

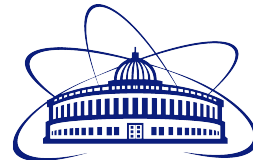


Определение центральности столкновений тяжелых ионов в эксперименте BM@N

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Научная сессия секции ядерной физики ОФН РАН
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Работа поддержана Министерством науки и высшего образования РФ, проект "Фундаментальные и прикладные исследования на экспериментальном комплексе класса мегасайенс NICA (ОИЯИ)"
№ FSWU-2025-0014



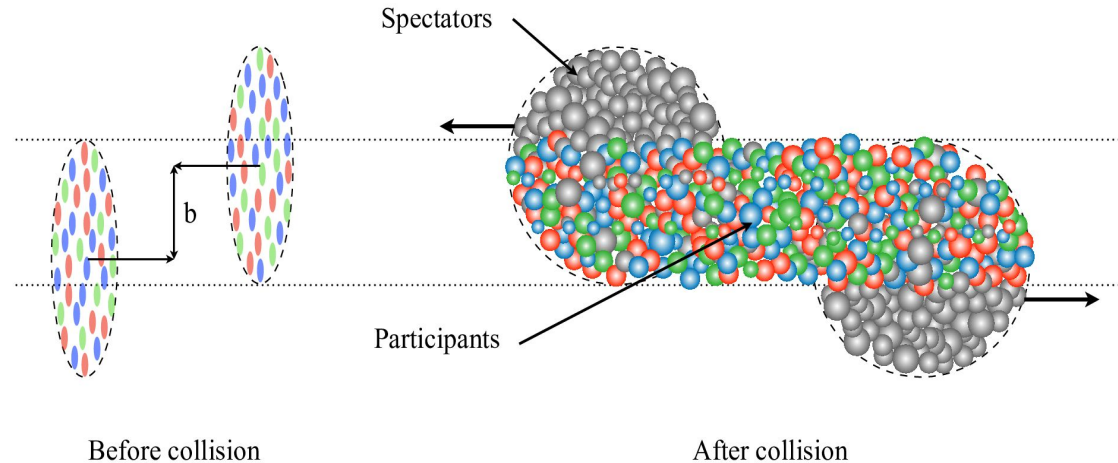
Motivation for centrality determination

- Evolution of matter produced in heavy-ion collisions depends on its initial geometry
- Impact parameters (**b**) - one of the important collision parameters
 - impossible to measure experimentally
- **Goal of centrality determination:** map (on average) the collision geometry parameters to experimental observables (centrality estimators)

Centrality class S_1 - S_2 :

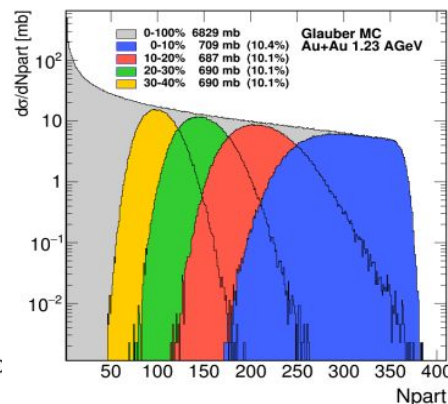
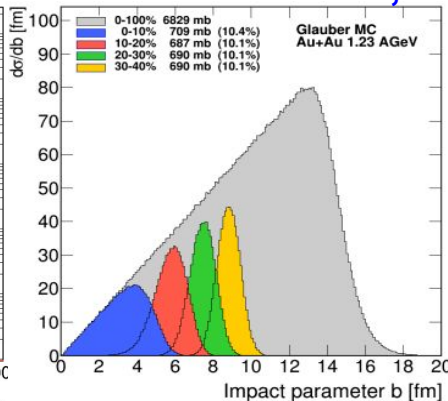
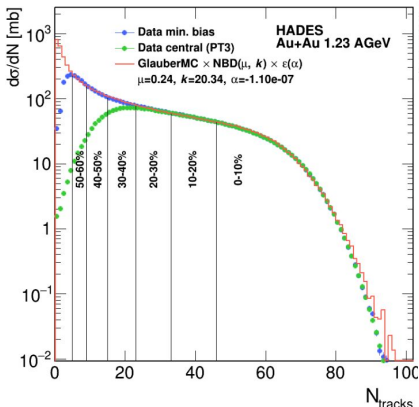
group of events corresponding to a given fraction (in %) of the total cross section:

$$C_S = \frac{1}{\sigma_{inel}^{AA}} \int_{S_1}^{S_2} \frac{d\sigma}{dS} dS$$



Centrality determination

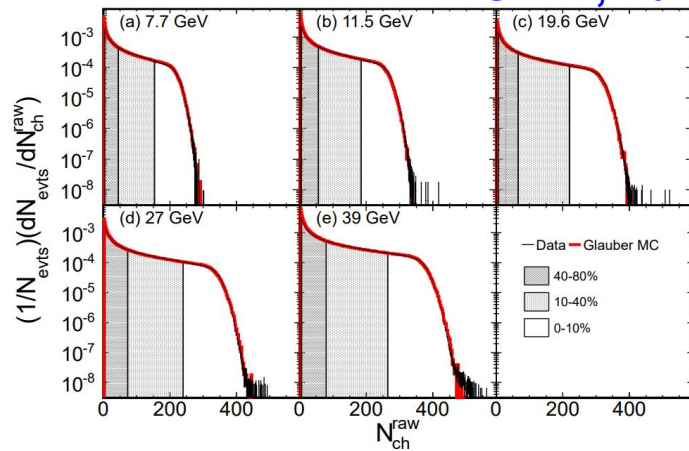
HADES, Au+Au 1.23A GeV



Eur. Phys. J. A (2018) 54: 85

Centrality Classes	b_{\min}	b_{\max}	$\langle b \rangle$
0 – 5 %	0.00	3.30	2.20
5 – 10 %	3.30	4.70	4.04
10 – 15 %	4.70	5.70	5.22
15 – 20 %	5.70	6.60	6.16
20 – 25 %	6.60	7.40	7.01
25 – 30 %	7.40	8.10	7.75
30 – 35 %	8.10	8.70	8.40
35 – 40 %	8.70	9.30	9.00
40 – 45 %	9.30	9.90	9.60
45 – 50 %	9.90	10.40	10.15
50 – 55 %	10.40	10.90	10.65
55 – 60 %	10.90	11.40	11.15

STAR, Au+Au, BES



Phys. Rev. C 86, 054908 (2012)

Centrality (%)	$\langle N_{\text{part}} \rangle$	$\langle N_{\text{coll}} \rangle$
0-5%	337 ± 2	774 ± 28
5-10%	290 ± 6	629 ± 20
10-20%	226 ± 8	450 ± 22
20-30%	160 ± 10	283 ± 24
30-40%	110 ± 11	171 ± 23
40-50%	72 ± 10	96 ± 19
50-60%	45 ± 9	52 ± 13
60-70%	26 ± 7	25 ± 9
70-80%	14 ± 4	12 ± 5

