# CORRELATIONS, CORRELATION INTEGRALS AND APPLICATION

TO BOSE-EINSTEIN INTERFEROMETRY<sup>1</sup>

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We review the basic notion of correlations in point processes, adapted to the language of high energy physicists. The measurement of accessible information on correlations by means of correlation integrals is summarized. Applications to the measurement of Bose-Einstein interferometry as well as some pitfalls are discussed.

#### 1. Introduction

In moving to ever higher energies in particle physics, the experimentalist faces a rapidily increasing complexity of the observed events. Beyond posing questions of physics alone, the task to compare experimental data with competing theoretical and phenomenological models raises many questions purely statistical in nature. The latter define a field of their own: multiparticle statistics.

The task of multiparticle statistics is to provide a framework for maximal utilization of the information provided by experiments. This includes questions of measurement of single and multiparticle spectra, correlations among two and more particles, statistical errors due to finite event-sample size, limited detector acceptance, misidentification of track and much more.

In these lectures, we summarize basic concepts of point processes, correlations and their measurement. To demonstrate the discriminative power of advanced correlation measurements, we discuss a recent test of dynamical assumptions in modeling Bose-Einstein correlations.

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### Cross sections, point processes and correlations

 $\sigma_{\rm incl}$  respectively. To be specific, we recall some common definitions [1]: pressed in terms of exclusive or inclusive differential cross sections denoted  $\sigma_{\text{excl}}$  and Correlations (in the widest sense) among final state particles are commonly ex-

$$j_N(p_1,...,p_N) \equiv \frac{1}{\sigma_{\text{tot}}} \frac{d^{3N}\sigma_{\text{excl}}}{d^3p_1d^3p_2...d^3p_N},$$
 (1)

$$\rho_q(p_1,\ldots,p_q) \equiv \frac{1}{\sigma_{\text{tot}}} \frac{d^3q\sigma_{\text{incl}}}{d^3p_1d^3p_2\ldots d^3p_q},\tag{2}$$

 $p_i$  is the three-momentum of the *i*-th particle, and  $\sigma_{\text{tot}} = \sum_N \sigma_N$  is the total (inelastic) cross section, with  $\sigma_N$  the integrated cross section for events with N final state particles the multiplicity distribution in full phase-space (here to be taken as identical for simplicity of notation). The ratio  $P_N=\sigma_N/\sigma_{
m tot}$  gives

The densities  $j_N$  and  $\rho_q$  are very different objects:  $j_N(p_1, \ldots, p_N)d^3p_1 \ldots d^3p_N$  is proportional to the probability that in an event with exactly N particles we find siparticles with momenta simultaneously within infinitesimal boxes centered on  $p_1, \dots, p_q$  per event (no matter what N is). Therefore  $\rho_p$  characterizes samples which include all By contrast,  $\rho_q(p_1,\ldots,p_q)d^3p_1\ldots d^3p_q$  is the average number of unordered q-tuples of of samples built exclusively from events with exactly N particles (exclusive samples). multaneously the first particle in a box of size  $d^3p_1$  centered in  $p_1$ , the second in a referring to  $j_N$  as a Janossy density and to  $\rho_q$  as a factorial moment density. events (inclusive samples). While various other names are in use, we follow Ref. [2] in box at  $p_2, \ldots$ , and the N-th at  $p_N$ . Thus,  $j_N$  characterizes the statistical properties

to do with quantum mechanical "indistinguishability" of particles, but reflects a mere physical information. Therefore it is customary to count all N!(q!) permutations of convention designed to simplify the resulting formalism. In fact, the labels "first" particle, "second" particle, ... are arbitrary and carry no labels as separate, independent events (q-tuples). This symmetrization has nothing Both functions are symmetrized with respect to permutations of their arguments.

density of N random variables  $p_i$ , it is a mistake to regard  $\rho_q$  as such. It is constructed from inclusive samples, where the number of random variables (N) is a again a random While  $j_N/N!P_N$ , normalized to unity, represents a conventional joint probability

