



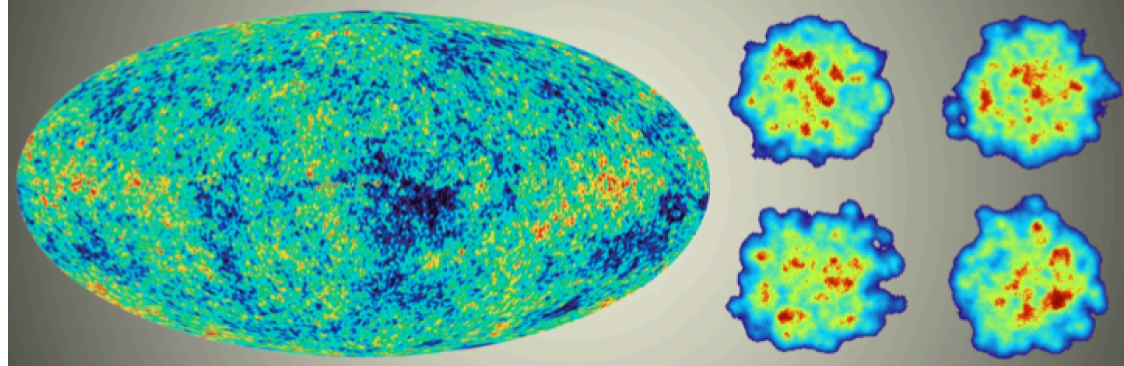
Correlations between harmonics

Jiangyong Jia

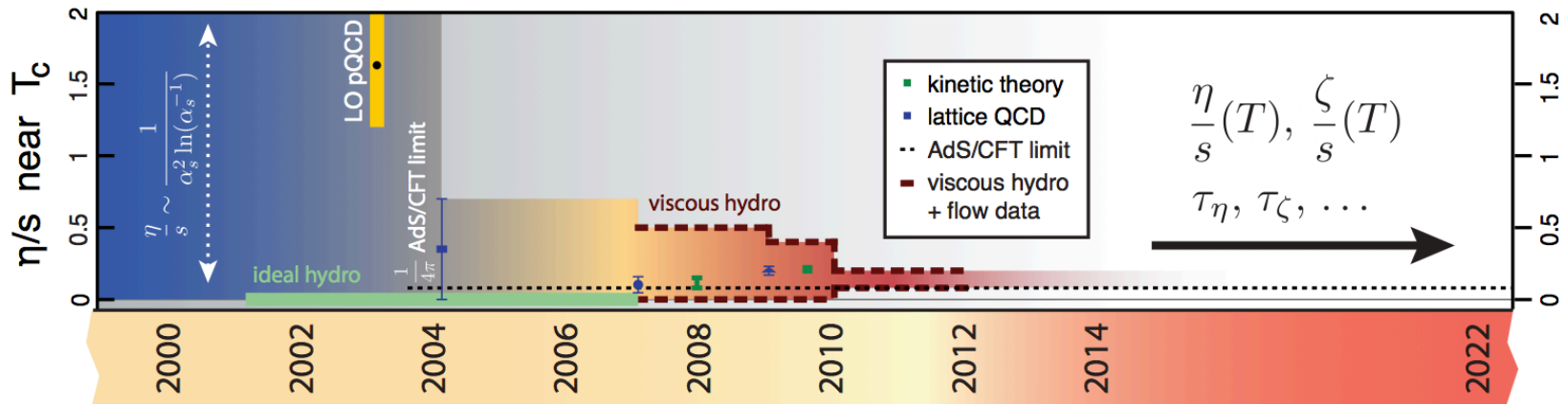
The 2nd International Conference on the
Initial Stages in High-Energy Nuclear Collisions

Recent accomplishments

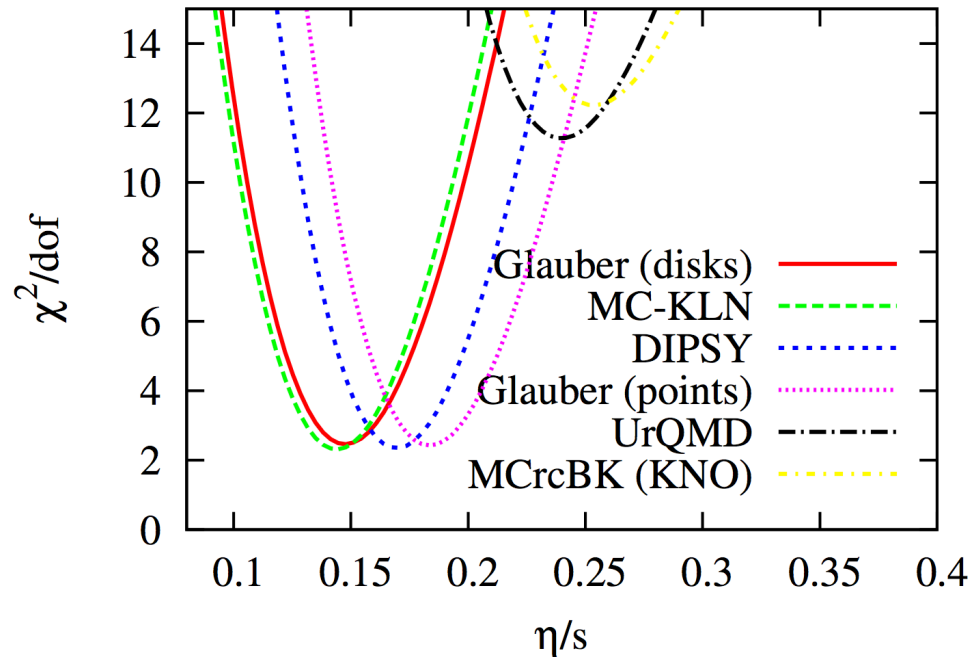
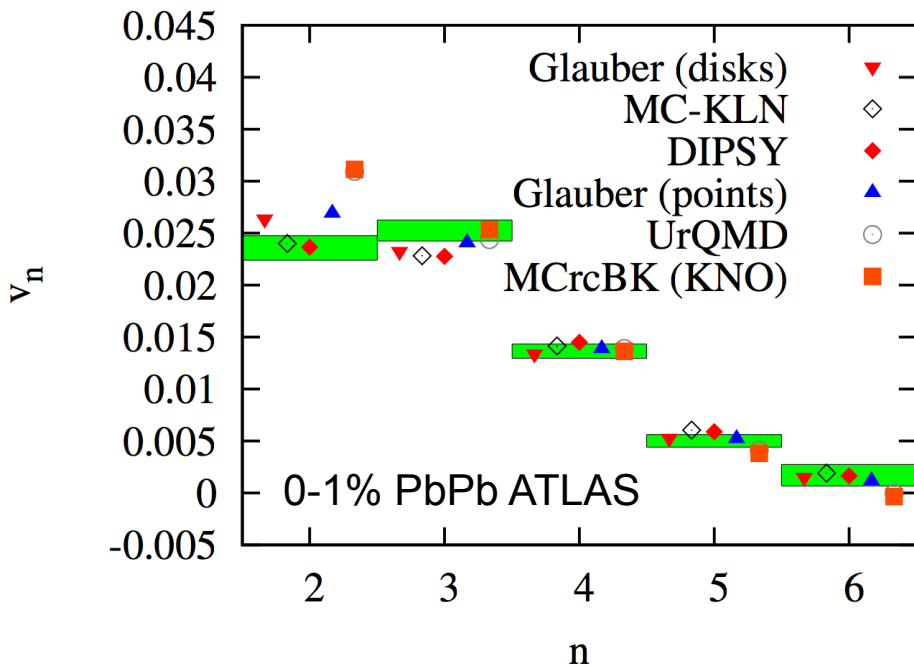
Improved the **space-time picture** of bulk matter evolution



Improved extraction of the **medium properties**



Challenge: simultaneous control of two unknowns



For traditional observables, i.e. $\langle v_n \rangle$,
 experimental error no-longer the limiting factors

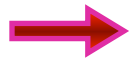
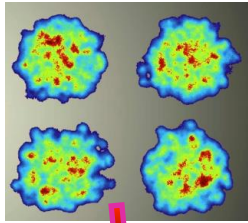
Need new observables with new intuitive insights!

- **Pb+Pb/Au+Au top energy: as complete as possible**
 - Event-by-event flow observables → initial fluctuation & flow correlation
 - Event shape engineer & engen-mode ana. → geometry model & hydro. response
 - Longitudinal fluctuations → particle production & early time dynamics
 - Medium response → relaxation mechanism for local energy deposition
 - PID v_n → hadronization and hadronic transport
- **System size scan**
 - e.g. Cu+Au or U+U → hydro response to extreme geometries.
 - p+A vs peripheral A+A → onset of collectivity? onset of hydrodynamics?
- **Energy scan of Au+Au collisions at RHIC**
 - Temperature dependence of medium properties, partonic vs hadronic.
 - However lower stats & p_T reach

Event-by-event observables

Many little bangs

1104.4740, 1209.2323, 1203.5095, 1312.3572



$$p(v_n, v_m, \dots, \Phi_n, \Phi_m, \dots) = \frac{1}{N_{\text{evts}}} \frac{dN_{\text{evts}}}{dv_n dv_m \dots d\Phi_n d\Phi_m \dots}$$

Moments:

$$\langle \cos(n_1 \phi_1 + n_2 \phi_2 \dots + n_m \phi_m) \rangle = \langle v_{n_1} v_{n_2} \dots v_{n_m} \cos(n_1 \Phi_{n_1} + n_2 \Phi_{n_2} \dots + n_m \Phi_{n_m}) \rangle \quad \sum n_i = 0$$

e.g. $\langle \cos(n\phi_1 - n\phi_2 + n\phi_3 - n\phi_4) \rangle = \langle v_n^4 \cos(n\Phi_n - n\Phi_n + n\Phi_n - n\Phi_n) \rangle = \langle v_n^4 \rangle$

$$\langle \cos(n\phi_1 - n\phi_2 + m\phi_3 - m\phi_4) \rangle = \langle v_n^2 v_m^2 \cos(n\Phi_n - n\Phi_n + m\Phi_m - m\Phi_m) \rangle = \langle v_n^2 v_m^2 \rangle$$

Cumulants obtained by combining with lower order correlators:

