



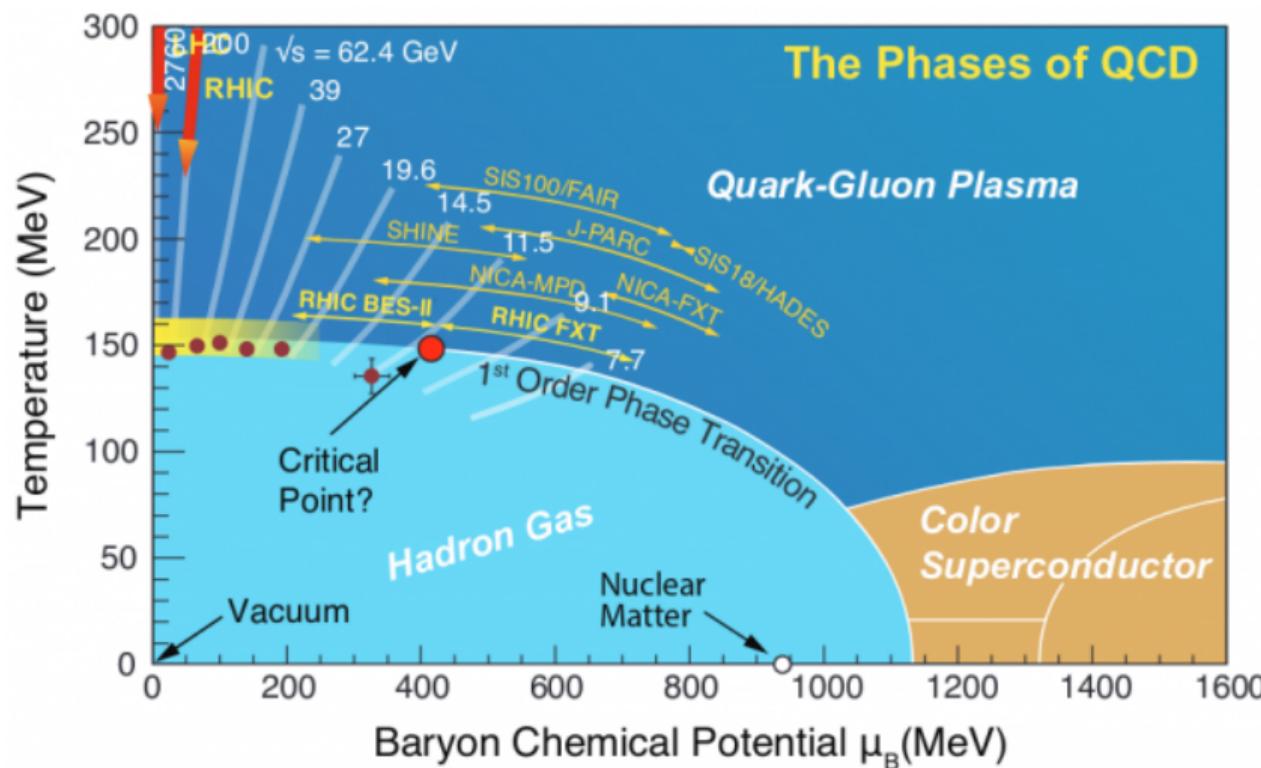
# Efficient solver for relativistic hydrodynamics with implicit Runge-Kutta method

Nathan Touroux<sup>1</sup>  
Marlene Nahrgang, Masakiyo Kitazawa, Koichi Murase

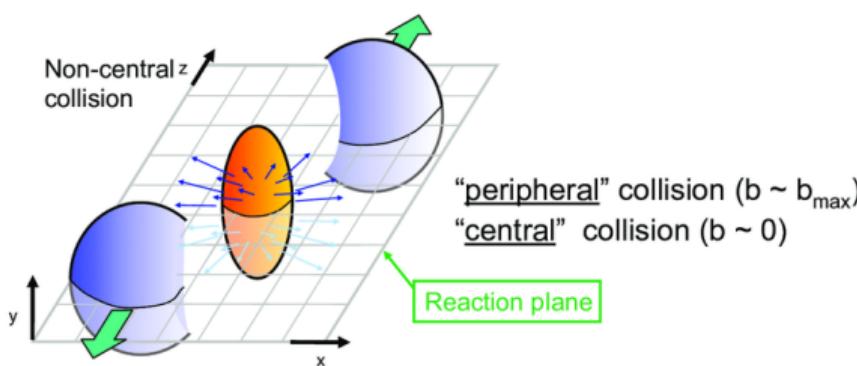
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<sup>1</sup>Subatech, Osaka University, YITP

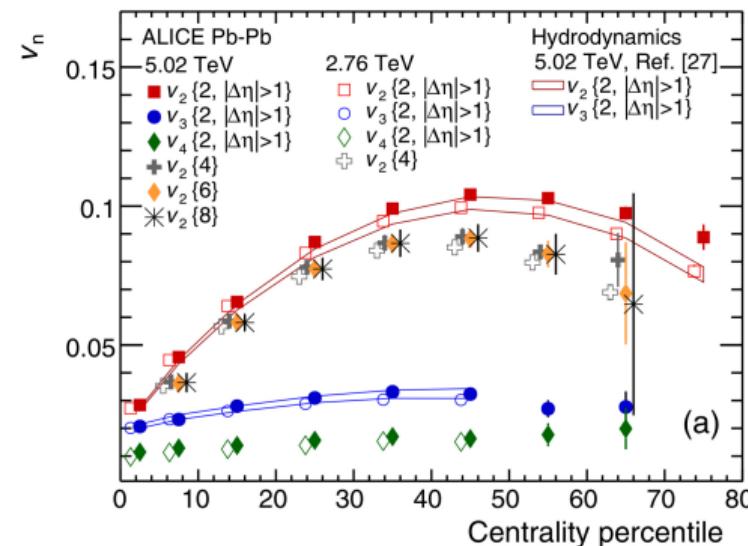
# QCD phase diagram



# Why hydrodynamics?



Almond shape → elliptic flow



J. Adam et al, PRL 116, 132302 (2016)

# Relativistic hydrodynamics

Energy-momentum tensor conservation:

$$\partial_{;\mu} T^{\mu\nu} = \partial_\mu T^{\mu\nu} + \Gamma_{\alpha\mu}^\mu T^{\alpha\nu} + \Gamma_{\alpha\mu}^\nu T^{\mu\alpha} = 0$$

