

Bayesian Model/Data Analysis Big Experiment vs Big Theory Lessons from Relativistic Heavy Ion Collisions

Models and Data Analysis Initiative
<http://madai.us>



MICHIGAN STATE
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Duke
UNIVERSITY



THE UNIVERSITY
of NORTH CAROLINA
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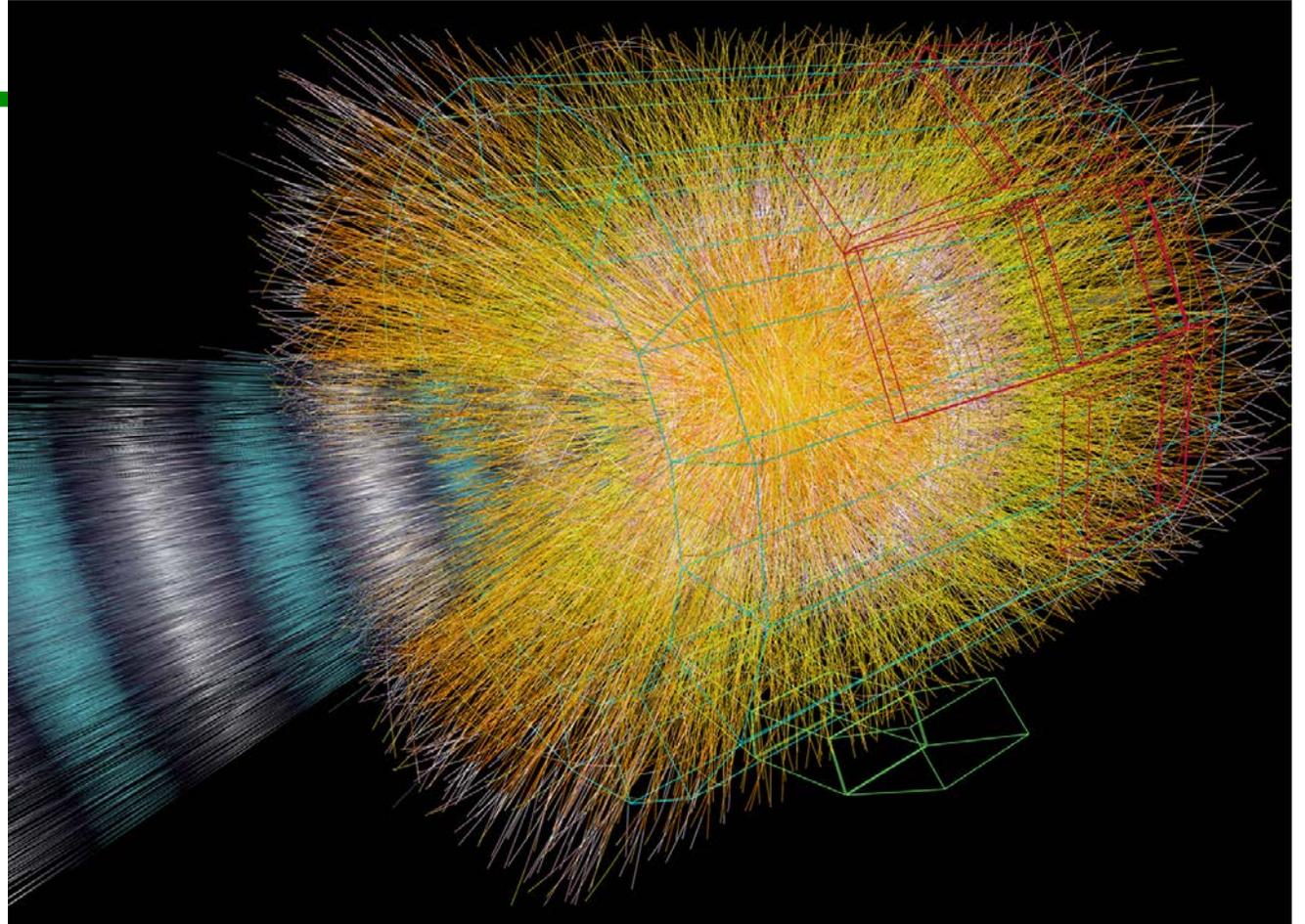
renci



1st MADAI Collaboration Meeting, SANDIA 2010

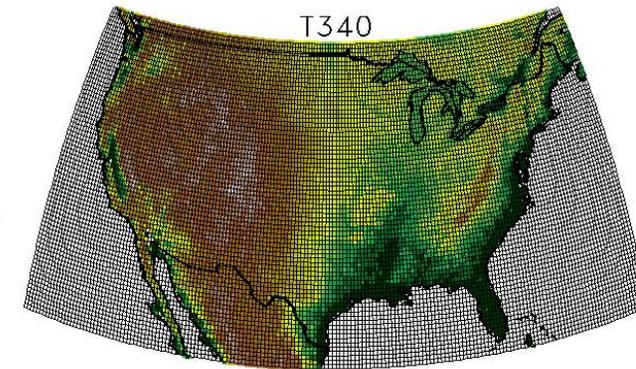
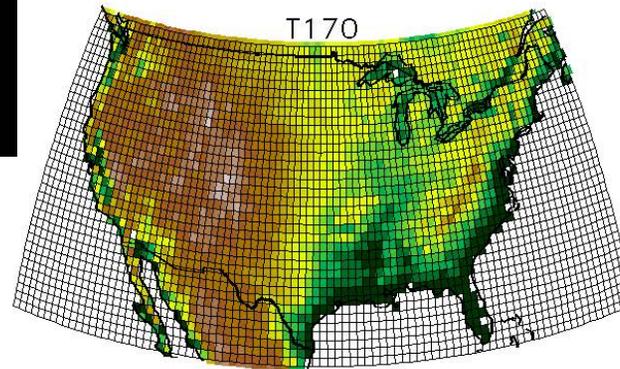
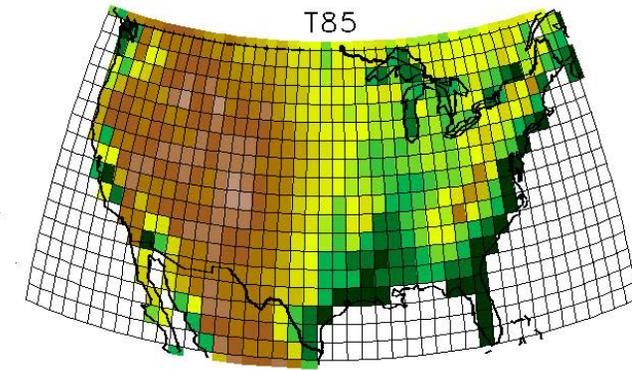
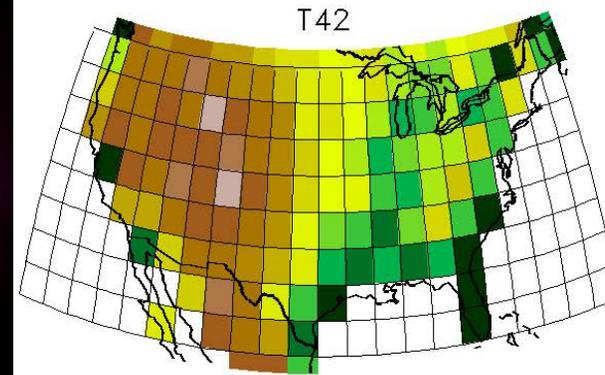
BIG EXPERIMENT

**Large,
Heterogenous
Data Sets**



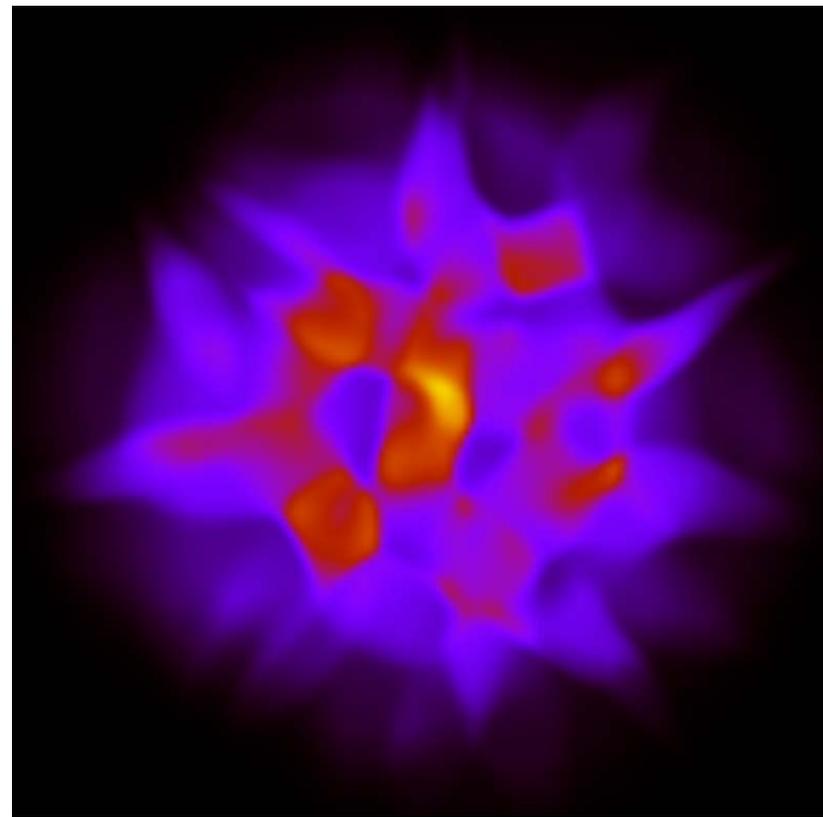
BIG MODELS

Numerous parameters
Assumptions



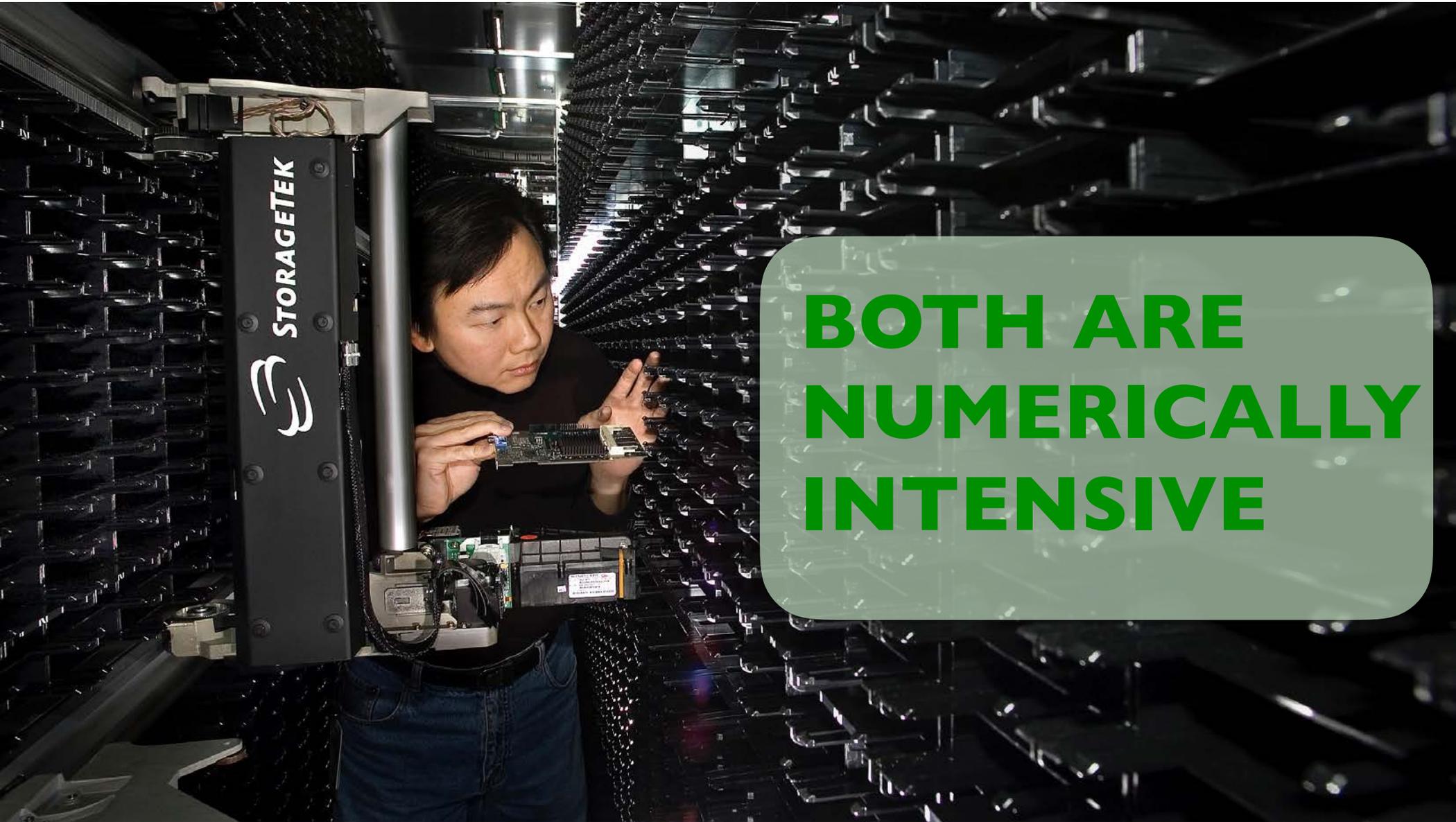
Atmospheric modeling

heavy ion collisions



GOALS

- **Determine parameters**
- **Test assumptions**
- **Identify connections between observables and parameters**