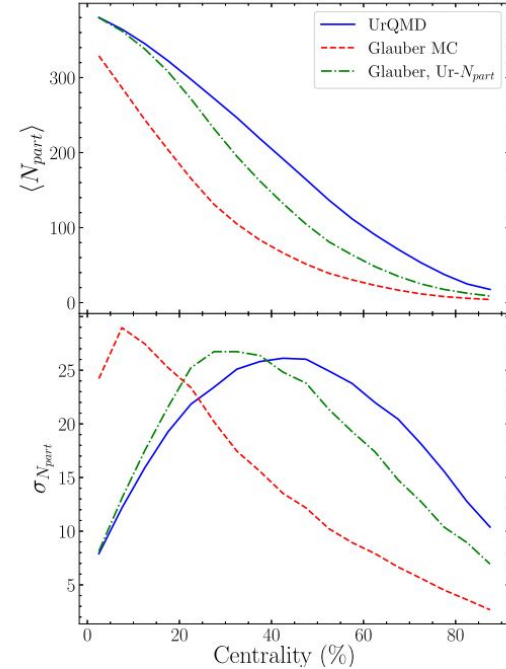
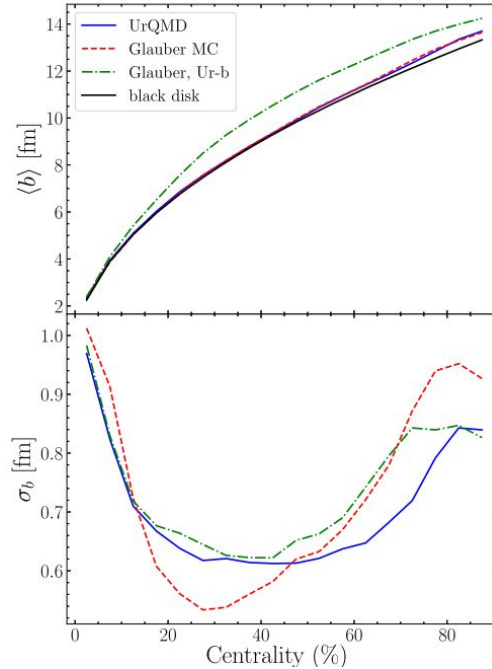
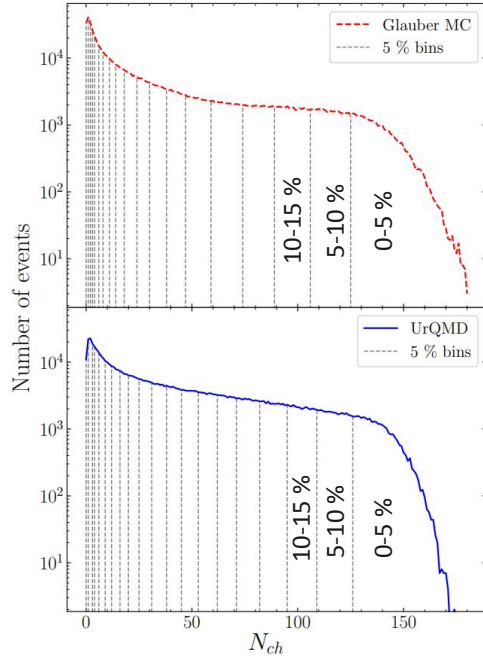
Centrality determination based on multiplicity provides with:

- impact parameter (b)
- number of participating nucleons (N_{part})

Similar centrality estimator is needed for comparisons with STAR, HADES, etc.

Model dependence of b , N_{part}

Eur. Phys. J. C 83, 792 (2023)



- MC-Glauber x NBD multiplicity fitting procedure is standard method for centrality determination
- The MC Glauber non-realistic N_{part} simulations at low energies
- Differences in of number of participant nucleons (N_{part}) distributions from UrQMD and MC
- The impact parameter (b) - model independent centrality estimator

The BM@N experiment

Simulation:

- DCM-QGSM-SMM, Xe-Cs
- GEANT4 transport

Data:

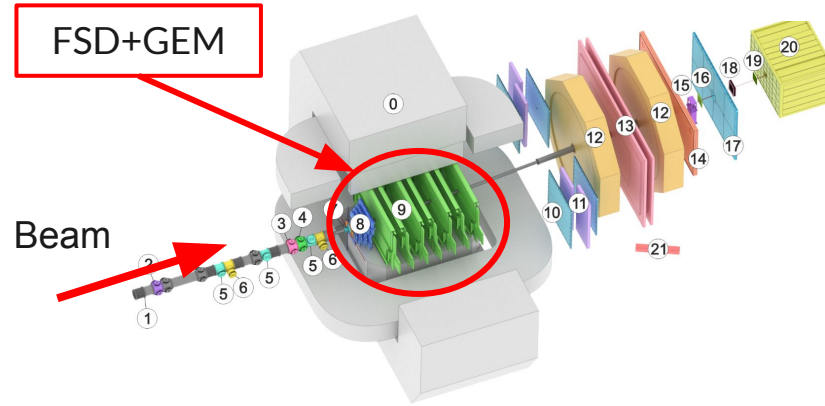
- run8 Xe-Csl @3.8A GeV
- Event selection:
 - Physical runs
 - Centrality trigger (CCT2)
 - More than 1 track in vertex reconstruction
 - $Vtx_R < 1.0$ cm
 - $Vtx_Z < 0.1$ cm

Multiplicity of charged particles from tracking system FSD+GEM

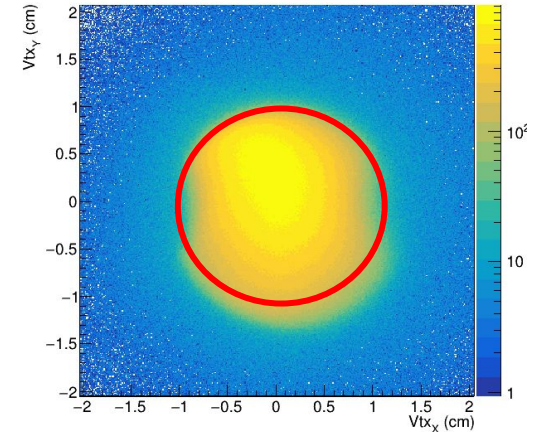
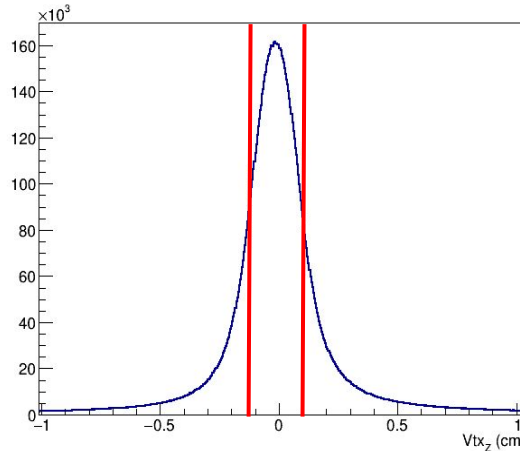
FSD+GEM

Beam

arXiv:2312.17573



- Magnet SP-41 (0)
- Vacuum Beam Pipe (1)
- BC1, VC, BC2 (2-4)
- SIBT, SiProf (5, 6)
- Triggers: BD + SiMD (7)
- FSD, GEM (8, 9)
- CSC 1x1 m² (10)
- TOF 400 (11)
- DCH (12)
- TOF 700 (13)
- ScWall (14)
- FD (15)
- Small GEM (16)
- CSC 2x1.5 m² (17)
- Beam Profilometer (18)
- FQH (19)
- FHCAL (20)
- HGN (21)



Centrality determination based on Monte-Carlo sampling of produced particles

For **multiplicity of produced particles** used in HADES, CBM, NA61/SHINE

Get (b, N_{part}, N_{coll}) from MC-Glauber

Evaluate number of ancestors (sources of produced particles)
$$N_a = f N_{part} + (1-f) N_{coll}$$

Sample multiplicity of produced particles (S_i) N_a times from $NBD(\mu, \mathbf{k})$

Multiplicities from two collision events are randomly superimposed with the probability p ("pileup" events)