In fact, integration over an arbitray region of phase-space  $\Omega$  gives the q-th factorial moment of the random multiplicity n in this domain:

$$\xi_q(\Omega) = \int_{\Omega} \rho_p(p_1, \dots, p_q) d^3 p_1 \dots d^3 p_q = \langle n^{[q]} \rangle_{\Omega}. \tag{3}$$

We use the common notation  $n^{[q]} \equiv n(n-1)\cdots(n-q+1)$  for factorials.

mental multiplicity and correlation measurements can be viewed as particular counting All the above definitions are given in terms of cross sections. However, all experi-

> moreover, how do we count most efficiently? tions: How do we have to count in order to obtain a certain type of information and, procedures of particles or particle-tuples in certain domains. Thus, we pose the ques-

generality, we subsequently denote the "positions" of the particles (=points) by  $\mathbf{X}_{i}^{a}$ , variables such as (p, s), with s labeling the charge, spin or species of the particle<sup>3</sup> are rapidity y, rapidity-azimuth  $(y, \Phi)$  or even a combination of continuous and discrete  $(i=1,\ldots,N)$ , where X can refer to any set of coordinates. Frequently used examples Consider a sample of  $N_{ev}$  events, each labeled by an index  $a=1,\ldots,N_{ev}$ . For greater The appropriate tool to tackle such questions is the theory of point processes [3, 2]

represented by the "random Dirac comb" The density of such points at x in one particular event a is most conveniently

$$\hat{
ho}_1^a(\mathbf{x}) = \sum_{i_1=1}^N \delta(\mathbf{x} - \mathbf{X}_{i_1}^a),$$

4

in that domain. More generally, the simultaneous behavior (=correlation) of q of these particles in that event is represented by the restricted tensor product of Dirac combs which, when integrated over a certain domain  $\Omega$ , just gives the number of particles n

$$\hat{\rho}_q^a(\mathbf{x}_1,\dots,\mathbf{x}_q) = \sum_{i_1 \neq i_2 \neq \dots \neq i_q = 1}^N \delta(\mathbf{x}_1 - \mathbf{X}_{i_1}^a)\delta(\mathbf{x}_2 - \mathbf{X}_{i_2}^a)\dots\delta(\mathbf{x}_q - \mathbf{X}_{i_q}^a),$$
 (5)

sum). the same particle more than once (hence the restriction of the indices in the multiple which just acts as a q-tuple counter in event a. Note that it doesn't make sense to count

yield the q-tuple density Meaningful results are extracted by averaging over the inclusive event sample, to

$$\rho_q = \langle \hat{\rho}_q \rangle = N_{ev}^{-1} \sum_{a=1}^{N_{ev}} \hat{\rho}_q^a, \tag{6}$$

which is nothing but the counting prescription for the factorial moment densities eq.

A point process is fully determined by the knowledge of either all  $\rho_q$ , all  $j_N$  or, most conveniently, by its generating functional. The latter is defined by

$$Z[\lambda(\mathbf{x})] = \left\langle \exp\left(\int \hat{\rho}(\mathbf{x}) \log[1 + \lambda(\mathbf{x})] d\mathbf{x} \right) \right\rangle \tag{7}$$

$$= \sum_{q\geq 0} \frac{1}{q!} \int \rho_q(\mathbf{x}_1, \dots, \mathbf{x}_q) \lambda(\mathbf{x}_1) \cdots \lambda(\mathbf{x}_q) d\mathbf{x}_1 \cdots d\mathbf{x}_q.$$
 (8)

Once we know  $Z[\lambda(\mathbf{x})]$ , the  $\rho_q$  as well as the  $j_N$  can be obtained via functional derivatives with respect to the test function  $\lambda$ :

$$\rho_q(\mathbf{x}_1, \dots, \mathbf{x}_q) = \frac{\delta^q Z[\lambda(\mathbf{x})]}{\delta \lambda(\mathbf{x}_1) \cdots \delta \lambda(\mathbf{x}_q)} \bigg|_{\lambda=0}$$
(9)