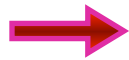
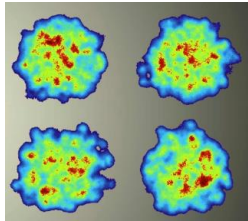
$$c_{n\{4\}} = \langle v_n^4 \rangle - 2 \langle v_n^2 \rangle^2 \quad p(v_n)$$

$$c_{n,m\{2,2\}} = \langle v_n^2 v_m^2 \rangle - \langle v_n^2 \rangle \langle v_m^2 \rangle \quad p(v_n, v_m)$$

Event-by-event observables

Many little bangs

1104.4740, 1209.2323, 1203.5095, 1312.3572



$$p(v_n, v_m, \dots, \Phi_n, \Phi_m, \dots) = \frac{1}{N_{\text{evts}}} \frac{dN_{\text{evts}}}{dv_n dv_m \dots d\Phi_n d\Phi_m \dots}$$

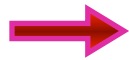
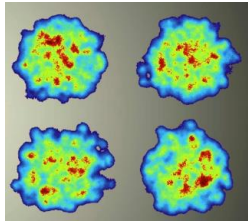
| | pdf's | cumulants |
|-------------------|---------------------------------|---|
| | $p(v_n)$ | $v_n \{2k\}, k = 1, 2, \dots$ |
| | $p(v_n, v_m)$ | $\langle v_n^2 v_m^2 \rangle - \langle v_n^2 \rangle \langle v_m^2 \rangle$ |
| Flow-amplitudes | $p(v_n, v_m, v_l)$ | $\langle v_n^2 v_m^2 v_l^2 \rangle + 2 \langle v_n^2 \rangle \langle v_l^2 \rangle - \langle v_n^2 v_m^2 \rangle \langle v_l^2 \rangle - \langle v_m^2 v_l^2 \rangle \langle v_n^2 \rangle - \langle v_n^2 v_l^2 \rangle \langle v_m^2 \rangle$ |
| | ... | Obtained as above |
| EP-correlation | $p(\Phi_n, \Phi_m, \dots)$ | $\langle v_n^{c_n} v_m^{c_m} \dots \cos(c_n n \Phi_n + c_m m \Phi_m + \dots) \rangle - \langle v_n^{c_n} \rangle \langle v_m^{c_m} \rangle \dots \cos(c_n n \Phi_n + c_m m \Phi_m + \dots)$ $\sum_k k c_k = 0$ |
| Mixed-correlation | $p(v_l, \Phi_n, \Phi_m, \dots)$ | $\langle v_l^2 v_n^{c_n} \dots \cos(c_n n \Phi_n + c_m m \Phi_m + \dots) \rangle - \langle v_l^2 \rangle \langle v_n^{c_n} v_m^{c_m} \dots \cos(c_n n \Phi_n + c_m m \Phi_m + \dots) \rangle$ $\sum_k k c_k = 0$ |

**a subset moments of PDF
only even modes**

Event-by-event observables

Many little bangs

1104.4740, 1209.2323, 1203.5095, 1312.3572



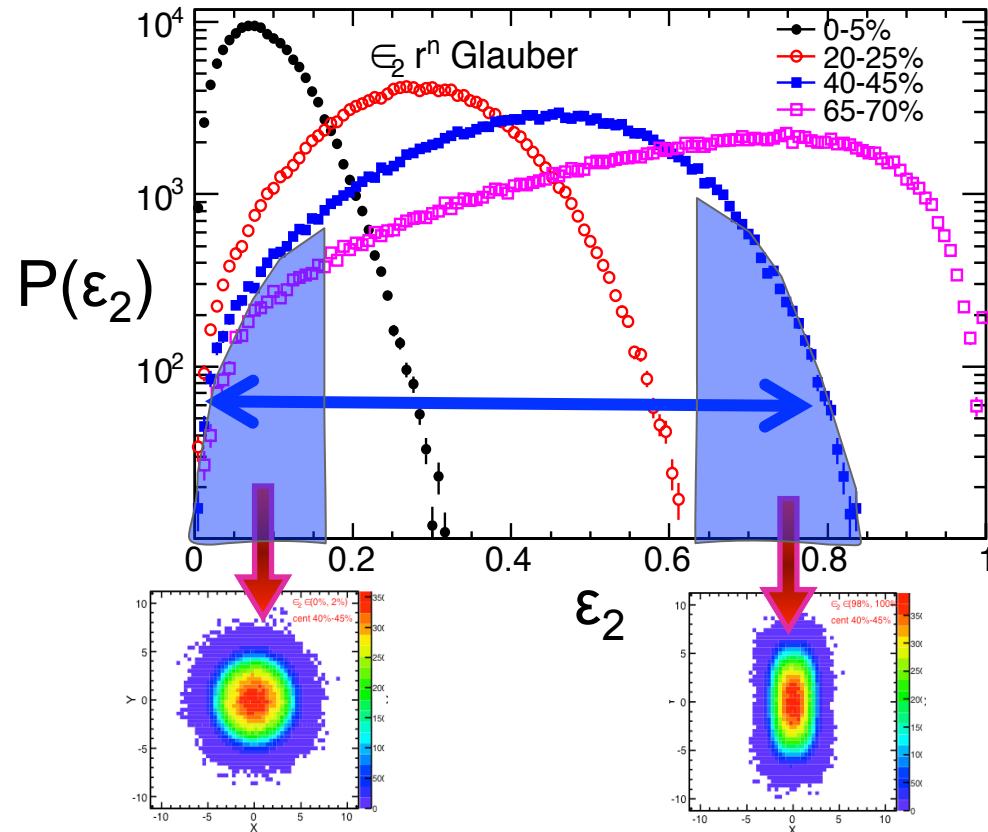
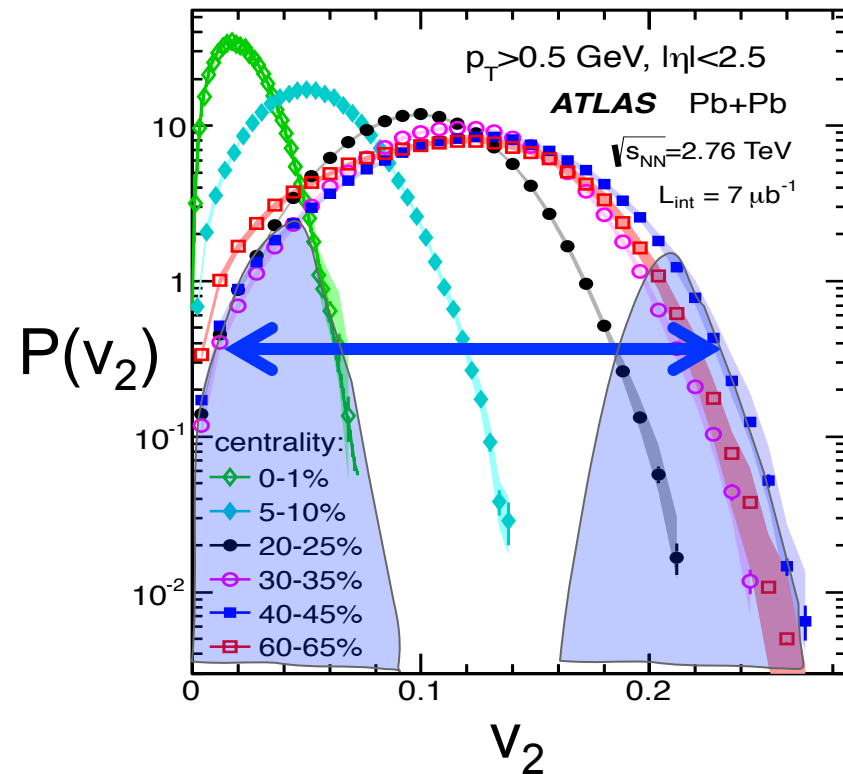
$$p(v_n, v_m, \dots, \Phi_n, \Phi_m, \dots) = \frac{1}{N_{\text{evts}}} \frac{dN_{\text{evts}}}{dv_n dv_m \dots d\Phi_n d\Phi_m \dots}$$

| | pdf's | cumulants | event-shape method |
|-------------------|---------------------------------|---|--------------------|
| | $p(v_n)$ | $v_n \{2k\}, k = 1, 2, \dots$ | NA |
| | $p(v_n, v_m)$ | $\langle v_n^2 v_m^2 \rangle - \langle v_n^2 \rangle \langle v_m^2 \rangle$ | yes |
| Flow-amplitudes | $p(v_n, v_m, v_l)$ | $\langle v_n^2 v_m^2 v_l^2 \rangle + 2 \langle v_n^2 \rangle \langle v_m^2 \rangle \langle v_l^2 \rangle - \langle v_n^2 v_m^2 \rangle \langle v_l^2 \rangle - \langle v_n^2 v_l^2 \rangle \langle v_m^2 \rangle - \langle v_m^2 v_l^2 \rangle \langle v_n^2 \rangle$ | yes |
| | ... | Obtained ¹ as above | yes |
| EP-correlation | $p(\Phi_n, \Phi_m, \dots)$ | $\langle v_n^{c_n} v_m^{c_m} \dots \cos(c_n n \Phi_n + c_m m \Phi_m + \dots) \rangle$ $\sum_k k c_k = 0$ | yes |
| Mixed-correlation | $p(v_l, \Phi_n, \Phi_m, \dots)$ | $\langle v_l^2 v_n^{c_n} \dots \cos(c_n n \Phi_n + c_m m \Phi_m + \dots) \rangle - \langle v_l^2 \rangle \langle v_n^{c_n} v_m^{c_m} \dots \cos(c_n n \Phi_n + c_m m \Phi_m + \dots) \rangle$ $\sum_k k c_k = 0$ | yes |

a subset moments of PDF
only even modes

Event shape engineering technique

$$p(v_n, v_m, \dots, \Phi_n, \Phi_m, \dots) = \frac{1}{N_{\text{evts}}} \frac{dN_{\text{evts}}}{dv_n dv_m \dots d\Phi_n d\Phi_m \dots}$$



