

# Prospects of Event Shape Sorting

Jakub Cimerman, Boris Tomášik

3 December 2019

## ZIMÁNYI SCHOOL '19



Győrfi András: Az úton (On the road)

**19. ZIMÁNYI SCHOOL**  
**WINTER WORKSHOP ON**  
**HEAVY ION PHYSICS**

**Dec. 2. - Dec. 6.,**  
**Budapest, Hungary**



József Zimányi (1931 - 2006)

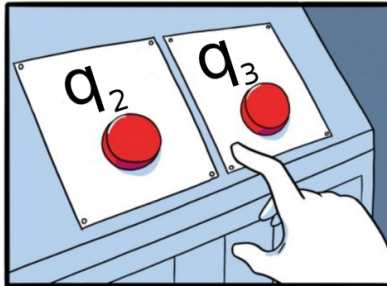
Have you ever been told by someone  
"You're doing too much averaging in that analysis. You lose a lot of  
information about anisotropies and so on."?



No problem. I'll just use

# Event Shape Engineering

But according to which variable I sort the events?



JAKE-CLARK.TUMBLR



Have you recognized yourself in this?

Then we have a perfect tool for you

Have you recognized yourself in this?

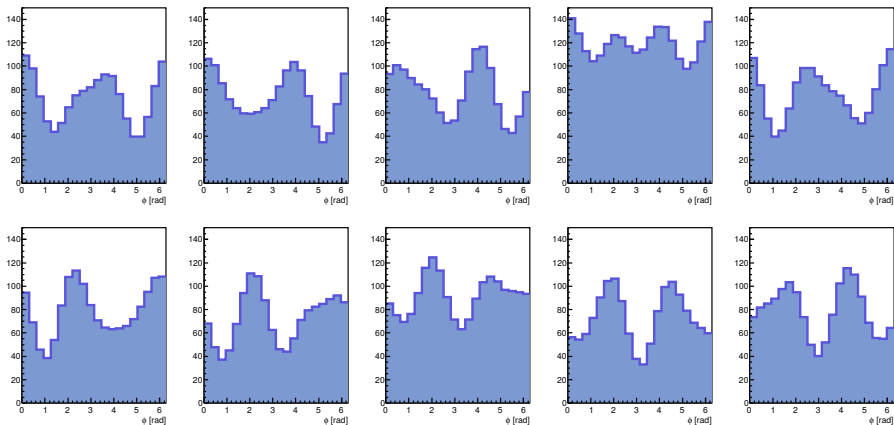
Then we have a perfect tool for you

Event Shape ~~Engineering~~  
Sorting

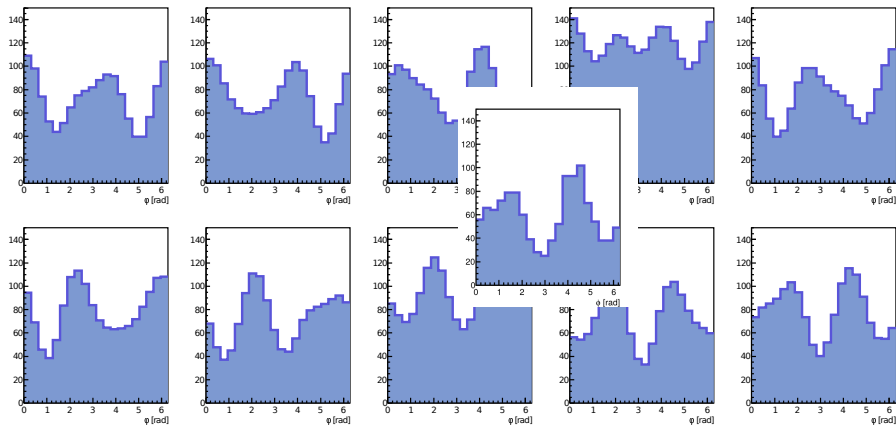
# Event Shape Sorting vs Event Shape Engineering