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

$$\Delta_{\mu\nu} = g_{\mu\nu} - u_\mu u_\nu, \quad u^\mu u_\mu = 1,$$

Israel-Stewart equations:

$$u^\lambda \partial_{;\lambda} \Pi = -\frac{\Pi - \Pi_{NS}}{\tau_\Pi} - \frac{4}{3} \Pi \partial_{;\lambda} u^\lambda$$

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

$$\Pi_{NS} = -\zeta \partial_{;\lambda} u^\lambda,$$

$$\pi_{NS}^{\mu\nu} = \eta (\Delta^{\mu\lambda} \partial_{;\lambda} u^\nu + \Delta^{\nu\lambda} \partial_{;\lambda} u^\mu) - \frac{2}{3} \eta \Delta^{\mu\nu} \partial_{;\lambda} u^\lambda,$$

$$\tau_\pi = \tau_\Pi = \frac{5\eta}{\epsilon + P}$$

# Relativistic Hydrodynamics

Rewriting the equations:

$$\partial_\tau (\tau T_{\text{id}}^{\tau\nu}) = -\partial_x (\tau T^{x\nu}) - \partial_y (\tau T^{y\nu}) - \partial_\eta (T^{\eta\nu}) + A(\partial_\tau u^\nu, \partial_\tau \Pi, \partial_\tau \pi^{\tau\nu}) \quad (1)$$

$$\partial_\tau (u^\tau \pi^{\mu\nu}) = -\partial_x (u^x \pi^{\mu\nu}) - \partial_y (u^y \pi^{\mu\nu}) - \partial_\eta (u^\eta \pi^{\mu\nu}) + B(\partial_\tau u^\nu, \partial_i u^\nu) \quad (2)$$

$$\partial_\tau (u^\tau \Pi) = -\partial_x (u^x \Pi) - \partial_y (u^y \Pi) - \partial_\eta (u^\eta \Pi) + C(\partial_\tau u^\nu, \partial_i u^\nu) \quad (3)$$

where

$$A(\partial_\tau u^\nu, \partial_\tau \Pi, \partial_\tau \pi^{\tau\nu}) = I_T^\nu - \partial_\tau (\tau [-\Pi \Delta^{\tau\nu} + \pi^{\tau\nu}]), \quad (4)$$

$$B(\partial_\tau u^\nu, \partial_i u^\nu) = -\frac{\tilde{\pi}^{\mu\nu} - \tilde{\pi}_{\text{NS}}^{\mu\nu}}{\tau_\pi} - \frac{1}{3} \tilde{\pi}^{\mu\nu} \left( \tilde{\partial}_\lambda \tilde{u}^\lambda - 4 \frac{\tilde{u}^\tau}{\tau} \right) \quad (5)$$

$$- (\tilde{u}^\nu \tilde{\pi}^{\mu\beta} + \tilde{u}^\mu \tilde{\pi}^{\nu\beta}) \tilde{u}^\lambda \tilde{\partial}_\lambda \tilde{u}_\beta + \frac{\tilde{u}^\eta}{\tau} I_\pi^{\mu\nu}, \quad (6)$$

$$C(\partial_\tau u^\nu, \partial_i u^\nu) = -\frac{\tilde{\Pi} - \tilde{\Pi}_{\text{NS}}}{\tau_\Pi} - \frac{1}{3} \tilde{\Pi} \left( \tilde{\partial}_\lambda \tilde{u}^\lambda - 4 \frac{\tilde{u}^\tau}{\tau} \right), \quad (7)$$

# Numerical scheme

Hydro equations with the dynamical variables  $y = (\tau T_{\text{id}}^{\tau\nu}, u^\tau \pi^{\mu\nu}, u^\tau \Pi)$ :

$$\partial_t y = f(t, y)$$

Kurganov-Tadmor space discretization:  $f(t, y) = f_{KT}(t, y) + O(\Delta x^2)$

- **Independent of time discretization**
- Numerical diffusion  $\propto \Delta x^2$
- Flux limiter → avoid numerical oscillations

Runge-Kutta time discretization

$$y(t + \Delta t) = y(t) + \Delta t \sum_{j=1}^n b_j k_j$$

$$k_i = f(t, y(t) + \Delta t \sum_{j=1}^n a_{ij} k_j) \quad \text{“}\vec{K} = \vec{F}(\vec{K})\text{”}$$

# Extending Runge-Kutta

Considering that

$$k_i = (\partial_\tau (\tau T_{\text{id}}^{\tau\nu}), \partial_\tau (u^\tau \pi^{\mu\nu}), \partial_\tau (u^\tau \Pi)), \quad (8)$$

we can hope to express the right-hand-side time derivatives by extending Runge-Kutta as either:

implicit

explicit

$$k_i = f(t, y(t) + \Delta t \sum_{j=1}^n a_{ij} k_j, k_i)$$

$$k_1 = f(t, y(t) + \Delta t, k_n(t - \Delta t))$$

$$k_2 = f(t, y(t) + \Delta t a_{21} k_1, k_1)$$

$$k_3 = f(t, y(t) + \Delta t(a_{31} k_1 + a_{32} k_2), k_2)$$

...

