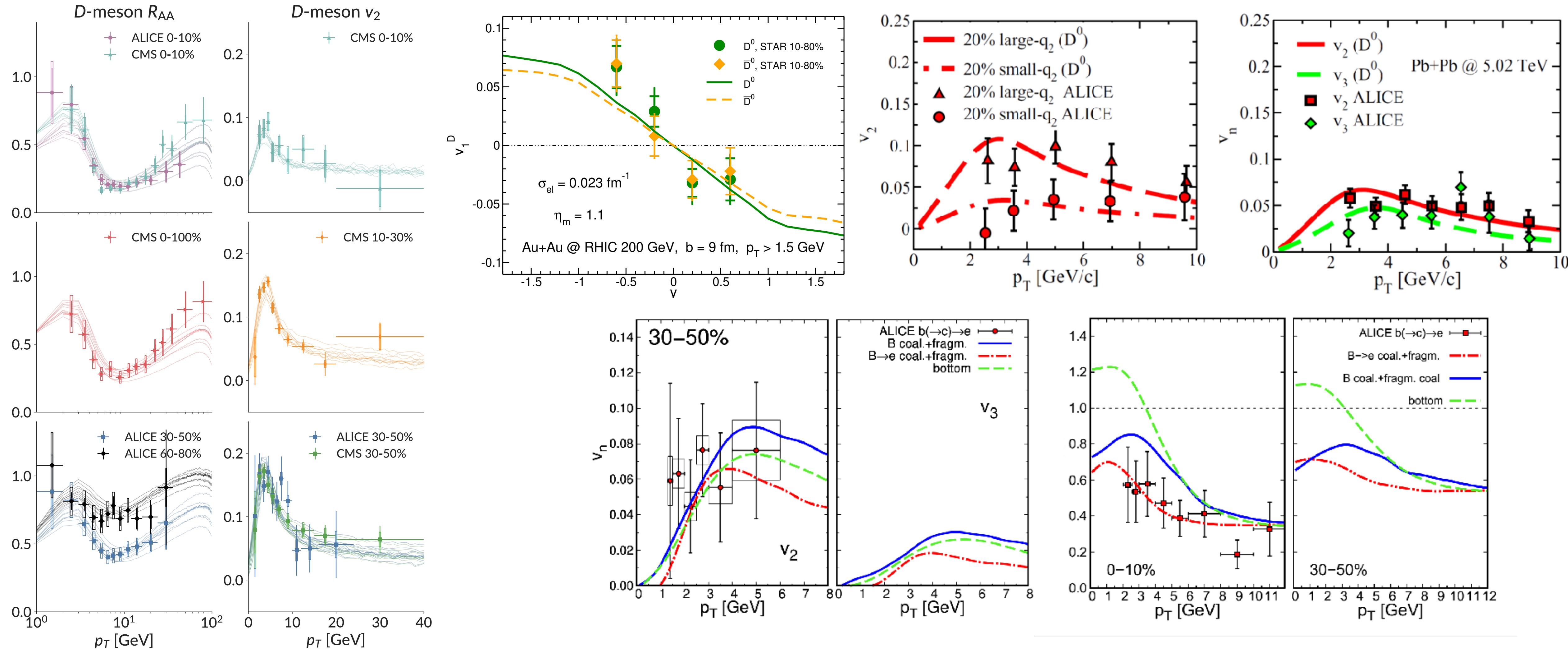
## HQ interaction & transport:

1.  $D_s$  from Lattice QCD
2. T-Matrix approach
3. Bayesian calibration
1. Langevin dynamics
2. Boltzmann dynamics
3. Hybrid approaches

## Hadronic afterburner:

- non-equilibrium evolution of an interacting hadron gas
- separation of chemical and kinetic freeze-out
- hadron gas shear & bulk viscosities are implicitly contained in calculation

# Heavy Quark Transport: comparison to data



- Y. Xu, J.E. Bernhard, S.A. Bass, M. Nahrgang & G-Y. Qin: Phys. Rev. C97 (2018) 014907
- L. Oliva, S. Plumari and V. Greco, JHEP 05 (2021) 034
- M.L. Sambataro, V. Minissale, S. Plumari, V. Greco, Phys.Lett.B 849 (2024) 138480
- M.L. Sambataro, Y. Sun, V. Minissale, S. Plumari, V. Greco, Eur.Phys.J.C 82 (2022) 9, 833

# Collaborative research: key to advancement

PHYSICAL REVIEW C 99, 054907 (2019)

## Toward the determination of heavy-quark transport coefficients in quark-gluon plasma

Shanshan Cao,<sup>1</sup> Gabriele Coci,<sup>2,3</sup> Santosh Kumar Das,<sup>4,2</sup> Weiyao Ke,<sup>5</sup> Shuai Y. F. Liu,<sup>6</sup> Salvatore Plumari,<sup>2</sup> Taesoo Song,<sup>7</sup> Yingru Xu,<sup>5</sup> Jörg Aichelin,<sup>8</sup> Steffen Bass,<sup>5</sup> Elena Bratkovskaya,<sup>9,10</sup> Xin Dong,<sup>11</sup> Pol Bernard Gossiaux,<sup>8</sup> Vincenzo Greco,<sup>2,3</sup> Min He,<sup>12</sup> Marlene Nahrgang,<sup>8</sup> Ralf Rapp,<sup>6</sup> Francesco Scardina,<sup>2,3</sup> and Xin-Nian Wang<sup>13,11,\*</sup>

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<sup>2</sup>Department of Physics and Astronomy, University of Catania, Via Santa Sofia 64, I-95125 Catania, Italy

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<sup>4</sup>School of Physical Science, Indian Institute of Technology Goa, Ponda, Goa, India

<sup>5</sup>Department of Physics, Duke University, Durham, North Carolina 27708, USA

<sup>6</sup>Cyclotron Institute and Department of Physics and Astronomy, Texas A&M University, College Station, Texas 77843, USA

<sup>7</sup>Institut für Theoretische Physik, Universität Gießen, Germany

<sup>8</sup>SUBATECH, IMT Atlantique, Université de Nantes, CNRS-IN2P3, Nantes, France

<sup>9</sup>Institute for Theoretical Physics, Johann Wolfgang Goethe Universität, Frankfurt am Main, Germany

<sup>10</sup>GSI Helmholtzzentrum für Schwerionenforschung GmbH, Darmstadt, Germany

<sup>11</sup>Nuclear Science Division, Lawrence Berkeley National Laboratory, Berkeley, California 94740, USA

<sup>12</sup>Department of Applied Physics, Nanjing University of Science and Technology, Nanjing 210094, China

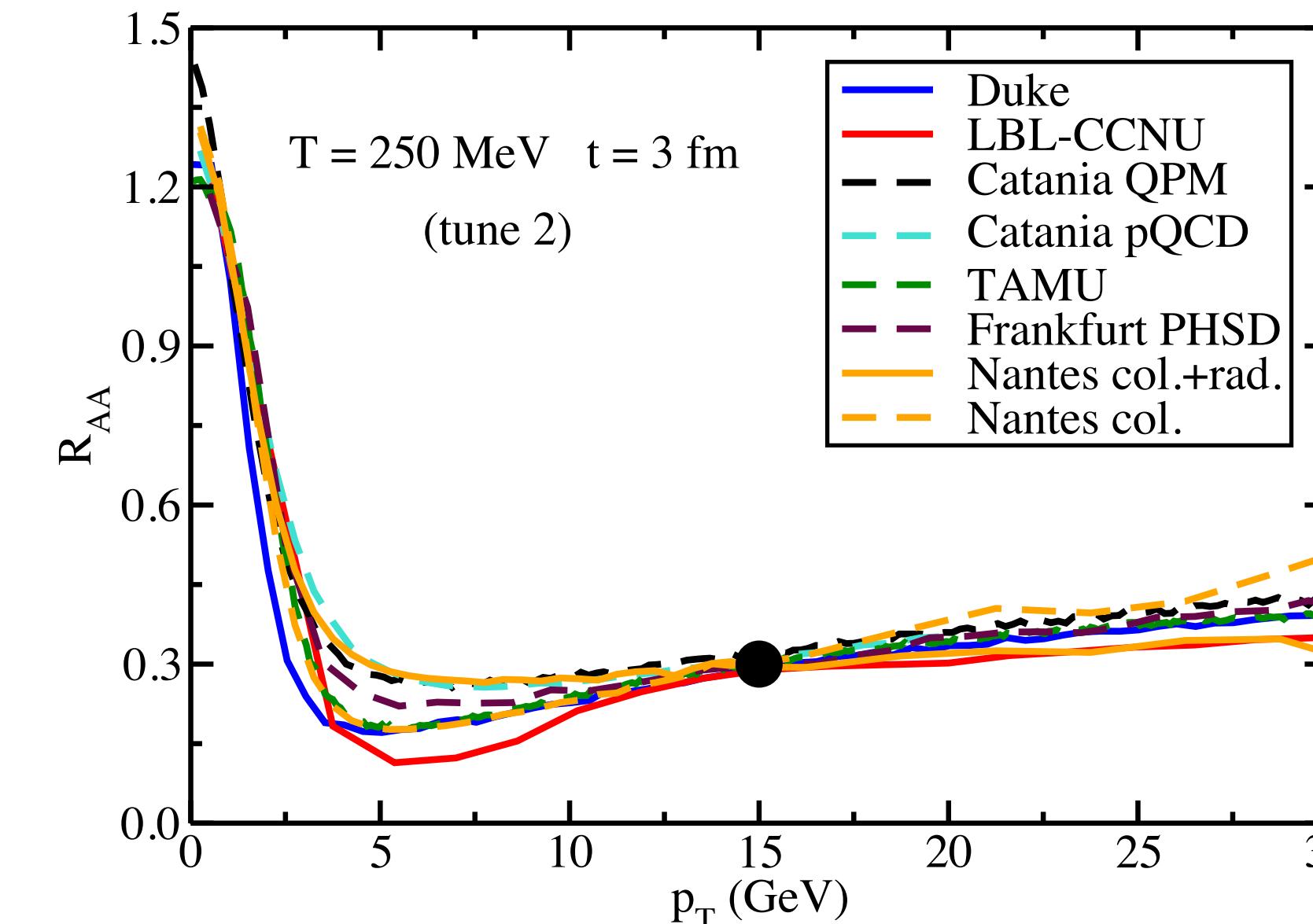
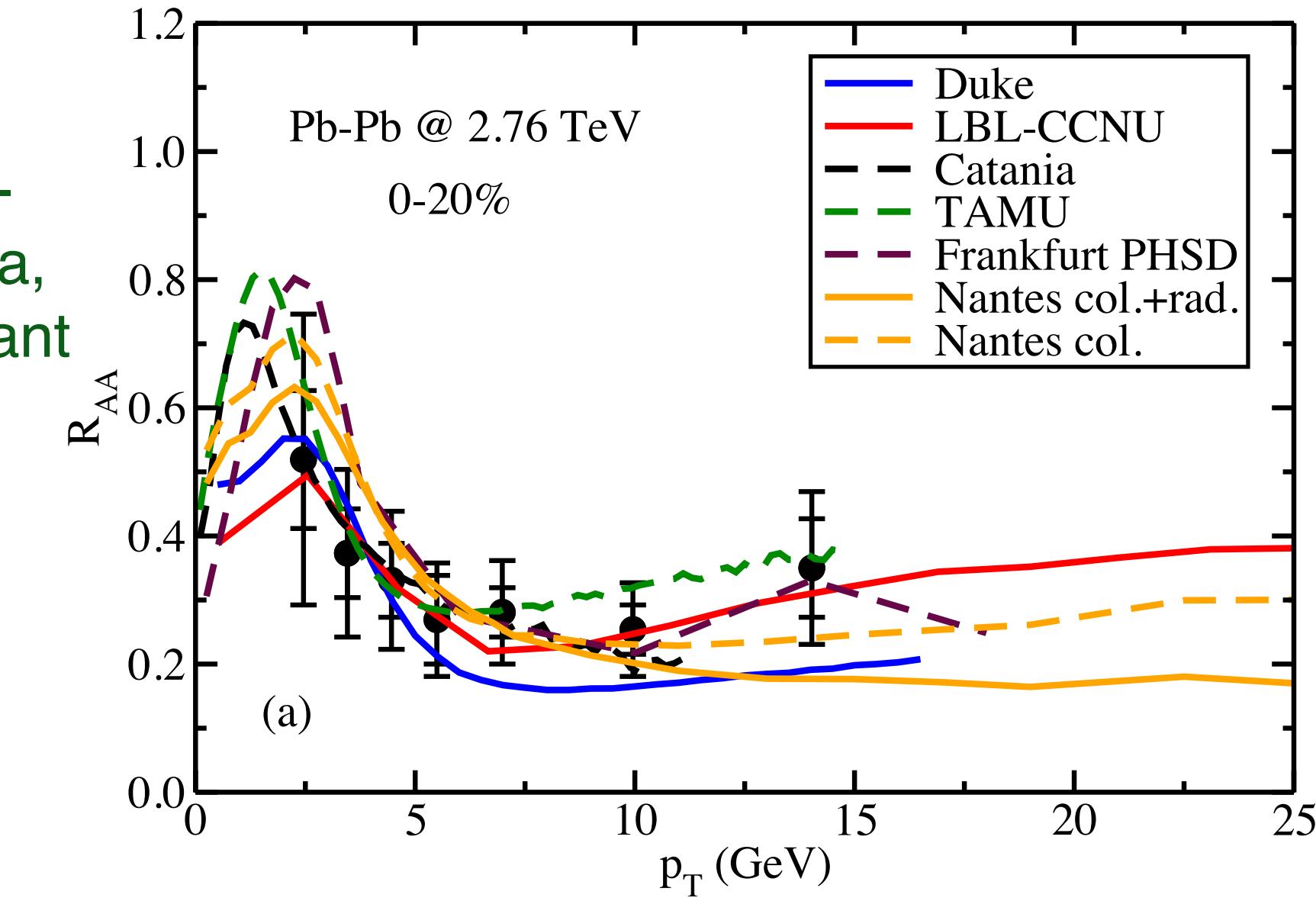
<sup>13</sup>Key Laboratory of Quark and Lepton Physics (MOE) and Institute of Particle Physics, Central China Normal University, Wuhan 430079, China