- Both are sorting algorithms
- ESE sorts events according to one chosen property ( $q_2, q_3, \dots$ ), but all the other properties will be averaged out
- ESS sorts events according to similarity of their shapes in a more complex way, so you can see combination of anisotropies of different orders at the same time

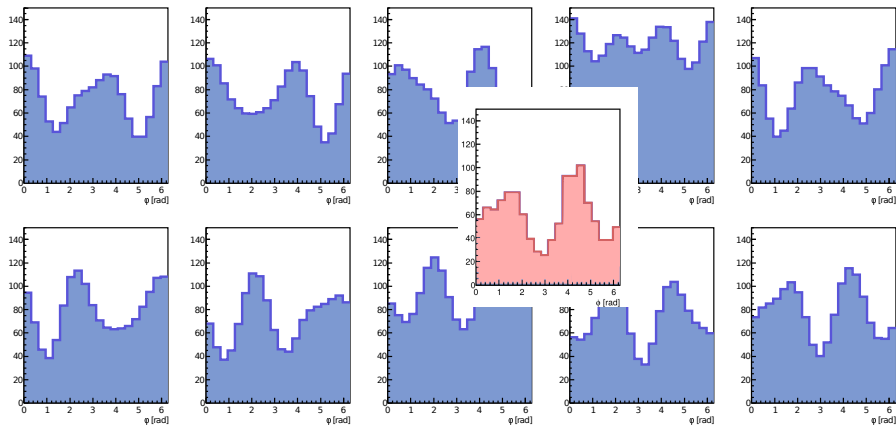
# Event Shape Sorting Algorithm



# Event Shape Sorting Algorithm

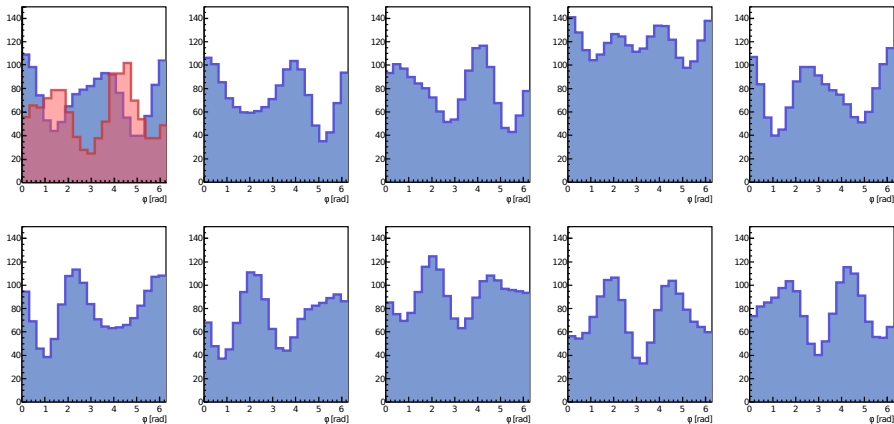


# Event Shape Sorting Algorithm

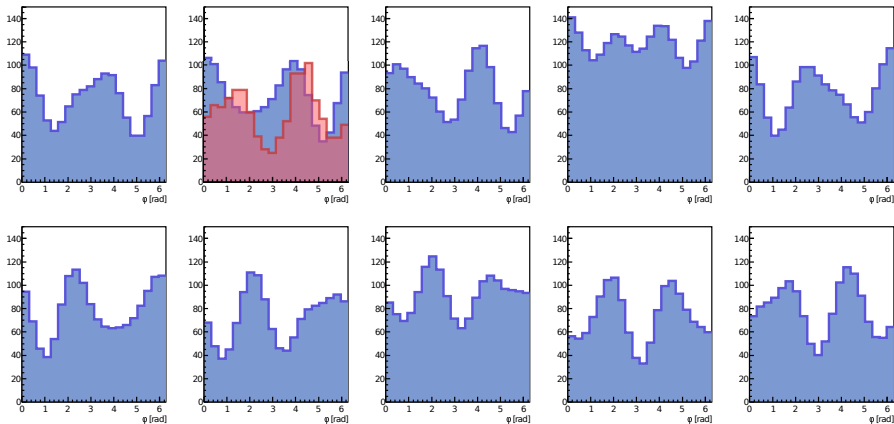




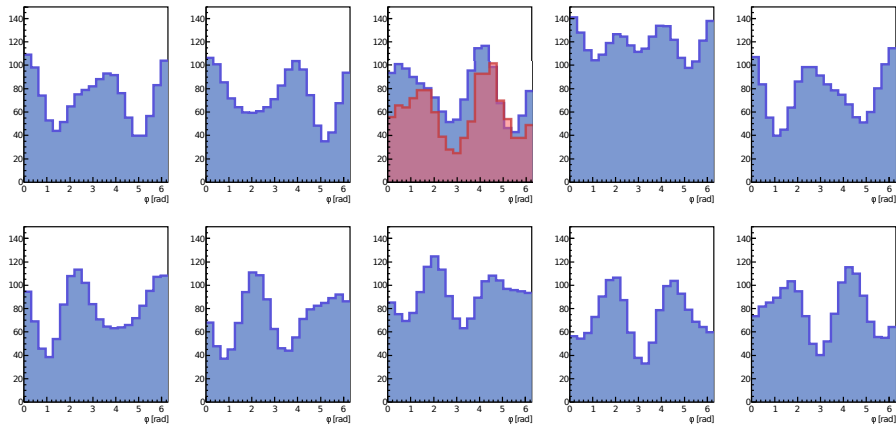
# Event Shape Sorting Algorithm



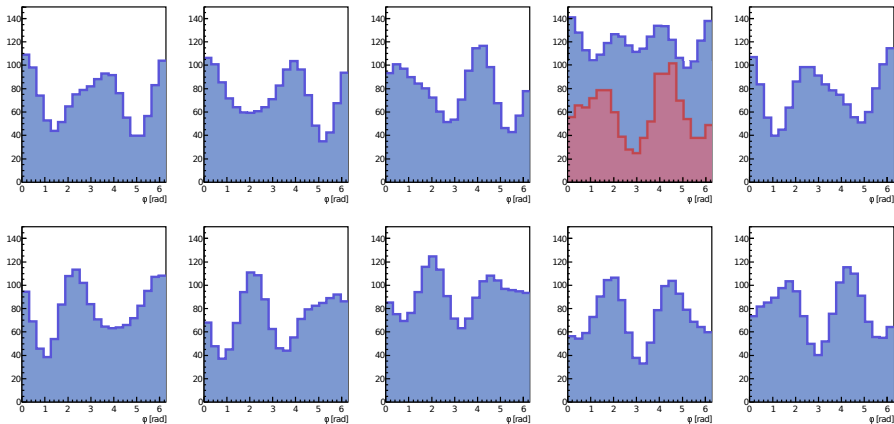