 $<sup>^3</sup>$ In this case all following integrations contain implicitly sums over s.

$$j_{N}(\mathbf{x}_{1},...,\mathbf{x}_{N}) = \frac{\delta^{q} Z[\lambda(\mathbf{x}) - 1]}{\delta \lambda(\mathbf{x}_{1}) \cdots \delta \lambda(\mathbf{x}_{N})} \bigg|_{\lambda=0}$$
(10)

of the factorial moment densities  $\rho_q(\mathbf{x}_1,\ldots,\mathbf{x}_q)=\rho_1(\mathbf{x}_1)\rho_1(\mathbf{x}_2)\cdots\rho_1(\mathbf{x}_q)$  for all q. In of statistical independence in point processes. The latter is defined by full factorization this case we can readily sum up the series (8) to obtain In order to understand what we mean by "correlations" we first discuss the meaning

$$Z[\lambda(\mathbf{x})] = \exp\left(\int \rho_1(\mathbf{x})\lambda(\mathbf{x})d\mathbf{x}\right),$$
 (11)

statistics of continuous random variables. process plays a similar central role for point processes as the Gaussian does for the multiplicity of points in any arbitrary domain  $\Omega$  follows a Poisson distribution. This which is the generating functional of the so-called Poisson process. In this process, the

by functional derivatives of log  $Z[\lambda]$ : Genuine correlations among points are then quantifiable by deviations from the Poisson process. Consider the family of functions, called cumulant densities, obtained

$$C_q(\mathbf{x}_1, \dots, \mathbf{x}_q) = \frac{\delta^q \log Z[\lambda(\mathbf{x})]}{\delta \lambda(\mathbf{x}_1) \dots \delta \lambda(\mathbf{x}_q)} \bigg|_{\lambda=0}$$
(12)

deviations from the Poisson process. For the Poisson process (11) we see that  $C_1 = \rho_1$  and all higher cumulant densities vanish. In this sense, nonvanishing  $C_q$  with  $(q \ge 2)$  quantify genuine correlations, i.e.

heavy ion reactions cumulants of third and higher orders are highly suppressed. Such processes can be modeled as poisson processes of "clusters", each cluster decaying into It is quite possible that only a few orders of cumulants are nonzero. For example in

The generating functional  $Z[\lambda]$  is a convenient bookkeeping device for establishing all kinds of relations among  $j_N$ ,  $\rho_q$  and  $C_q$ , for example one or 2 daughter particles.

$$C_2(\mathbf{x}_1, \mathbf{x}_2) = \rho_2(\mathbf{x}_1, \mathbf{x}_2) - \rho_1(\mathbf{x}_1)\rho_1(\mathbf{x}_2),$$
 (13)

$$\hat{r}_3(\mathbf{x}_1, \mathbf{x}_2, \mathbf{x}_3) = \rho_3(\mathbf{x}_1, \mathbf{x}_2, \mathbf{x}_3) - \rho_1(\mathbf{x}_1)\rho_2(\mathbf{x}_2, \mathbf{x}_3) - \rho_1(\mathbf{x}_2)\rho_2(\mathbf{x}_3, \mathbf{x}_1) - \rho_1(\mathbf{x}_3)\rho_2(\mathbf{x}_1, \mathbf{x}_2) + 2\rho_1(\mathbf{x}_1)\rho_1(\mathbf{x}_2)\rho_1(\mathbf{x}_3) \text{ etc.}$$

ground". It is this property that makes cumulants a favourable tool for discriminative in the construction of cumulants, sometimes called "removal of combinatorial backexperimental analysis, but at the same time more difficult to measure. Notice the subtration of various products of lower order factorial moment densities

#### 3. Correlation integrals

can never, of course, be achieved in practice. However, while higher-order correlation all orders, the multiparticle production process would be completely determined. This If we had experimental knowledge of correlations in terms of either  $j_N$ ,  $\rho_q$  or  $C_q$  to

