

A general procedure for detector response correction of higher order cumulants

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Based on [arXiv:1805.00279](https://arxiv.org/abs/1805.00279), submitted to *Nucl.Instrum.Meth. A*

Outline

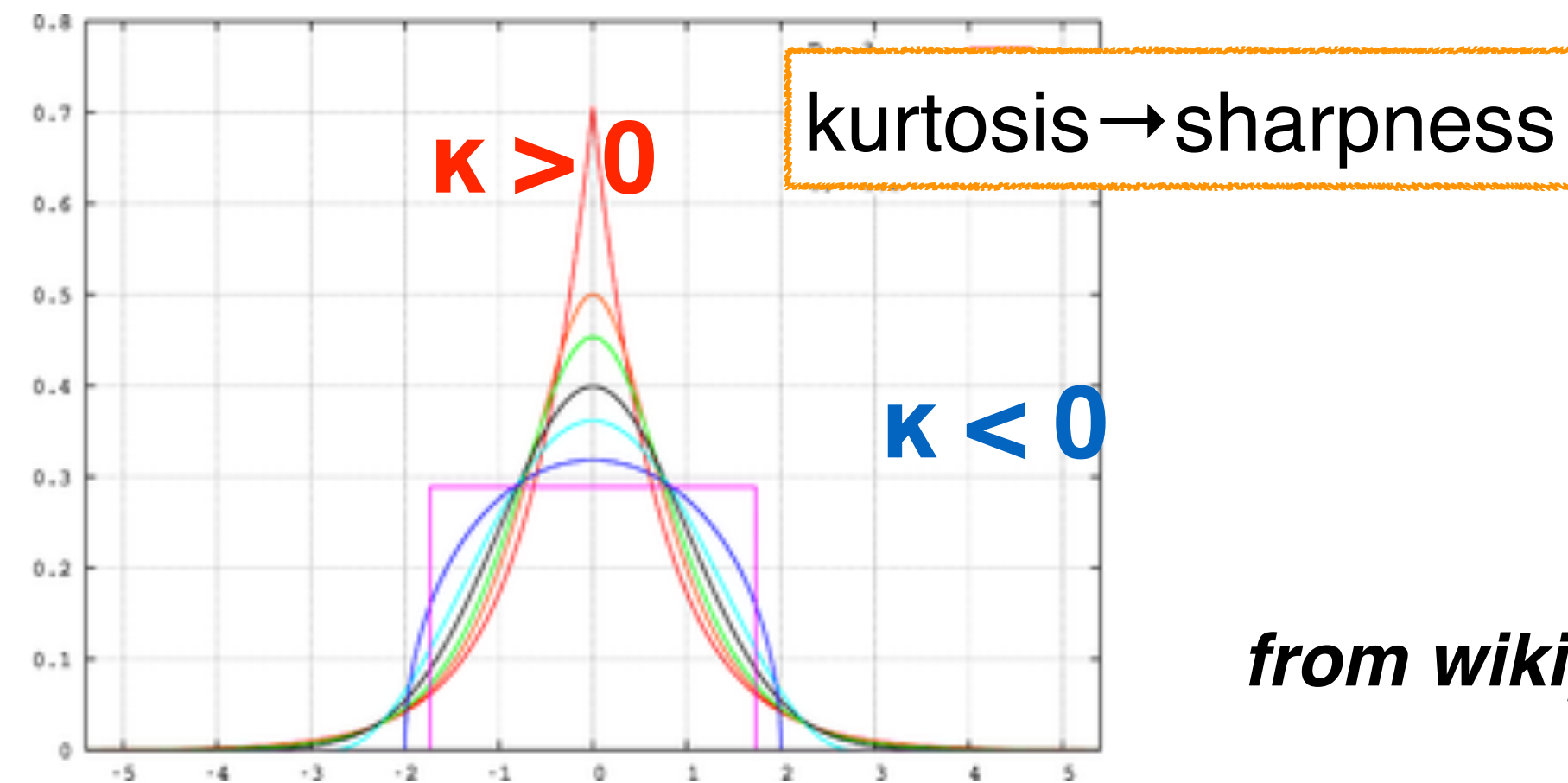
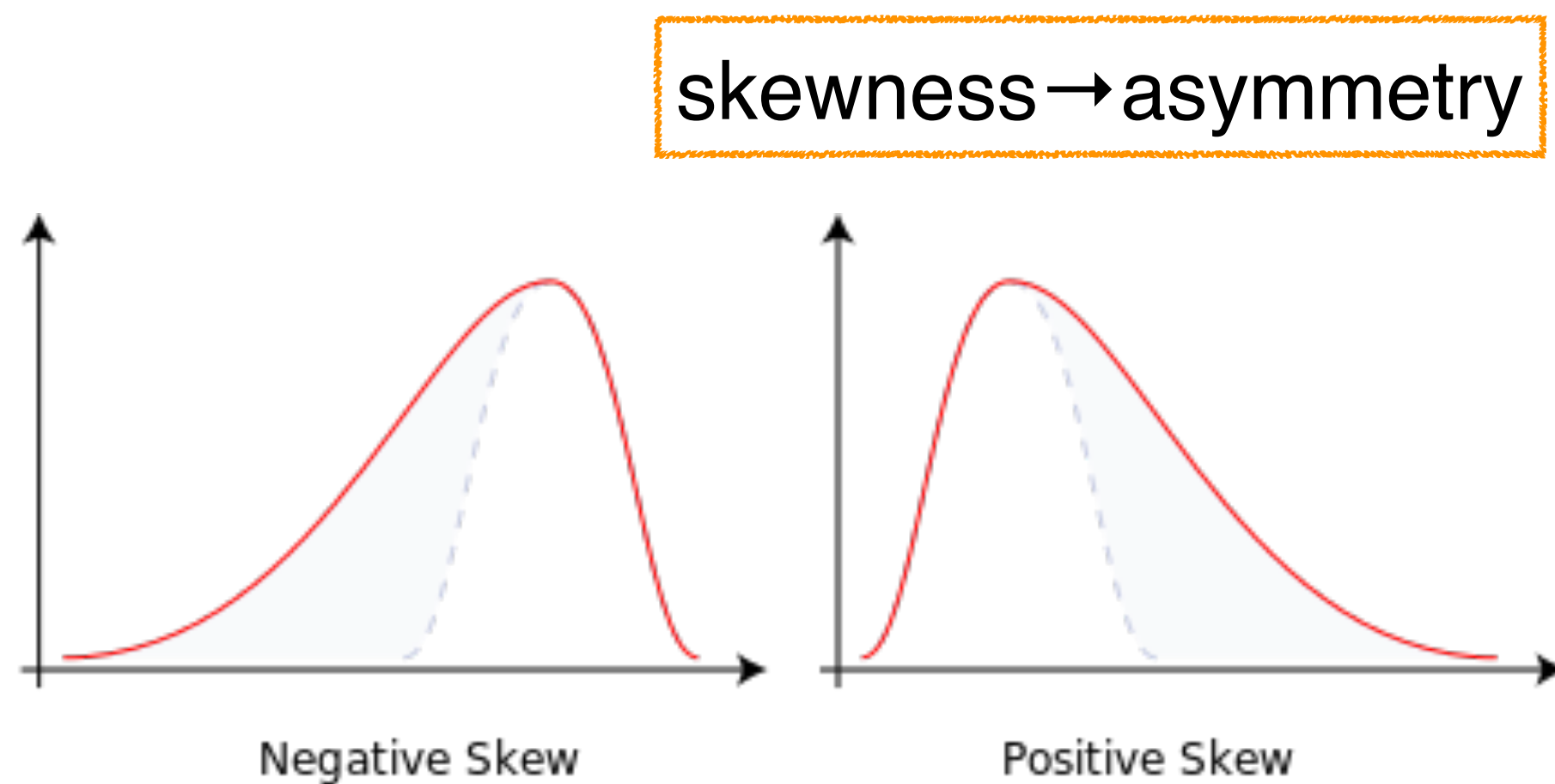
- Introduction
- Efficiency correction with binomial assumption
- More general procedures
- Summary

Introduction

Higher order fluctuation

◆ Moments and cumulants are mathematical measures of “shape” of a distribution which probe the fluctuation of observables.

- ✓ Moments: mean (M), standard deviation (σ), skewness (S) and kurtosis (κ).
- ✓ S and κ are non-gaussian fluctuations.



from wikipedia

✓ Cumulant \Leftrightarrow Moment

$$\langle \delta N \rangle = N - \langle N \rangle$$

$$C_1 = M = \langle N \rangle$$

$$C_2 = \sigma^2 = \langle (\delta N)^2 \rangle$$

$$C_3 = S\sigma^3 = \langle (\delta N)^3 \rangle$$

$$C_4 = \kappa\sigma^4 = \langle (\delta N)^4 \rangle - 3 \langle (\delta N)^2 \rangle^2$$

✓ Cumulant : additivity

$$C_n(X + Y) = C_n(X) + C_n(Y)$$

\rightarrow proportional to volume

Cumulants of conserved quantities

◆ Net-baryon, net-charge and net-strangeness

“Net” : positive - negative

$$\Delta N_q = N_q - N_{\bar{q}}, \quad q = B, Q, S$$

No. of **positively charged** particles in one collision

No. of **negatively charged** particles in one collision

Fill in histograms over many collisions



(1) Sensitive to correlation length

$$C_2 = \langle (\delta N)^2 \rangle_c \approx \xi^2 \quad C_5 = \langle (\delta N)^5 \rangle_c \approx \xi^{9.5}$$

$$C_3 = \langle (\delta N)^3 \rangle_c \approx \xi^{4.5} \quad C_6 = \langle (\delta N)^6 \rangle_c \approx \xi^{12}$$

$$C_4 = \langle (\delta N)^4 \rangle_c \approx \xi^7$$

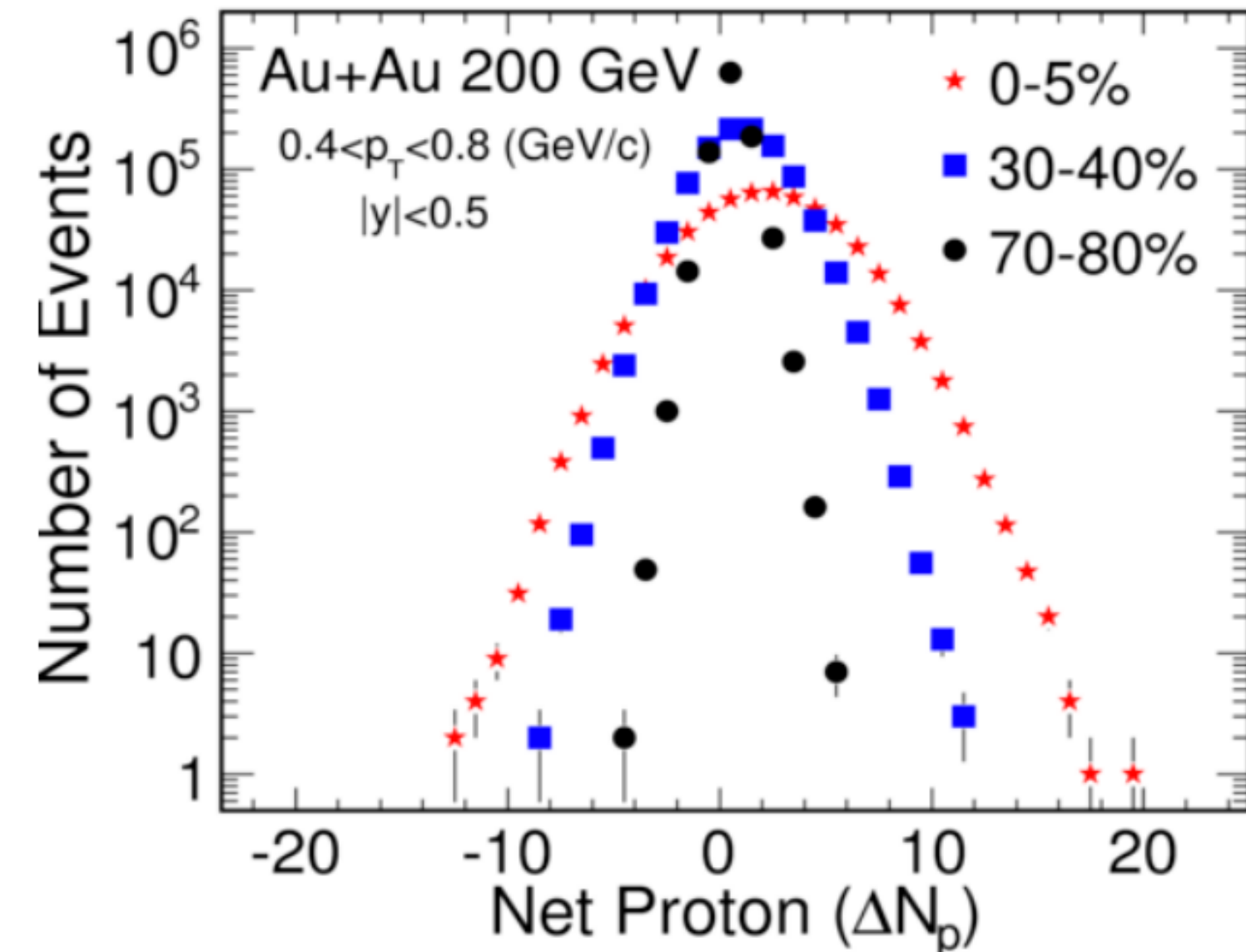
(2) Direct comparison with susceptibilities.