**BOTH ARE
NUMERICALLY
INTENSIVE**

BAYES theorem

$$P(A \& B) = P(A|B)P(B) = P(B|A)P(A)$$

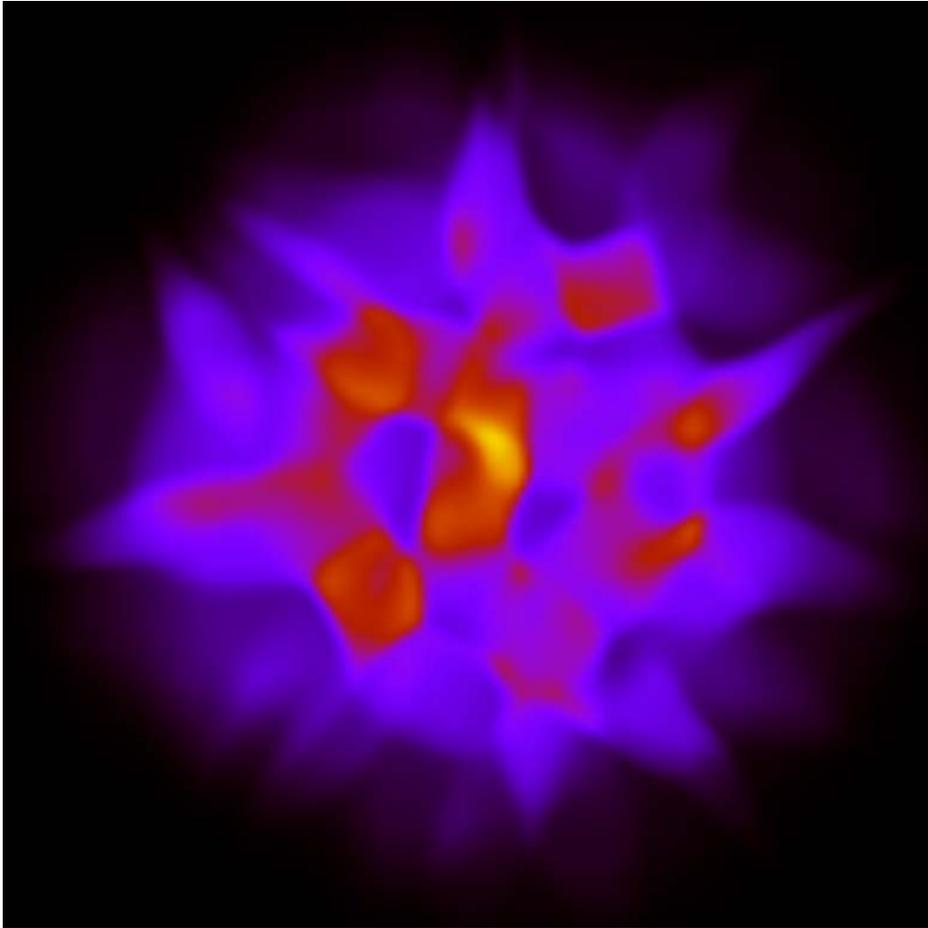
$$P(B|A) = \frac{P(A|B)P(B)}{P(A)}$$



P(A|B) is probability of data given theory(parameters)

P(B|A) is probability of theory(parameters) given data

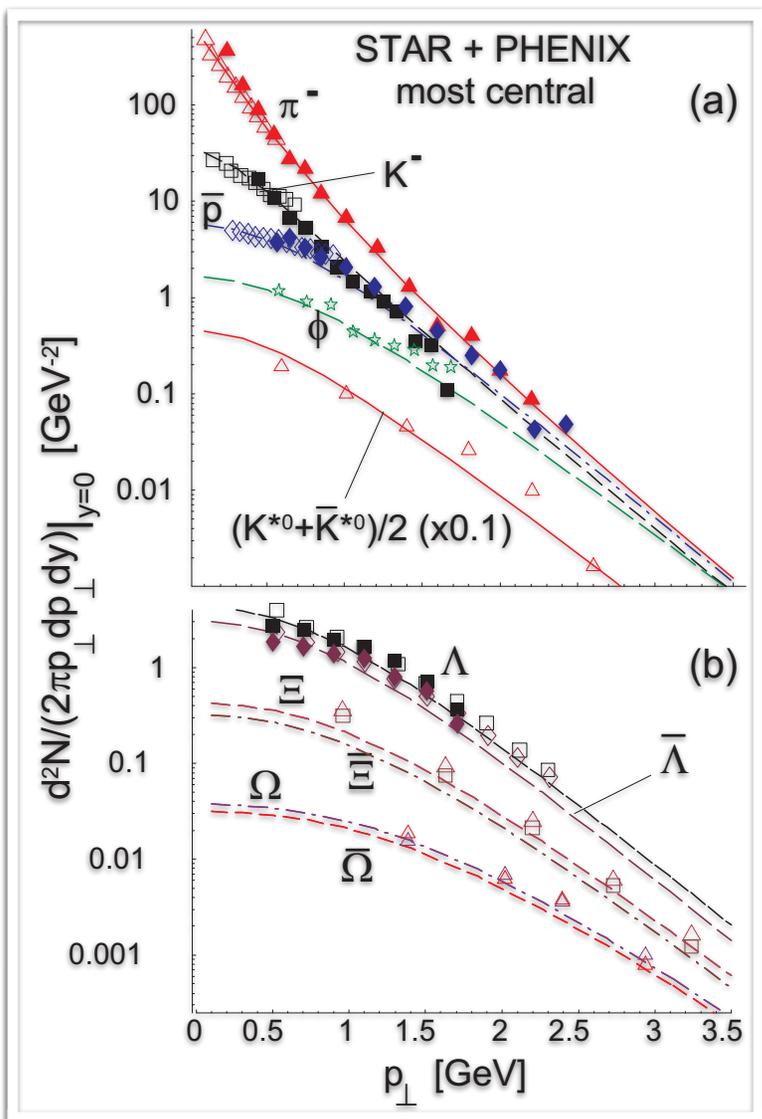
MODEL COMPONENTS



MODEL COMPONENTS

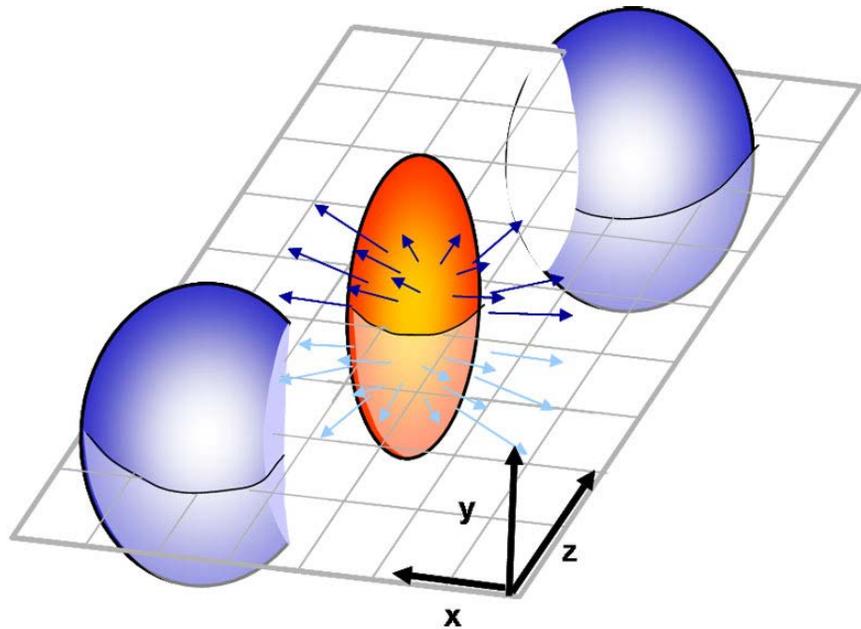
- ◆ **Thermalization**
(First fm/c)
- ◆ **Viscous Hydrodynamics**
(First ~5-10 fm/c)
- ◆ **Hadron Simulation**
(Dissolution & Breakup)
- ◆ Numerous parameters
(up to few dozen)
- ◆ ~Days of CPU to study one
point in parameter space

OBSERVABLES: Spectra

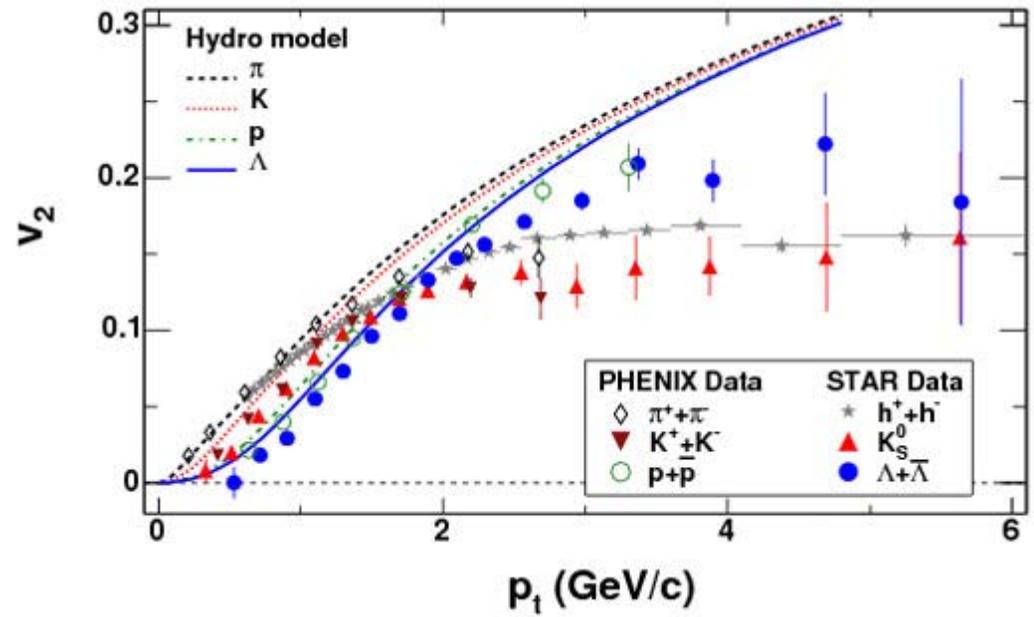


feature	infer
Yields	entropy
$\langle p_t \rangle$	T and flow
$\langle p_t \rangle$ of heavy particles	more flow than T

OBSERVABLES: v_2 (elliptic flow)



$$v_2 = \langle \cos 2\phi \rangle$$

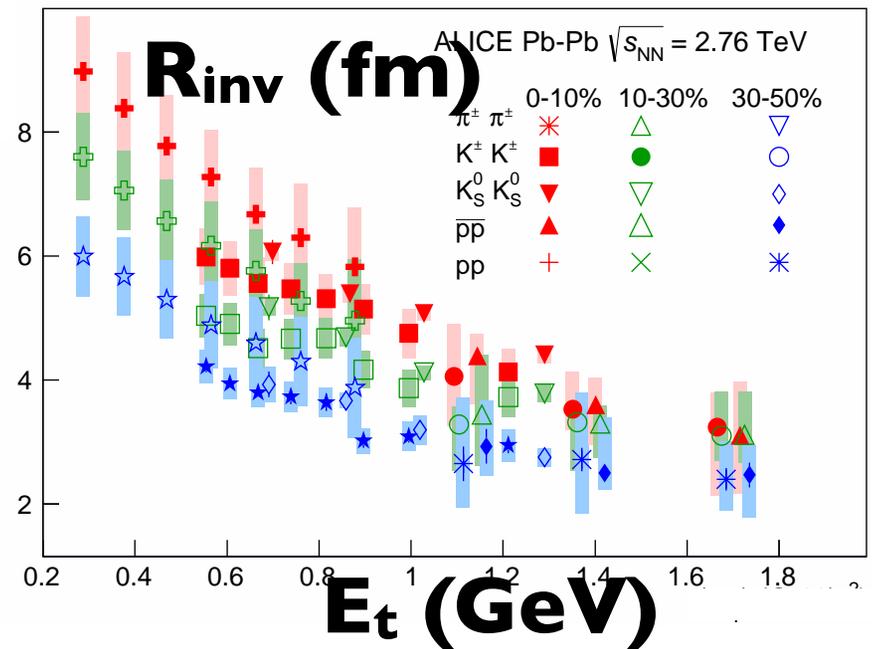
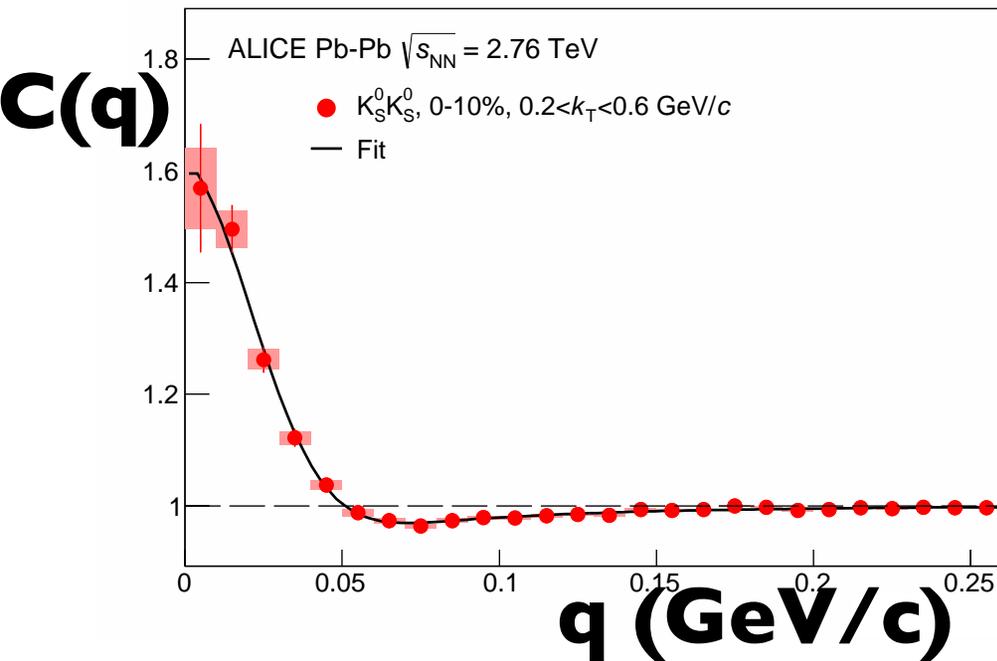