Result: total S_{tot}

MC-Glauber distribution

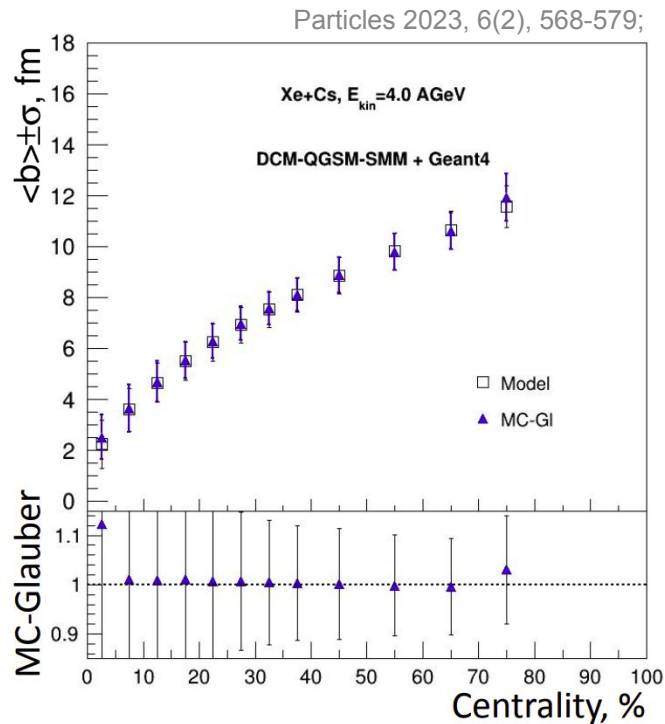
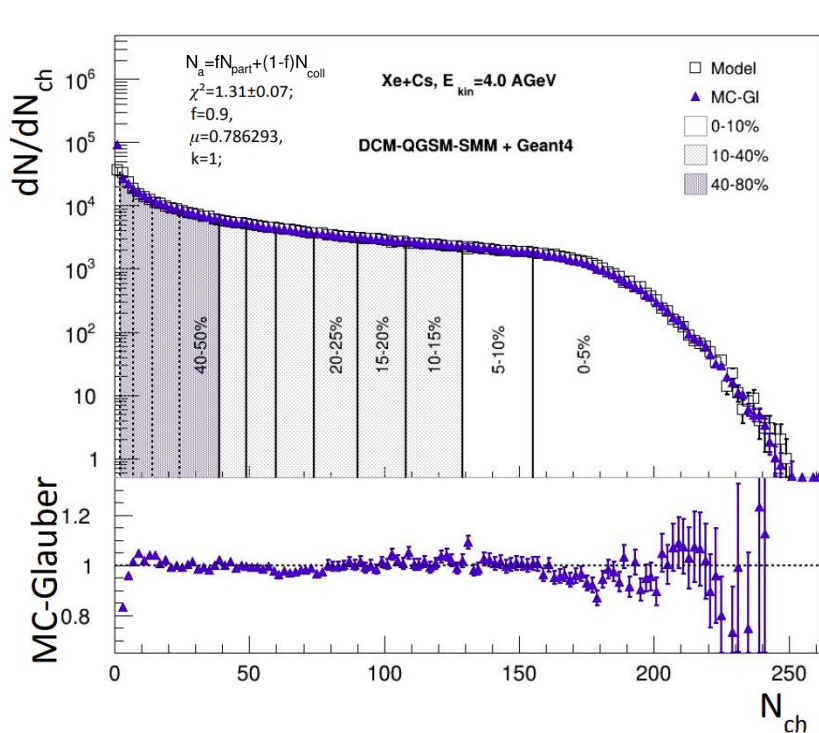
Full Monte-Carlo (real data) distribution

Evaluate χ^2 between $N/dN_{MC/data}$ and N/dN_{GI}

Scan phase space of parameters to find their values for minimum of χ^2

Extract relation between geometry parameters and centrality estimator

MC-Glauber fit result Xe-Cs



- Good agreement between model data and fit
- Impact parameter distributions in different centrality classes reproduces ones from DCM-QGSM-SMM

The Bayesian inversion method (Γ -fit)

Relation between multiplicity N_{ch} and impact parameter b is defined by the fluctuation kernel:

$$P(N_{ch}|c_b) = \frac{1}{\Gamma(k(c_b))\theta^k} N_{ch}^{k(c_b)-1} e^{-N_{ch}/\theta}$$

$$\frac{\sigma^2}{\langle N_{ch} \rangle} = \theta \approx const, k = \frac{\langle N_{ch} \rangle}{\theta}$$

$$c_b = \int_0^b P(b') db' \quad - \text{centrality based on impact parameter}$$

Mean multiplicity as a function of c_b can be defined as follows:

$$\langle N_{ch} \rangle = N_{knee} \exp\left(\sum_{j=1}^3 a_j c_b^j\right) \quad N_{knee}, \theta, a_j - 5 \text{ parameters}$$

$$\text{Fit function for } N_{ch} \text{ distribution: } P(N_{ch}) = \int_0^1 P(N_{ch}|c_b) dc_b$$

$$\text{b-distribution for a given } N_{ch} \text{ range: } P(b|n_1 < N_{ch} < n_2) = P(b) \frac{\int_{n_1}^{n_2} P(N_{ch}|b) dN_{ch}}{\int_{n_1}^{n_2} P(N_{ch}) dN_{ch}}$$

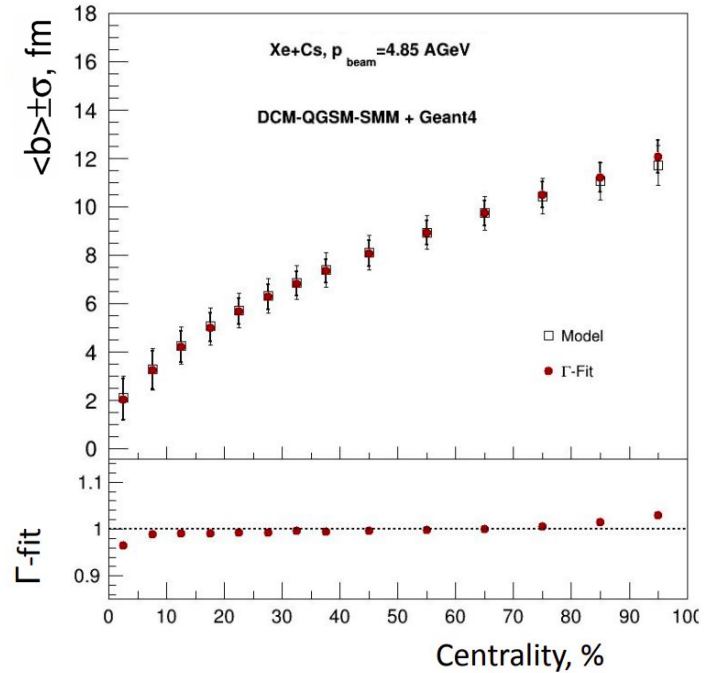
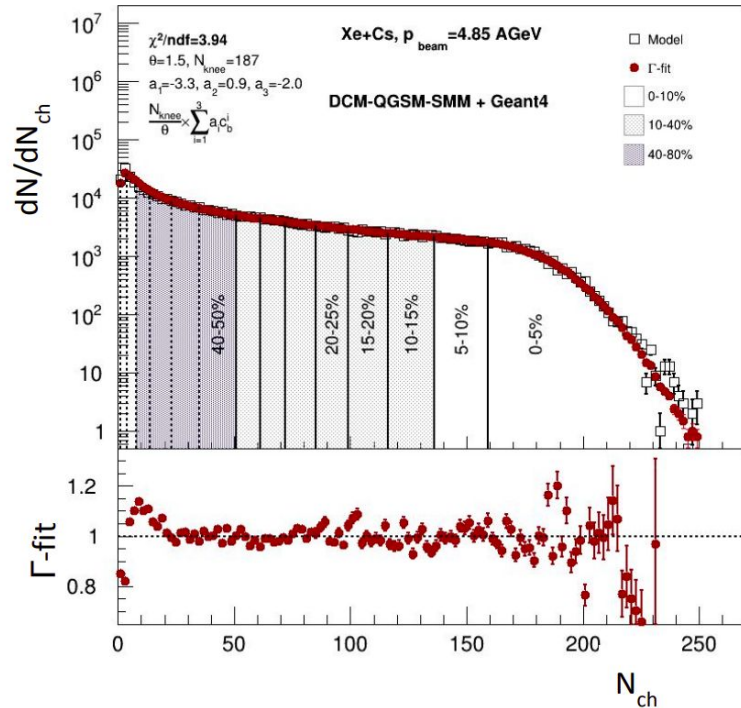
2 main steps of the method:

Fit experimental (model) distribution with $P(N)$



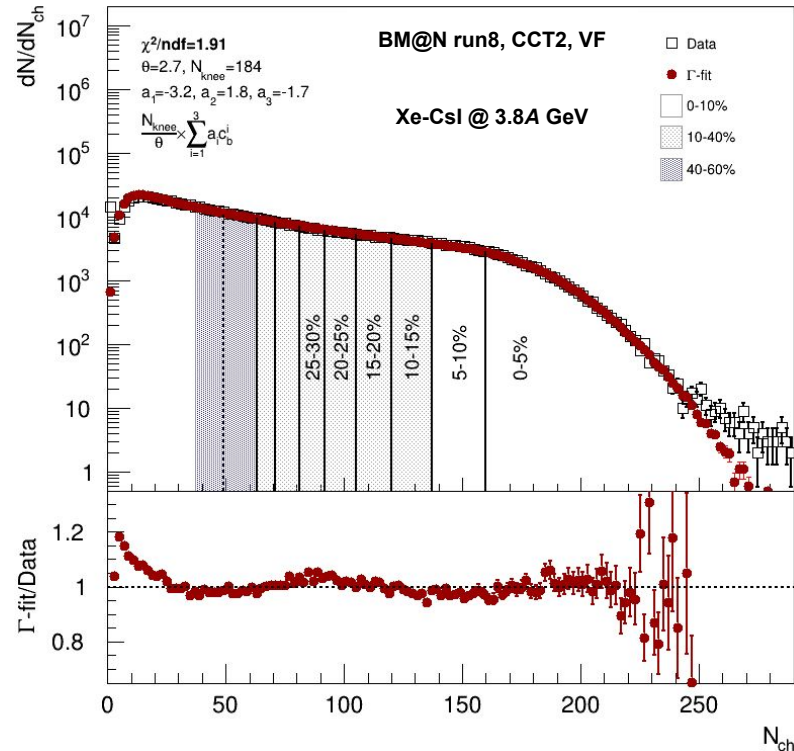
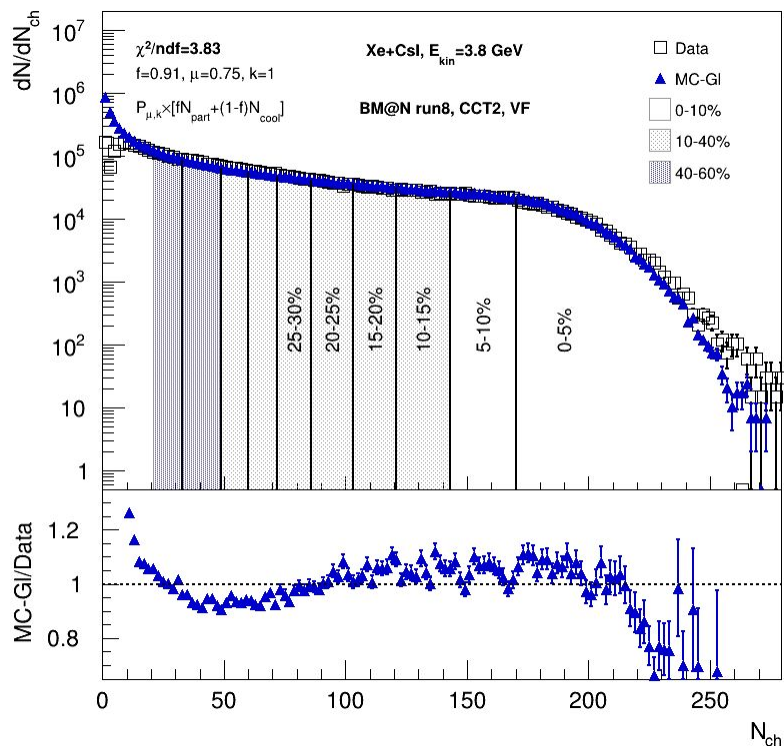
Construct $P(b|N)$ using Bayes' theorem:
 $P(b|N) = P(b)P(N|b)/P(N)$

Γ -fit result Xe-Cs



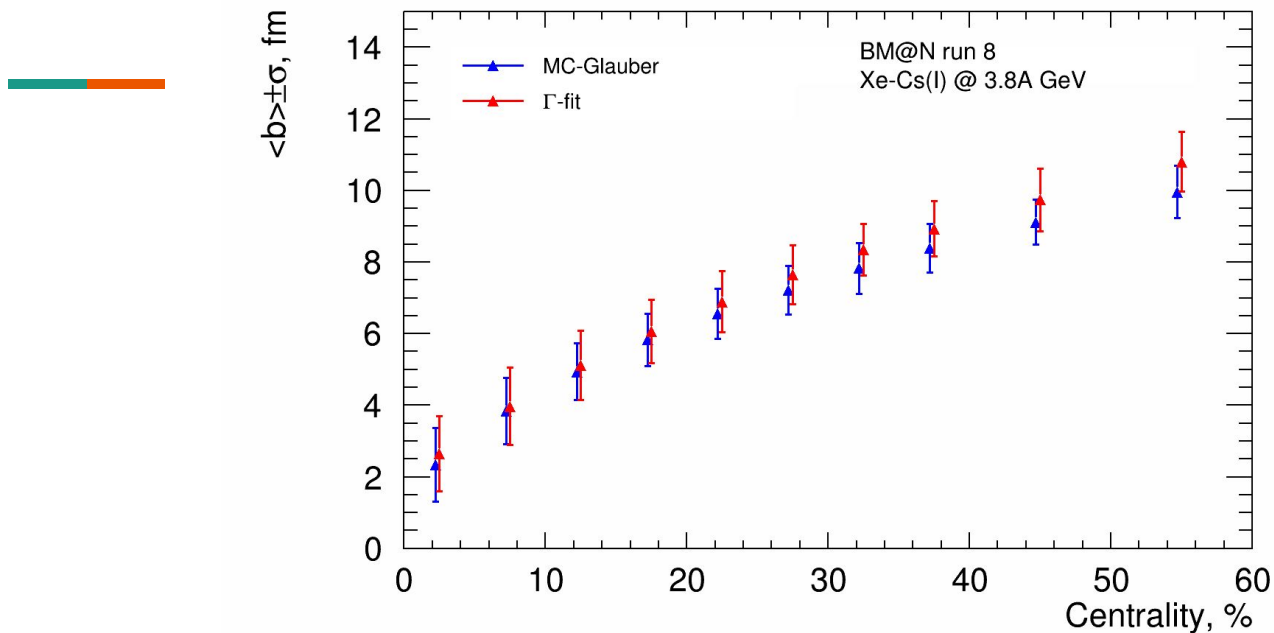
- Good agreement between model data and fit
- Impact parameter distributions in different centrality classes reproduces ones from DCM-QGSM-SMM

Result of centrality determination at Xe-CsI @ 3.8 AGeV



- Centrality determination methods were applied on experimental Xe-CsI data
- Good agreement between data and fit for both methods
- New centrality classes is used in analysis (see talk by M.Mamaev)