# Event Shape Sorting Algorithm



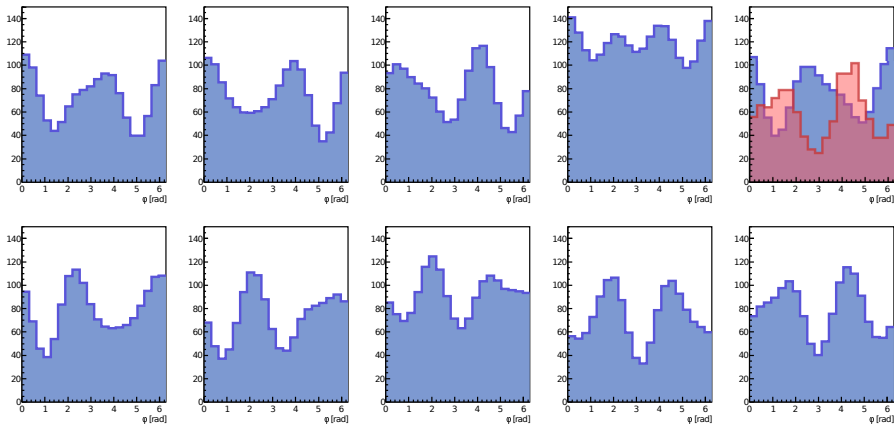
# Event Shape Sorting Algorithm



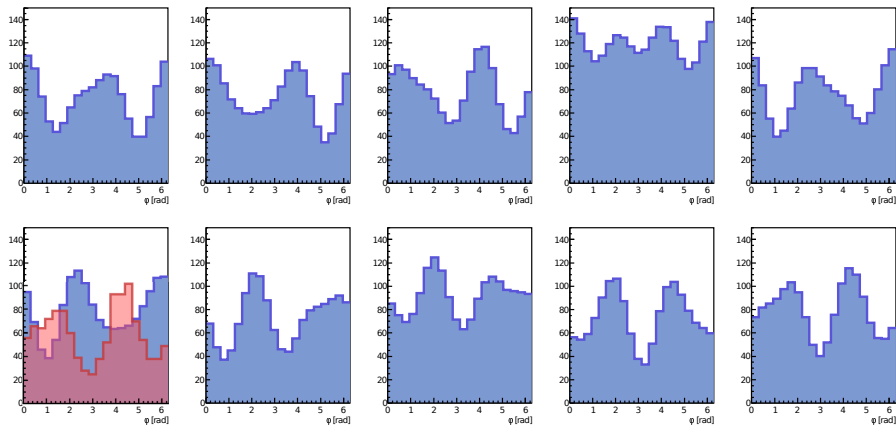
# Event Shape Sorting Algorithm



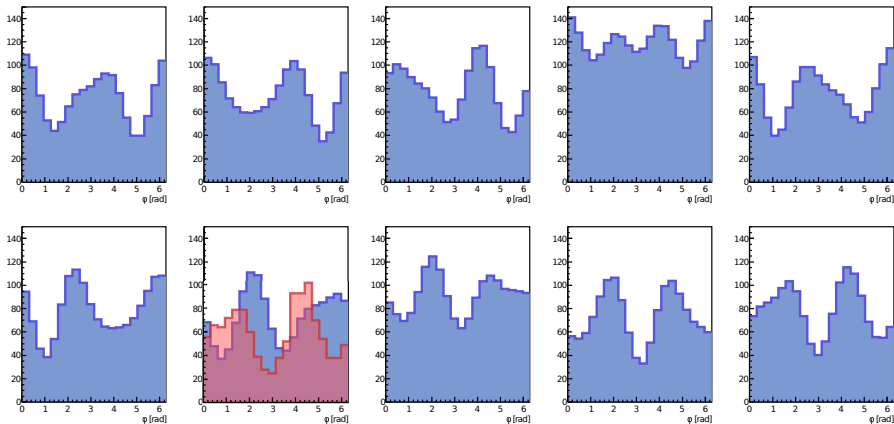
# Event Shape Sorting Algorithm



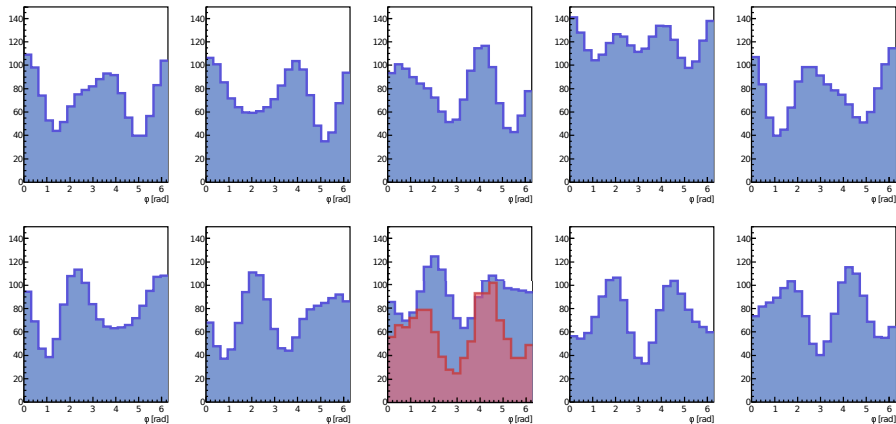