- Measure v_n or EP correlation for different v_2 at fixed centrality

$$p(v_n, v_2)$$

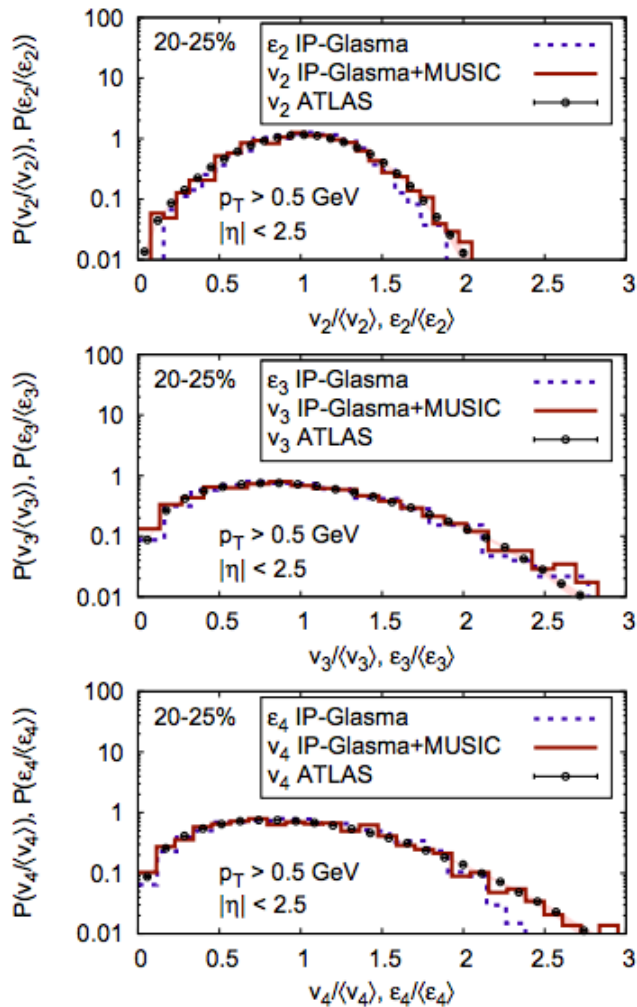
$$p(\Phi_n, \Phi_m, v_2)$$

Applications

| | pdf's | cumulants | event-shape method |
|-------------------|---------------------------------|--|--------------------|
| | $p(v_n)$ | $v_n\{2k\}, k = 1, 2, \dots$ | NA |
| | $p(v_n, v_m)$ | $\langle v_n^2 v_m^2 \rangle - \langle v_n^2 \rangle \langle v_m^2 \rangle$ | yes |
| Flow-amplitudes | $p(v_n, v_m, v_l)$ | $\langle v_n^2 v_m^2 v_l^2 \rangle + 2\langle v_n^2 \rangle \langle v_m^2 \rangle \langle v_l^2 \rangle - \langle v_n^2 v_m^2 \rangle \langle v_l^2 \rangle - \langle v_m^2 v_l^2 \rangle \langle v_n^2 \rangle - \langle v_l^2 v_n^2 \rangle \langle v_m^2 \rangle$ | yes |
| | ... | Obtained recursively as above | yes |
| EP-correlation | $p(\Phi_n, \Phi_m, \dots)$ | $\langle v_n^{c_n} v_m^{c_m} \dots \cos(c_n n \Phi_n + c_m m \Phi_m + \dots) \rangle$ $\sum_k k c_k = 0$ | yes |
| Mixed-correlation | $p(v_l, \Phi_n, \Phi_m, \dots)$ | $\langle v_l^2 v_n^{c_n} v_m^{c_m} \dots \cos(c_n n \Phi_n + c_m m \Phi_m + \dots) \rangle - \langle v_l^2 \rangle \langle v_n^{c_n} v_m^{c_m} \dots \cos(c_n n \Phi_n + c_m m \Phi_m + \dots) \rangle$ $\sum_k k c_k = 0$ | yes |

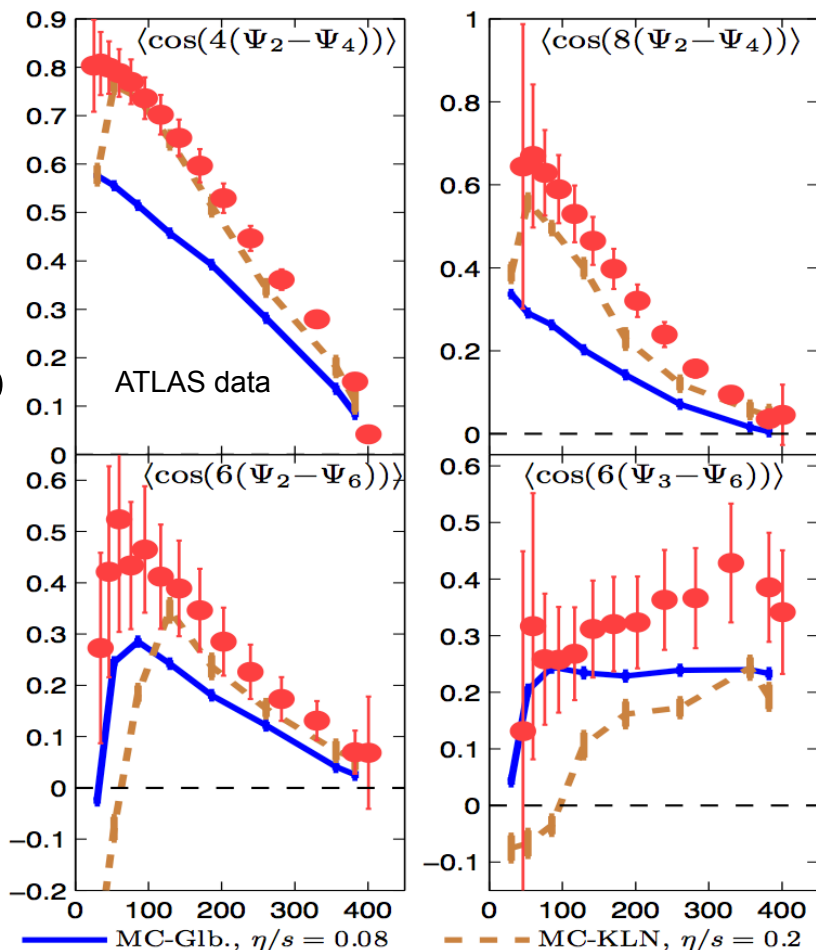
Current measurements of E-by-E distributions

$p(v_2), p(v_3), p(v_4)$



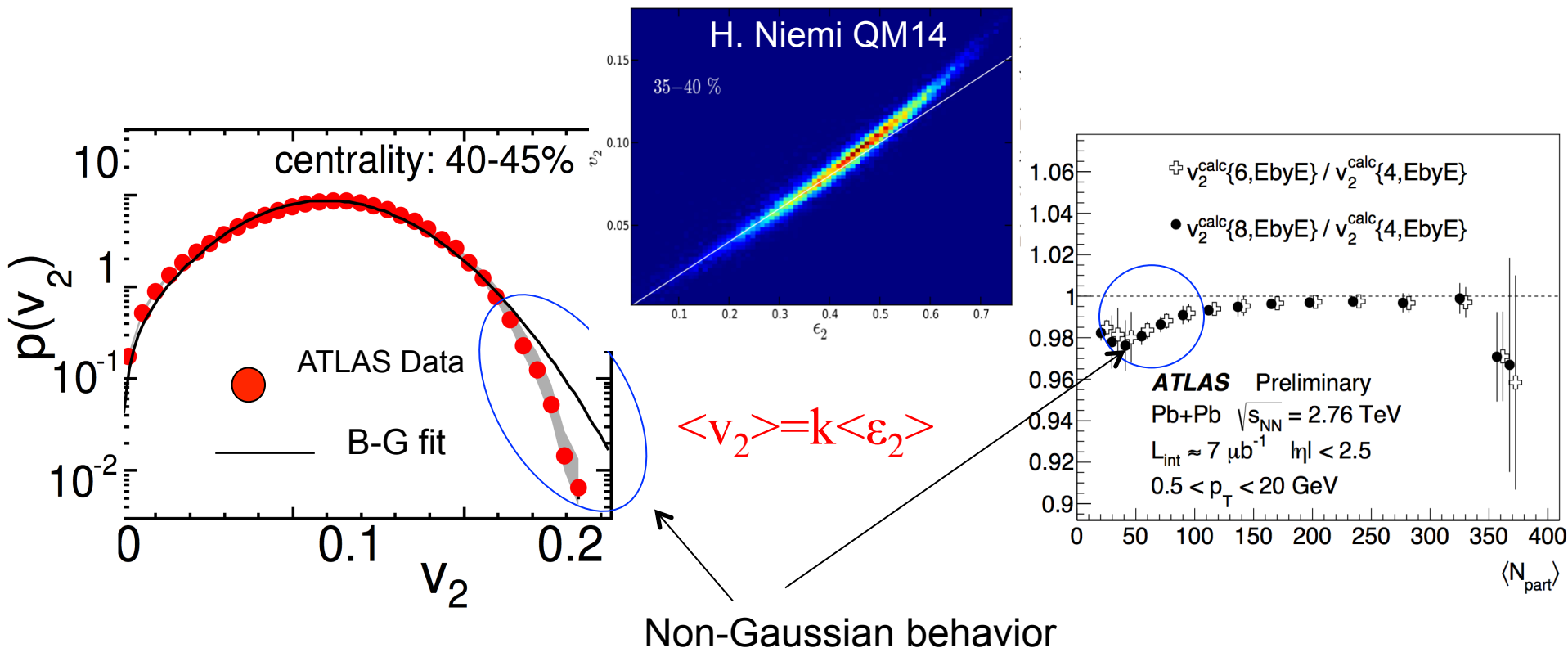
$p(\Phi_n, \Phi_m)$ and $p(\Phi_n, \Phi_m, \Phi_L)$

Pb+Pb



Measured distributions quantitatively described by hydro.

Non-Gaussianity in the $p(v_2)$ distribution



- Reflected by a 1-2% change beyond 4th order cumulants
 - Note: $p(v_2)$ contains more information than first few cumulants.

Non-BG in $p(\epsilon_2)$ or non-linearity of response for large ϵ_2

EP correlation: how $(\varepsilon_n, \Phi_n^*)$ are transferred to (v_n, Φ_n) ?

- Flow response is linear for v_2 and v_3 : $v_n \propto \varepsilon_n$ and $\Phi_n \approx \Phi_n^*$ i.e.

$$v_2 e^{-i2\Phi_2} \propto \varepsilon_2 e^{-i2\Phi_2^*}, \quad v_3 e^{-i3\Phi_3} \propto \varepsilon_3 e^{-i3\Phi_3^*}$$

- Higher-order can arise from EP correlations (**mode mixing**), e.g. :

$$v_4 e^{i4\Phi_4} \propto \varepsilon_4 e^{i4\Phi_4^*} + c v_2^2 e^{i4\Phi_2} + \dots$$

$$v_5 e^{i5\Phi_5} \propto \varepsilon_5 e^{i5\Phi_5^*} + c v_2 v_3 e^{i(2\Phi_2+3\Phi_3)} + \dots$$

$$v_6 e^{i6\Phi_6} \propto \varepsilon_6 e^{i6\Phi_6^*} + c_1 v_2^3 e^{i6\Phi_2} + c_2 v_3^2 e^{i6\Phi_3} + c_3 v_2 \varepsilon_4 e^{i(2\Phi_2+4\Phi_4^*)} \dots$$

Ollitrault, Luzum, Teaney, Li, Heinz, Chun....

