

# Transverse momentum fluctuation in ultra-central Pb+Pb collision

Rupam Samanta

AGH University of Krakow, Poland

in collaboration with Jean-Yves Ollitrault, Matthew Luzum, Jiangyong Jia,  
Somadutta Bhatta, Joao-Paulo Pichhetti

...based on arXiv:2303.15323 and arXiv:2306.09294



BIALASOWKA HEP Seminar

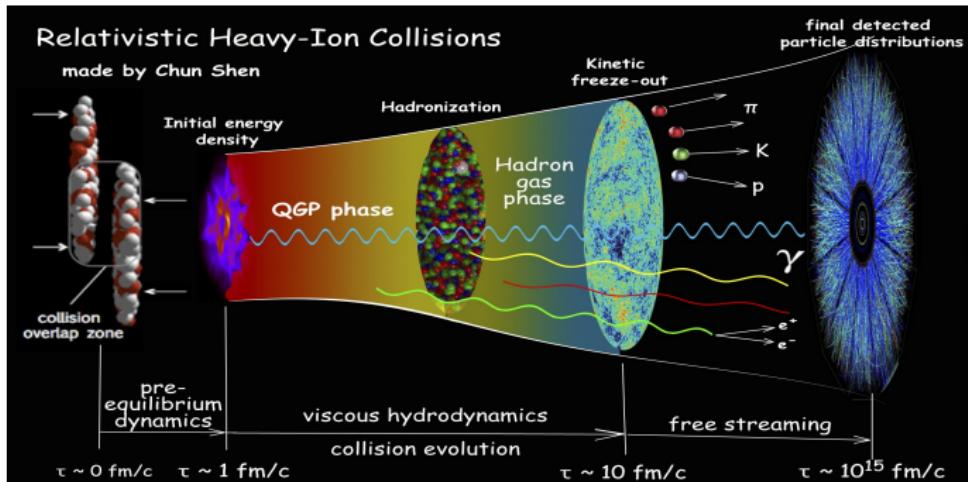
Krakow, December 1, 2023



AGH UNIVERSITY OF SCIENCE  
AND TECHNOLOGY



# High energy heavy ion(HI) collision: "The Little Bang"



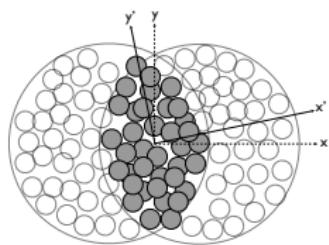
Shen, Heinz, arXiv:1507.01558



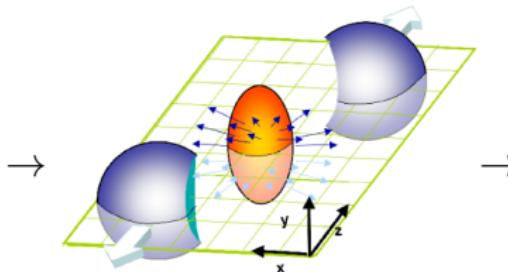
Boiling water :  $10^2$  K   Core of the Sun :  $10^7$  K

QGP :  $10^{12}$  K !!

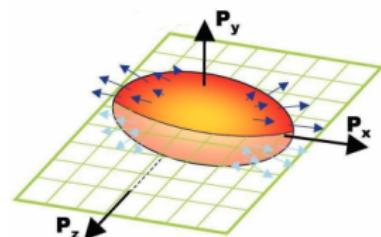
## Evidence of QGP : Collective flow in HI Collisions



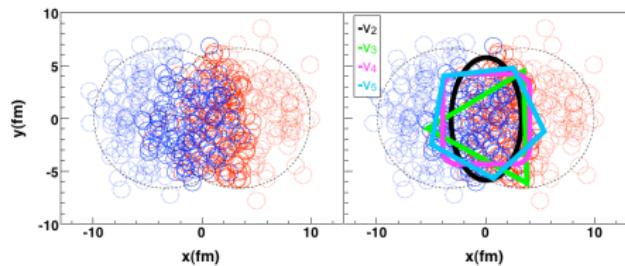
PHOBOS arXiv:0711.3724



U. Heinz, arXiv:0810.5529



BNL: RHIC



Alver, Baker, Loizides, Steinberg arXiv:0805.4411

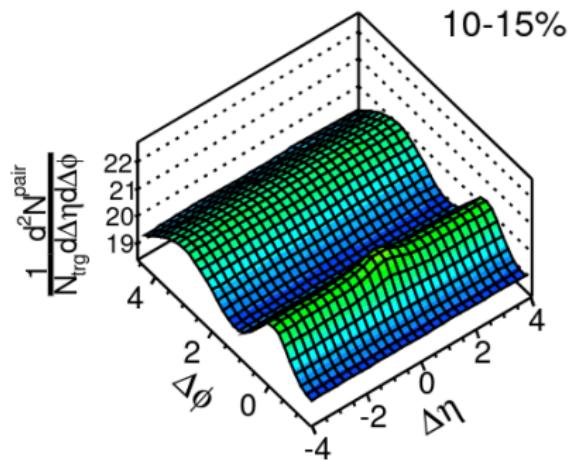
### Momentum anisotropy : Harmonic flow

$$\frac{dN}{dp_T d\phi} = \frac{dN}{2\pi dp_T} \left( 1 + 2 \sum_{n=1}^{\infty} v_n(p_T) \cos [n(\phi - \Psi_n)] \right)$$

$V_2 \rightarrow$  Elliptic flow &  $V_3 \rightarrow$  Triangular flow

## Experimental evidence !

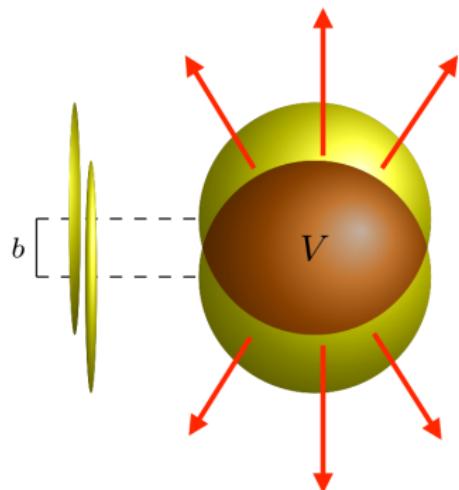
- Experimental evidence for the formation of a **little fluid** in Pb+Pb collision → **azimuthal correlations between particles** seen in detectors.



CMS:1201.3158

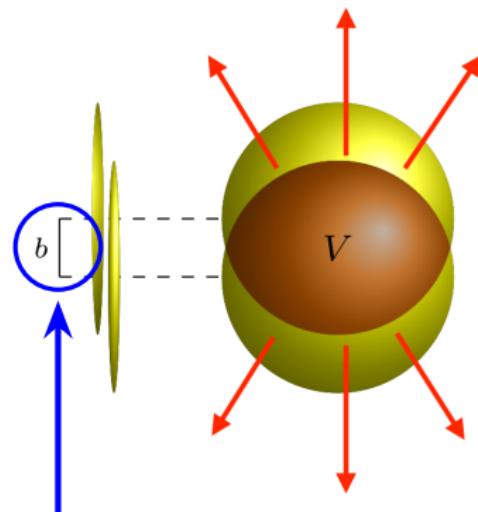
## Experimental evidence !

- Experimental evidence for the formation of a **little fluid** in Pb+Pb collision → **azimuthal correlations between particles** seen in detectors.
- Evidence is **indirect** ! → azimuthal distribution of particles is **not isotropic** → **anisotropy driven by pressure gradients** within a fluid.



## Experimental evidence !

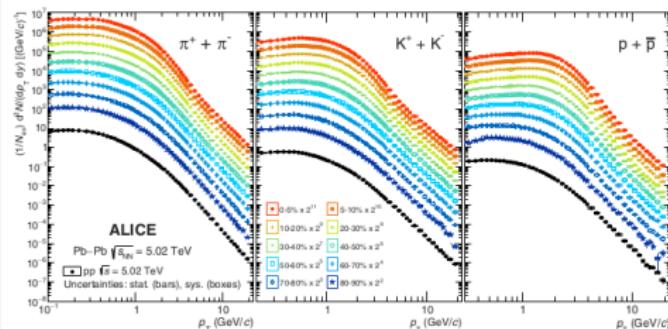
- Experimental evidence for the formation of a **little fluid** in Pb+Pb collision → azimuthal correlations between particles seen in detectors.
- Evidence is **indirect** ! → azimuthal distribution of particles is not isotropic → anisotropy driven by pressure gradients within a fluid.
- We report **more direct evidence** of local thermalization in Pb+Pb collisions → does not involve directions of outgoing particles, but solely their momenta.



**$b = \text{impact parameter}$**   
**(important in this talk ! )**

# Multiplicity ( $N_{ch}$ ) and Transverse momentum per particle ( $[p_t]$ )

- Charged particle spectra;  $\frac{dN}{dp_t}$  vs  $p_t$  observed in experiments.



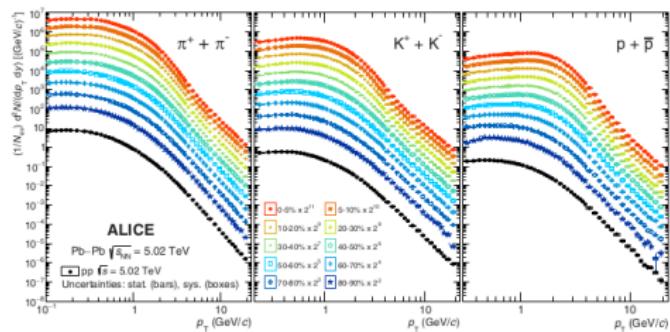
$p_t$  spectra , arXiv:1910.07678

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- Total number of charged particle (multiplicity),

$$N_{ch} = \int_{p_{min}}^{p_{max}} \frac{dN}{dp_t} dp_t$$



$p_t$  spectra , arXiv:1910.07678

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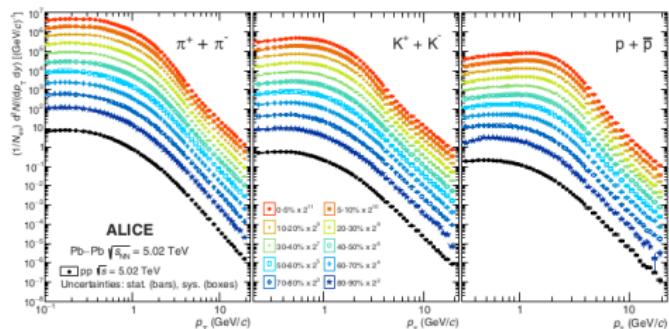
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- Total number of charged particle (multiplicity),

$$N_{ch} = \int_{p_{min}}^{p_{max}} \frac{dN}{dp_t} dp_t$$

- Average transverse momentum per particle,

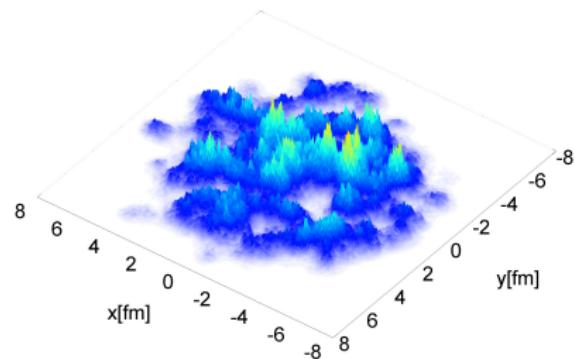
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## Fluctuation in HI collision

- Event-by-event fluctuation of initial state.