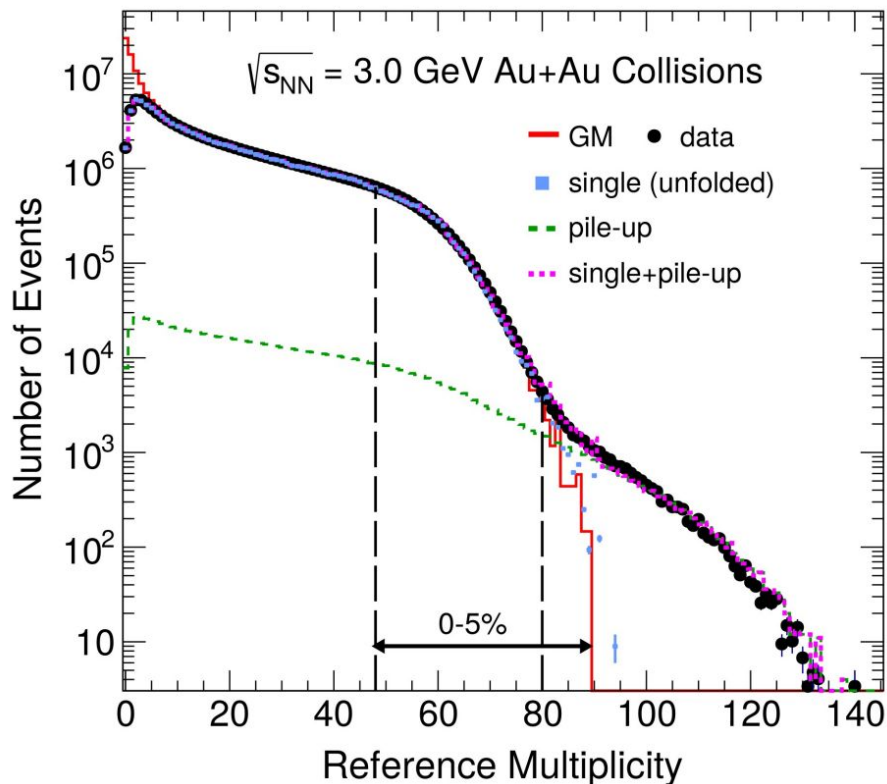
Comparison between impact parameter distributions



- For Γ -fit all centrality classes are comparable
- Γ -fit and MC-Glauber fit are now in more agreement with each other

Implementation of “pileup” in the centrality determination procedure

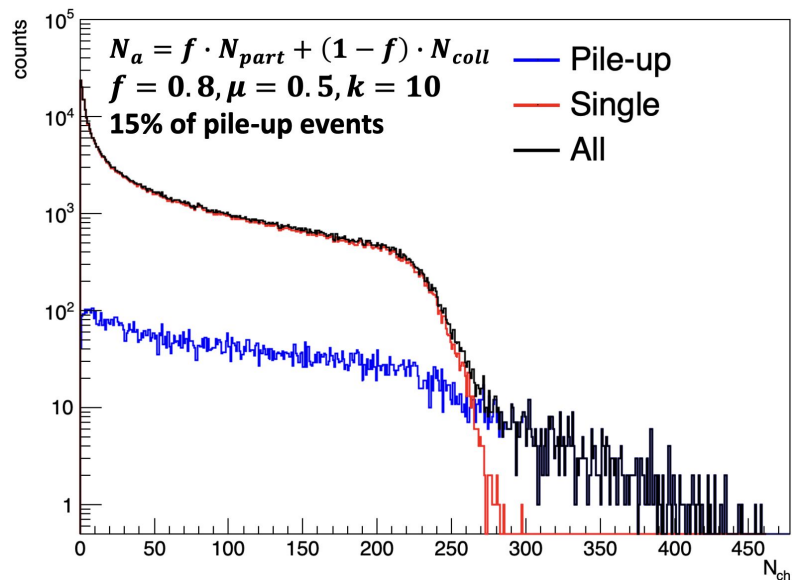
<https://arxiv.org/abs/2112.00240>



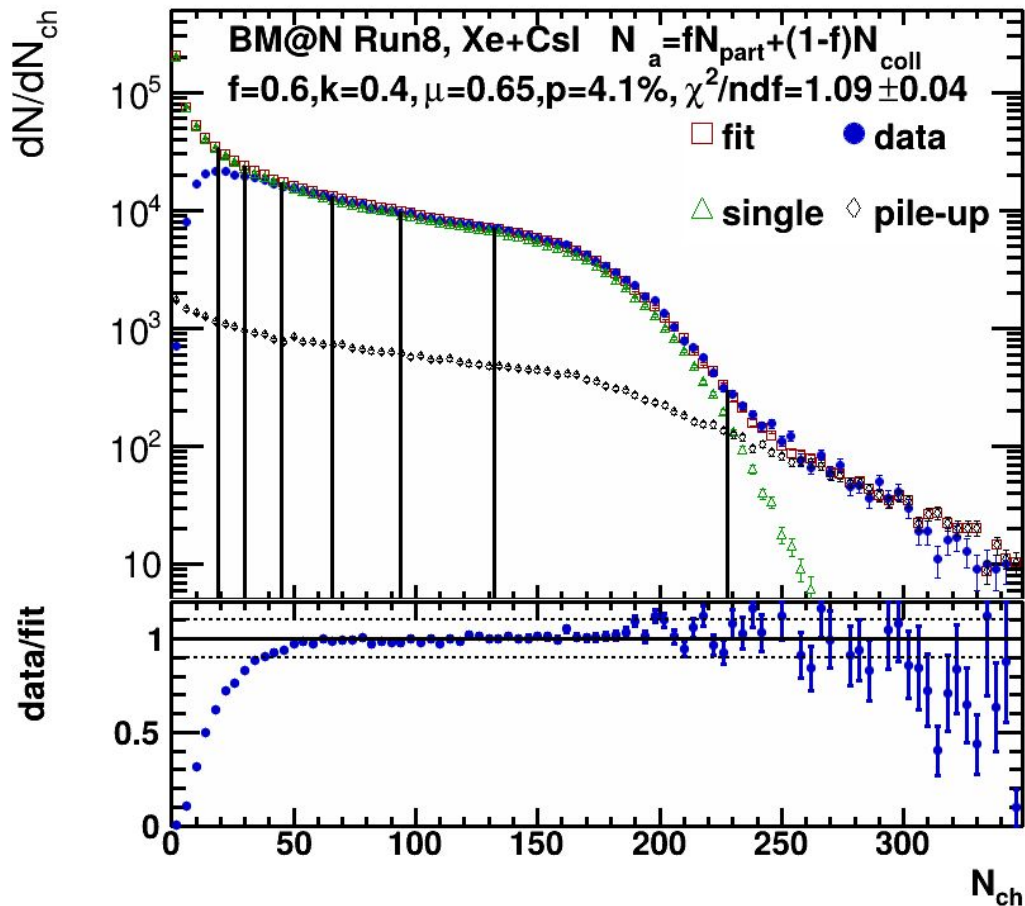
Pileup events occur with the probability α_m at the m multiplicity bin.

The probability to find N particles of interest at multiplicity m with the pileup effects is given by:

$$P_m(N) = (1 - \alpha_m)P_m^{\text{single}}(N) + \alpha_m P_m^{\text{pileup}}(N)$$



Result of centrality determination (with “pileup”)



RunId: 8120-8170

Multiplicity Cuts:

- CCT2
- $N_{\text{vtxTr}} > 1$
- (Sts digi vs N_{tr}) cut
- $V_r < 1$ cm
- $V_z < 0.1$ cm

Fit predicts 4% pileup events

Good agreement with experimental data

Summary and outlook

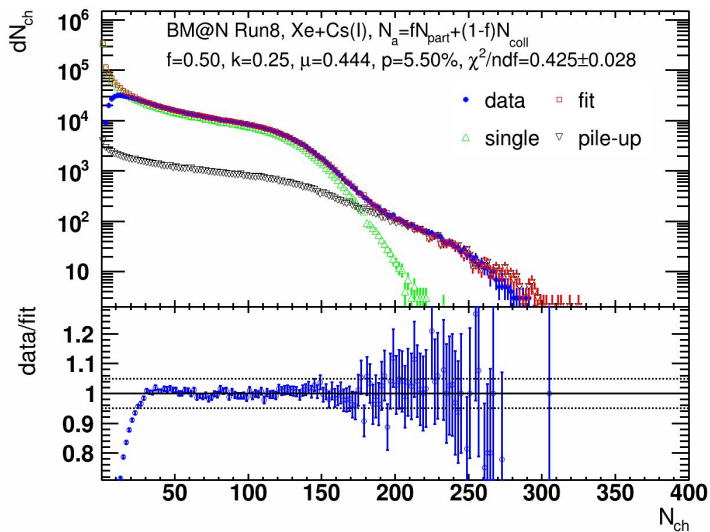
- The first version of it is performed
 - A new approach to accounting for pileup is considered
 - The MC-Glauber and the Bayesian inversion method reproduce charged particle multiplicity for fixed-target experiment at BM@N
 - Relation between impact parameter and centrality classes is extracted
 - These results are used in the physics analysis
- ❖ Consider the multiplicity h^- / π^{+-} to determine centrality

Thank you for your attention!