# Right-hand-side time derivatives partial solving

By considering  $\partial_\tau(\tau T_{\text{id}}^{\tau\nu})$ ,  $\partial_\tau(u^\tau \pi^{\mu\nu})$  and  $\partial_\tau(u^\tau \Pi)$  known, we can express  $\partial_\tau u^\nu$ ,  $\partial_\tau \Pi$  and  $\partial_\tau \pi^{\tau\nu}$ <sup>2</sup>:

$$\partial_\tau \pi^{\tau\nu} = \frac{\partial_\tau(u^\tau \pi^{\tau\nu}) - \pi^{\tau\nu} \partial_\tau u^\tau}{u^\tau} \quad (9)$$

$$\partial_\tau \Pi = \frac{\partial_\tau(u^\tau \Pi) - \Pi \partial_\tau u^\tau}{u^\tau} \quad (10)$$

$$\partial_\tau u^\tau = -\frac{u^x \partial_\tau u^x + u^y \partial_\tau u^y + u^\eta \partial_\tau u^\eta}{u^\tau} \quad (11)$$

$$(\partial_\tau \epsilon \quad \partial_\tau u^x \quad \partial_\tau u^y \quad \partial_\tau u^\eta)^T = \mathbb{M}^{-1} \partial_\tau T_{\text{id}}^{\tau\nu} \quad (12)$$

$$\partial_\tau T_{\text{id}}^{\tau\nu} = \frac{\partial_\tau(\tau T_{\text{id}}^{\tau\nu}) - T_{\text{id}}^{\tau\nu}}{\tau} \quad (13)$$

where the 4 by 4 matrix  $\mathbb{M} = \left[ \frac{\partial(\partial_\tau T_{\text{id}}^{\tau\nu})}{\partial(\partial_\tau \epsilon \quad \partial_\tau u^x \quad \partial_\tau u^y \quad \partial_\tau u^\eta)} \right]$  can be inverted analytically.

<sup>2</sup>Inspired by rfh code (Koichi Murase)

# Choice of Runge-Kutta coefficients

Accuracy order:

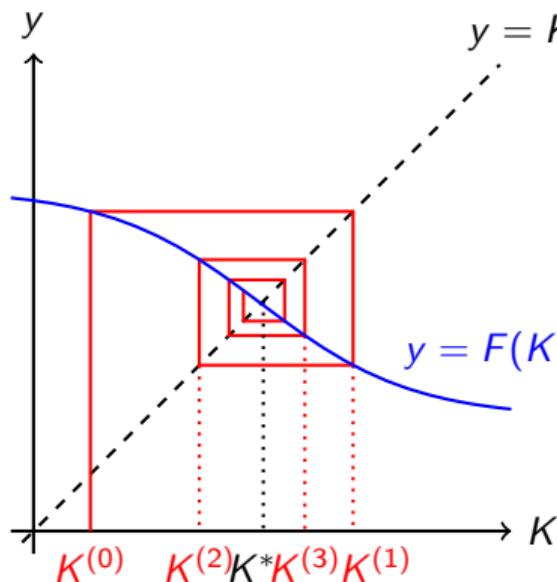
$$\|y^*(t) - y(t)\| < C\Delta t^p$$

Second order choices:

	Heun	Gauss-Legendre 1 (GL1)
Type	Explicit	Implicit
Stage $S$	2	1
Order $p$	2	2
$c_n$	0   0 0	0.5   0.5
$a_{nm}$	1   1 0	
$b_m$	0.5   0.5	1

# Implicit solver

Solve implicit equation " $\vec{K} = \vec{F}(\vec{K})$ " by the fixed-point solver



$$\vec{K}^{(l+1)} = \vec{F}(\vec{K}^{(l)})$$

$\vec{K}^{(0)} = \vec{0}$  or last time step solution

Solve iteratively

$$\vec{K}^{(0)} \rightarrow \vec{K}^{(1)} \rightarrow \vec{K}^{(2)} \rightarrow \dots \rightarrow \vec{K}^{(l+1)}$$

# Local optimization

- ➊ Update all cells once to obtain  $\vec{K}^{(1)}$
- ➋ check the convergence in every cell  $[\bullet]_j$ :

$$\|[\vec{F}(\vec{K}^{(l)})]_j - [\vec{K}^{(l)}]_j\| < e \frac{\langle T^{\tau\tau} \rangle}{\Delta\tau} \left( \frac{\Delta\tau}{\Delta x} \right)^{(p+1)}$$