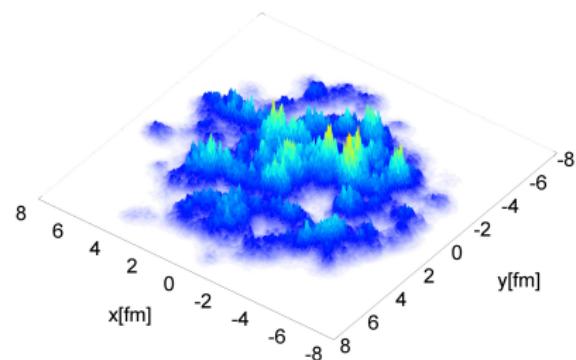


Lumpy structure of the initial density

Schenke, Tribedy, Venugopalan arXiv: 1206.6805

## Fluctuation in HI collision

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- All final state collective observables,  $N_{ch}$ ,  $[p_t]$ ,  $V_n$  fluctuates event-by-event.

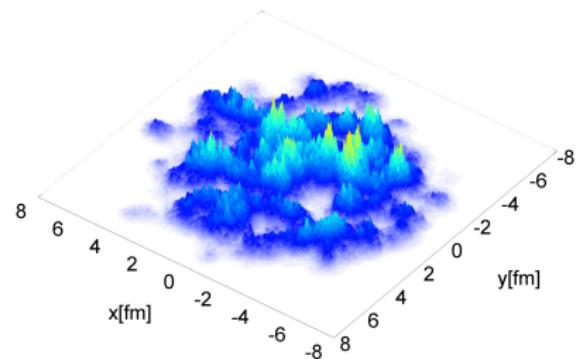


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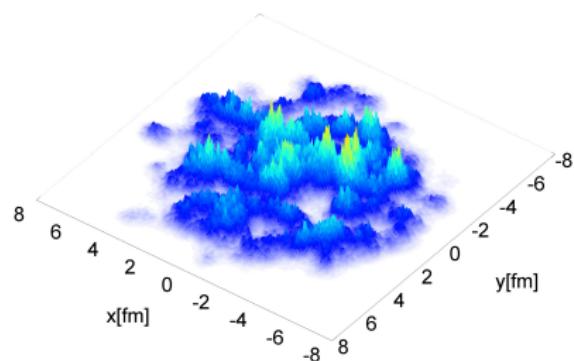


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- Could be a combination of **classical or geometrical** ( $b$  fluctuation), **quantum or intrinsic** (at fixed  $b$ ) and **statistical fluctuation**.
- In this talk :  $[p_t]$ -fluctuation  $\Rightarrow$  separating geometrical and intrinsic fluctuation !

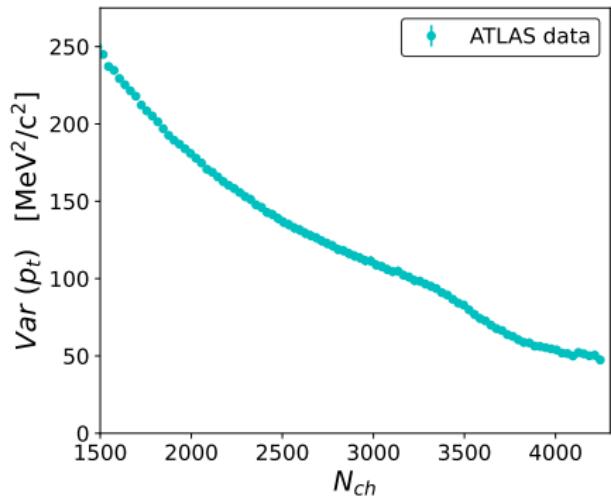


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## ATLAS data for $[p_t]$ fluctuation

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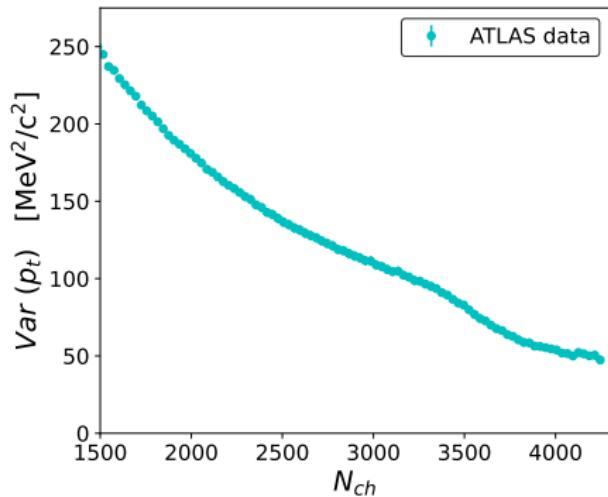
Variance of  $[p_t]$  for Pb+Pb @ 5.02 TeV

PhysRevC.107.054910

Table 374 in <https://www.hepdata.net/record/ins2075412>

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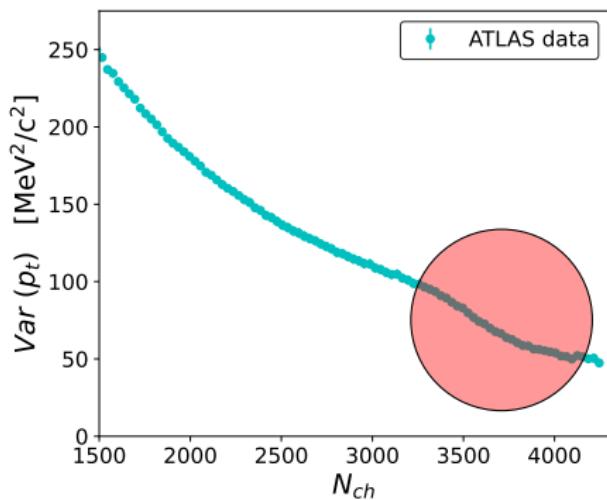
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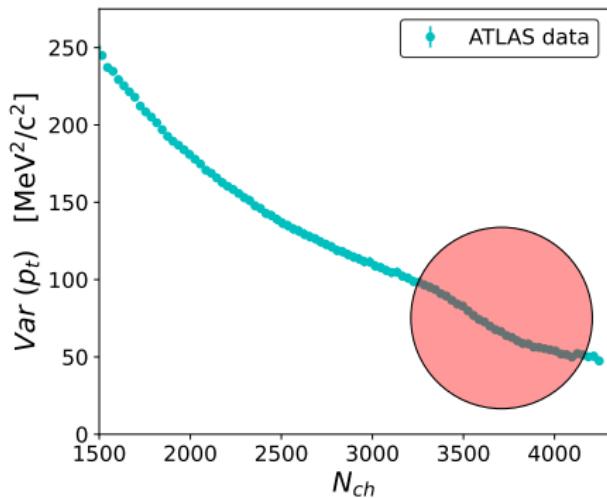
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- The relative dynamical fluctuation of  $[p_t]$  is very small  $\sim 1\%$
- Puzzling behavior in ATLAS data : steep decrease over a narrow range of  $N_{ch}$
- We will show that this could be a consequence of thermalization !

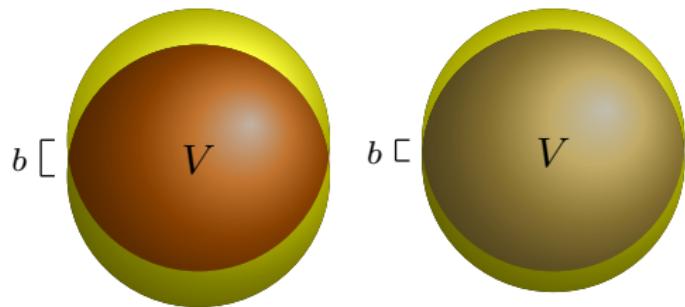


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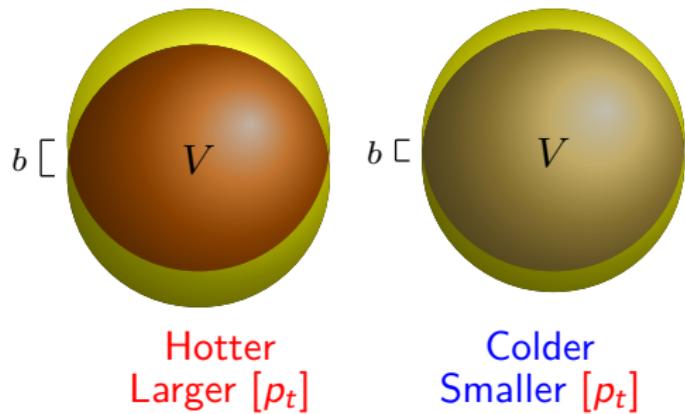
## Impact parameter ( $b$ ) is important !

- In experiment  $b$  is not known !  $\Rightarrow [p_t]$  fluctuation is measured for fixed  $N_{ch}$



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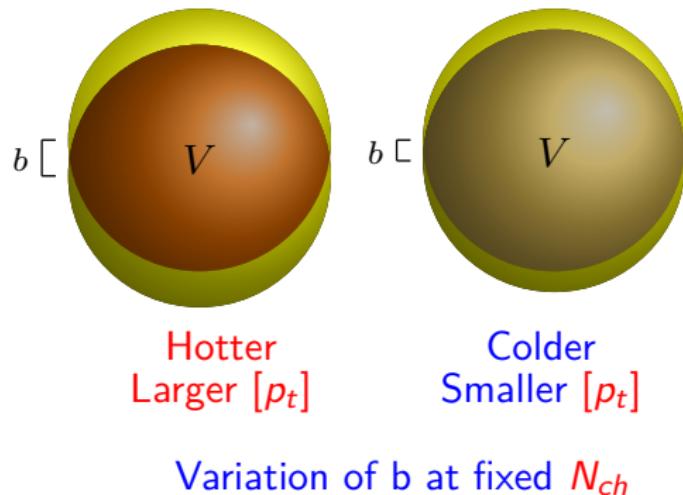
- In experiment  $b$  is not known !  $\Rightarrow [p_t]$  fluctuation is measured for fixed  $N_{ch}$
- Fixed  $N_{ch}$   $\Rightarrow$  finite range of  $b$  !