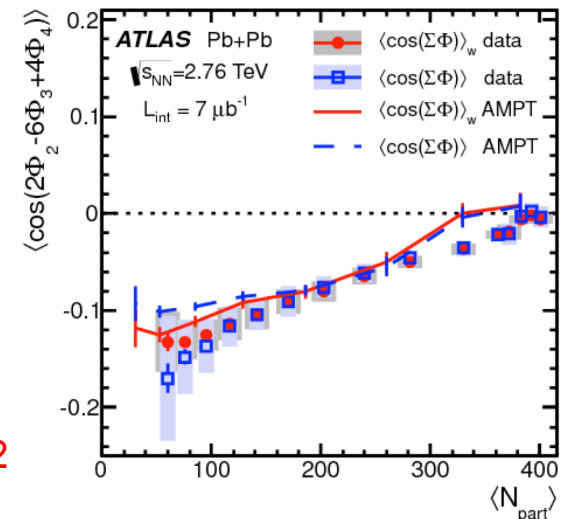
(14 correlators from ATLAS)

- Some correlators lack intuitive explanation

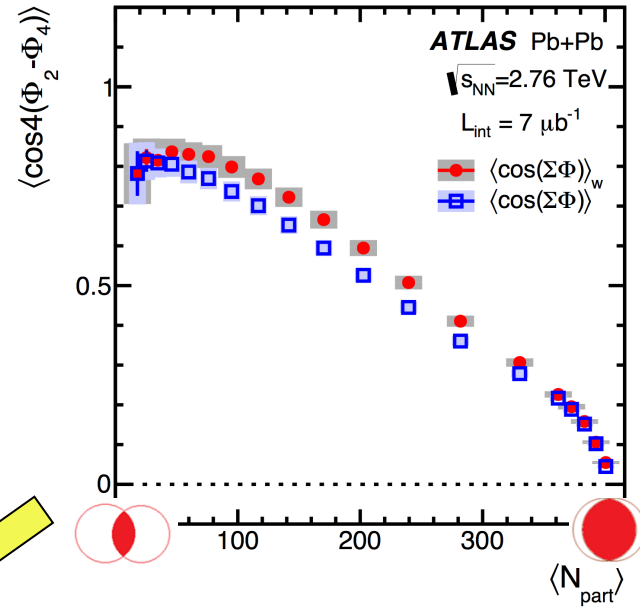
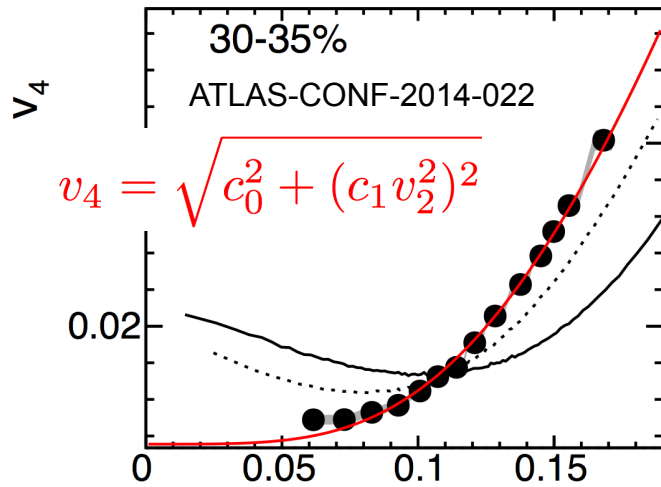
e.g. $\langle \cos(2\Phi_2 - 6\Phi_3 + 4\Phi_4) \rangle$ correlation

- Although described by EbyE hydro and AMPT

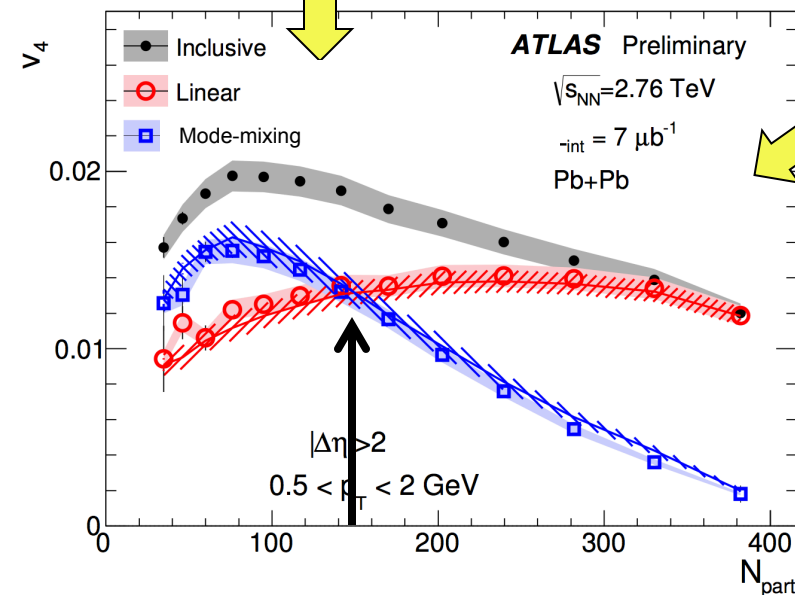
Naturally expect correlation between v_n and v_2



v_4-v_2 correlation from event-shape engineering 13



$$v_4 e^{i4\Phi_4} = c_0 e^{i4\Phi_4^*} + c_1 (v_2 e^{i2\Phi_2})^2 + \dots$$



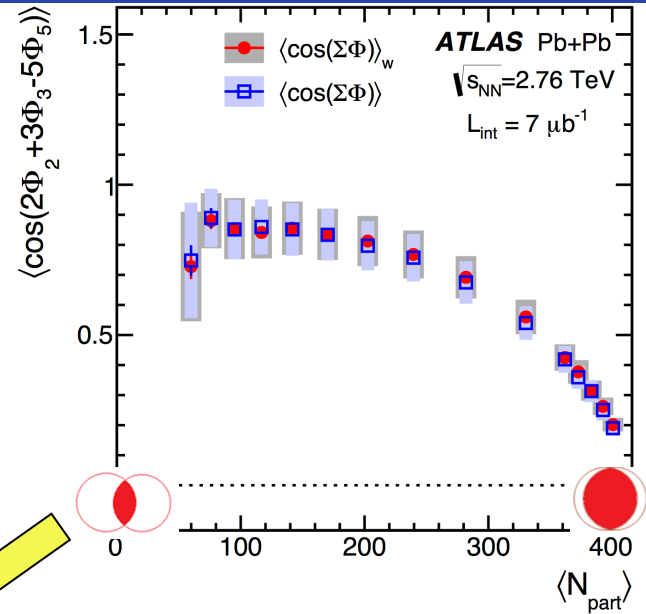
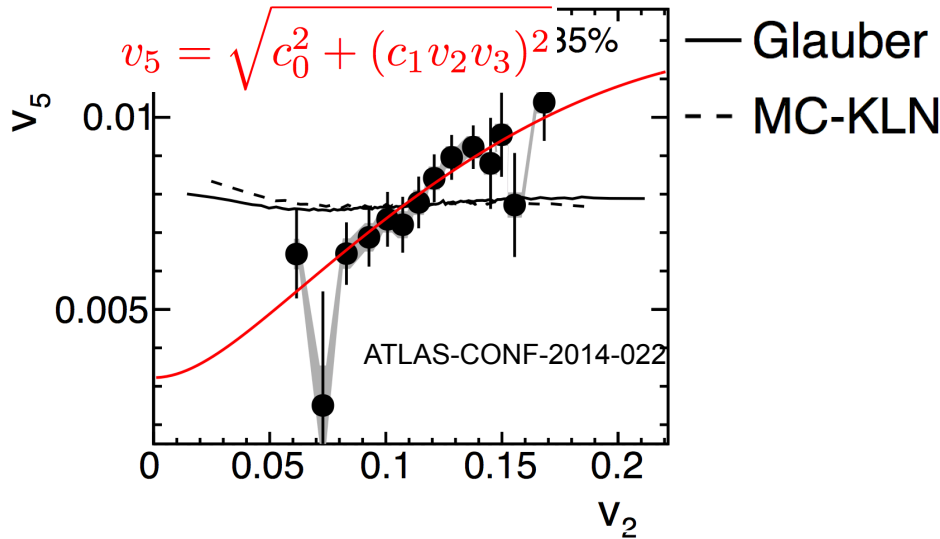
$$v_4^{NL} = v_4 \langle \cos 4(\Phi_2 - \Phi_4) \rangle$$

$$v_4^L = \sqrt{v_4^2 - (v_4^{NL})^2}$$

Compare well with expectation from EP correlations.
higher-order mode-mixing negligible

Separate geometry and mode-mixing components!!