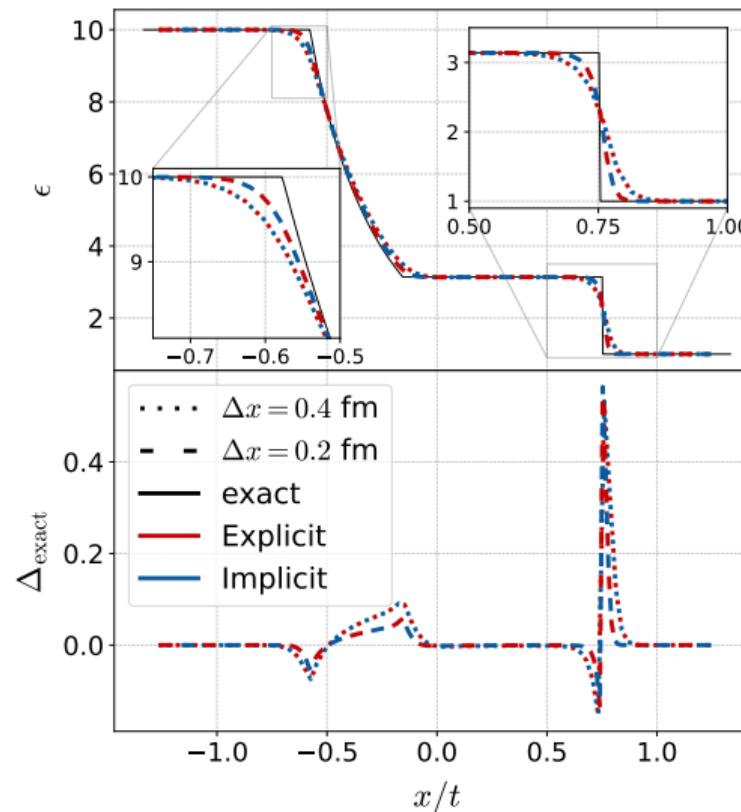
- ➌ To obtain  $\vec{K}^{(l+1)}$  ( $l+1 \geq 2$ ), only update a cell if itself or any surrounding cells does not satisfy the threshold
- ➍ Repeat (2) and (3) until all cells satisfy the threshold

**This dramatically reduces the computation cost.**

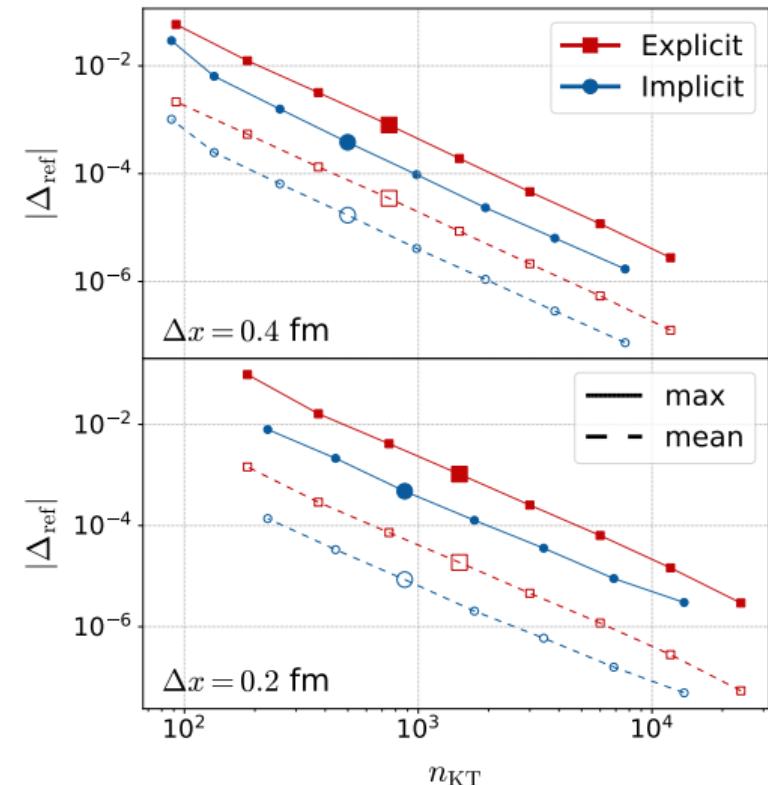
Disclaimer: This partial update breaks conservation and leads to inconsistencies, but this is controlled within the error threshold.