Variation of  $b$  at fixed  $N_{ch}$

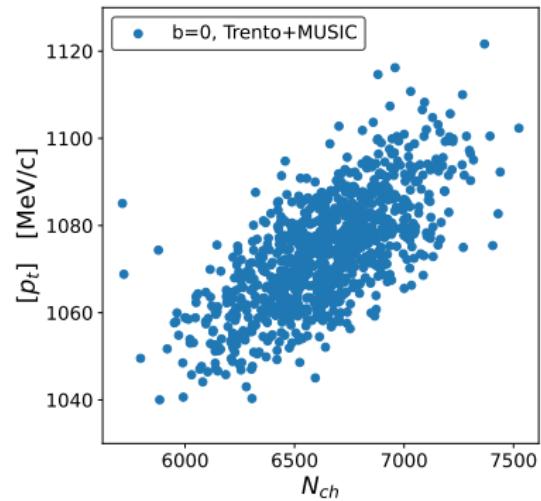
## Impact parameter ( $b$ ) is important !

- In experiment  $b$  is not known !  $\Rightarrow [p_t]$  fluctuation is measured for fixed  $N_{ch}$
- Fixed  $N_{ch} \Rightarrow$  finite range of  $b$  !
- Variation of  $b$  gives a contribution to the variation of  $[p_t] \Rightarrow$  goes to 0 in ultracentral collisions !



## Hydrodynamic simulation: $b$ is known !

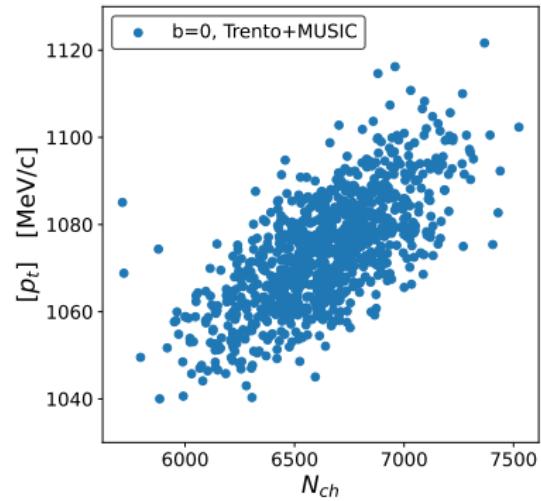
- Hydro : assumes thermalization !  
⇒ We simulate Pb+Pb collisions at fixed  $b$  ( $=0$ ) with TRENTO (initial condition) + MUSIC (hydro)



Pb+Pb @ 5.02 TeV

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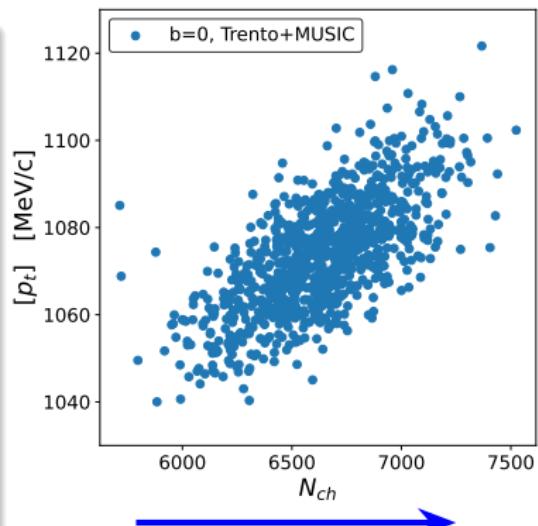
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- ▶ Significant fluctuation of  $N_{ch}$  and modest fluctuation of  $[p_t]$ . Strong correlation between  $[p_t]$  and  $N_{ch}$



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- ▶ Fixed  $b \Rightarrow$  fixed collision volume  
Larger  $N_{ch} \Rightarrow$  larger density  
⇒ larger temperature  
⇒ larger energy per particle  
⇒ larger  $[p_t]$

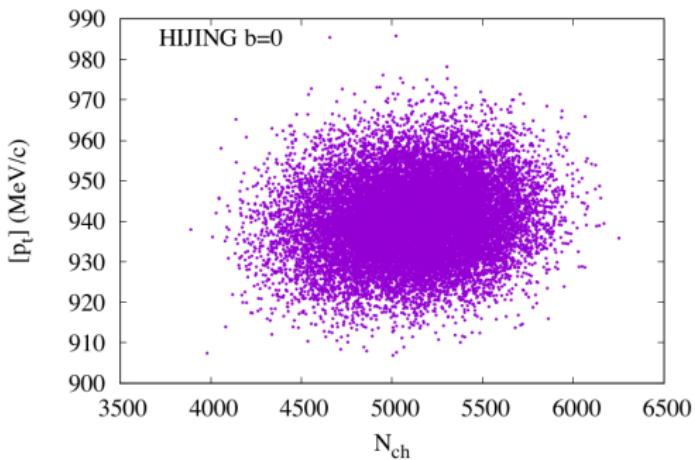


density increases  
⇒ temperature increases

## Comparing other models : HIJING simulation

Wang, Gyulassy, arXiv:nucl-th/9502021

- HIJING: microscopic model of HI collision  $\Rightarrow$  the system doesn't thermalize !

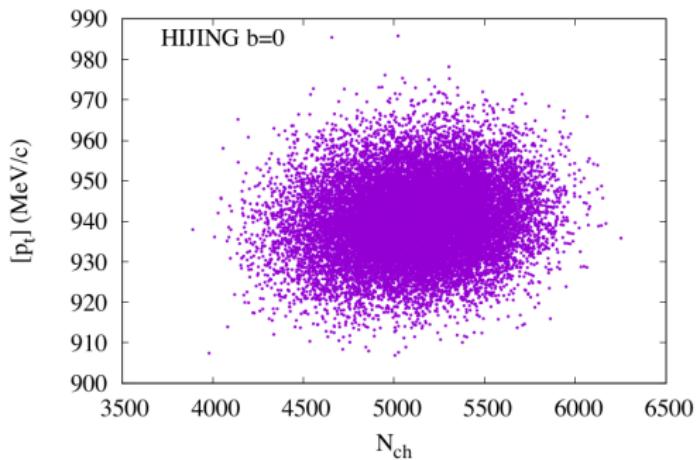


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- Very small correlation between  $N_{ch}$  and  $[p_t] \sim 10 \times$  smaller !!



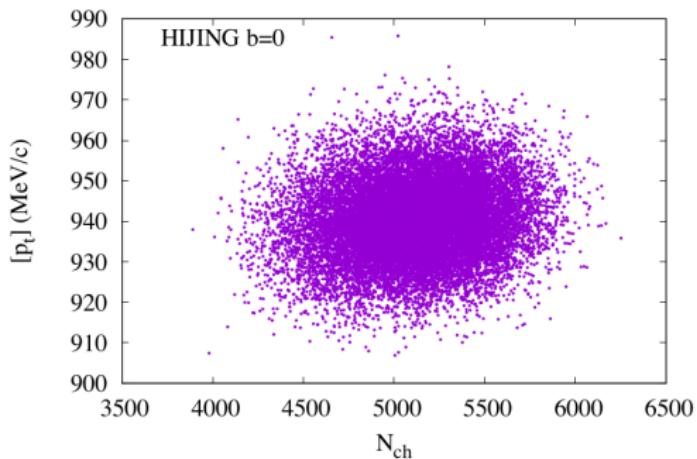
Pb+Pb @ 5.02 TeV

No thermalization  
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- HIJING: microscopic model of HI collision  $\Rightarrow$  the system doesn't thermalize !
- Very small correlation between  $N_{ch}$  and  $[p_t] \sim 10 \times$  smaller !!
- Hence the correlation could be a signature of thermalization !

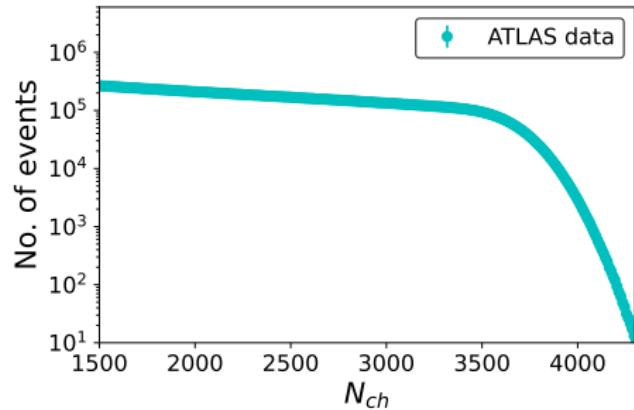


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## Bayesian reconstruction of $P(\mathbf{b} | N_{ch})$

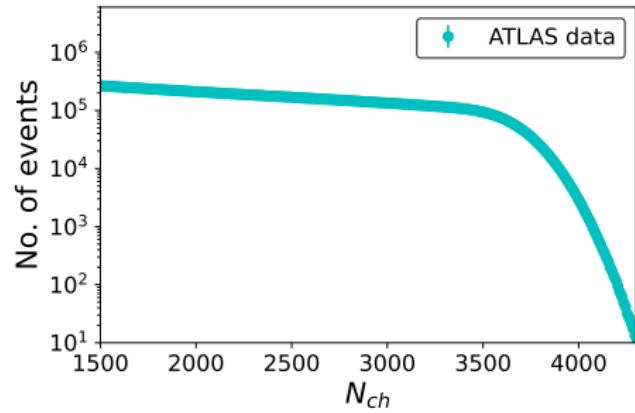
- First we solve the inverse problem:  
what is the distribution of  $N_{ch}$  at fixed  $\mathbf{b}$  i.e.  $P(N_{ch}|\mathbf{b})$  ?



$N_{ch}$  distribution  
for centrality classification !

