



Examination of nucleon distribution with Bayesian imaging for isobar collisions

based on Phys. Rev. C 107, 064909

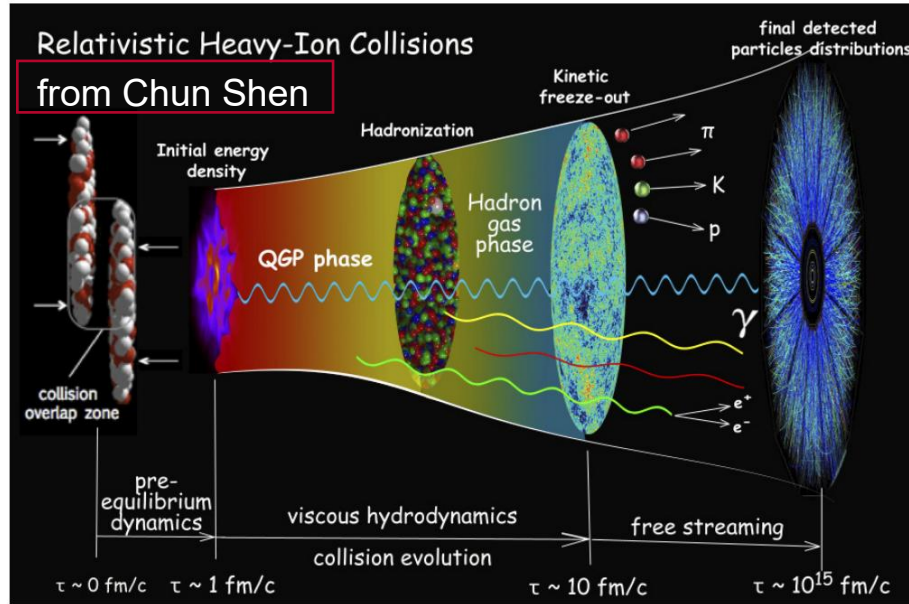
Yilin Cheng

Co-authors: Kai Zhou, Shuzhe Shi, Yugang Ma, Horst Stöcker



Motivation:

- Understanding Heavy-ion collisions
- Mapping between the initial and final stage

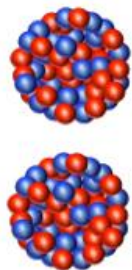


Observables of final stage have a strong relationship with the initial state

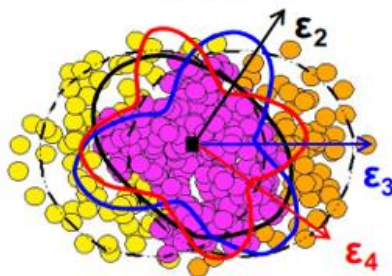
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Nuclear Structure



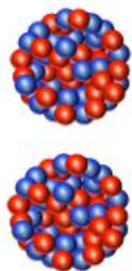
Initial State



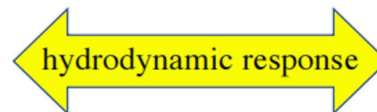
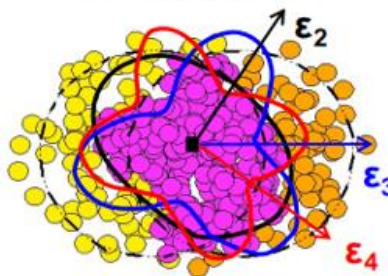
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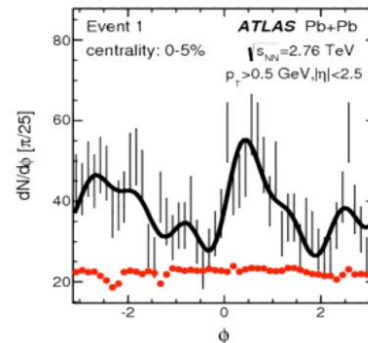
Nuclear Structure



Initial State



Produced Particle Flow



Radial Flow Anisotropic Flow

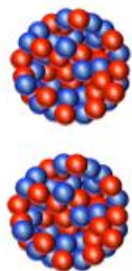
$$\frac{d^2N}{d\phi dp_T} = N(p_T) \left(\sum_n V_n e^{-in\phi} \right)$$

