

# Bayesian inference of the EoS of dense nuclear matter from heavy-ion collision data

Manjunath Omana Kuttan, Jan Steinheimer, Kai Zhou, Horst Stöcker

# Density dependent EoS in UrQMD

- ♦ A density dependent potential enters the QMD equations:  $\mathbf{r}_i = \frac{1}{\partial \mathbf{p}_i}, \quad \mathbf{p}_i = \frac{1}{\partial \mathbf{p}_i}$ 
  - $\succ$  The potential energy term of the Hamiltonian  ${f H}$  is density dependent.
- The potential energy  $V(n_B)$  is related to the pressure as:

$$P(n_B) = P_{
m id}(n_B) + \int_0^{n_B} n' rac{\partial U(n')}{\partial n'} dn' \ , \ U(n_B) = rac{\partialig(n_B\cdot V(n_B)ig)}{\partial n_B}$$

 $U(n_B) \Rightarrow$ single particle potential,  $P_{id}(n_B) \Rightarrow$  pressure of an ideal Fermi gas of baryons

constraining the potential energy  $V(n_B) \Rightarrow$  constraining the EoS

 $\partial \mathbf{H}$ 

EPJ C 82.5 (2022): 1-12.

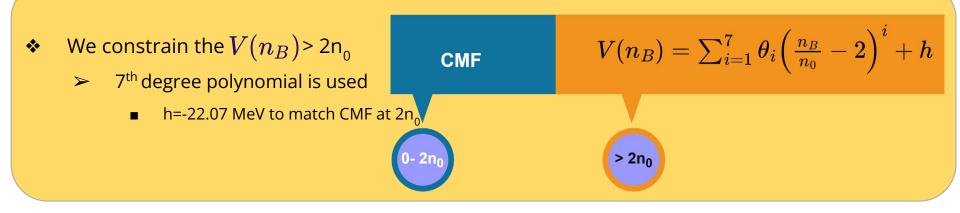
 $\partial \mathbf{H}$ 

 $\partial \mathbf{r}_i$ 

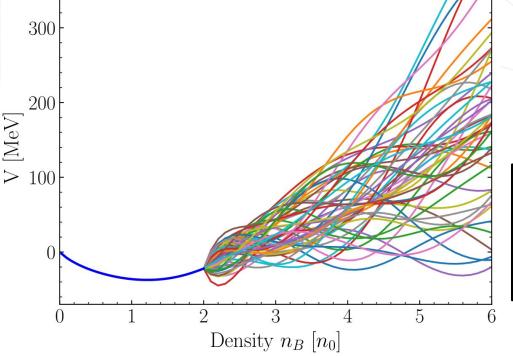
EPJ C 82.10 (2022):

# Parameterisation of the potential energy

- Upto 2n<sub>0</sub>, EoS reasonably constrained by
  - flow data at moderate energies P. Danielewicz, Et al Science 298, 1592 (2002), H. Kruse, Et al. Phys. Rev. Lett. 54, 289 (1985)
  - nuclear incompressibility data Y. Wang, Et al. PLB 778, 207 (2018)
  - bayesian analysis s. Huth et al., Nature 606, 276 (2022)
- Upto 2n<sub>0</sub>, CMF model-fit is used A. Motornenko et al., PRC 103.5 (2021), A. Motornenko et al., PRC 101.3 (2020)



# Few examples...



$$V(n_B) = \sum_{i=1}^7 heta_i \left(rac{n_B}{n_0} - 2
ight)^i + h_i$$

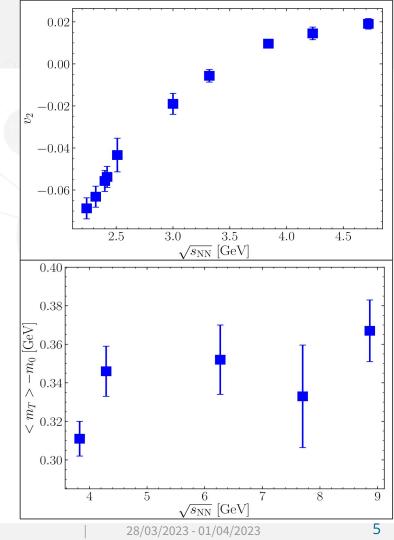
We constrain the polynomial coefficients

$$oldsymbol{ heta} = \{ heta_1, heta_2, \dots, heta_7\}$$

# The experimental data

- Proton observables (mid rapidity)
  - > Elliptic flow : 10 data points
    - E895, CERES, FOPI, STAR, HADES
    - Mid-central collisions
  - Transverse kinetic energy: 5 data points
    - E802, NA49, STAR
    - Central collisions

The data, 
$$\mathbf{D}=\{v_2^{exp},\langle m_T
angle^{exp}-m_0\}$$
 is used to constrain the parameters  $oldsymbol{ heta}$  .



