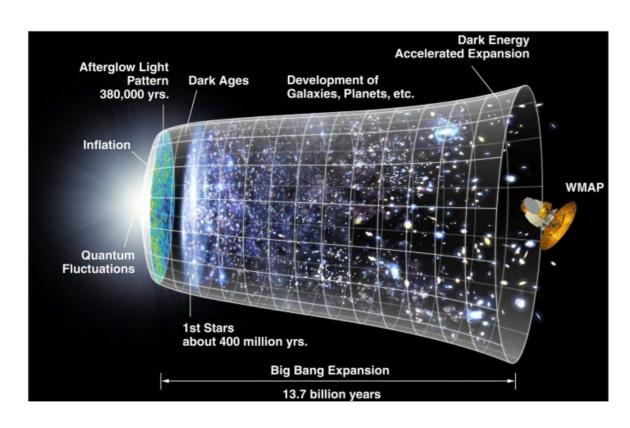
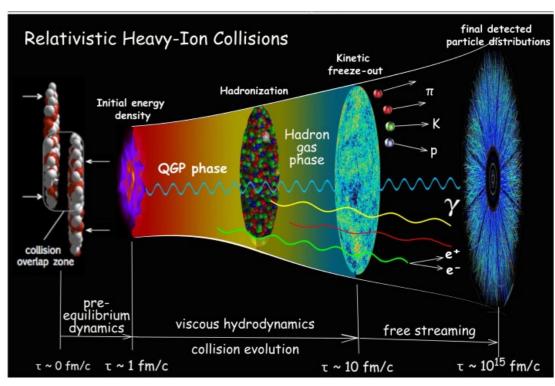
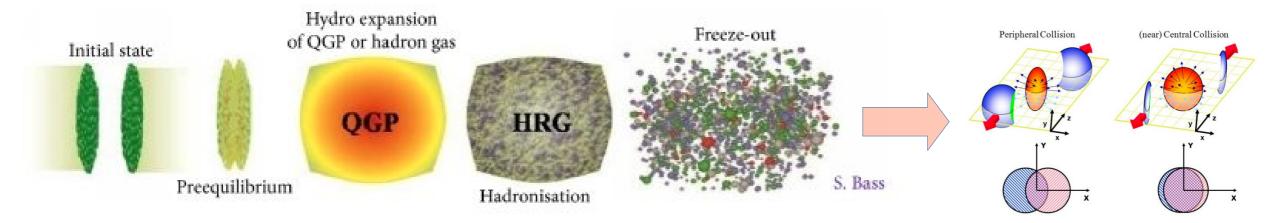
# G.G. Barnaföldi *et al*: Deep learning predicted elliptic flow of identified particles at the RHIC & LHC energies







## QGP signature: elliptic flow (v<sub>2</sub>) in HIC



Elliptic flow describes the azimuthal momentum space anisotropy of particle emission for a non-central heavy-ion collision.

$$E\frac{d^{3}N}{dp^{3}} = \frac{d^{2}N}{p_{T}dp_{T}dy} \frac{1}{2\pi} \left( 1 + 2\sum_{n=1}^{\infty} v_{n} \cos[n(\phi - \psi_{n})] \right)$$

The 2<sup>nd</sup> harmonic coefficient of the Fourier expansion of azimuthal momentum distribution:  $v_2(p_T, y) = \langle \cos(2(\phi - \psi_2)) \rangle$ 

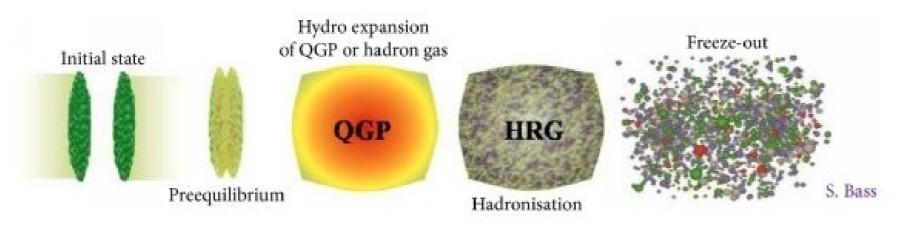
Azimuthal momentum space anisotropy

Spatial anisotropy

# QGP signature: elliptic flow (v<sub>2</sub>) in HIC

(near) Central Collision

TIDITISVERPETSI



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### Results: how ML can help with this?

### It is possible to estimate the elliptic flow by ML

- Get best Min. Bias. Monte Carlo simulation data and train the well-designed DNN system...
  - → AMPT & DNN correlates well for all centrality
  - → Best correlation is for the highest statistic
  - → Energy scaling is well preserved (non-linear)
  - → The  $v_2(p_T)$  is also preserved with PID & NCQ

#### See more on poster #105

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Refs.: PRD 105, 114022 (2022) PRD 107, 094001(2023)

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