D. Teaney and L. Yan, Phys. Rev. C 86, 044908

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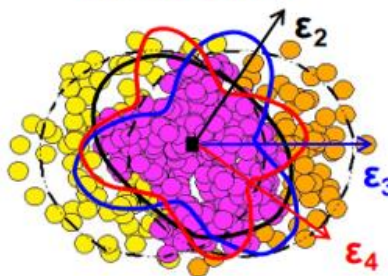
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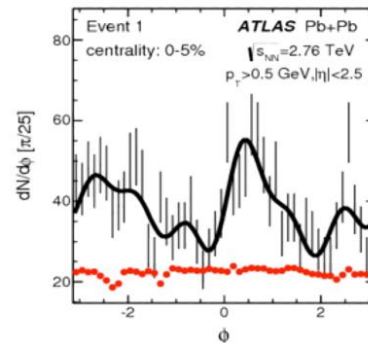
Imaging?

Initial State



hydrodynamic response

Produced Particle Flow



Radial Flow Anisotropic Flow

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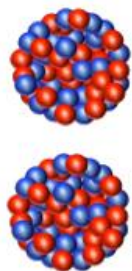
Approximate linear response
in each event:

$$\frac{\delta[p_T]}{[p_T]} \propto -\frac{\delta R_\perp}{R_\perp} \quad V_n \propto \mathcal{E}_n$$

Motivation:

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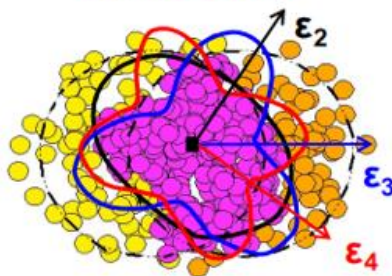
Nuclear Structure



Imaging?

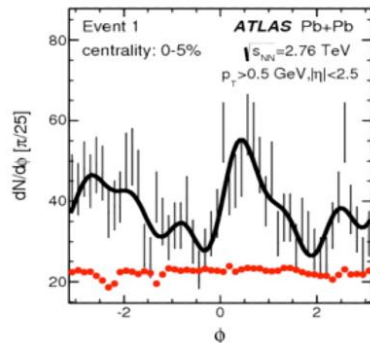
$$\rho(r, \theta, \phi) = \frac{\rho_0}{1 + e^{(r-R_0(1+\sum_n \beta_n Y_n^0(\theta, \phi)))/a_0)}$$

Initial State



hydrodynamic response

Produced Particle Flow



Radial Flow Anisotropic Flow

$$\frac{d^2 N}{d\phi dp_T} = N(p_T) \left(\sum_n V_n e^{-in\phi} \right)$$

D. Teaney and L. Yan, Phys. Rev. C 86, 044908

$\beta_2 \rightarrow$ Quadrupole deformation

$\beta_3 \rightarrow$ Octupole deformation

$a_0 \rightarrow$ Surface diffuseness

$R_0 \rightarrow$ Nuclear size

Initial Size

Initial Shape

$$R_\perp^2 \propto \langle r_\perp^2 \rangle$$

$$\mathcal{E}_n \propto \langle r_\perp^n e^{in\phi} \rangle$$

R_0

a_0

β_n

?

Approximate linear response
in each event:

$$\frac{\delta[p_T]}{[p_T]} \propto -\frac{\delta R_\perp}{R_\perp}$$

$$V_n \propto \mathcal{E}_n$$

Motivation: -Understanding Heavy-ion collisions
-Mapping between the initial and final stage

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Monte-Carlo Glauber as Estimator