## Bayesian reconstruction of $P(\mathbf{b} | N_{ch})$

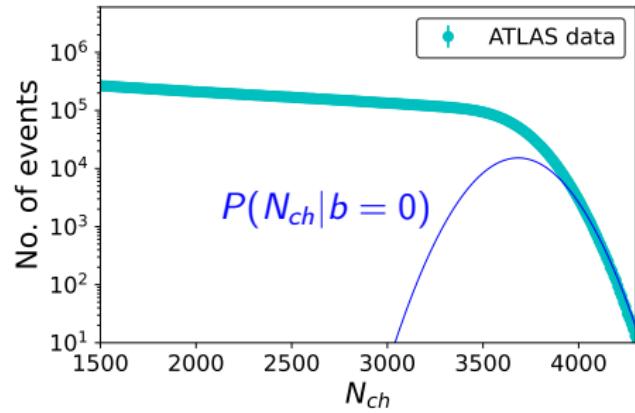
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$$P(\mathbf{b} | N_{ch}) P(N_{ch}) = P(N_{ch}|\mathbf{b}) P(\mathbf{b})$$



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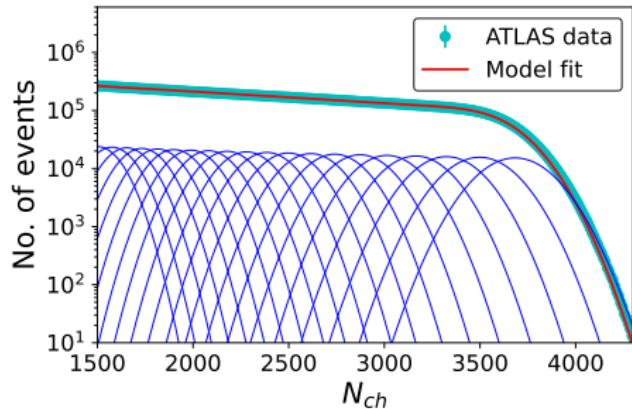
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$$P(\mathbf{b} | N_{ch}) P(N_{ch}) = P(N_{ch}|\mathbf{b}) P(\mathbf{b})$$
- We assume  $P(N_{ch}|\mathbf{b})$  to be Gaussian !



$N_{ch}$  distribution at fixed b  
Gaussian assumption !

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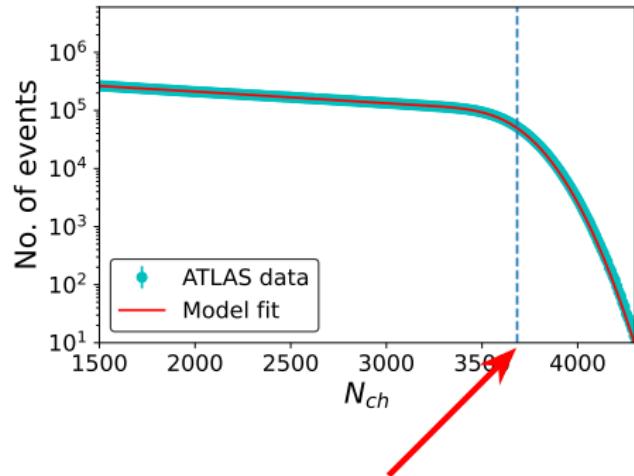


Sum of Gaussians at fixed  $\mathbf{b}$

Das, Giacalone, Monard, Ollitrault  
arXiv:1708.00081

## Bayesian reconstruction of $P(\mathbf{b} | N_{ch})$

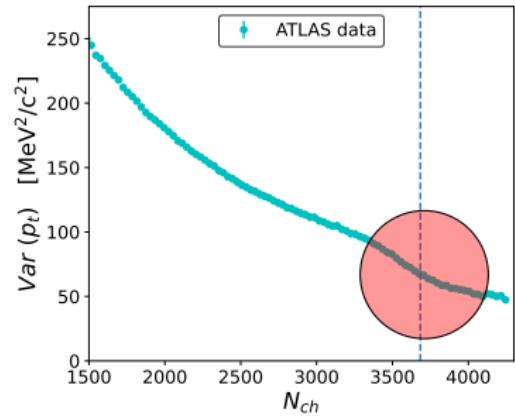
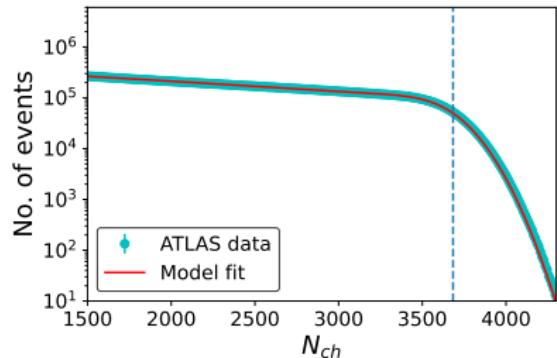
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- Fit  $P(N_{ch})$  as sum of Gaussians
- We precisely reconstruct the knee (mean  $N_{ch}$  at  $\mathbf{b}=0$ )



Precise construction of knee  
 $\langle N_{ch} | \mathbf{b} = 0 \rangle$

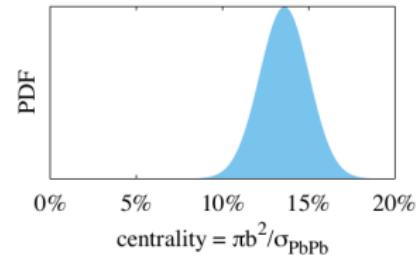
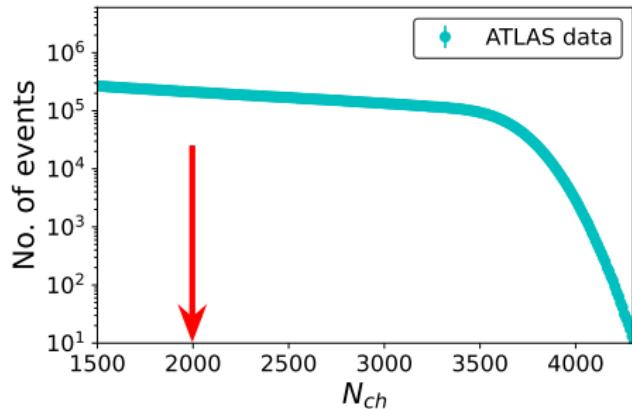
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- We assume  $P(N_{ch}|\mathbf{b})$  to be **Gaussian !**
- Fit  $P(N_{ch})$  as **sum of Gaussians**
- We precisely reconstruct the **knee** (mean  $N_{ch}$  at  $\mathbf{b}=0$ )
- The **steep fall** of the variance precisely **occur at the knee !**



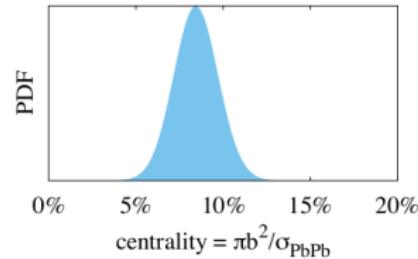
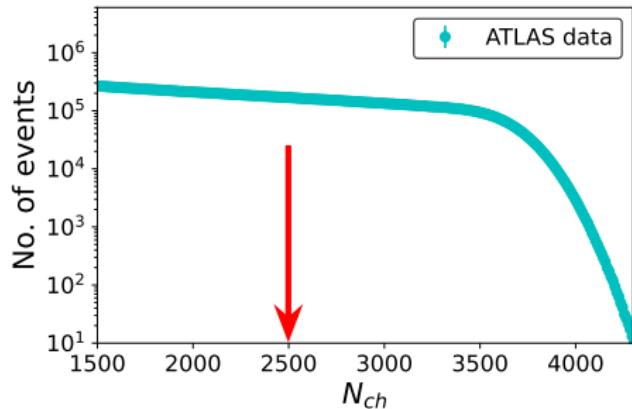
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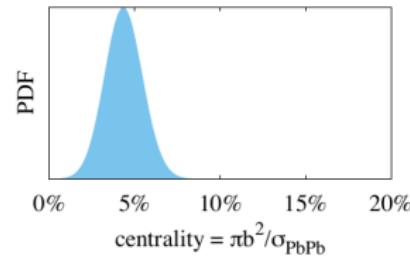
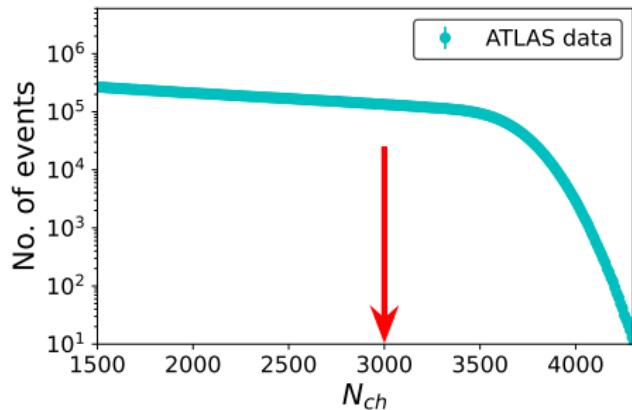
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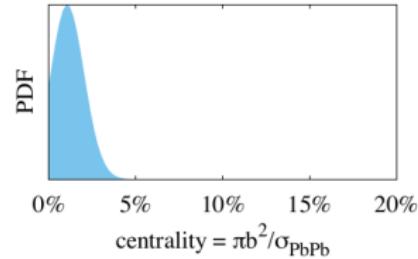
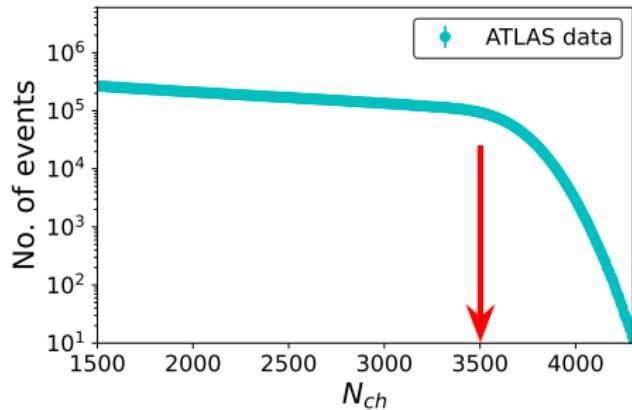
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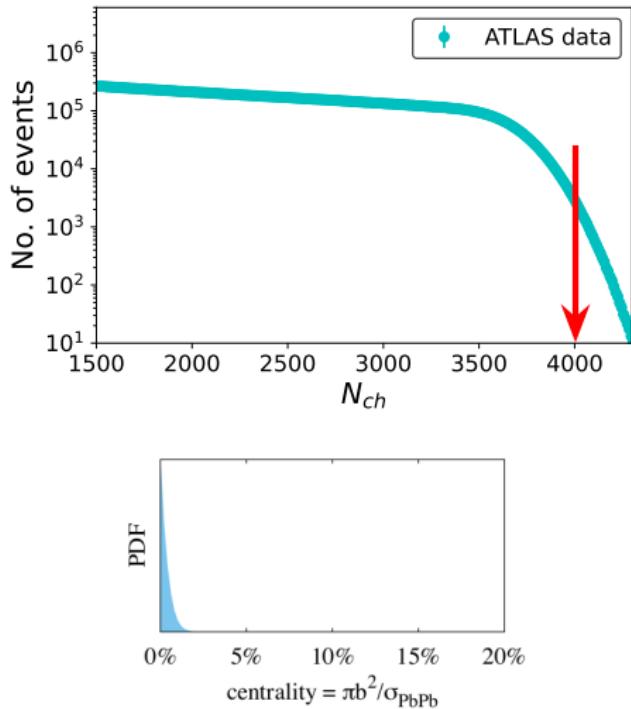
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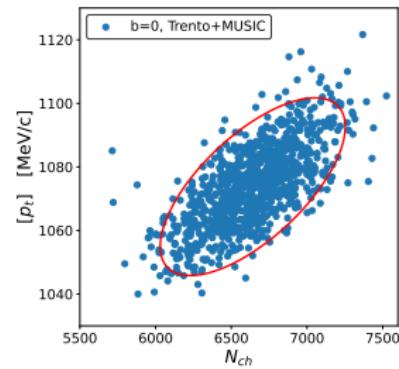
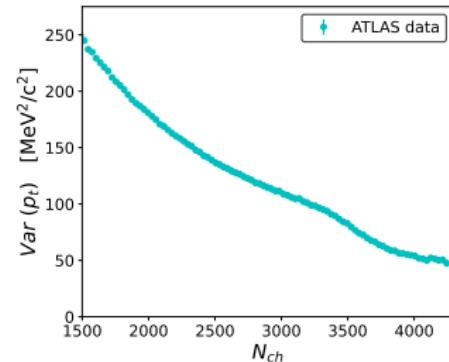
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- Above the knee it gets extremely truncated  $\Rightarrow$  the impact parameter fluctuation gradually disappears !