# 1D ideal Riemann problem

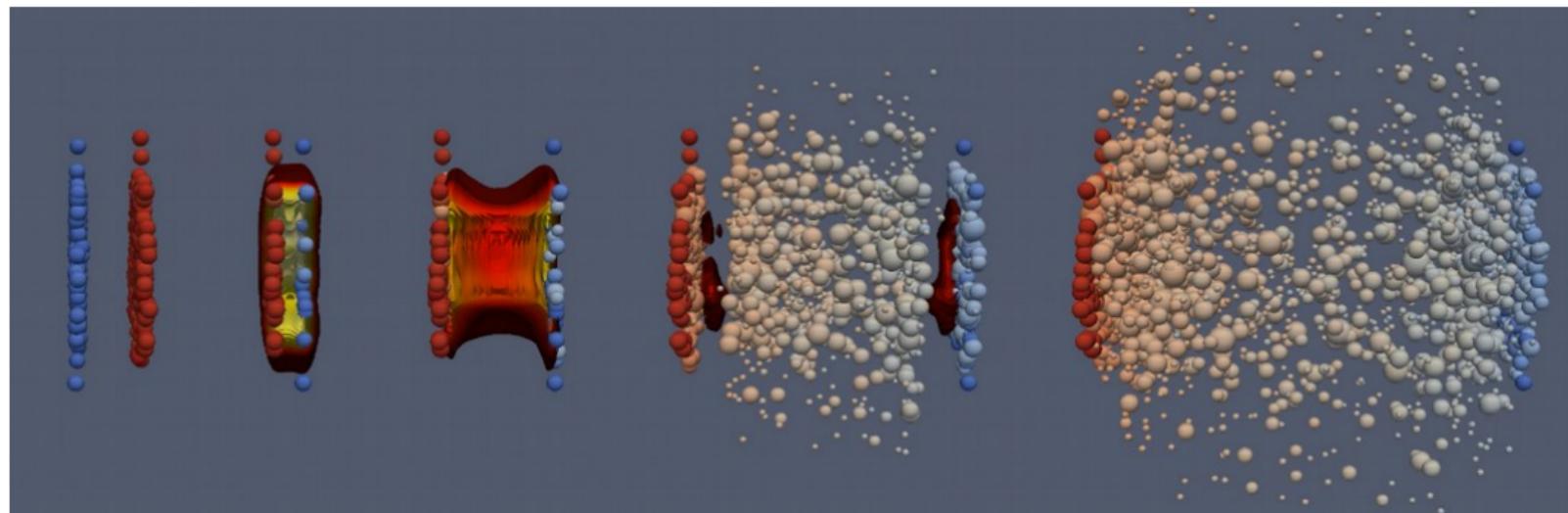
Vanishing  $\Delta t$  comparison



Error-cost comparison



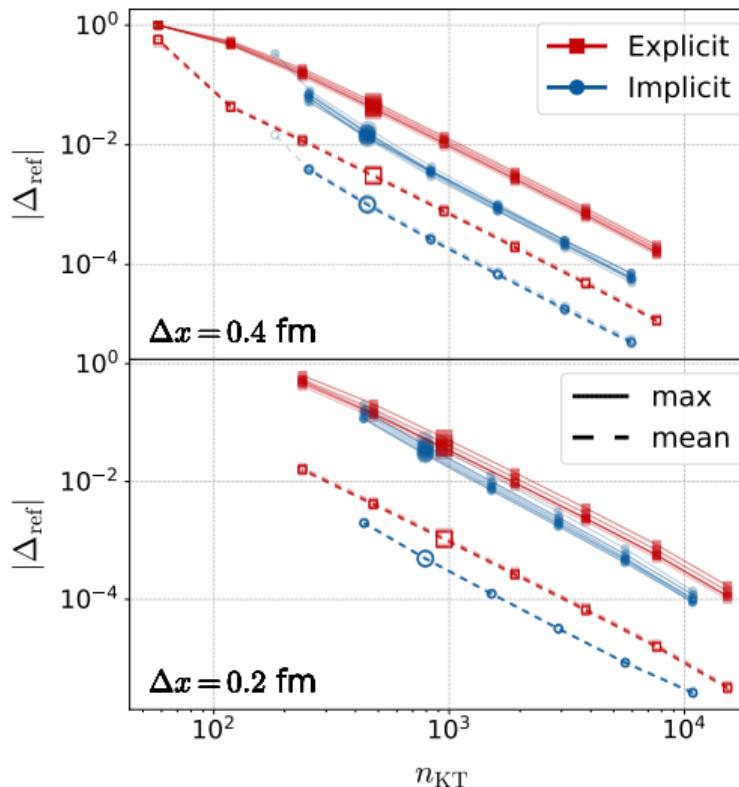
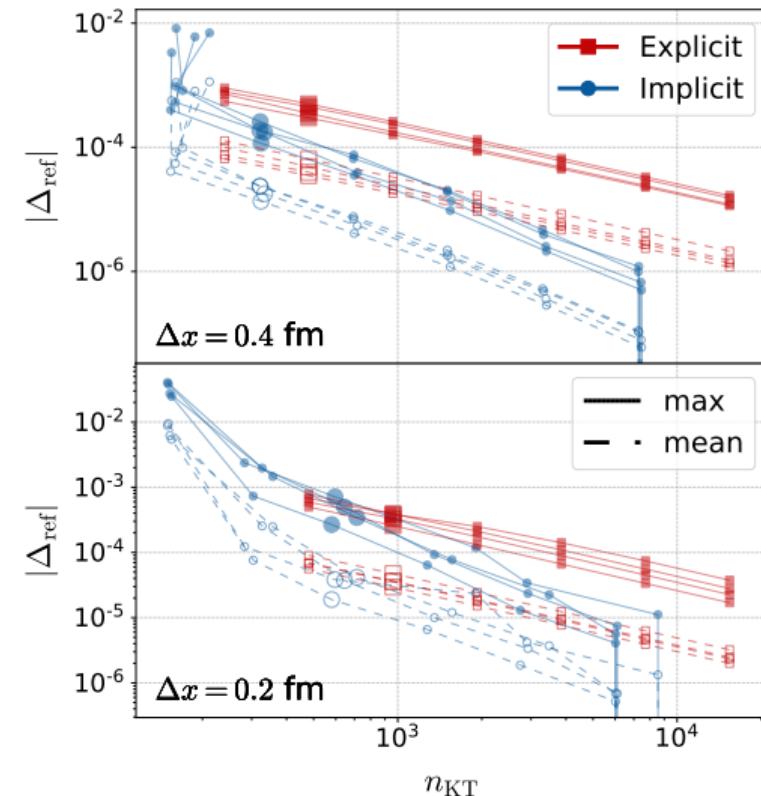
# Heavy ion collisions evolution



Collision	thermalization Pre-hydrodynamics	QGP Hydrodynamics	Particilization Cooper-Frye	Decay/free-streaming Cascade/Kinetic
Trento	skipped	ImplHydro	frzout	UrQMD

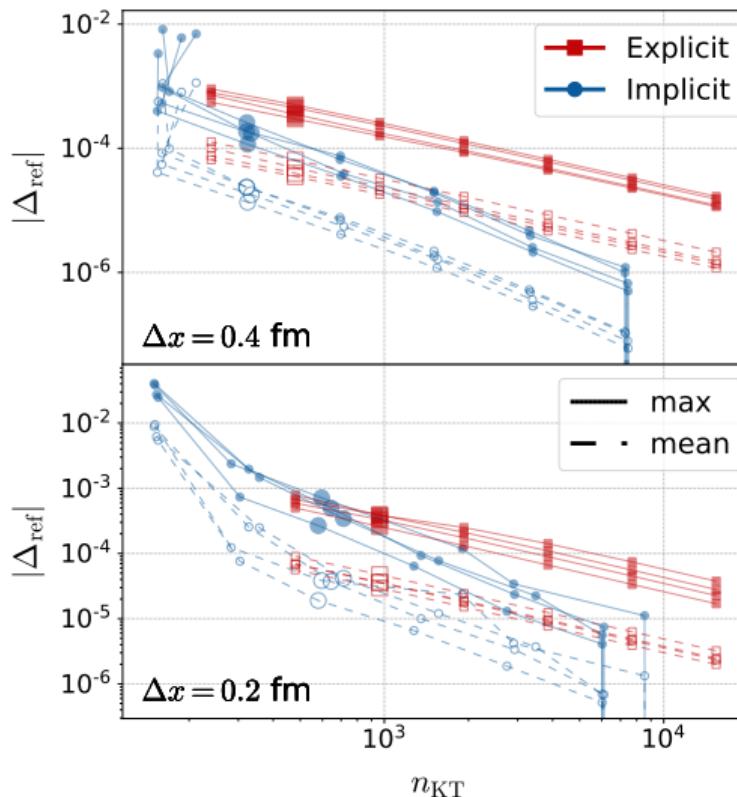
Parameters from: PRC 101, 024911 (2020)