Initial state

$E, \epsilon_2, \epsilon_3, d_{\perp}$

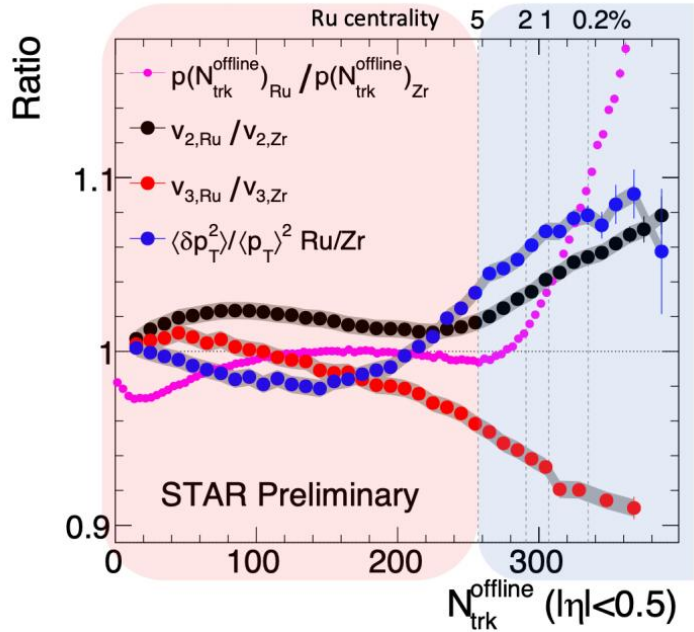


Final state

$N_{\text{ch}}, v_2\{2\}, v_3\{2\}, \langle p_T \rangle$

Motivation:

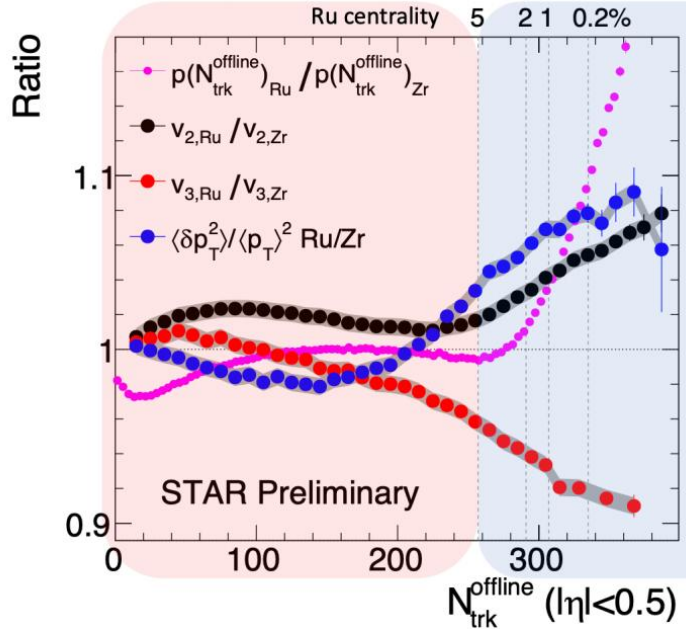
- Understanding Heavy-ion collisions
- Mapping between the initial and final stage



Plot from Chunjian Zhang's talk

Motivation:

- Understanding Heavy-ion collisions
- Mapping between the initial and final stage

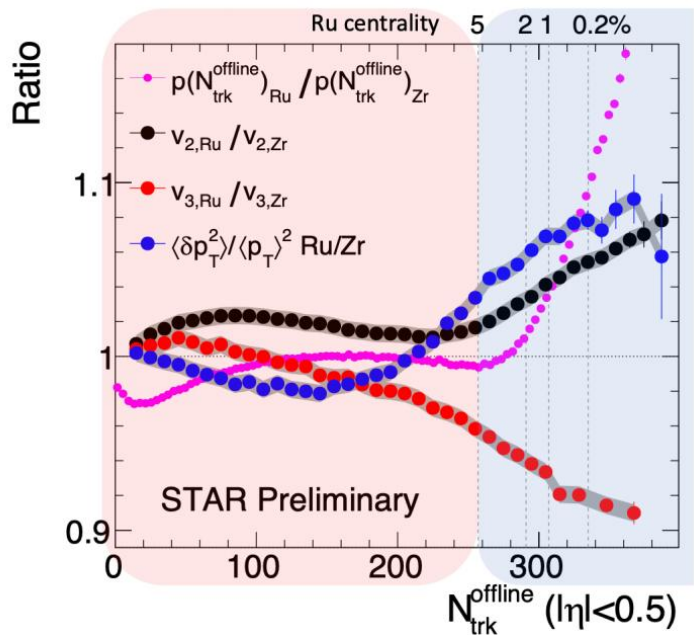


Plot from Chunjian Zhang's talk

Infer the **nuclear structure** from **final state** observables

- In single collision system ?
- Simultaneously for isobar systems with ratios ?