(Received 24 September 2018; published 28 May 2019)

Several transport models have been employed in recent years to analyze heavy-flavor meson spectra in high-energy heavy-ion collisions. Heavy-quark transport coefficients extracted from these models with their default parameters vary, however, by up to a factor of 5 at high momenta. To investigate the origin of this large theoretical uncertainty, a systematic comparison of heavy-quark transport coefficients is carried out between various transport models. Within a common scheme devised for the nuclear modification factor of charm quarks in a brick medium of a quark-gluon plasma, the systematic uncertainty of the extracted drag coefficient among these models is shown to be reduced to a factor of 2, which can be viewed as the smallest intrinsic systematical error band achievable at present time. This indicates the importance of a realistic hydrodynamic evolution constrained by bulk hadron spectra and of heavy-quark hadronization for understanding the final heavy-flavor hadron spectra and extracting heavy-quark drag coefficient. The transverse transport coefficient is less constrained due to the influence of the underlying mechanism for heavy-quark medium interaction. Additional constraints on transport models such as energy loss fluctuation and transverse-momentum broadening can further reduce theoretical uncertainties in the extracted transport coefficients.

DOI: 10.1103/PhysRevC.99.054907

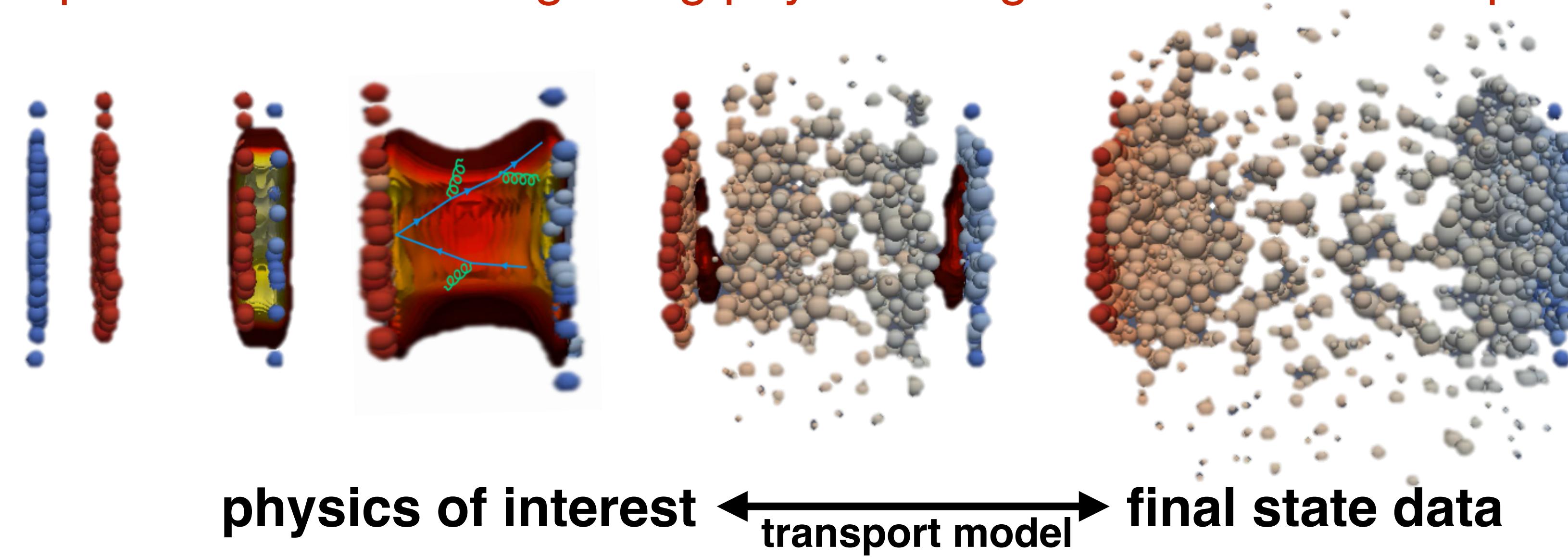
- models show reasonable agreement to D-meson RAA data, but with significant variations



- reduce calculation to quark level with same transport coefficient at fixed temperature & fix normalization: very good agreement!
- details (different medium evolution, ICs etc.) can lead to significant variation!

# Transport as discovery tool:

- The goal of constructing a transport model is to test its underlying assumptions via a comparison to data and gaining physics insight from such a comparison



- provides access to quantities that are not directly accessible by experiment:
  - QCD transport coefficients (viscosities, diffusion coefficients etc.) & relaxation times
  - structure of the initial state
- probe the underlying physics of observed phenomena
  - jet energy-loss
  - dynamics of thermalization & hadronization

# Model to Data Comparison: Parametric Nightmare

Transport Models have multiple parameters encoding its underlying physics that are sensitive to experimental data

## Model Parameter:

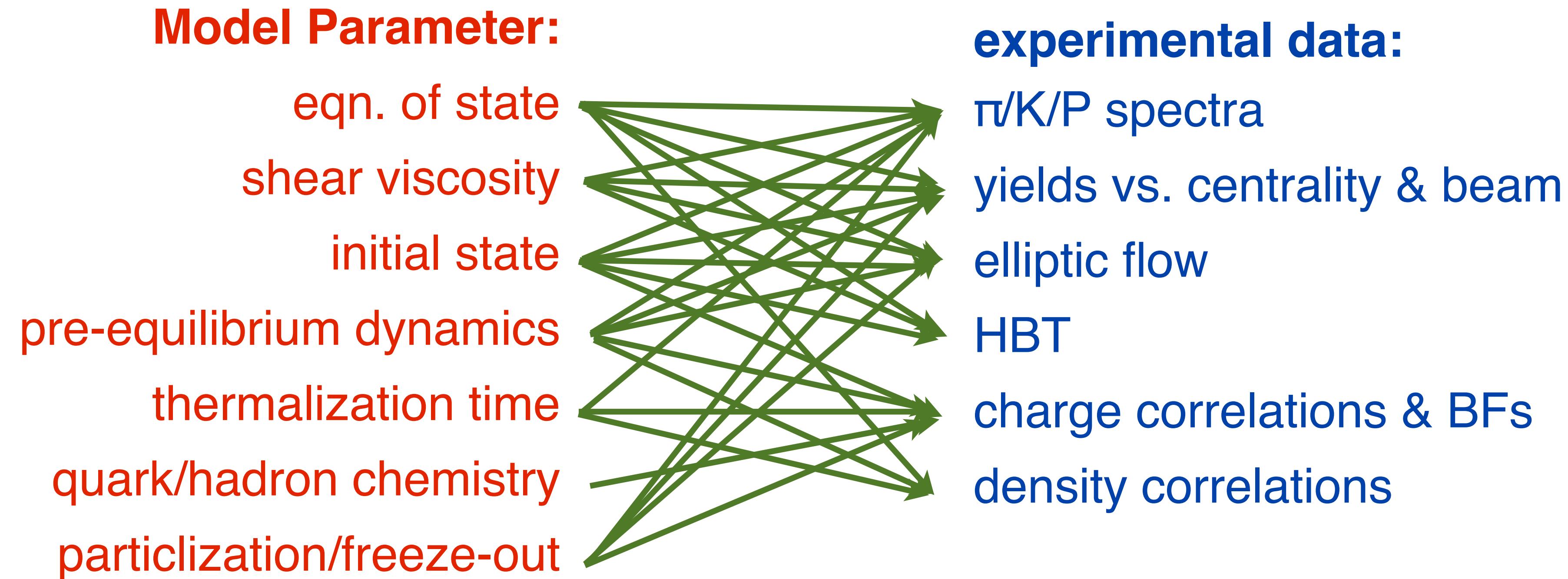
- eqn. of state
- shear viscosity
- initial state
- pre-equilibrium dynamics
- thermalization time
- quark/hadron chemistry
- particlization/freeze-out

## experimental data:

- $\pi/K/P$  spectra
- yields vs. centrality & beam
- elliptic flow
- HBT
- charge correlations & BFs
- density correlations

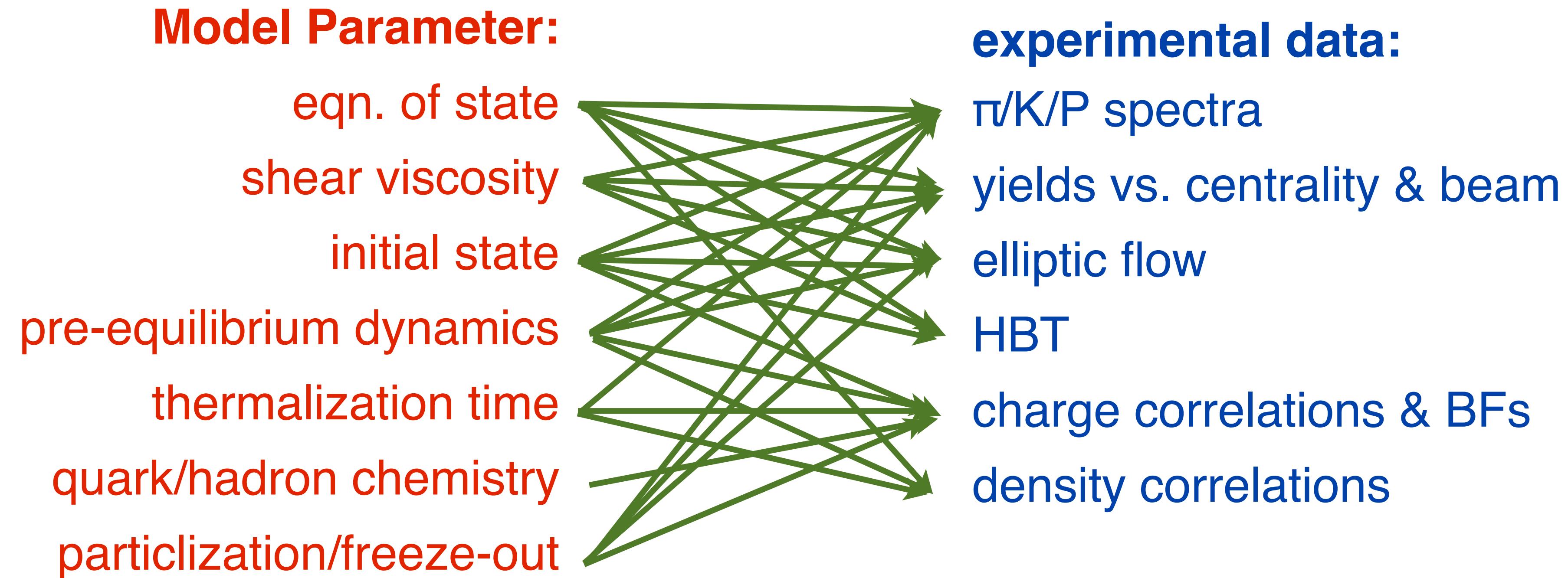
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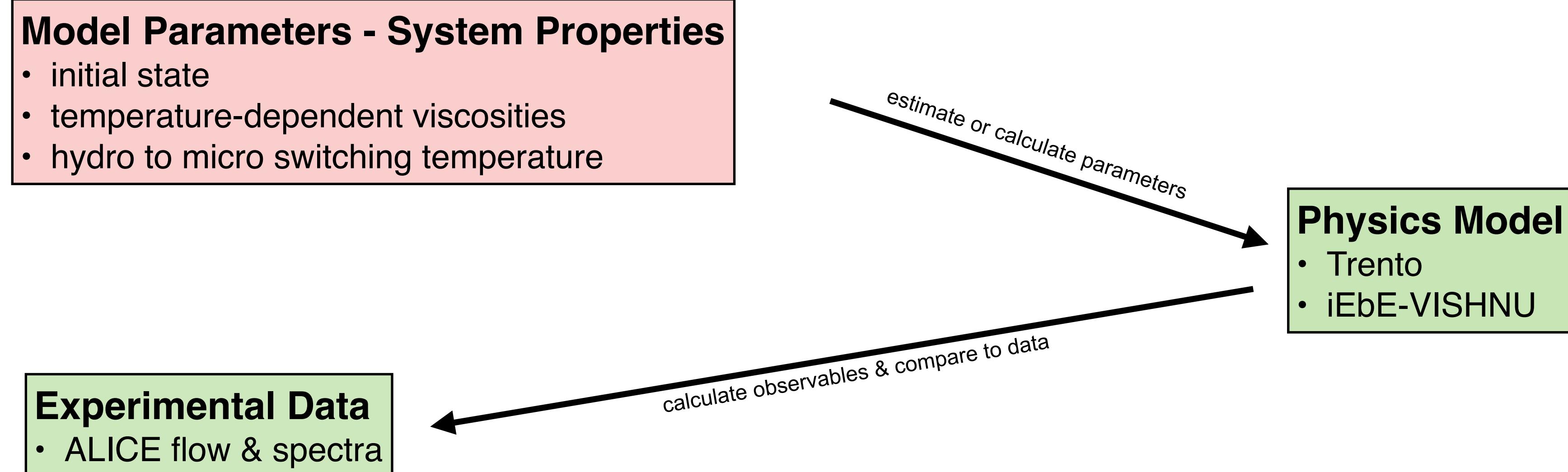