> functions can never be sampled fully differentially, one can still try to extract as much domains  $\Omega$  as in eq. (3). information as possible from a give data sample by measuring integrals over various

counts  $n_m^{[q]}$  in every bin [4], variable X ("binning the data") and then to find  $\xi_q$  by averaging over all events the Conventional measurements of correlations proceed first to discretize the continuous

$$\xi_q^{conv} = \left\langle \sum_{\text{hins } m} n_m^{[q]} \right\rangle. \tag{15}$$

generally, every correlation integral of order q assigns "size" to every possible q-tuple of particles. The way this assignment is done distinguishes the different versions of  $X_{i_1i_2} \equiv |X_{i_1} - X_{i_2}|$  rather than counting particles in predefined bins [4, 5]. More large data sample, this corresponds to an integration over  $\rho_q$  in specific domains  $\Omega(\epsilon)$ . correlation integrals. Finally, they count the number of q-tuples with a given size  $\epsilon$ ("differential forms) or the ones smaller than a given size ("integral forms"). For a By contrast, so-called correlation integrals rely on distances between pairs of points

particle at the center  $X_{i_1}$ , number of particles ("sphere count") within one of these spheres is, not counting the radius  $\epsilon$ , each centered at one of the N particles in the event. For a given event a the For the Star integral [6], the domain  $\Omega$  is given by the collection of N spheres of

$$a \equiv \hat{n}(\mathbf{X}_{i_1}, \epsilon) \equiv \sum_{i_2=1}^{N} \Theta(\epsilon - \mathbf{X}_{i_1 i_2}), \quad i_2 \neq i_1,$$
(16)

where a is an "ultra short" notation needed for some lengthy formulae below. With this elementary counter the factorial moment of order q is simply obtained by

$$\xi_q^{\text{Star}}(\epsilon) = \langle \sum_{i_1} \hat{n}(\mathbf{X}_{i_1}, \epsilon)^{[q-1]} \rangle = \langle \sum_{i_1} a^{[q-1]} \rangle. \tag{17}$$

One can show [6] that the above counting prescription corresponds to integrating eq. (3) using for  $\Omega$  a particular "Star" domain implemented via theta functions  $\Theta_{1j} \equiv$  $\Theta(\epsilon - |\mathbf{x}_1 - \mathbf{x}_j|)$ , restricting all q - 1 coordinates  $\mathbf{x}_j$  to within a distance  $\epsilon$  of  $\mathbf{x}_1$ :

$$\xi_q^{\text{Star}}(\epsilon) = \int \rho_q(\mathbf{x}_1, \dots, \mathbf{x}_q) \Theta_{12} \Theta_{13} \dots \Theta_{1q} d\mathbf{x}_1 \dots d\mathbf{x}_q. \tag{18}$$

integral have order  $N_{ev}*N^q$  complexity, which quickly becomes unmanageable for higher advantage of the particular "Star" domain lies in the fact that eq. (17) requires typ-Peschanski factorial moments [4] is discussed and demonstrated in [5]. The further ically  $N_{ev}*N^2$  computation steps for any order q, whereas other types of correlation The superiority of correlation integrals in general over the conventional Bialas-

become customary in high energy physics to measure normalized factorial moments [4] In order to eliminate, among other things, the overall total cross section, it has

background,  $\rho_1^q$ . While it can be implemented in a number of ways, we prefer the "vertical" normalization, in which  $\rho_1^q$  is integrated over exactly the same domain  $\Omega$  as the inclusive density  $\rho_q$  in the numerator. Thus for the Star integral, the normalized The denominator used for such normalization should be made up of the uncorrelated

$$F_q^{\text{Star}}(\epsilon) \equiv \frac{\xi_q^{\text{Star}}}{\xi_q^{\text{norm}}} = \frac{\int \rho_q(\mathbf{x}_1, \dots, \mathbf{x}_q) \Theta_{12} \Theta_{13} \dots \Theta_{1q} d\mathbf{x}_1 \dots d\mathbf{x}_q}{\int \rho_1(\mathbf{x}_1) \dots \rho_1(\mathbf{x}_q) \Theta_{12} \Theta_{13} \dots \Theta_{1q} d\mathbf{x}_1 \dots d\mathbf{x}_q}.$$
 (19)