Motivation: -Understanding Heavy-ion collisions
-Mapping between the initial and final stage



Plot from Chunjian Zhang's talk

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Bayesian inference

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$$P(\theta|\mathcal{D}) \propto P(\mathcal{D}|\theta)P(\theta)$$

Bayesian inference

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nuclear structure

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nuclear structure

heavy-ion collision observables

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Bayesian inference

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nuclear structure

heavy-ion collision observables

$$P(\theta | \mathcal{D}) \propto P(\mathcal{D} | \theta) P(\theta)$$

$$P(\mathcal{D} | \theta) = \frac{1}{\sqrt{(2\pi)^m \det \Sigma}} \exp\left(-\frac{1}{2} [y_m(\theta) - y_e]^T \Sigma^{-1} [y_m(\theta) - y_e]\right)$$

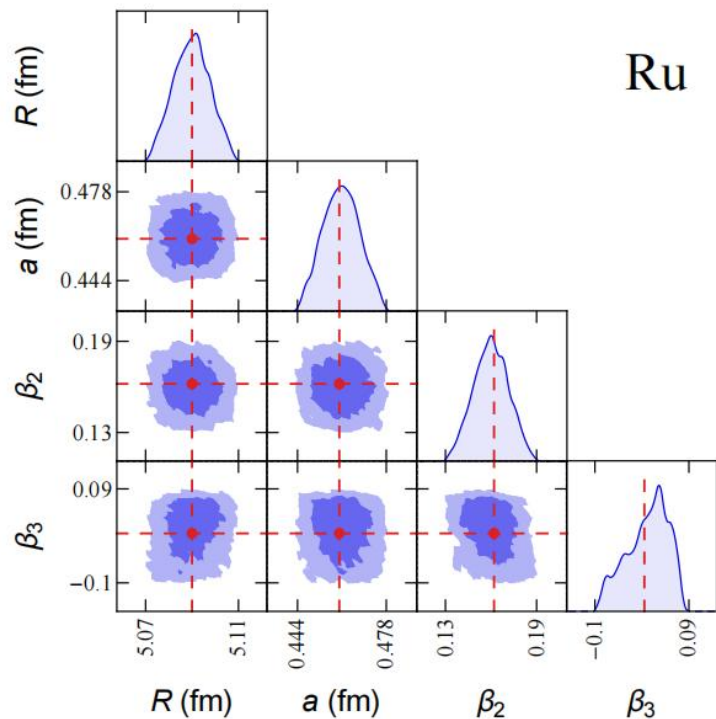
model

experiment

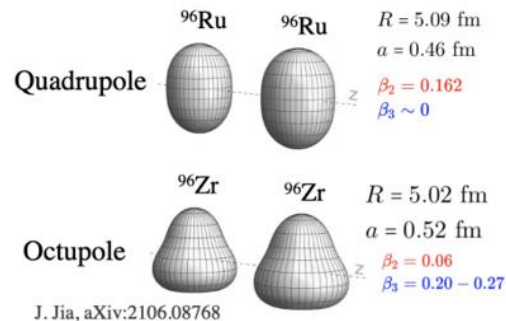
MCMC

Bayesian inference : Single system construction

It is possible to reconstruct the nuclear structure from the final state observables in heavy-ion collisions