Sensitive to viscosity

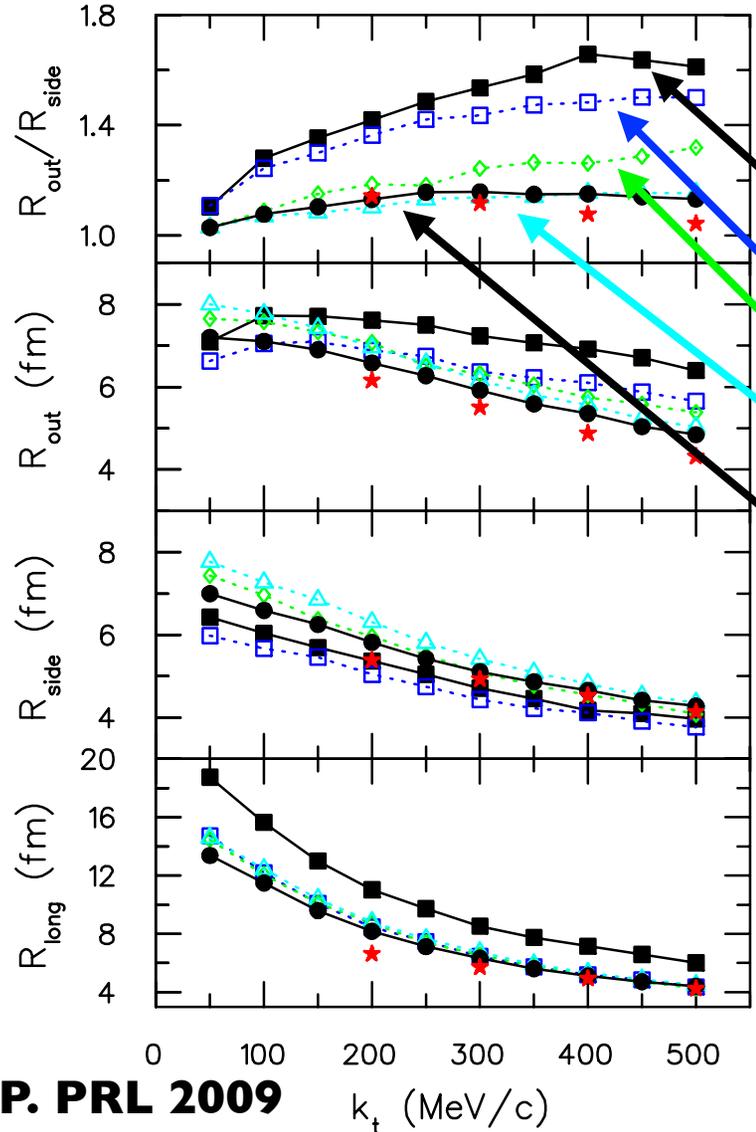
OBSERVABLES: Femtoscopic Correlations

$$C(P = p_1 + p_2, q = p_1 - p_2) = \frac{d^2 N / dp_1 dp_2}{(dN / dp_1)(dN / dp_2)}$$

$$C(P, q) \rightarrow S(P, r) = \frac{\int dr_1 dr_2 f(P/2, r_2) f(P/2, r_2) \delta(r - [r_1 - r_2])}{\int dr_1 dr_2 f(P/2, r_2) f(P/2, r_2)}$$



OBSERVABLES: Femtoscopic



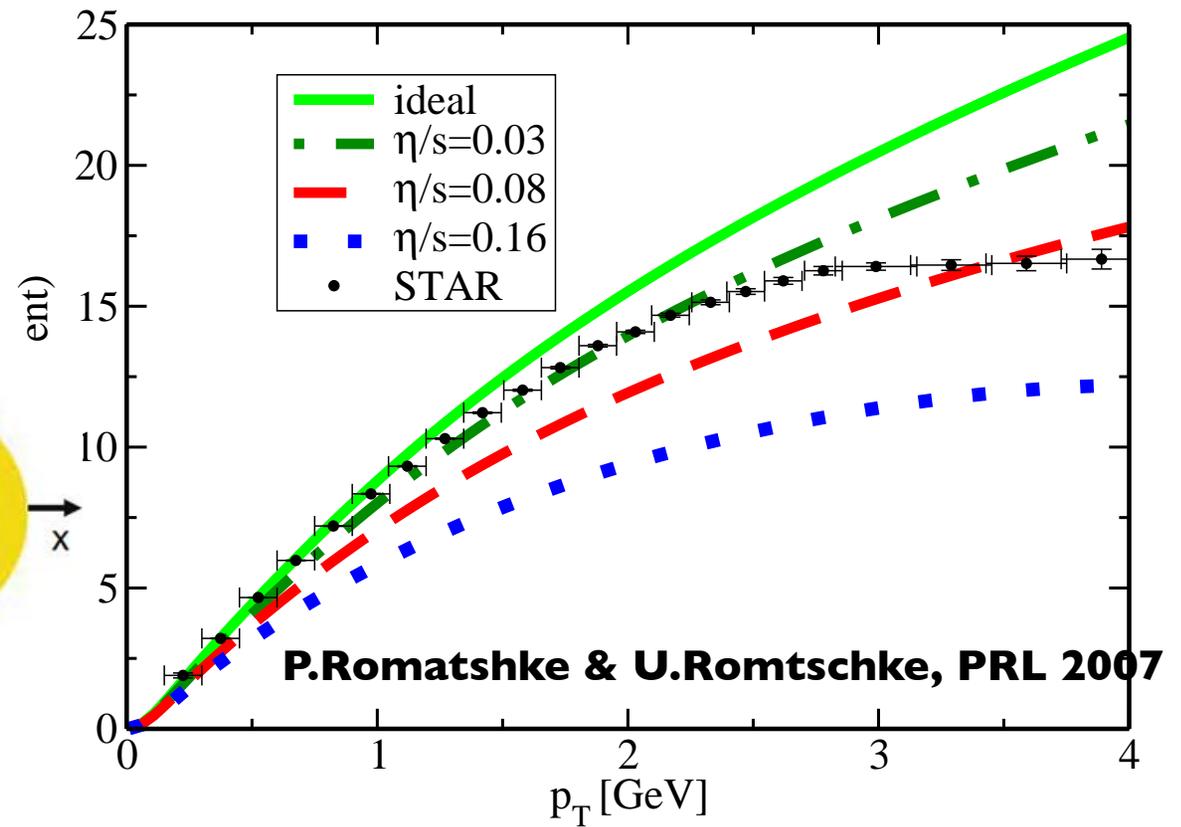
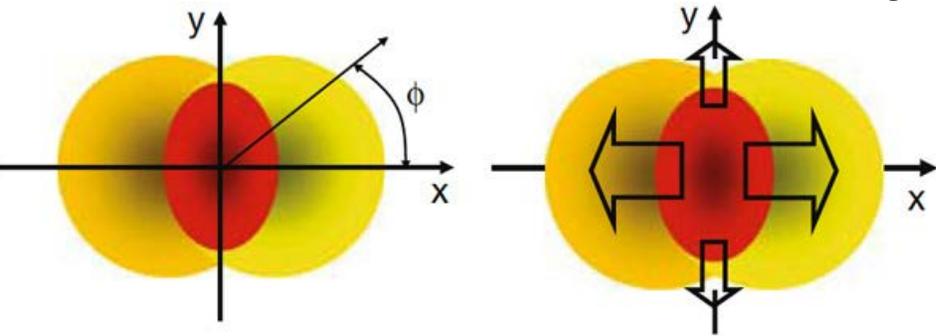
- 1st generation hydro model
- + initial acceleration
- + realistic Eq. of State
- + viscosity
- + improved wave function



S.P. PRL 2009

How this was done before (v_2 and η/s)

$$v_2 \equiv \langle \cos 2\phi \rangle$$



PROBLEM

v2 depends on

- **viscosity**
- **saturation model**
- **pre-thermal flow**
- **Eq. of State**
- **T-dependence of η/s**
- **initial T_{xx}/T_{zz}**
- **. . . .**

e.g. Drescher, Dumitru, Gombeaud and Ollitrault
PRC 2007

Correct Way (MCMC)

- ◆ Simultaneously vary N model parameters \mathbf{x}_i
- ◆ Perform random walk weight by likelihood

$$\mathcal{L}(\mathbf{x}|\mathbf{y}) \sim \exp \left\{ - \sum_a \frac{(y_a^{(\text{model})}(\mathbf{x}) - y_a^{(\text{exp})})^2}{2\sigma_a^2} \right\}$$

- ◆ Use all observables \mathbf{y}_a
- ◆ Obtain representative sample of posterior

MCMC Metropolis algorithm

Imagine $N \rightarrow \infty$ instances (samplings) of parameters x with probability $P(x)$

Consider two point in parameter space x_i and x_j

Rates of $i \rightarrow j$ and $j \rightarrow i$ are

$$R(i \rightarrow j) = P(i)R(i \rightarrow j|i)$$

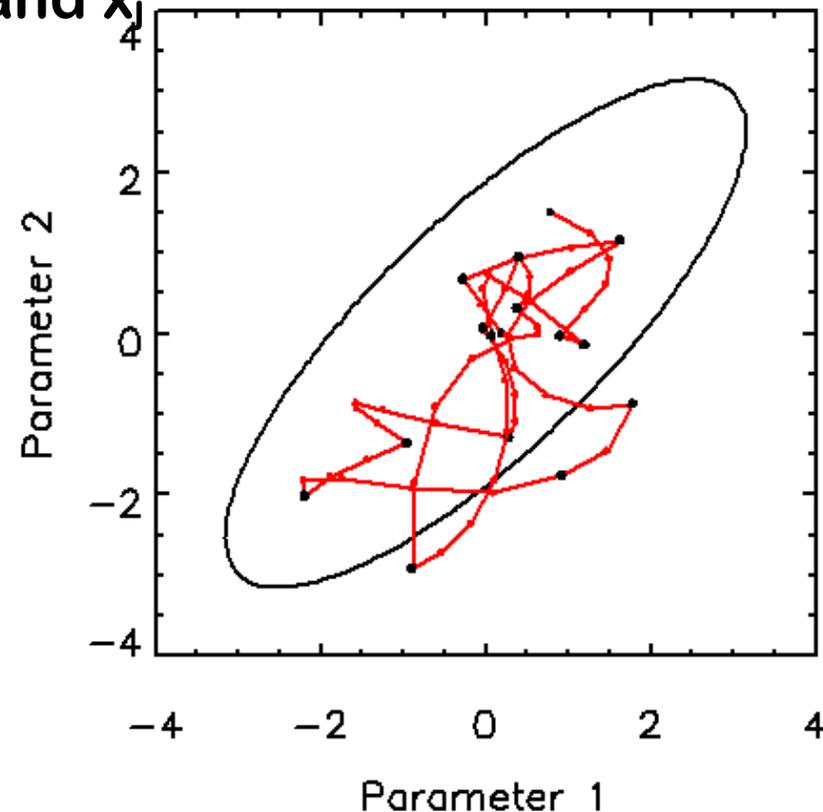
$$R(j \rightarrow i) = P(j)R(j \rightarrow i|j)$$

$$R(j \rightarrow i|j) = R(i \rightarrow j|i) \frac{P(i)}{P(j)}$$

If in state “i” do random step

if $P(j) > P(i)$, keep 100%

if $P(i) > P(j)$, keep with prob. $P(j)/P(i)$



Difficult Because...