average: with  $\mathbf{X}_{i_1i_2}^{ab} \equiv |\mathbf{X}_{i_1}^a - \mathbf{X}_{i_2}^b|$  measuring the distance between two particles taken from different events a and b, and the "ultra short" notation We have shown [6] that the denominator  $\xi_q^{\rm norm}$  is given by the following double event

$$b \equiv \hat{n}_b(\mathbf{X}_{i_1}^a, \epsilon) = \sum_{i_2} \Theta(\epsilon - \mathbf{X}_{i_1 i_2}^{ab})$$
 (20)

$$\xi_q^{\text{norm}}(\epsilon) \equiv \left\langle \sum_{i_1} \hat{\xi}_q^{\text{norm}}(i) \right\rangle = \left\langle \sum_{i_1} \langle b \rangle^{q-1} \right\rangle.$$
 (21)

Note that the outer event average and sum over  $i_1$  are taken over the center particle taken from event a, each of which is used as the center of sphere counts  $\hat{n}_b(\mathbf{X}_{i_1}^a,\epsilon)$  taken only  $A * N_{ev} * N^2$  computation steps. This procedure we call "reduced" event mixing. small subsample  $b = a - 1, a - 2, \dots, a - A$ , consisting only of A - 1 events and requiring over other events  $b \neq a$  in the inner event average. We thus see the natural emergence of in practical applications it is sufficient to reduce the inner average over b-events to a prescription has complexity  $N_{\rm ev}^2 * N^2$  – unmangeable for large event samples. However, the heuristic procedure of normalization known as "event mixing" [6, 5]. This counting

i are performed. Integrating the  $C_q$  over the Star integral domain, we define normalized (factorial) cumulants  $K_q^{\rm Star}(\epsilon) \equiv f_q(\epsilon)/\xi_q^{\rm norm}(\epsilon)$ , with we get the integrals of cumulants (almost) for free, once the counts a and b per particle Another essential advantage of the Star domain is the fact that on top of the  $F_q^{Star}$ 

$$f_q(\epsilon) \equiv \int C_q(\mathbf{x}_1, \dots, \mathbf{x}_q) \Theta_{12} \Theta_{13} \dots \Theta_{1q} d\mathbf{x}_1 \dots d\mathbf{x}_q.$$
 (22)

The latter can be written entirely in terms of the sphere counts a and b introduced previously! Defining for convenience the "i-particle cumulant"  $\hat{f}_q(i)$  so that

 $\left\langle \sum_{i} \hat{f}_{q}(i) \right\rangle = f_{q}$ , we find [6]

$$\hat{f}_2(i) = a - \langle b \rangle, \tag{23}$$

$$\hat{f}_3(i) = a^{[2]} - \langle b^{[2]} \rangle - 2a\langle b \rangle + 2\langle b \rangle^2, \quad \text{etc.}$$
 (24)

ison of theoretical models with data dictate a specific choice of variables and integration tude, other types of integration domains  $\Omega$  are also useful, in particular when a compar-While computationally more expensive than the Star integral by orders of magni-

> a preferred variable is the 4-momentum difference  $q_{ij} = [(\mathbf{p}_i - \mathbf{p}_j)^2 - (E_i - E_j)^2]^{1/2}$ . domain. This is typically the case in measurements of Bose-Einstein correlations, where

metrical manner (called GHP topology) [7] such as We then may study differential forms of integrations over  $\rho_q$  or  $C_q$  in a fully sym-

$$C_q(\epsilon) = \int C_q(\mathbf{p}_1, \dots, \mathbf{p}_q) \delta(\epsilon - \sum_{i < j=1}^q q_{ij}^2) d^3 \mathbf{p}_1 \cdots d^3 \mathbf{p}_q.$$
 (25)