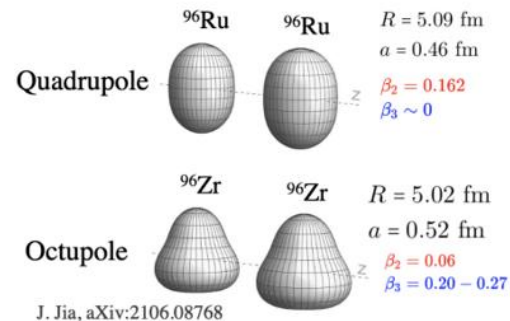
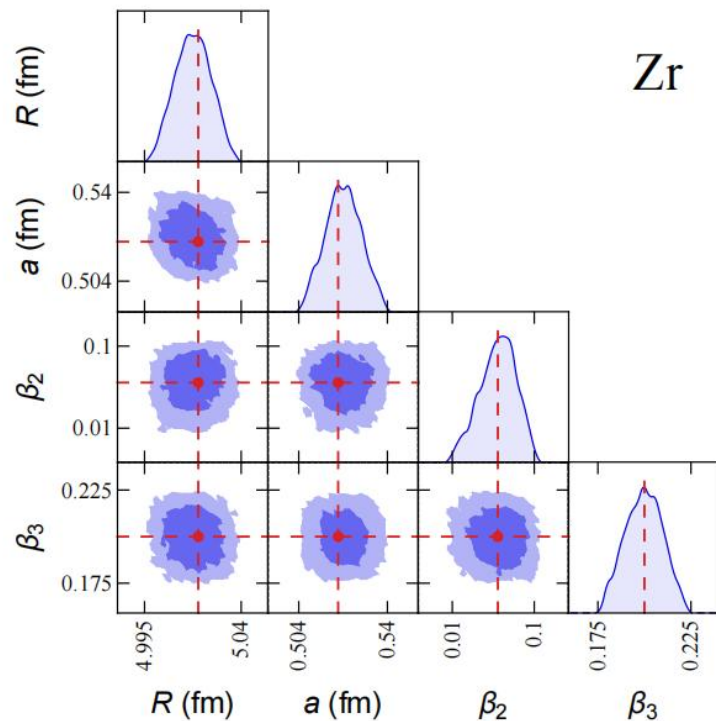


$$\mathbf{y}_{\text{Ru}} \equiv \{P_a^{\text{Ru}}, \varepsilon_{2,a}^{\text{Ru}}, \varepsilon_{3,a}^{\text{Ru}}, d_{\perp,a}^{\text{Ru}}\}_{a=1,\dots,40}$$