MAGIC23, Kerala, India

# Fast emulators are necessary

- Bayesian inference involves numerous UrQMD simulations
  - UrQMD ~100 s/event

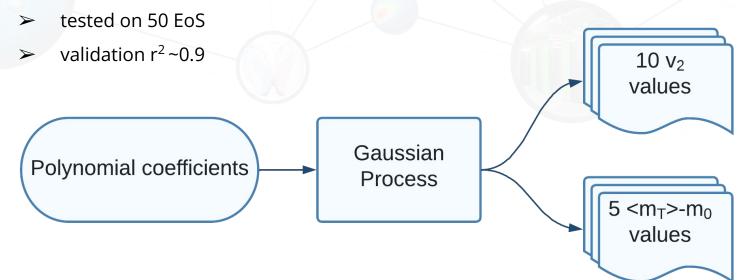
observable	No of beam energies	No of events per beam energy
V <sub>2</sub>	10	12000
<m<sub>T&gt;-m<sub>0</sub></m<sub>	5	1000

- For one parameter set ~125000 events\* 100 s = ~3500 hrs
- MCMC sampling requires random walk through 1000s of parameter sets

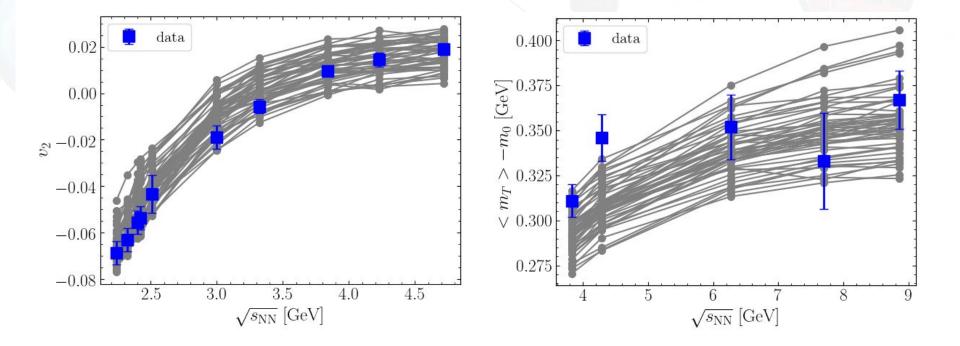
#### not feasible to run UrQMD for MCMC !

# Training the GP models

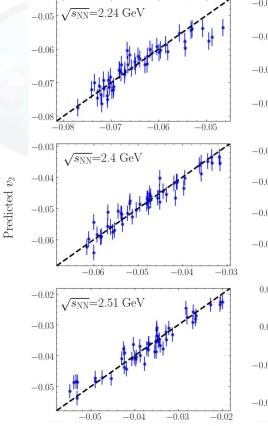
- Gaussian Process models are trained as fast emulators
  - trained on 200 randomly generated EoSs