I. Too Many Model Runs

Requires running model $\sim 10^6$ times

II. Many Observables

Could be hundreds of plots,
each with dozens of points

Complicated Error Matrices

Model Emulators

1. **Run the model ~1000 times**
Semi-random points (LHS sampling)

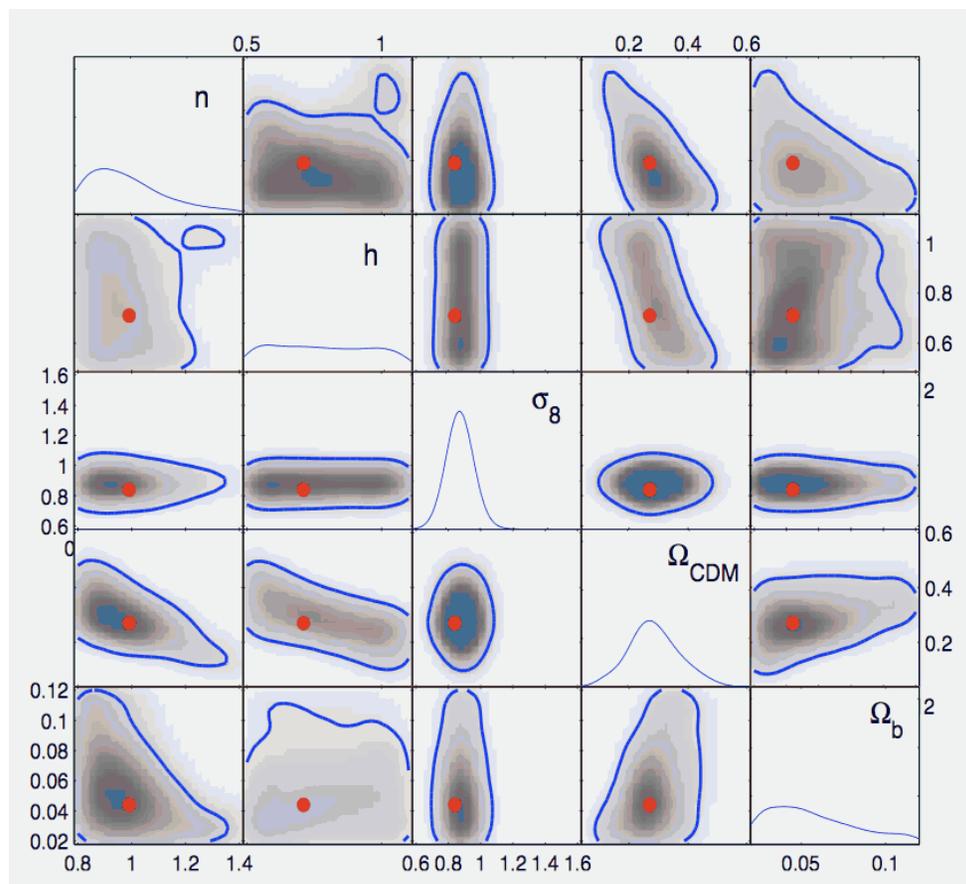
2. **Determine Principal Components**

$$(y_a - \langle y_a \rangle) / \sigma_a \rightarrow z_a$$

3. **Emulate z_a (Interpolate) for MCMC**

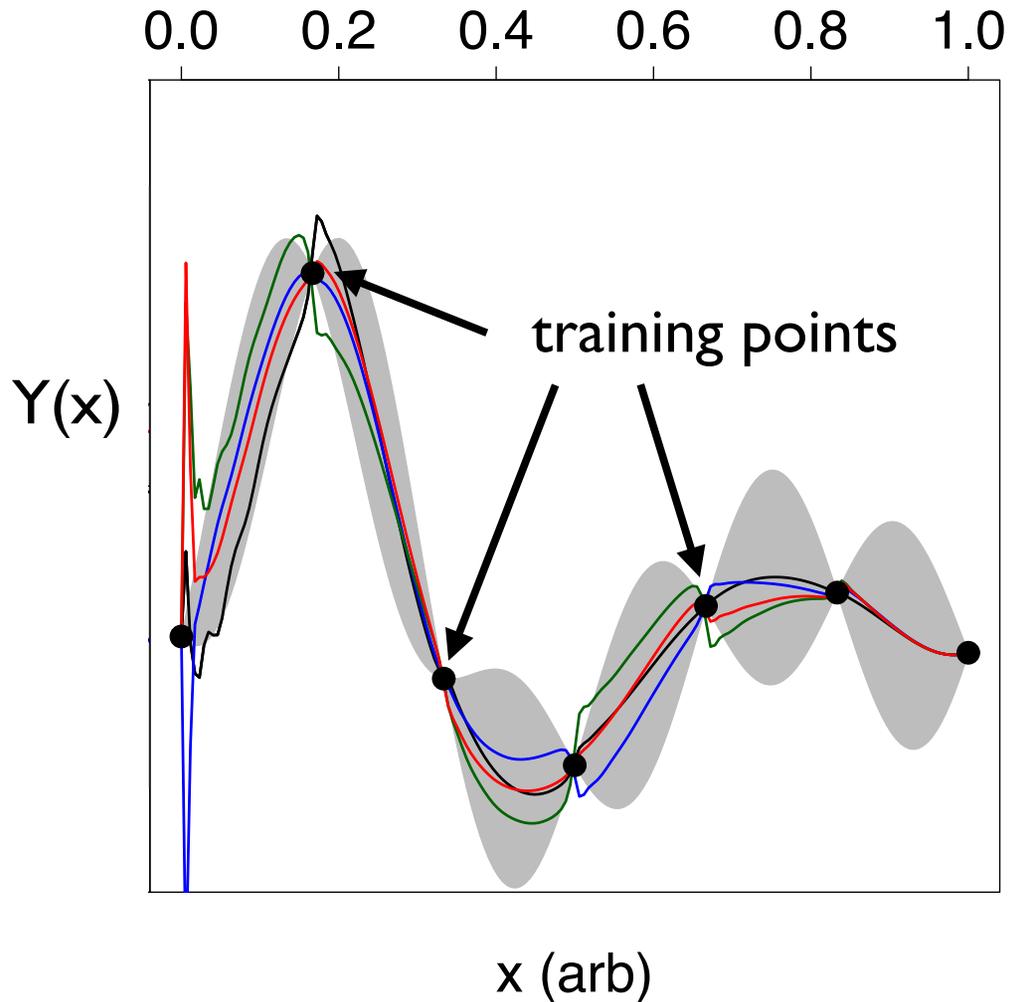
Gaussian Process...

$$\mathcal{L}(\mathbf{x}|\mathbf{y}) \sim \exp \left\{ -\frac{1}{2} \sum_a (z_a^{(\text{emulator})}(\mathbf{x}) - z_a^{(\text{exp})})^2 \right\}$$



S. Habib, K. Heitman, D. Higdon, C. Nakhleh & B. Williams, PRD(2007)

Emulator



- ◆ **Gaussian Process**
 - Reproduces training points
 - Assumes localized Gaussian covariance
 - Must be trained, i.e. find “hyper parameters”
- ◆ **Other methods also work**

14 Parameters

- ◆ 5 for Initial Conditions at RHIC
- ◆ 5 for Initial Conditions at LHC
- ◆ 2 for Viscosity
- ◆ 2 for Eq. of State

30 Observables

- ◆ π, K, p Spectra
 $\langle p_t \rangle$, Yields
- ◆ Interferometric Source Sizes
- ◆ v_2 Weighted by p_t

Initial State Parameters

$$\epsilon(\tau = 0.8\text{fm}/c) = f_{\text{wn}}\epsilon_{\text{wn}} + (1 - f_{\text{wn}})\epsilon_{\text{cgc}},$$

$$\epsilon_{\text{wn}} = \epsilon_0 T_A \frac{\sigma_{\text{nn}}}{2\sigma_{\text{sat}}} \{1 - \exp(-\sigma_{\text{sat}} T_B)\} + (A \leftrightarrow B)$$

$$\epsilon_{\text{cgc}} = \epsilon_0 T_{\text{min}} \frac{\sigma_{\text{nn}}}{\sigma_{\text{sat}}} \{1 - \exp(-\sigma_{\text{sat}} T_{\text{max}})\}$$

$$T_{\text{min}} \equiv \frac{T_A T_B}{T_A + T_B},$$

$$T_{\text{max}} \equiv T_A + T_B,$$

$$u_{\perp} = \alpha\tau \frac{\partial T_{00}}{2T_{00}}$$

$$T_{zz} = \gamma P$$

5 parameters for RHIC, 5 for LHC

Equation of State and Viscosity

$$c_s^2(\epsilon) = c_s^2(\epsilon_h) + \left(\frac{1}{3} - c_s^2(\epsilon_h) \right) \frac{X_0 x + x^2}{X_0 x + x^2 + X'^2},$$

$$X_0 = X' R c_s(\epsilon) \sqrt{12},$$

$$x \equiv \ln \epsilon / \epsilon_h$$

$$\frac{\eta}{s} = \left. \frac{\eta}{s} \right|_{T=165} + \kappa \ln(T/165)$$

2 parameters for EoS, 2 for η/s

DATA Distillation



1. **Experiments reduce PBs to 100s of plots**
2. **Choose which data to analyze**
Does physics factorize?
3. **Reduce plots to a few representative numbers, y_a**
4. **Transform to principal components, z_a**
$$\mathcal{L} \sim \exp \left\{ \frac{-1}{2} \sum_a (z_a - z_a^{(\text{exp})})^2 \right\}$$
5. **Resolving power of RHIC/LHC data reduced to ≈ 10 numbers!**

Principal Component Analysis (PCA)

Many observables y_a

All change as function of parameters x_i

BUT, some linear combinations change — some don't

$$\delta\tilde{y}_a = \frac{(y_a - \langle y_a \rangle)}{\sigma_a} \text{ average over } x$$

$$M_{ab} = \langle \delta\tilde{y}_a \delta\tilde{y}_b \rangle$$

Diagonalize M

$$M \rightarrow \begin{pmatrix} \lambda_1 & 0 & 0 & 0 \\ 0 & \lambda_2 & 0 & 0 \\ 0 & 0 & \lambda_3 & 0 \\ 0 & 0 & 0 & \lambda_4 \end{pmatrix}$$

Principal Component Analysis (PCA)

$$\delta\tilde{y}_a = \frac{(y_a - \langle y_a \rangle)}{\sigma_a}$$
$$M_{ab} = \langle \delta\tilde{y}_a \delta\tilde{y}_b \rangle$$

Diagonalize M

$$M \rightarrow \begin{pmatrix} \lambda_1 & 0 & 0 & 0 \\ 0 & \lambda_2 & 0 & 0 \\ 0 & 0 & \lambda_3 & 0 \\ 0 & 0 & 0 & \lambda_4 \end{pmatrix}$$

