Towards a more robust intermittency analysis technique in heavy ion collisions: achievements and challenges.

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Robust intermittency analysis

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Outline

- QCD Phase Diagram and Critical Phenomena
- 2 Intermittency analysis methodology
- Intermittency analysis results
- 4 Challenges & possible solutions in intermittency analysis
- 5 Critical Monte Carlo Simulations
- 6 Assessing models through PCA
- 7 Conclusions & Outlook

The phase diagram of QCD



- Phase transitions from hadronic matter to quark-gluon plasma:
 - Low μ_B & high $T \rightarrow$ cross-over (lattice QCD)
 - High µ_B & low T → 1st order (effective models)
 - \Rightarrow 1st order transition line ends at Critical Point (CP) \rightarrow 2nd order transition
- At the CP: scale-invariance, universality, collective modes ⇒ good physics signatures



- Detection of the QCD Critical Point (CP): Main goal of many heavy-ion collision experiments (in particular the SPS NA61/SHINE experiment)
- Look for observables tailored for the CP; Scan phase diagram by varying energy and size of collision system.

Critical Observables & the Order Parameter (OP)



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Self-similar density fluctuations near the CP



Observing power-law fluctuations through intermittency



[Csorgo, Tamas, PoS CPOD2009 (2009) 035]

Experimental observation of local, power-law distributed fluctuations of net baryon density

Intermittency in transverse momentum space at mid-rapidity (Critical opalescence in ion collisions)

[F.K. Diakonos, N.G. Antoniou and G. Mavromanolakis, PoS (CPOD2006) 010, Florence]

• Net proton density carries the same critical fluctuations as the net baryon density, and can be substituted for it.

[Y. Hatta and M. A. Stephanov, PRL91, 102003 (2003)]

• Furthermore, antiprotons can be ignored (their multiplicity is negligible compared to protons), and we can analyze just the proton density.

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Observing power-law fluctuations: Factorial moments

- Pioneered by Białas and others, as a method to detect non-trivial dynamical fluctuations in high energy nuclear collisions
- Transverse momentum space is partitioned into *M*² cells
- Calculate second factorial moments F₂(M) as a function of cell size ⇔ number of cells M:

$$F_2(M) \equiv \frac{\left\langle \frac{1}{M^2} \sum_{i=1}^{M^2} n_i (n_i - 1) \right\rangle}{\left\langle \frac{1}{M^2} \sum_{i=1}^{M^2} n_i \right\rangle^2},$$

where $\langle \ldots \rangle$ denotes averaging over events.

[A. Bialas and R. Peschanski, Nucl. Phys. B 273 (1986) 703-718]
 [A. Bialas and R. Peschanski, Nucl. Phys. B 308 (1988) 857-867]
 [J. Wosiek, Acta Phys. Polon. B 19 (1988) 863-869]
 [A. Bialas and R. Hwa, Phys. Lett. B 253 (1991) 436-438]
 [Z. Burda, K. Zalewski, R. Peschanski, J. Wosiek, Phys. Lett. B 314 (1993) 74-78]



 $p_{x,y}$ range in present analysis: -1.5 $\leq p_{x,y} \leq$ 1.5 GeV/c $M^2 \sim$ 10 000

Background subtraction – the correlator $\Delta F_2(M)$

Background of non-critical pairs must be subtracted from experimental data;

Partitioning of pairs into critical/background

$$\langle n(n-1)\rangle = \underbrace{\langle n_c(n_c-1)\rangle}_{\text{critical}} + \underbrace{\langle n_b(n_b-1)\rangle}_{\text{background}} + \underbrace{2\langle n_b n_c\rangle}_{\text{cross term}}$$
$$\underbrace{\Delta F_2(M)}_{\text{correlator}} = \underbrace{F_2^{(d)}(M)}_{\text{data}} - \lambda(M)^2 \cdot \underbrace{F_2^{(b)}(M)}_{\text{background}} - 2 \cdot \underbrace{\lambda(M)}_{\text{ratio}} \cdot (1 - \lambda(M)) f_{bc}$$

• If $\lambda(M) \leq 1$ (dominant background) \Rightarrow cross term negligible & $F_2^{(b)}(M) \sim F_2^{\min}(M)$ (Critical Monte Carlo* simulations), then:

 φ_2 : intermittency index

Theoretical prediction^{*} for
$$\varphi_2$$

 $\varphi_{2,cr}^{(p)} = \frac{5}{6} (0.833...)$

*[Antoniou *et al*, PRL **97**, 032002 (2006)]

Intermittency restored in
$$\Delta F_2(M)$$
:

 $\Delta F_2(M) \simeq F_2^{\text{data}}(M) - F_2^{\text{mix}}(M)$

$$\Delta F_2(M) \sim \left(M^2\right)^{\varphi_2}, \ M \gg 1$$

The correlation integral C(R) as an aid to intermittency

• A computationally faster alternative to lattice averaging on a fixed grid, the correlation integral is defined as:

$$C(R) = \frac{2}{\langle N_{mul} (N_{mul} - 1) \rangle_{ev}} \left\langle \sum_{\substack{i,j \\ i < j}} \Theta \left(|x_i - x_j| \le R \right) \right\rangle_{ev}$$

[P. Grassberger and I. Procaccia (1983). "Measuring the strangeness of strange attractors". Physica. 9D: 189–208]
 [F. K. Diakonos and A. S. Kapoyannis, Eur. Phys. J. C 82, 200 (2022)]



*F*₂(*M*) can be obtained from *C*(*R*), or vice-versa, by the relations:

$$C(R_M) = \frac{\langle N_{mul} \rangle_{ev}^2}{\langle N_{mul} (N_{mul} - 1) \rangle_{ev}} \frac{F_2(M)}{M^2}$$
$$F_2(M) = \frac{\langle N_{mul} (N_{mul} - 1) \rangle_{ev}}{\langle N_{mul} \rangle_{ev}^2} M^2 C(R_M),$$

where $\pi R_M^2 = a^2$.

NA49 C+C, Si+Si, Pb+Pb @ $\sqrt{s_{NN}} \simeq 17$ GeV – protons

Factorial moments of proton transverse momenta analyzed at mid-rapidity



[T. Anticic et al., Eur. Phys. J. C 75:587 (2015), arXiv:1208.5292v5]

• $F_2(M)$, $\Delta F_2(M)$ errors estimated by the bootstrap method [W.J. Metzger, "Estimating the Uncertainties of Factorial Moments", HEN-455 (2004).]

- Fit with $\Delta F_2^{(e)}(M \ ; \ C, \phi_2) = 10^C \cdot \left(\frac{M^2}{M_0^2}\right)^{\phi_2}$, for $M^2 \ge 6000 \ (M_0^2 \equiv 10^4)$
- Evidence for intermittency in "Si"+Si but large statistical errors.

NA61/SHINE intermittency: ⁷Be + ⁹Be @ $\sqrt{s_{NN}} \simeq 17$ GeV

- Intermittency analysis is pursued within the framework of the NA61/SHINE experiment, inspired by the positive, if ambiguous, NA49 Si+Si result.
 [T. Anticic et al., Eur. Phys. J. C 75:587 (2015), arXiv:1208.5292v5]
- Two NA61/SHINE systems were initially examined: ⁷Be + ⁹Be and ⁴⁰Ar + ⁴⁵Sc @ 150A GeV/c ($\sqrt{s_{NN}} \simeq 17$ GeV)



- $F_2(M)$ of data and mixed events **overlap** \Rightarrow
- Subtracted moments $\Delta F_2(M)$ fluctuate around zero \Rightarrow No intermittency effect is observed in Be+Be.

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NA61/SHINE 40 Ar + 45 Sc @ $\sqrt{s_{NN}} \simeq 17$ GeV

- First indication of intermittency in mid-central Ar+Sc 150A GeV/c collisions (CPOD2018); In 2019, an extended event statistics set was analysed;
- A scan in centrality was performed (maximum range: 0-20% most central), as centrality may influence the system's freeze-out temperature;
- Event statistics: ~ 400K events per 10% centrality interval;



- Some signal indication in c.10-20% ("mid-central"), but this (final) correlated bin analysis was inconclusive;
- Afterwards, NA61/SHINE turned to **independent bin analysis** (more on this later...)

Challenges in proton intermittency analysis

- Particle species, especially protons, cannot be perfectly identified experimentally; candidates will always contain a small percentage of impurities;
- Experimental momentum resolution sets a limit to how small a bin size (large M) we can probe;
- A finite (small) number of usable events is available for analysis; the "infinite statistics" behaviour of ΔF₂(M) must be extracted from these;
- Proton multiplicity for medium-size systems is low (typically ~ 2 3 protons per event, in the window of analysis) and the demand for high proton purity lowers it still more;
- M-bins are correlated the same events are used to calculate all F₂(M)! This biases fits for the intermittency index φ₂, and makes confidence interval estimation hard.