M. Cheng et al, PRD 79, 074505 (2009)

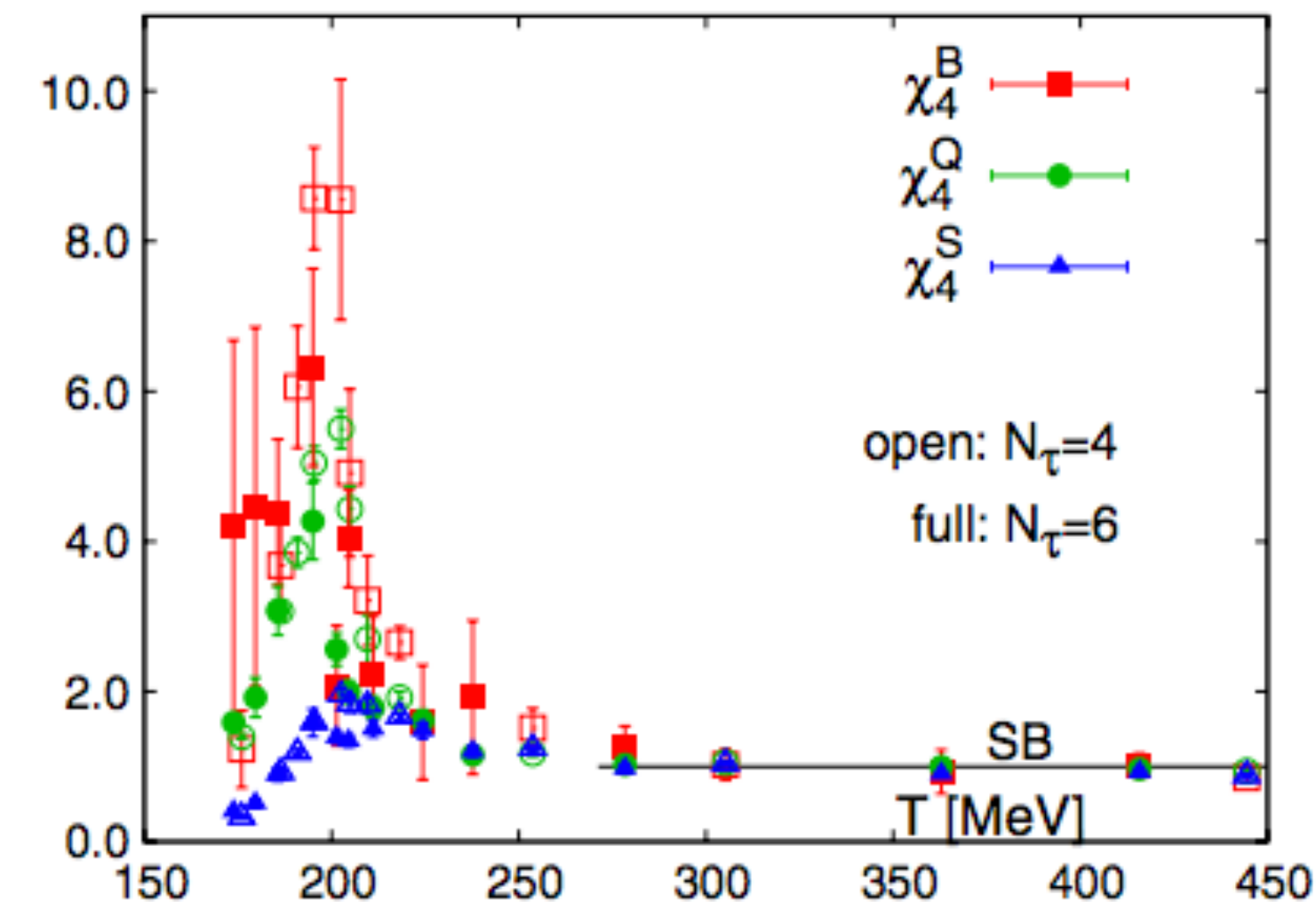
$$S\sigma = \frac{C_3}{C_2} = \frac{\chi_3}{\chi_2} \quad \kappa\sigma^2 = \frac{C_4}{C_2} = \frac{\chi_4}{\chi_2}$$

$$\chi_n^q = \frac{1}{VT^3} \times C_n^q = \frac{\partial^n p/T^4}{\partial \mu_q^n}, \quad q = B, Q, S$$

Volume dependence can be canceled by taking ratio.



→ neutrons cannot be measured



Two difficulties

✓ **Two issues which have not been fully understood yet.**

- **Initial volume fluctuation**

- It is known that event-by-event participant (impact parameter) fluctuation affects the value of cumulants.
- Several model dependent correction methods.

- **Non-binomial detector effects**

- Detectors miss some particles with finite probability called “efficiency”, which distorts the value of cumulants.
- Conventional correction methods rely on the assumption that efficiencies are binomial.
- **Experimentally, however, it may not be binomial.**

Efficiency correction with binomial assumption

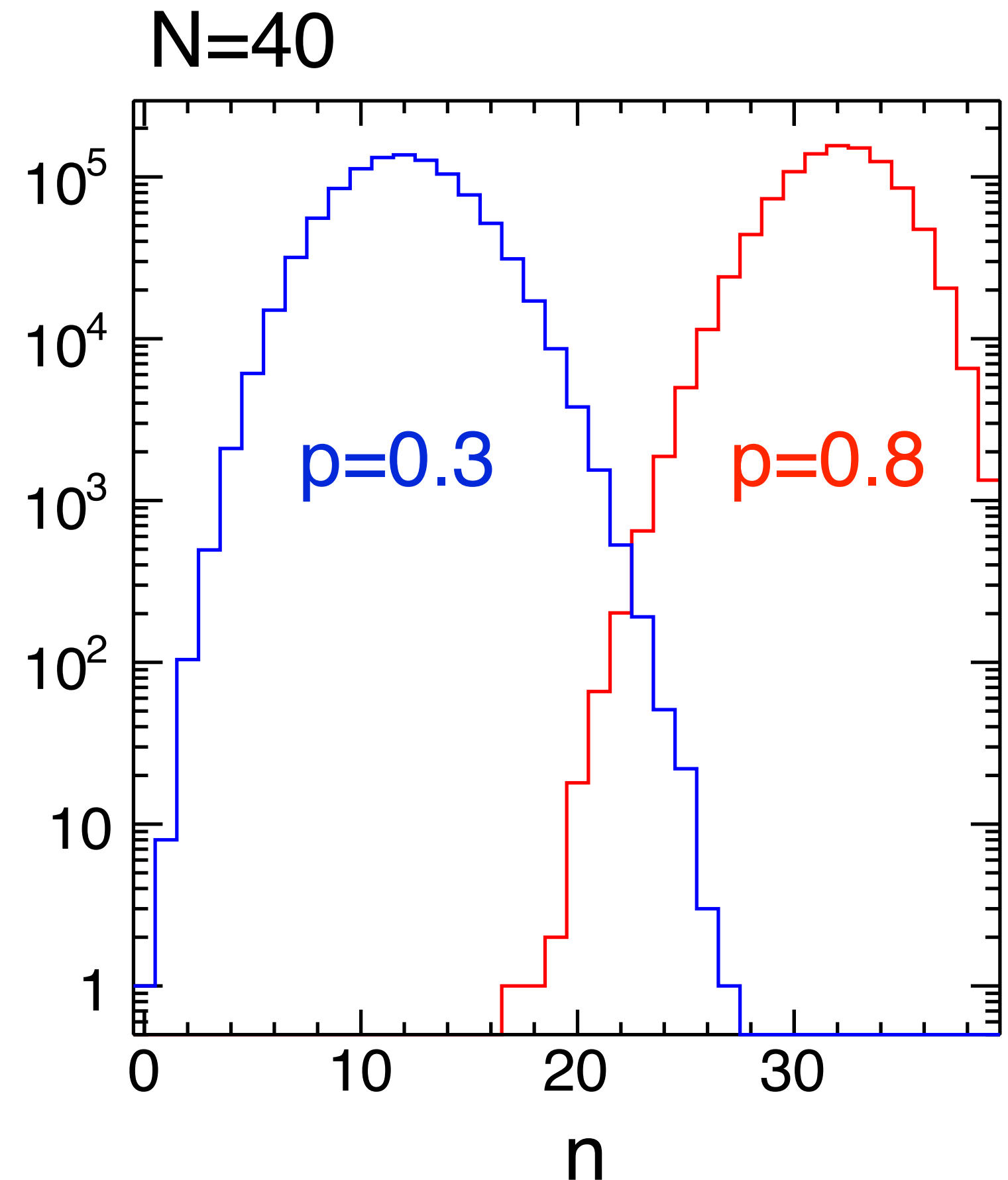
M. Kitazawa and M. Asakawa : PRC.86.(2012)024904

$$\begin{aligned}\langle N \rangle &= \xi_1^{-1} \langle n \rangle, \\ \langle N^2 \rangle_c &= \xi_1^{-2} \langle n^2 \rangle_c - \xi_2 \xi_1^{-3} \langle n \rangle, \\ \langle N^3 \rangle_c &= \xi_1^{-3} \langle n^3 \rangle_c - 3\xi_2 \xi_1^{-4} \langle n^2 \rangle_c + (3\xi_2^2 \xi_1^{-5} - \xi_3 \xi_1^{-4}) \langle n \rangle, \\ \langle N^4 \rangle_c &= \xi_1^{-4} \langle n^4 \rangle_c - 6\xi_2 \xi_1^{-5} \langle n^3 \rangle_c + (15\xi_2^2 \xi_1^{-6} - 4\xi_3 \xi_1^{-5}) \langle n^2 \rangle_c \\ &\quad - (15\xi_2^3 \xi_1^{-7} - 10\xi_2 \xi_3 \xi_1^{-6} + \xi_4 \xi_1^{-5}) \langle n \rangle.\end{aligned}$$

A. Bzdak and V. Koch : PRC.(2012)86.044904

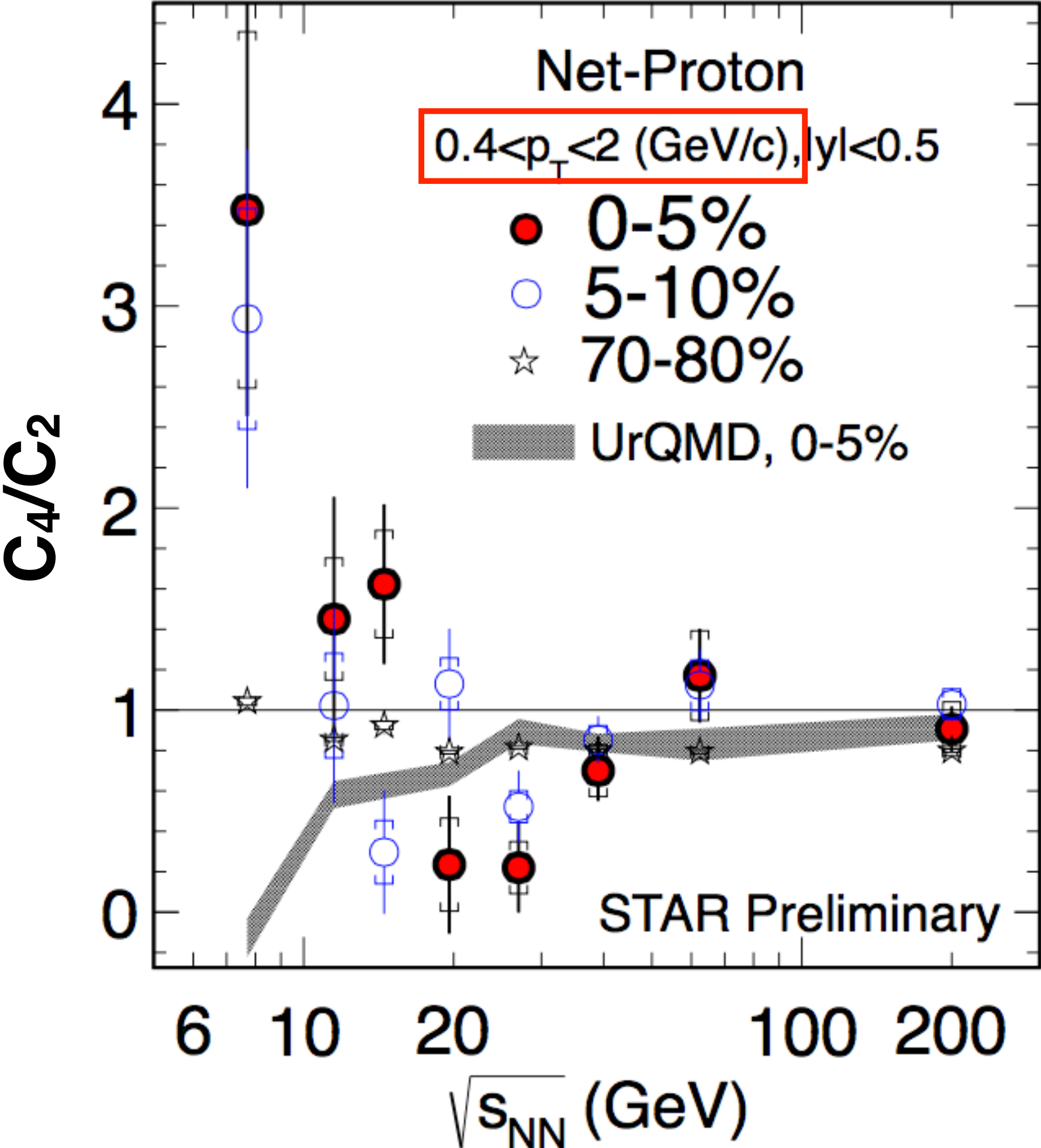
$$\begin{aligned}pK_1 &= c_1, \\ p^2K_2 &= c_2 - n(1-p), \\ p^3K_3 &= c_3 - c_1(1-p^2) - 3(1-p)(f_{20} - f_{02} - nc_1), \\ p^4K_4 &= c_4 - np^2(1-p) - 3n^2(1-p)^2 - 6p(1-p)(f_{20} + f_{02}) \\ &\quad + 12c_1(1-p)(f_{20} - f_{02}) - (1-p^2)(c_2 - 3c_1^2) - 6n(1-p)(c_1^2 - c_2) \\ &\quad - 6(1-p)(f_{03} - f_{12} + f_{02} + f_{20} - f_{21} + f_{30}).\end{aligned}$$