In new basis, $y_a \leftrightarrow z_a$

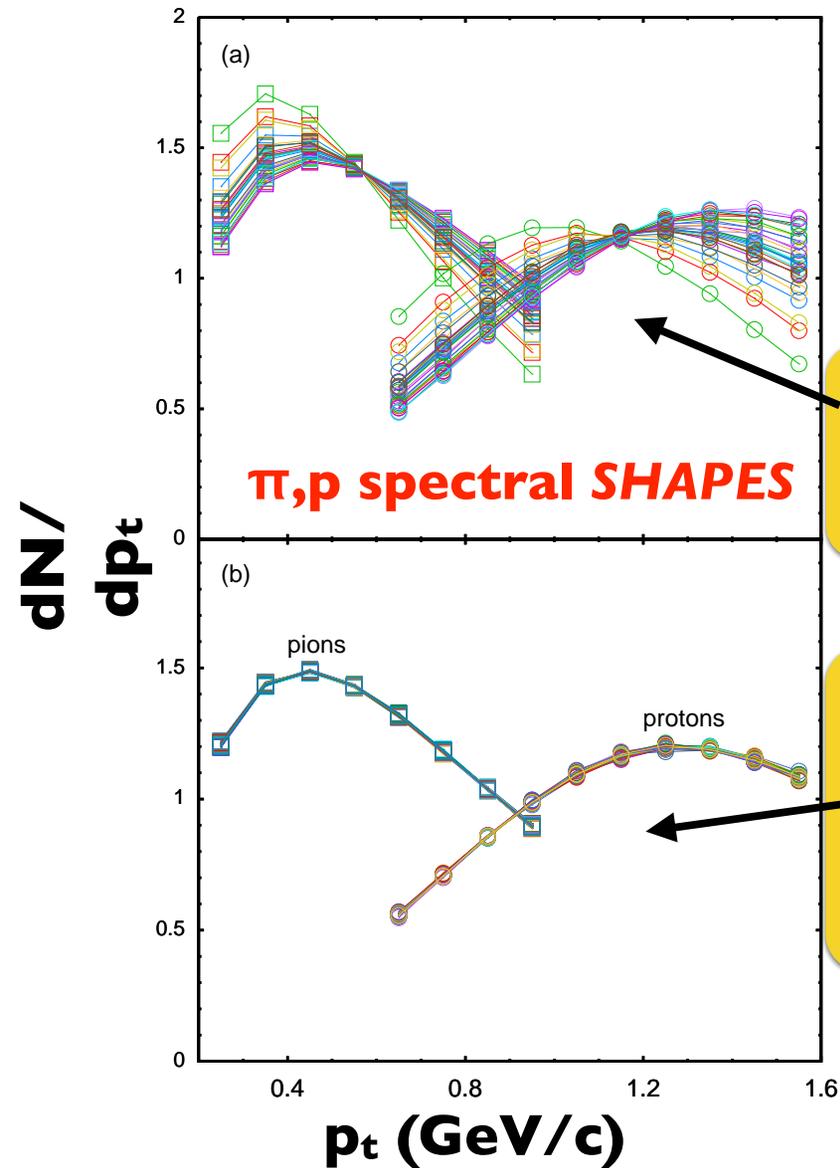
z_a are known as principal components

If $\lambda_a \gg 1$, good resolving power

If $\lambda_a \ll 1$, little resolving power, no need to analyze

Checking the Distillation

Spectral information
encapsulated by two numbers,
 dN/dy & $\langle p_t \rangle$



model spectra from
30 random points in
parameter prior

74 pion spectra:
with $573 < \langle p_t \rangle_\pi < 575$ MeV

44 proton spectra:
with $1150 < \langle p_t \rangle_p < 1152$ MeV

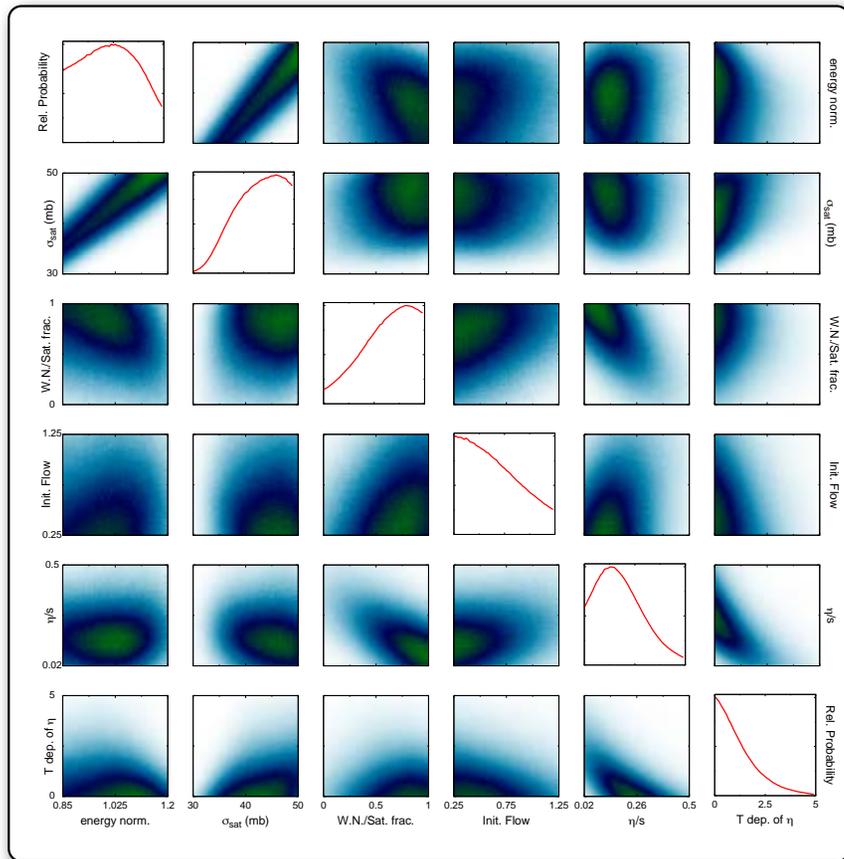
Review the Grand Plan

- 1. Choose observables**
- 2. Distill Data**
- 3. Parameterize model**
- 4. Run full model hundreds of times**
(Latin hyper-cube sampling)
- 5. Build & Tune emulator**
- 6. Perform MCMC with emulator**
- 7. Analyze sensitivities**

Two Calculations

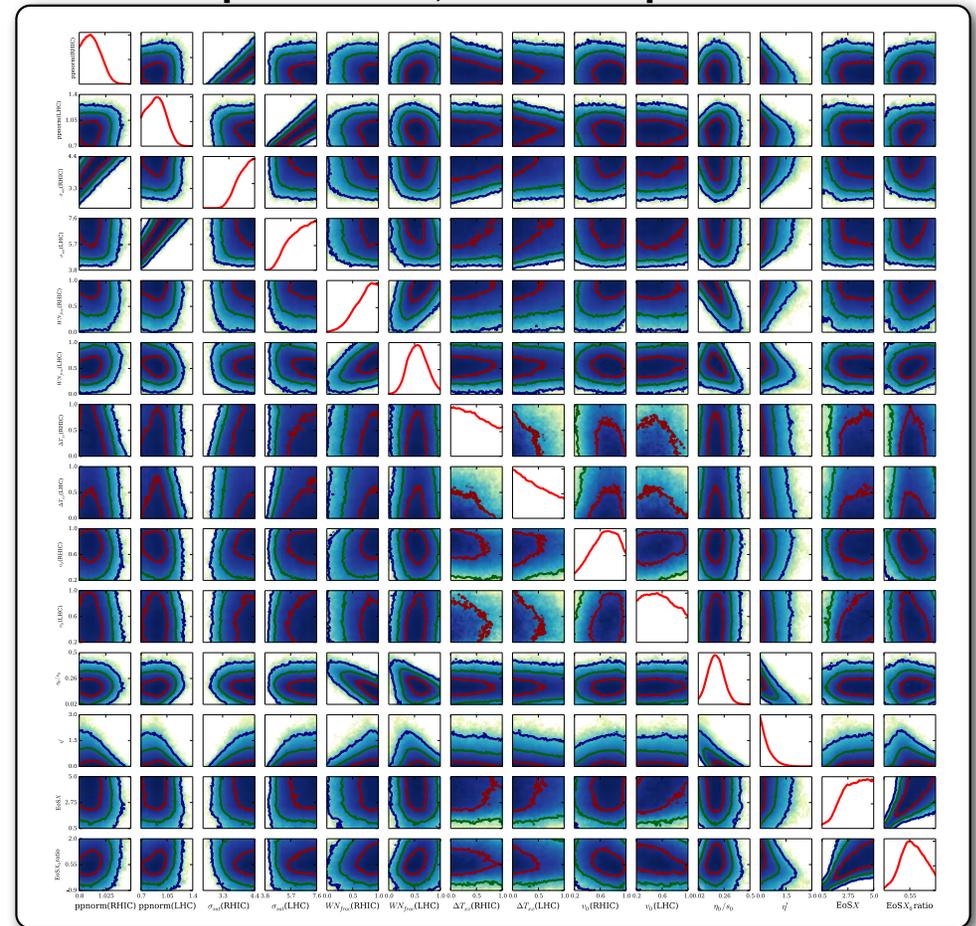
J.Novak, K. Novak, S.P., C.Coleman-Smith & R.Wolpert, PRC 2014

RHIC Au+Au Data
6 parameters

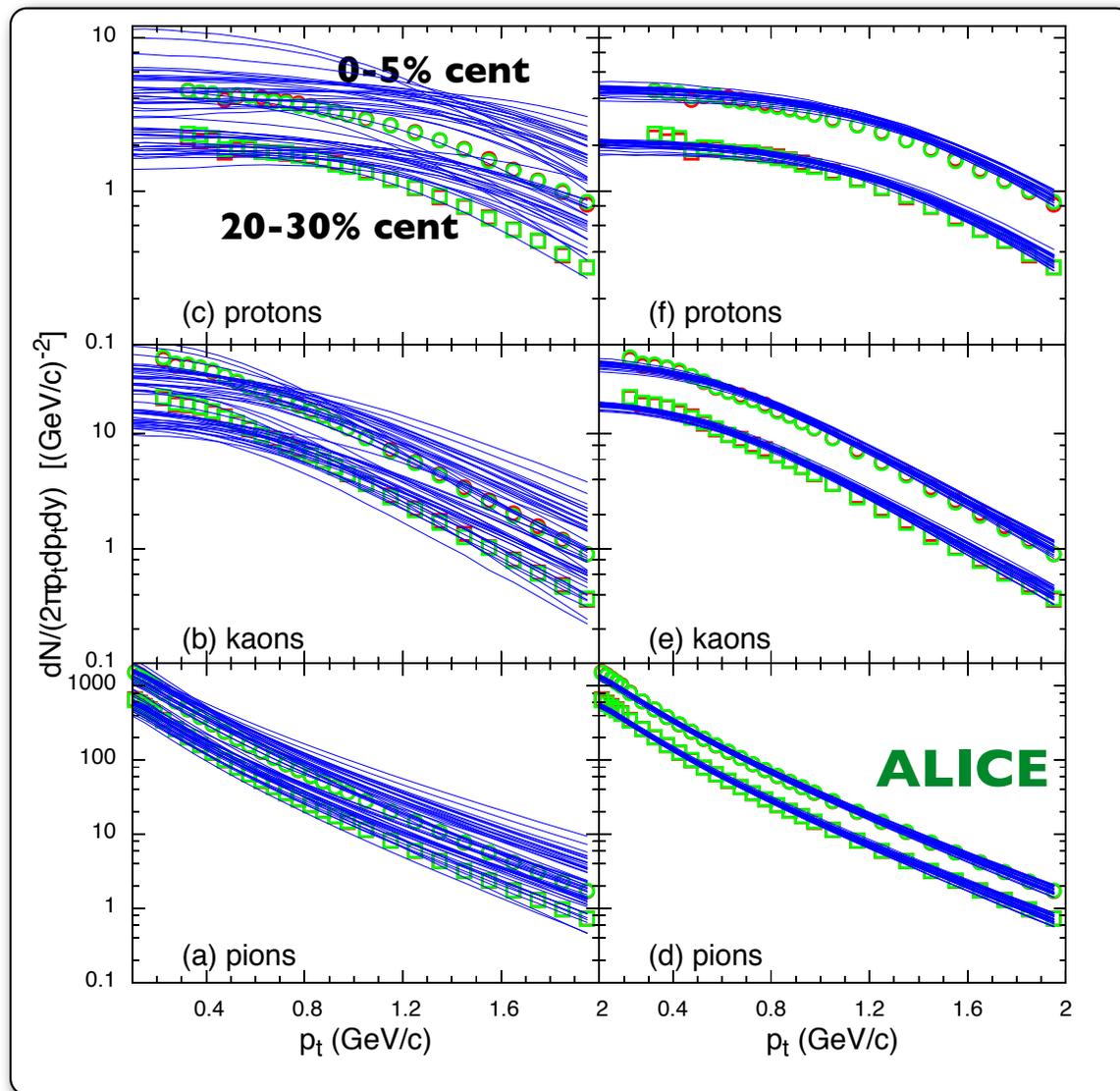


S.P., E.Sangaline, P.Sorensen & H.Wang, PRL 2015

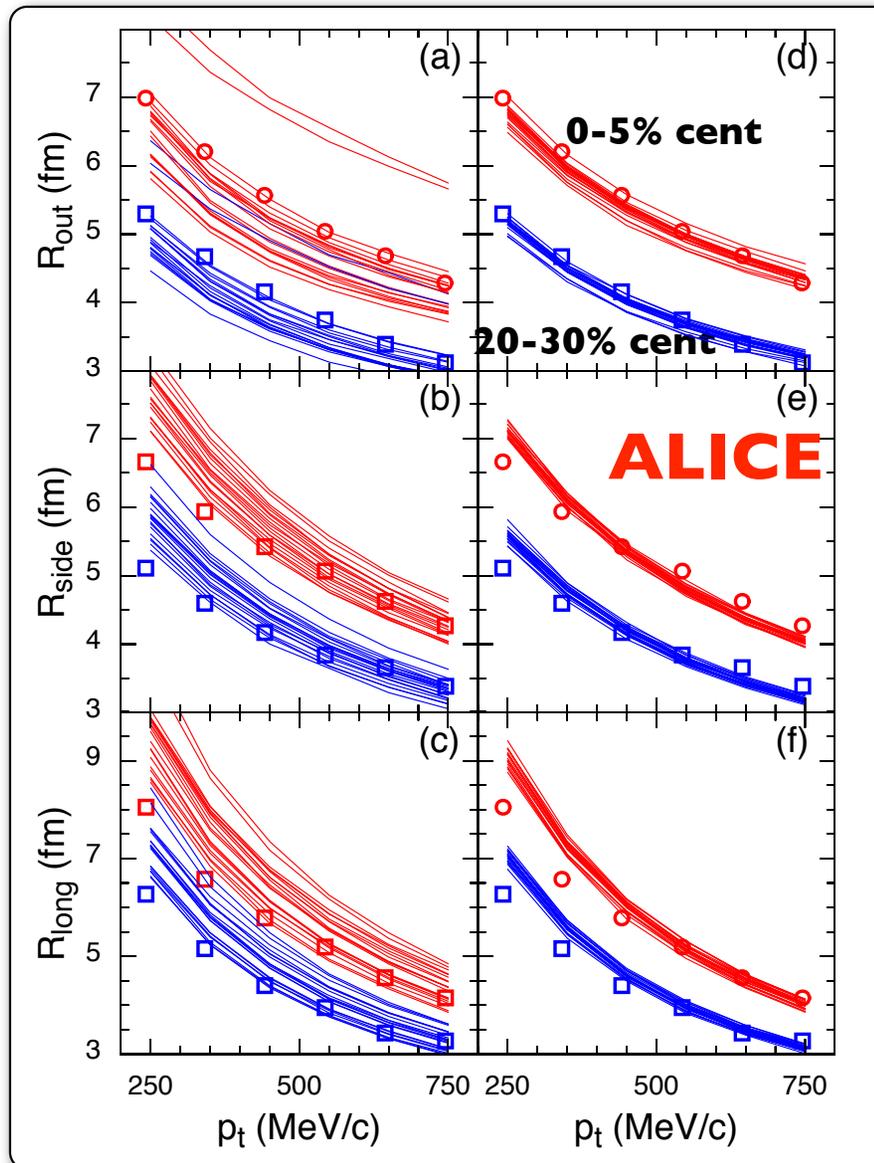
RHIC Au+Au and LHC Pb+Pb Data
14 parameters, include Eq. of State



