# Partial correlation analysis in ultra-relativistic nuclear collisions

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NATIONAL SCIENCE CENTRE

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#### Outline

- Partial correlations (PC) analysis, physical and control random variables (meaning of centrality)
- PC in a superposition approach placing constraints on sources
- Test on a hydro solution: a working scheme



# Partial correlations

# Kindergarden

#### Sample of children:

- weight
- intelligence

Pearson's correlation matrix:

$$\rho = \left(\begin{array}{cc} 1 & 0.62\\ 0.62 & 1 \end{array}\right)$$

 $\rightarrow \rho(\text{weight}, \text{intelligence}) \simeq 0.6 - \text{large}$ 

Hints to wrong conclusions

[W. Krzanowski, Principles of Multivariate Analysis, Oxford U. Press, 2000]

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# Kindergarden

Sample of children:

- weight
- intelligence
- 3 age control (external, nuisance) variable

Pearson's correlation matrix:

$$\rho = \left(\begin{array}{ccc}
1 & 0.62 & 0.84 \\
0.62 & 1 & 0.74 \\
0.84 & 0.74 & 1
\end{array}\right)$$

 $\rightarrow \rho(\text{weight}, \text{intelligence}) \simeq 0.6 - \text{large}$ 

Partial correlation (defined shortly) gives  $\rho(\text{weight}, \text{intelligence} \bullet \text{age}) \simeq 0$ 

[W. Krzanowski, Principles of Multivariate Analysis, Oxford U. Press, 2000]

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#### Partial correlation

Two physical variables X, Y and one control variable Z:

$$c(X, Y \bullet Z) = c(X, Y) - \frac{c(X, Z)c(Z, Y)}{v(Z)}$$

Pearson's-like partial correlation coefficient is

$$\rho(X,Y\bullet \mathbf{Z}) = \frac{c(X,Y\bullet Z)}{\sqrt{c(X,X\bullet Z)c(Y,Y\bullet Z)}} = \frac{\rho(X,Y) - \rho(X,Z)\rho(Z,Y)}{\sqrt{1-\rho(X,Z)^2}\sqrt{1-\rho(Z,Y)^2}}$$

 $\rho(\text{weight, intelligence} \bullet \text{age}) \simeq 0$ 

One often uses the correlation = covariance scaled with the multiplicities:

$$C(X,Y) = \frac{c(X,Y)}{\langle X \rangle \langle Y \rangle}, \quad V(X) \equiv c(X,X) = \frac{v(X)}{\langle X \rangle^2}$$

Then

$$C(X, Y \bullet Z) = C(X, Y) - \frac{C(X, Z)C(Z, Y)}{V(Z)}$$

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#### Relation to conditional covariance

 $c(X_i, X_j | \mathbf{Z})$  - evaluate at fixed  $\mathbf{Z}$  and then average over  $\mathbf{Z}$ 

[Lawrance 1976]: if a sample satisfies  $E\left(\mathbf{X}|\mathbf{Z}\right)=\alpha+\mathbf{BZ}$ , with  $\alpha$  a constant and  $\mathbf{B}$  a constant matrix  $\Rightarrow$ 

$$c(X_i, X_j \bullet \mathbf{Z}) = c(X_i, X_j | \mathbf{Z})$$

 $\Leftarrow$  shown by [Baba et al. 2005]

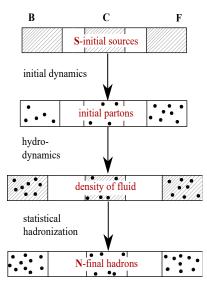
Application of conditinal covariance by [STAR 2008], where Z is hadron multiplicity in the reference bin R:

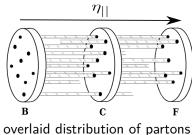
- lacktriangle Divide R into very narrow subsamples (centrality classes) according to Z
- ② Evaluate the covariance between  $X_i$  and  $X_j$  in each subsample
- 4 Average obtained covariances over the subsamples

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# Superposition model

### Superposition model





deterministic

overlaid distribution of hadrons

overlaid detector efficiency

# Superposition model (cont.)

$$N_A = \sum_{i=1}^{S_A} m_i, \quad A = F, B, C$$

$$\langle N_A \rangle = \langle S_A \rangle \langle m \rangle$$

$$v(N_A) = \langle m \rangle^2 v(S_A) + v(m) \langle S_A \rangle$$

$$c(N_A, N_{A'}) = \langle m \rangle^2 c(S_A, S_{A'}), \quad A \neq A'$$

$$c(N_A, S_{A'}) = \langle m \rangle c(S_A, S_{A'})$$

$$\begin{array}{ccc} C(S_A,S_{A'}) & = & C(N_A,N_{A'}) - \delta^{AA'} \frac{\omega(m)}{\langle N_A \rangle} \equiv \overline{C}(N_A,N_{A'}) \\ \\ & \omega(m) = \frac{\mathrm{v}(m)}{\langle m \rangle} & \text{(for Poisson } \omega(m) = 1) \end{array}$$