$$B(n; p, N) = p^n (1-p)^{N-n} \frac{N!}{n!(N-n)!}$$



Efficiency correction with binomial assumption

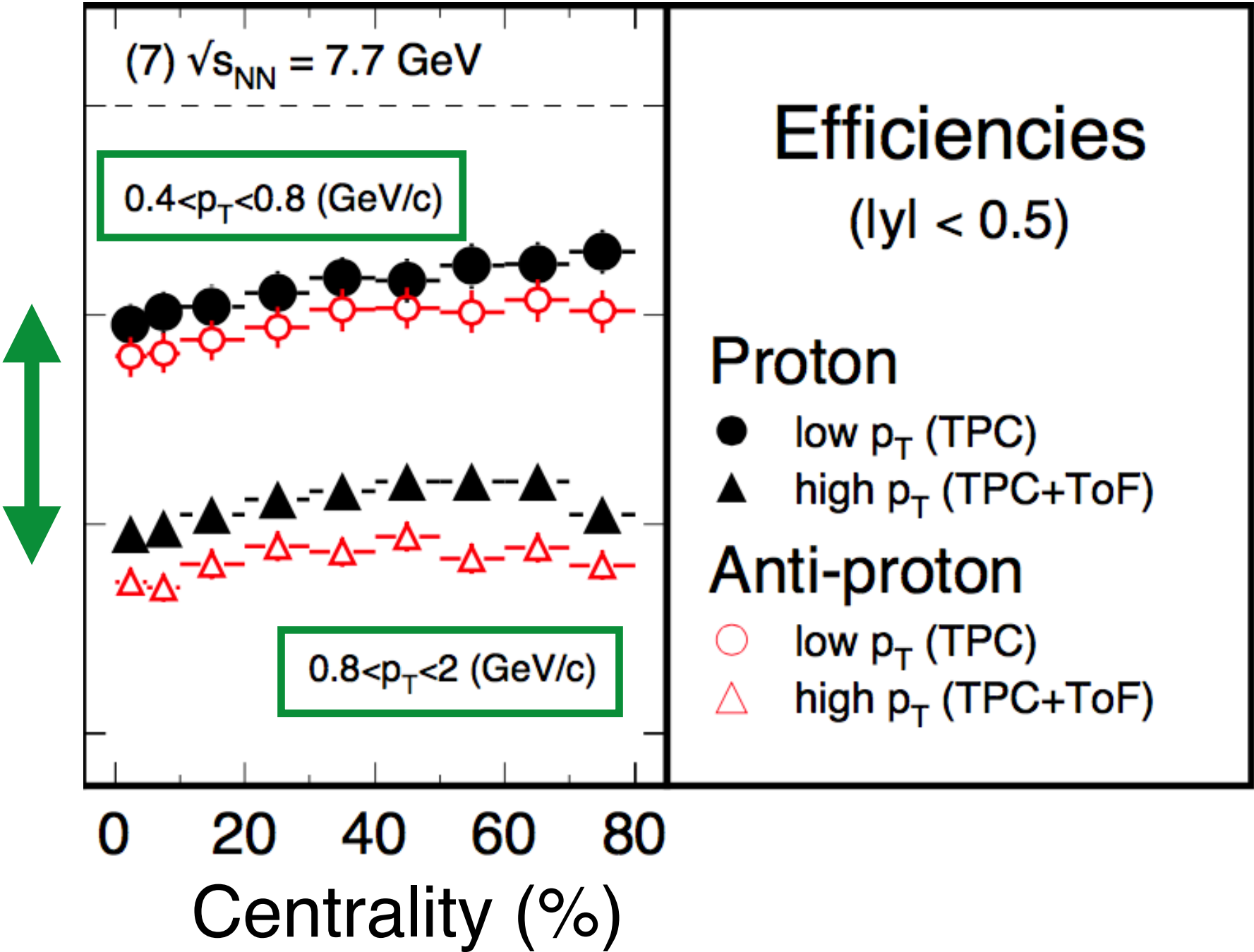
X. Luo (STAR collaboration) arXiv:1503.02558v2



✓ Efficiency correction was extended to multi-variable case (many efficiency bins).

A. Bzdak and V. Koch : PRC.91.(2015)027901

Different efficiency between two p_T regions due to different PID



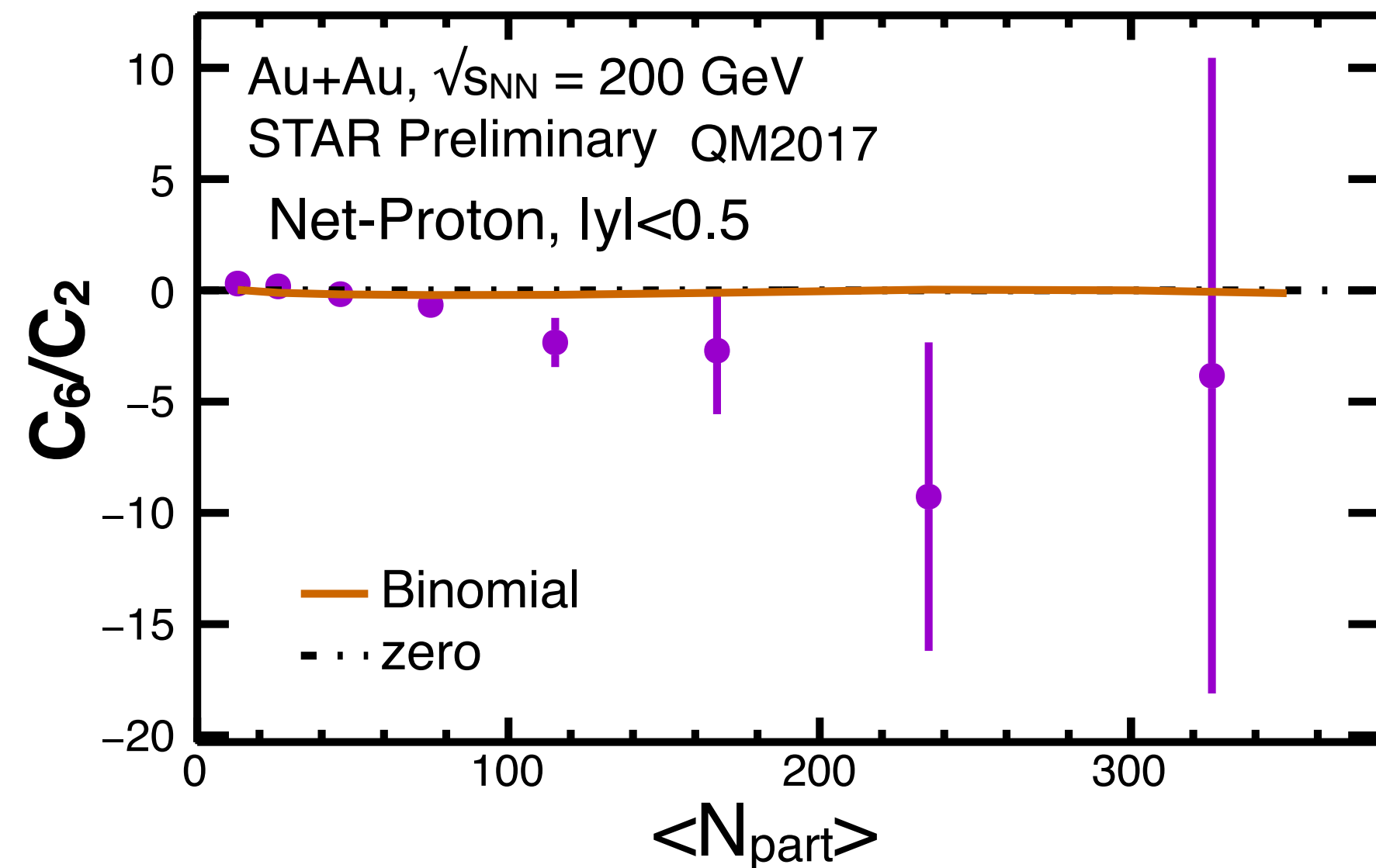
Efficient formulas

✓ Huge calculation cost for higher order with many efficiency bins.

$\sim M^m$ for large M .
 m : order of cumulant
 M : # of efficiency bins

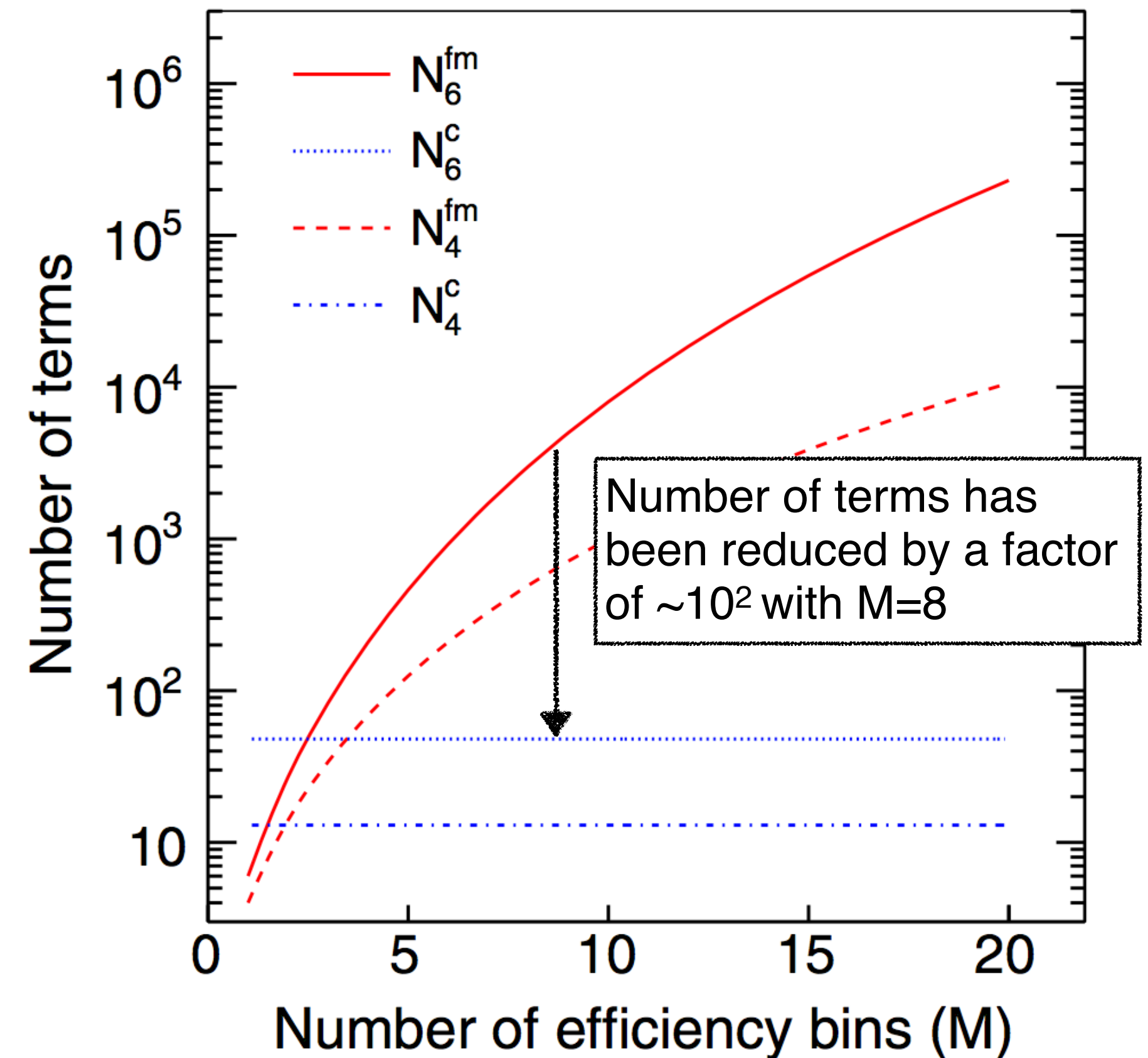
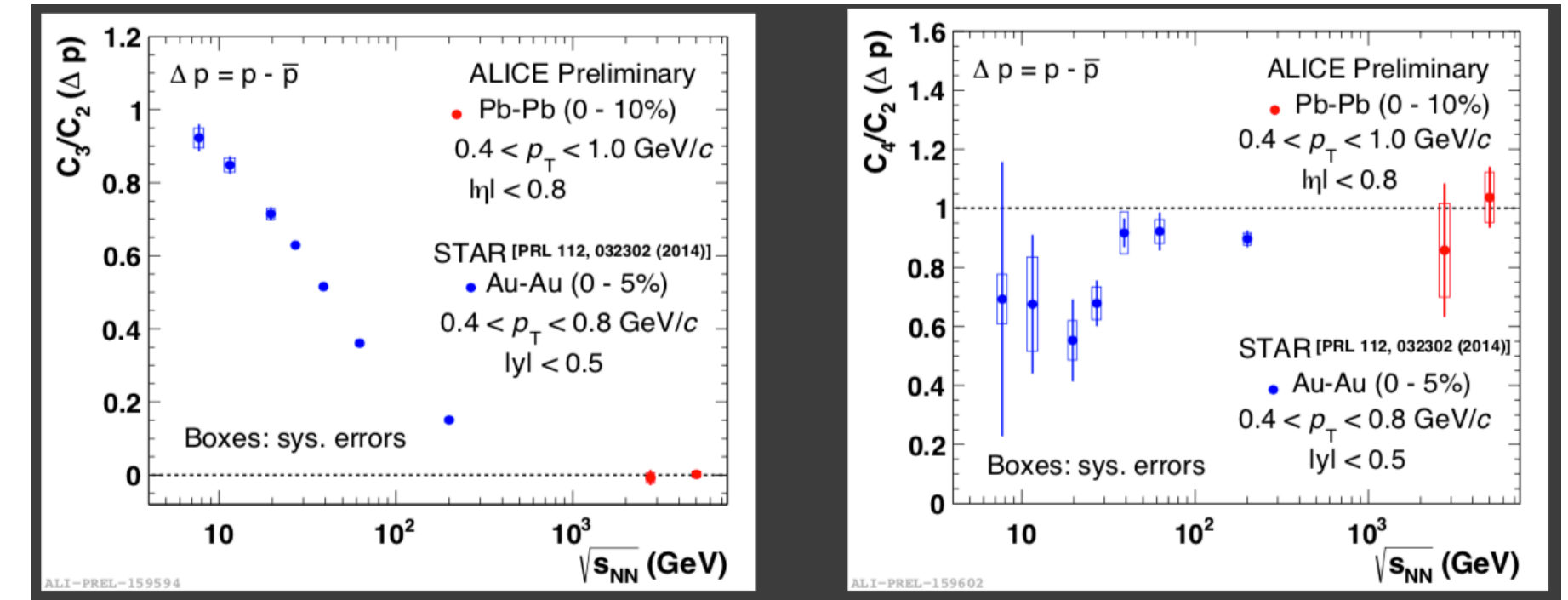
✓ More efficient formulas have been developed.