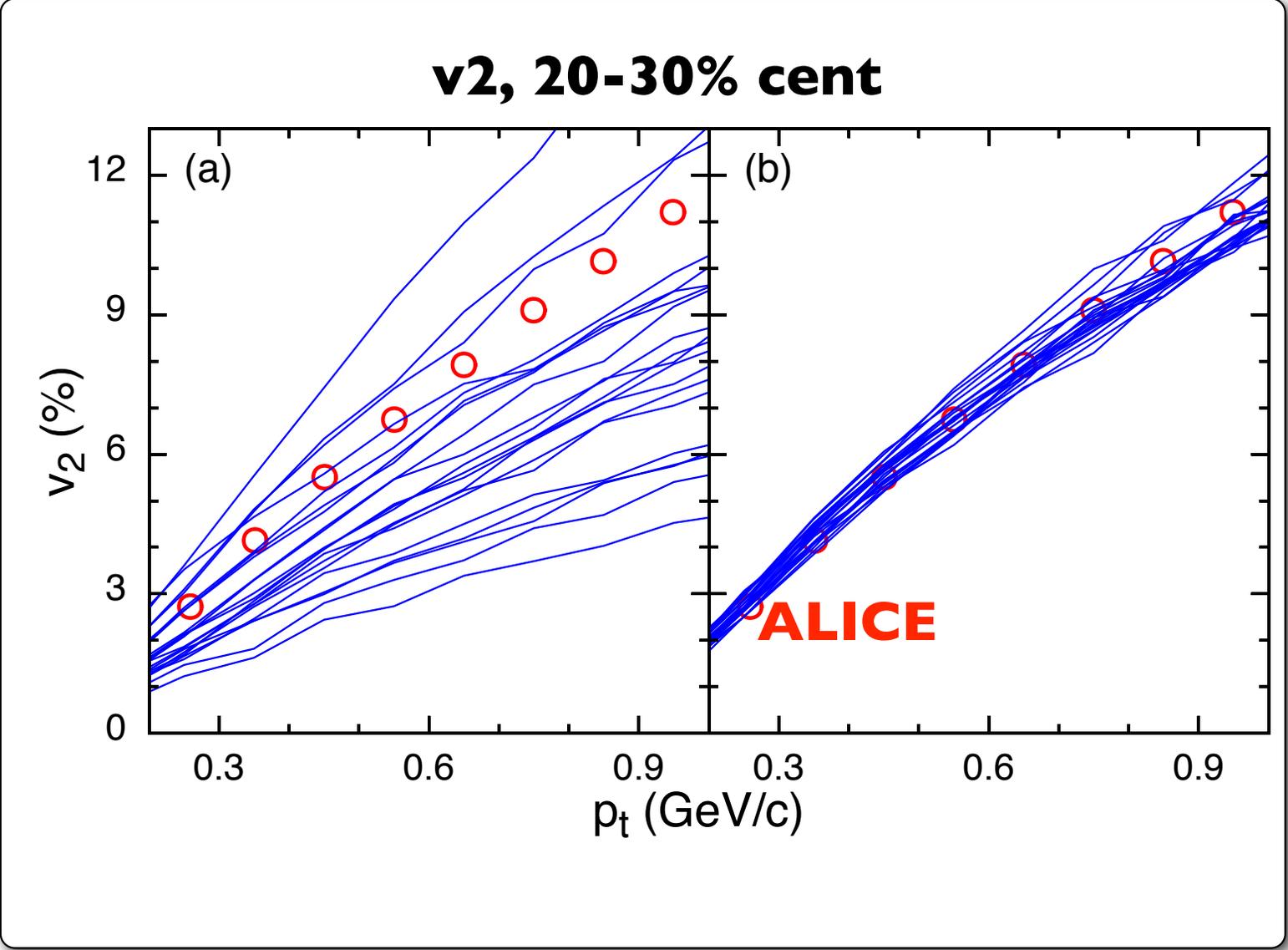
Sample Spectra from Prior and Posterior



**Sample
HBT from
Prior and
Posterior**



**Sample V_2
from Prior
and
Posterior**

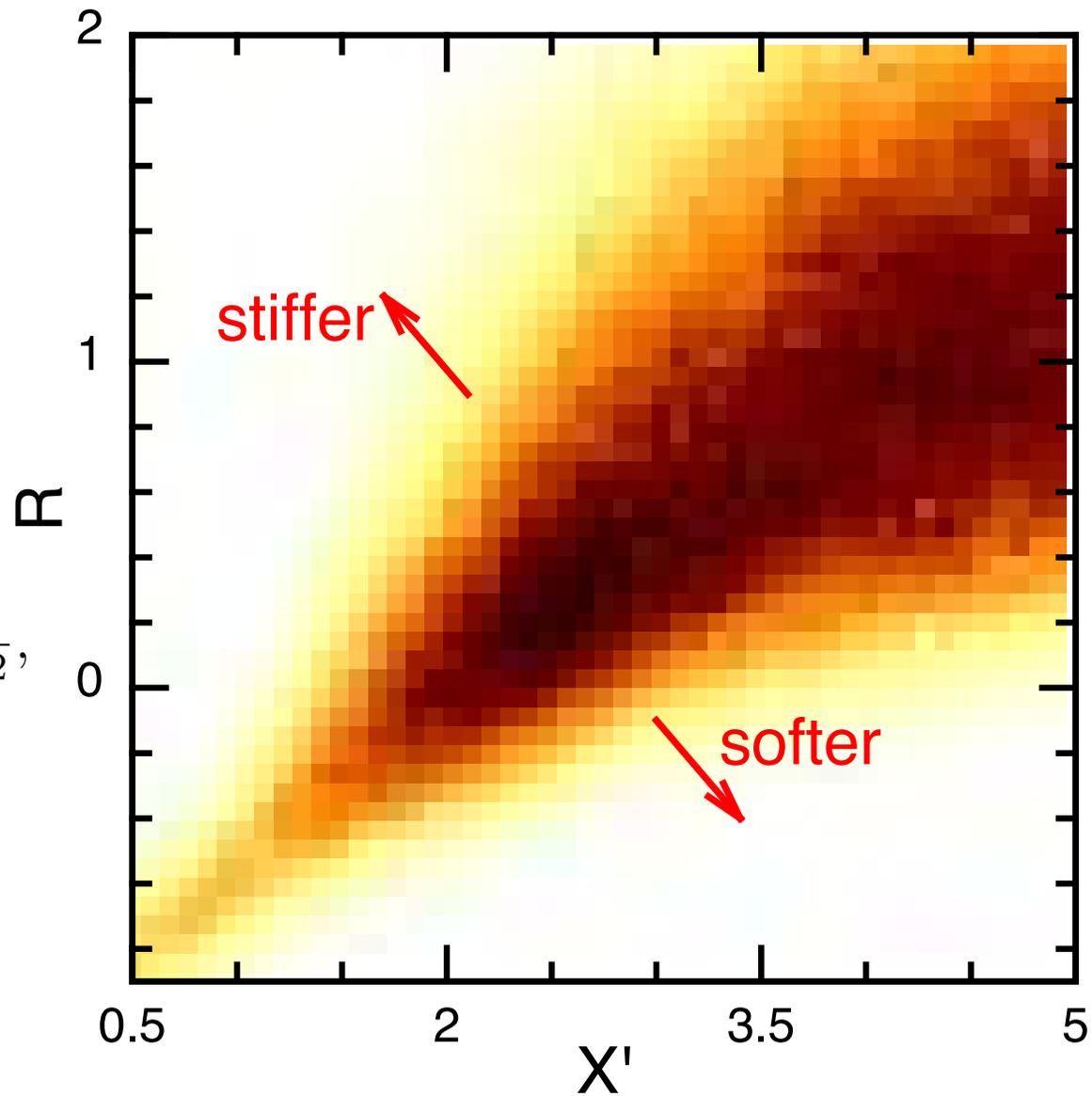


Eq. of State

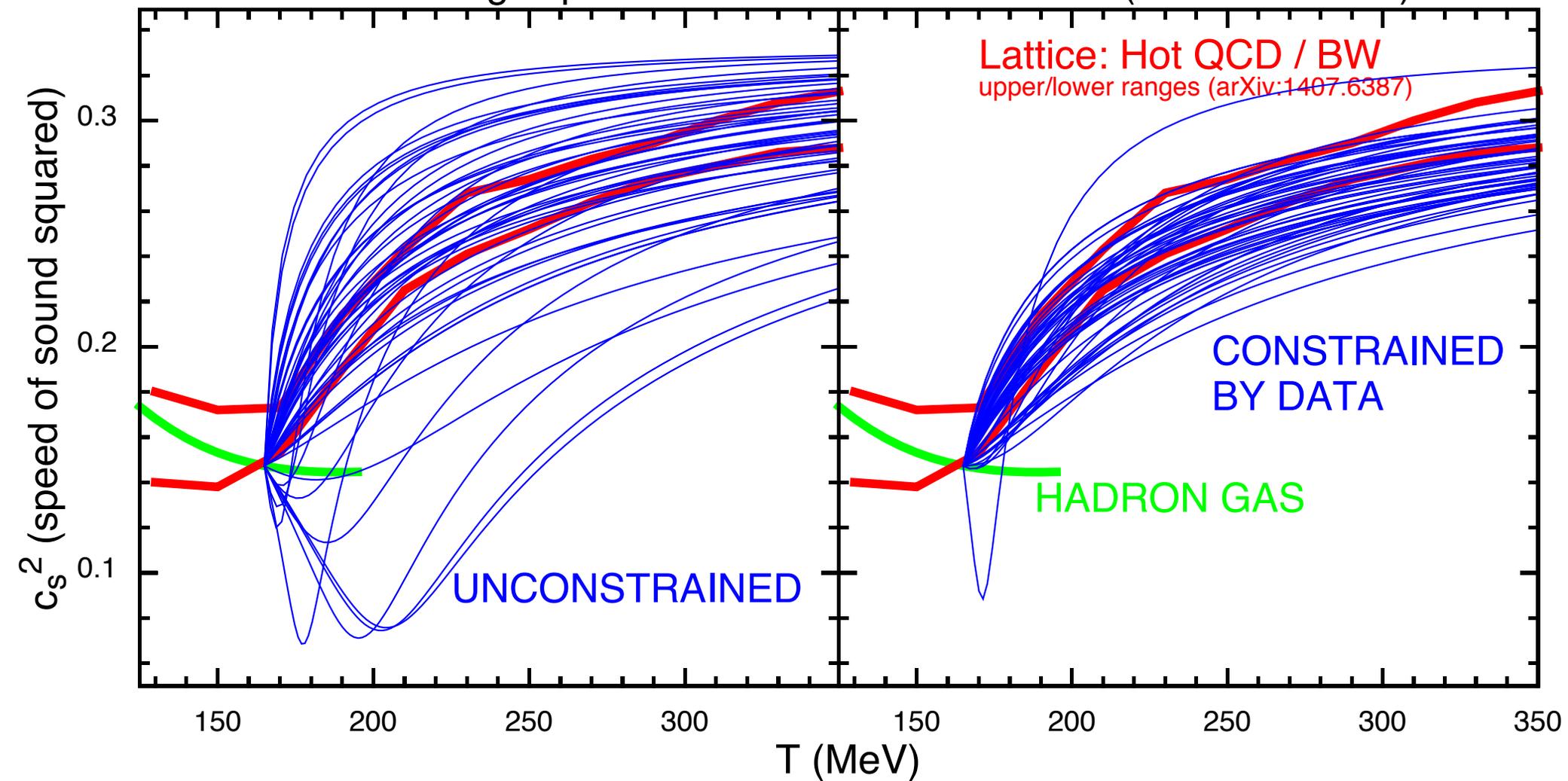
$$c_s^2(\epsilon) = c_s^2(\epsilon_h) + \left(\frac{1}{3} - c_s^2(\epsilon_h) \right) \frac{X_0 x + x^2}{X_0 x + x^2 + X'^2},$$

$$X_0 = X' R c_s(\epsilon) \sqrt{12},$$

$$x \equiv \ln \epsilon / \epsilon_h$$



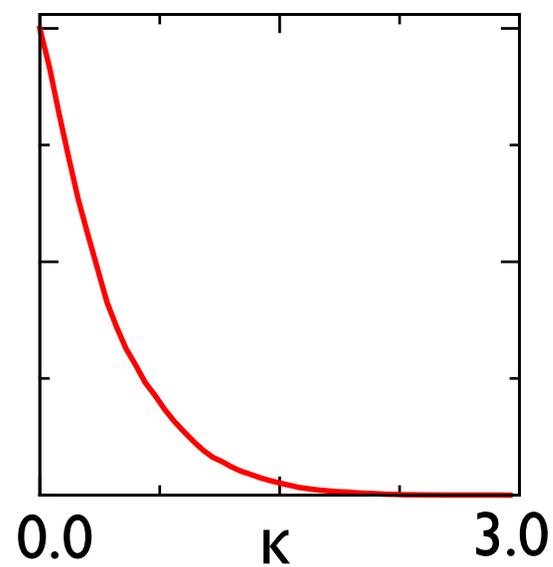
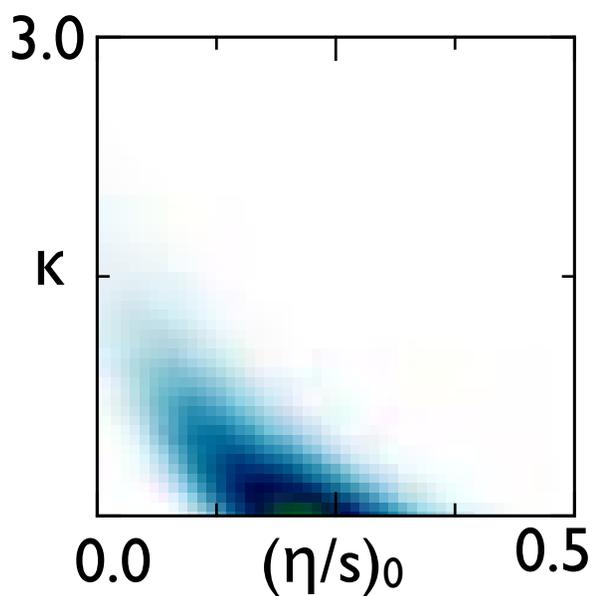
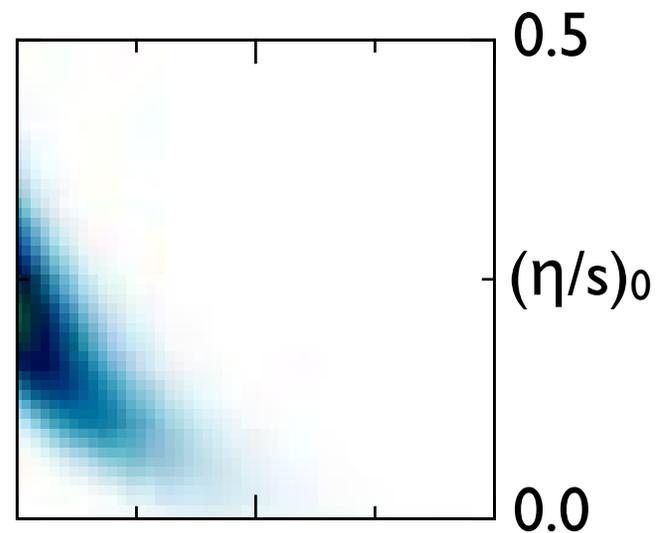
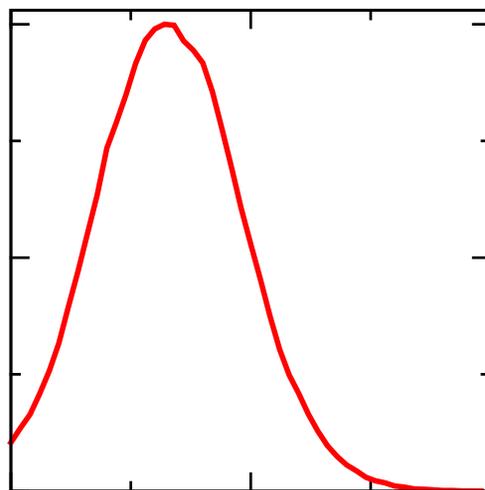
Constraining Eq. of State with RHIC/LHC Data (MADAI Collab.)



