



Uniwersytet
Wrocławski



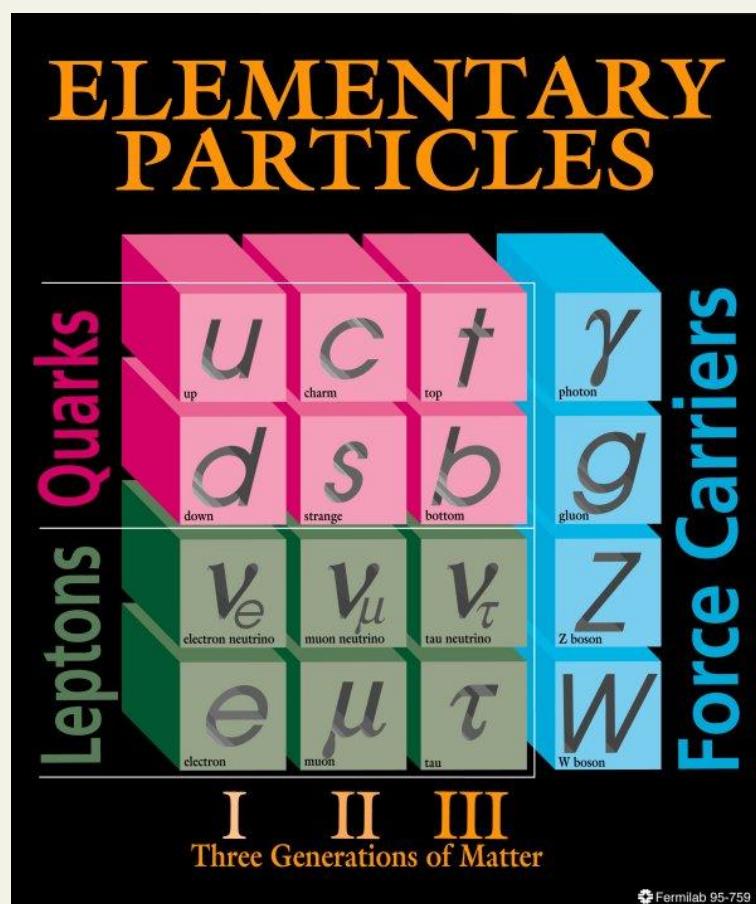
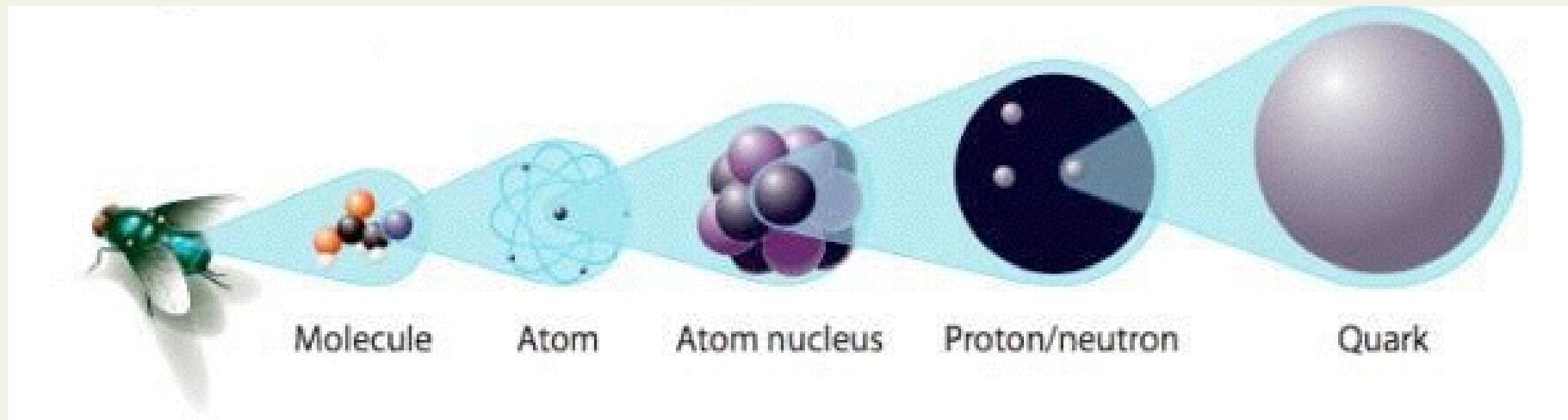
Hydrodynamic simulations of superdense fluids

Pasi Huovinen

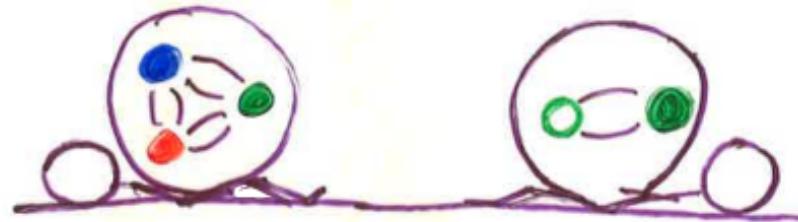
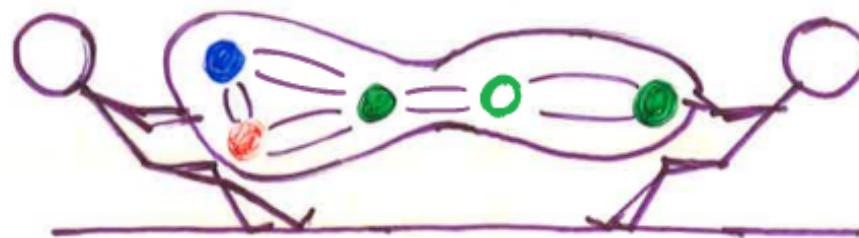
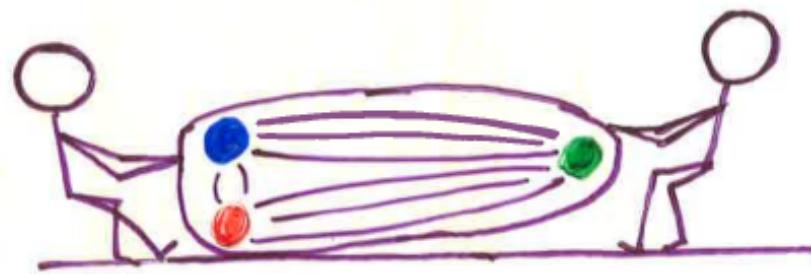
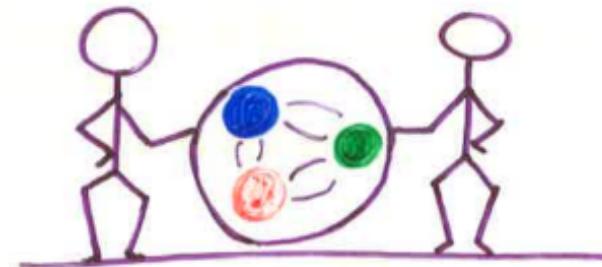
Incubator of Scientific Excellence—Centre for Simulations of Superdense Fluids
University of Wrocław

Many particle systems under extreme conditions

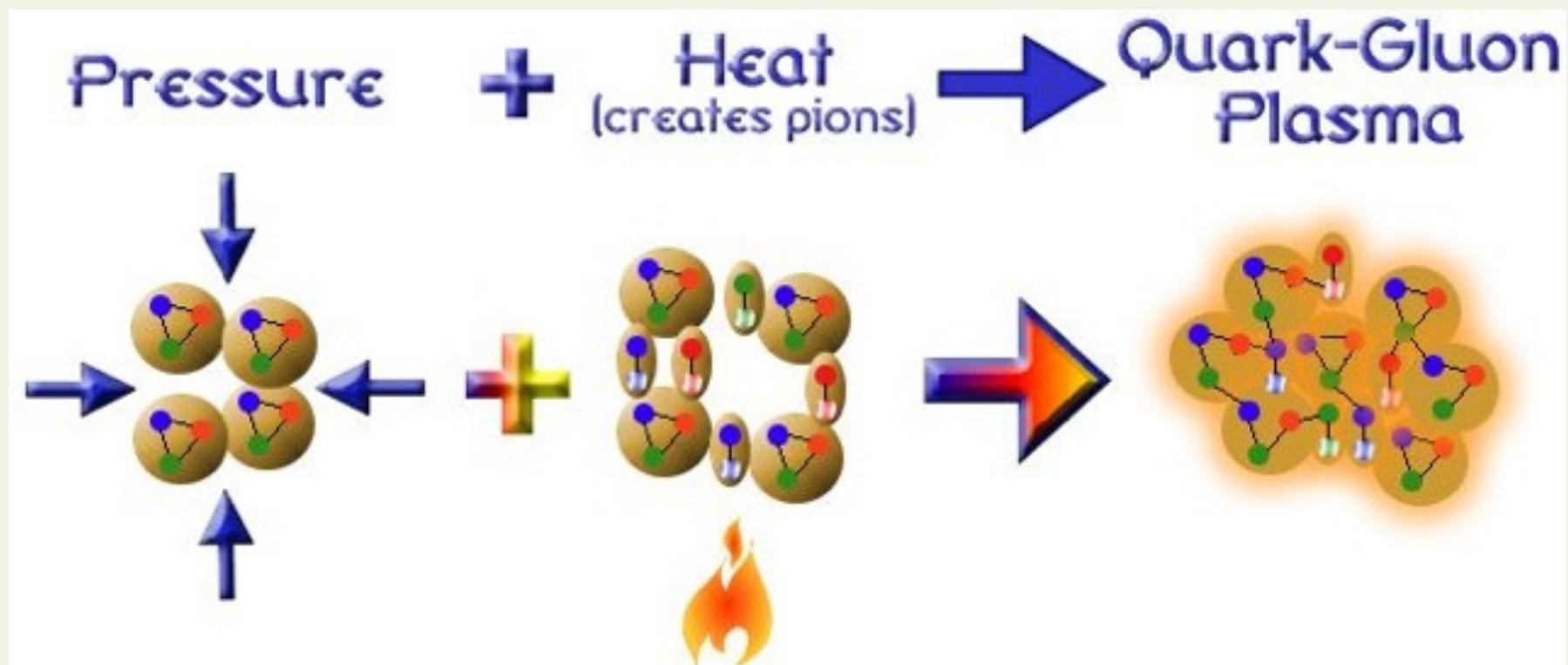
December 4, 2023, Görlitz, Germany



- 12 building blocks of matter (+antiparticles)
- observed matter consists only of three of these (u, d, e)