- large number of interconnected parameters w/ non-factorizable data dependencies
- data have correlated uncertainties
- develop novel optimization techniques: Bayesian Statistics and MCMC methods
- transport models require too much CPU: need new techniques based on emulators
- general problem, not restricted to RHIC Physics

→**collaboration with Statistical Sciences**

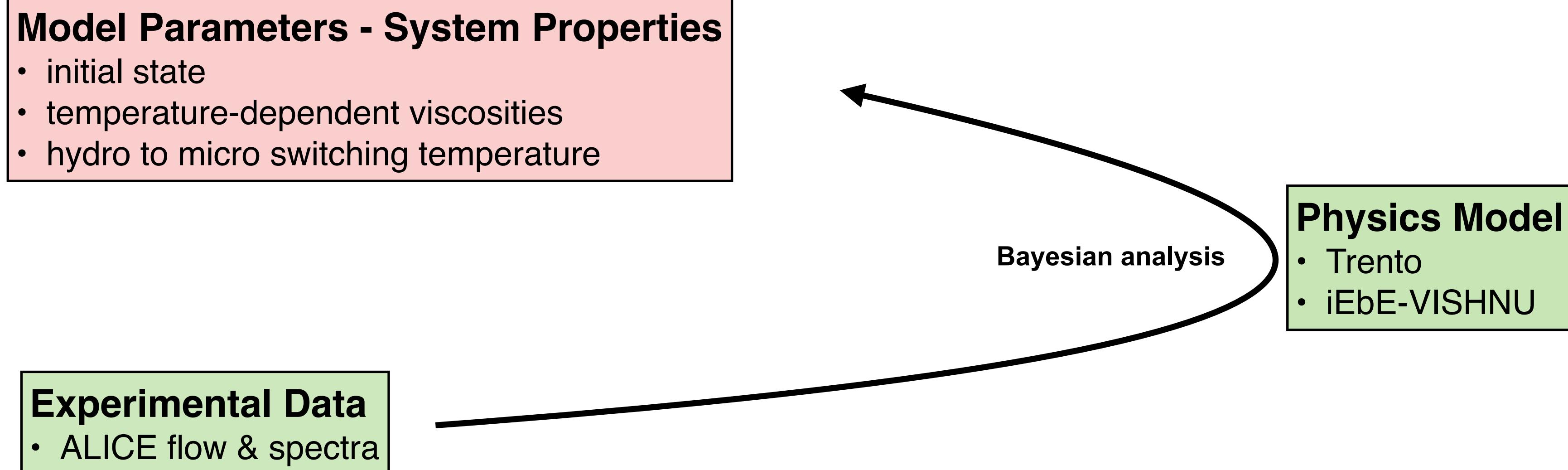
# State of the Art for model to data comparisons: Bayesian inference

Each computational model relies on a set of physics parameters to describe the dynamics and properties of the system. These physics parameters act as a representation of the information we wish to extract from comparison to data.



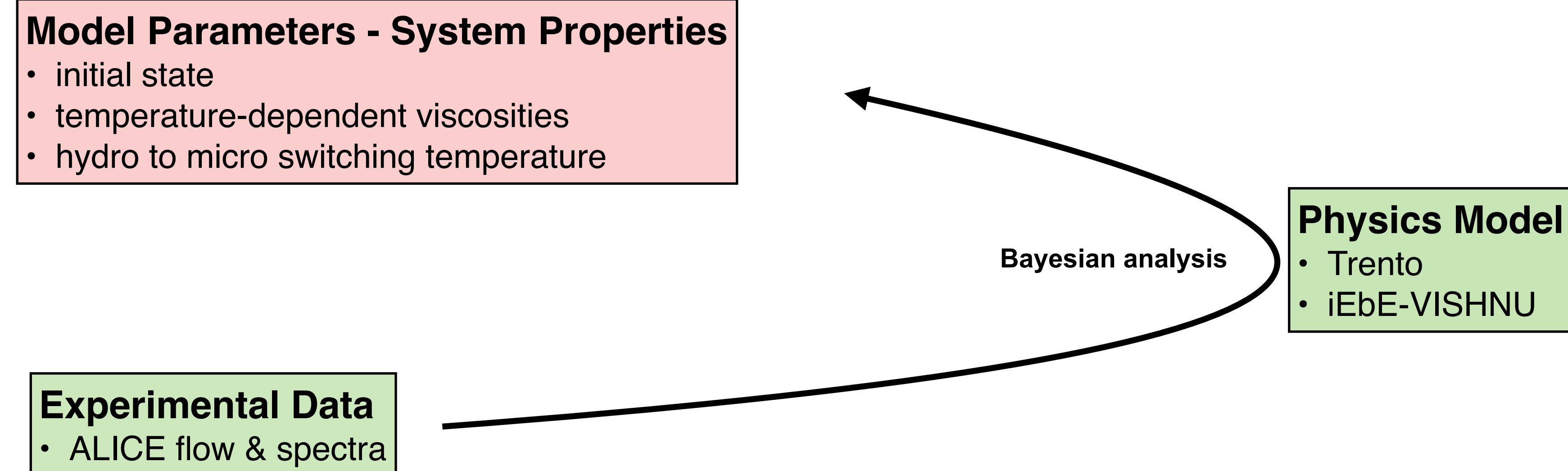
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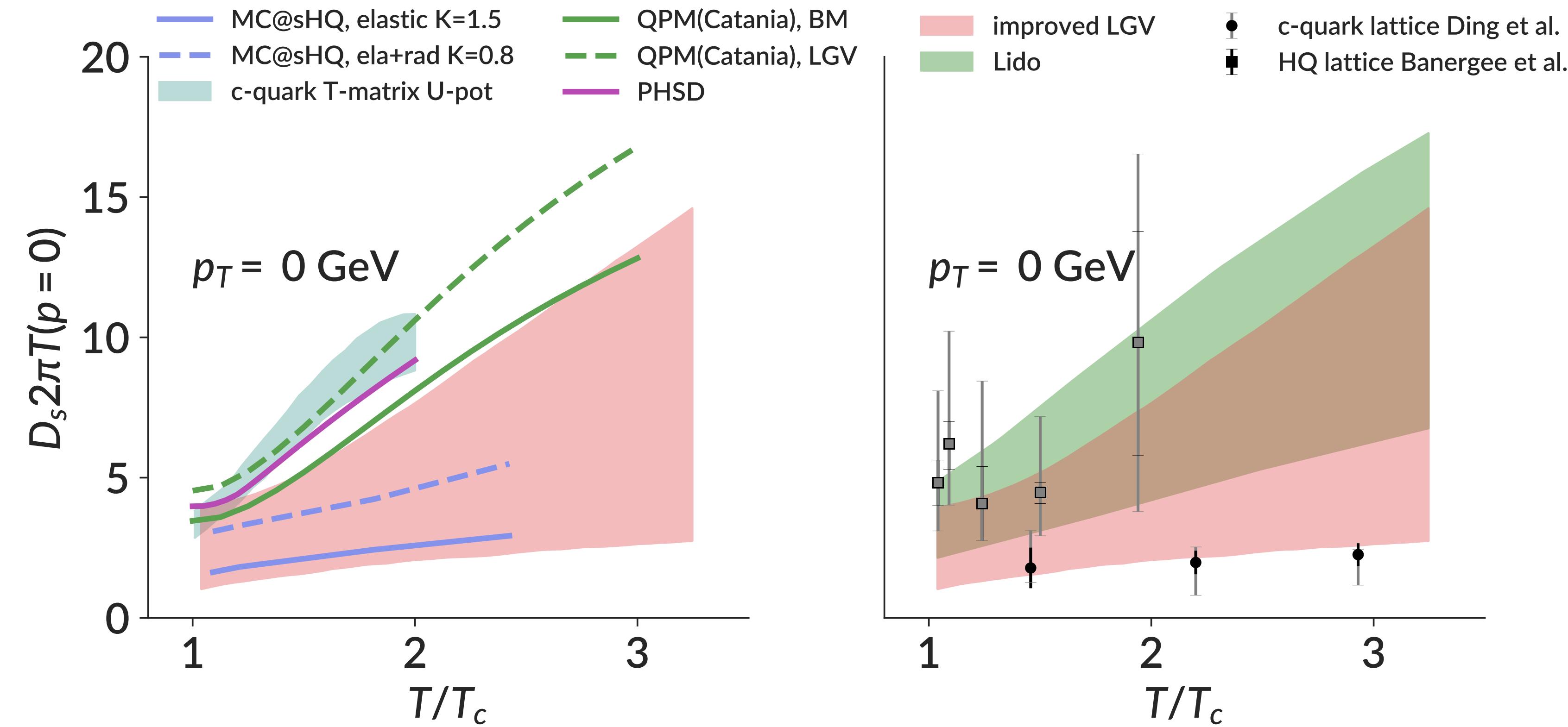
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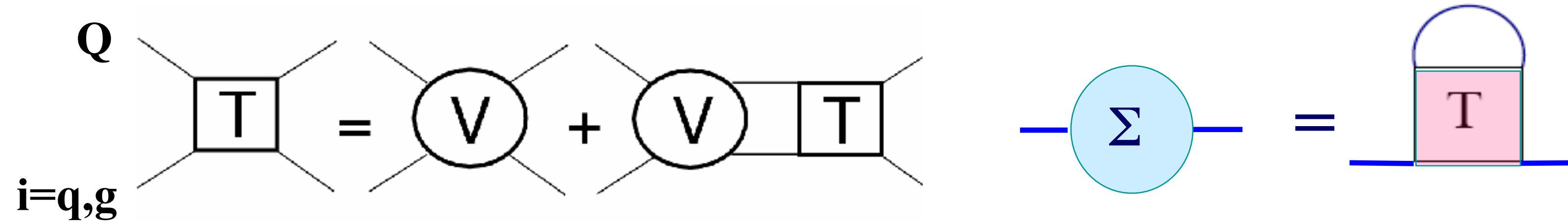
- Bayesian analysis allows us to simultaneously calibrate all model parameters via a model-to-data comparison
- determine parameter values such that the model best describes experimental observables
- extract the probability distributions of all parameters

# Bayesian analysis: heavy flavor transport coefficient



- comparison shows large variability in  $D_s$  between different heavy quark transport/interaction models
- Lattice results favor data-driven extraction (within large uncertainties)
- Lido vs. radiation improved Langevin:
  - large overlap in extracted  $D_s$  band
  - Lido trends to larger  $D_s$  values (influence of pQCD contribution in model)