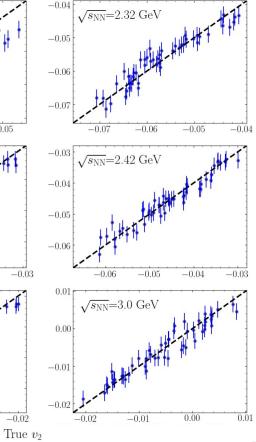
# Potentials for training the GP models



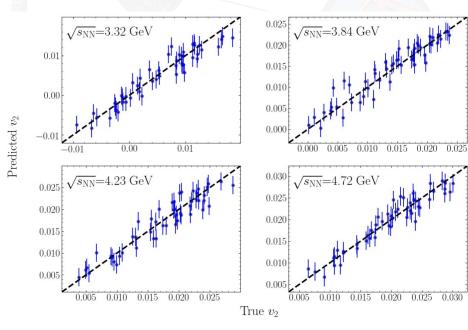
## GP models: performance



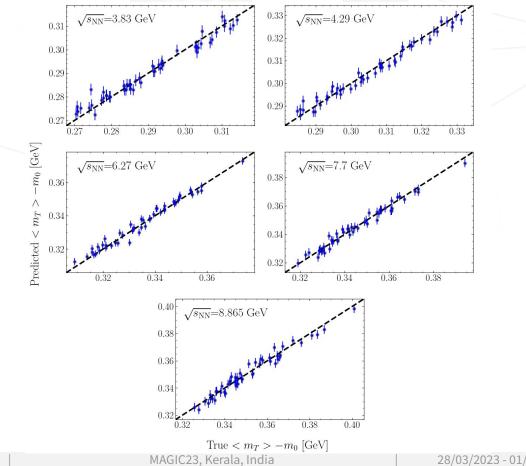
Manjunath Omana Kuttan



MAGIC23, Kerala, India



## GP models: performance

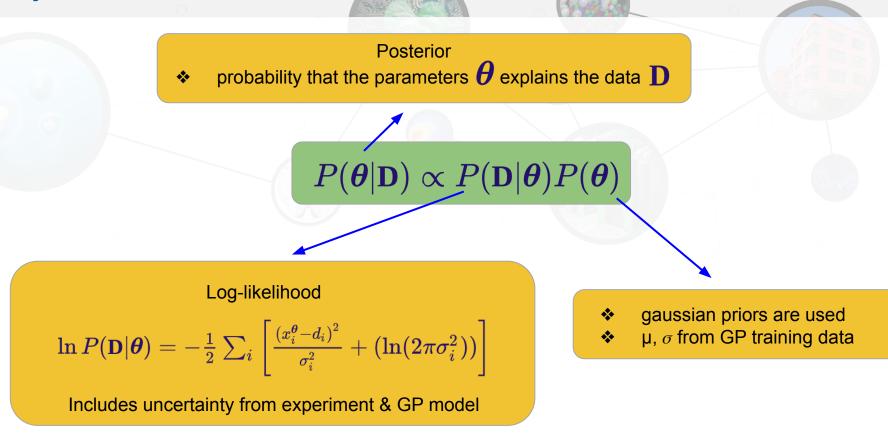


Manjunath Omana Kuttan

28/03/2023 - 01/04/2023

10

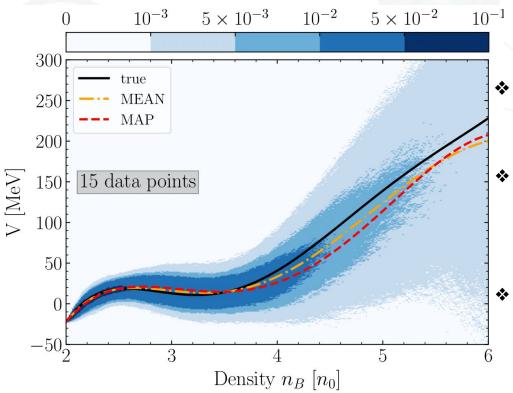
# Bayesian inference



# Closure test

- 1. Consider a random EoS as "ground-truth"
- 2. Calculate observables from UrQMD
  - a. 10 values of  $v_2$  and 5 values of  $\langle m_T \rangle m_0$
- 3. Assume the UrQMD observables for this EoS to be the "data"
  - a. assume uncertainty similar to experimental data
- 4. Construct the posterior
- 5. Compare with the "ground-truth"

# **Closure test: results**

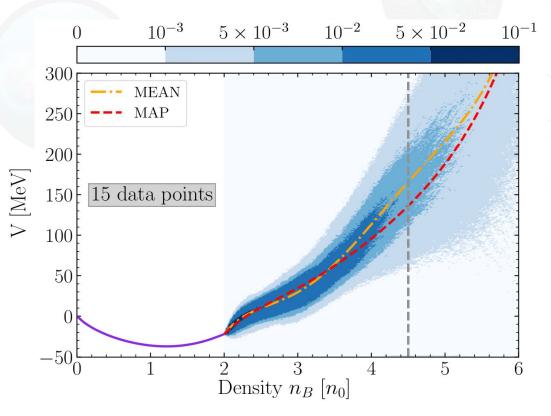


- Tight constraints up to 4n<sub>0</sub>
  - > large uncertainty above  $4n_0$
  - yet mean closely follows "ground-truth"

#### • Two curves extracted:

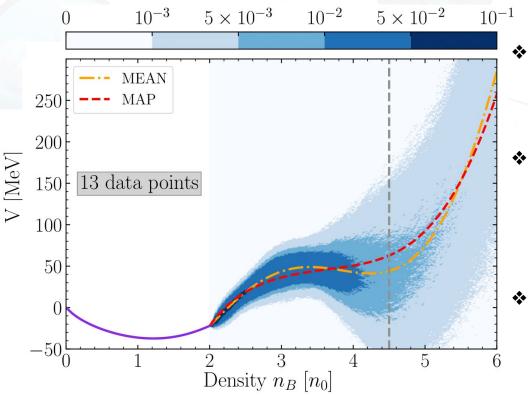
- "MAP": mode of posterior
- "MEAN": mean of posterior
- MEAN and MAP closely follows "ground-truth" upto 6n<sub>0</sub>