- M. Kitazawa : PRC.93.(2016)044911
- T. Nonaka, M. Kitazawa, S. Esumi, PRC.94.(2017)034909
- Used in STAR C_6 and ALICE C_4



T. Nonaka, QM2017, poster

N. K. Behera (ALICE), QM2018 → ~20 efficiency bins



Efficient formulas

- ✓ Huge calculation cost for higher order with many efficiency bins.

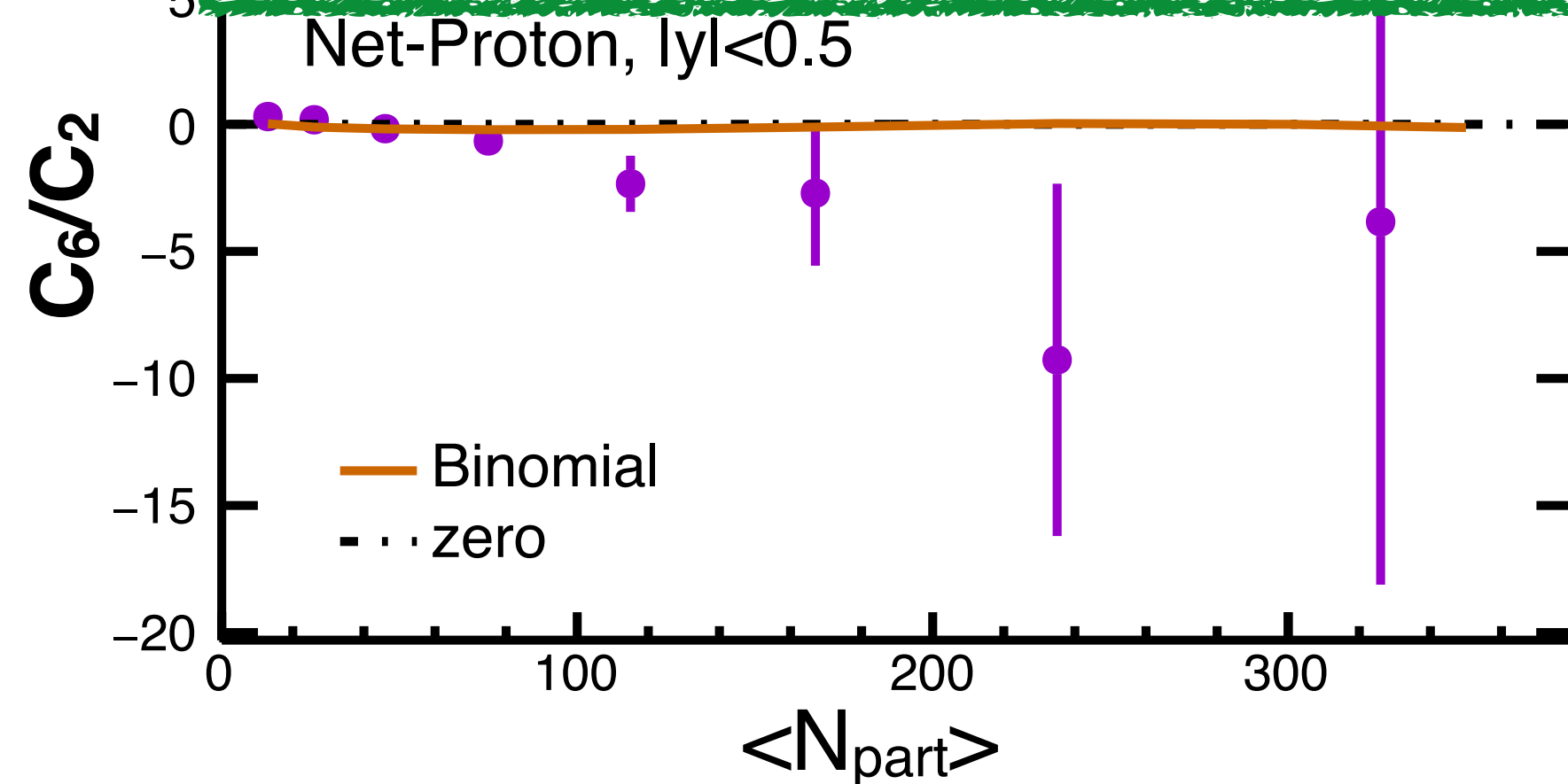
$\sim M^m$ for large M .
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- ✓ More efficient formulas have been developed.

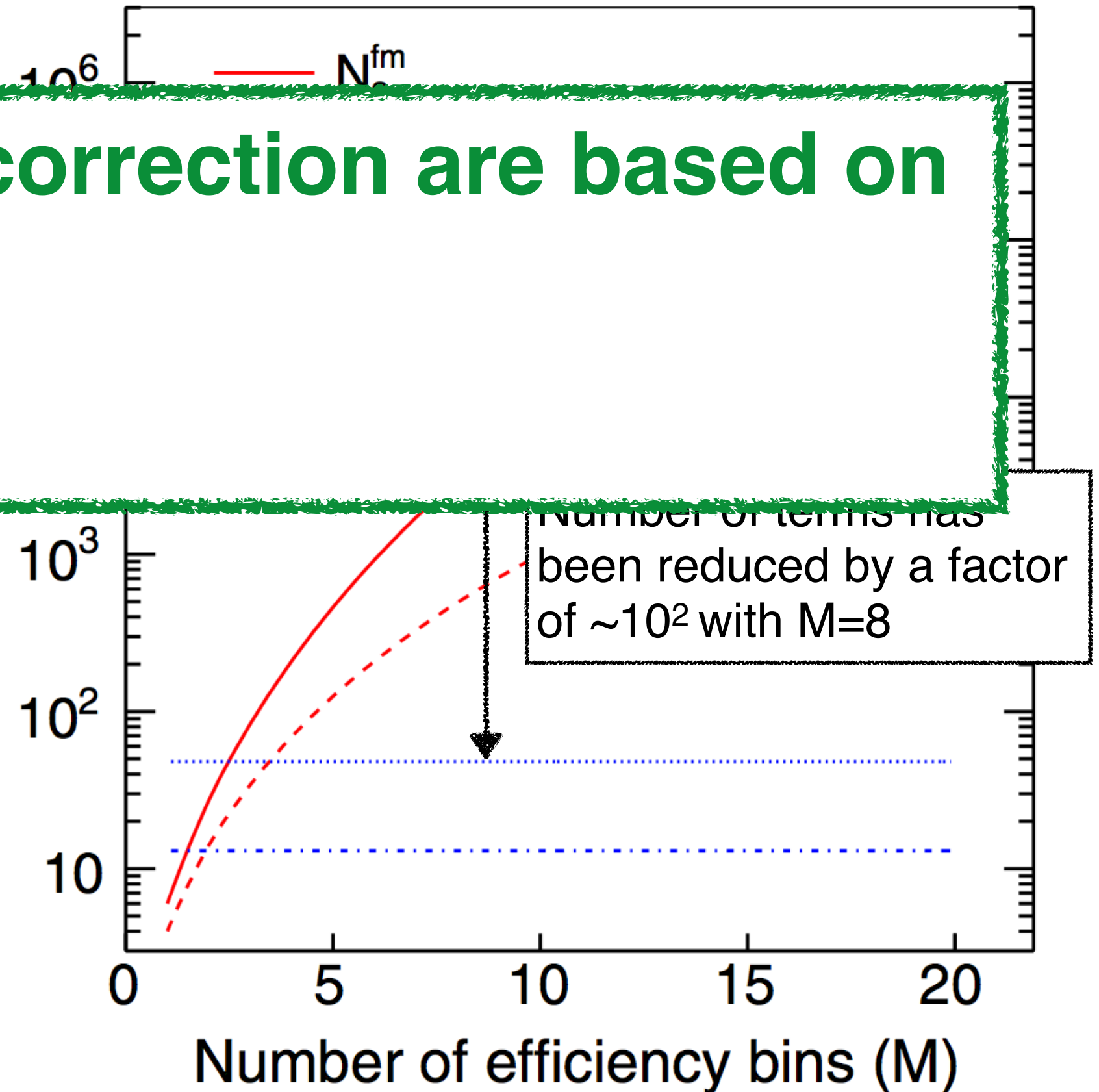
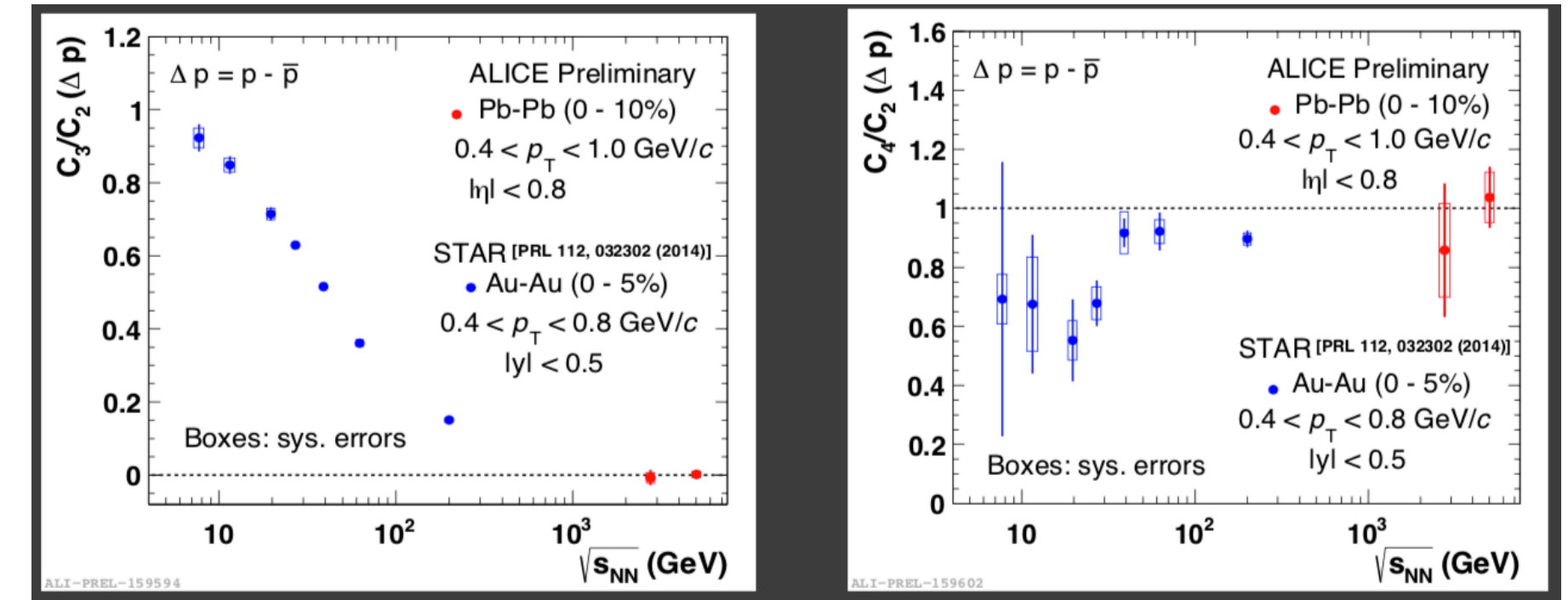
- M. Ki
- T. No
- Used

So far, all methods for efficiency correction are based on the binomial assumption.

Is experiment really binomial?