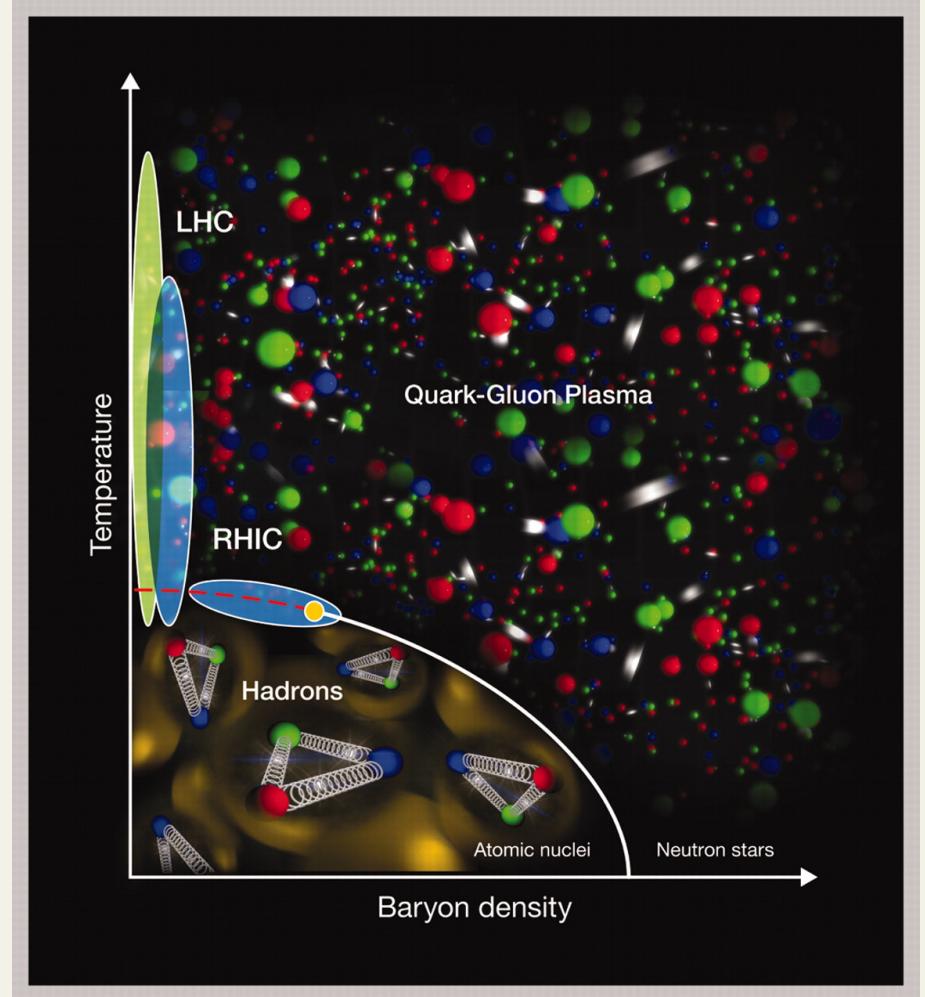
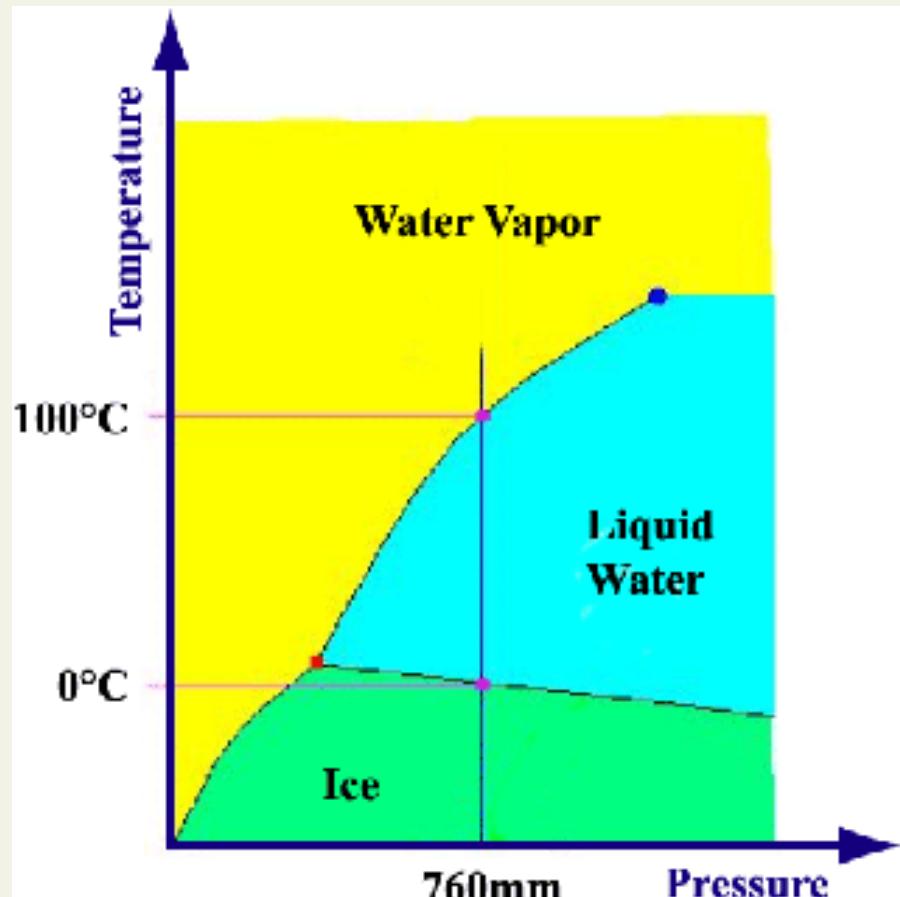


What happens when you compress nuclear matter to very high temperatures and densities?