# Trento initial conditions: 2D ideal vs viscous

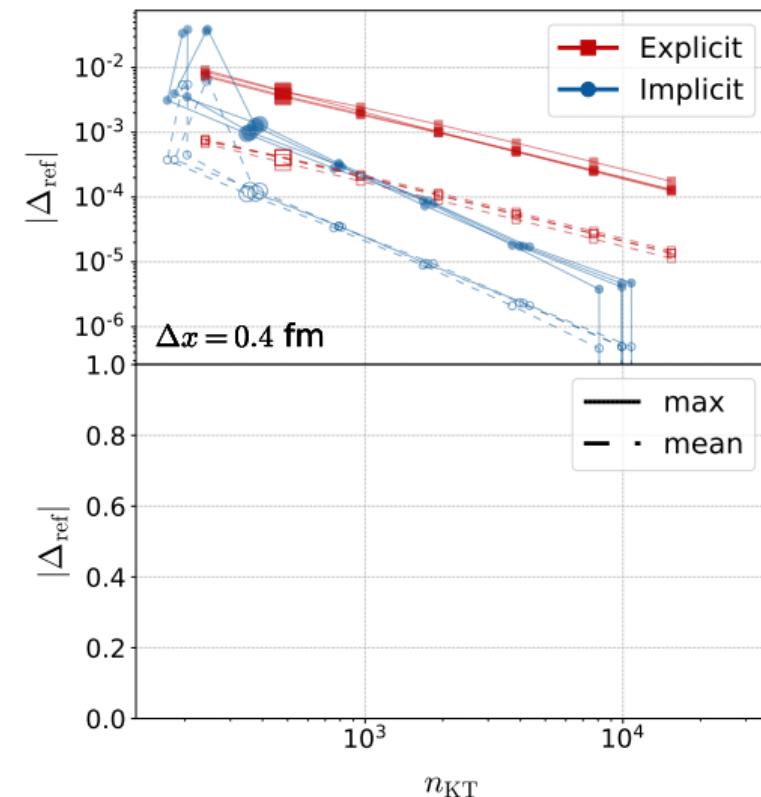
**Ideal****Viscous**

# Trento initial conditions: viscous 2D vs 3D

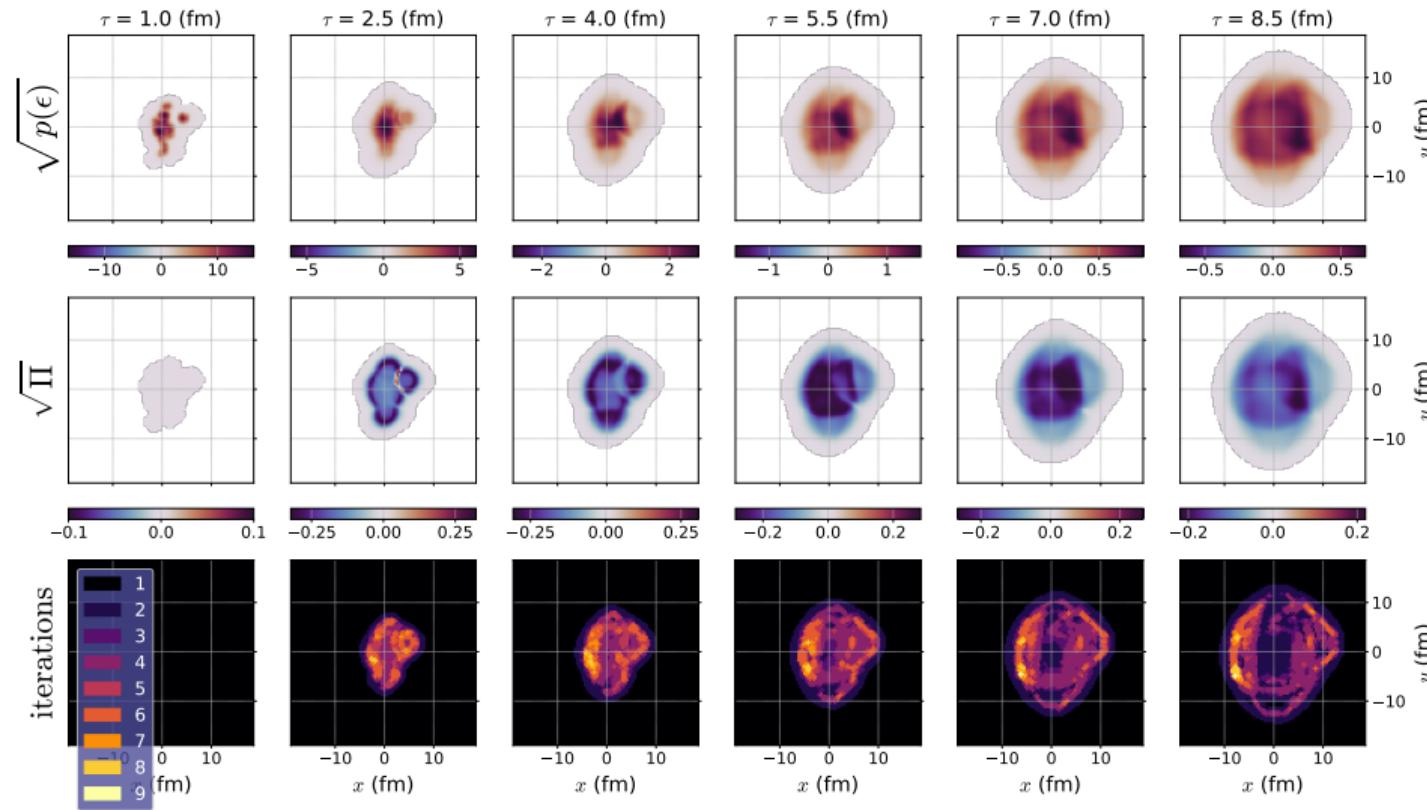
2D



3D

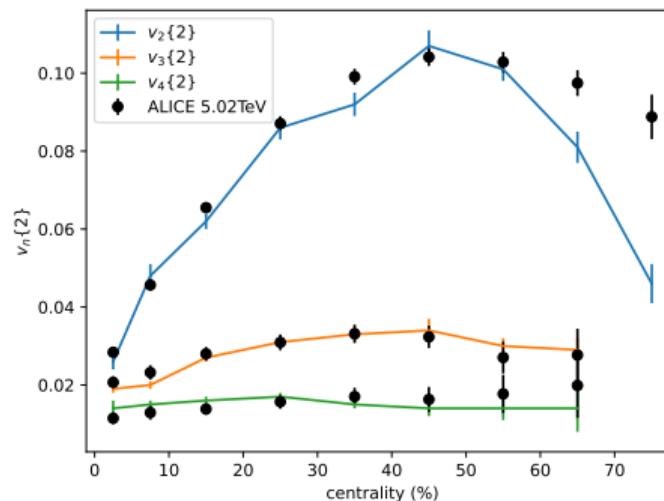


# Trento initial conditions: viscous 3D

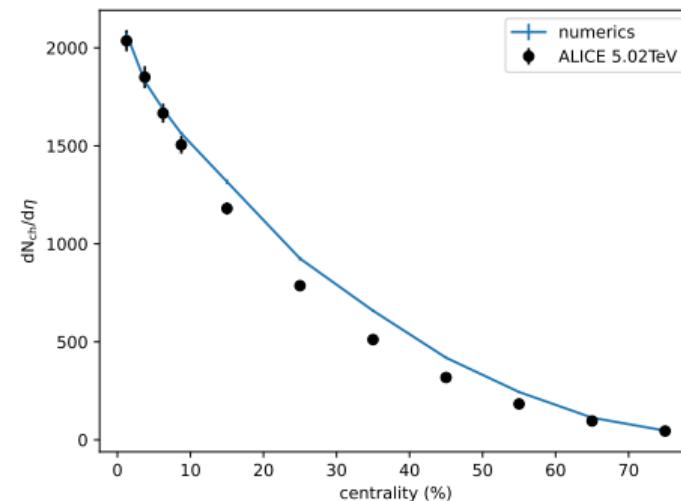


# Physical results

## Elliptic flow



## Particle spectra



# Conclusion

## Achievement

- Implicit method can be more efficient and accurate than explicit
- First true 2<sup>nd</sup> order time integrator for viscous hydrodynamics
- Detects stiffness through convergence of the fixed-point iterator