T. Nonaka, QM2017, poster

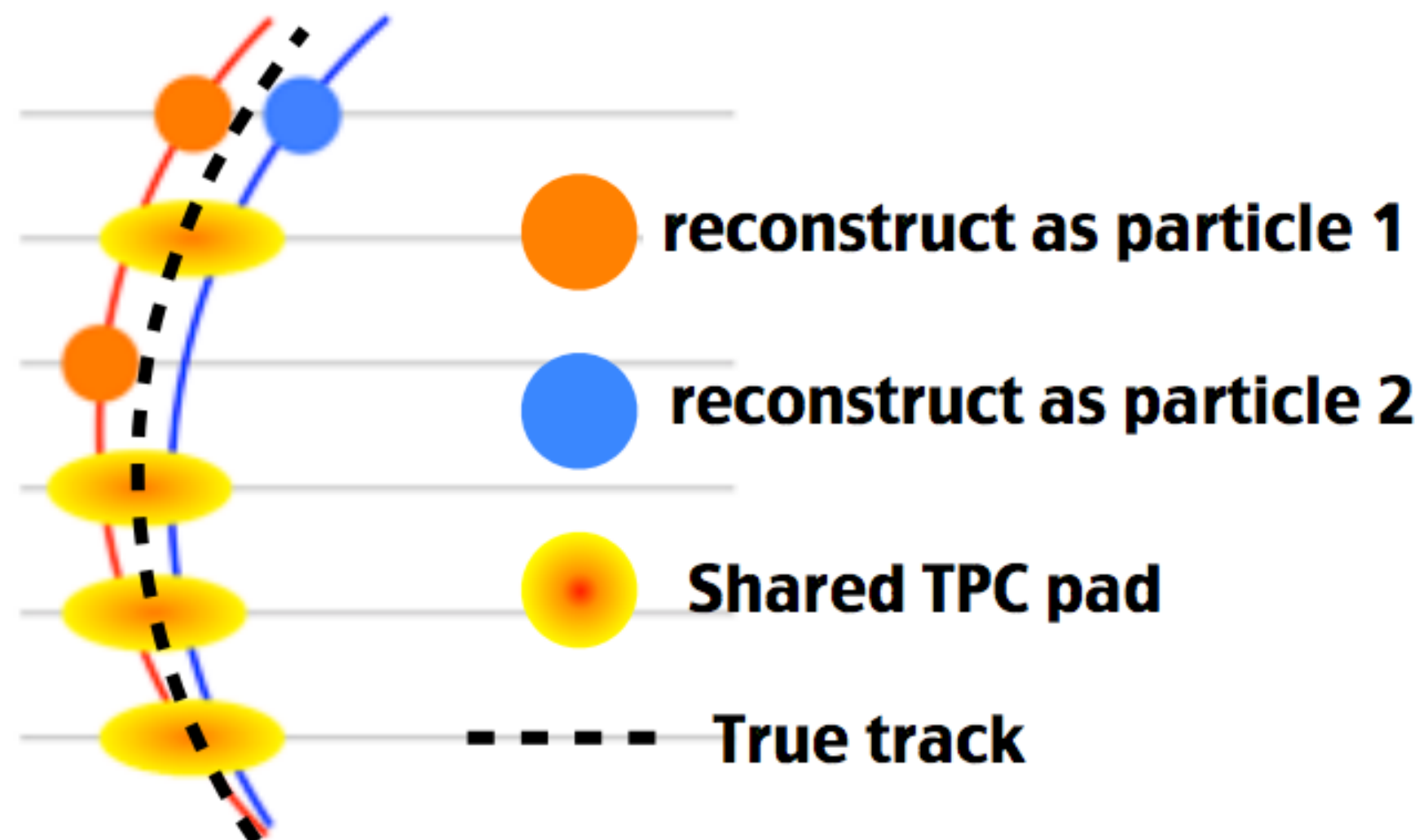


Where does non-binomial come from?

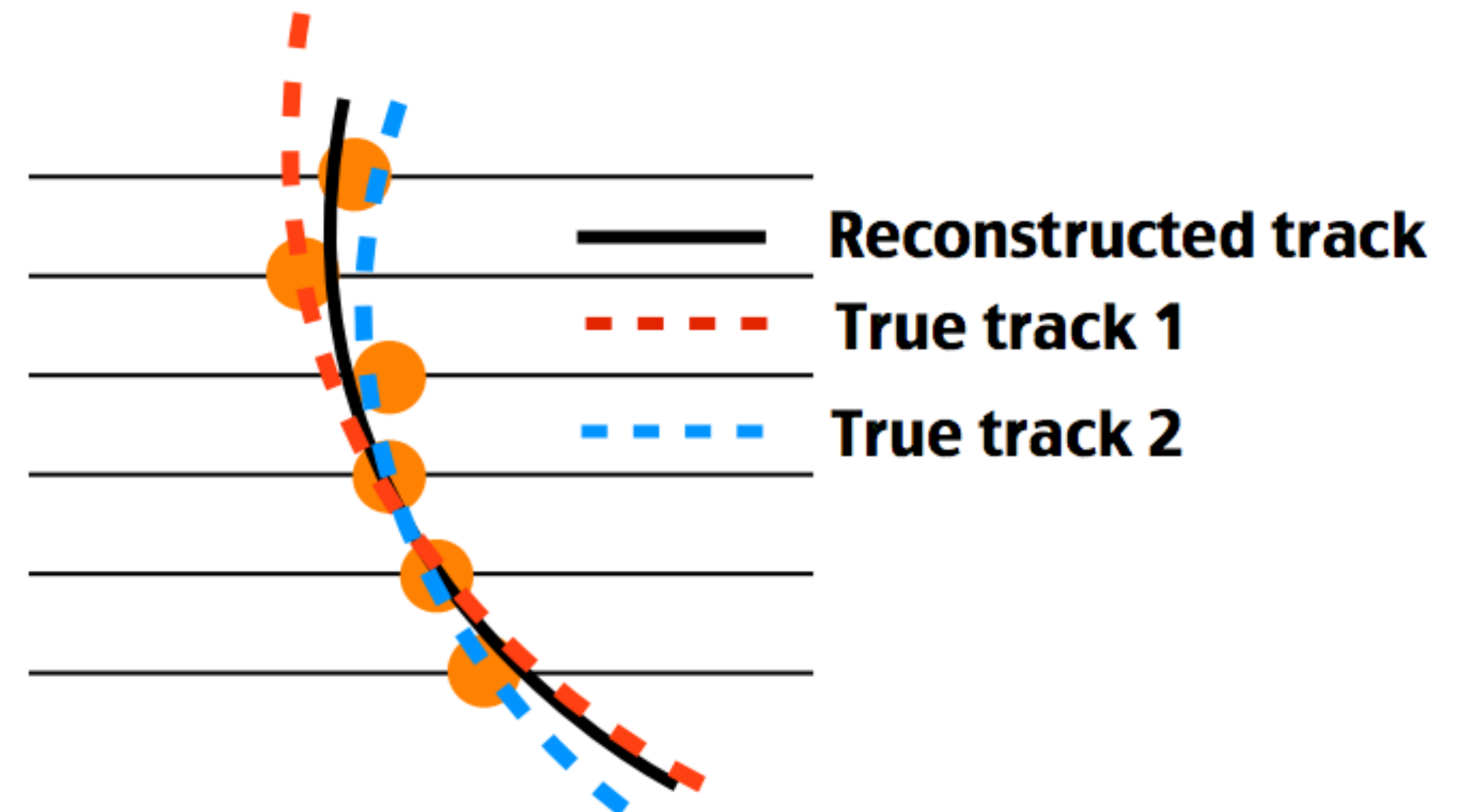
✓ There might be some reasons to make the response matrix non-binomial, which can be understood only by MC.

- Track splitting/merging
- Time dependence of efficiency
- Particle misidentification

Track splitting



Track merging



If experiment is non-binomial...

A.Bzdak, R. Holzmann, V. Koch : PRC.94.(2016)064907

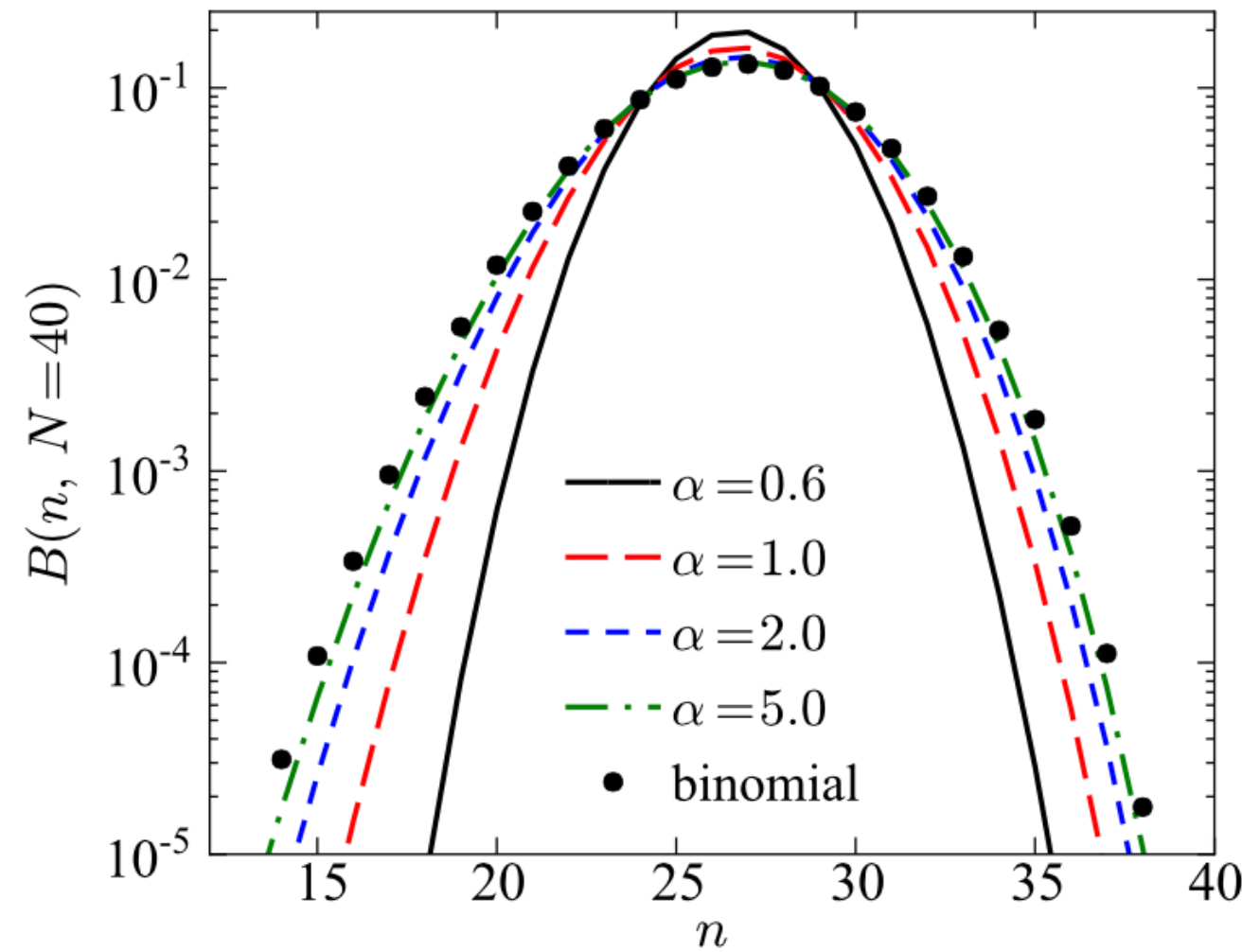


FIG. 2. The hypergeometric distribution for different values of α compared with the binomial distribution (black points). Here $N = 40$ and $\epsilon = 2/3$.

Hypergeometric

TABLE I. The obtained values of K_n/K_2 for the hypergeometric distribution, using $F_i = f_i/\epsilon^i$ with $\epsilon = 2/3$, for different values of α as presented in Fig. 2.

Hypergeometric	$\alpha = 0.6$	$\alpha = 1.0$	$\alpha = 2.0$	$\alpha = 5.0$
K_3/K_2	1.16	1.12	1.07	1.03
K_4/K_2	0.66	0.88	0.98	1.00
K_5/K_2	2.19	1.68	1.23	1.05
K_6/K_2	-3.99	-1.38	0.31	0.89

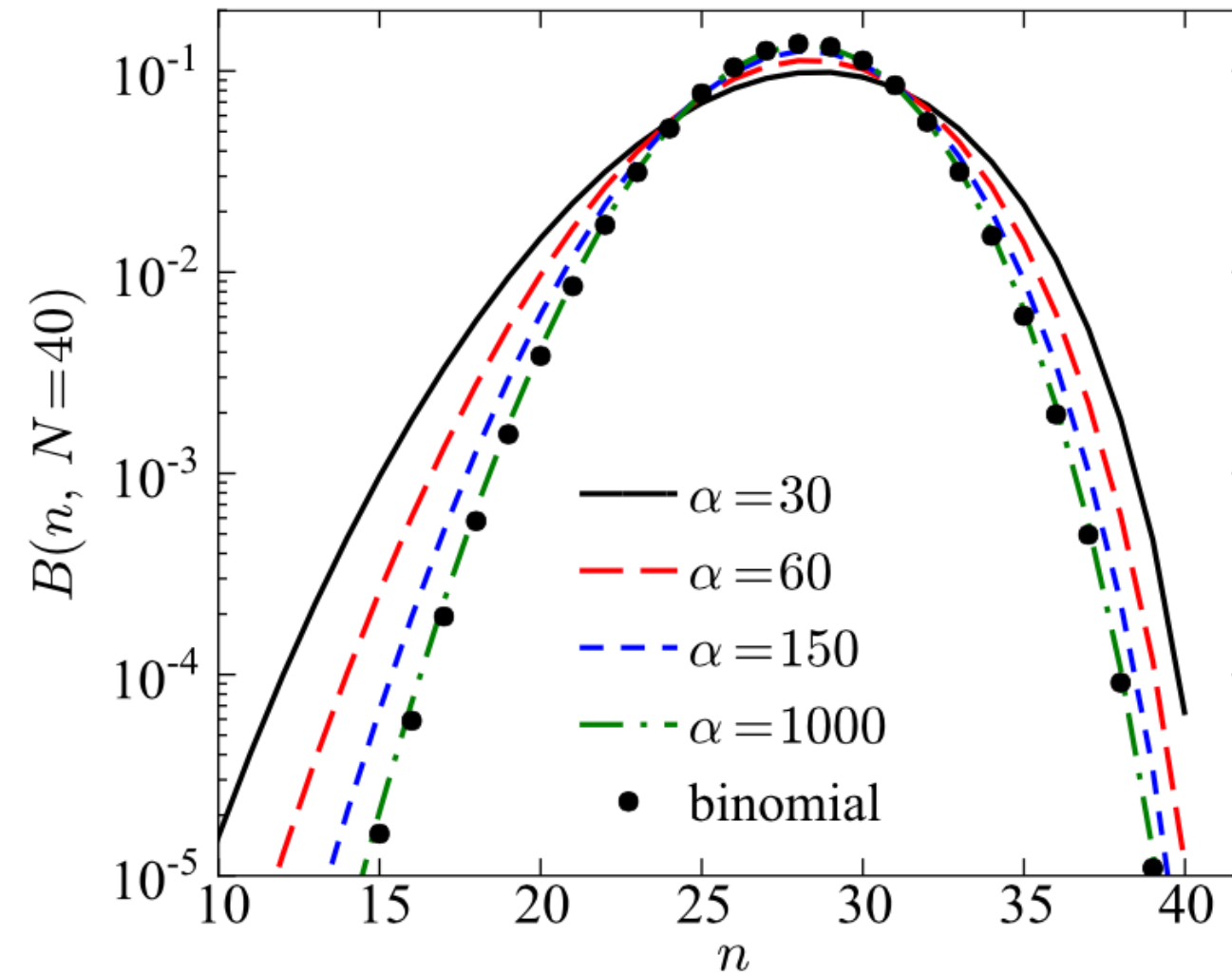


FIG. 3. The beta-binomial distribution for different values of α compared with the binomial distribution (black points). Here $N = 40$ and $\epsilon = 0.7$.