# Result from experimental data



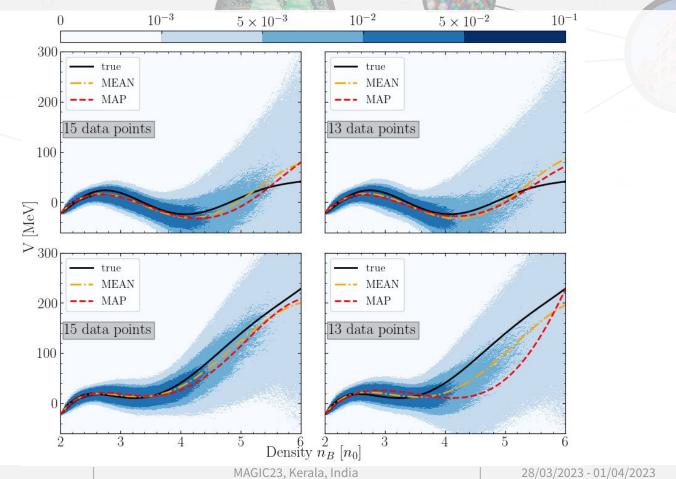
- Posterior from experimental data
  - > 10 measurements of  $v_2$
  - > 5 measurements of  $< m_T > -m_0$
- Tight constraints upto 4n<sub>0</sub>
  - MEAN, MAP suggests stiff EoS
  - > No phase transition

# Sensitivity to choice of observables



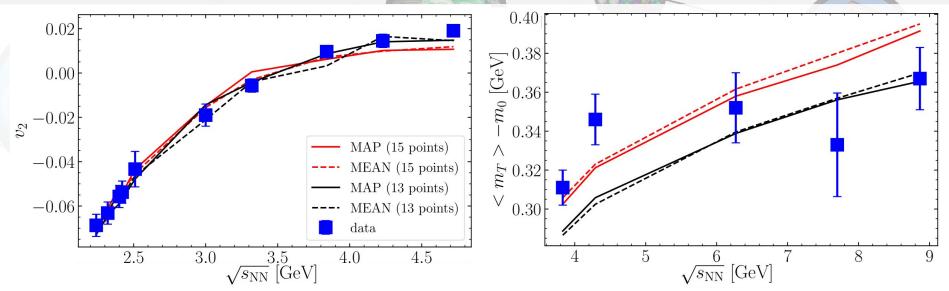
- Only 13 data points are used
  - <m<sub>T</sub>>-m<sub>0</sub> at 3.83, 4.29 GeV not used
  - Significant differences in posterior
    - ➤ softening at 3- 5n<sub>0</sub>
    - phase transition
- Beyond 3n<sub>0</sub> strong dependence to choice of observables

# More closure test results



Manjunath Omana Kuttan

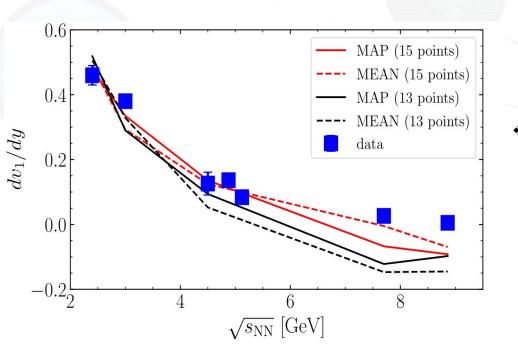
# Reconstructed EoSs: v<sub>2</sub>, <m<sub>7</sub>>-m<sub>0</sub>



• better  $v_2$  predictions at high energies when 2 data points are removed

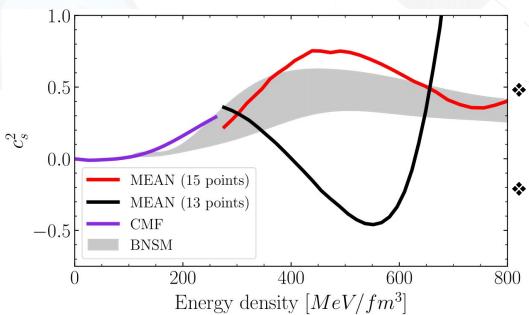
- > but also results in lower  $< m_t > -m_0$
- large  $< m_t > -m_0$  values for the stiff EoS (extracted using all data points)
- possible tension in data at ~4 GeV!
  - Measurement uncertainty? or limitation of the model?