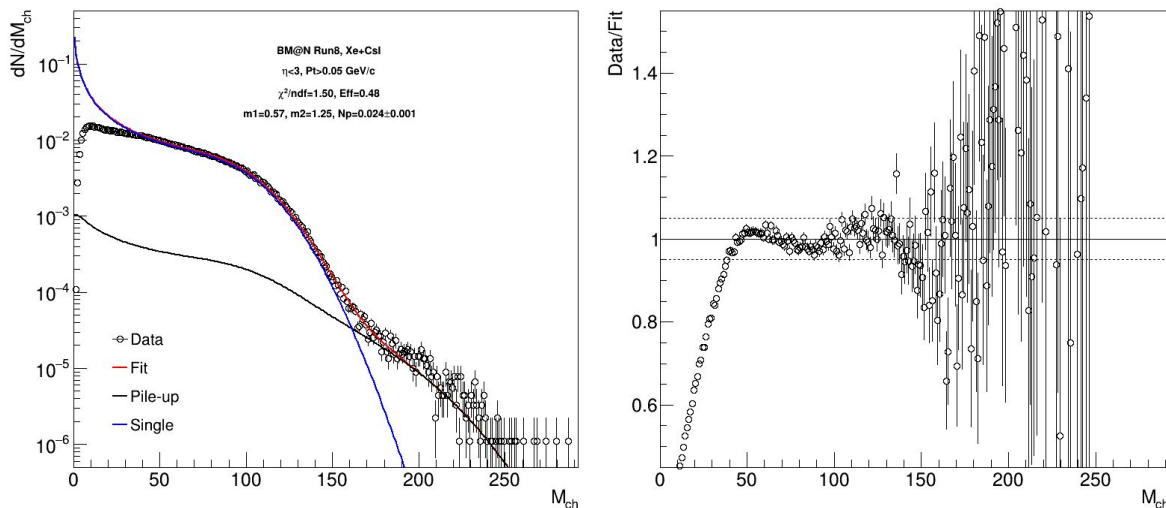


Result of centrality determination at Xe-CsI @ 3.8 AGeV

MC-Glauber



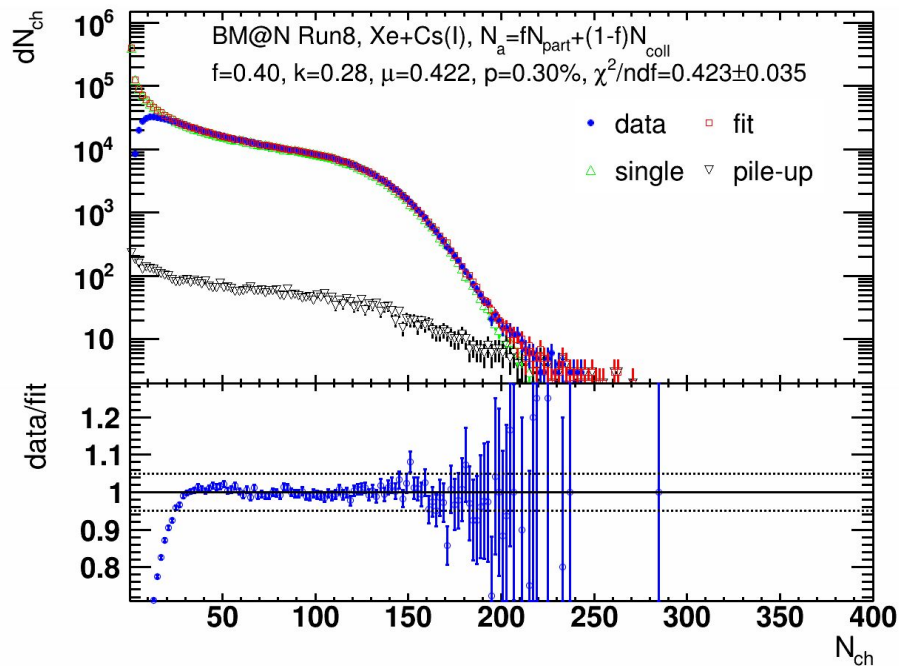
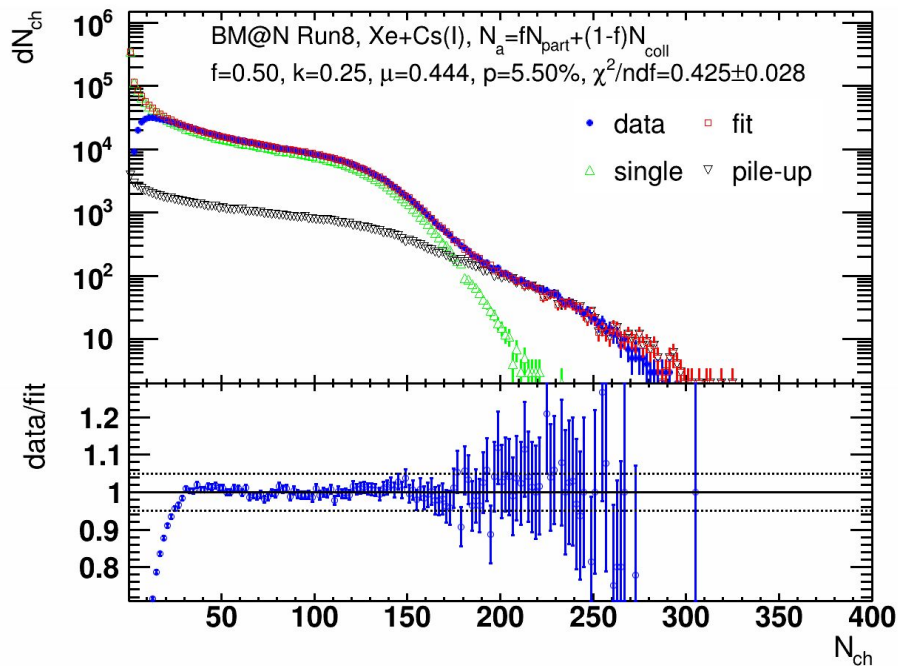
Γ -fit



D. Idrisov, 13th Collaboration Meeting of the BM@N Experiment at NICA, ICPPA-2024, "Comparison of different centrality determination methods at the BM@N experiment"

- Centrality determination methods were applied on experimental Xe-CsI data
- Good agreement between data and fit for both methods
- New centrality classes are used in physics analysis (see talk by M.Mamaev tomorrow)

Centrality determination after remove "pileup"