Beta-binomial

TABLE II. The obtained values of K_n/K_2 for the beta-binomial distribution, using $F_i = f_i/\epsilon^i$ with $\epsilon = 0.7$, for different values of α as presented in Fig. 3.

beta binomial	$\alpha = 30$	$\alpha = 60$	$\alpha = 150$	$\alpha = 1000$
K_3/K_2	1.28	1.24	1.13	1.02
K_4/K_2	0.82	1.45	1.35	1.07
K_5/K_2	-1.11	1.15	1.63	1.16
K_6/K_2	5.71	-0.44	1.80	1.32

- ✓ Unfolding is necessary. Reconstruct the distribution itself numerically.
- Estimation of systematic uncertainties is difficult.
- Huge calculation cost due to iterations.
- Distributions are not so necessary.

More general method to correct for detector effects by cumulants directly?

True cumulants are unity, but all results deviate.

Notations

$$\tilde{P}(n) = \sum_N \mathcal{R}(n; N) P(N). \quad (1)$$

$\mathcal{R}(n; N)$ Response matrix

$$\langle\langle n^m \rangle\rangle = \sum_n n^m \tilde{P}(n), \quad \text{Measured moments}$$

$$\langle N^m \rangle = \sum_N N^m P(N). \quad \text{True moments}$$

$$\begin{aligned} \langle\langle n^m \rangle\rangle &= \sum_N P(N) \sum_n n^m \mathcal{R}(n; N) \\ &= \sum_N P(N) R_m(N), \end{aligned} \quad (4)$$

$$R_m(N) = \sum_n n^m \mathcal{R}(n; N). \quad \text{Moments of response matrix with fixed } N$$

$$R_m(N) = \sum_{j=0}^{\infty} r_{mj} N^j. \quad (6)$$

- ✓ If Eq. (6) closes at finite order, Eq. (9) can be **exactly solved**.
- ✓ If not, some **truncations** are necessary.

$$\langle\langle n^m \rangle\rangle = \underline{r_{m0}} + \sum_{j=1}^{\infty} \underline{r_{mj}} \langle N^j \rangle. \quad (7)$$

Property of the detector

$$\begin{bmatrix} \langle N \rangle \\ \langle N^2 \rangle \\ \vdots \end{bmatrix} = \mathbf{R}^{-1} \begin{bmatrix} \langle\langle n \rangle\rangle \\ \langle\langle n^2 \rangle\rangle \\ \vdots \end{bmatrix} - \mathbf{R}^{-1} \begin{bmatrix} r_{10} \\ r_{20} \\ \vdots \end{bmatrix}. \quad (9)$$

General procedures

- Exactly solvable case
- Truncation

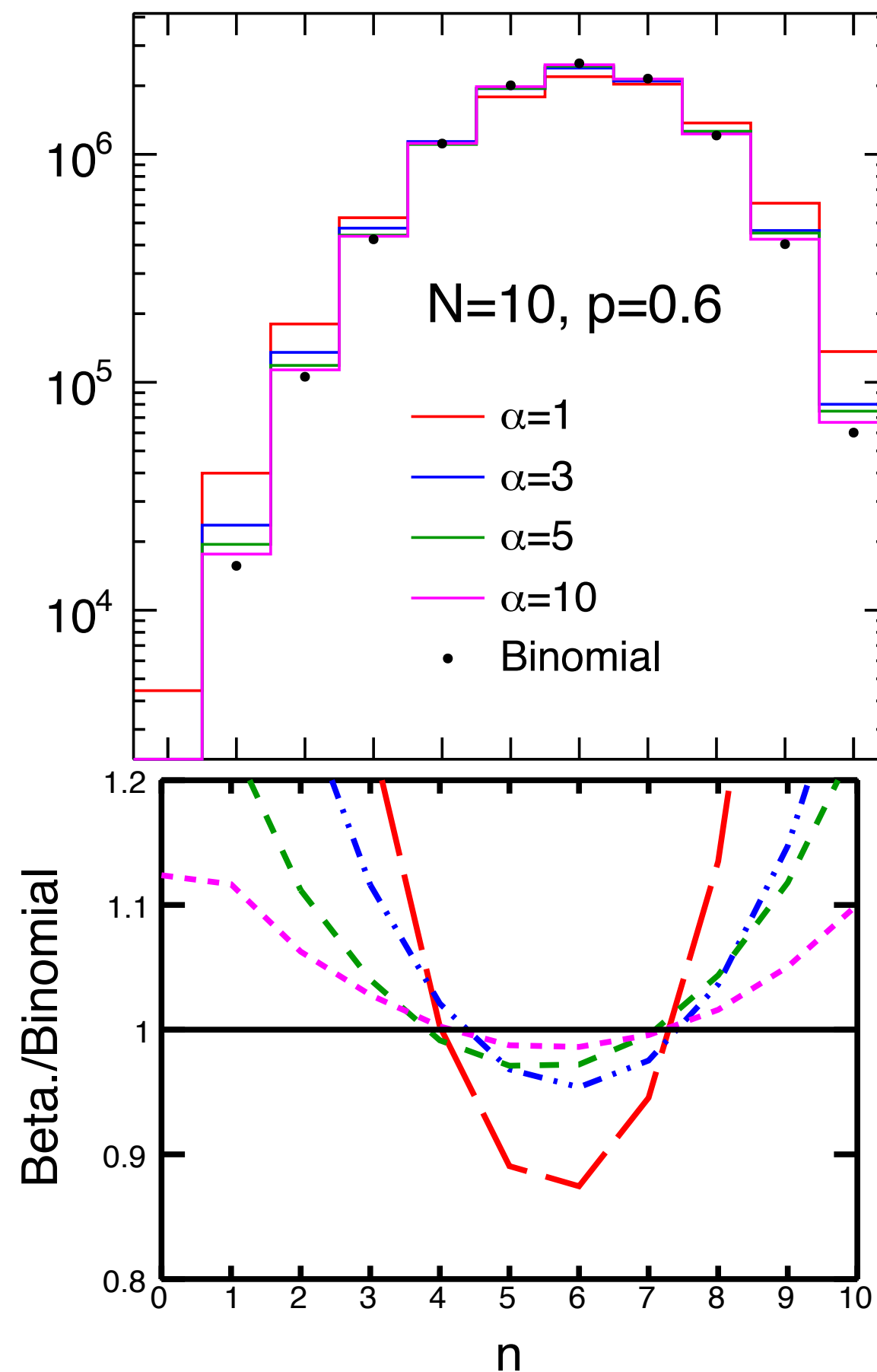
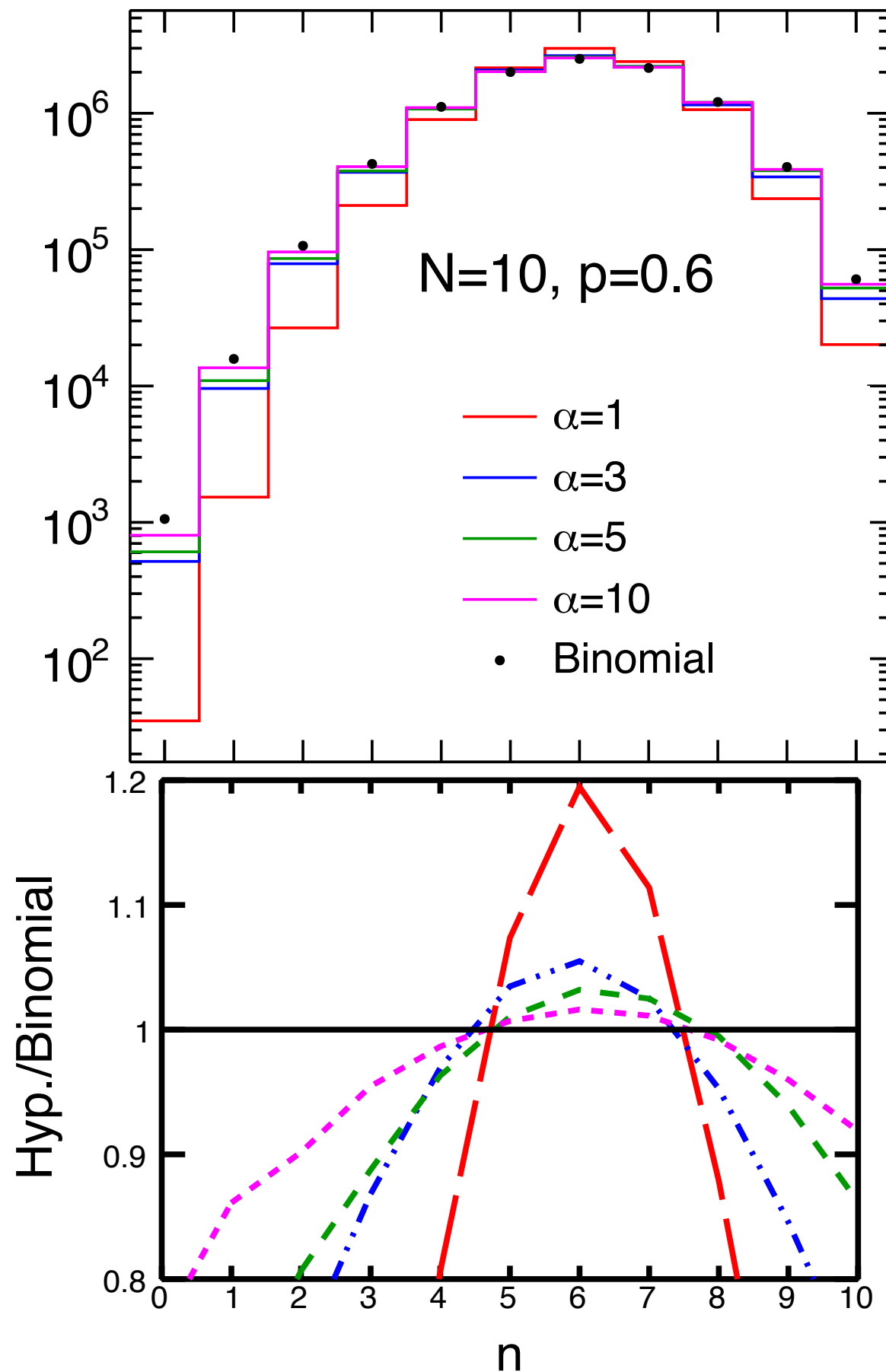
Non-binomial distribution

Hypergeometric distribution

Draw a ball from urn, if it is white, count particle. This is repeated **without replacement**.

Beta-binomial distribution

Draw a ball from urn, if it is white, count particle. And **return two white balls to urn** (similar for black balls).



- $X = N_w, Y = N_{\text{tot}} = N_w + N_b$
- $p = N_w / N_{\text{tot}} = X / Y$

$$\mathcal{R}_{\text{HG}}(n; N) = H(n; N, X, Y), \quad (18)$$

$$\mathcal{R}_{\beta}(n; N) = \beta(n; N, X, Y - X), \quad (19)$$

Close to binomial with $Y \rightarrow \infty$

✓ Use these two response distributions to check our new method.

✓ Also, these examples can be solved analytically, which is discussed in Appendix D.

Non-binomial RM

- ✓ Assume non-binomial response matrix, hypergeometric and beta-binomial distribution, with efficiency $p=0.7$.
- ✓ Generate $P(N)$ with Poisson distribution, then sample $\tilde{P}(n)$ by hypergeometric and beta-binomial response function.
- ✓ 10 times larger statistics are generated for MC.