Intermittency analysis tools: correlated fit

Possible to perform correlated fits for φ₂, with *M*-correlation matrix estimated via bootstrap;
 Correlated fit
 Uncorrelated fit

NA61/SHINE Ar+Sc 150, cent.10 - 20%, pur > 90% NA61/SHINE Ar+Sc 150, cent.10 - 20%, pur > 90% 1.5corr.fit, $\phi_2 = 0.553 \pm 0.283$ fit, $\phi_2 = 0.389 \pm 0.073$ ndf = 72.4 / 67 /ndf = 8.3 / 67 99.7% C I NA61/SHINE Ar+Sc 150 #0 NA61/SHINE Ar+Sc 150 #0 0.5 0.5 ∆F₂(M) $\Delta F_2(M)$ -0.5 -0.5 5000 10000 15000 20000 5000 10000 15000 20000

- Replication of events means bootstrap sets are not independent of the original: magnitude of variance and covariance estimates can be trusted, but central values will be biased to the original sample;
- Correlated fits for φ₂ are known to be unstable;

[B. Wosiek, APP B21, 1021 (1990); C. Michael, PRD49, 2616 (1994)]

The approach of independent bins decimates event statistics.

[NA61/SHINE Collaboration, EPJC 83 (2023) 881; EPJC 84 (2024) 7, 741]

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Intermittency analysis tools: Monte Carlo model scan

Avoid fitting, use model weighting!

- Build Monte Carlo models incorporating background & fluctuations;
- Compare them against experimental moments ΔF₂(M);
- Models are parametrized in critical exponent strenght (φ₂ value), critical component (% of critical to total protons), and possibly other parameters (e.g. detector effects);
- Ideally, a wide scan of model parameters should be performed against the experimental data.



Critical Monte Carlo (CMC) algorithm for baryons

- Simplified version of CMC* code:
 - Only protons produced;
 - One cluster per event, produced by sampling random Lévy walk of adjustable dimension d_F, e.g.: d^B_F = 1/3 ⇒ φ₂ = 1 − d^B_F/2 = 5/6
 - Lower / upper bounds of Lévy walks *p_{min,max}* plugged in;
 - Cluster center adjustable to experimental set mean proton p_T per event;
 - **Poissonian** proton multiplicity distribution.

Input parameters (example)

Parameter	$p_{\min} ({\rm MeV})$	p_{\max} (MeV)	$\lambda_{Poisson}$
Value	0.1 → 1	800 → 1200	$\langle ho angle_{ m non-empty}$

*[Antoniou, Diakonos, Kapoyannis and Kousouris, Phys. Rev. Lett. 97, 032002 (2006).]



CMC – background simulation & detector effects

- Non-critical background simulation: replace critical tracks by uncorrelated (random) tracks, with fixed probability: $\mathcal{P}_{track} = 1 \mathcal{P}_{crit}$, where \mathcal{P}_{crit} is the percentage of critical component;
- *p*_T distribution of background tracks plugged in to match experimental data;
- y_{CM} rapidity value generated orthogonal to p_T, matching experimental distribution;
- *p_T*, *y_{CM}*, quality & acceptance cuts applied, modeled after NA61/SHINE cuts; [NA61/SHINE Collaboration, EPJC 83 (2023) 881]



CMC scan $\Delta F_2(M)$ – examples

- Results shown for CMC ΔF₂(M), with (p) = 2.562, corresponding to SHINE Ar+Sc @ 150A GeV/c, cent.10-20%;
- 2 settings:
 - $\phi_2 = 0.125$, crit.% = 1.60%;
 - 2 $\phi_2 = 0.750$, crit.% = 1.60%;
- For each setting, ~ 8K independent samples of ~ 400K events are generated; event statistics selected to match SHINE data.



Weighting models: Goodness-of-fit function



 Calculate the residuals for each bin M_i between model & experiment:

$$res(M_i) \equiv -\frac{F_2^{\text{exper.}}(M_i) - F_2^{\text{model}}(M_i)}{1\sigma},$$

 $\sigma \sim$ uncertainties (e.g. by bootstrap);



Scan parameter space, weighting models on a grid.

• Weight models by χ^2 metric:

$$\chi^{2} = \sum_{i} res^{2}(M_{i}) \Rightarrow$$
Model Weight ~ $e^{-\frac{\chi^{2}}{2}}$

Handling bin correlations through PCA

- While CMC samples (events) are independent, *M*-bins in a sample are not; they are strongly correlated;
- Additionally, there are ~ 150 bins, i.e. dimensions to consider, and we have
 N_s = 8K independent samples too few to probe the joint distribution;
- We need to reduce the effective dimensionality and untangle correlations;
- We can do this via Principal Component Analysis (PCA): center and scale sample points in *M*-space, then rotate the axes to make independent linear combinations of *M*-bins. Finally, keep only few significant components.