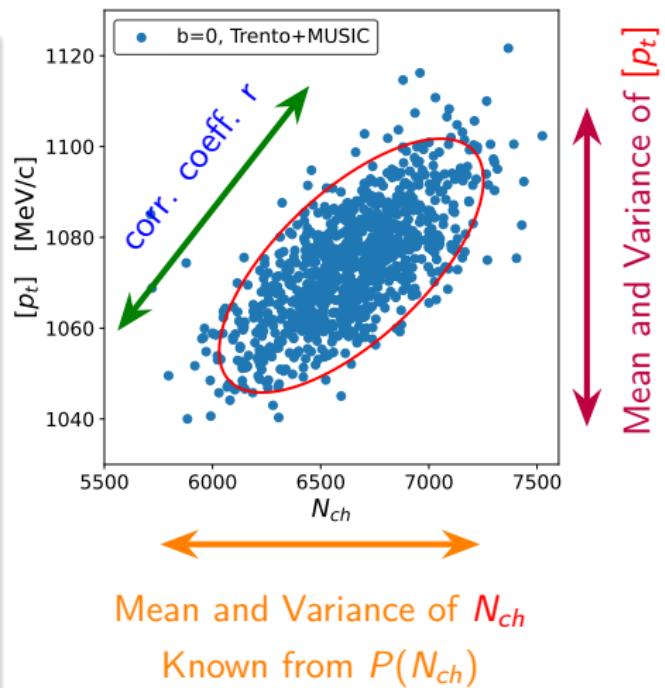
## Understanding $[p_t]$ fluctuation data : Parametrizing $P(N_{ch}, [p_t] | b)$

- We assume a simple 2D correlated Gaussian between  $[p_t]$  and  $N_{ch}$  at fixed impact parameter  $b$  :  $P([p_t], N_{ch} | b)$ .



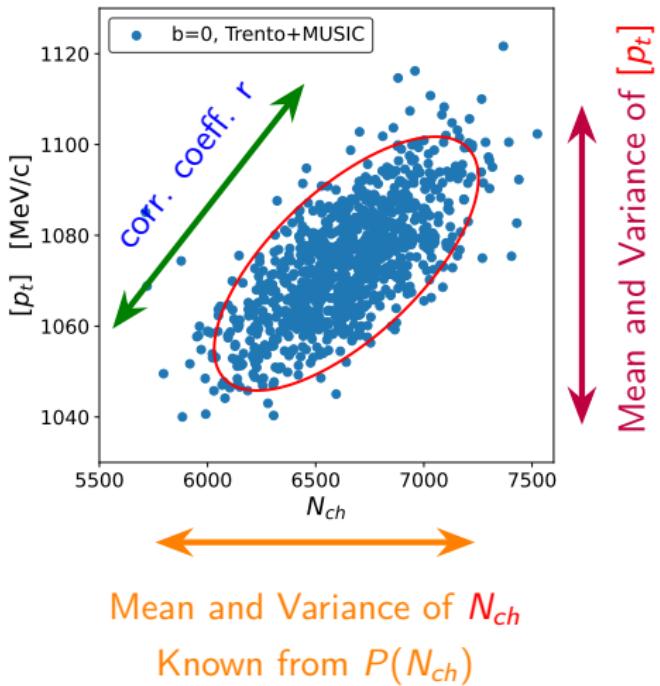
## Understanding $[p_t]$ fluctuation data : Parametrizing $P(N_{ch}, [p_t] | b)$

- ▶ We assume a simple 2D correlated Gaussian between  $[p_t]$  and  $N_{ch}$  at fixed impact parameter  $b$  :  $P([p_t], N_{ch} | b)$ .
- ▶ The distribution has 5 parameters : Mean and variance of  $N_{ch}$ , Mean and variance of  $[p_t]$  and correlation coefficient  $r$  between  $N_{ch}$  and  $[p_t]$ .



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- ▶ Mean value of  $[p_t]$  is constant at fixed  $b$  and assuming it is independent of  $b \implies$  we fit  $P(\delta p_t, N_{ch} | b)$   
 $\delta p_t = [p_t] - \langle [p_t] \rangle$



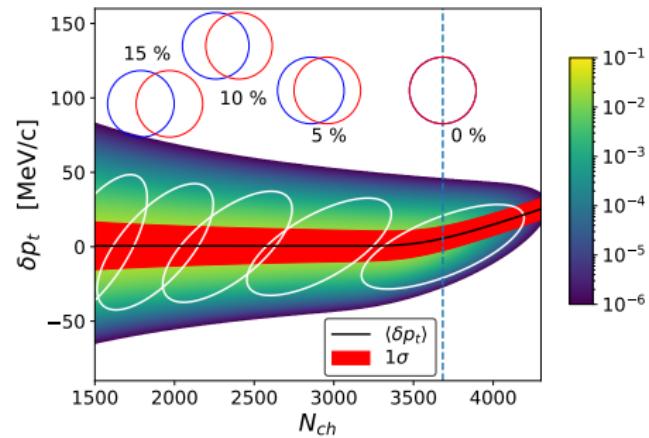
## b-dependence of the fit parameters

- We assume mean  $[p_t]$  to be independent of  $b$
- We assume  $\text{Var}([p_t])$  is a smooth function of mean multiplicity :

$$\sigma p_t^2 \left( \frac{\langle N_{ch}(0) \rangle}{\langle N_{ch}(b) \rangle} \right)$$

- We also assume  $r$  to be independent of  $b$  for simplicity

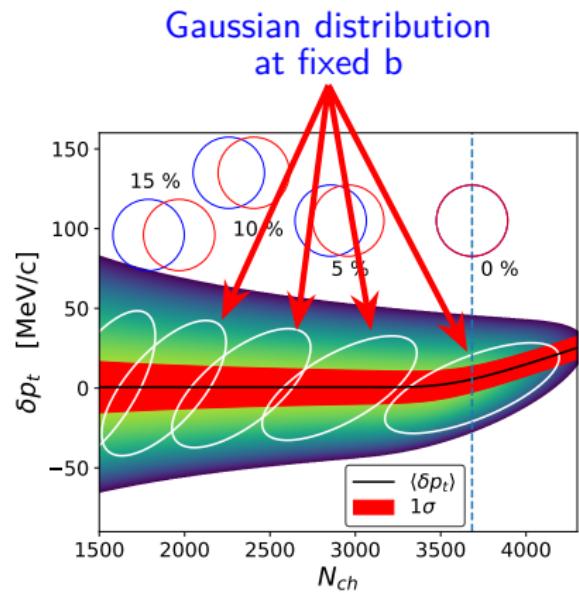
Fit result :  $P(N_{ch}, \delta p_t)$



2D correlated gaussian  
distribution of  $\delta p_t$  and  $N_{ch}$

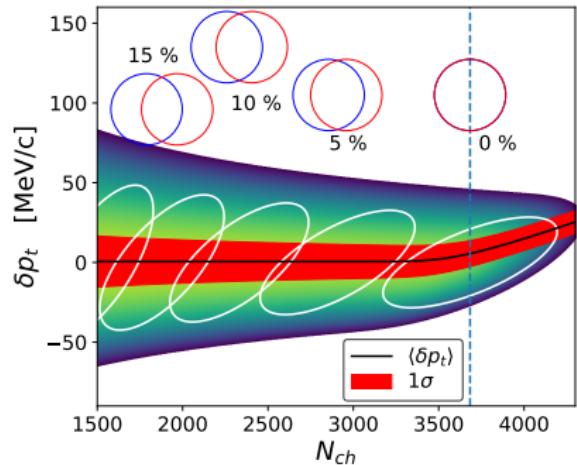
**Fit result :  $P(N_{ch}, \delta p_t)$**

- We get,  $P(N_{ch}, \delta p_t) = \int P(N_{ch}, \delta p_t | b) P(b) db$



**Fit result :**  $P(N_{ch}, \delta p_t)$

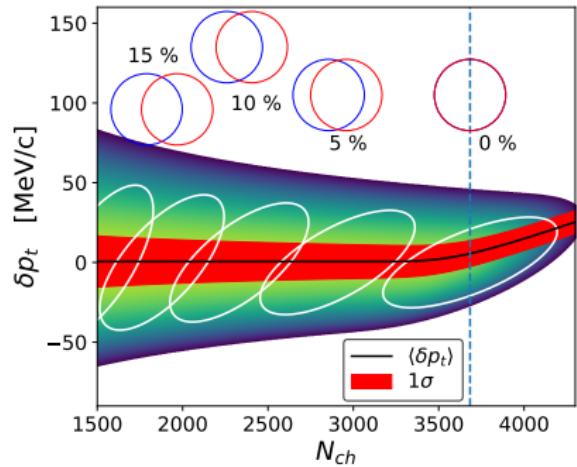
- We get,  $P(N_{ch}, \delta p_t)$   
 $= \int P(N_{ch}, \delta p_t | b)P(b)db$
  - By conditional probability  
 $P(\delta p_t | N_{ch}) = \frac{P(N_{ch}, \delta p_t)}{P(N_{ch})}$   
 $\implies \text{Var}([\delta p_t] | N_{ch})$  is the squared width of  $P(\delta p_t | N_{ch})$