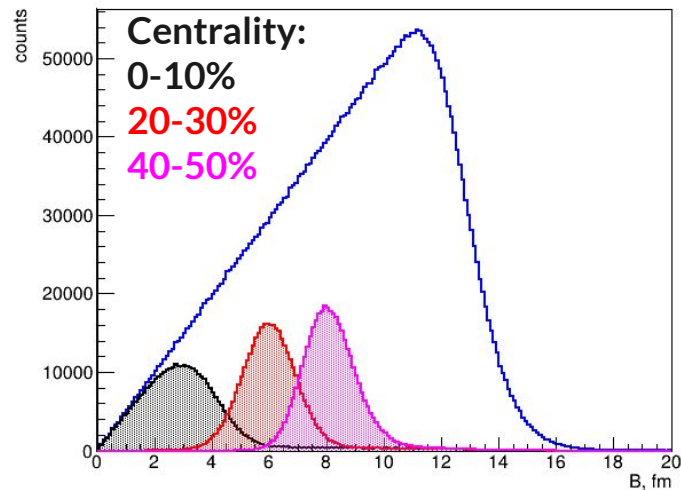
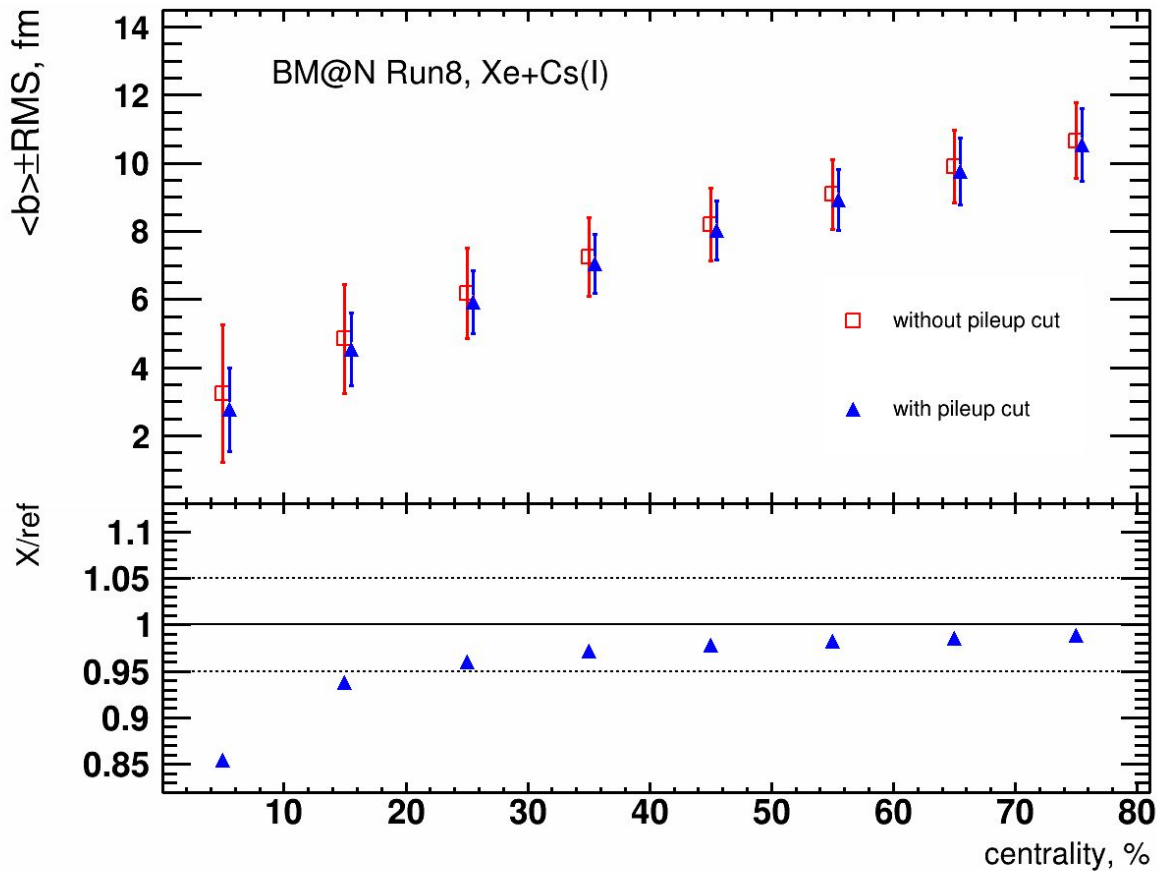


Change fit result

- $f: 0.5 \rightarrow 0.4$
- $k: 0.25 \rightarrow 0.28$
- $\mu: 0.44 \rightarrow 0.42$
- pileup: $5.5\% \rightarrow 0.3\%$

After pileup rejection the "pileup" events contribution is less 1%

Centrality determination after refMult correction: $\langle b \rangle$ vs cent

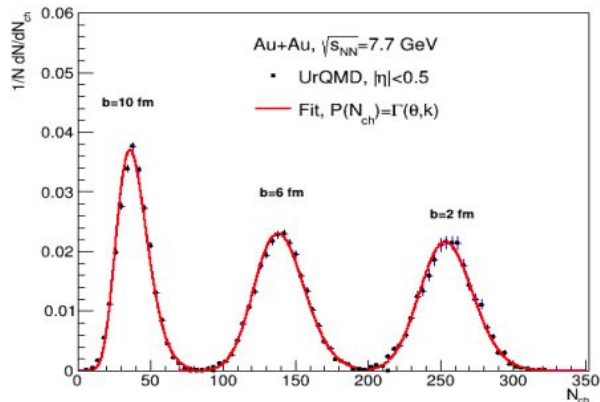


The Bayesian inversion method (Γ -fit): main assumptions

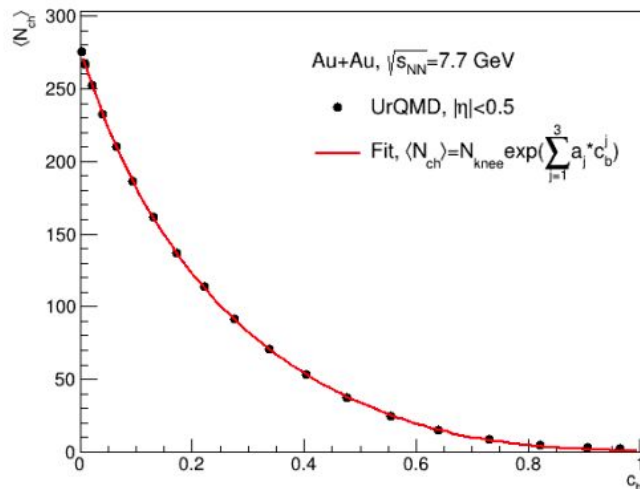
- Relation between multiplicity N_{ch} and impact parameter b is defined by the fluctuation kernel:

$$P(N_{ch}|c_b) = \frac{1}{\Gamma(k(c_b))\theta^k} N_{ch}^{k(c_b)-1} e^{-N_{ch}/\theta}$$

$$c_b = \int_0^b P(b') db' \simeq \frac{\pi b^2}{\sigma_{inel}} \quad \text{-- centrality based on impact parameter}$$



The results of fitting the multiplicity distribution for a fixed impact parameter



The dependence of the average value of multiplicity on centrality and the results of its fit

$$\frac{\sigma^2}{\langle N_{ch} \rangle} = \theta \simeq const$$

$$\langle N_{ch} \rangle = N_{knee} \exp\left(\sum_{j=1}^3 a_j c_b^j\right), \quad k = \frac{\langle N_{ch} \rangle}{\theta}$$

Five fit parameters

N_{knee}, θ, a_j

Reconstruction of b

- Normalized multiplicity distribution $P(N_{ch})$

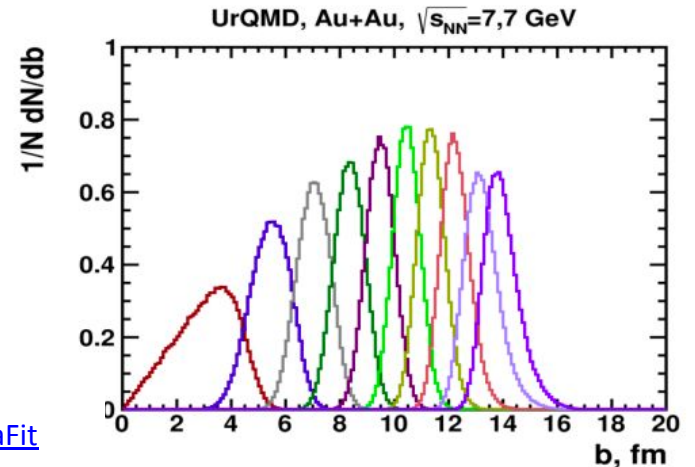
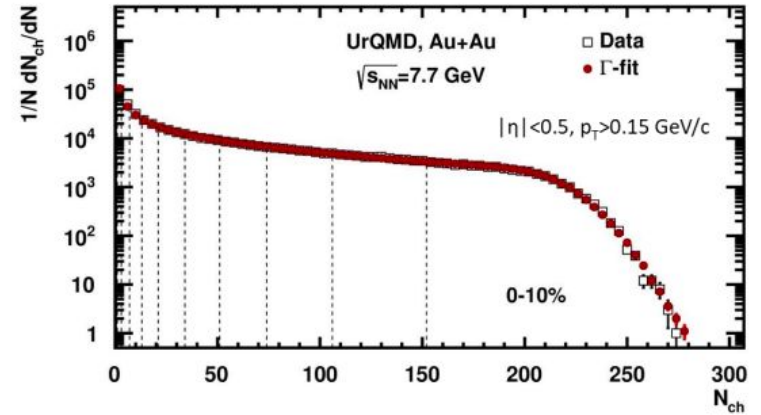
$$P(N_{ch}) = \int_0^1 P(N_{ch}|c_b)dc_b$$