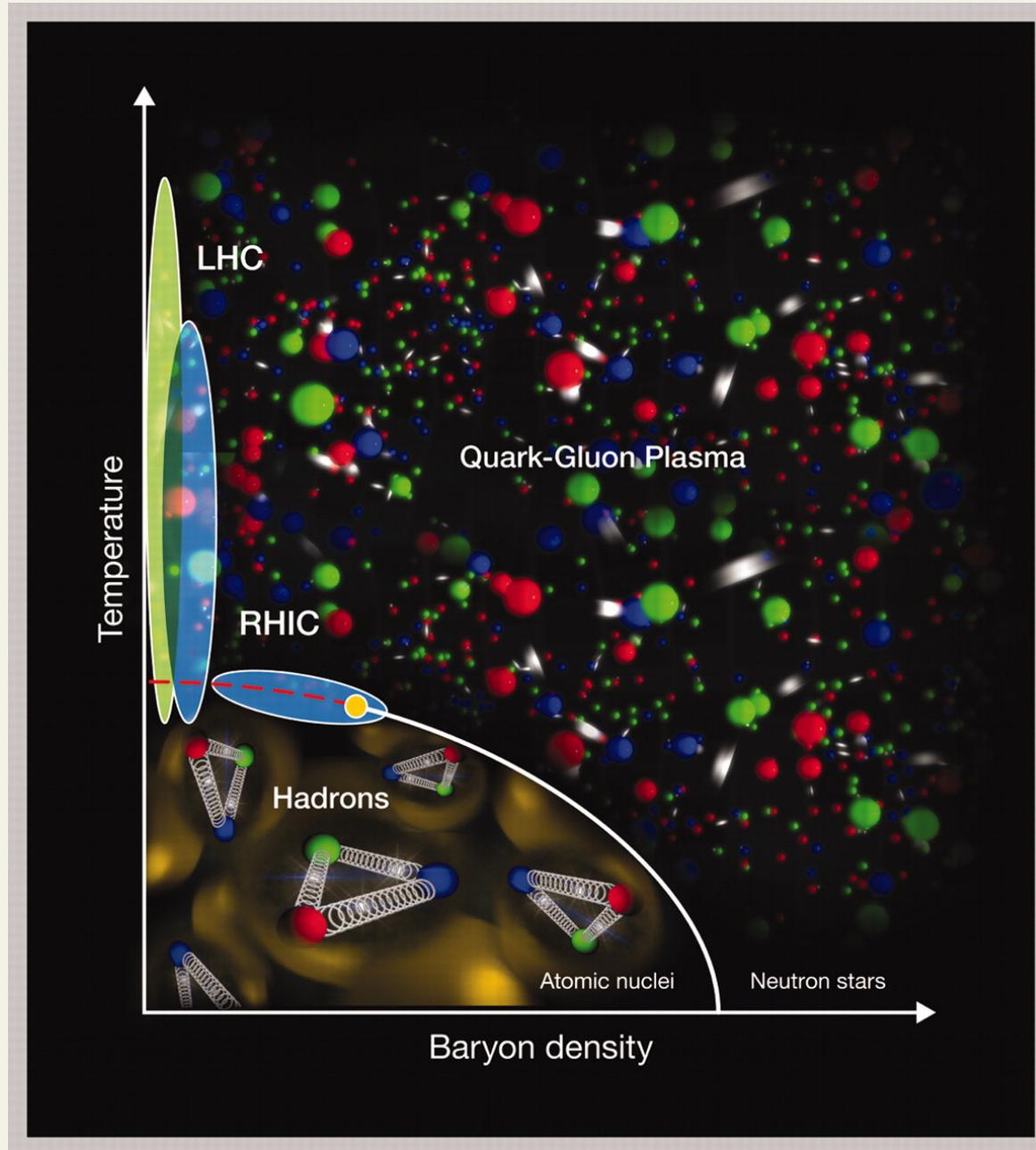


- phase transition to quark-gluon plasma, QGP

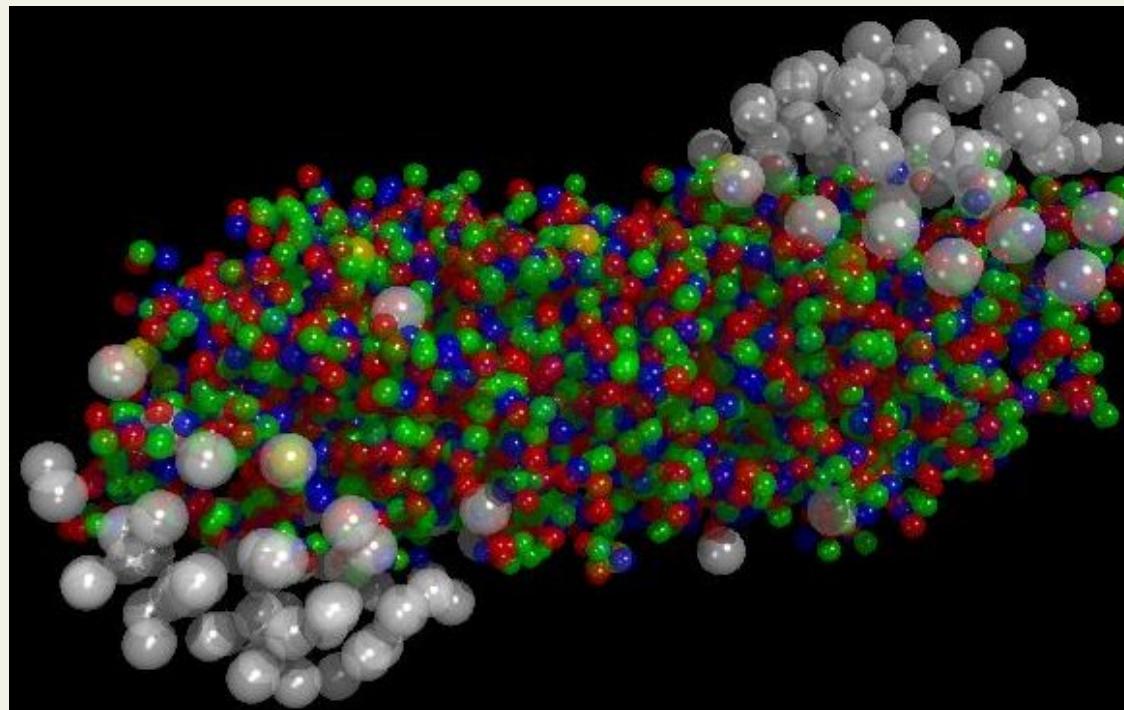
Phase diagram



Phase diagram

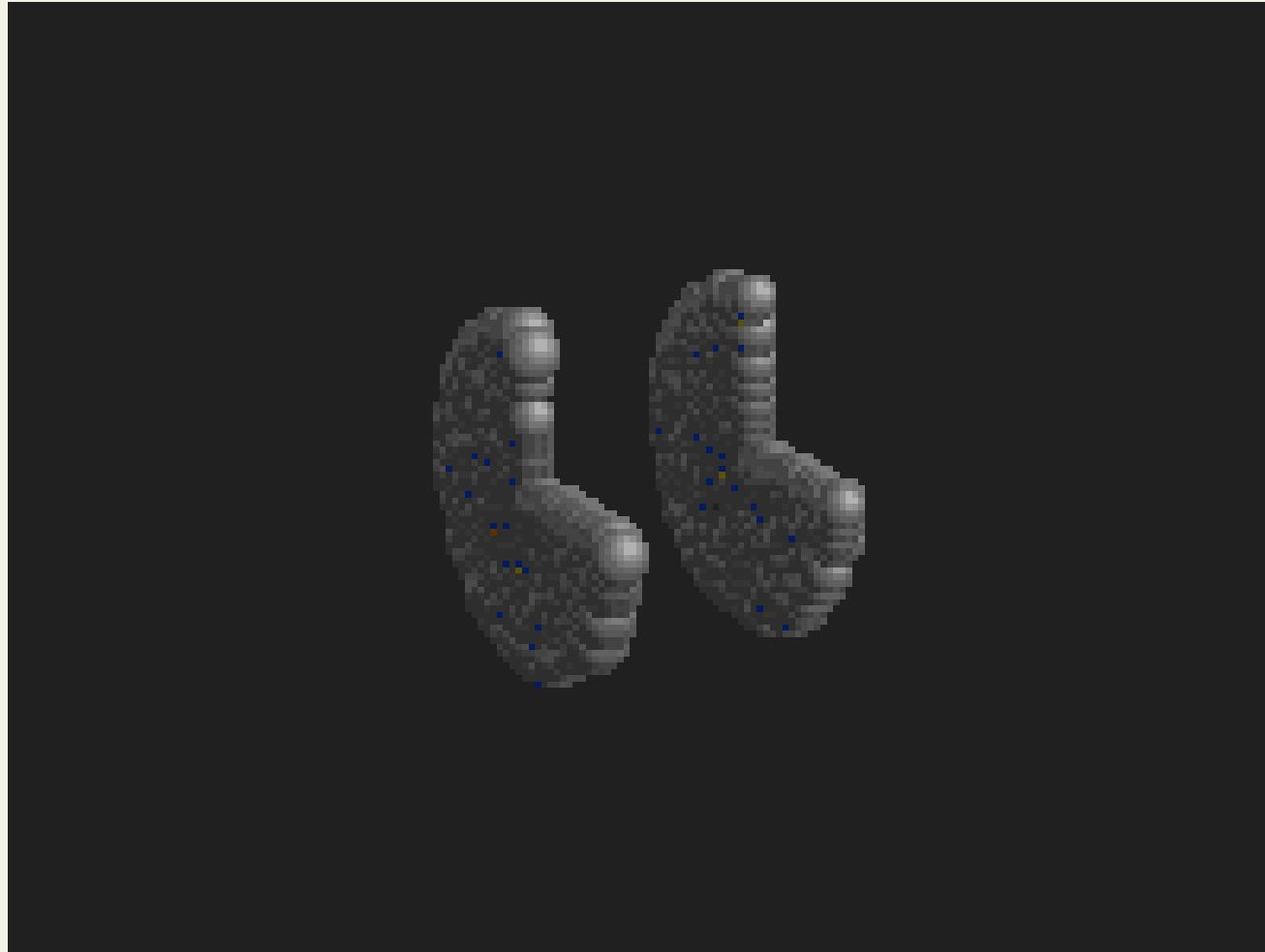


Collisions of heavy nuclei: Little Bang



- Brookhaven National Laboratory (BNL): RHIC,
 $Au + Au$ at $\sqrt{s_{NN}} = 200$ GeV
- CERN LHC,
 $Pb + Pb$ at $\sqrt{s_{NN}} = 2.76$ and 5.02 TeV

Heavy-ion collision



©Harri Niemi

Heavy-ion collision



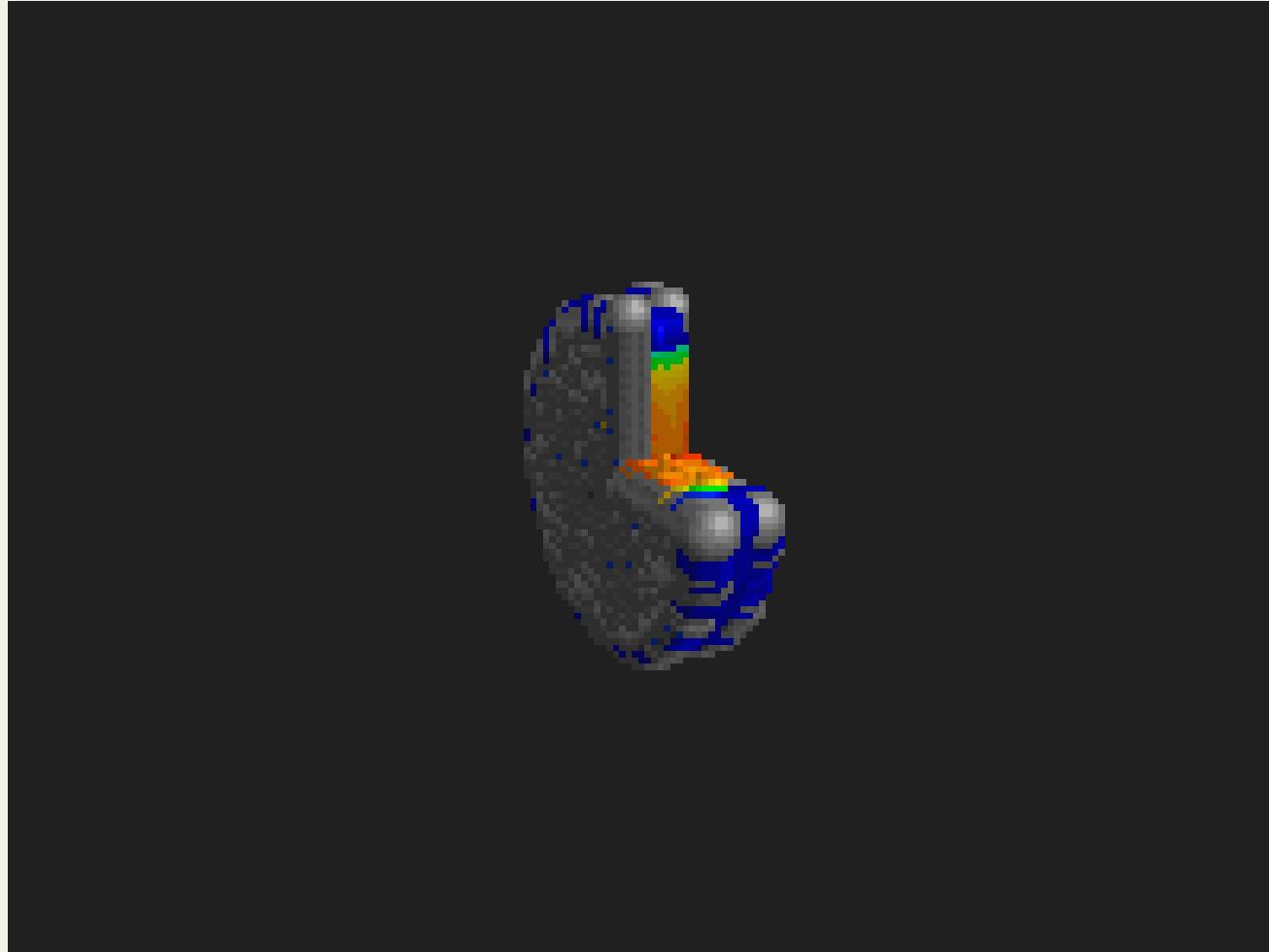
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Heavy-ion collision



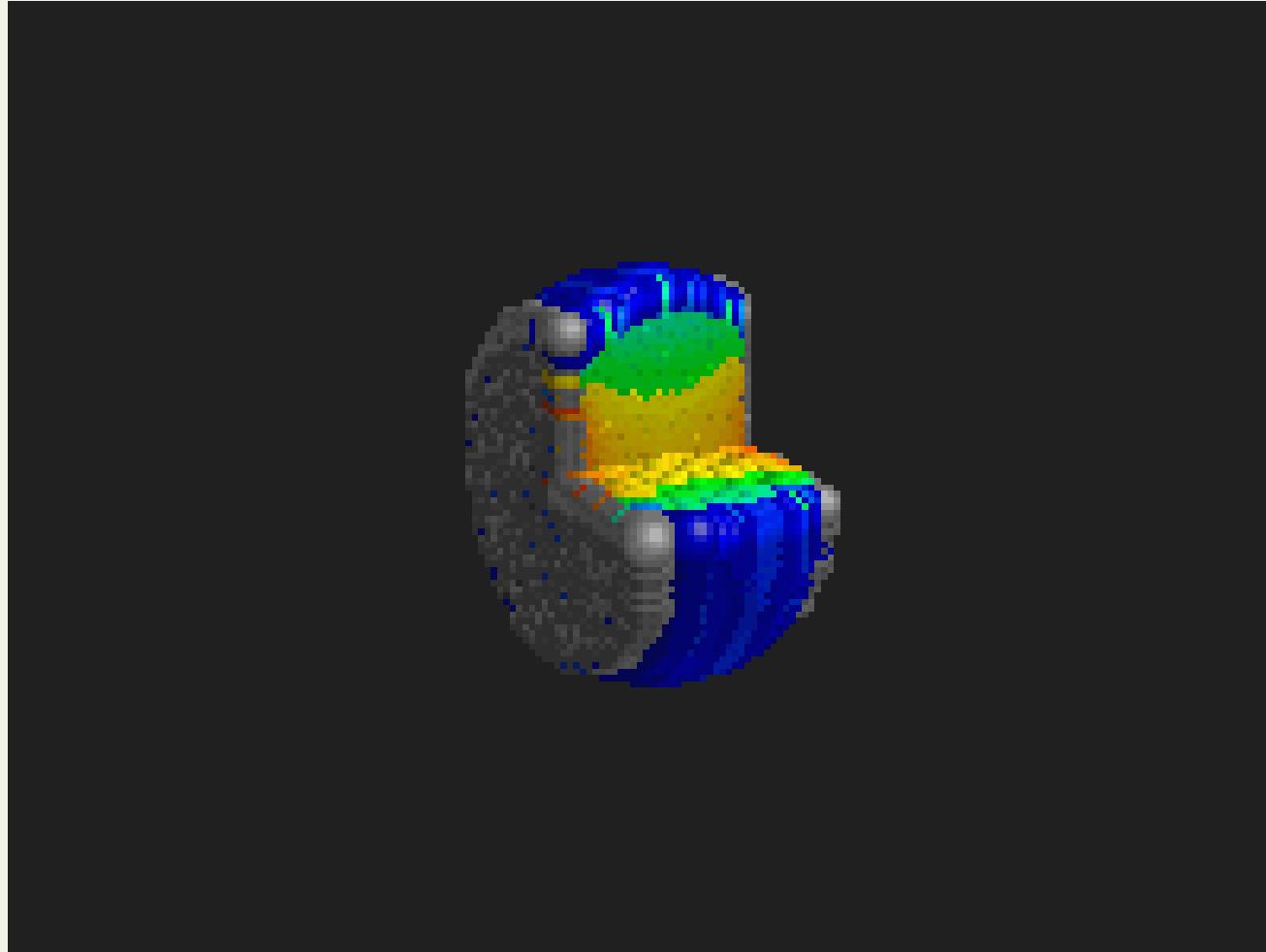
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Heavy-ion collision