## 2D correlated gaussian distribution of $\delta p_t$ and $N_{ch}$

**Fit result :**  $P(N_{ch}, \delta p_t)$

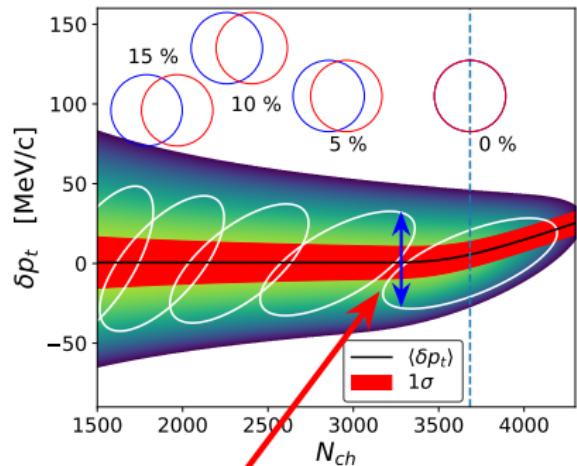
- We get,  $P(N_{ch}, \delta p_t) = \int P(N_{ch}, \delta p_t | b) P(b) db$
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 $P(\delta p_t | N_{ch}) = \frac{P(N_{ch}, \delta p_t)}{P(N_{ch})}$   
 $\implies \text{Var}([\delta p_t] | N_{ch})$  is the squared width of  $P(\delta p_t | N_{ch})$
  - The width of  $[p_t]$  fluctuation has two contributions :



## 2D correlated gaussian distribution of $\delta p_t$ and $N_{ch}$

**Fit result :**  $P(N_{ch}, \delta p_t)$

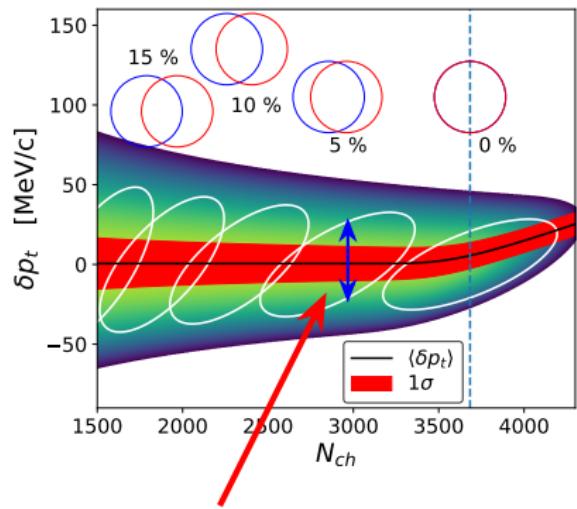
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⇒ Var( $\delta p_t | N_{ch}$ ) is the squared width of  $P(\delta p_t | N_{ch})$
  - The width of  $\delta p_t$  fluctuation has two contributions :
    - ① due to fluctuation of impact parameter b



fluctuation from the variation of b  
(several ellipses contribute)

## Fit result : $P(N_{ch}, \delta p_t)$

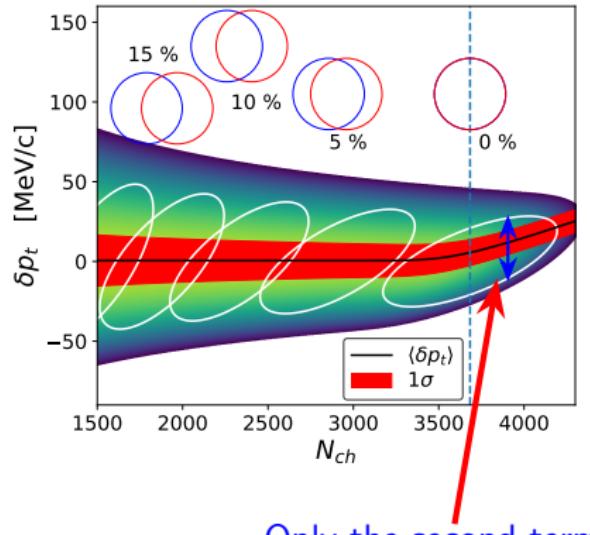
- We get,  $P(N_{ch}, \delta p_t)$   
 $= \int P(N_{ch}, \delta p_t | b) P(b) db$
- By conditional probability  
 $P(\delta p_t | N_{ch}) = \frac{P(N_{ch}, \delta p_t)}{P(N_{ch})}$   
 $\Rightarrow \text{Var}([\delta p_t] | N_{ch})$  is the squared width of  $P(\delta p_t | N_{ch})$
- The width of  $[\delta p_t]$  fluctuation has two contributions :
  - I due to fluctuation of impact parameter  $b$
  - II the true intrinsic fluctuation



fluctuation of  $[\delta p_t]$  at fixed  $b$  and fixed  $N_{ch}$   
 (height of a single ellipse)

**Fit result :**  $P(N_{ch}, \delta p_t)$

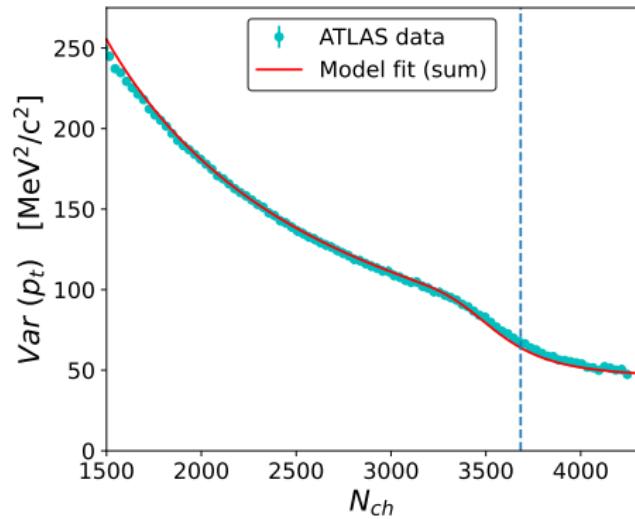
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  - The width of  $[\delta p_t]$  fluctuation has two contributions :
    - ① due to fluctuation of impact parameter b
    - ② the true intrinsic fluctuation
  - Only the second term contributes above knee in the ultracentral regime.



Only the second term  
remains in  
ultracentral collisions

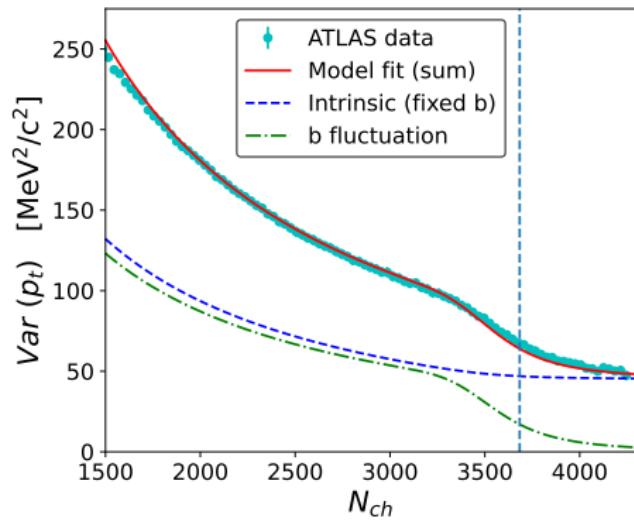
## Fit result : $\text{Var}([p_t])$ vs $N_{ch}$

- Our simple model naturally reproduces the steep fall in the ATLAS data very well !



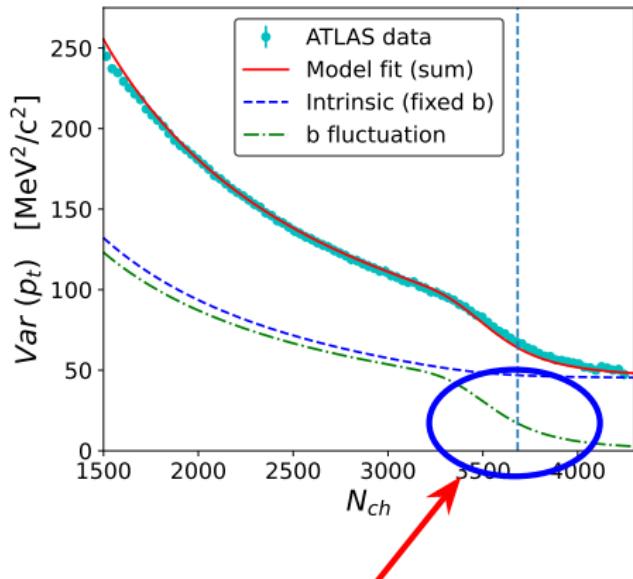
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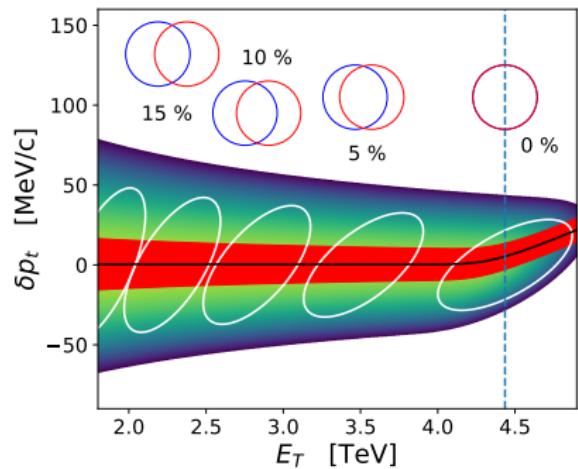
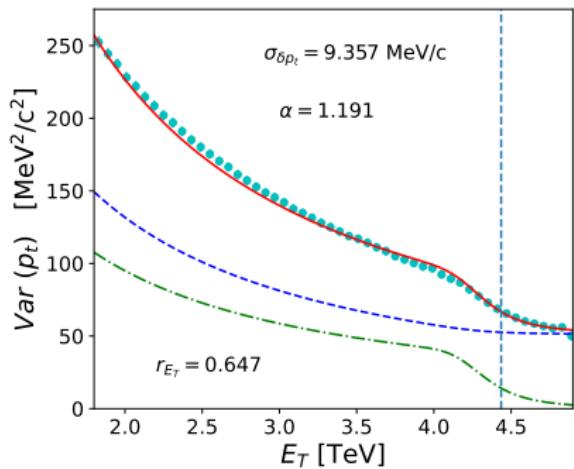
## Fit result : $\text{Var}([p_t])$ vs $N_{ch}$

- Our simple model naturally reproduces the steep fall in the ATLAS data very well !
- Below the knee, half of the contribution is from impact parameter fluctuation
- The contribution gradually disappears around the knee !