- Find probability of b for fixed range of N_{ch} using Bayes' theorem:

$$P(b|n_1 < N_{ch} < n_2) = P(b) \frac{\int_{n_1}^{n_2} P(N_{ch}|b)dN_{ch}}{\int_{n_1}^{n_2} P(N_{ch})dN_{ch}}$$

- The Bayesian inversion method consists of 2 steps:**

- Fit normalized multiplicity distribution with $P(N_{ch})$
- Construct $P(b|N_{ch})$ using Bayes' theorem with parameters from the fit



R. Rogly, G. Giacalone and J. Y. Ollitrault, Phys.Rev. C98 (2018) no.2, 024902

Implementation for MPD and BM@N by D. Idrisov: <https://github.com/Dim23/GammaFit>

Example of application in MPD: P. Parfenov et al., Particles 4 (2021) 2, 275-287

The BM@N experiment

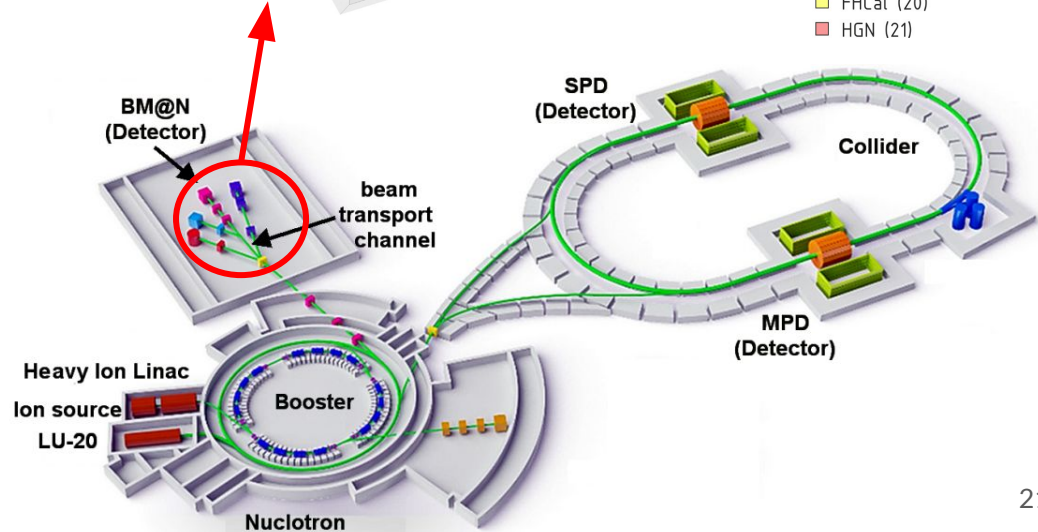
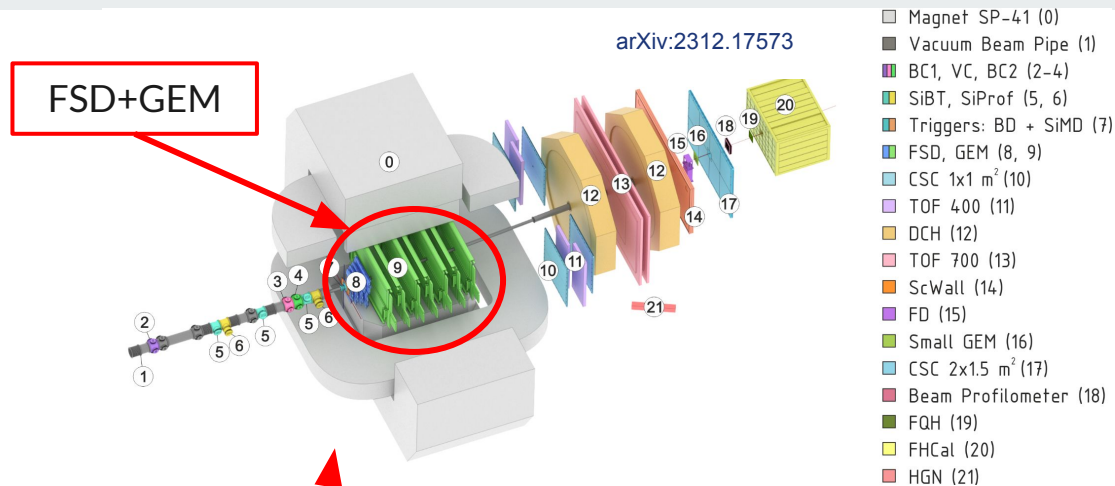
Simulation:

- Xe-Cs
- GEANT4 transport

Data:

- run8 Xe-CsI @3.8A GeV
- Physical runs

Multiplicity of charged particles from tracking system FSD+GEM

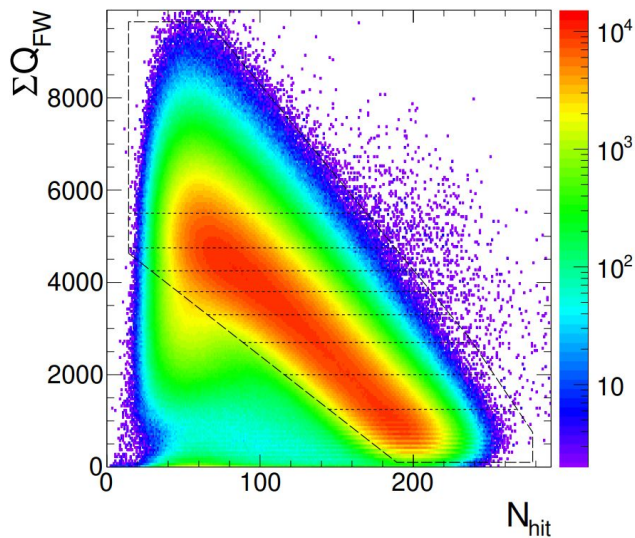


Why several alternative centrality estimators

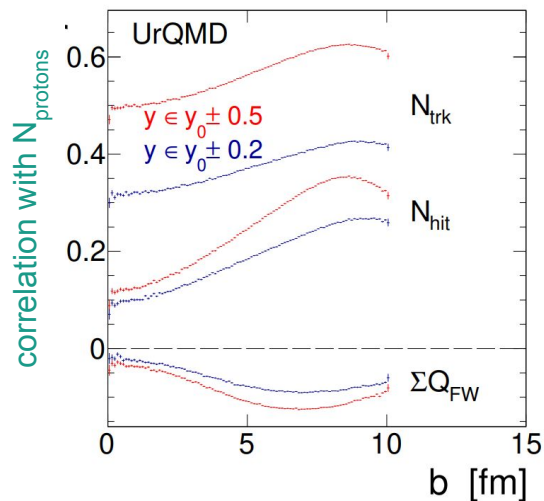
Anticorrelation between charge of the spectator fragments (FW) and particle multiplicity (hits)

A number of produced protons is stronger correlated with the number of produced particles (track & RPC+TOF hits) than with the total charge of spectator fragments (FW)

HADES; Phys.Rev.C 102 (2020) 2, 024914



HADES; Phys.Rev.C 102 (2020) 2, 024914



Avoid self-correlation biases when using spectators fragments for centrality estimation