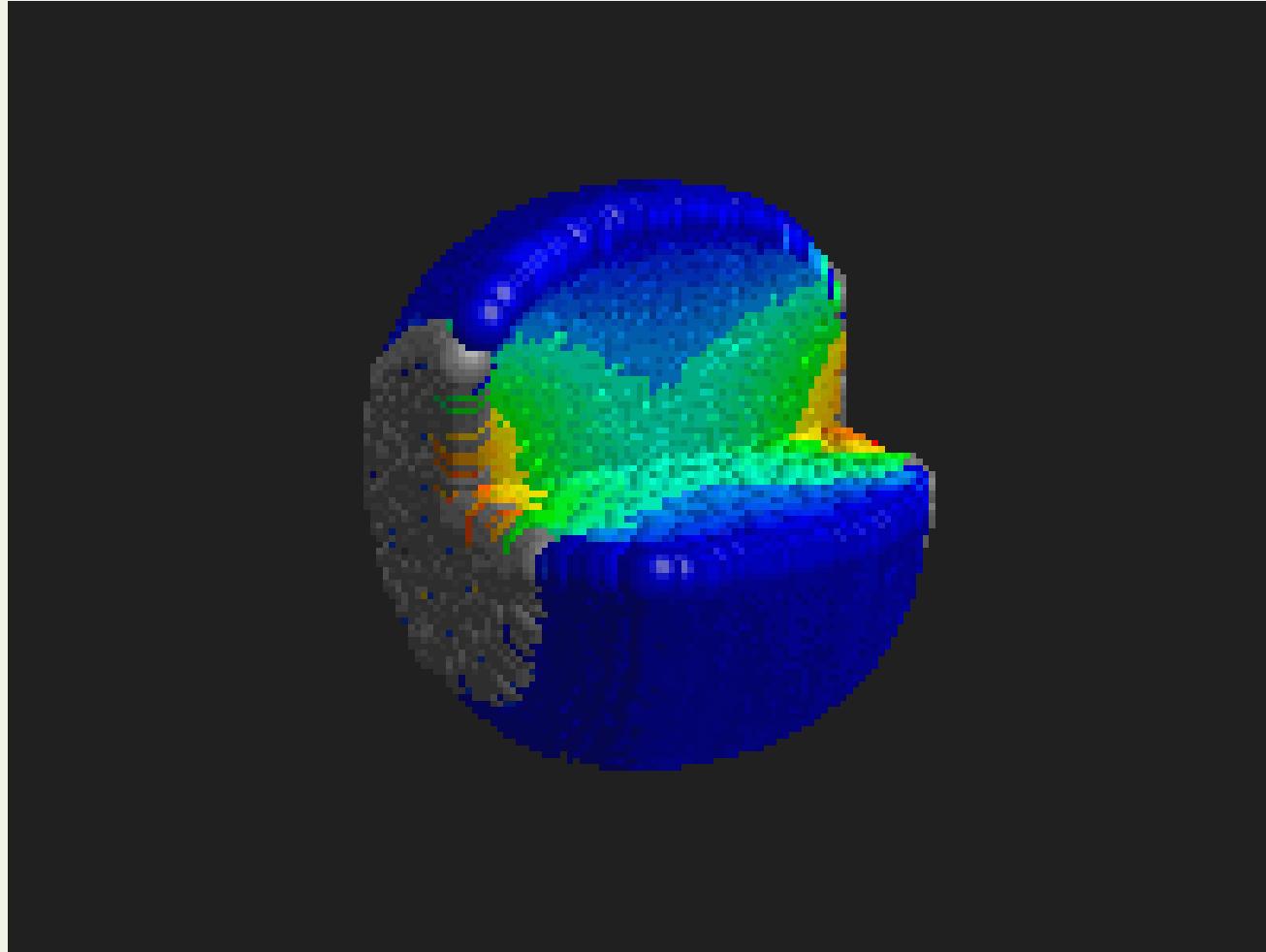
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Heavy-ion collision



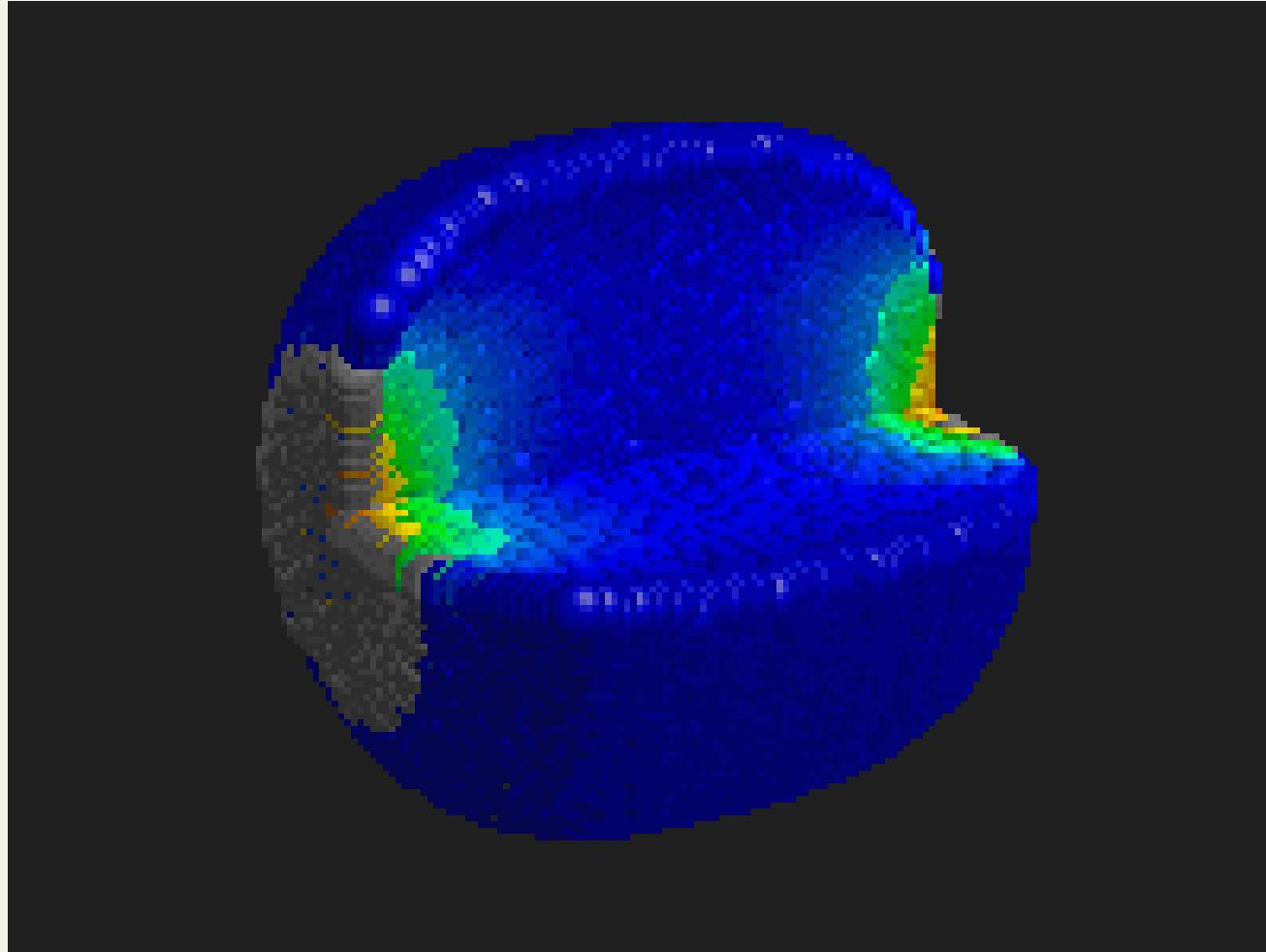
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Heavy-ion collision

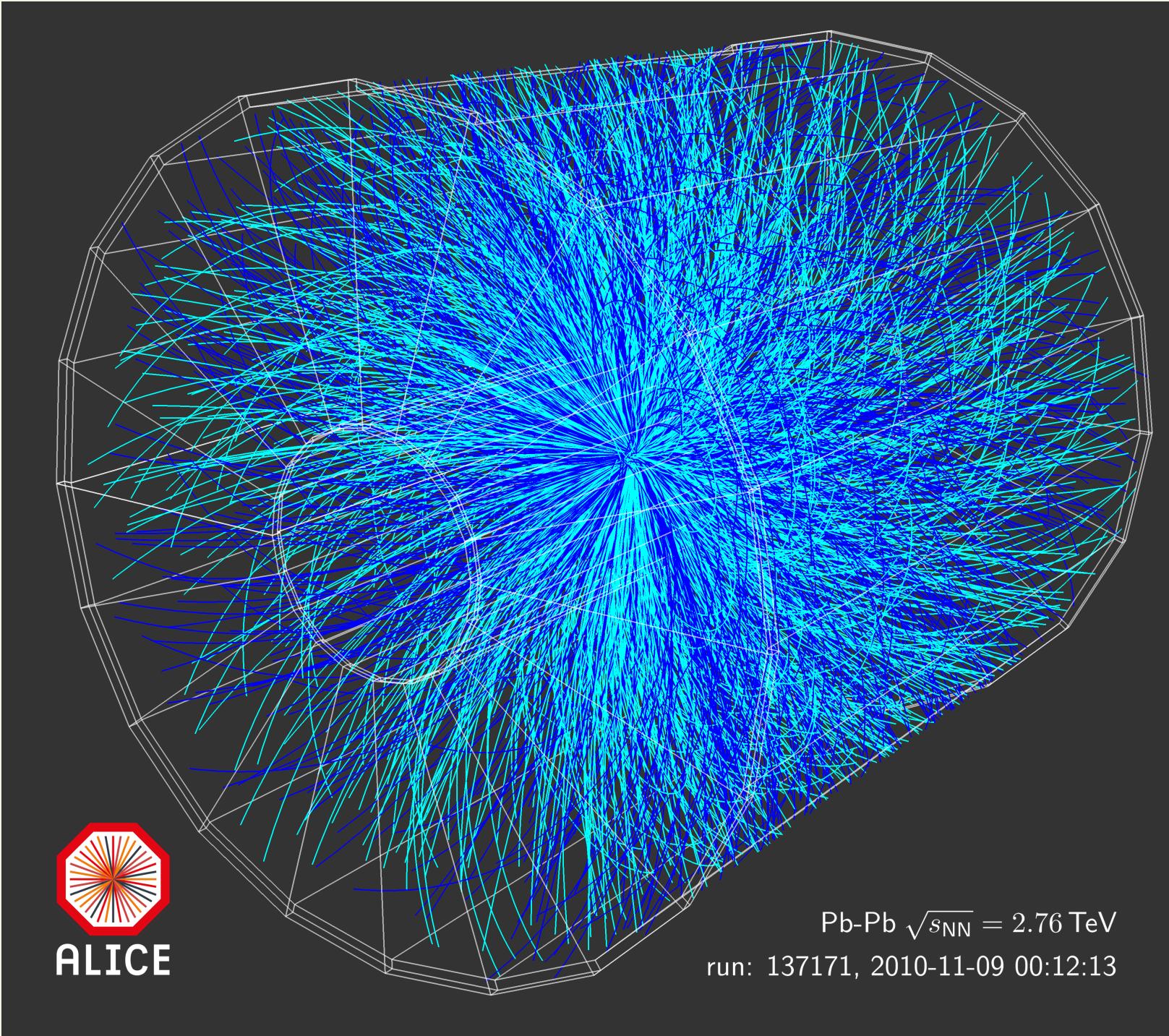


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Heavy-ion collision



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Fluid dynamics

local conservation of energy, momentum and baryon number:

$$\partial_\mu T^{\mu\nu}(x) = 0 \quad \text{and} \quad \partial_\mu N^\mu(x) = 0$$

$$\begin{aligned} T^{\mu\nu} &= (\epsilon + P + \Pi) u^\mu u^\nu - (P + \Pi) g^{\mu\nu} + \pi^{\mu\nu} \\ N^\mu &= n u^\mu + \nu^\mu \end{aligned}$$

local, macroscopic variables: **energy density** $\epsilon(x)$
pressure $P(x)$
flow velocity $u^\mu(x)$

matter characterized by: **equation of state** $P = P(T, \{\mu_i\})$
transport coefficients $\eta = \eta(T, \{\mu_i\})$
 $\zeta = \zeta(T, \{\mu_i\})$
 $\kappa = \kappa(T, \{\mu_i\})$