# T-Matrix formalism for heavy quark interactions (TAMU)

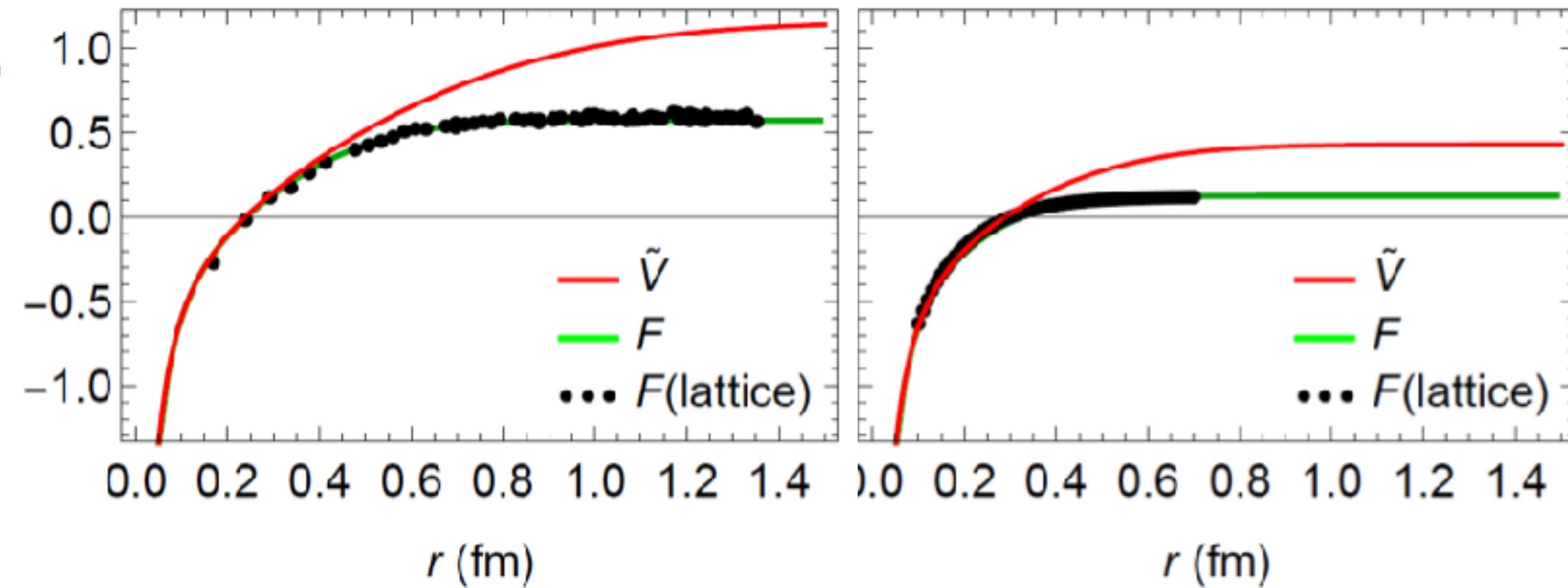
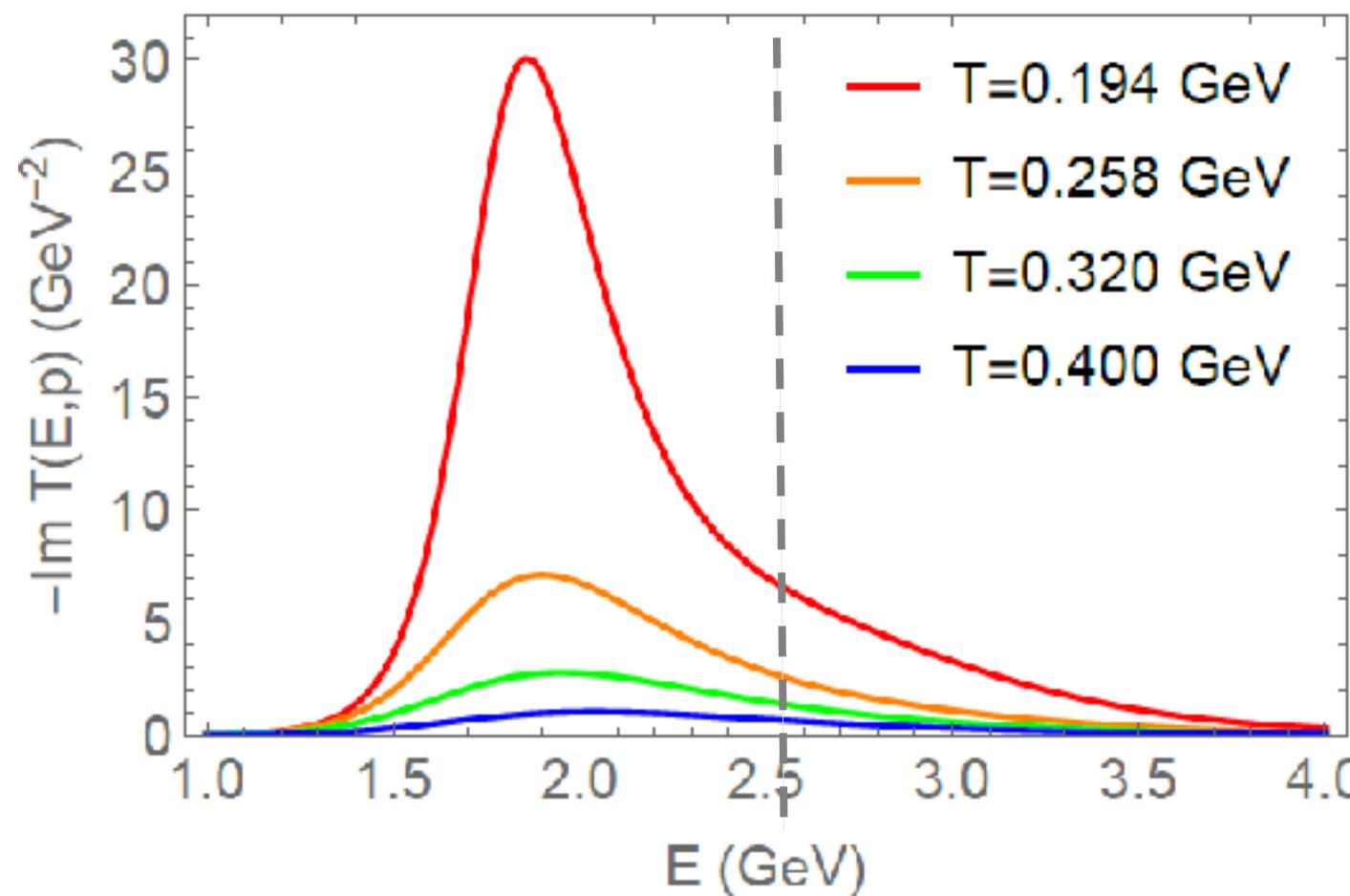


- **strong** coupling → resummation:  $T_{Qi} = V_{Qi} + \int V_{Qi} D_Q D_i T_{Qi}$

- **key input:** in-med potential  $V_{Qi}$ ,

constrained by lattice QCD

→ “string” force above  $T_c$



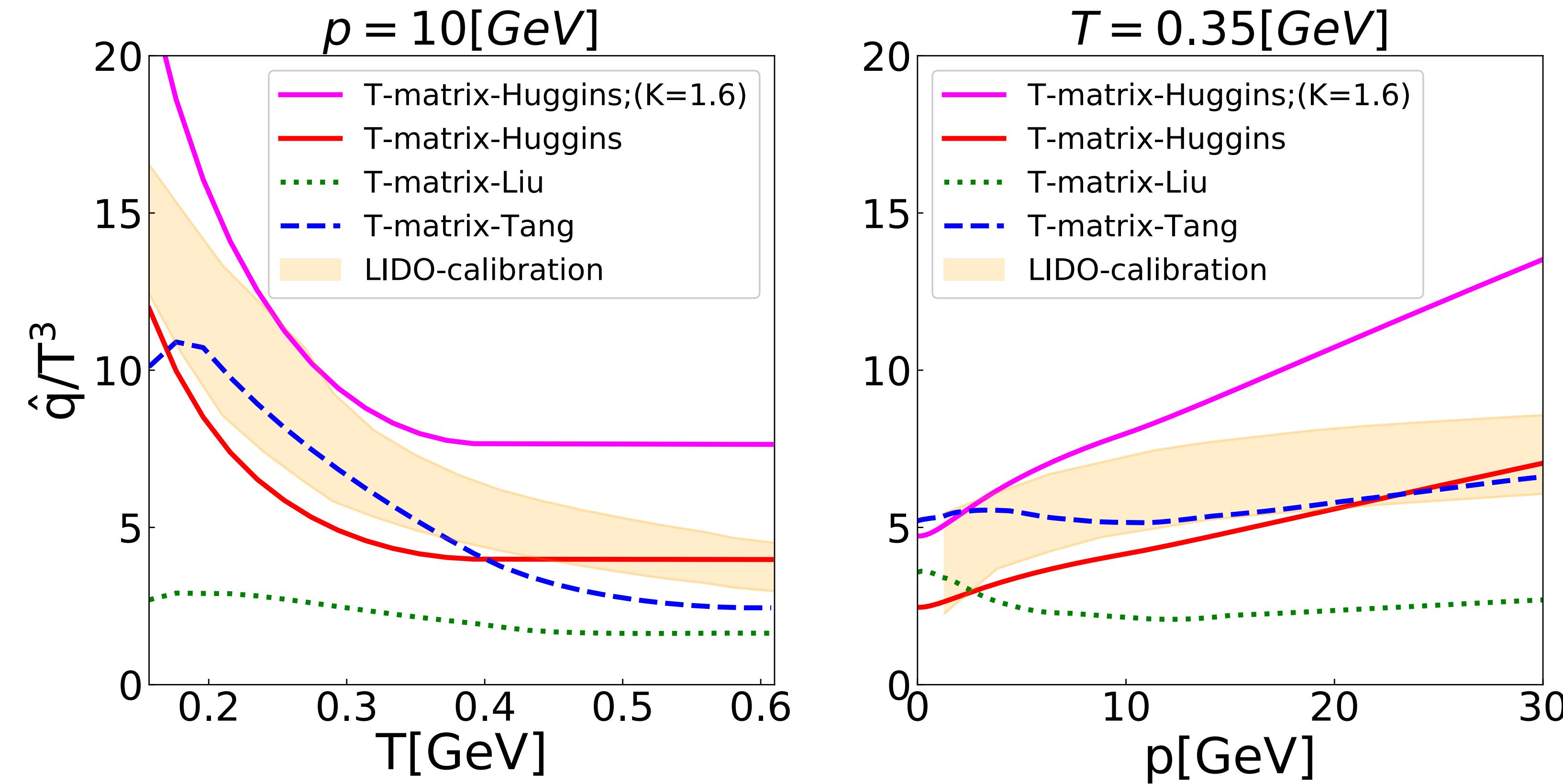
- Non-perturbative effects:
  - thermalization time  $\tau_c \approx 3\text{-}4 \text{ fm/c}$
  - hadronization

- K. Huggins & R. Rapp: Nucl. Phys. A896 (2012) 24-45
- S.Y.F. Liu & R. Rapp: Phys. Rev. C106 (2022) 055201
- Z. Tang, S. Mukherjee, P. Petreczky & R. Rapp: Eur. Phys. Journal A60 (2024) 92

# Transport as bridge between theory and data

Two methods for establishing a connection between theory and experiment:

1. Theory to Data: calculate transport coefficients, run evolution model, compare prediction to data
2. Data to Theory: parametrize transport coefficient, perform Bayesian calibration on data

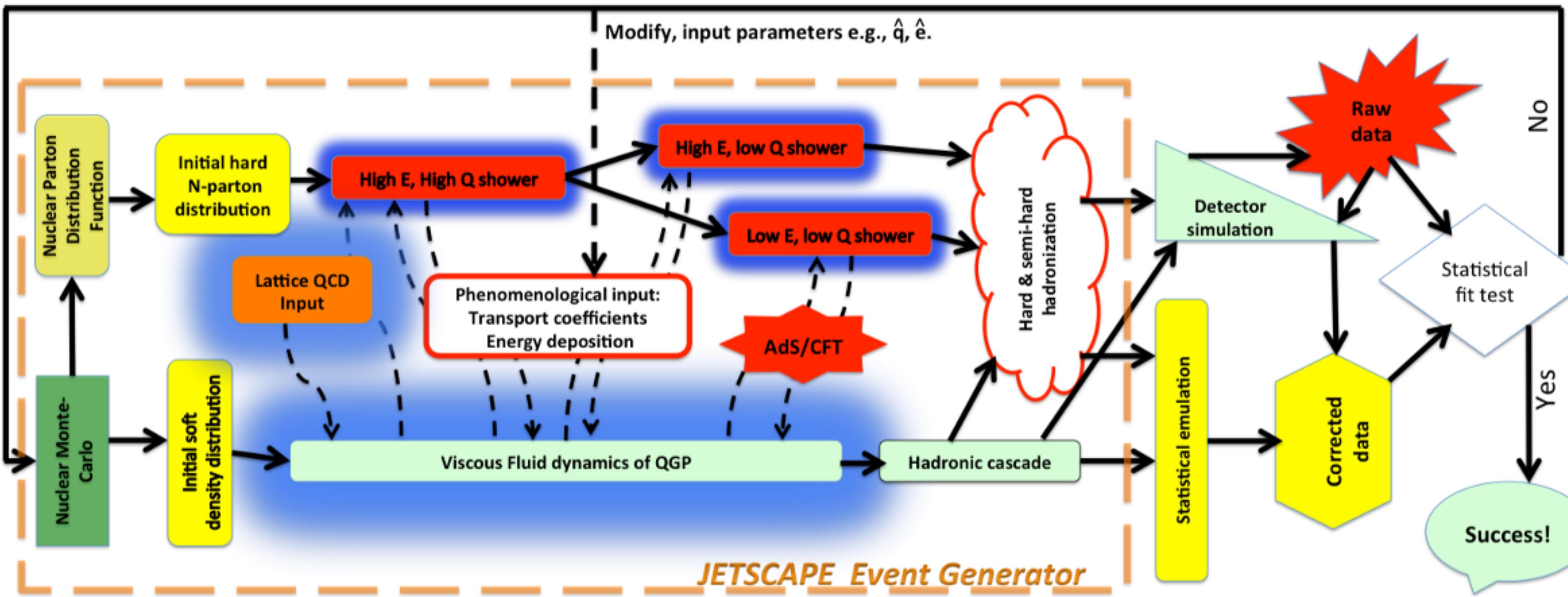


- comparison of theory-calculated transport coefficient to the calibration extraction allows for a quick assessment on whether the theory will be able to describe the data