# Event Shape Sorting Algorithm



# Event Shape Sorting Algorithm

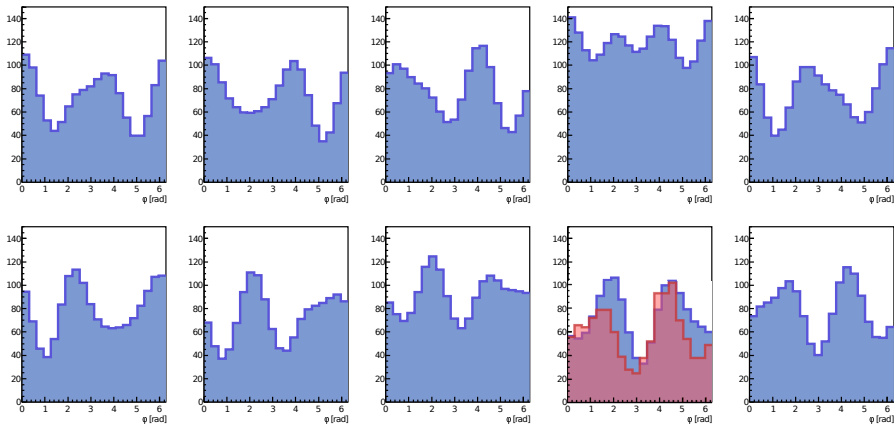


# Event Shape Sorting Algorithm

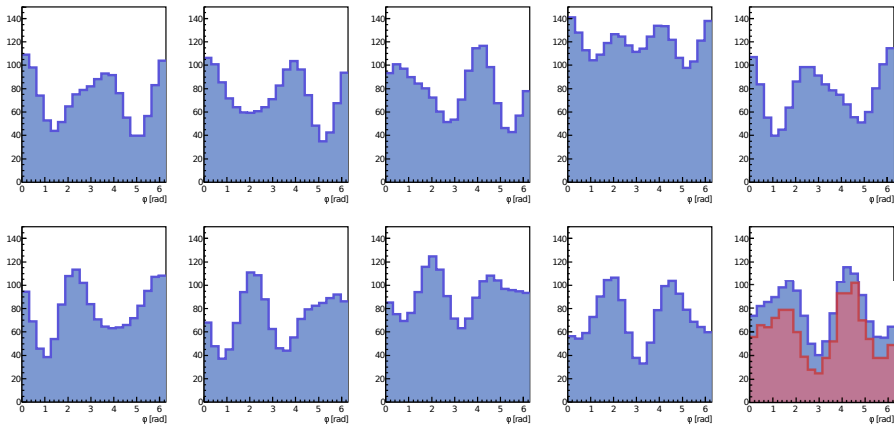




# Event Shape Sorting Algorithm



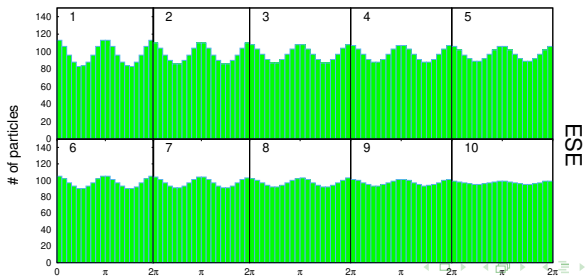
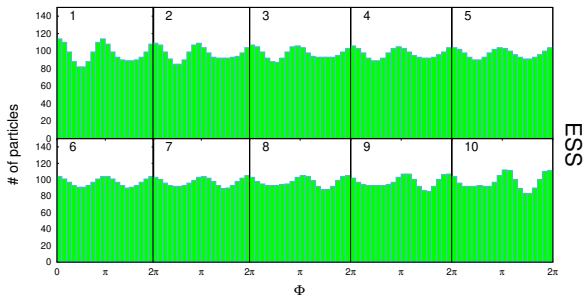
# Event Shape Sorting Algorithm