Contribution of b-fluctuation disappears above knee

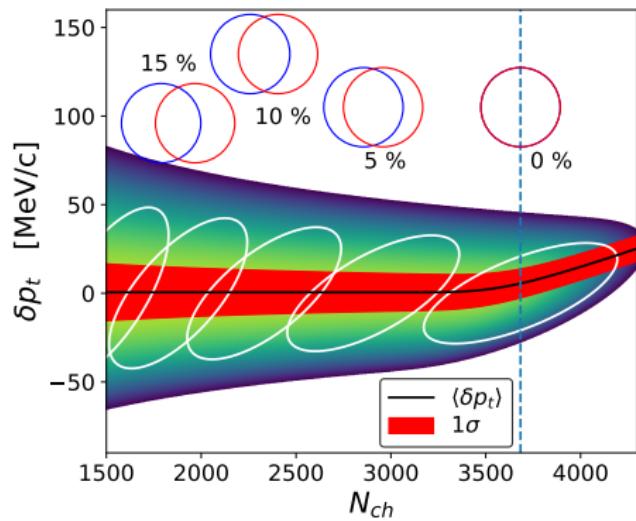
# $E_T$ -dependent $[p_t]$ -fluctuation (ATLAS)



**Impact parameter fluctuation is small !**

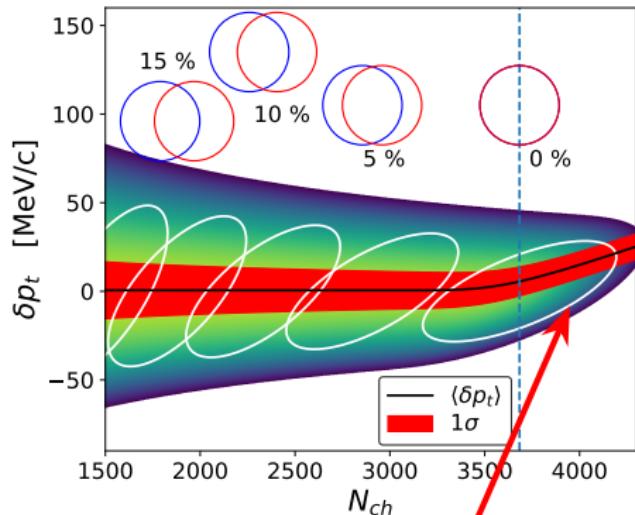
## Hint of Thermalization !

- Our model fit returns  $r = 0.676$  !



## Hint of Thermalization !

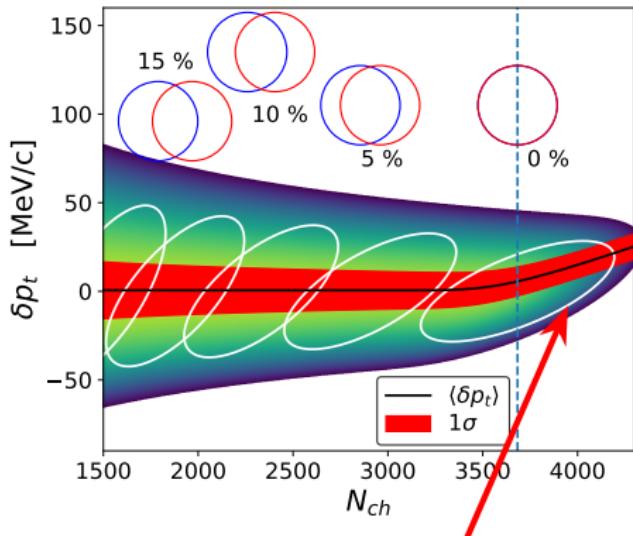
- Our model fit returns  $r = 0.676$  !
- It suggests strong correlation between  $[p_t]$  and  $N_{ch}$  at fixed  $b$



Strong correlation between  
 $[p_t]$  and  $N_{ch}$  at fixed  $b$   
from our model fit

## Hint of Thermalization !

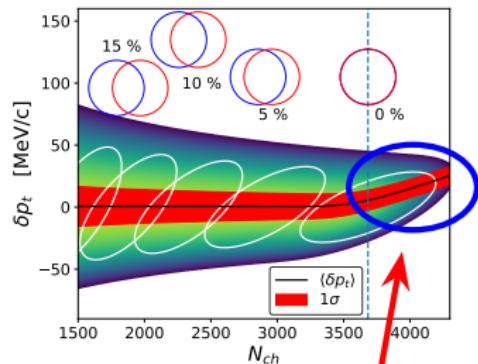
- Our model fit returns  $r = 0.676$  !
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- thermalization is at work ?



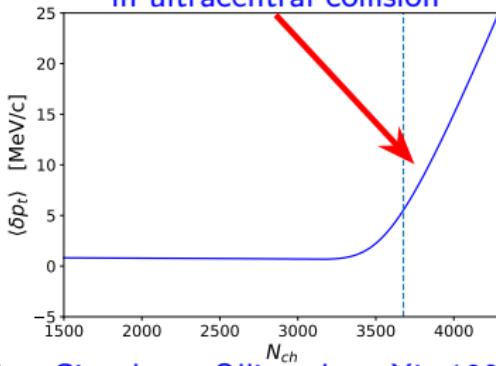
Strong correlation between  
 $[p_t]$  and  $N_{ch}$  at fixed  $b$   
from our model fit

### Further predictions : Mean, Skewness and kurtosis

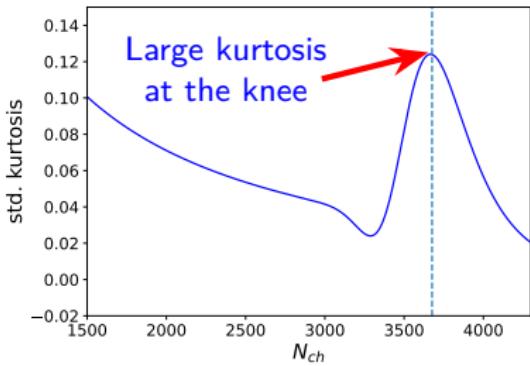
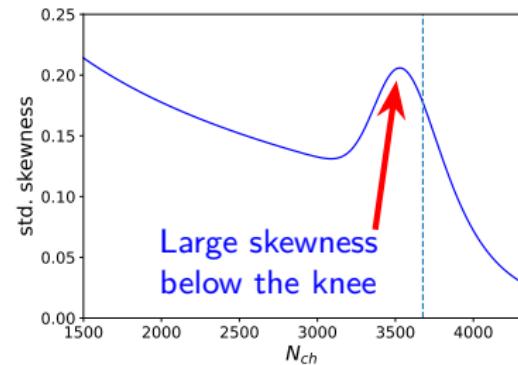
RS, Picchetti, Luzum, Ollitrault, arXiv:2306.09294



Slight increase of mean  $[p_t]$   
in ultracentral collision



Gardim, Giacalone, Ollitrault, arXiv:1909.11609



$P(\delta p_t | N_{ch}, c_b)$  in terms of k1 and k2

$$P(\delta p_t | N_{ch}, c_b) = \frac{1}{\sqrt{2\pi\kappa_2(c_b)}} \exp\left(-\frac{(\delta p_t - \kappa_1(c_b))^2}{2\kappa_2(c_b)}\right)$$

$$\kappa_1(c_b) = r \frac{\sigma_{p_t}(c_b)}{\sigma_{N_{ch}}(c_b)} (N_{ch} - \overline{N_{ch}}(c_b)),$$

$$\kappa_2(c_b) = (1 - r^2) \sigma_{p_t}^2(c_b).$$

## Moments and cumulants of $[p_t]$ -fluctuation

$$\langle \delta p_t \rangle = \langle \kappa_1 \rangle,$$

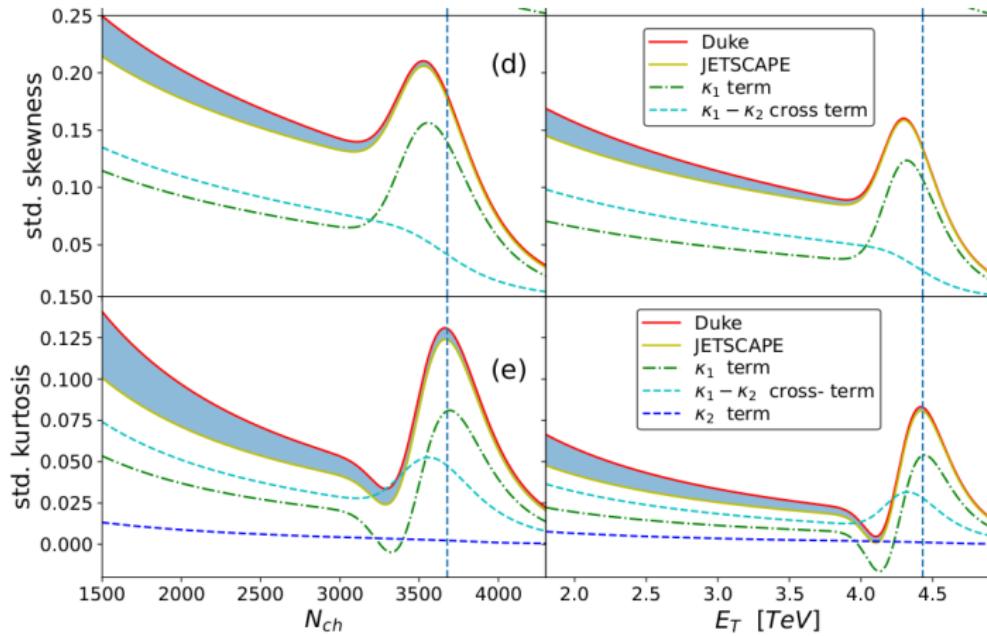
$$\langle \delta p_t | c_b \rangle = \kappa_1, \quad \text{Var}(p_t) = (\langle \kappa_1^2 \rangle - \langle \kappa_1 \rangle^2) + \langle \kappa_2 \rangle,$$

$$\langle \delta p_t^2 | c_b \rangle = \kappa_1^2 + \kappa_2, \quad \text{Skew}(p_t) = \langle \kappa_1^3 \rangle - 3\langle \kappa_1^2 \rangle \langle \kappa_1 \rangle + 2\langle \kappa_1 \rangle^3$$

$$\langle \delta p_t^3 | c_b \rangle = \kappa_1^3 + 3\kappa_2\kappa_1, \quad + 3(\langle \kappa_2\kappa_1 \rangle - \langle \kappa_2 \rangle \langle \kappa_1 \rangle),$$