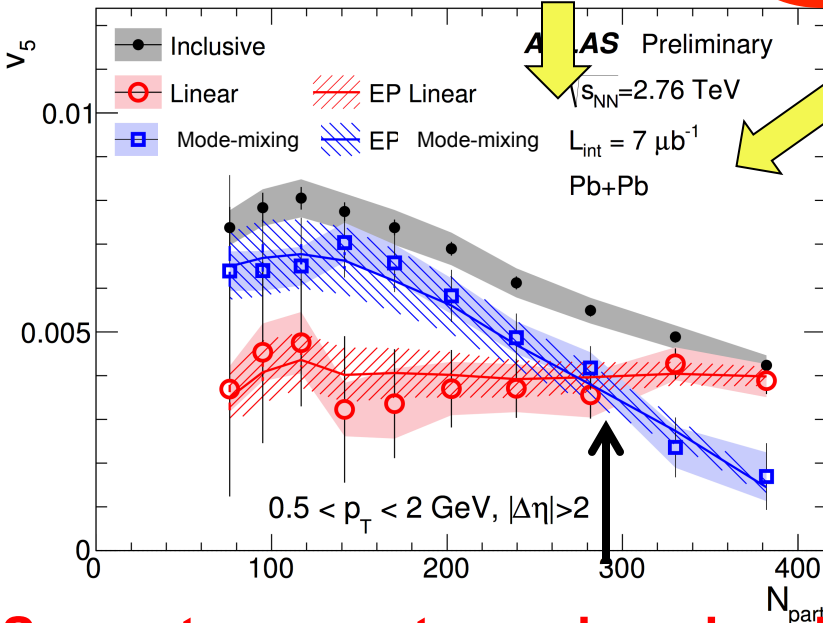
v_5-v_2 correlation from event-shape engineering ¹⁴



$$v_5 e^{i5\Phi_5} \propto c_0 e^{i5\Phi_5^*} + c_1 v_2 v_3 e^{i(2\Phi_2 + 3\Phi_3)} + \dots$$

$$v_5^{NL} = v_5 \langle \cos(2\Phi_2 + 3\Phi_3 - 5\Phi_5) \rangle$$

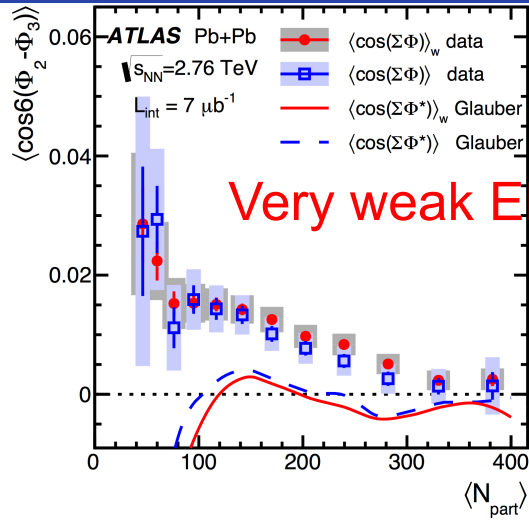
$$v_5^L = \sqrt{v_5^2 - (v_5^{NL})^2}$$



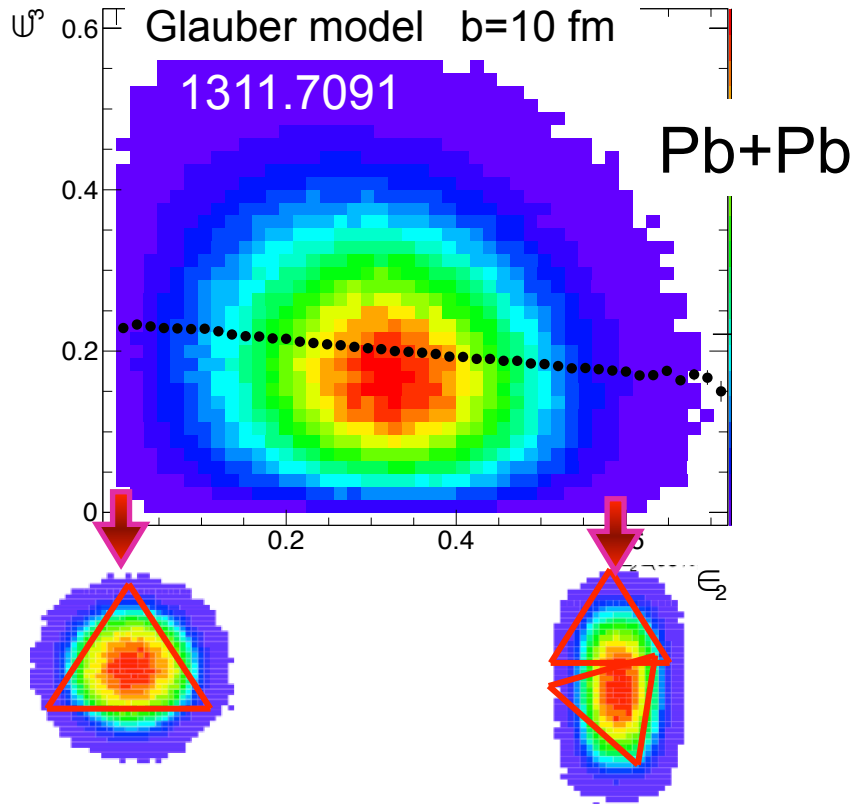
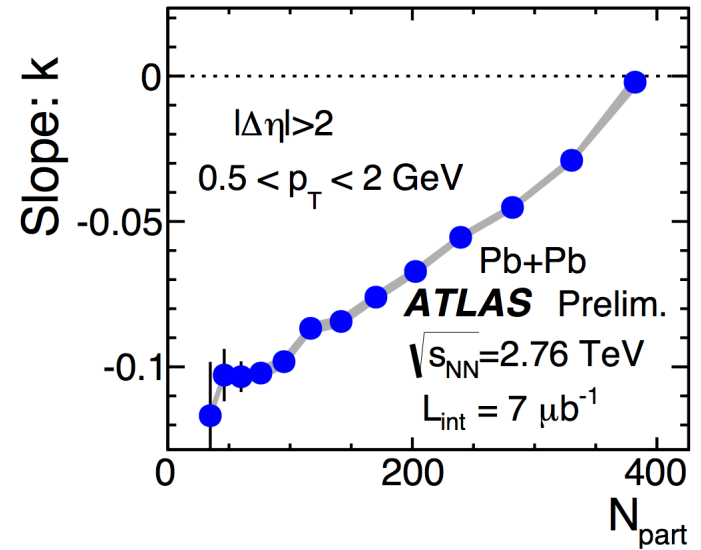
Compare well with expectation from EP correlations.
 higher-order mode-mixing negligible

Separate geometry and mode-mixing components!!

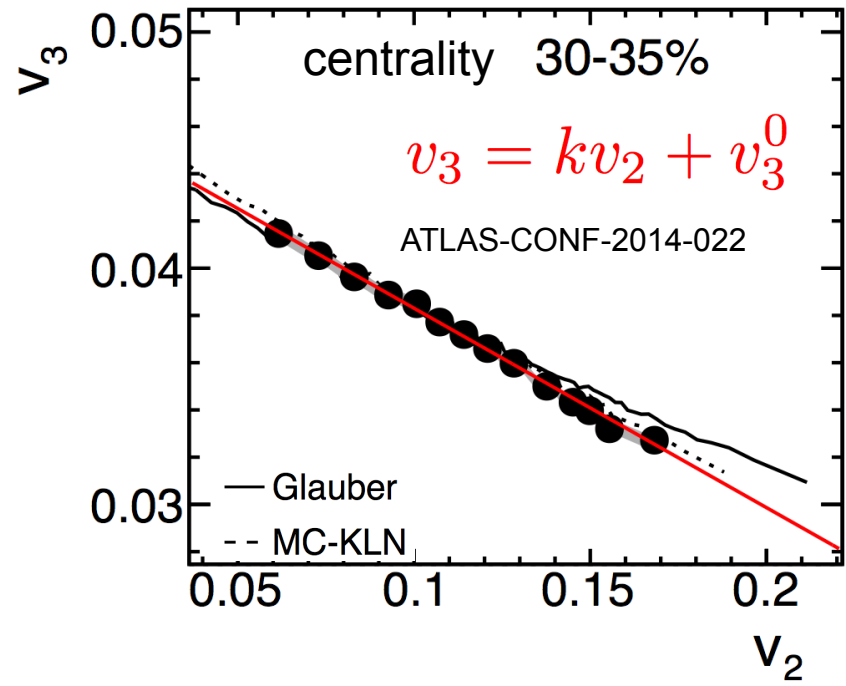
v_3-v_2 correlation from event-shape engineering 15



Very weak EP correlations



Direct constraint on initial geometry



■ Basar & Teaney's conformal scaling:

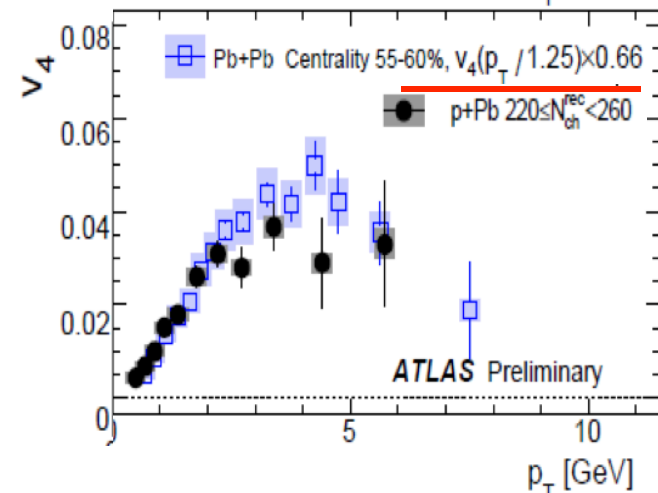
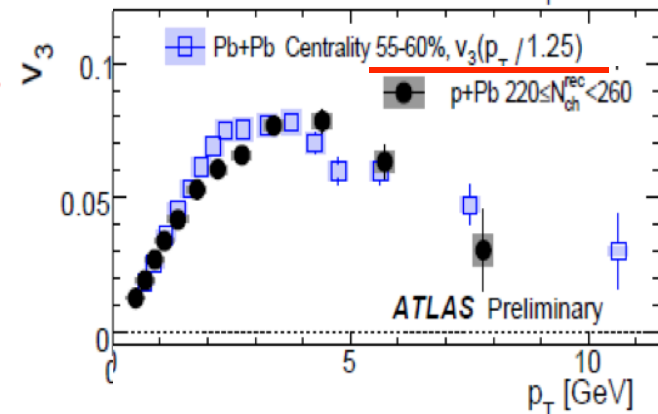
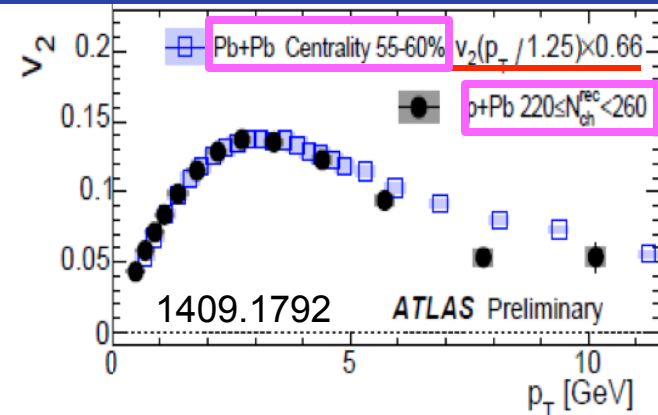
$$V_n(p_T)_{pA} \sim a v_n(p_T/k)_{AA} \text{ at same multiplicity}$$