Evolution equations

- four coupled differential equations, $\mu \in \{0, 1, 2, 3\}$:

$$\frac{\partial}{\partial t} T^{0\mu} = -\frac{\partial}{\partial x} T^{1\mu} - \frac{\partial}{\partial y} T^{2\mu} - \frac{\partial}{\partial z} T^{3\mu}$$

which can be cast in the form

$$\frac{\partial}{\partial t} \rho_0 + \nabla \cdot (\vec{v}\rho) = X$$

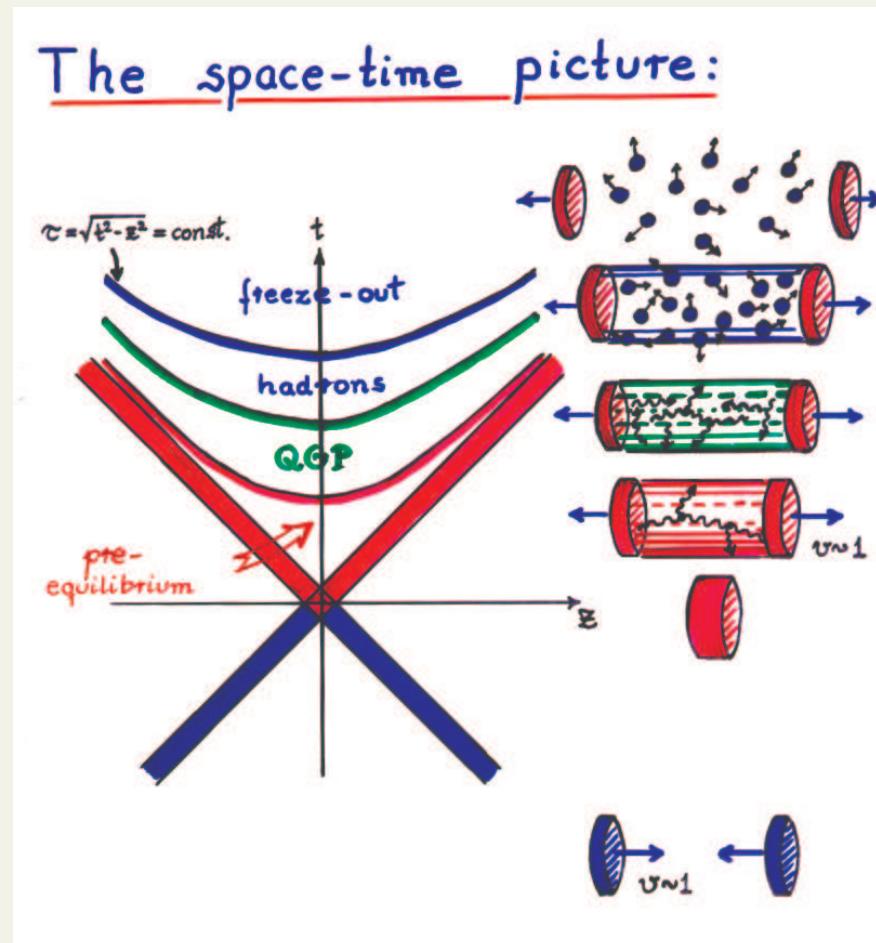
- must be solved numerically
- can be solved using conventional algorithms of fluid dynamics

usually

- fixed grid
- finite volume with flux correction

Boundary conditions

- spatial: expansion to vacuum \Rightarrow vacuum “far away”
- temporal, early: after primary collisions!
- temporal, late: once the system has cooled



©Dirk H. Rischke

Dissipative relativistic fluid dynamics

$$T^{\mu\nu} = (\epsilon + P + \Pi)u^\mu u^\nu - (P + \Pi)g^{\mu\nu} + \pi^{\mu\nu}$$

$\pi^{\mu\nu}$: shear-stress tensor, Π : bulk pressure

relativistic Navier-Stokes: $\pi^{\mu\nu} = 2\eta\nabla^{\langle\mu}u^{\nu\rangle}$

- parabolic equations of motion!
- unstable and acausal

Dissipative relativistic fluid dynamics

$$T^{\mu\nu} = (\epsilon + P + \Pi)u^\mu u^\nu - (P + \Pi)g^{\mu\nu} + \pi^{\mu\nu}$$

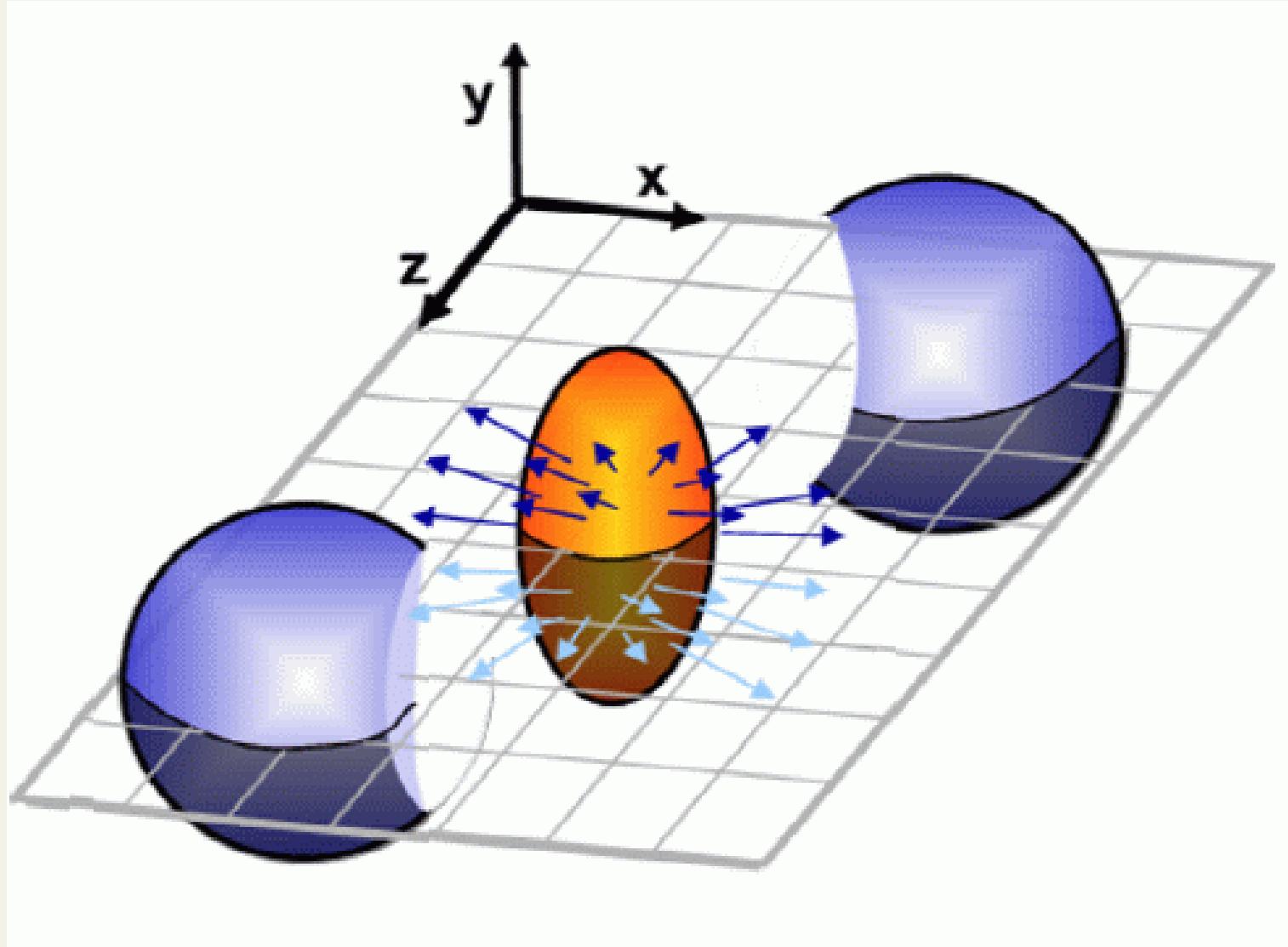
$\pi^{\mu\nu}$: shear-stress tensor, Π : bulk pressure

Israel-Stewart: $\pi^{\mu\nu}$ independent dynamical variable

$$\langle u^\lambda \partial_\lambda \pi^{\mu\nu} \rangle = \frac{\pi_{\text{NS}}^{\mu\nu} - \pi^{\mu\nu}}{\tau_\pi} - \frac{4}{3} \pi^{\mu\nu} \partial_\lambda u^\lambda + \dots$$

- η : shear viscosity coefficient
- τ_π : shear relaxation time, known for massless particles
- solved numerically
- finite differencing, same grid than for equations of motion

Elliptic flow v_2



Elliptic flow v_2

spatial anisotropy → final azimuthal momentum anisotropy

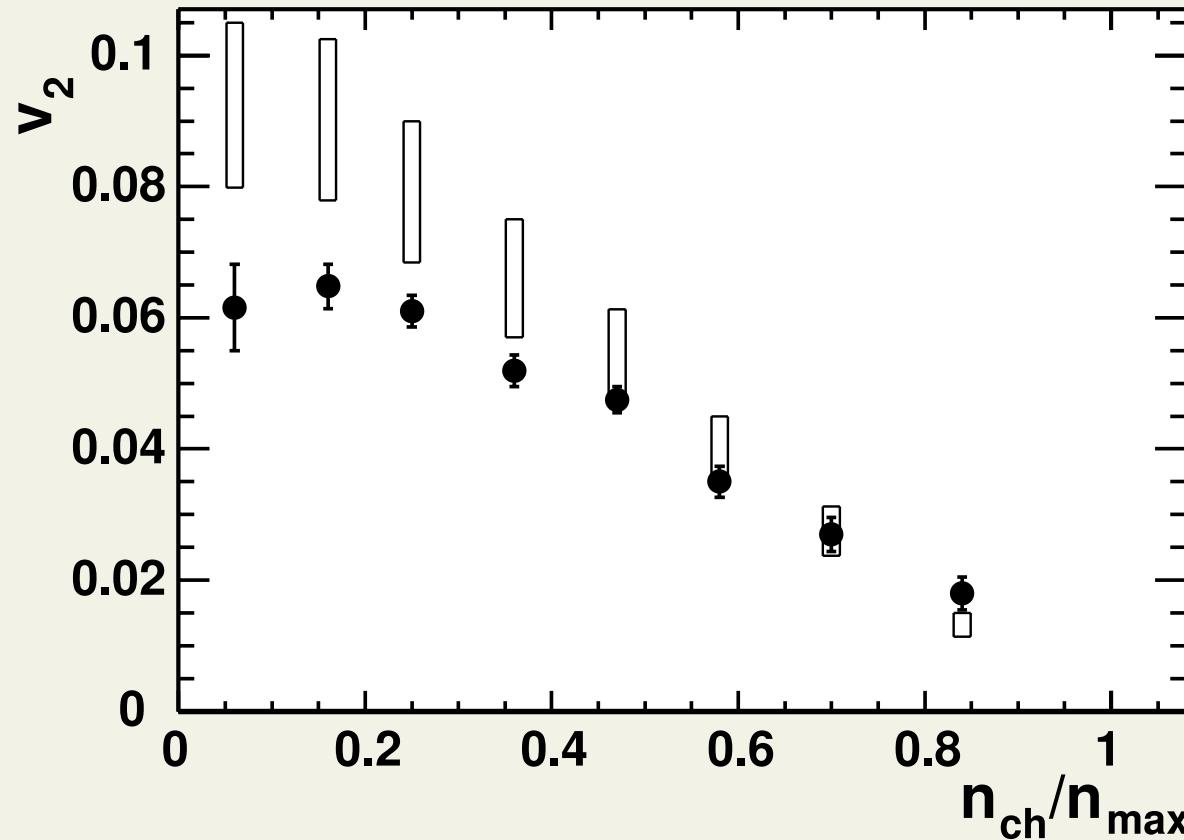


- Anisotropy in coordinate space + rescattering
⇒ Anisotropy in momentum space
- sensitive to speed of sound $c_s^2 = \partial p / \partial e$ and shear viscosity η
- Quantified with Fourier expansion of momentum distribution:

$$\frac{dN}{dy p_T dp_T d\phi} = \frac{1}{2\pi} \frac{dN}{dy p_T dp_T} (1 + 2v_1(y, p_T) \cos \phi + 2v_2(y, p_T) \cos 2\phi + \dots)$$

Success of ideal hydrodynamics

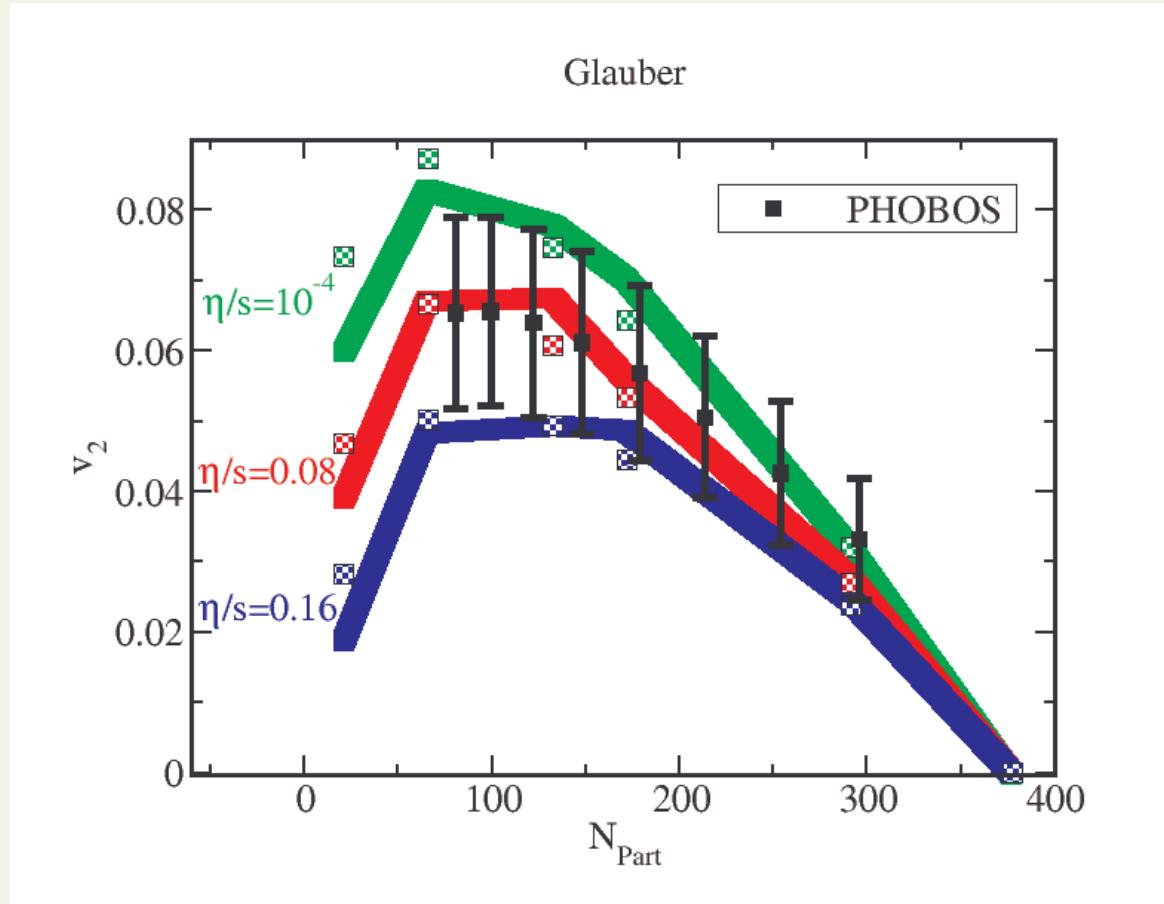
- *p_T-averaged v₂* of charged hadrons:
- points: data, boxes: hydrodynamical calculations



- works beautifully in central and semi-central collisions

Dissipative hydrodynamics

- Luzum & Romatschke, Phys.Rev.C78:034915,2008



- $\eta/s = 0.08$ favoured

Modeling problem

Model parameters (input): $\vec{x} = (x_1, \dots, x_n)$

$(\tau_0, \epsilon_{\text{init}}, \eta/s, T_{\text{dec}} \dots)$



Model output $\vec{y} = (y_1, \dots, y_m) \Leftrightarrow$ **Experimental values** \vec{y}^{exp}

$(dN/dy, \langle p_T \rangle, v_n, \dots)$

- Which values of input parameters \vec{x} give the best reproduction of experimental output \vec{y}^{exp} ?
- What is the level of uncertainty of these values?

Bayesian analysis

Model parameters (input): $\vec{x} = (x_1, \dots, x_n)$

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Model output $\vec{y} = (y_1, \dots, y_m) \Leftrightarrow$ **Experimental values** \vec{y}^{exp}

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Bayes' theorem:

Posterior probability \propto **Likelihood** · **Prior knowledge**

Bayesian analysis

Model parameters (input): $\vec{x} = (x_1, \dots, x_n)$

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Model output $\vec{y} = (y_1, \dots, y_m) \Leftrightarrow$ **Experimental values** \vec{y}^{exp}

$(dN/dy, \langle p_T \rangle, v_n, \dots)$

Bayes' theorem:

Posterior probability \propto Likelihood · Prior knowledge

- Prior knowledge: Range of parameter values

Bayesian analysis

Model parameters (input): $\vec{x} = (x_1, \dots, x_n)$

$(\tau_0, \epsilon_{\text{init}}, \eta/s, T_{\text{dec}} \dots)$



Model output $\vec{y} = (y_1, \dots, y_m) \Leftrightarrow$ **Experimental values** \vec{y}^{exp}

$(dN/dy, \langle p_T \rangle, v_n, \dots)$

Bayes' theorem:

Posterior probability \propto **Likelihood** \cdot **Prior knowledge**

- **Likelihood:** $\mathcal{L}(\vec{x}) \propto \exp\left(-\frac{1}{2}(\vec{y}(\vec{x}) - \vec{y}^{\text{exp}})\Sigma^{-1}(\vec{y}(\vec{x}) - \vec{y}^{\text{exp}})^T\right)$,

where Σ is the covariance matrix

Bayesian analysis

Model parameters (input): $\vec{x} = (x_1, \dots, x_n)$

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where Σ is the covariance matrix
- evaluation of the likelihood function $\mathcal{O}(10^8)$ runs...
- use Gaussian emulator instead
= stochastic, non-parametric interpolation of the model

Bayesian analysis

Model parameters (input): $\vec{x} = (x_1, \dots, x_n)$

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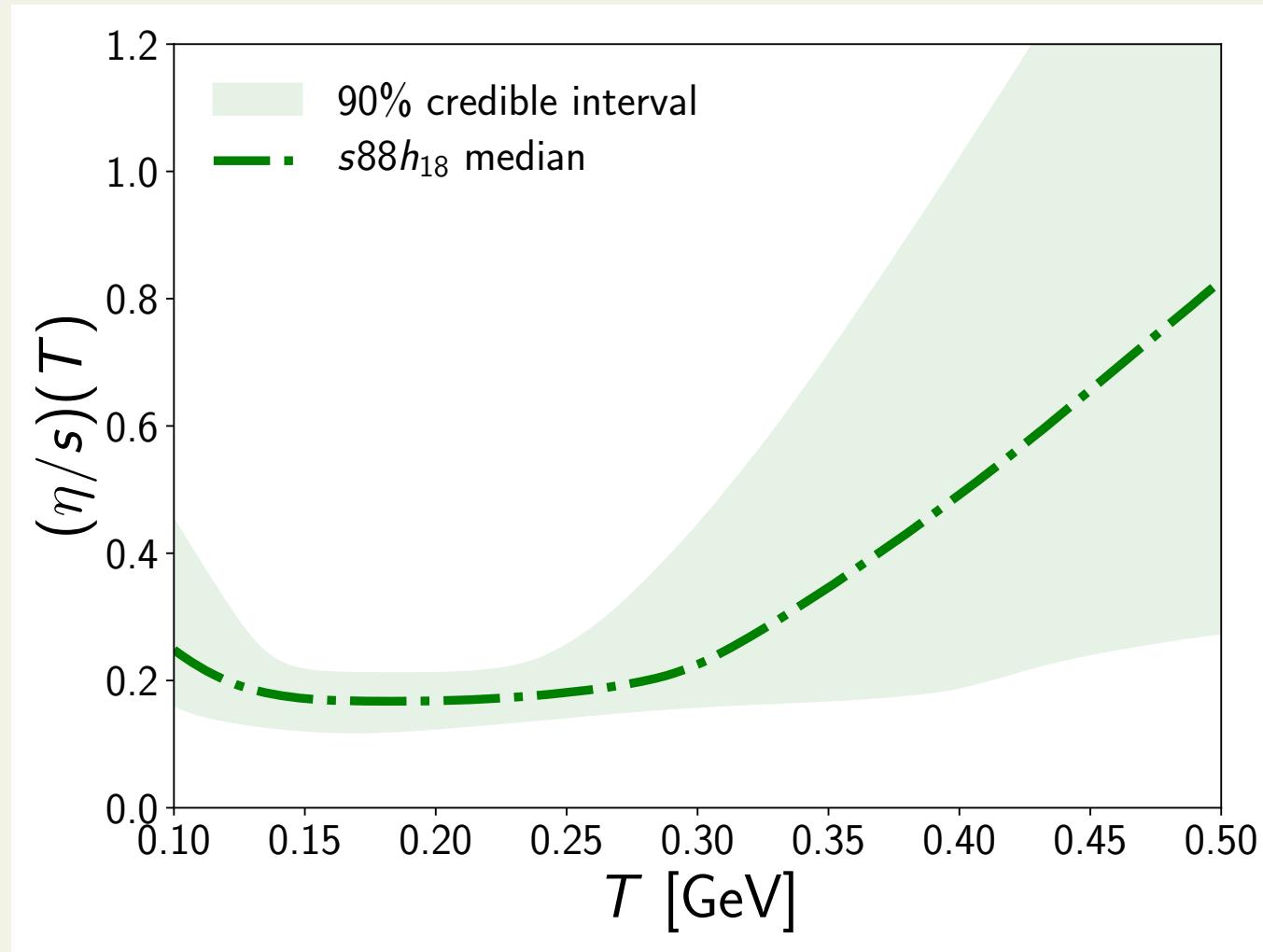
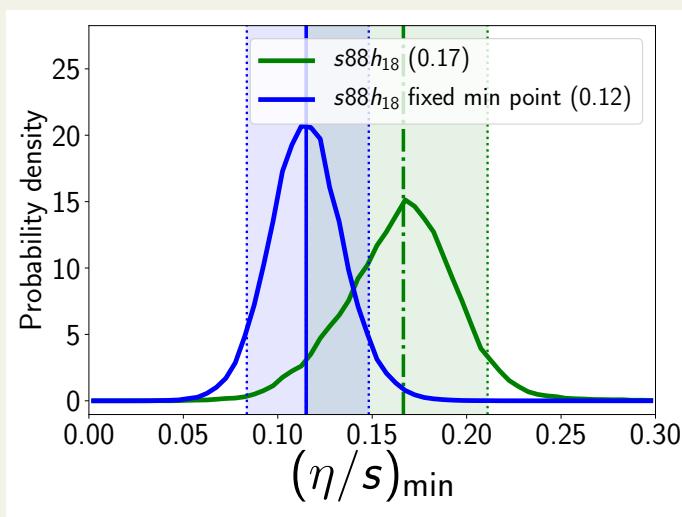
Model output $\vec{y} = (y_1, \dots, y_m) \Leftrightarrow$ Experimental values \vec{y}^{exp}
 $(dN/dy, \langle p_T \rangle, v_n, \dots)$