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# Partial correlations in the superposition model

Multiplicities in  ${\bf F},{\bf B}$  are physical, multiplicity in  ${\bf C}$  is a control variable  $N_C$  constraint:

$$C(S_F, S_B \bullet N_C) = \overline{C}(N_F, N_B) - \frac{\overline{C}(N_F, N_C)\overline{C}(N_B, N_C)}{v(N_C)}$$

 $S_C$  constraint:

$$C(S_F, S_B \bullet S_C) = \overline{C}(N_F, N_B) - \frac{\overline{C}(N_F, N_C)\overline{C}(N_B, N_C)}{\overline{v}(N_C)}$$

Only measured quantities (hadron multiplicities) on r.h.s.!

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$$C(S_F, S_B \bullet N_C)$$
 vs  $C(S_F, S_B \bullet S_C) \leftrightarrow \mathrm{v}(N_C)$  vs  $\overline{\mathrm{v}}(N_C)$ 

Method allows us to impose constraints at the level of initial sources, based on experimentally available info

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# Test of the method

#### Test on actual simulations

- ullet Wounded quark model with GLISSANDO at centrality 30-40%
- Bzdak-Teaney model with triangular emission functions
- 3+1D viscous hydrodynamics
- Statistical hadronization via THERMINATOR
- Results for
  - **1** all charged particles  $\pi^{\pm}$ ,  $K^{\pm}$ , p and  $\overline{p}$ ,
  - 2 primordial particles before resonance decays
  - $\bullet$   $\pi^+$
- $\bullet$  Wide acceptance,  $|\eta_{\parallel}| \leq 5.1,$  divided into 51 bins with  $\Delta \eta = 0.2$

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# Bzdak-Teaney (BT) model

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Use the triangle emission profiles, then:

$$C(S_F, S_B) = \frac{\mathbf{v}(Q_+)}{\langle Q_+ \rangle^2} + \frac{\mathbf{v}(Q_-)}{\langle Q_+ \rangle^2} u_F u_B,$$

where  $u_{F,B}=\eta_{F,B}/y_b$ ,  $Q_{\pm}=Q_A\pm Q_B$  – numbers of wounded quarks In the central (reference) bin  $S_C$  we have  $\eta=0$ , which yields

$$C(S_{F,B}, S_C) = C(S_C, S_C) = \frac{v(Q_+)}{\langle Q_+ \rangle^2}$$

$$C(S_F, S_B \bullet S_C) = \frac{\mathbf{v}(Q_-)}{\langle Q_+ \rangle^2} u_F u_B$$

Partial correlations

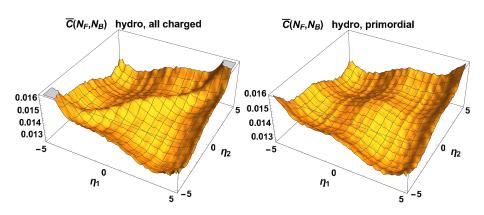
(the same result follows via the condition fixing  $Q_+ \to \mathrm{v}(Q_+) = 0$ )

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#### Scaled covariance

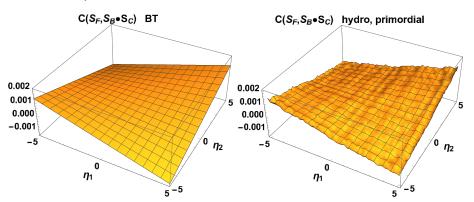


Covariance matrices with the auto-correlations removed Hallmark ridge along the diagonal from resonance decays

(looks as nothing ...)

# Partial: BT vs primordial

$$C: -0.1 < \eta < 0.1$$



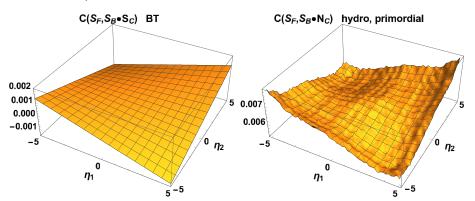
Remarkable agreement of BT and primordial partial correlations

$$C(S_F, S_B \bullet S_C) = \overline{C}(N_F, N_B) - \frac{\overline{C}(N_F, N_C)\overline{C}(N_B, N_C)}{\overline{v}(N_C)}$$

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# Partial: BT vs primordial

$$C: -0.1 < \eta < 0.1$$

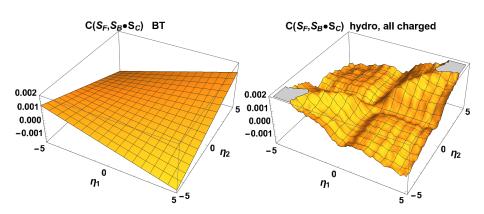


No agreement for the  ${\cal N}_{\cal C}$  constraint

$$C(S_F, S_B \bullet N_C) = \overline{C}(N_F, N_B) - \frac{\overline{C}(N_F, N_C)\overline{C}(N_B, N_C)}{v(N_C)}$$