i.e. collective response controlled multiplicity

- $\langle p_T \rangle_{pPb} = k \langle p_T \rangle_{PbPb}$, $k=1.25$
- $a \approx 0.66$ for $v_2 \rightarrow$ average elliptic geometry $V_2 = a v_2$
- $a \approx 1.0$ for $v_3 \rightarrow$ dominated by fluctuations. $V_3 \approx v_3$
- $a \approx 0.66$ for $v_4 \rightarrow$ mode-mixing $\propto v_2^2$. **but why also 0.66?**

$$\text{Pb+Pb: } v_4^2 = \left(v_4^L \right)^2 + \left(v_4^{NL} \right)^2 \quad \text{p+Pb: } V_4^2 = \left(V_4^L \right)^2 + \left(V_4^{NL} \right)^2$$

linear
Mode-mixing
linear
Mode-mixing



Basar & Teaney's conformal scaling:

$$V_n(p_T)_{pA} \sim a v_n(p_T/k)_{AA} \text{ at same multiplicity}$$

i.e. collective response controlled multiplicity

- $\langle p_T \rangle_{pPb} = k \langle p_T \rangle_{PbPb}$, $k=1.25$
- $a \approx 0.66$ for $v_2 \rightarrow$ average elliptic geometry $V_2 = a v_2$
- $a \approx 1.0$ for $v_3 \rightarrow$ dominated by fluctuations. $V_3 \approx v_3$
- $a \approx 0.66$ for $v_4 \rightarrow$ mode-mixing $\propto v_2^2$. **but why also 0.66?**

$$\begin{array}{ccc} \text{linear} & \text{Mode-mixing} & \\ \text{Pb+Pb: } v_4^2 = (v_4^L)^2 + (c v_2^2)^2 & & \text{P+Pb: } V_4^2 = (V_4^L)^2 + (c V_2^2)^2 \\ \text{linear} & \text{Mode-mixing} & \end{array}$$

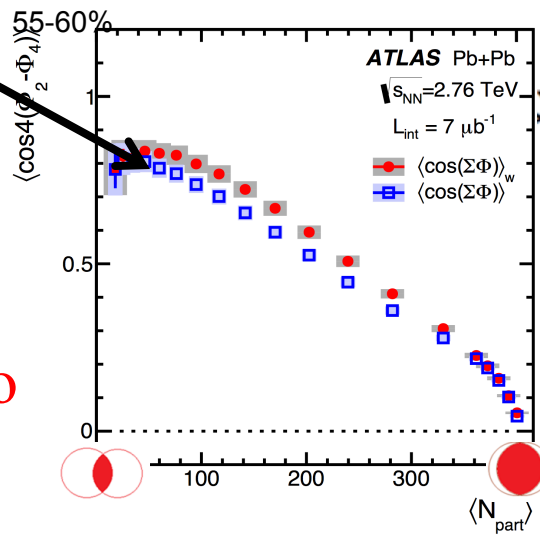
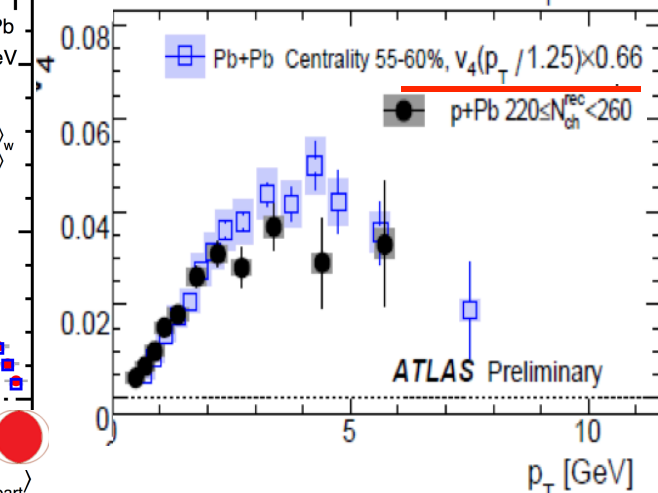
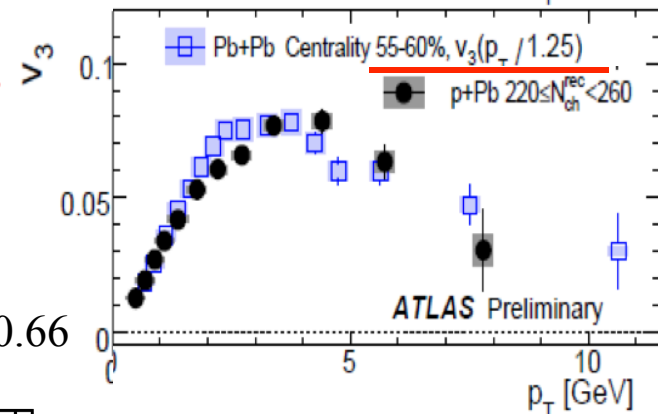
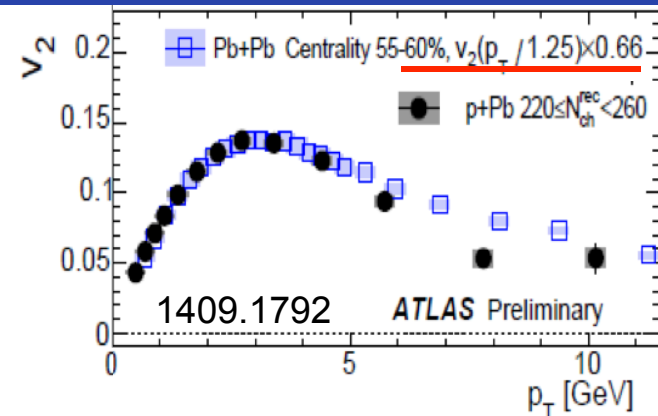
$$\text{Since: } c v_2^2 = b v_4 \quad b = 0.837, \quad v_4^L = \sqrt{1 - b^2} v_4 \quad V_4^L \approx v_4^L \quad V_2 = a v_2 \quad a = 0.66$$

$$\Rightarrow c V_2^2 \approx a^2 c v_2^2 = a^2 b v_4$$

$$\Rightarrow V_4^2 = [1 - b^2 + a^4 \times b^2] v_4^2$$

$$\Rightarrow V_4 \approx 0.66 v_4$$

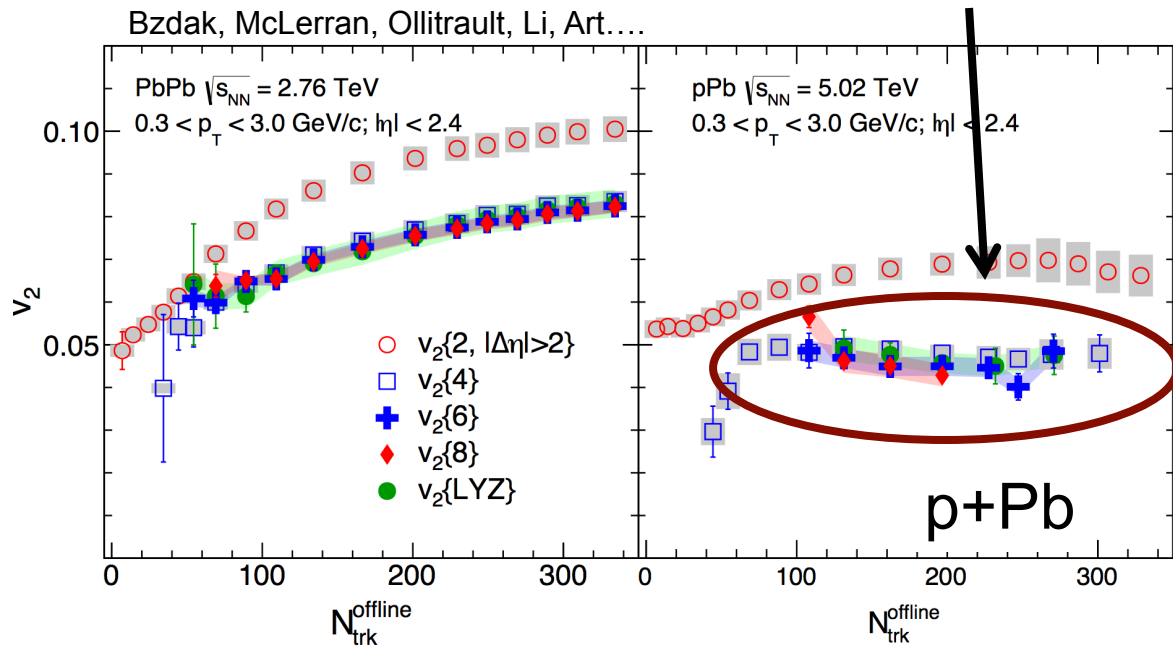
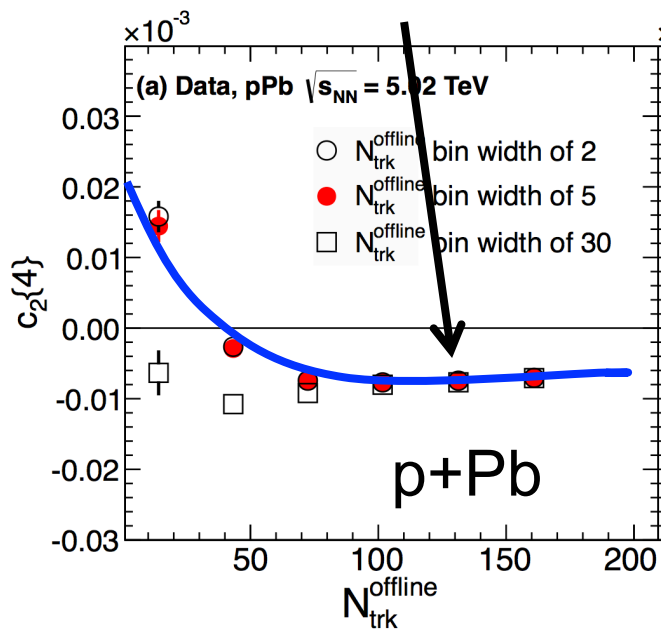
Why extrapolation of hydro prediction works so well?



A detour: The meaning of cumulants?

negative $c_2\{4\}$ or real $v_2\{4\}$
Onset of collectivity?