Bayes' theorem:

Posterior probability \propto Likelihood · Prior knowledge

- Likelihood: $\mathcal{L}(\vec{x}) \propto \exp\left(-\frac{1}{2}(\vec{y}(\vec{x}) - \vec{y}^{\text{exp}})^T \Sigma^{-1} (\vec{y}(\vec{x}) - \vec{y}^{\text{exp}})\right)$,
where Σ is the covariance matrix
- evaluation of the likelihood function $\mathcal{O}(10^8)$ runs...
- use Gaussian emulator instead
= stochastic, non-parametric interpolation of the model
- Sample the likelihood function using Markov chain Monte Carlo
= random walk in parameter space constrained to favour high likelihood
→ distribution of Markov chain steps \equiv probability distribution

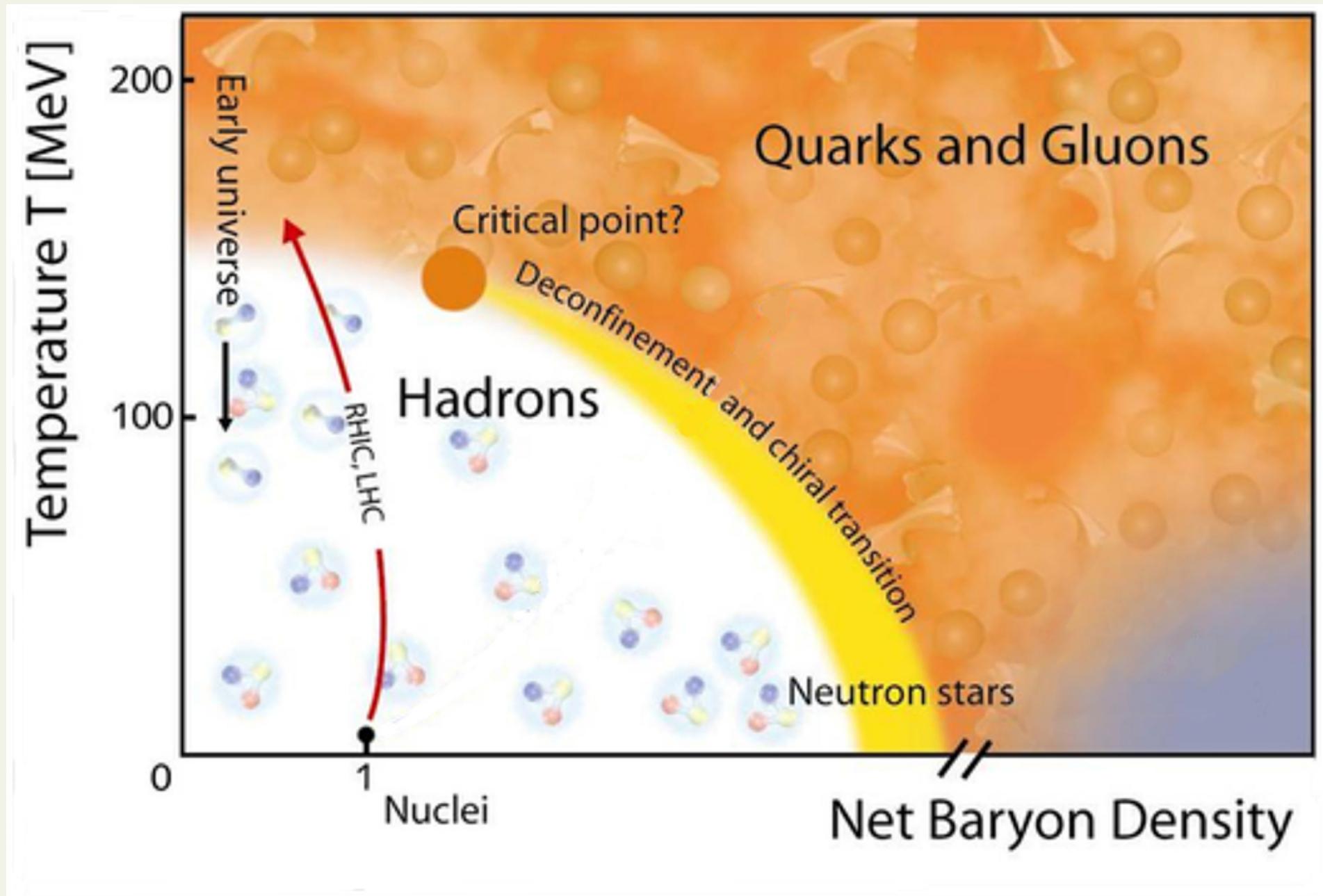
$(\eta/s)(T)$



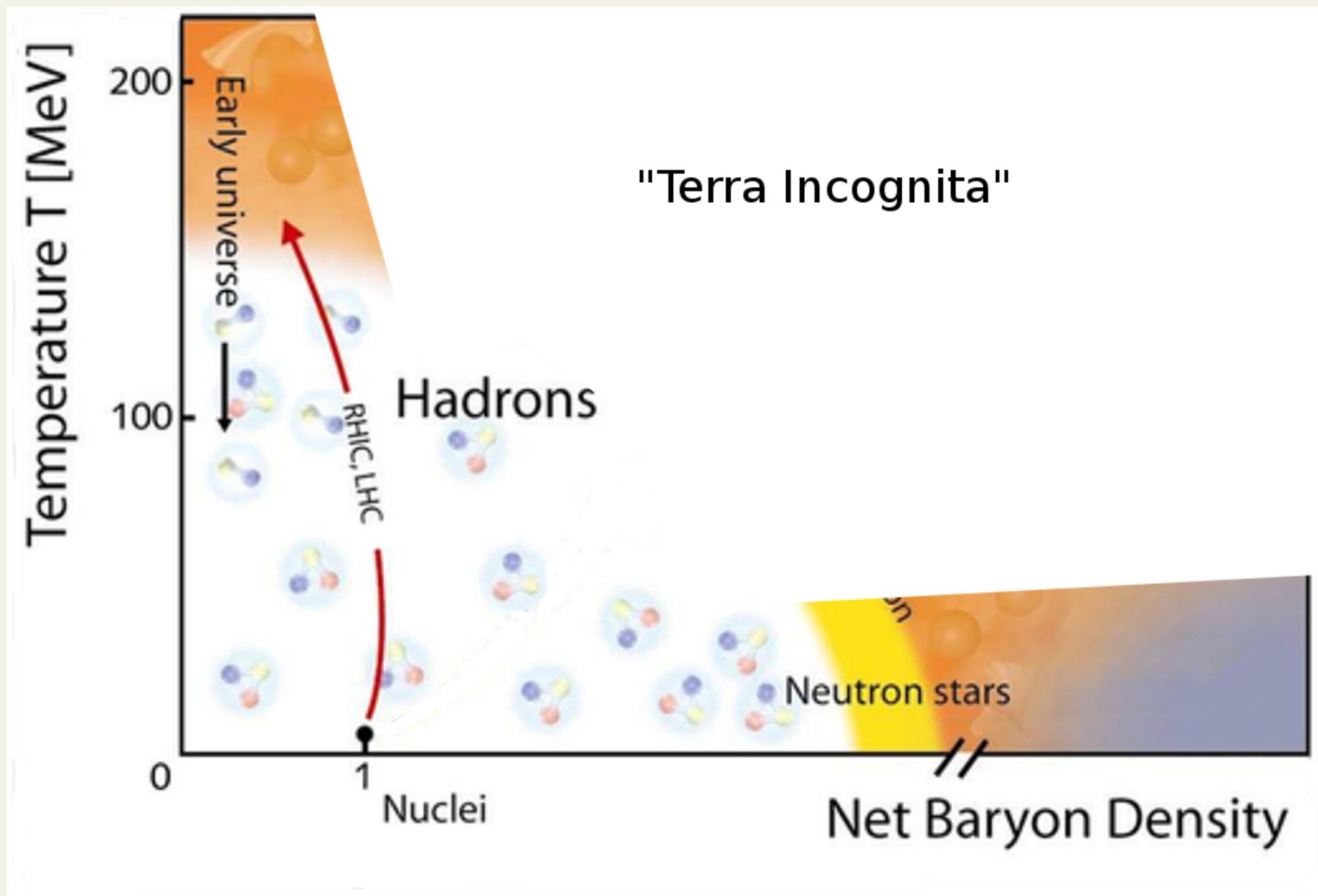
Auvinen *et al.*, PRC 102, 044911 (2020)