# The Jetscape framework

## JETSCAPE: Jet Energy Loss Tomography with a Statistically and Computationally Advanced Program Envelope

- provide a tool (modular software library) to study the physics of energy-loss
- large area of research, many different approaches exist, no single group or PI has the capability to do them all
- collaboration of theoretical and experimental physicists, computer scientists and statisticians



- Trento (2+1) + free Streaming
- Medium evolution:
  - MUSIC (2+1, 3+1),
  - external reader
  - brick
  - Gubser
- Pythia8 (parton gun, string fragmentation)
- MATTER
- Martini
- AdS/CFT
- LBT
- Cooper Frye
- SMASH
- Custom and HepMC output

- JETSCAPE package interfaces with leading community tools that are publicly available and well-tested
- additional functions and codes can be linked as external modules (e.g. Lido) or utilize the framework for their own energy-loss kernel (e.g. Tequila)

# Heavy-Flavor Theory for QCD Matter (HEFTY)



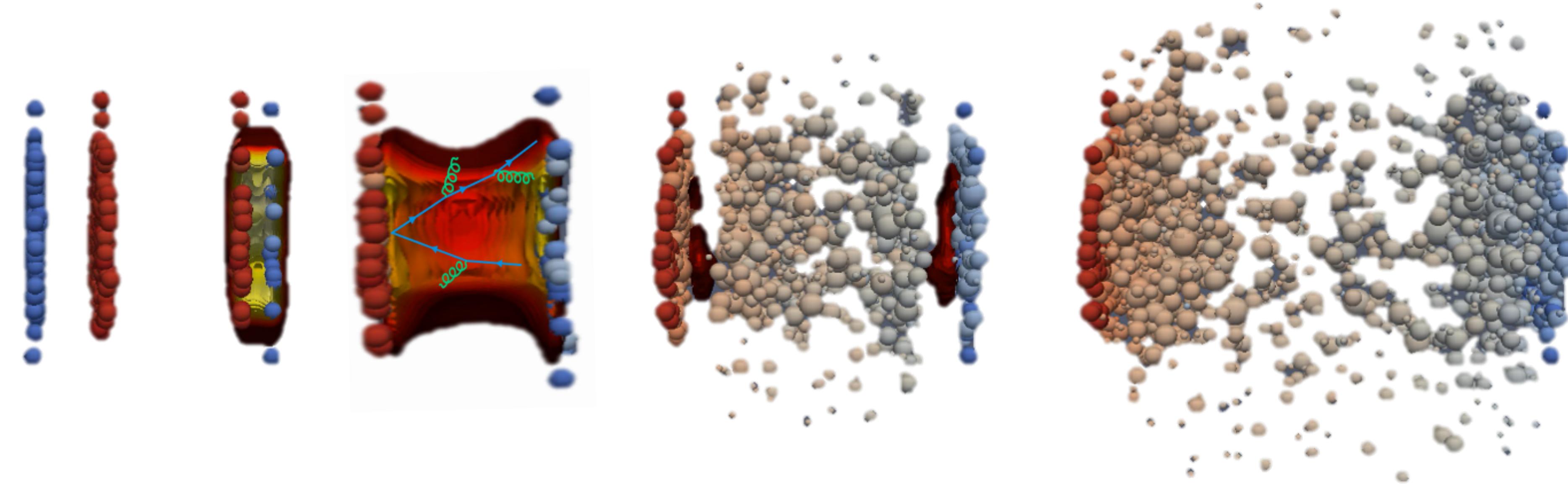
U.S. DEPARTMENT OF  
**ENERGY**

Office of  
Science

**Ralf Rapp**, Steffen A. Bass, Thomas Mehen, Swagato Mukherjee, Peter Petreczky, Jianwei Qiu,  
Mike Strickland, Ivan Vitev, Ramona Vogt, Yen-Jie Lee, Xin Dong and Anthony Frawley



# Concluding remarks



- Transport models are versatile tools to connect final state data to underlying physics phenomena and to extract physical quantities not directly accessible via measurements
- Transport theory provides a rich set of concepts to design models for the different epochs and regimes of excited QCD matter created in (relativistic) heavy-ion collisions
- Statistical tools (Bayesian analysis, ML) are crucial for model calibration and uncertainty quantification
- The future is bright - many more exciting applications, analyses and results to come!