# Reconstructed EoSs: dv<sub>1</sub>/dy



- dv<sub>1</sub>/dy data was not used for inference
  - yet consistent with reconstructed EoSs
    - especially with all 15 data points

# Reconstructed EoSs: c<sup>2</sup>



- 15 points, predicts a stiff EoS
  - consistent with astrophysical constraints arXiv:2203.14974 (2022)
    - broad peak structure
- 13 points, drastic drop in c<sub>s</sub><sup>2</sup>
  - ➤ first order phase transition

# Summary

- Bayesian inference using polynomial parameterization of the density dependence of EoS
  - > v<sub>2</sub>, <m<sub>t</sub>>-m<sub>0</sub> of protons are used for inference
- inference using all 15 data points:
  - > constraints the EoS upto  $4n_0$
  - > stiff EoS upto  $4n_0$ , no phase transition
  - consistent with BNSM constraint, dv<sub>1</sub>/dy data
- strong dependence on choice of observables for  $> 3n_0$
- tension in data at ~4 GeV
  - measurement uncertainty or model limitation?

# For stricter, robust constraints on the EoS below $4n_0$ , significant improvements and consistency in flow measurements are necessary for E\_lab = 2-10 A GeV

	tps://arxiv.c	- 0	11 11670	
	uprxiv.C	orglabs122		
nt	tps://arxiv			

# Backup slides

# Microscopic transport with density dependent potential

- Non-equilibrium MD part of UrQMD is used
- UrQMD:
  - > Propagation of hadrons on classic trajectories
    - stochastic binary scattering , color string formation, resonance excitation and decays
  - Imaginary part of interactions:
    - geometric interpretation of cross section
      - Experiment, detailed balance
  - Hadronic cascade
    - effective EoS of HRG with respective dof
- Real part of interactions in UrQMD
  - QMD + density dependent potential
    - Unlike other mean field models, QMD is an n-body theory of interactions between n nucleons

# Microscopic transport with density dependent potential

A density dependent potential enters QMD equations

 $\dot{\mathbf{r}}_i = rac{\partial \mathbf{H}}{\partial \mathbf{p}_i}, \quad \dot{\mathbf{p}}_i = -rac{\partial \mathbf{H}}{\partial \mathbf{r}_i}.$ 

The total hamiltonian function is sum over all hamiltonians of the i baryons

$$\mathbf{H} = \sum_i H_i, \;\; H_i = E_i^{kin} + V_i$$

This include KE and total potential energy V  $\mathbf{V} = \sum_i V_i \equiv \sum_i Vig(n_B(r_i)ig)$ 

The change in momentum for baryon 'i' is then

The local interaction density  $n_B at r_k$  is calc by assuming each particle as gaussian wave packet

$$egin{aligned} n_B(r_k) &= n_k = \sum_{j,j
eq k} n_{j,k} \ &= ig(rac{lpha}{\pi}ig)^{3/2} \sum_{j,j
eq k} B_j \exp\left(-lpha(\mathbf{r_k}-\mathbf{r_j})^2
ight) \ &lpha$$
=1/2L, L= 2 fm²

$$egin{aligned} \dot{\mathbf{p}}_i &= -rac{\partial \mathbf{H}}{\partial \mathbf{r}_i} = -rac{\partial \mathbf{V}}{\partial \mathbf{r}_i} \; \; n_{\{i,j\}} \equiv n_B(r_{\{i,j\}}) \ &= -\left(rac{\partial V_i}{\partial n_i} \cdot rac{\partial n_i}{\partial \mathbf{r}_i}
ight) - \left(\sum_{j 
eq i} rac{\partial V_j}{\partial n_j} \cdot rac{\partial n_j}{\partial r_i}
ight) \end{aligned}$$

Force on i<sup>th</sup> baryon depends on change in potential energy at point  $r_i$  due to local gradient of  $n_B(r_i)$  and change in potential at positions  $r_i$  of all baryons j due to change in  $r_i$ 

-solved in timestep 0.2fm/c

Manjunath Omana Kuttan

28/03/2023 - 01/04/2023

$$egin{aligned} P(n_B) &= P_{ ext{id}}(n_B) + \int_0^{n_B} n' rac{\partial U(n')}{\partial n'} dn' \,, \, U(n_B) &= rac{\partial ig(n_B \cdot V(n_B)ig)}{\partial n_B} \ \mu_B'(n_B) &= \mu_B^{id}(n_B) + U(n_B) \ \epsilon(n_B) &= -P(n_B) + \mu_B' n_B + sT \end{aligned}$$