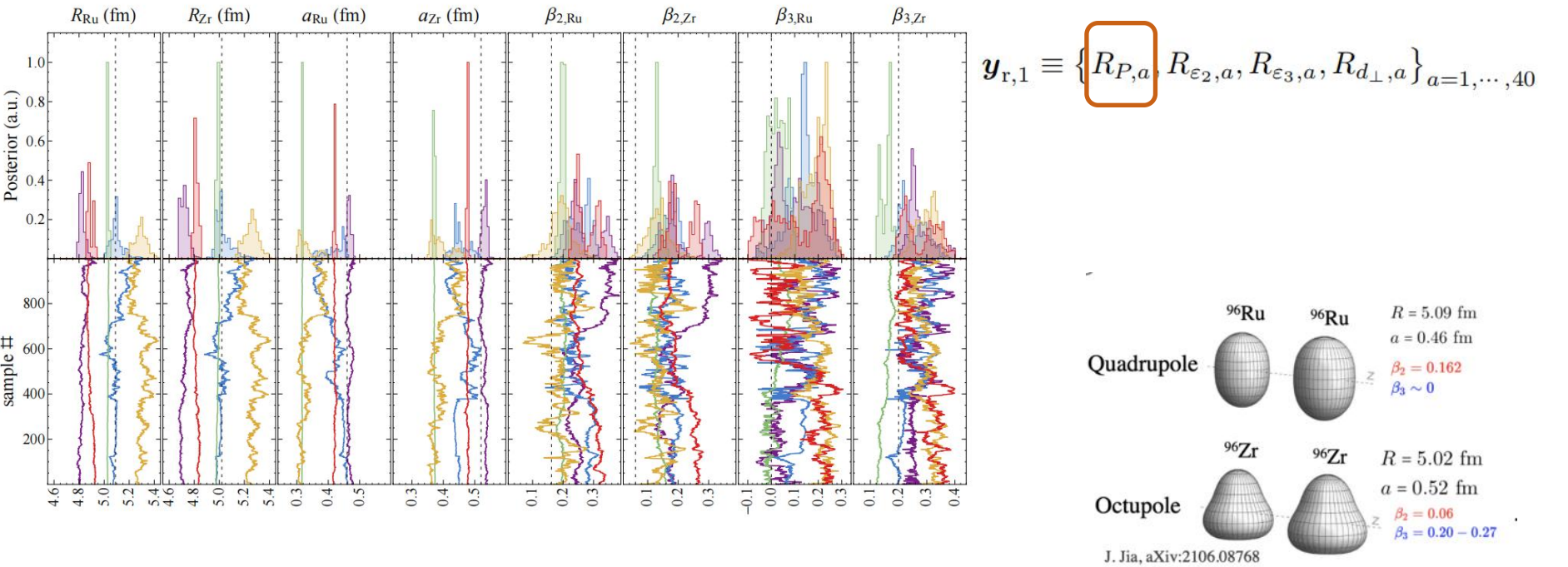
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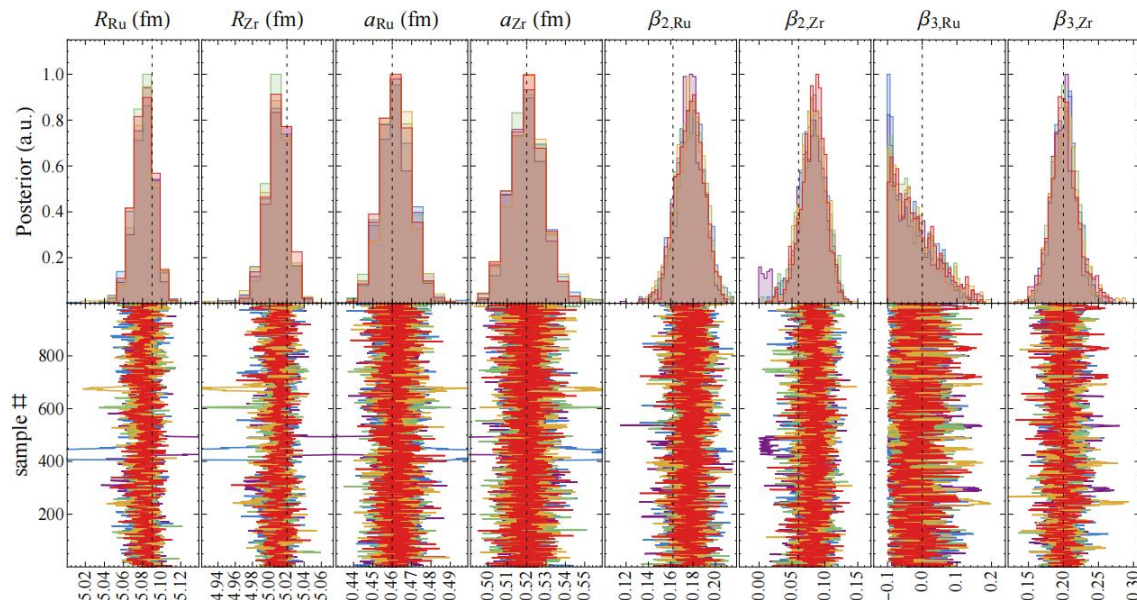
Bayesian inference : Simultaneous Reconstruction for Isobar Systems

Starting from purely the ratios, one can not simultaneously determine the nuclear structure s of the two isobar systems.

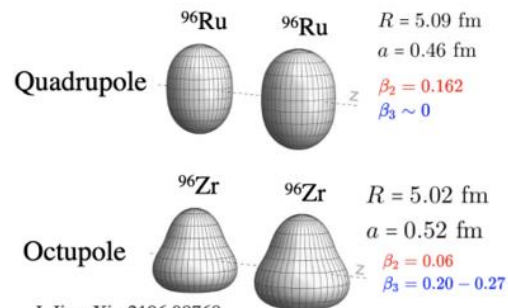


Bayesian inference : Simultaneous Reconstruction for Isobar Systems

Taking the multiplicity distributions of the two isobar systems together with the ratios of ϵ_2 , ϵ_3 , and $d \perp$, one can infer the isobar nuclear structures to very high precision.