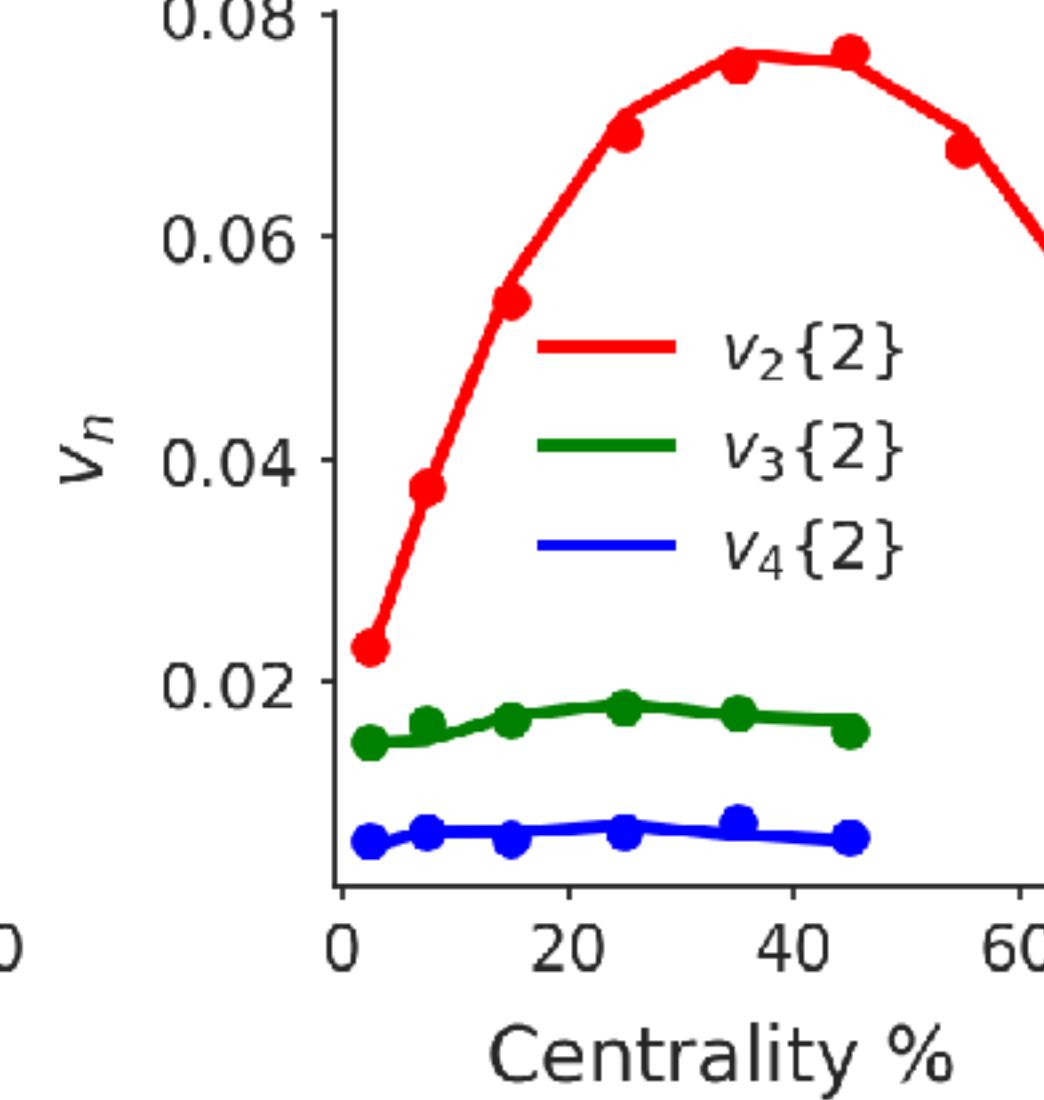
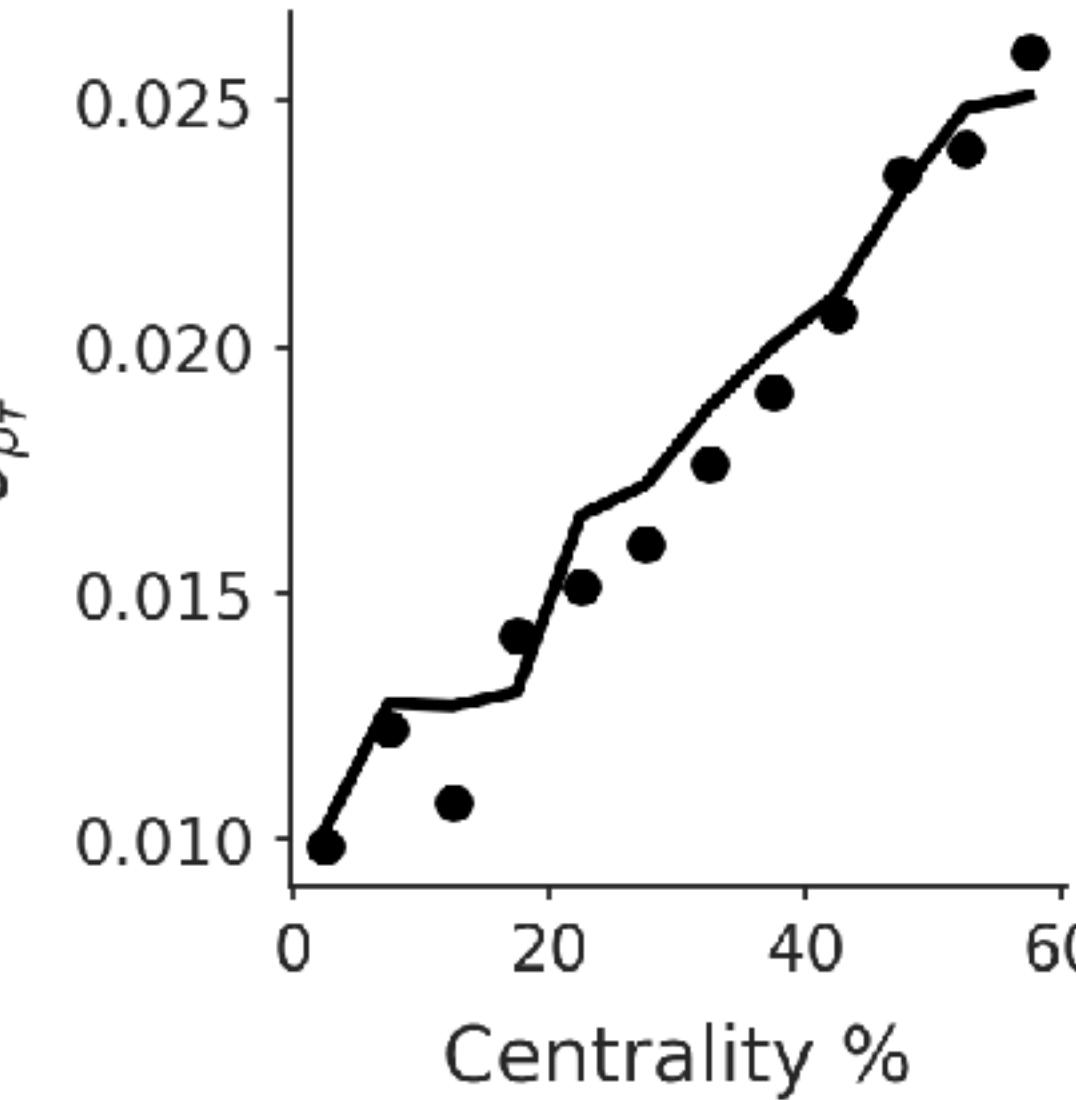
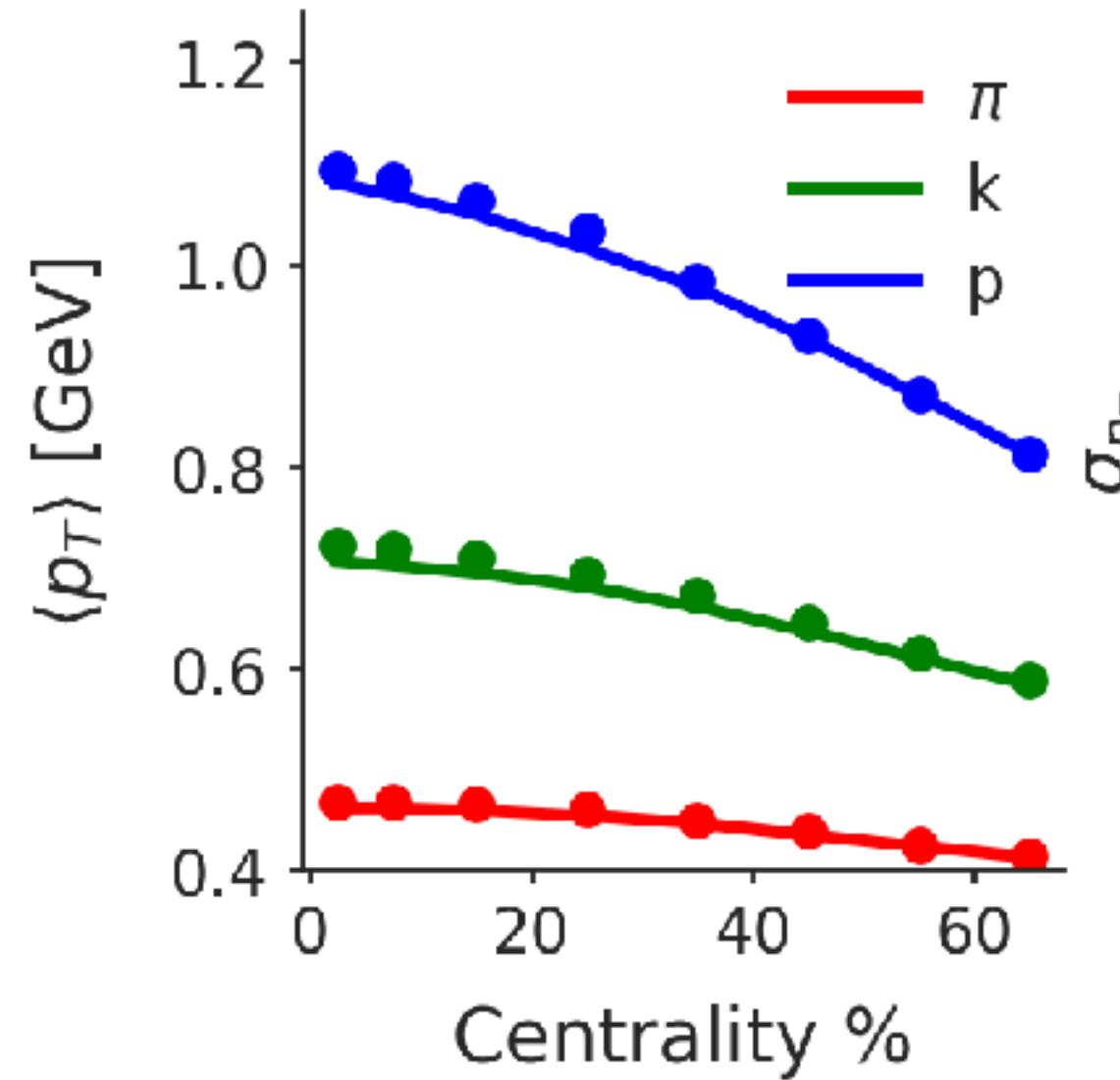
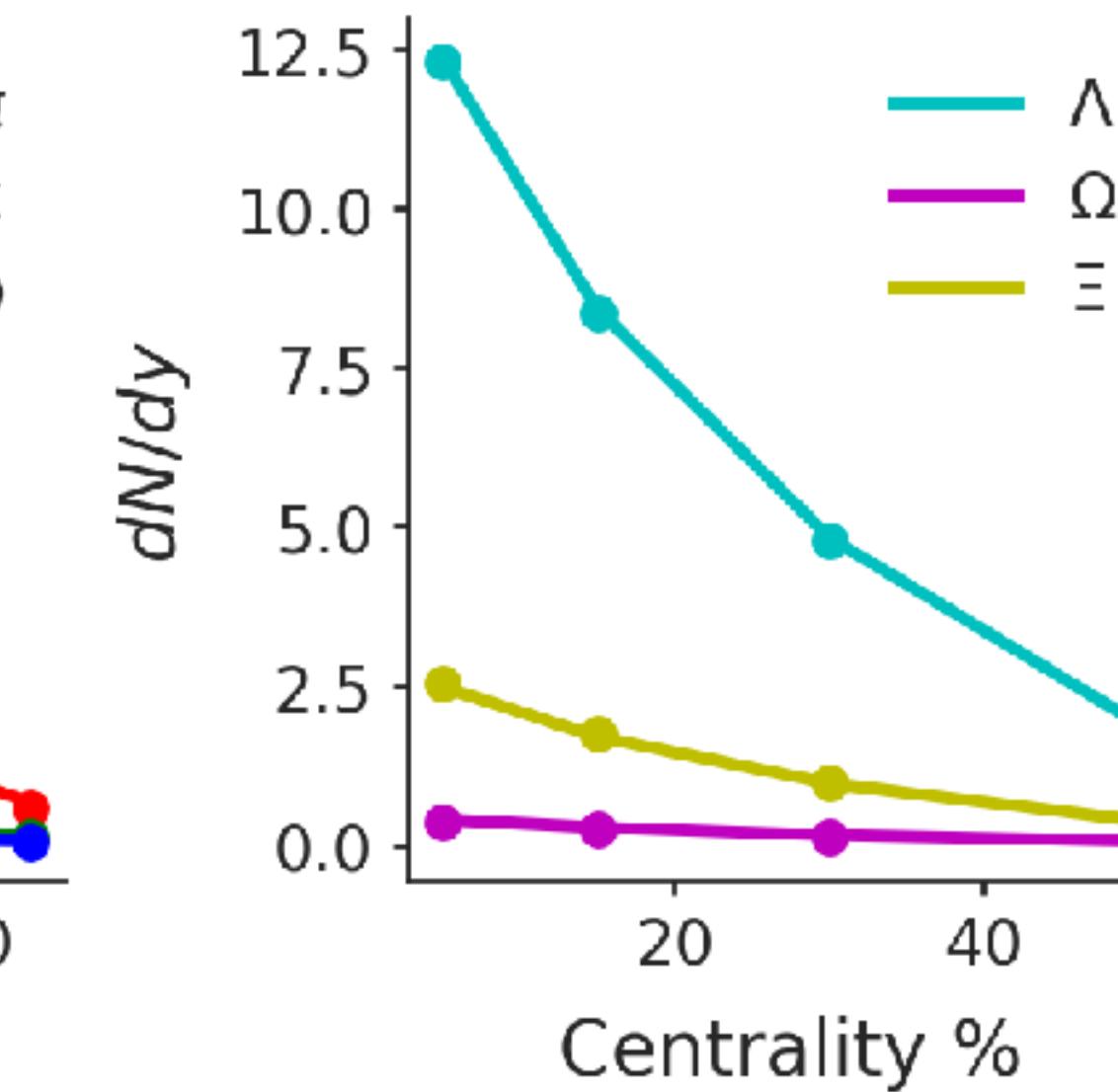
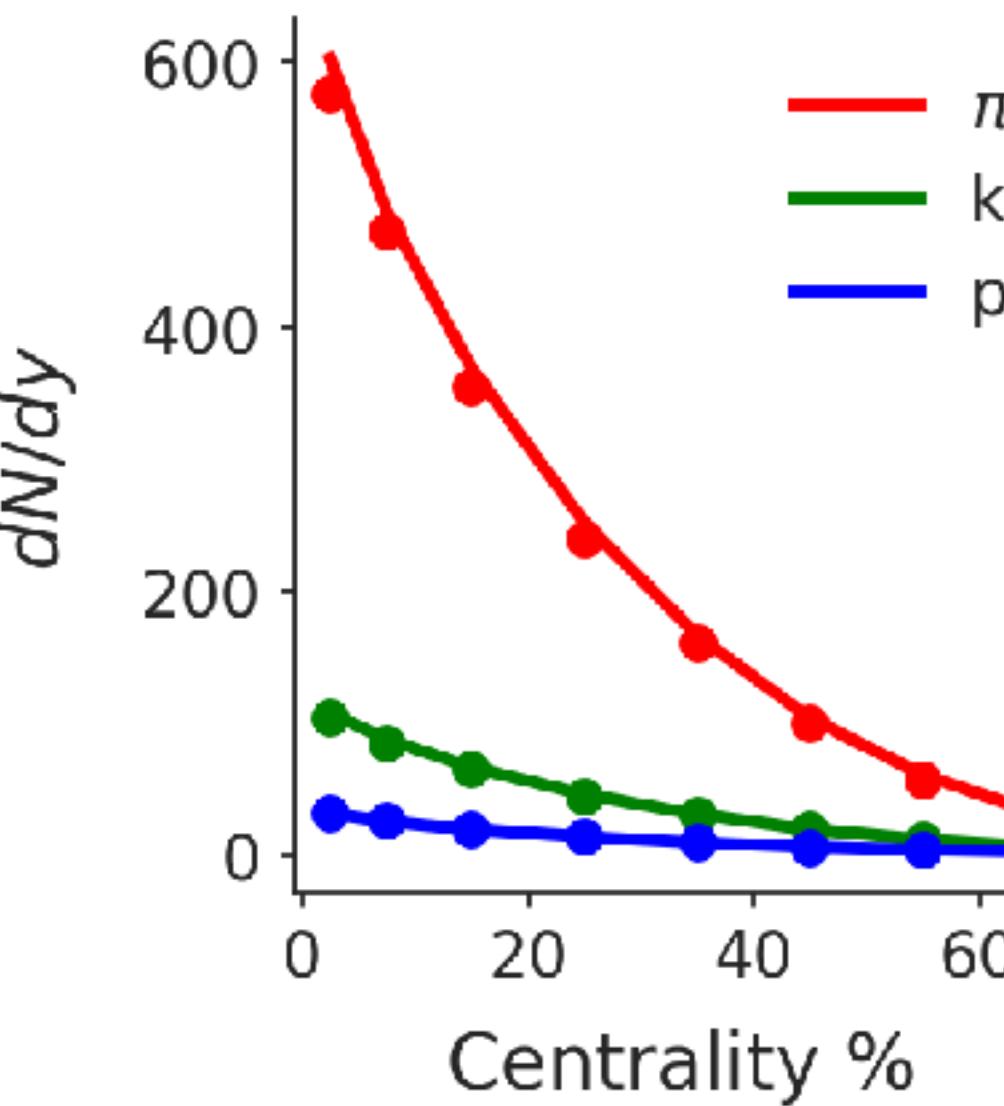
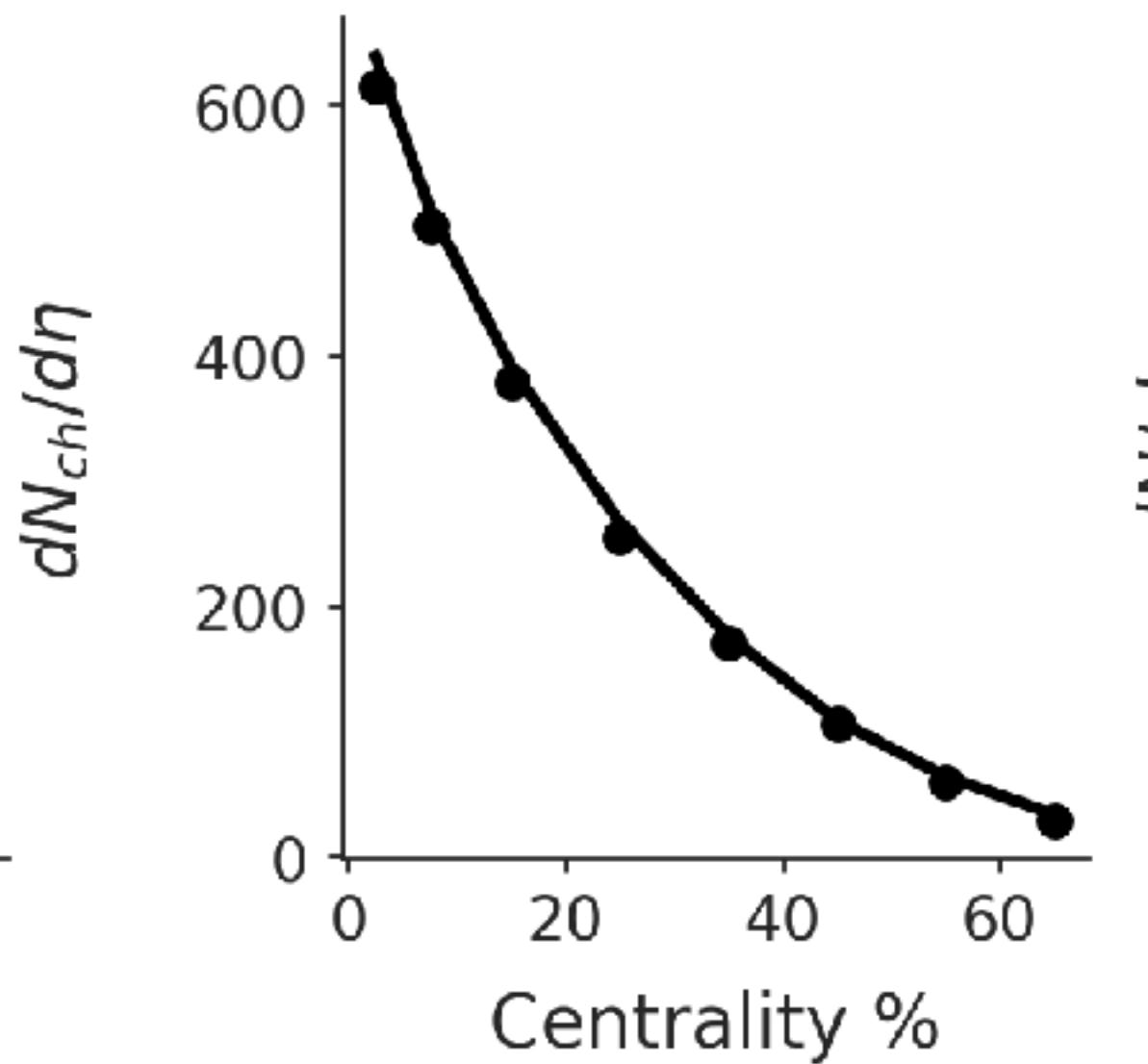
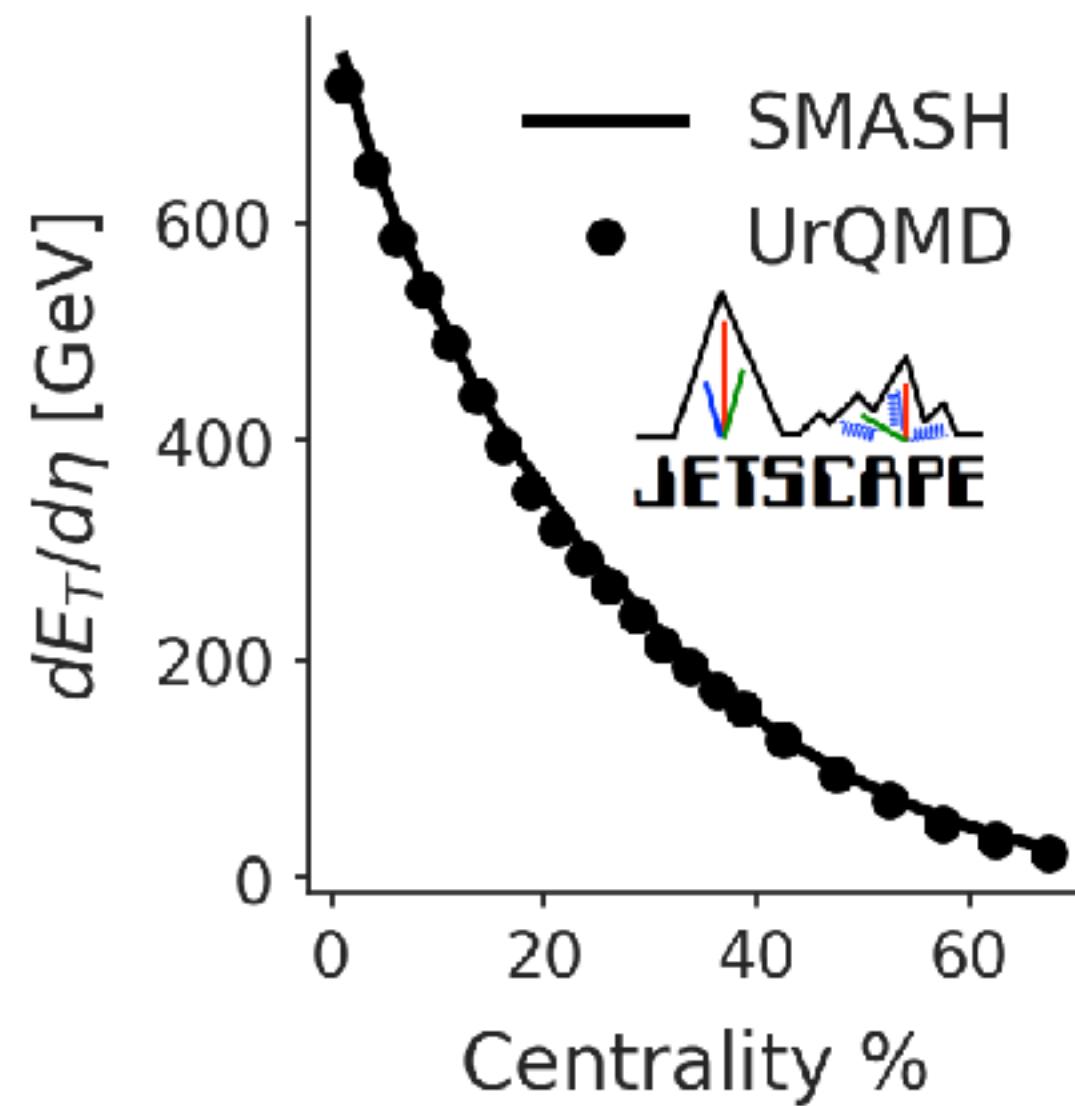
**The End**