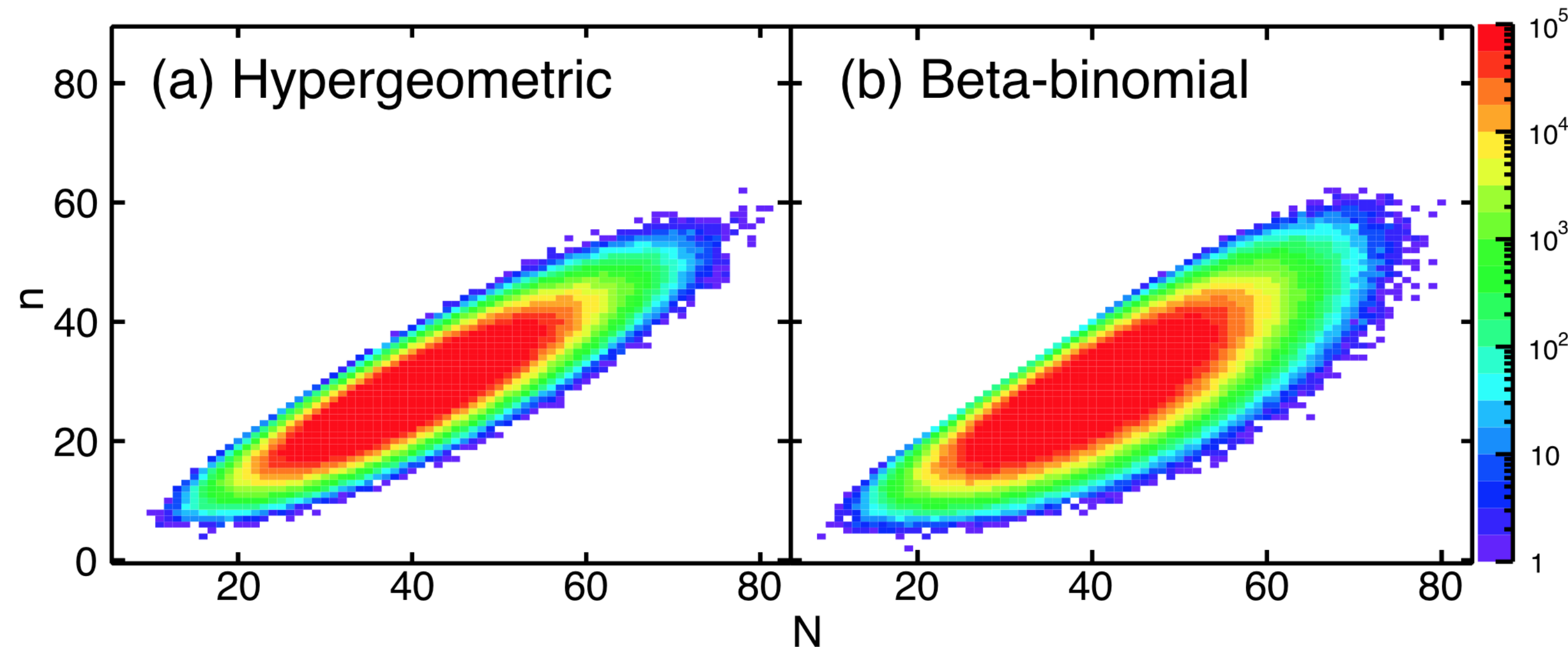
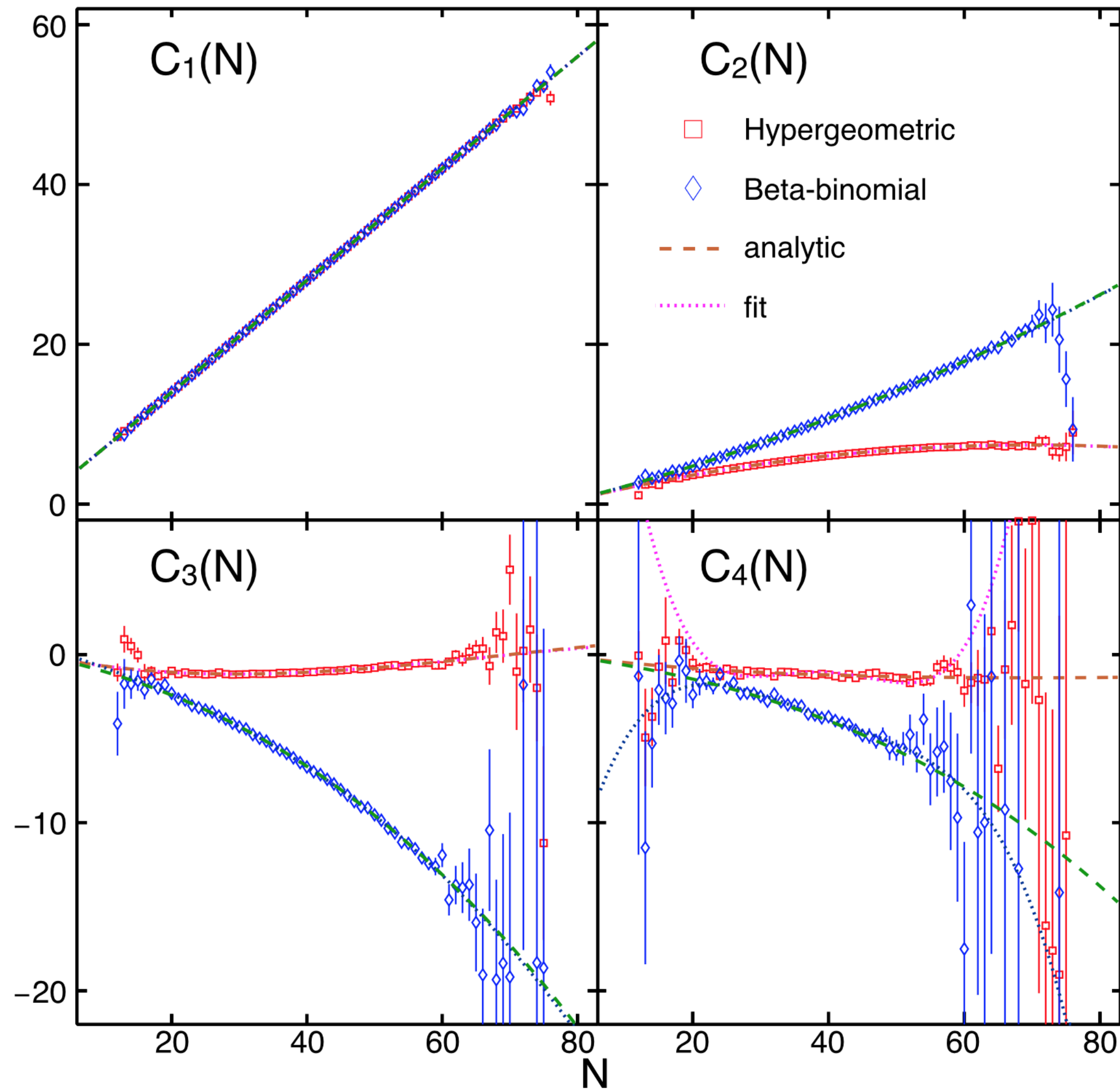


FIG. 1. Correlation between n and N on the sample events, i.e. the magnitude of $\mathcal{R}(n; N)P(N)$, for the response matrices $\mathcal{R}_{\text{HG}}(n; N)$ (hypergeometric) and $\mathcal{R}_{\beta}(n; N)$ (beta-binomial) with $p = X/Y = 0.7$ and $Y = 140$.

Extract detector properties



✓ Calculate m^{th} order moments of n at each N .

- Cumulants are shown here for better presentation.

✓ Fit by m^{th} polynomial, to extract r_{mj}

$$R_m(N) = \sum_{j=0}^m r_{mj} N^j. \quad (10)$$

✓ Solve the equations.

$$\langle\langle n^m \rangle\rangle = r_{m0} + \sum_{j=1}^{\infty} r_{mj} \langle N^j \rangle. \quad (7)$$

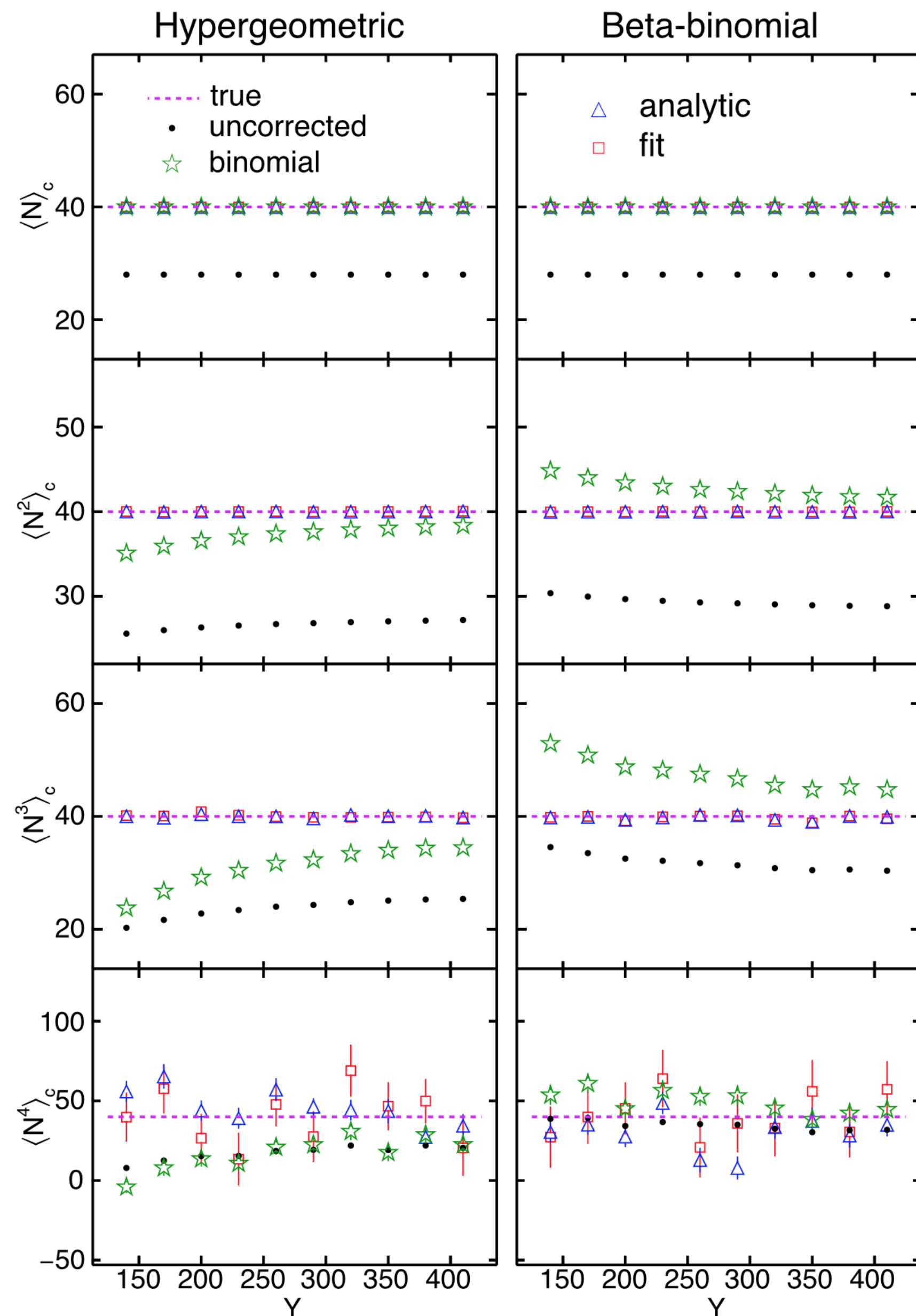
$$\langle\langle n \rangle\rangle = r_{10} + r_{11} \langle N \rangle$$

$$\langle\langle n^2 \rangle\rangle = r_{20} + r_{21} \langle N \rangle + r_{22} \langle N^2 \rangle$$

$$\langle\langle n^3 \rangle\rangle = r_{30} + r_{31} \langle N \rangle + r_{32} \langle N^2 \rangle + r_{33} \langle N^3 \rangle$$

$$\langle\langle n^4 \rangle\rangle = r_{40} + r_{41} \langle N \rangle + r_{42} \langle N^2 \rangle + r_{43} \langle N^3 \rangle + r_{44} \langle N^4 \rangle$$