## Outlook

- Adapt  $\Delta t$  depending on convergence for unconditional stability
- Including other charges (baryon number, electric charge, ...)
- Including fluctuations

## ImplHydro:

- arXiv:2306.12696
- Open source: <https://github.com/xayon40-12/ImplHydro.git>



## Enforcing positivity of the effective pressure

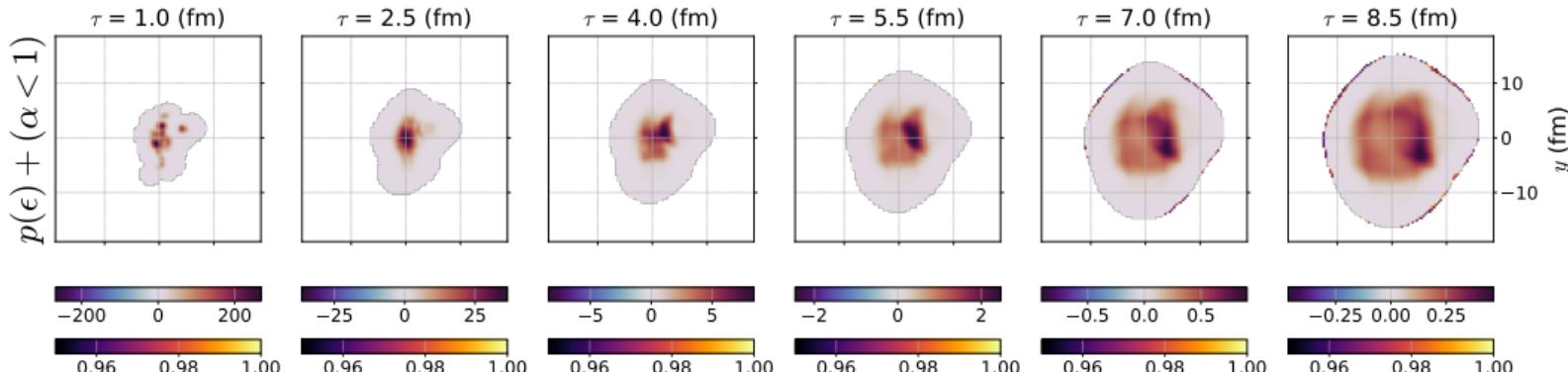
We need to garanty that

$$\epsilon + P(\epsilon) + \Pi + \lambda_{\min} \geq 0, \quad (14)$$

where  $\lambda_{\min} = \min_\mu \{\lambda^\mu\}$  is the smallest of the eigenvalues  $\lambda^\mu$  of the shear tensor  $\pi^{\mu\nu}$ . If it is not the case, we rescale the bulk pressure and shear tensor as