# Transport Model vs. Monte-Carlo:

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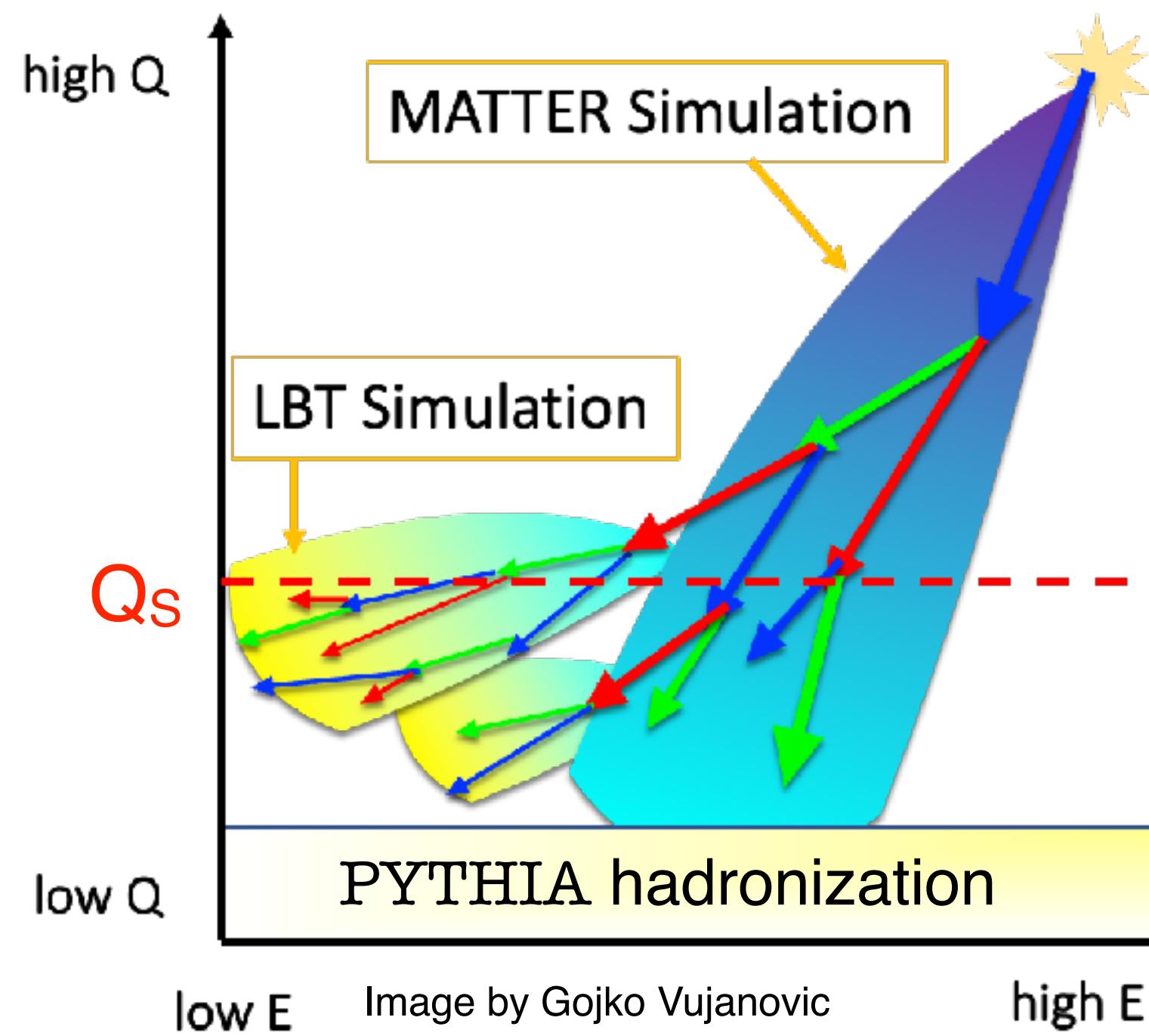
- A transport model is based on some underlying physics model - this is not necessarily the case for a Monte-Carlo
- A Monte-Carlo is designed to describe data under some specific constraints/assumptions. The way it accomplishes this does not necessarily rely on an underlying physics model, instead it can directly parametrize the desired output and/or generate it from sampling some pre-determined distribution.
- The goal of constructing a transport model is to test its underlying assumptions via a comparison to data and gaining physics insight from such a comparison whereas a Monte-Carlo often is utilized as a stand-in for real data.
- A transport model may utilize monte-carlo techniques for sampling of its physics processes.