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Performing PCA on CMC vs itself

CMC, $\langle p \rangle = 2.562$, crit.=1.00%, $\varphi_2 = 0.825$



• PCA decouples bins; χ^2 of CMC sample vs itself can be summed per PC.

CMC model scan & the role of event statistics

- In order to illustrate the behavior of CMC samples vs CMC models, it is useful to focus on one single row of the parameter grid, roughly corresponding to critical $\phi_2 = 5/6$;
- We choose φ₂ = 0.825, and plot the p-values along this row of ~ 1600 samples (CMC simulations) of:
 - OMC with crit. comp. = 0%, ϕ_2 = 0.1 (no signal case)
 - Summarial CMC with crit. comp. = 1%, ϕ_2 = 0.825 (signal case)
- The result is ~ 1600 curves of p-value as a function of % crit. comp., and we study their behavior: peaks correspond to more likely models (% crit. comp.); the width of a peak indicates the resolution (power) of our scan in telling models apart;
- We then repeat the scan with ~ 1600 CMC samples of ~ 4M events each, i.e. 10× the usual Ar+Sc @ 150 GeV/c SHINE event statistics;
- We want to investigate the effect of larger event statistics on the accuracy of the scan; ideally, the model scan should produce curves narrowly peaked around the true % crit. comp.

Scan of models - normal & 10× stats

Example p-value curves, (No signal, PCA)



Example p-value curves, (1% signal, PCA)



• Almost twice the % crit. comp. resolution for 10× statistics!

×10 stats

normal stats

Conclusions & Outlook

- Proton intermittency analysis is a promising tool for detecting the critical point of strongly interacting matter; however, large uncertainties and bin correlations cannot be handled by the conventional analysis methods;
- The Principal Component Analysis technique is able to handle statistical & systematic uncertainties, as well as bin correlations, without sacrificing event statistics;
- We build suitable Monte Carlo models and evaluate them against data via a scan in parameter space; rotation from original bins to principal components ensures that bin correlations do not invalidate the analysis;



Conclusions & Outlook

- Critical Monte Carlo (CMC) investigation of PCA analysis performance indicates that large F₂(M) fluctuations are inherent to proton intermittency analysis; thus, the effect of event statistics on model parameter scan resolution is decisive;
- We cannot distinguish between no-signal and reasonable (≤ 2%) levels of critical component, with a statistics of ~ 500K events;
- We could discern as weak as a ~ 1% critical component from no-signal with 10× statistics, i.e. ~ 5M events; this is well within reach of e.g. the NA61/SHINE experiment;
- Results of this study soon to appear in methodological paper(s), along with extensive discussion; stay tuned! :-)



Thank You!



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Backup Slides

Backup Slides Outline

NA49 intermittency results

- NA61/SHINE intermittency results
- Oritical Monte Carlo
- 1 Intermittency analysis challenges
- 12 Remedies to intermittency problems
- 3 2D exclusion plots
- Selecting the optimal number of PCs

NA49 C+C, Si+Si, Pb+Pb @ $\sqrt{s_{NN}} \simeq 17$ GeV – dipions

• 3 sets of NA49 collision systems at 158A GeV/c ($\sqrt{s_{NN}} \simeq 17$ GeV)

[T. Anticic et al, Phys. Rev. C 81, 064907 (2010); T. Anticic et al., Eur. Phys. J. C 75:587 (2015)]

 Intermittent behaviour (φ₂^(σ) ≃ 0.35) of dipion pairs (π⁺, π⁻) in transverse momentum space observed in central Si+Si collisions at 158A GeV.



[T. Anticic et al, Phys. Rev. C 81, 064907 (2010)]

 No such power-law behaviour observed in central C+C and Pb+Pb collisions at the same energy.

NA49 C+C, Si+Si, Pb+Pb @ $\sqrt{s_{NN}} \simeq 17$ GeV



• No intermittency detected in the "C"+C, Pb+Pb datasets.

NA49 C+C, Si+Si, Pb+Pb @ $\sqrt{s_{NN}} \simeq 17$ GeV

- Evidence for intermittency in "Si"+Si but large statistical errors.
- Distribution of φ_2 values, $P(\varphi_2)$, and confidence intervals for φ_2 obtained by fitting individual bootstrap samples [B. Efron, *The Annals of Statistics* 7,1 (1979)]



- Bootstrap distribution of ϕ_2 values is highly asymmetric (due to closeness of $F_2^{(d)}(M)$ to $F_2^{(m)}(M)$).
- Uncorrelated fits used, but errors between M are correlated!
- Estimated intermittency index: $\phi_{2,B} = 0.96^{+0.38}_{-0.25}$ (stat.) ± 0.16(syst.)