$\eta/s(T)$

$$\eta/s = (\eta/s)_0$$

$$+ \kappa \ln(T/165)$$

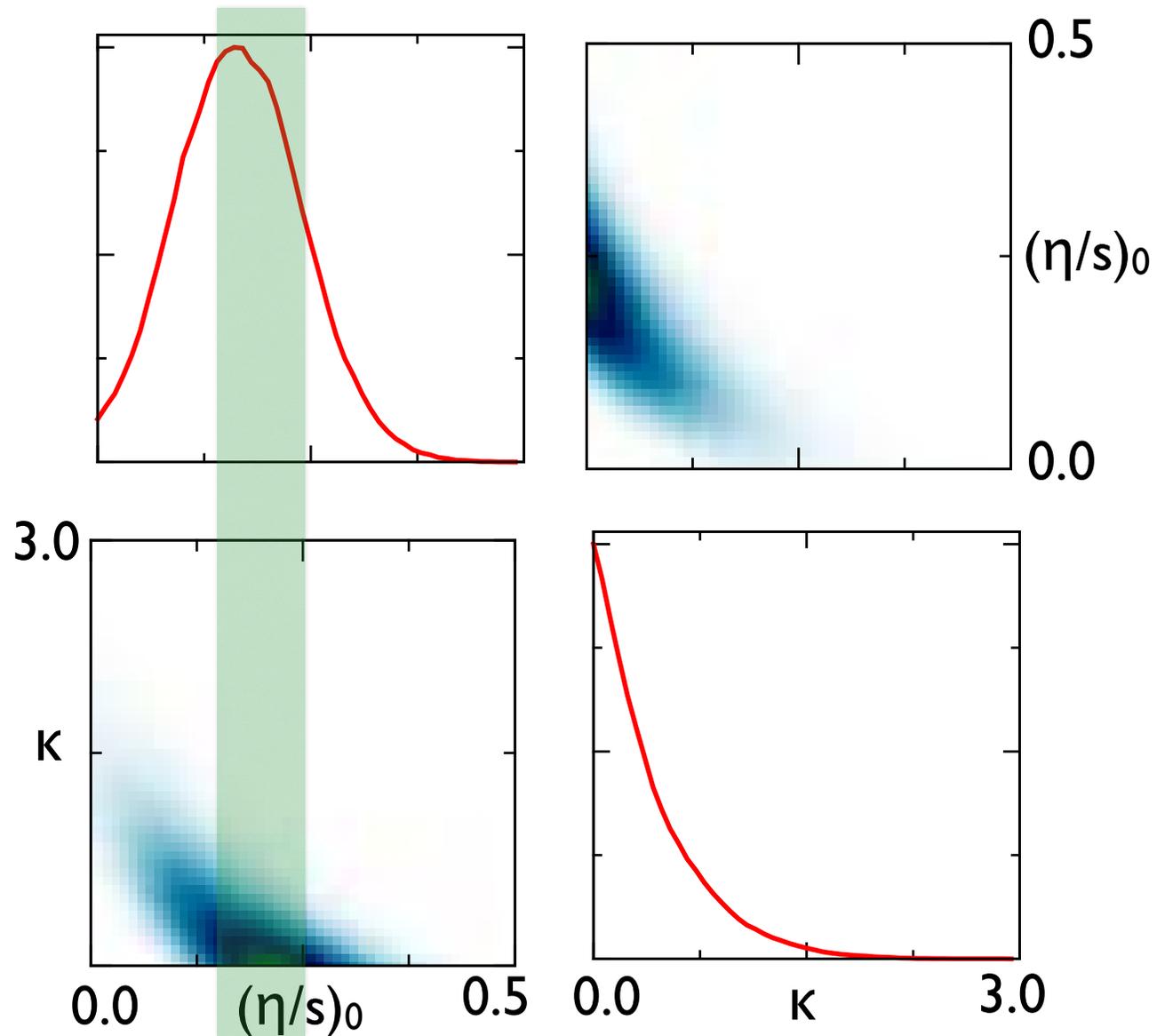


What should you expect for η/s at $T=165$ MeV?

- ADS/CFT: 0.08
- Perturbative QCD: > 0.5 ($\sigma \approx 3$ mb)
- Hadron Gas: ≈ 0.25 ($\sigma \approx 30$ mb)

**Extracted η/s at
T=165 MeV
consistent with
expectations for
hadron gas!**

**Does not rise
strongly in QGP**



RESOLVING POWER OF OBSERVABLES

**How does changing
 $y_{a,\text{exp}}$ or σ_a
alter $\langle\langle x_i \rangle\rangle$ or
 $\langle\langle \delta x_i \delta x_j \rangle\rangle$?**

We need

$$\frac{\partial}{\partial y_a^{(\text{exp})}} \langle\langle x_i \rangle\rangle$$

NOT

$$\frac{\partial}{\partial x_i} y_a^{(\text{mod})}$$

RESOLVING POWER OF OBSERVABLES

$$\langle\langle x_i \rangle\rangle = \frac{\langle x_i \mathcal{L} \rangle}{\langle \mathcal{L} \rangle}$$

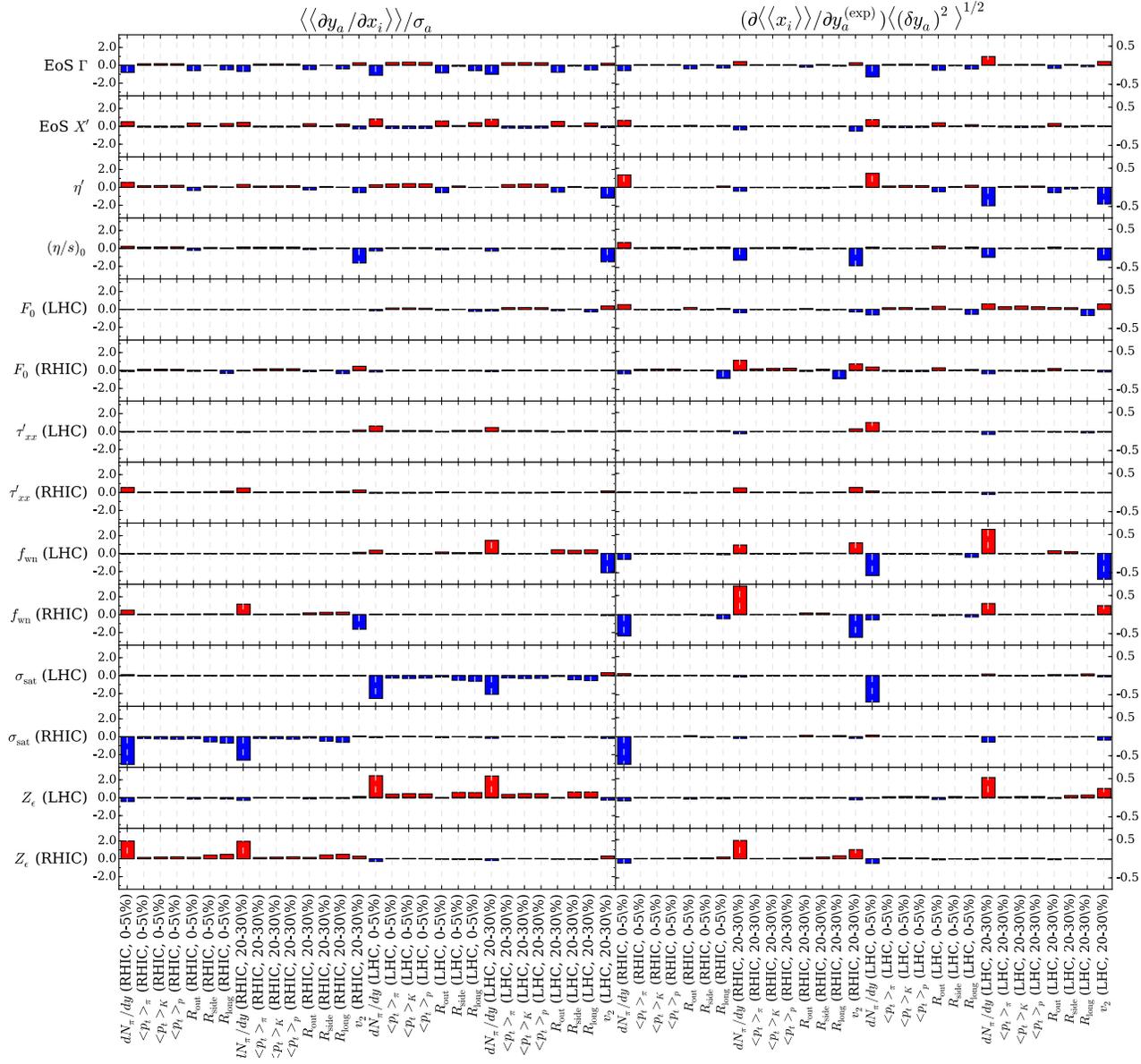
$$\begin{aligned} \frac{\partial}{\partial y_a^{(\text{exp})}} \langle\langle x_i \rangle\rangle &= \langle\langle x_i (\partial_a \mathcal{L}) / \mathcal{L} \rangle\rangle - \langle\langle x_i \rangle\rangle \langle\langle (\partial_a \mathcal{L}) / \mathcal{L} \rangle\rangle \\ &= \langle\langle \delta x_i (\partial_a \mathcal{L}) / \mathcal{L} \rangle\rangle \\ &= -\Sigma_{ab}^{-1} \langle\langle \delta x_i \delta y_b \rangle\rangle \quad (\text{for Gaussian}) \end{aligned}$$

$$\delta x_i = x_i - \langle\langle x_i \rangle\rangle, \quad \delta y_a = y_a - y_a^{(\text{exp})}$$

can find similar relation for $\frac{\partial}{\partial \sigma_a} \langle\langle \delta x_i \delta x_j \rangle\rangle$

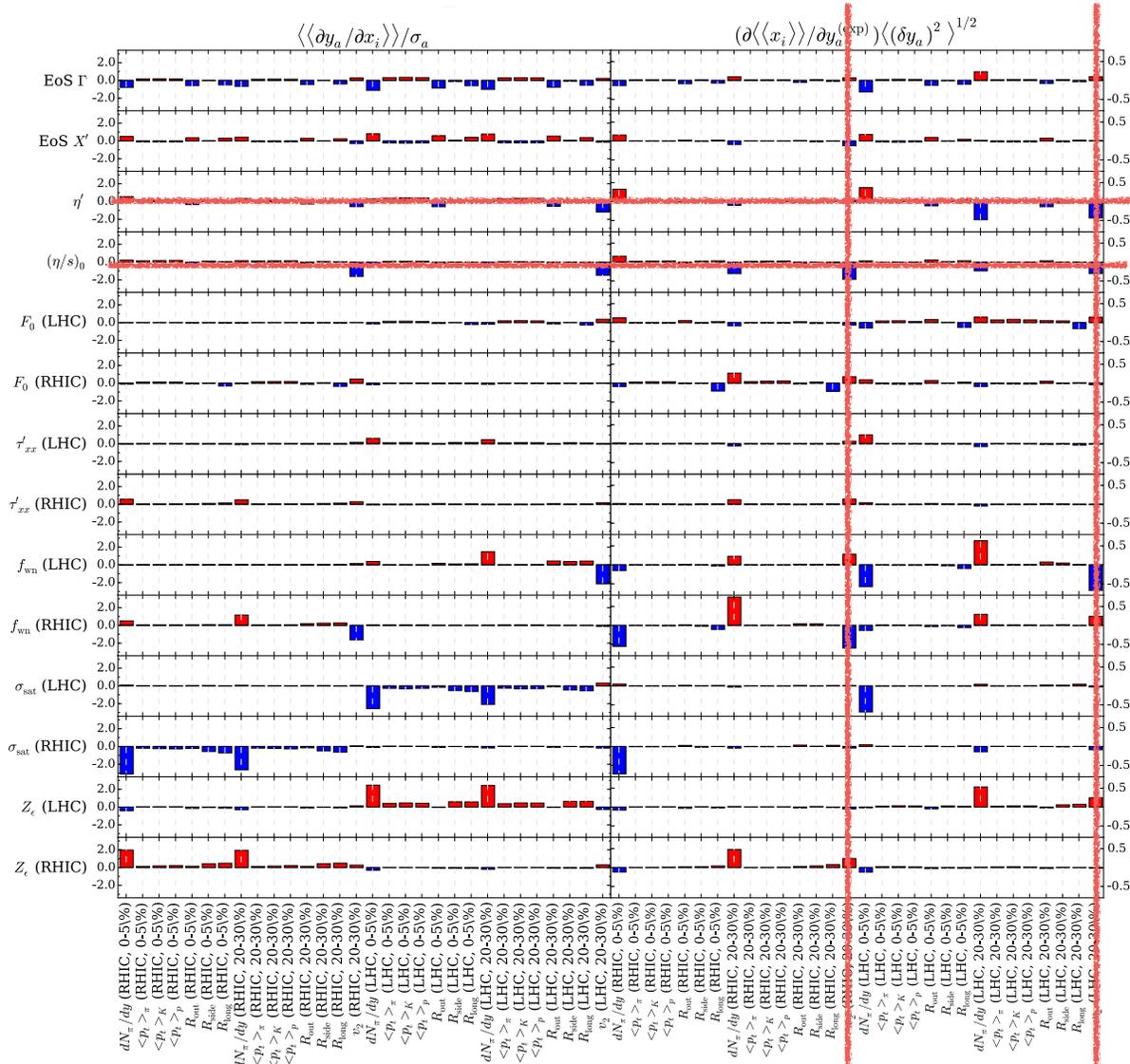
E.Sangaline and S.P., PRC 2016

$$\frac{1}{\sigma_a} \left. \frac{\partial y_a}{\partial x_i} \right|_{y_b \neq a}$$