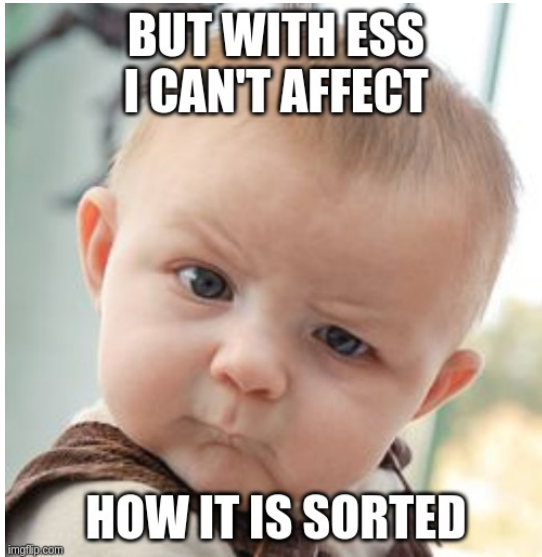


# Animation of Event Shape Sorting

# Animation of Event Shape Sorting

# Results of Sorting: ESS vs. ESE





But that is a good thing

But that is a good thing

ESS can find things you don't know are there



But that is a good thing

ESS can find things you don't know are there

Let's take a look at some example

# Correlation between the Second and Third-Order Event Plane

- For uncorrelated event planes we get

$$\left\langle e^{i\{\theta_2 - \theta_3\}_{min}} \right\rangle = \frac{\int_0^\pi d\theta_2 \int_0^{\frac{2\pi}{3}} d\theta_3 e^{i\{\theta_2 - \theta_3\}_{min}}}{\int_0^\pi d\theta_2 \int_0^{\frac{2\pi}{3}} d\theta_3} = \frac{3}{\pi}$$

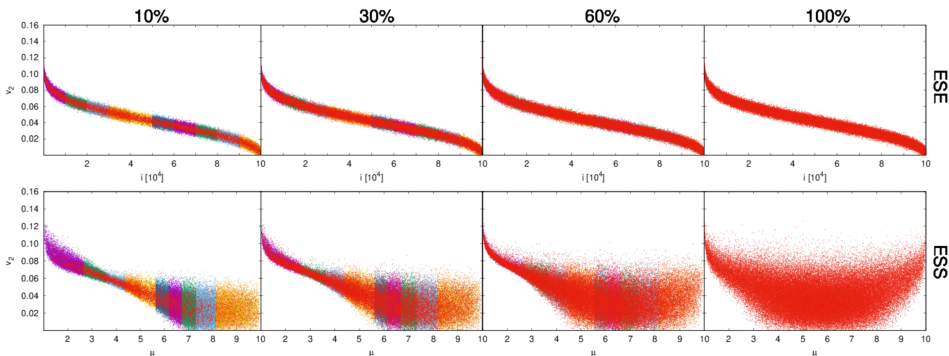
- The same value of the correlator can be obtained with event planes which only differ by one of two possible values

$$\left\langle e^{i\{\theta_2 - \theta_3\}_{min}} \right\rangle = \frac{1}{2} \left( e^{i\delta_1} + e^{i\delta_2} \right) = \frac{3}{\pi}$$

$$\delta_{1,2} = \pm 0.301374$$

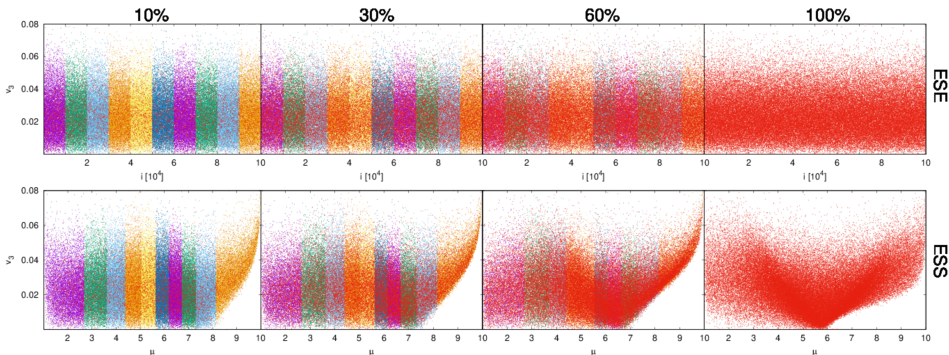
- To illustrate this, we generated 100 000 events using DRAGON with 10%, 30%, 60% and 100% events with correlated event planes