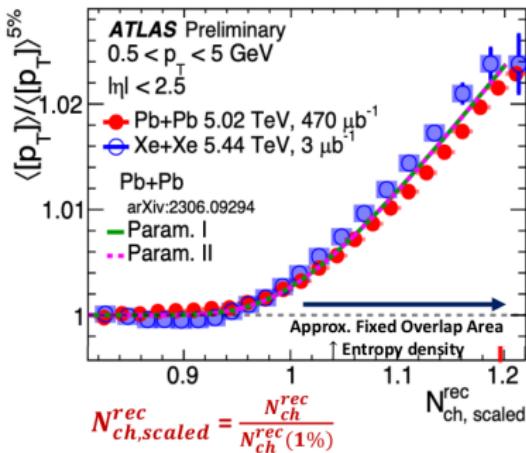
$$\langle \delta p_t^4 | c_b \rangle = \kappa_1^4 + 6\kappa_2\kappa_1^2 + 3\kappa_2^2, \quad \text{Kurt}(p_t) = \langle \kappa_1^4 \rangle - 4\langle \kappa_1^3 \rangle \langle \kappa_1 \rangle + 6\langle \kappa_1^2 \rangle \langle \kappa_1 \rangle^2 - 3\langle \kappa_1 \rangle^4$$
$$+ 6(\langle \kappa_2\kappa_1^2 \rangle - \langle \kappa_2 \rangle \langle \kappa_1^2 \rangle - 2\langle \kappa_2\kappa_1 \rangle \langle \kappa_1 \rangle$$
$$+ 2\langle \kappa_2 \rangle \langle \kappa_1 \rangle^2) + 3(\langle \kappa_2^2 \rangle - \langle \kappa_2 \rangle^2),$$

## Anatomy of skewness and kurtosis prediction !

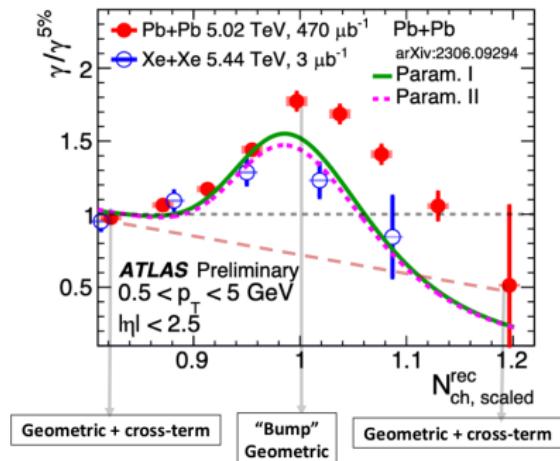


ATLAS Preliminary, presented in QM2023

T.Bold's Bialasowka talk on Oct 27

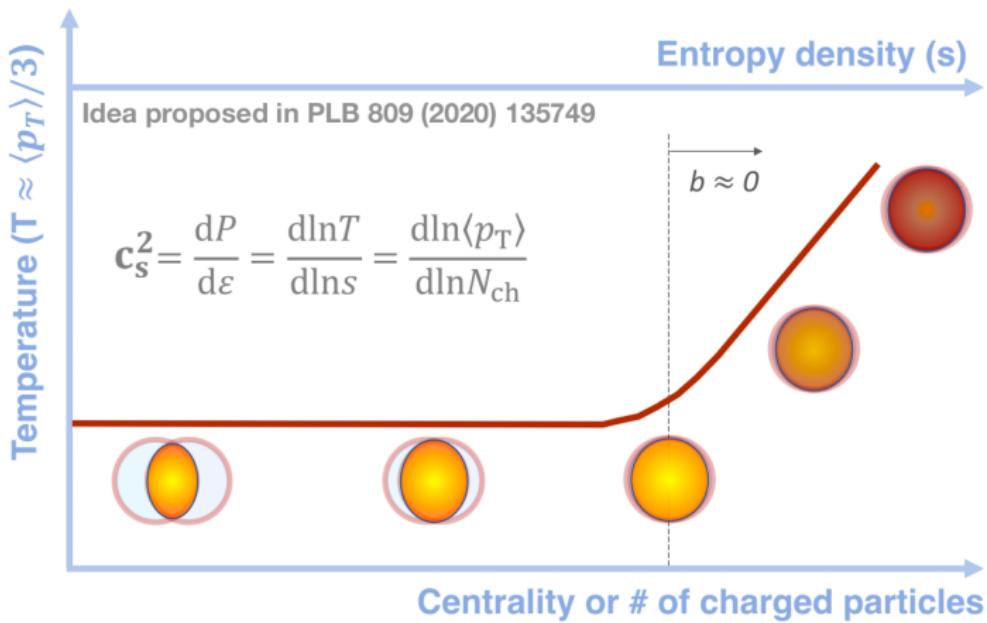


mean  $[p_t]$



$[p_t]$ - skewness

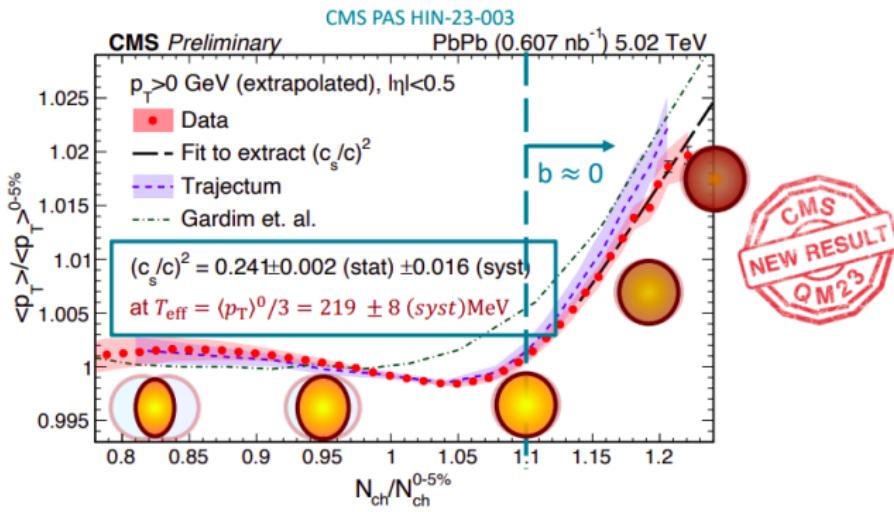
## Application : Extraction of speed of sound in QGP from mean $\langle p_t \rangle$



## CMS Result on mean $\langle p_T \rangle$ !

See CMS preliminary in QM 2023

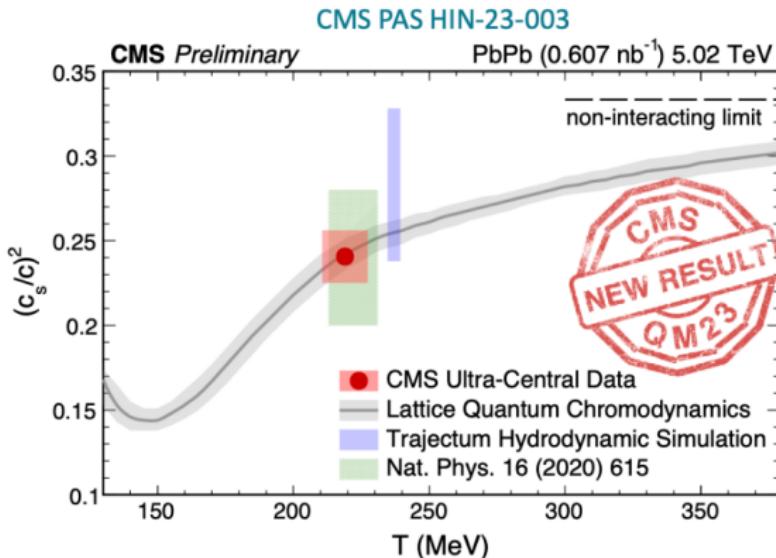
Significant increase of  $\langle p_T \rangle$  toward UCC events as predicted by the simulations



Speed of sound extracted from the fit and  $T_{\text{eff}}$  from  $\langle p_T \rangle^0$

## How precise is the measurement ?

See CMS preliminary in QM 2023



Speed of sound in QGP is predicted and measured with great precision !!

## Summary and Outlook

- Two separate contributions to  $[p_t]$  fluctuation :
  - ➊ intrinsic fluctuation → originates from quantum fluctuation in the initial state
  - ➋ impact parameter fluctuation at fixed  $N_{ch}$  → disappears in ultracentral region → causes the steep fall at the knee

Our methodology paves a way to separate the geometrical and quantum fluctuations

- We present predictions for mean  $[p_t]$ , skewness and kurtosis of  $[p_t]$ -fluctuation → the unique patterns of the cumulants of  $[p_t]$  fluctuation at the ultracentral regime originates mostly due to b-fluctuation !
- Prediction of increase of mean  $[p_t]$  with  $N_{ch}$  leads to the precise extraction of speed of sound ( $c_s^2$ ) in QGP
- Transverse momentum fluctuation in ultra central collision could be a probe of the thermalization !

## Summary and Outlook

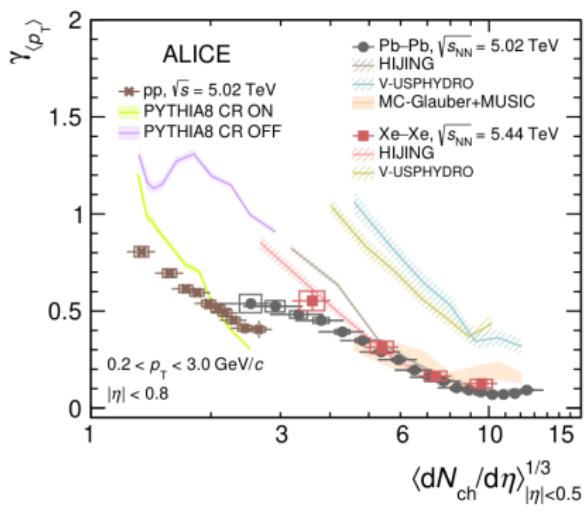
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... Thank you for your attention !

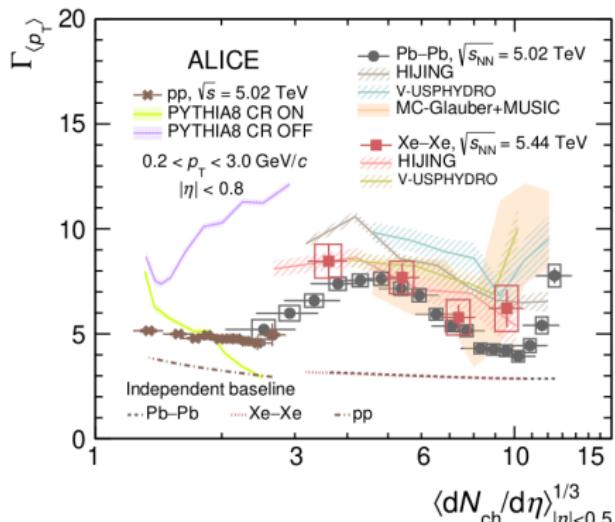


# ALICE Measurements of $[p_t]$ -skewness !

arXiv: 2308.16217



standardized  $[p_t]$ -skewness



intensive  $[p_t]$ -skewness