- $0.12 < \eta/s < 0.23$ when $150 \lesssim T/\text{MeV} \lesssim 220$

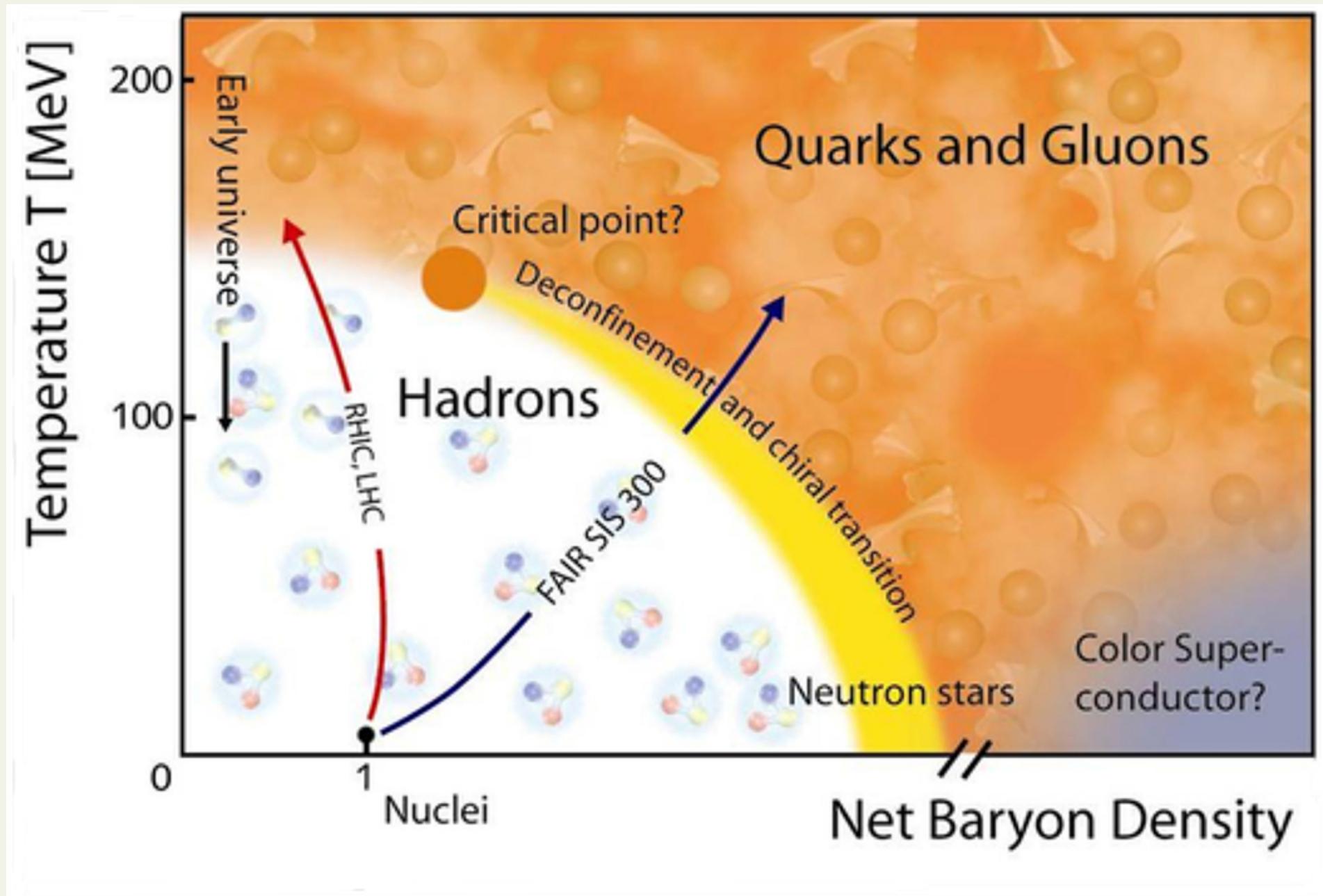
Phase diagram



Phase diagram

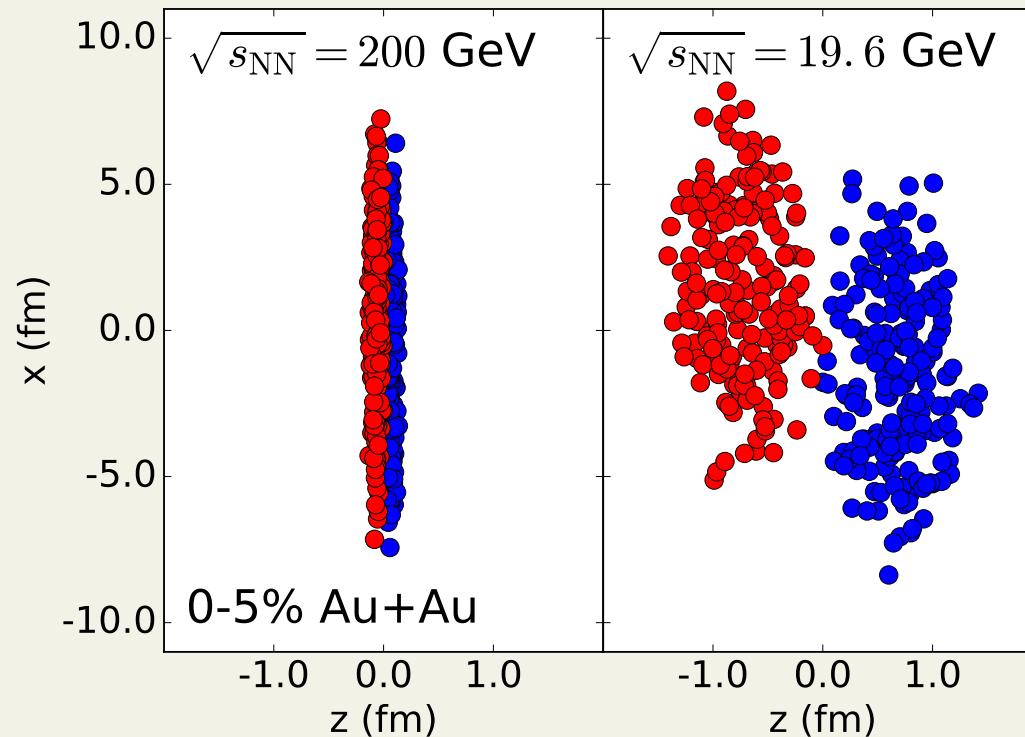


Phase diagram



Challenges

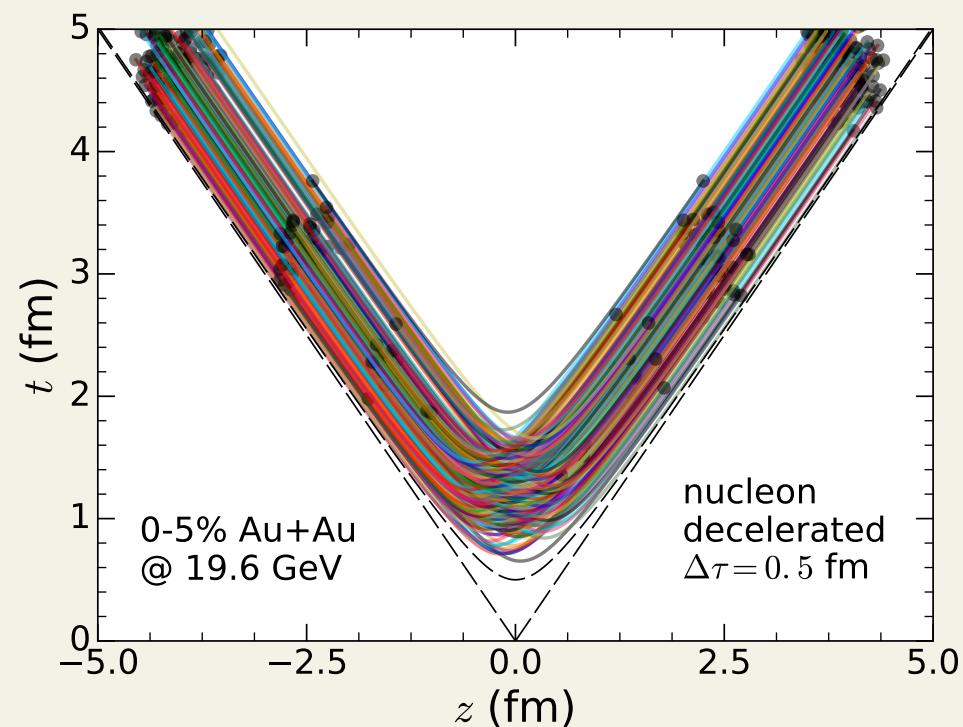
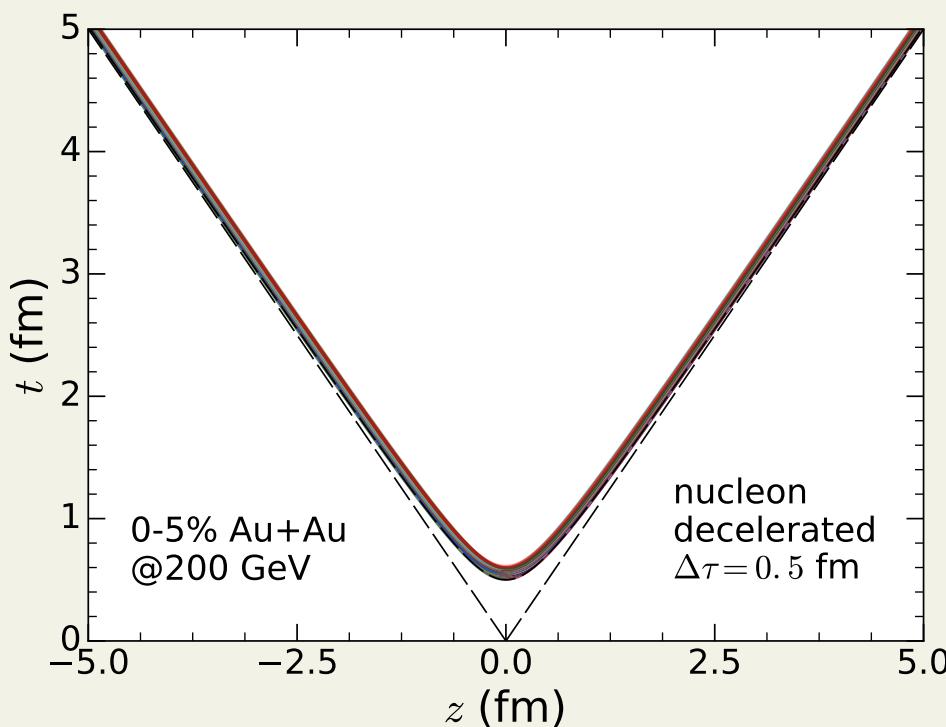
1. primary collisions overlap with secondary collisions



Shen & Schenke, PRC97, 024907 (2018)

Challenges

1. lower multiplicity \implies smaller system
 \implies larger deviations from equilibrium?
2. primary collisions overlap with secondary collisions



Shen & Schenke, PRC97, 024907 (2018)

3-fluid dynamics

$$0 = \partial_\mu T^{\mu\nu}$$

3-fluid dynamics

$$\begin{aligned} 0 &= \partial_\mu T^{\mu\nu} \\ &= \partial_\mu T_t^{\mu\nu} + \partial_\mu T_p^{\mu\nu} + \partial_\mu T_{fb}^{\mu\nu} \end{aligned}$$

$T_t^{\mu\nu}$ = target fluid

$T_p^{\mu\nu}$ = projectile fluid

$T_{fb}^{\mu\nu}$ = fireball fluid

- target and projectile represent colliding nucleons
- fireball (loosely) represents produced particles

3-fluid dynamics

$$\partial_\mu T_t^{\mu\nu}(x) = -F_t^\nu(x) + F_{ft}^\nu(x)$$

$$\partial_\mu T_p^{\mu\nu}(x) = -F_p^\nu(x) + F_{fp}^\nu(x)$$

$$\partial_\mu T_{fb}^{\mu\nu}(x) = F_p^\nu(x) + F_t^\nu(x) - F_{fp}^\nu(x) - F_{ft}^\nu(x)$$

- interaction between **target** and **projectile**:
friction terms $-F_t^\nu(x)$ and $-F_p^\nu(x)$
- interaction between **fireball** and **target/projectile**:
friction terms $F_{fp}^\nu(x)$ and $F_{ft}^\nu(x)$

Summary

- at (extremely) large temperatures and densities hadrons “melt” to quark-gluon plasma where quarks and gluons are the basic degrees of freedom
- quark-gluon plasma behaves like fluid with very low specific shear viscosity
- models for lower collision energies/higher baryon densities under construction