Event shape analysis in ultrarelativistic nuclear collisions

Renata Kopečná in collaboration with Boris Tomášik

Faculty of Nuclear Sciences and Physical Engineering CTU in Prague

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Outline

Motivation

• Event shape engineering

ESSTER

- Toy Model Monte Carlo Generator
- Event Shape Analysis

Results

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Motivation

- Every event undergoes different initial conditions, different fluctuations
- Fluctuations lost during summation
- Mixing really similar events
- Single event femtoscopy?
- Studying effects of the initial geometry
 - Sorting events according to different initial conditions
 - Centrality: problematic for small ranges
 - Differently sized and shaped nuclei
- What is a good observable for 'sorting' events?

Event Shape Engineering

- Two subevents
 - Subevent a: event selection
 - Subevent b: physical analysis
- Helps avoiding nonphysical biases (nonflow effects)
- Information loss
- Event selection according to the magnitude of the *reduced flow vector* q_n

$$\vec{Q}_n = \left(\sum_{i=1}^M \cos(n\phi_i), \sum_{i=1}^M \sin(n\phi_i)\right),$$

$$q_n = |ec{Q}_n|/\sqrt{M}$$
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J. Schukraft, A. Timmins, S. A. Voloshin Phys. Lett. B 719 (2013) 394-398

Ordering example



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Ordering example



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Ordering example

- Color: Same v₂
- Line/Dashed: Same *v*₃
- Row: Same Ψ_3



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ESSTER (Event Shape SorTER)

- Toy Model MC generator
 - Azimuthal angle distribution $\frac{dN}{d\phi} = \frac{N}{2\pi} \left(1 + \sum_{n=1}^{\infty} 2v_n \cos[n(\phi \Psi_n)]\right)$
 - *M* ∈ (300, 3000)
 - Flow multiplicity dependent: $v_n = a_n M^2 + b_n M + c_n$
 - Gaussian smearing
 - Ψ_n independent, uniform distribution
 - Generated 5000 events



Event-Sorting

- Order events according the similarity of the shape of their azimuthal angle histogram using Bayesian statistics
- Dividing events into deciles according to a chosen variable
- Final arrangement in the terms of $\hat{\mu}$ (1 10)

$$\hat{\mu} = \sum_{\mu} \mu P(\mu | \{ n_i \})$$

- Initial assignment error matrix
- Correlation of $\hat{\mu}$ and several variables
- How to initially rotate events?

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Average histograms, simple example with v_1 and v_2 only



Average histograms, simple example after sorting



Simple example: comparison with q_2

- Initial rotation: Ψ_2
- Sort: *q*₂
- Each point represents one event
- Lines around integers: Distinct bins



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Simple example: Errors for initially q_2 sorted events

- Error of the 'original sorting variable' (in this case q₂) is the probability that event from bin β (before) should be in bin α (after)
- If the original sorting was good, we expect \sim 1 around diagonal, \sim 0 elsewhere.
- Not a diagonal
- Possible influence of quadratic dependence v₂ = v₂(M²)
- Dark corners: finite number of bins



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Simple example: Elliptic flow

- Correlation v_2 and $\hat{\mu}$: 0.959
- Obvious linear dependence
- v_2 might be a better measure than q_2



Simple example: Elliptic flow

- Correlation v_2 and $\hat{\mu}$: 0.959
- Obvious linear dependence
- v_2 might be a better measure than q_2
- Does not depend on initial sorting! (showing random initial assignment)



Advanced example: Flow up to v_5

- Included flow up to v_5
- Initial rotation: Ψ_2
- Linear v₂ dependence lost
 - Interplay of other flow contributions



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Advanced example: Average histograms before sorting



Advanced example: Average histograms after sorting



Rotation according to Ψ_2 and Ψ_3

- Included flow up to v₅
- Initial rotation:
 - Angle bisector between 𝒱₂ and 𝒱₃
 - Ψ₂ less than π/2 counterclockwise from Ψ₃
 - Flip problem solved



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Average histograms before sorting



Average histograms after sorting



Conclusion & Outlook

- ESSTER sorts events based on their similarity: appropriate sorting
- q_2 might not be a good measure to sort events
- v_2 might be a better measure than q_2
- Femtoscopy, U+U, Au+Co, He+Au collisions
- Further study of higher harmonics, histogram rotation, AMPT or other models, implementation improvement
- Renata Kopečná, Boris Tomášik: Event shape sorting arXiv: 1506.06776

Backup



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Multiplicity for initially q_2 sorted events

- Does not depend on multiplicity, only *shape*!
- Follows v_2 distribution





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- For every event make $\frac{dN}{d\phi}$ histogram
- Denoting bins as *i*
- Order events according to something (q₂) into deciles

 Calculate the probability that particle is in *ith* bin given the event is in event-bin μ:



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$$P(i|\mu) = \frac{\text{\# of particles in } i^{th} \text{ bin for all events in } \mu}{\text{\# of all particles in all events in } \mu}$$

- Full description of any event *α* in *μ* is set of numbers {*n_i*}
 - Event α in μ is 'binned' according to $\frac{dN}{d\phi}$ as $\{n_1, \dots, n_i, \dots, n_{\#bin}\}$
- For each event we calculate the probability that event in bin μ will have dN/dφ histogram with n_i particles in each angle bin:

$$P(\{n_i\}|\mu) = N! \prod_i \frac{P(i|\mu)^{n_i}}{n_i!}$$



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 We want to know the probability that an event with record {n_i} belongs to the bin μ

$$P(\mu|\{n_i\}) = \frac{P(\{n_i\}|\mu)p(\mu)}{P(\{n_i\})} = \frac{P(\{n_i\}|\mu)p(\mu)}{\sum_{\mu'} P(\{n_i\}|\mu')p(\mu')}$$

• $p(\mu)$ is a *prior*; for deciles $p(\mu) = 1/10$

• This probability uses *all* data: fluctuations caused by 'rare' events are reduced



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• For every event we calculate 'average bin number'

$$\hat{\mu} = \sum_{\mu} \mu P(\mu | \{ n_i \})$$

• Sort according to $\hat{\mu}$



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• Return to 1)

• Repeat until the μ bins remains unchanged



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