[T. Anticic et al., Eur. Phys. J. C 75:587 (2015), arXiv:1208.5292v5]

The NA61/SHINE experiment



- Fixed-target, high-energy collision experiment at CERN SPS;
- Reconstruction & identification of emitted protons in an extended regime of rapidity, with precise evaluation of their momentum vector;
- Centrality of the collision measured by a forward Projectile Spectator Detector (PSD);

- Direct continuation of NA49
- Search for Critical Point signatures



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Independent bin analysis with cumulative variables

- M-bin correlations complicate uncertainties estimations for $\Delta F_2(M) \& \phi_2$; one way around this problem is to use independent bins – a different subset of events is used to calculate $F_2(M)$ for each M;
- Advantage: correlations are no longer a problem;
 Disadvantage: we break up statistics, and can only calculate F₂(M) for a handful of bins.
- Furthermore, instead of p_x and p_y, one can use cumulative quantities: [Bialas, Gazdzicki, PLB 252 (1990) 483]

$$Q_x(x) = \int_{min}^{x} P(x)dx \bigg| \int_{min}^{max} P(x)dx;$$
$$Q_y(x,y) = \int_{ymin}^{y} P(x,y)dy \bigg| P(x)$$

- transform any distribution into **uniform** one (0, 1);
- remove the dependence of F₂ on the shape of the single-particle distribution;
- approximately preserves ideal power-law correlation function. [Antoniou, Diakonos, https://indico.cern.ch/event/818624/]





P(x, y)

v10⁻³

SHINE ⁴⁰Ar + ⁴⁵Sc independent bin proton intermittency

No signal indicating the critical point



 $1^2 \leq M^2 \leq 150^2$

 $1^2 \leq M^2 \leq 32^2$

number of subdivisions in cumulative transverse momentum space

[NA61/SHINE, EPJC 83 (2023) 881] [NA61/SHINE, EPJC 84 (2024) 7, 741]

Simulating fractal sets through random Lévy walks

 In D-dimensional space, we can simulate a fractal set of dimension d_F, D - 1 < d_F < D, through a random walk with step size Δr distribution:



CMC model scan (zoomed)



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Intermittency analysis tools: the bootstrap

- Random sampling of events, with replacement, from the original set of events;
- k bootstrap samples (k ~ 1000) of the same number of events as the original sample;
- Each statistic (ΔF₂(M), φ₂) calculated for bootstrap samples as for the original; [B. Efron, *The Annals of Statistics* 7,1 (1979)]
- Variance of bootstrap values estimates standard error of statistic.





Proton selection

NA61/SHINE preliminary

NA61/SHINE preliminary



- Particle ID through energy loss dE/dx in the Time Projection Chambers (TPCs);
- Employ p_{tot} region where Bethe-Bloch bands do not overlap (3.98 GeV/c ≤ p_{tot} ≤ 126 GeV/c);
- Mid-rapidity region ($|y_{CM}| < 0.75$) selected for present analysis.

Momentum resolution: effect on intermittency





- CMC + background + Gaussian noise (5 MeV radius);
- A 5 MeV Gaussian error in p_x, p_y leads to ~ 10% discrepancy in the value of φ₂.
- For very large backround values (> 99%), momentum resolution matters little to the overall distortion.

AMIAS on NA49 & NA61/SHINE data – ϕ_2 vs $N_{wounded}$

- ϕ_2 AMIAS confidence intervals calculated for NA49 & NA61/SHINE systems with indications of intermittency
- Corresponding mean number of participating ("wounded") nucleons N_w estimated via geometrical Glauber model simulation



[N. G. Antoniou (N. Davis) et. al., Decoding the QCD critical behaviour in A + A collisions, NPA 1003 (2020) 122018]

Effect of event statistics on exclusion significance

- 2D plots comparing CMC with 1% signal to a grid of models in φ₂ and % crit. comp. ("exclusion plots");
- Plotted: percentage of samples that cannot be excluded at p-value > 0.05 level;



Selecting an optimal number of PCs

- We must select an optimal # of PCs; too few, and we lose information on the moments distribution; too many, and we retain noise from the particular set of samples;
- One criterion is to pick the # of PCs that minimizes the loss in reconstructing the original distribution from the PCs but we have to be cautious!



- We use the ΔF₂(M) values of all but one M-bin to predict the missing value in one sample ("leave-one-out" predictor) using the model; then we aggregate the score over all samples;
- Scores are cross-validated in sub-samples for added confidence;
- About \sim 35 components should be kept by leave-one-out metric.