# Macro + Micro Hybrid: hadronic afterburner comparison



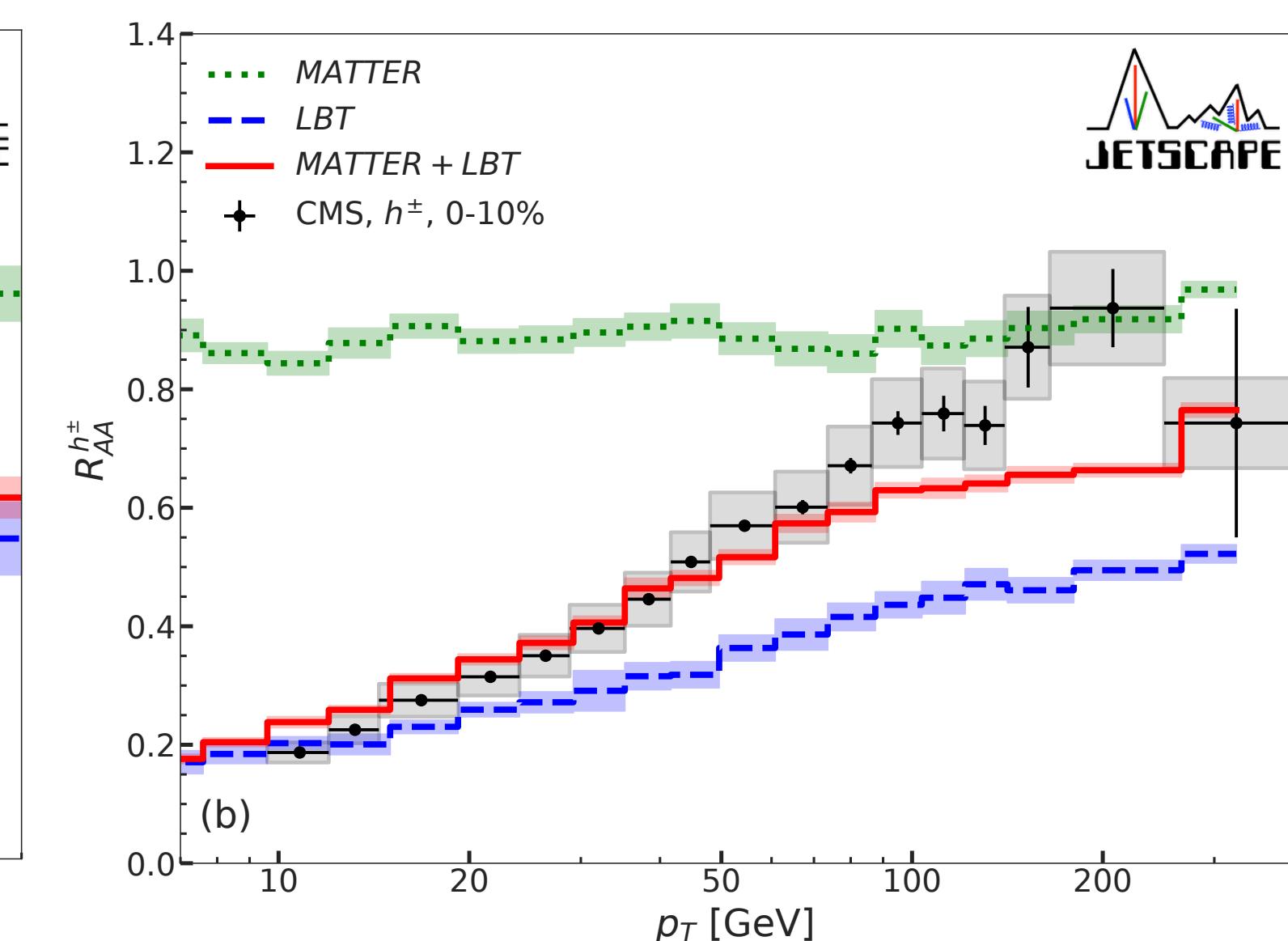
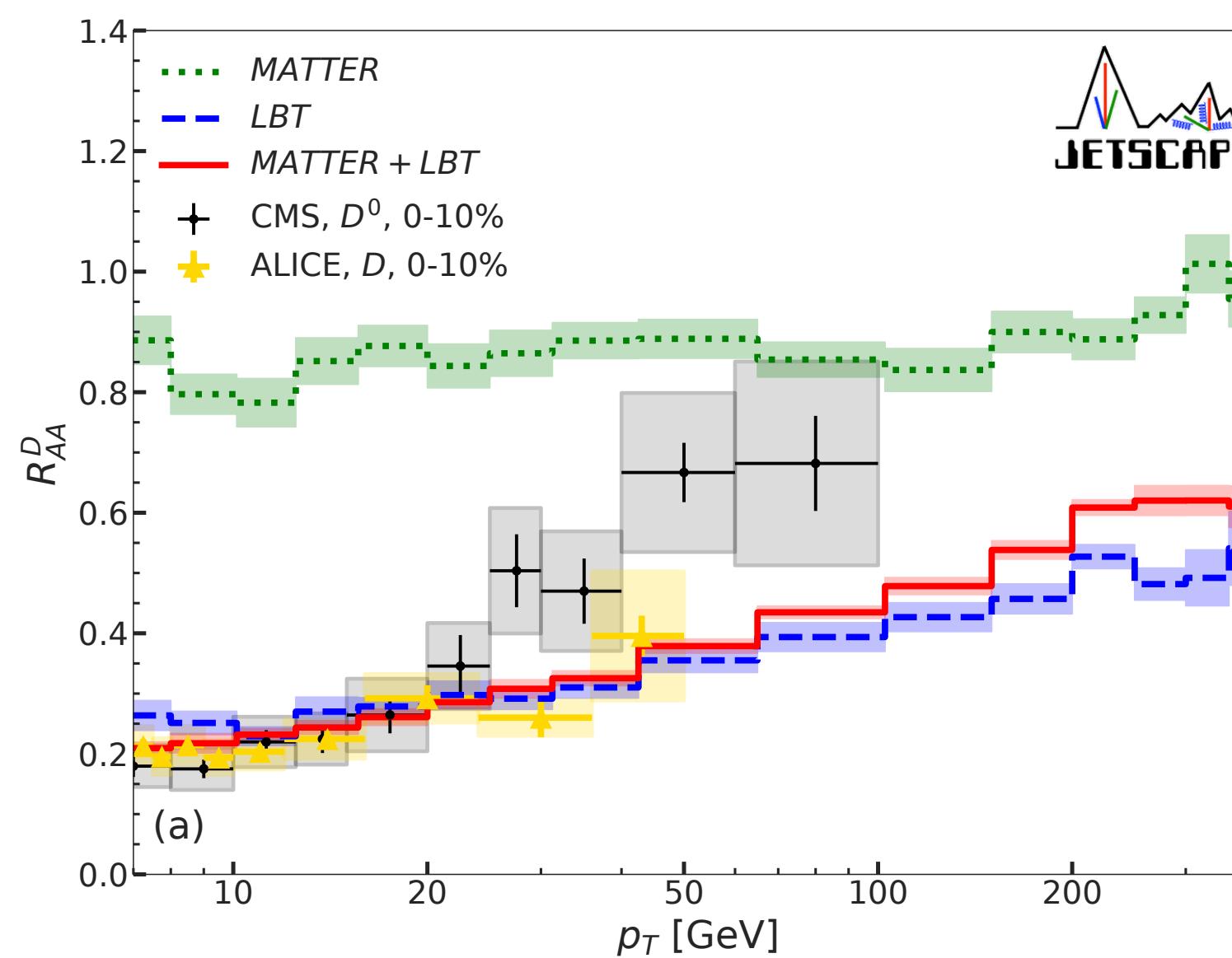
- for the purpose of modeling hadronic rescattering at RHIC & LHC, UrQMD and SMASH behave nearly identical

# Energy-Loss in Matter/LBT



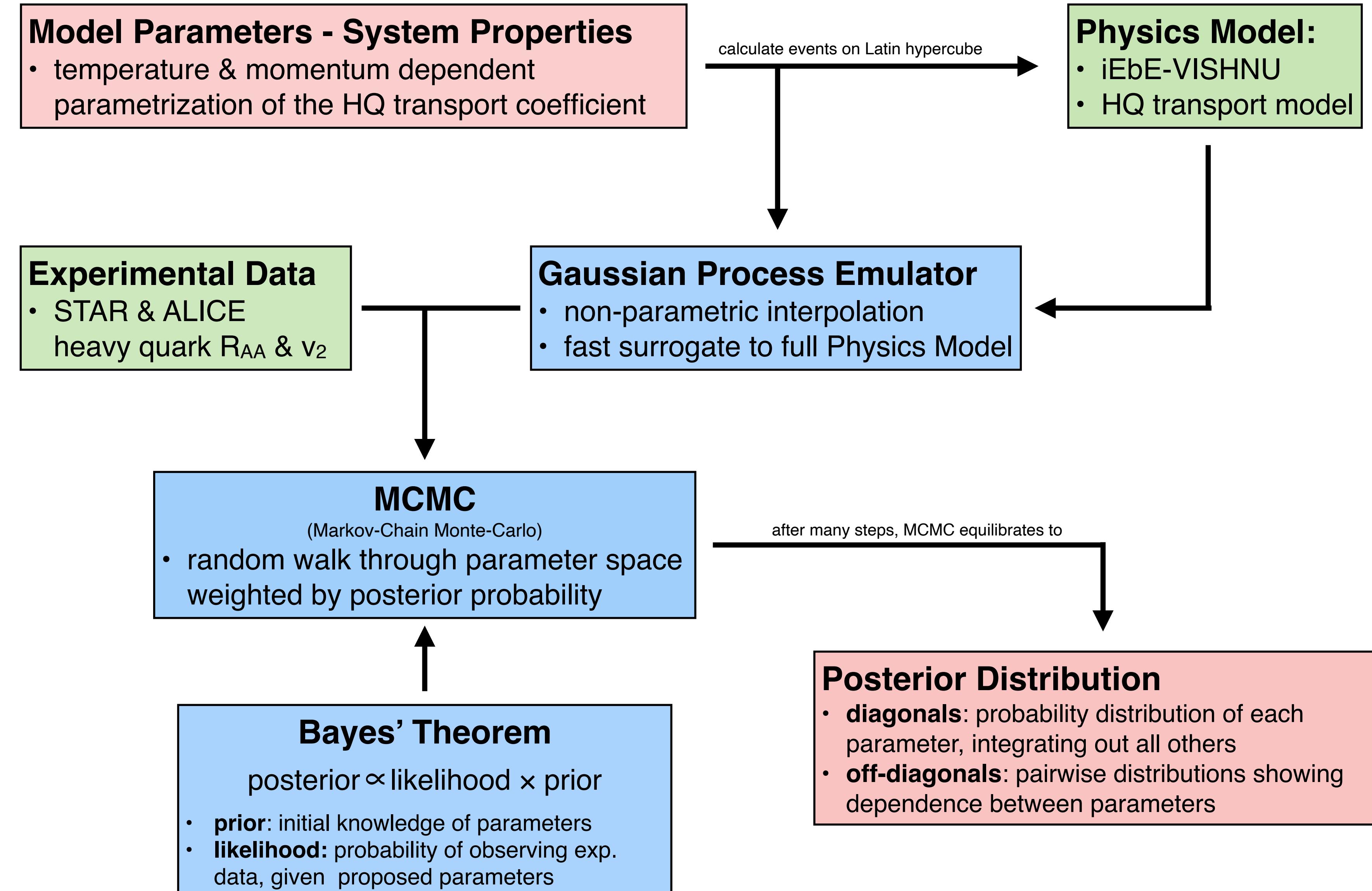
- high virtuality **in medium** parton showering is solved by the Matter model which employs the Higher Twist formalism. It generates a virtuality ordered shower with splittings above  $Q \gg Q_0$
- low virtuality parton showering is solved via a Linearized Boltzmann Transport (LBT model):

$$p_1^\mu \partial_\mu f_1(x_1, p_1) = \mathcal{C}_{\text{el}}[f_1] + \mathcal{C}_{\text{inel}}[f_1]$$



- Matter/LBT is embedded into the Jetscape framework that provides a calibrated hydro evolution and initial state
- LBT on-shell parton evolution dominates for D-mesons while Matter evolution is necessary for a good description of charged hadron data

# Setup of a Bayesian Statistical Analysis:



# Calibration

Vector of input parameters:  $\mathbf{x} = [p, k, w, (\eta/s)_{\min}, (\eta/s)_{\text{slope}}, (\zeta/s)_{\text{norm}}, T_{sw}]$

- assume true parameters  $\mathbf{x}_*$  exist  $\Rightarrow$  find probability distribution for  $\mathbf{x}_*$

$$\text{Bayes' Theorem: } P(\mathbf{x}_* | \mathbf{X}, \mathbf{Y}, \mathbf{y}_{\text{exp}}) \propto P(\mathbf{X}, \mathbf{Y}, \mathbf{y}_{\text{exp}} | \mathbf{x}_*) P(\mathbf{x}_*)$$

- $\mathbf{X}$ : training data design points
- $\mathbf{Y}$ : model output on  $\mathbf{X}$
- $P(\mathbf{x}_*)$  = prior  
 $\Rightarrow$  initial knowledge of  $\mathbf{x}_*$
- $P(\mathbf{X}, \mathbf{Y}, \mathbf{y}_{\text{exp}} | \mathbf{x}_*)$  = likelihood  
 $\Rightarrow$  probability of observing  $(\mathbf{X}, \mathbf{Y}, \mathbf{y}_{\text{exp}})$  given proposed  $\mathbf{x}_*$
- $P(\mathbf{x}_* | \mathbf{X}, \mathbf{Y}, \mathbf{y}_{\text{exp}})$  = posterior  
 $\Rightarrow$  probability of  $\mathbf{x}_*$  given observations  $(\mathbf{X}, \mathbf{Y}, \mathbf{y}_{\text{exp}})$

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 $\Rightarrow$  probability of  $\mathbf{x}_*$  given observations  $(\mathbf{X}, \mathbf{Y}, \mathbf{y}_{\text{exp}})$

## Markov-Chain Monte-Carlo:

- random walk through parameter space weighted by posterior
- large number of samples  $\Rightarrow$  chain equilibrates to posterior distribution
- flat prior within design range, zero outside
- likelihood:  $\log[P(\mathbf{X}, \mathbf{Y}, \mathbf{y}_{\text{exp}} | \mathbf{x}_*)] \sim -(\mathbf{y}(\mathbf{x}_*) - \mathbf{y}_{\text{exp}})^2 / (2\sigma^2)$ 
  - $\sigma=0.1$  on principal components (includes correlations)
- posterior  $\sim$  likelihood within design range, zero outside

# Initial Conditions: Soft - Hard Correlations

## QGP medium: Trento

- effective, parametric, description of entropy production prior to thermalization
- entropy deposition  $dS/dy$  parameterized in terms of  $T_A$ ,  $T_B$ :

$$dS/dy |_{\tau=\tau_0} \propto T_R(p; T_A, T_B) \equiv \left( \frac{T_A^p + T_B^p}{2} \right)^{1/p}$$

- choose  $p=0$ : EKRT & IP-Glasma scaling

## Heavy Quarks:

- initial spatial production probability:  $\propto T_A T_B$ , consistent with soft QGP medium
- momentum space: use PYTHIA to generate HQ momenta