# Amplitude of Second Order Anisotropy $v_2$



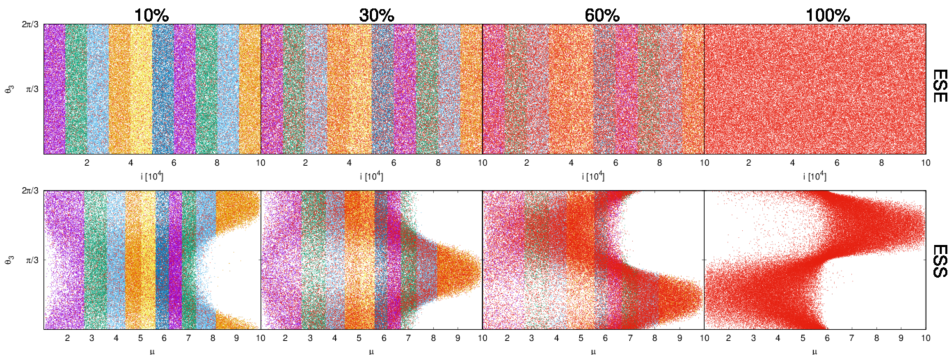
- Correlated events are tagged with red dots
- $\mu$  is the sorting variable in ESS algorithm

# Amplitude of Third Order Anisotropy $v_3$



- Correlated events are tagged with red dots
- $\mu$  is the sorting variable in ESS algorithm

# Angle of Rotation of the Third Order Plane $\theta_3$



- Correlated events are tagged with red dots
- $\mu$  is the sorting variable in ESS algorithm

# Conclusions

- We have shown the advantages of ESS against ESE
  - ESS takes into account both second- and third-order anisotropies at the same time, which ESE can not
  - ESS can find out some informations which can not be found using ESE



Backup slides

# Event Shape Sorting algorithm

- 1. Split particles in each event into  $k$  azimuthal angle bins
- 2. Initial sorting according to  $|\vec{q}_2|$
- 3. Calculate  $P(i|\mu)$  for each angle bin in each class

$$P(i|\mu) = \frac{\sum_{\text{events in } \mu\text{-th class}} (n_i)_j}{\sum_{\text{events in } \mu\text{-th class}} N_j}$$

- 4. Calculate  $P(\mu | \{n_i\}_j)$  for each event

$$P(\mu | \{n_i\}_j) = \frac{\prod_{i=1}^k P(i|\mu)^{(n_i)_j}}{\sum_{\mu'=1}^{\omega} \prod_{i=1}^k P(i|\mu')^{(n_i)_j}}$$

- 5. Calculate mean class number  $\bar{\mu}$

$$\bar{\mu} = \sum_{\mu=1}^{\omega} \mu P(\mu | \{n_i\}_j)$$

- 6. Sort events according to  $\bar{\mu}$
- 7. Repeat from step 3, until order of events stay unchanged



- How statistical fluctuations influence the sorting?
- To find out we split the particles according to their pseudorapidity into reference particles  $|\eta| < 0.4$  and test particles  $0.5 < |\eta| < 0.8$
- This was done with three sets of simulated data
  - **DRAGON:** 150 000 events with anisotropies  $a_2, \rho_2 \in (-0.1; 0.1)$ ,  $a_3, \rho_3 \in (-0.03; 0.03)$
  - **UrQMD:** 100 000 events in Au+Au collisions,  $\sqrt{s_{NN}} = 200$  GeV, impact parameter  $b \in (7; 10)$  fm
  - **AMPT:** 10 000 events in Au+Au collisions,  $\sqrt{s_{NN}} = 200$  GeV, impact parameter  $b \in (7; 10)$  fm

# Amplitude of Second Order Anisotropy $v_2$