What $v_2\{2k\} \approx$ same means?
 \sim Gaussian?, \sim power?



$$a_n\{2\} \equiv c_n\{2\} = \langle v_n^2 \rangle$$

$$-a_n\{4\} \equiv c_n\{4\} = \langle v_n^4 \rangle - 2 \langle v_n^2 \rangle^2$$

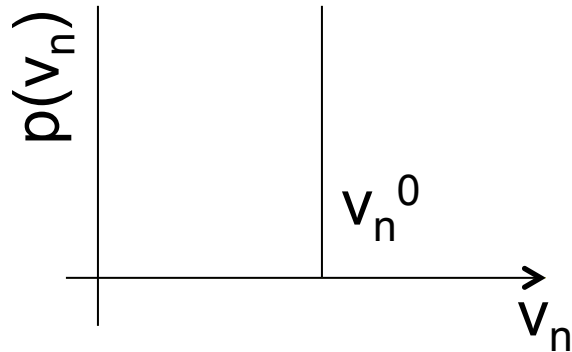
$$4a_n\{6\} \equiv c_n\{6\} = \langle v_n^6 \rangle - 9 \langle v_n^4 \rangle \langle v_n^2 \rangle + 12 \langle v_n^2 \rangle^3$$

$$-33a_n\{8\} \equiv c_n\{8\} = \langle v_n^8 \rangle - 16 \langle v_n^6 \rangle \langle v_n^2 \rangle - 18 \langle v_n^4 \rangle^2 + 144 \langle v_n^4 \rangle \langle v_n^2 \rangle^2 - 144 \langle v_n^2 \rangle^4$$

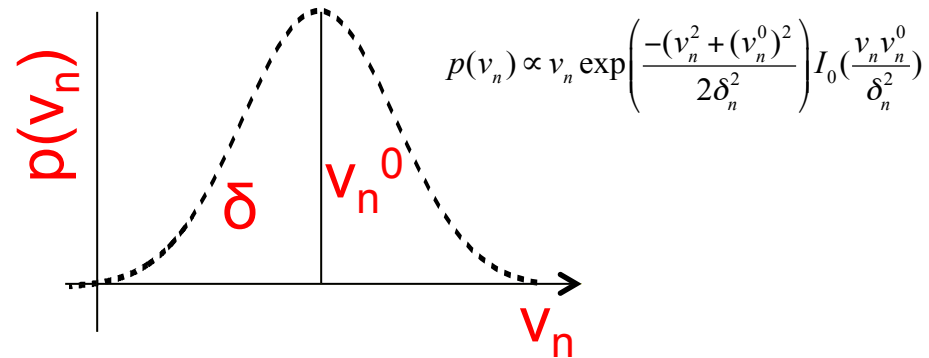
$$v_n\{2k\} \equiv \text{sgn}(a_n\{2k\}) \sqrt[2k]{|a_n\{2k\}|}$$

Cumulants ? $p(v_n)$

No fluctuation



Bessel-Gaussian fluctuation



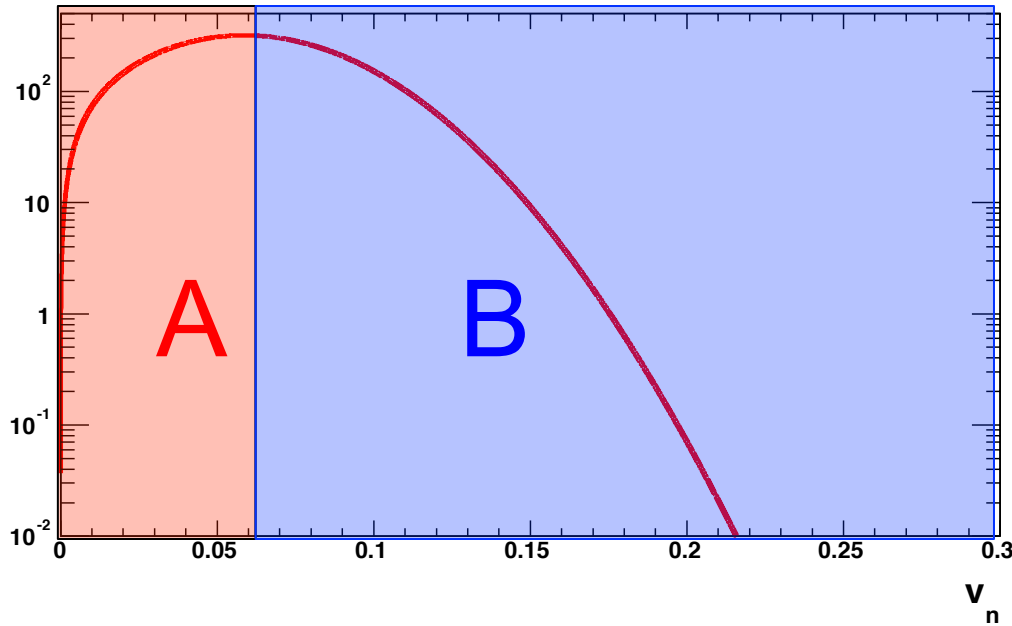
Same answer!:

$$c_n\{2\} = (v_n^0)^2 + 2\delta^2 \quad c_n\{4\} = -(v_n^0)^4$$

$$c_n\{6\} = 4(v_n^0)^6 \quad c_n\{8\} = -33(v_n^0)^8$$

$$v_n\{4\} = v_n\{6\} = v_n\{8\} = \dots = v_n^0$$

Can cumulants capture all details of $p(v_n)$?



$$p(v_n) \propto v_n \exp\left(\frac{-(v_n^2 + (v_n^0)^2)}{2\delta_n^2}\right) I_0\left(\frac{v_n v_n^0}{\delta_n^2}\right)$$

$v_n^0 = \delta\sqrt{2}$ as example

- Divide B-G distri. to 2 equal parts, and calculate cumulants separately.

| | $v_n\{2\}$ | $v_n\{4\}$ | $v_n\{6\}$ | $v_n\{8\}$ | In units of v_n^0 |
|-----|------------|--------------|--------------|--------------|---------------------|
| all | 1.414 | 1 | 1 | 1 | |
| A | 0.851 | <u>0.759</u> | <u>0.746</u> | <u>0.744</u> | |
| B | 1.809 | <u>1.690</u> | <u>1.701</u> | <u>1.701</u> | |

- $v\{2\}$ and $v\{4\} \rightarrow$ width and mean of $p(v)$, higher-order diff. very small
Higher-order cumulants not sensitive to $p(v_n)$ beyond mean and width?

Cumulants for narrow distribution

- Simple central moment expansion

$$\theta_j = \int \left(\frac{v_n - \langle v_n \rangle}{\langle v_n \rangle} \right)^j p(v_n) dv_n$$

$$\langle v_n^{2k} \rangle = \langle v_n \rangle^{2k} \int \left(1 + \frac{v_n - \langle v_n \rangle}{\langle v_n \rangle} \right)^{2k} p(v_n) dv_n = \langle v_n \rangle^{2k} \left(1 + \sum_{j=2}^{2k} C_{2k}^j \theta_j \right)$$

- Keep up to θ_4 :

$$v_n \{2\} / \langle v_n \rangle \approx 1 + \frac{1}{2} \theta_2 - \frac{3}{8} \theta_2^2$$

$$v_n \{4\} / \langle v_n \rangle \approx 1 - \frac{1}{2} \theta_2 - \theta_3 - \frac{1}{4} \theta_4 + \frac{1}{8} \theta_2^2$$

$$v_n \{6\} / \langle v_n \rangle \approx 1 - \frac{1}{2} \theta_2 - \frac{2}{3} \theta_3 + \frac{1}{4} \theta_4 - \frac{11}{8} \theta_2^2$$

$$v_n \{8\} / \langle v_n \rangle \approx 1 - \frac{1}{2} \theta_2 - \frac{7}{11} \theta_3 + \frac{31}{132} \theta_4 - \frac{117}{88} \theta_2^2$$

- Some useful relations (again only true for narrow distribution)

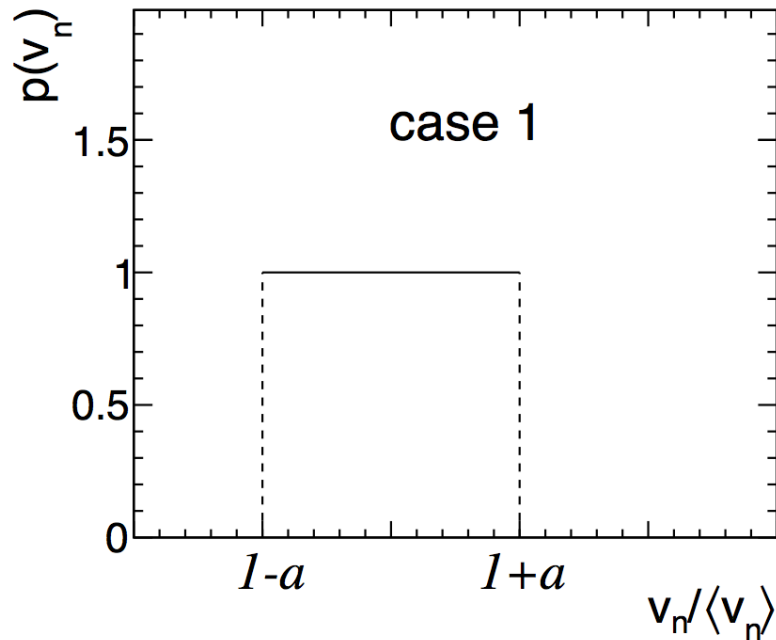
$$\frac{\sigma_n}{\langle v_n \rangle} \equiv \sqrt{\theta_2} \approx \sqrt{\frac{v_n^2 \{2\} - v_n^2 \{4\}}{v_n^2 \{2\} + v_n^2 \{4\}}}$$

Voloshin etc

$$v_n \{6\} - v_n \{4\} \approx \langle v_n \rangle \left(\frac{1}{3} \theta_3 + \frac{\theta_4 - 3\theta_2^2}{2} \right)$$