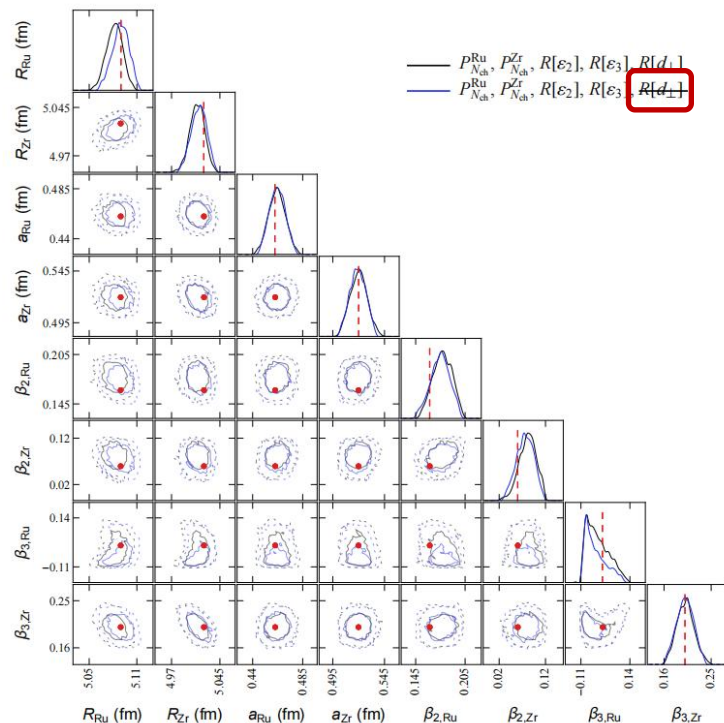


$$y_{r,2} \equiv \{P_a^{\text{Ru}}, P_a^{\text{Zr}}, R_{\epsilon_2,a}, R_{\epsilon_3,a}, R_{d \perp,a}\}_{a=1,\dots,40}$$

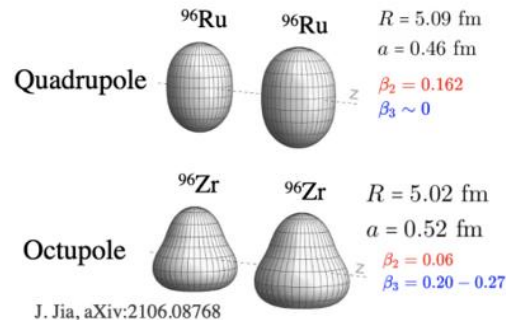


Bayesian inference : Simultaneous Reconstruction for Isobar Systems

Radial flow ($\langle p_T \rangle$), which can be estimated by $d_\perp$, carries redundant information as the ratios of elliptic/triangular flows.



$$\mathbf{y}_{r,3} \equiv \{P_a^{\text{Ru}}, P_a^{\text{Zr}}, R_{\epsilon_2,a}, R_{\epsilon_3,a}\}_{a=1,\dots,40}$$



Summary and outlook

- Infer nuclear structure from final observables : paves the way for precise predict non-CME background
- In single systems:
 - works well
- In isobar systems:
 - only from all of the ratios can not work
 - single-system multiplicity distributions are provided can work
 - ratio of radial flow is found to be nonessential
- Outlook: more realistic model needed; AMPT-based in progress.

Thank you !

$$\delta O_a^{\text{rel}} \equiv \sqrt{\frac{1}{d} \sum_{i=1}^d \left(\frac{O_{a,i}^{\text{pred}} - O_{a,i}^{\text{truth}}}{O_{a,i}^{\text{truth}}} \right)^2}.$$

Comparison of relative difference between the ground truth and predicted values using Gaussian Processor with **linear**(green), **quadratic**(blue), **4th-order**(red) and **RBF**(orange) kernels. As references, **gray** curves represent the differences due to the PCA truncation. Statistical errors over the mean value in the MC-Glauber modeling are also presented as black curves.