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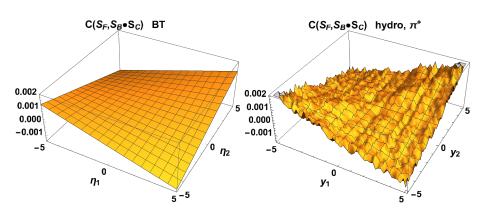
### Partial: BT vs all charged



Short-range correlations spoil the agreement

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#### Partial: BT vs $\pi^+$



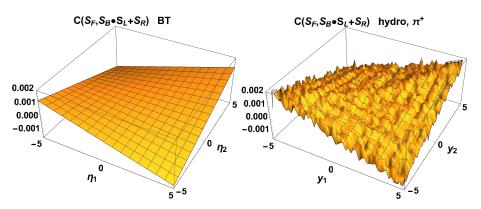
Reduce correlations from resonance decays - no direct decays to  $\pi^+\pi^+$ 

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# Left+right constraint

$$L: -6.1 < \eta < -5.1, \quad R: 5.1 < \eta < 6.1$$

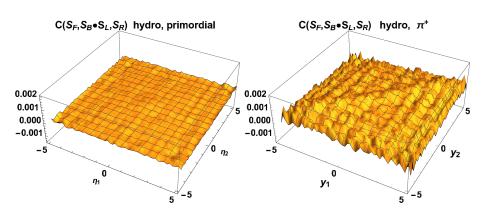


(for BT the same effect as from the central constraint)

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### Independent left- and right constraints



This correlation vanishes in BT (fixes both  $Q_A$  and  $Q_B$ , so nothing is left to fluctuate)

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# Conclusions

#### Conclusions

- Partial correlations+superposition model possibility of imposing constraints at the level of sources, gaining insight into the initial stage
- Contraining (event strictly) the number of particles leaves the fluctuation of sources!
- Feasibility of the method demonstrated on simulated data (wounded quarks, hydrodynamics, THERMINATOR) - would be great to use on actual data!
- Need to reduce the short-range correlations (e.g., by looking at  $\pi^+$ ), nice to have a large pseudorapidity acceptance
- Several simultaneous constraints possible, generalization of the concept of centrality

#### Conclusions

- Partial correlations+superposition model possibility of imposing constraints at the level of sources, gaining insight into the initial stage
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- Need to reduce the short-range correlations (e.g., by looking at  $\pi^+$ ), nice to have a large pseudorapidity acceptance
- Several simultaneous constraints possible, generalization of the concept of centrality

# Thank you!



# Definition of partial covariance

n physical variables  $\mathbf{X}=(X_1,\ldots,X_n),\ m$  control variables  $\mathbf{Z}=(Z_1,\ldots,Z_m)$   $X_i,\ Z_j$  are vectors in the space of events, i.e.,  $X_1=(X_1^{(1)},X_1^{(2)}\ldots X_1^{(N_{\mathrm{ev}})})$   $\langle\mathcal{O}\rangle\equiv\frac{1}{N_{\mathrm{ev}}}\sum_{k=1}^{N_{\mathrm{ev}}}\mathcal{O}^{(k)}$ 

#### Partial covariance:

$$c(\mathbf{X},\!\mathbf{X}\bullet\mathbf{Z})\equiv c(\mathbf{X},\!\mathbf{X})-c(\mathbf{X},\!\mathbf{Z})c^{-1}(\mathbf{Z},\!\mathbf{Z})c(\mathbf{Z},\!\mathbf{X})$$

where  $c(\mathbf{A},\mathbf{B})$  is the usual covariance  $c(A_i,B_j)=\langle A_iB_j\rangle - \langle A_i\rangle\langle B_j\rangle$ . Diagonalizing  $c(\mathbf{Z},\mathbf{Z})$  (orthonormal eigenvectors  $U_k$ ) yields

$$c(X_{i}, X_{j} \bullet \mathbf{Z}) = c(X_{i}, X_{j}) - \sum_{k=1}^{m} c(X_{i}, U_{k}) c(U_{k}, X_{j})$$
$$= c(X_{i} - c(X_{i}, U_{k}) U_{k}, X_{j} - c(X_{j}, U_{k'}) U_{k'})$$

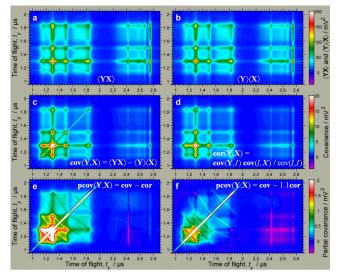
Components of **X** belonging to the space spanned by **Z** are projected out

[H. Cramer, Mathematical methods of statistics, Princeton U. Press, 1946]

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# Example: Coulomb explosion of $N_2$ molecule at FEL

- correlated product
- uncorrelated product
- covariance map
- spurious correlations
- partial covariance
- + corrections



L. J. Frasinski, 2016]