Other q-tuple size prescriptions (used below) are obtained by changing the argument of the  $\delta$ -function appropriately, e.g.  $\delta(\epsilon - \sum_{i < j=1}^q q_{ij})$  or  $\delta(\epsilon - \max(q_{12}, \dots, q_{q-1,q}))$ . The corresponding counting prescriptions are conveniently written in terms of the

$$I_{i_1i_2...i_q}^{e_1e_2...e_q}(\epsilon) = \begin{cases} 1 & \text{if the "size" of the } q\text{-tuple is within}\epsilon + \delta\epsilon \\ 0 & \text{otherwise,} \end{cases}$$
 (20)

where the q-tuple is composed of particle  $i_1$  taken from event  $e_1$ , particle  $i_2$  taken from a different event  $e_2$ , etc., and the "size" is evaluated, e.g., according to one of above quoted prescriptions.

The counting algorithms for  $\rho_q(\epsilon)$  and the first  $C_q(\epsilon)$  are

$$\rho_q(\epsilon)\delta\epsilon = \left\langle \sum_{i_1 \neq \cdots \neq i_q} I_{i_1 i_2 \cdots i_q}^{aa \cdots a}(\epsilon) \right\rangle$$
 (27)

$$C_2(\epsilon)\delta\epsilon = \left\langle \sum_{i \neq j} I_{ij}^{aa}(\epsilon) \right\rangle_a - \left\langle \left\langle \sum_{i,j} I_{ij}^{ab}(\epsilon) \right\rangle_b \right\rangle_a \tag{28}$$

$$C_{3}(\epsilon)\delta\epsilon = \left\langle \sum_{i\neq j\neq k} I_{ijk}^{aaa}(\epsilon) \right\rangle_{a} - 3 \left\langle \left\langle \sum_{i\neq j,k} I_{ijk}^{aab}(\epsilon) \right\rangle_{b} \right\rangle_{a} + 2 \left\langle \left\langle \left\langle \sum_{i,j,k} I_{ijk}^{abc}(\epsilon) \right\rangle_{c} \right\rangle_{b} \right\rangle_{a}$$
(29)
Integrals (25) over uncorrelated tensor products of  $\rho_{1}$ , needed for normalization, are sampled in similar ways [7]:

$$\rho_1 \otimes \rho_1 \otimes \dots \otimes \rho_1(\epsilon) \delta \epsilon = \left\langle \left\langle \dots \left\langle \sum_{i_1, i_2, \dots, i_q} I_{i_1 i_2 \dots i_q}^{e_1 e_2 \dots e_q}(\epsilon) \right\rangle \right\rangle \dots \right\rangle \dots \left\langle 30 \right\rangle$$

Note that multiple event averages, e.g.  $\langle (\dots \rangle_a)_b = \sum_{a \neq b} / N_{ev}(N_{ev} - 1)$ , always run over unequal events to avoid noticeable sampling biases. Again, inner event averages can run over a small fraction of the full sample to keep computing times in manageable ranges [6]

## Applications to Bose-Einstein correlation measurements

specific form of the generating functional (7) with  $C_2$  as freely parametrizable function. parametrization of  $C_2$ . from experiment with the model predictions, without having to rely on a particular validity of the postulated generating functional by comparing higher order cumulants fied with no further adjustable parameter. One therefore can test experimentally the Once  $C_2$  is determined from experiment, all higher cumulant densities are fully specifor Bose-Einstein correlations among identical pions [9, 10, 11]. The latter postulate a As an application, we discuss a recent test [8] of quantum statistical (QS) models

of interest are [11] When relative phases are neglected, the second and third (reduced) "QS cumulants"

$$c_2 \equiv \frac{C_2}{\rho_1 \otimes \rho_1} = 2p(1-p)d_{12} + p^2 d_{12}^2,$$
 (31)