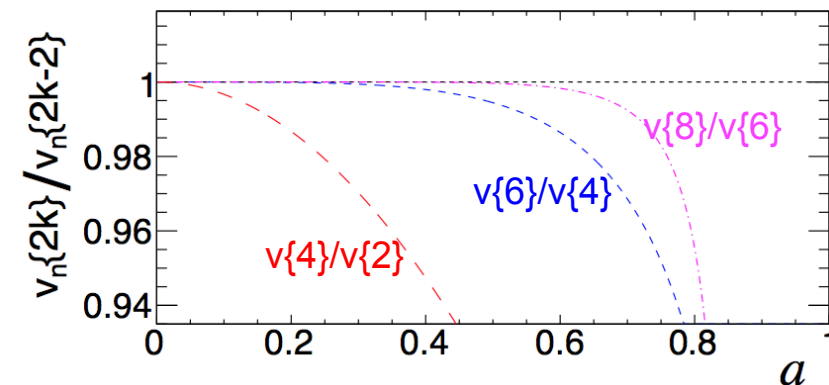
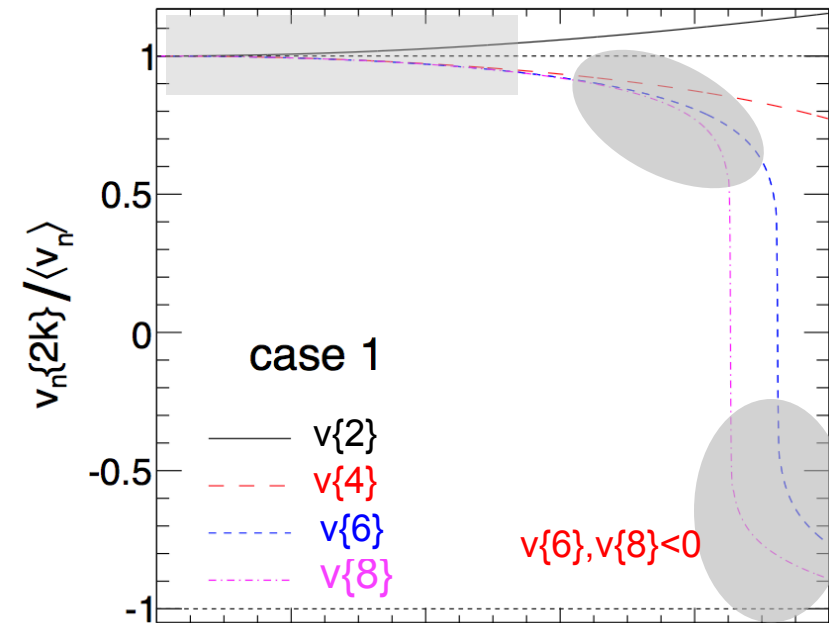
$$v_n \{8\} - v_n \{6\} \approx \langle v_n \rangle \left(\frac{1}{33} \theta_3 - \frac{\theta_4 - 3\theta_2^2}{66} \right)$$

Broad distribution - I



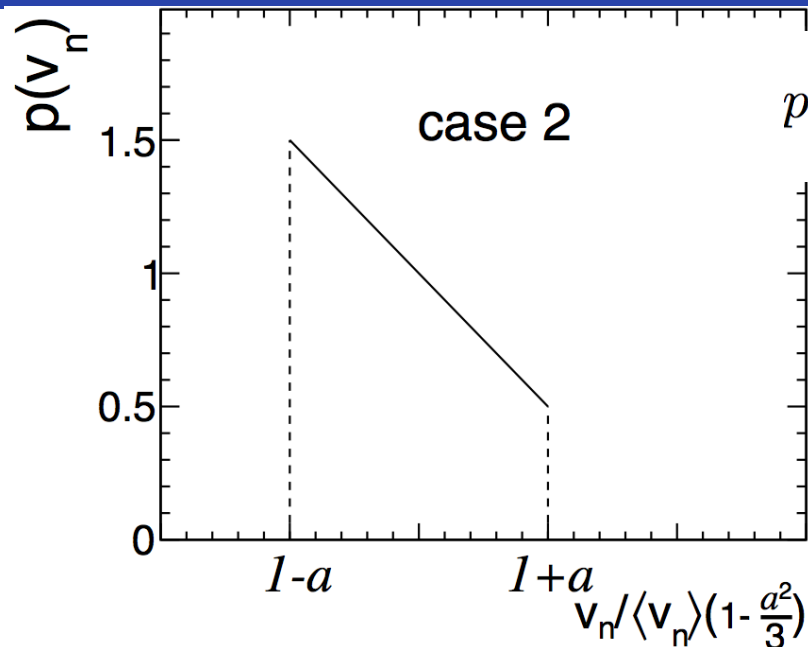
$$p(x; a) = \begin{cases} 1 & |x - 1| \leq a \\ 0 & |x - 1| > a \end{cases}, \quad x \equiv \frac{v_n}{\langle v_n \rangle}$$

$$v_n\{2k\} \equiv \text{sgn}(a_n\{2k\}) \sqrt[2k]{|a_n\{2k\}|}$$



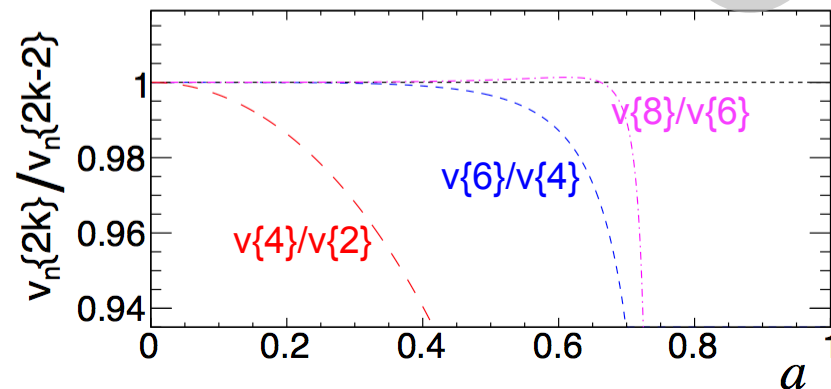
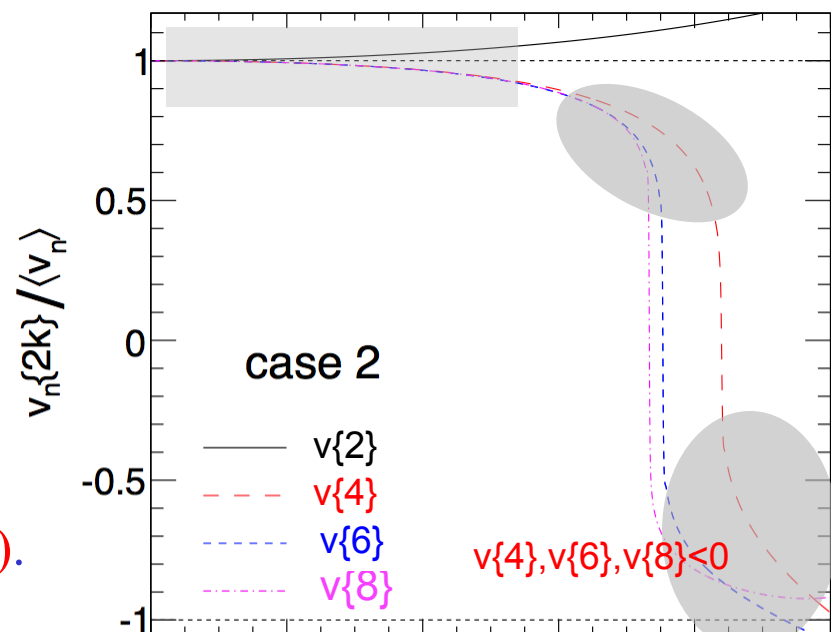
- Width controlled by a .
- $v\{2k\}$, $k > 1$ close to each other for $a < 0.5$. then deviates and flip sign ($v\{6\}, v\{8\} < 0$).

Broad distribution -II



$$p(x; a) = \begin{cases} 2 - x & |x - 1| \leq a \\ 0 & |x - 1| > a \end{cases}, \quad x \equiv \frac{v_n}{\langle v_n \rangle} \left(1 - \frac{a^2}{3}\right)$$

$$v_n\{2k\} \equiv \text{sgn}(a_n\{2k\}) \sqrt[2k]{|a_n\{2k\}|}$$



- Width controlled by a .
- $v\{2k\}$, $k > 1$ close to each other for $a < 0.5$. then deviates and flip sign ($v\{4\}, v\{6\}, v\{8\} < 0$).
- Unless have a-priori knowledge of its shape, i.e. close to BG, cumulants
 - Difficult to constrain $p(v)$ beyond mean and width
 - In principle can't use $c_n\{4\} < 0$ as sign of onset of collectivity

Summary

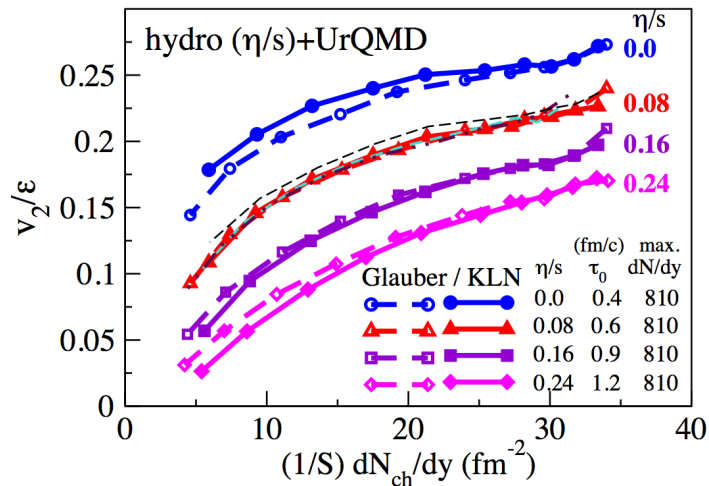
- Future measurements should provide direct physics insights
 - Obtain a complete picture at top \sqrt{s} , then apply to other systems and energies.
- Explore new observables sensitive to event-shape fluctuations:

$$p(v_n, v_m, \dots, \Phi_n, \Phi_m, \dots) = \frac{1}{N_{\text{evts}}} \frac{dN_{\text{evts}}}{dv_n dv_m \dots d\Phi_n d\Phi_m \dots}$$

- Augmented by the event-shape engineering technique
- Applications:
 - $p(v_2)$ is sensitive to non-linear response to $p(\varepsilon_2)$.
 - Event plane correlations and v_n - v_2 correlations allow quantitative separation of initial geometry and mode-mixing components
 - v_2 and v_3 are mainly controlled by ε_2 and ε_3
 - $v_4, v_5,$ and v_6 contains both geometry and mode-mixing components, can be separated by data-driven methods.
 - Novel anti-correlation between v_3 and v_2 reflects the ε_3 - ε_2 anti-correlation
 - Taking mode-mixing into account, conformal scaling explains the similarity of v_4 in pA and AA (in addition to v_2 and v_3)
- Cumulants not very sensitive to the $p(v_n)$ beyond its mean and width.

Backup

Eccentricity scaling: v_n/ϵ_n



Teaney and Yan

Viscous correlation

Linear:

$$-\frac{\Delta w_n}{w_n^{\text{id}}} \sim n^2 \frac{\eta}{s}$$

Viscous damping expected to be larger for larger n