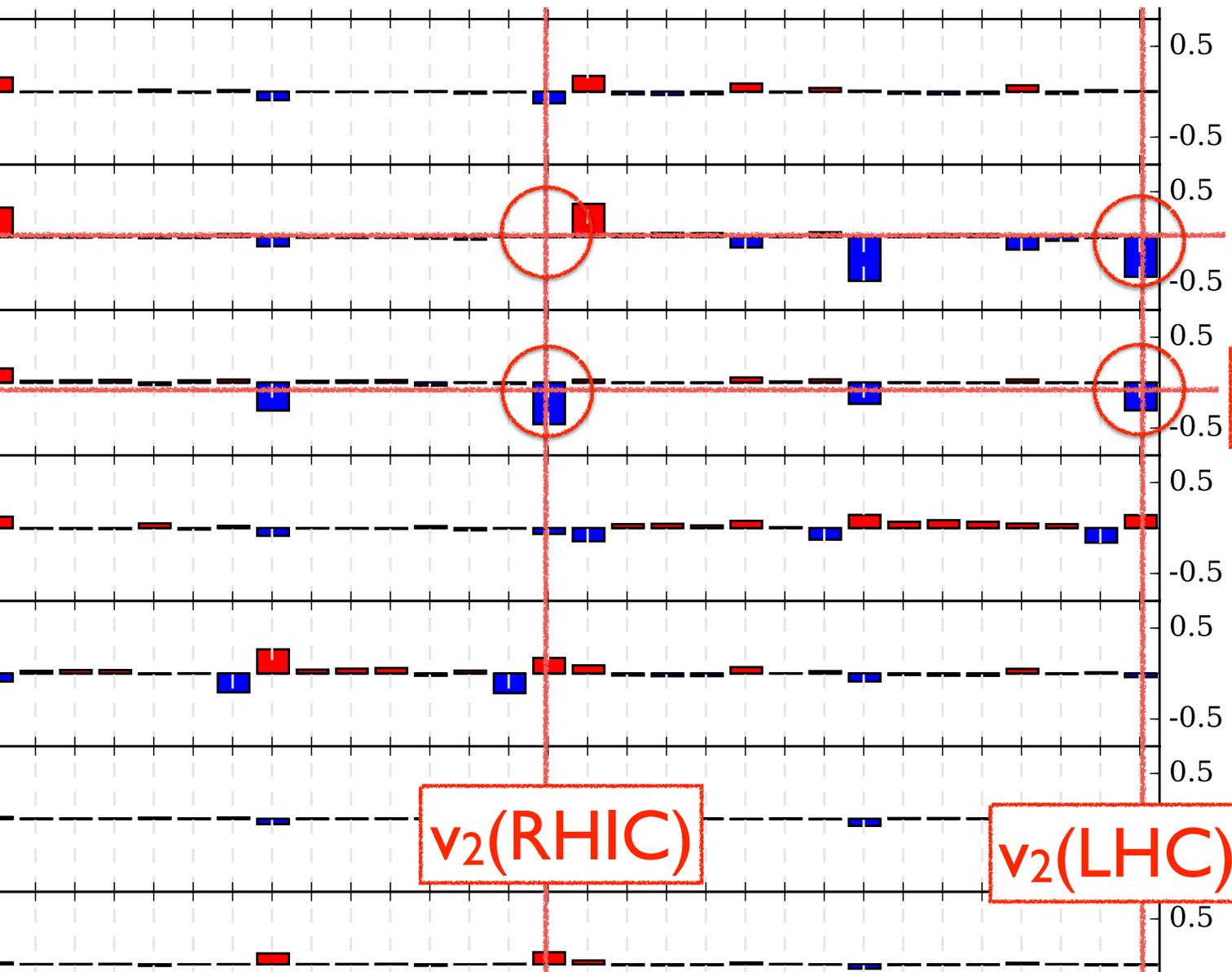


$$\langle \delta y_a \delta y_a \rangle^{1/2} \left. \frac{\partial x_i}{\partial y_a} \right|_{y_b \neq a}$$

$$\frac{1}{\sigma_a} \left. \frac{\partial y_a}{\partial x_i} \right|_{y_{b \neq a}}$$



$$\langle \delta y_a \delta y_a \rangle^{1/2} \left. \frac{\partial x_i}{\partial y_a} \right|_{y_{b \neq a}}$$



η'

η_0

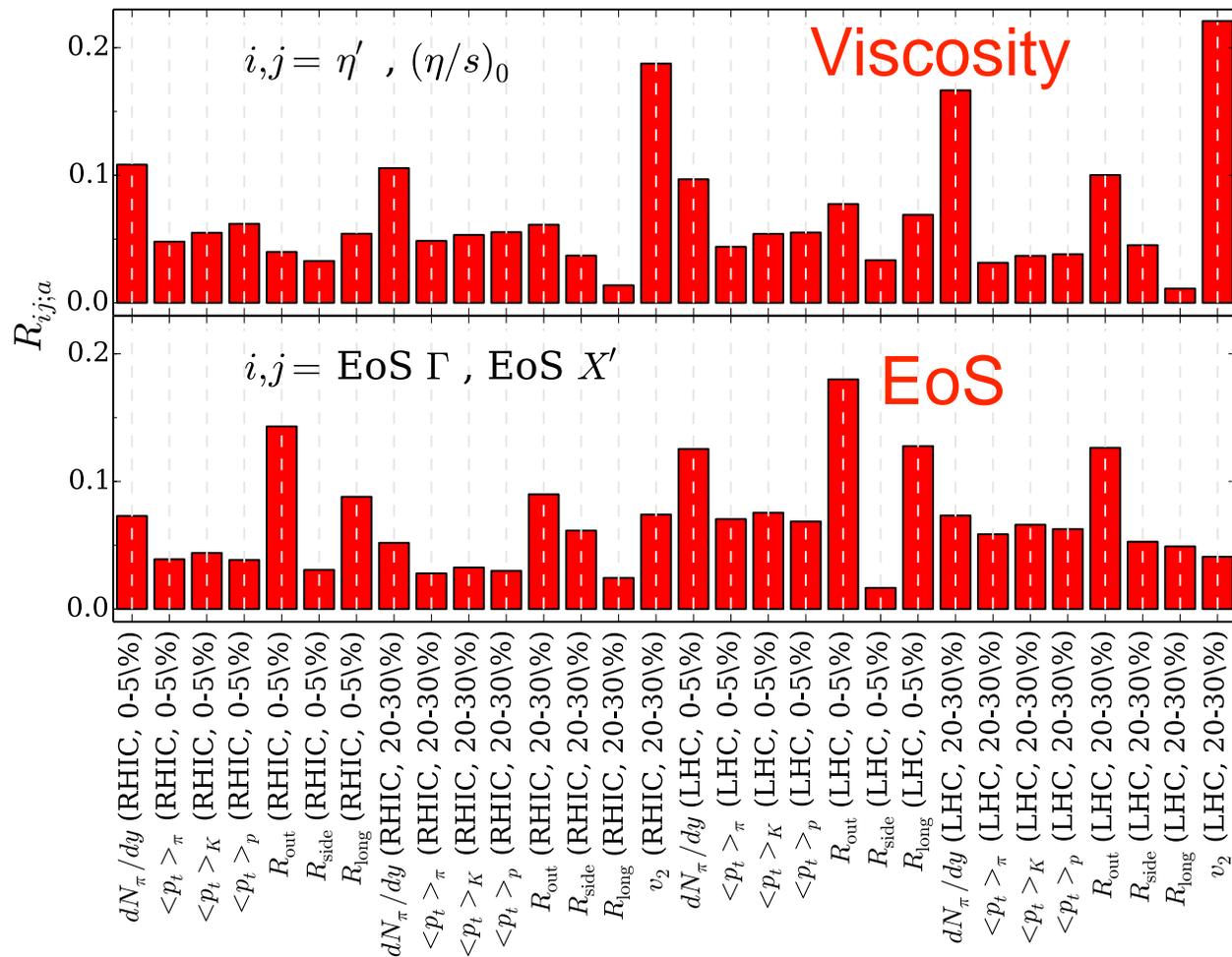
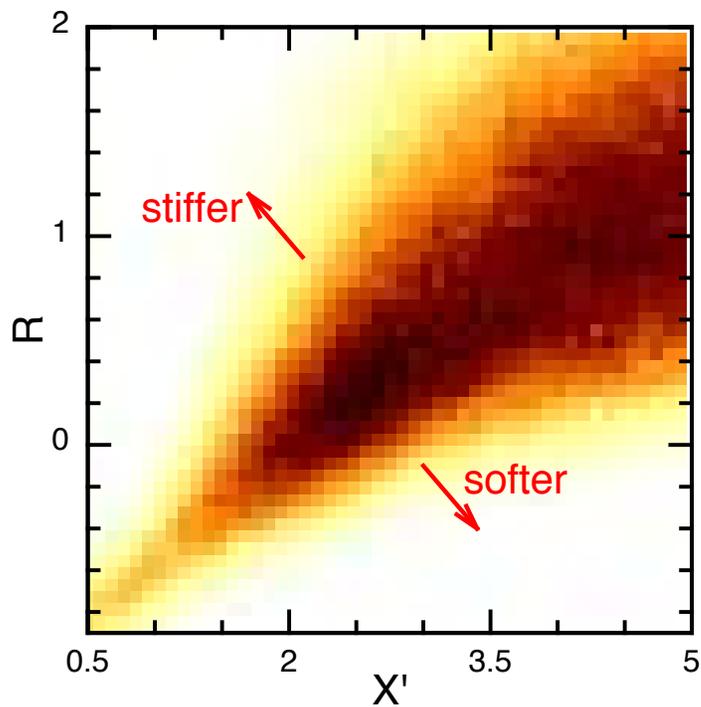
$v_2(\text{RHIC})$

$v_2(\text{LHC})$

$$\langle \delta y_a \delta y_a \rangle^{1/2} \frac{\partial x_i}{\partial y_a} \Big|_{y_b \neq a}$$

$$\frac{d}{d\sigma_y} \sqrt{\begin{vmatrix} \langle\langle \delta x_1 \delta x_1 \rangle\rangle & \langle\langle \delta x_1 \delta x_2 \rangle\rangle \\ \langle\langle \delta x_1 \delta x_2 \rangle\rangle & \langle\langle \delta x_2 \delta x_2 \rangle\rangle \end{vmatrix}} \langle\delta y \delta y\rangle^{1/2}$$

2-Parameter Resolving Power



What determines viscosity?

- Both v_2 and multiplicities
- T-dependence comes from LHC v_2

What determines EoS?

- **Lots of observables**
- **Femtoscopic radii are important**

CONCLUSIONS

- ◆ **Robust, emulation works splendidly**
- ◆ **Scales well to more parameters & more data**
- ◆ **Eq. of State and Viscosity can be extracted from data**
- ◆ **Eq. of State consistent with lattice gauge theory**
- ◆ **Other parameters not as well constrained**
- ◆ **Heavy-Ion Physics can be a Quantitative Science!!!!**

FUTURE

- ◆ Improve statement of uncertainties
- ◆ Add parameters
 - many related to hadronization region
- ◆ Consider more data
 - more observables
 - Beam energy scan (Yikes!!!!)
- ◆ Improve models
 - Lumpy initial conditions
 - 3D calculations for lower energies
 - Fill in missing physics (e.g. bulk viscosity)

THINGS TO KEEP IN MIND...

- ◆ Remember what you're trying to do
 - parameter constraint?
 - identify weakness of models?
 - predictivity?
 - not draw lines through data!
 - more parameters(physics) are better
- ◆ Data can be redundant
 - correlated uncertainties
 - underestimates of uncertainty
- ◆ Models have systematic uncertainties
 - requires objective self doubt

Additional slide: Charge BFs and charge susceptibilities