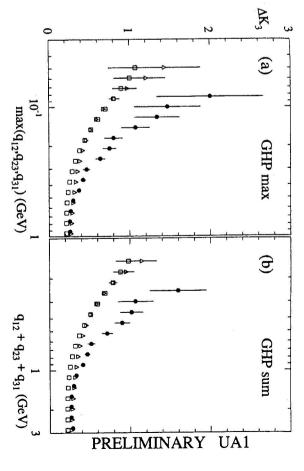
$$k_3 \equiv \frac{C_3}{\rho_1 \otimes \rho_1 \otimes \rho_1} = 2p^2(1-p)[d_{12}d_{23} + d_{23}d_{31} + d_{31}d_{12}] + 2p^3d_{12}d_{23}d_{31}, \tag{32}$$

where  $\rho_1\otimes\rho_1=\rho_1(p_1)\rho_1(p_2)$  and the  $d_{ij}=d(q_{ij})$  are functions of the 4-momentum differences qij

 $q_{ij}^{-\alpha}$ ). Since the Gaussian does not give viable fits, only the latter two parametrizations  $\exp(-r^2q_{ij}^2)$ , exponential  $(d_{ij}=\exp(-rq_{ij}))$  and power-law parametrizations  $(d_{ij}=\exp(-rq_{ij}))$ are used below. usually parametrized in a plausible and/or convenient way, such as Gaussian (dij While in principle calculable from a given density matrix, the functions  $d_{ij}$  are

at  $\sqrt{s} = 630 \text{ GeV}$ ) with QS-predictions (32). Besides the GHP-sum topology used in cumulant data measured differs appreciably from that predicted by the QS formulae curves may be shifted up and down. It is clear, though, that the shape of third-order theoretical points shown are determined only up to an additive constant, so that the were taken from the QS fit (31) to  $\Delta K_2$  obtained from the same data sample. All third order cumulant  $\Delta K_3$  (29) obtained from UA1 minimum-bias data ( $\bar{p}p$ -reactions and parameter values from  $\Delta K_2$ . This conclusion holds independently of the topology Fit parameter values used for the respective power law and exponential parametrizations triplets according to the largest of the three momentum differences,  $\max(q_{12}, q_{23}, q_{31})$ . used and of the functional form taken for d. (b), we show in (a) a separate analysis using the "GHP max" topology [6], which bins In Fig. 1, we show the results of a preliminary<sup>4</sup> comparison of the normalized

correlations and thus contain mostly redundant information. rect measurements of cumulants, which are considerably more sensitive than moments. that the recent improvement of measurement techniques have permitted the present dihigher-order factorial moments. The apparent discrepancy is explained by pointing out sion [10], based on an earlier UA1 paper [12], that QS theory was compatible with The latter are dominated numerically by the combinatorial background of lower order The results of this analysis may appear, at first sight, to contradict the conclu-



show predictions for  $\Delta K_3$  from QS theory and parameter values obtained from fits to  $\Delta K_2$ . sum (b) topologies [8]. Filled circles represent UA1 minimum-bias data, while open symbols Open triangles are predictions based on the QS power-law parametrization, open squares are QS exponential predictions. Fig. 1. Third-order differential cumulants  $\Delta K_3(\epsilon)$  integrated with GHP max (a) and GHP

#### 5. Final remarks

lants are promising and sensitive tools to obtain refined insight into various production likely reveal more detailed structure of the underlying dynamics. In particular, cumu-The use of correlation integrals permits much accurate measurements and hence will

one has to use unbiased estimators, witch correct for this effect (cf. [6]). due to the finite size of event samples and can have quite noticeable size. In practice, level than before. One such bias arising generally in the measurement of correlations is Greater accuracy requires, however, that possible biases be understood on a deeper

according to the experimental uncorrelated one-particle distribution; this and other differential ratios can be quite substantial. A procedure to overcome this problem correlation functions such as (31) and (32). Experimentally, one can never measure predictions, when the latter are given in terms of differentially normalized (reduced) details are explained in [8] amounts to a Monte Carlo integration of a theoretical correlation function sampled bin of finite size  $\Omega$  (however small) before the ratio is taken. The discrepancy to fully fully differential ratios; rather, the numerator and denominator are averaged over some Another caveat is the comparsion of experimental correlation data with theoretical

good detector acceptance. The analysis is currently being extended to an enlarged phase space region <sup>4</sup>This preliminary analysis is based on like-sign particles in a restricted phase space region with

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