$$\pi^{\mu\nu} \leftarrow \alpha\pi^{\mu\nu}$$

$$\alpha = -\frac{\epsilon + P(\epsilon)}{\Pi + \lambda_{\min}}$$



## Right hand side space derivatives

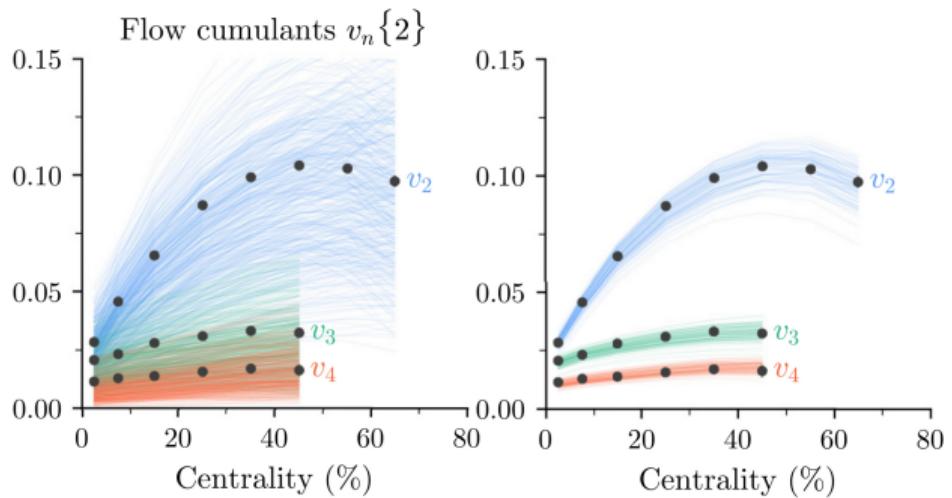
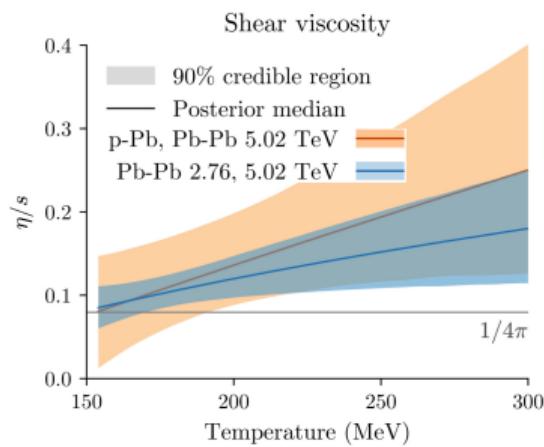
In order to approximate the space derivatives  $\partial_i u^\mu$  in  $A$ ,  $B$  and  $C$ , we first reconstruct the values of the fields at the boundaries of cells using the cubic interpolation (here in the  $x$  direction by considering that  $j$  indexes the  $x$  coordinate):

$$[u^\mu]_{j \pm 1/2} \approx -\frac{1}{16}[u^\mu]_{j \pm 1/2 - 3/2} + \frac{9}{16}[u^\mu]_{j \pm 1/2 - 1/2} + \frac{9}{16}[u^\mu]_{j \pm 1/2 + 1/2} - \frac{1}{16}[u^\mu]_{j \pm 1/2 + 3/2} \quad (15)$$

and then perform the finite difference to approximate

$$\partial_x u^\mu \approx \frac{[u^\mu]_{j+1/2} - [u^\mu]_{j-1/2}}{\Delta x}. \quad (16)$$

# Simulation framework



PRC 101, 024911 (2020)

## KT 2<sup>nd</sup> order algorithm

$$\begin{aligned}
 KT[f, \rho(f), y, \Delta x]_i &= \frac{H_{i+1/2} - H_{i-1/2}}{\Delta x} \\
 H_{i+1/2} &= \frac{f(y_{j+1/2}^+) + f(y_{j+1/2}^-)}{2} + \frac{a_{i+1/2}}{2} (y_{j+1/2}^+ - y_{i+1/2}^-) \\
 a_{i+1/2} &= \max \left\{ \rho \left( \frac{\partial f}{\partial y} \left( y_{i+1/2}^+ \right) \right), \rho \left( \frac{\partial f}{\partial y} \left( y_{i+1/2}^- \right) \right) \right\} \\
 y_{i+1/2}^+ &= y_{i+1} - \frac{\Delta x}{2} (\partial_x y)_{i+1} \\
 y_{i+1/2}^- &= y_i + \frac{\Delta x}{2} (\partial_x y)_i \\
 (\partial_x y)_i &= \text{minmod} \left( \theta \frac{y_i - y_{i-1}}{\Delta x}, \frac{y_{i+1} - y_{i-1}}{2\Delta x}, \theta \frac{y_{i+1} - y_i}{\Delta x} \right) \\
 1 \leq \theta &\leq 2
 \end{aligned}$$

needs  $y_{i-2}, y_{i-1}, y_i, y_{i+1}, y_{i+2}$