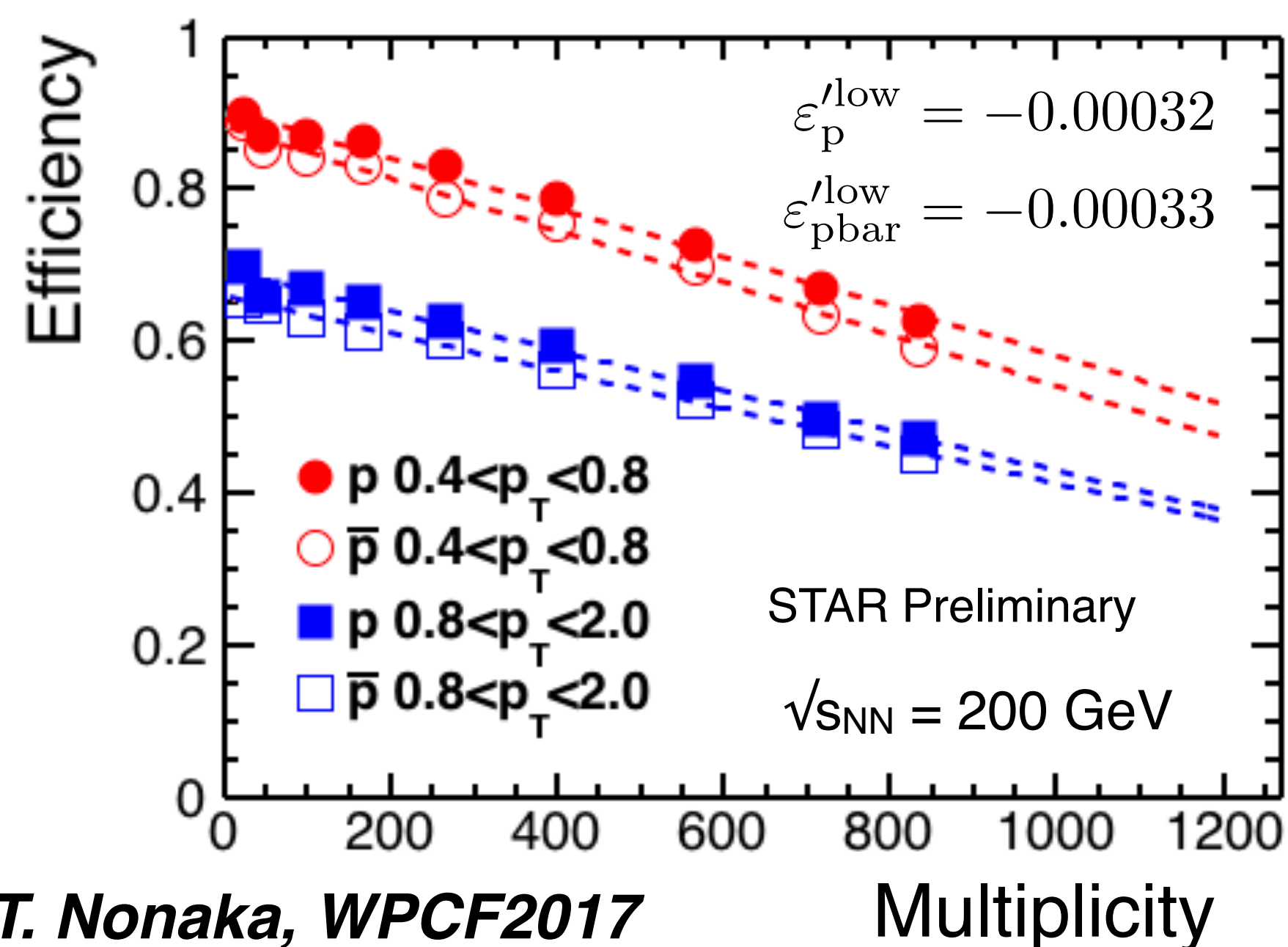
Results



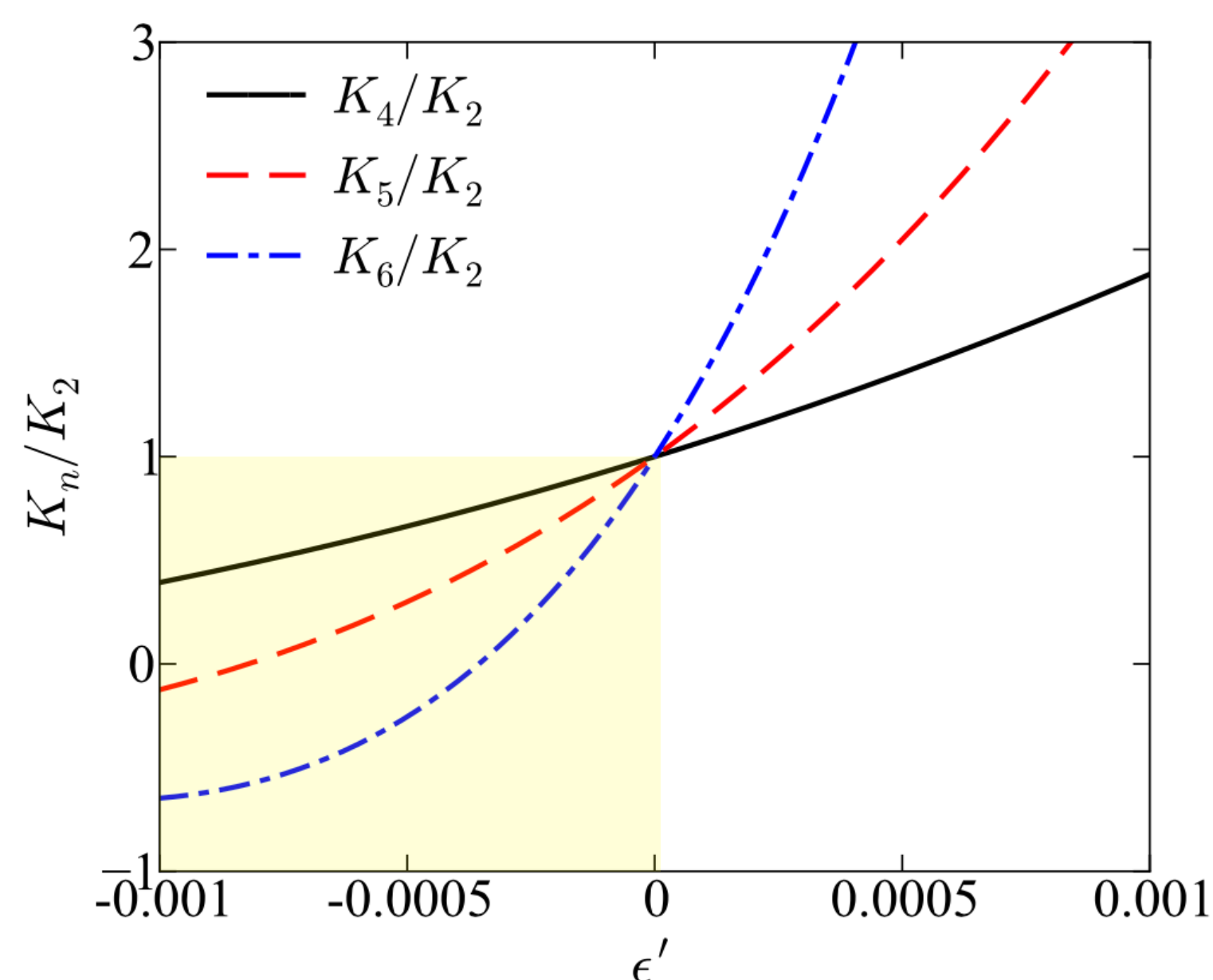
- ✓ **Check the correction as a function of different non-binomial parameter γ .**
- ✓ **Conventional correction assuming binomial efficiencies does not work.**
- ✓ **Our new method works well.**
- ✓ **Hypergeometric and beta-binomial examples can be also solved analytically.**

Multiplicity dependent efficiency

- ✓ Efficiency decreases with increasing multiplicity, which artificially decreases the value of cumulants.
- ✓ Residual dependence of efficiency inside one multiplicity bin (for centrality) needs to be taken into account.



T. Nonaka, WPCF2017



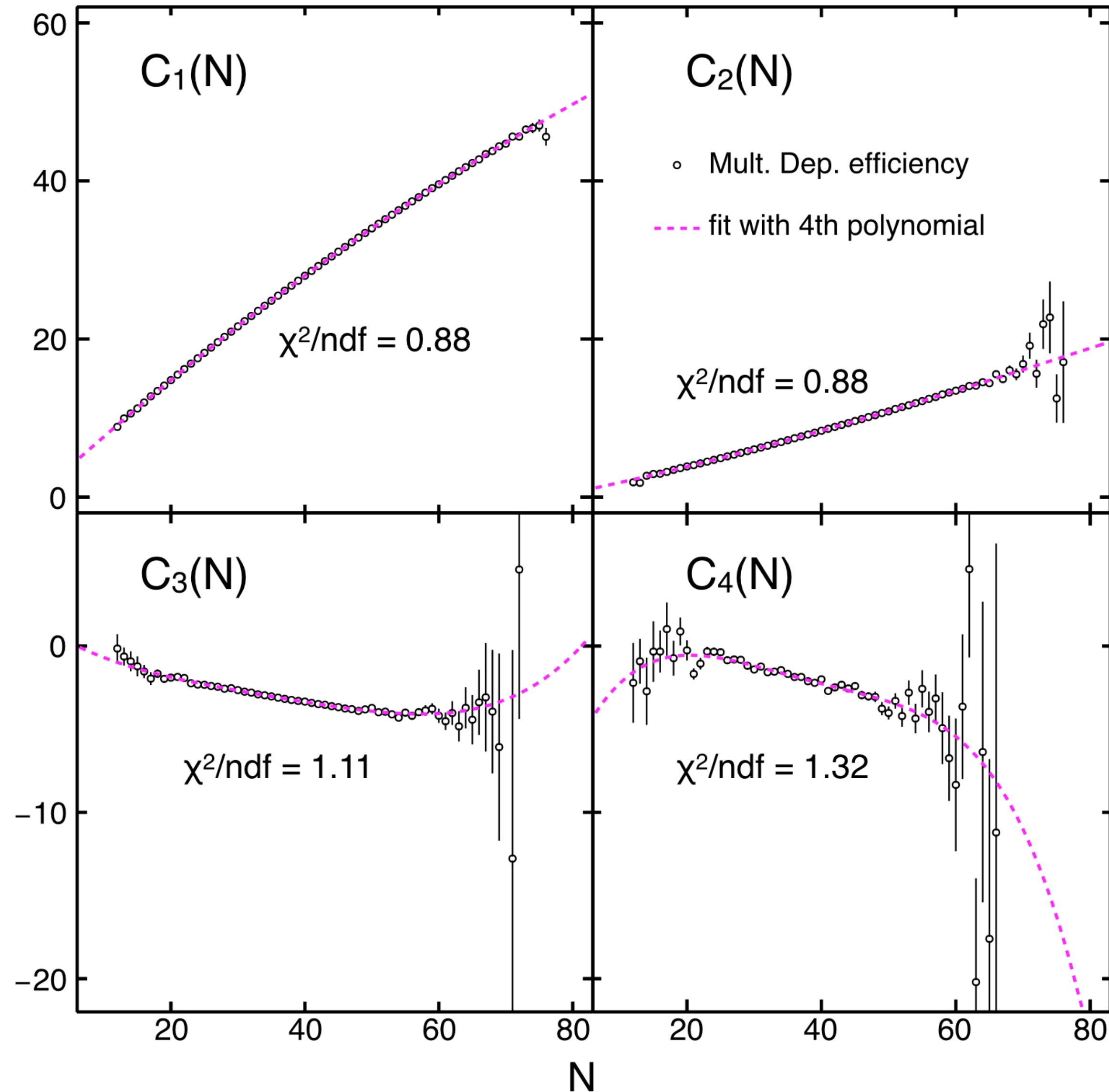
A. Bzdak, R. Holzmann, V. Koch :
PRC.94.064907

$$P(N) = \frac{\langle N \rangle^N}{N!} e^{-\langle N \rangle},$$

$$\epsilon(N) = \epsilon_0 + \epsilon'(N - \langle N \rangle),$$

Multiplicity dependent efficiency

$$p(N) = 0.7 - \varepsilon(N - \langle N \rangle) \quad \varepsilon = 0.002$$



✓ $C_1(N) \propto N^2$, $C_2(N) \propto N^4$, $C_3(N) \propto N^6$

- Cannot be exactly solved.
- Truncation is necessary.

✓ Fit by 4th polynomial, to extract r_{mj}

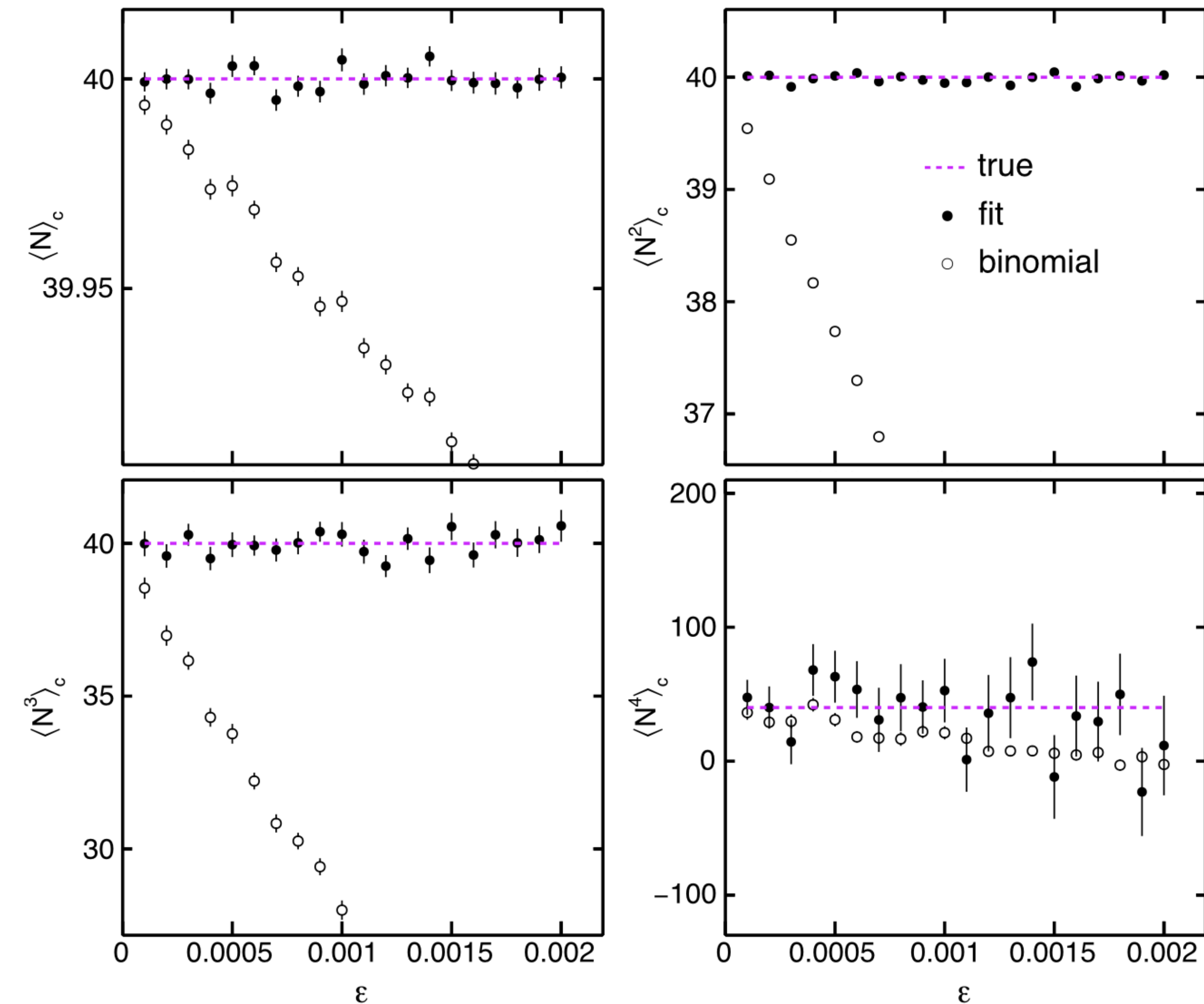
$$R_m(N) = \sum_{j=0}^{m-L=4} r_{mj} N^j. \quad (10)$$

✓ R will be L x L matrix

$$\begin{bmatrix} \langle\langle n \rangle\rangle \\ \langle\langle n^2 \rangle\rangle \\ \vdots \\ \langle\langle n^L \rangle\rangle \end{bmatrix} = \begin{bmatrix} r_{10} \\ r_{20} \\ \vdots \\ r_{L0} \end{bmatrix} + \mathbf{R} \begin{bmatrix} \langle N \rangle \\ \langle N^2 \rangle \\ \vdots \\ \langle N^L \rangle \end{bmatrix}, \quad (17)$$

Multiplicity dependent efficiency

$$p(N) = 0.7 - \varepsilon(N - \langle N \rangle) \quad \varepsilon = 0.002$$



- ✓ Check the correction as a function of efficiency slope ε .
- ✓ Conventional method dose not work.
- ✓ Truncation at 4th order works well.
- ✓ With high statistics in MC, the truncation at 4th order would lose it's validity
- ✓ Validity of truncation can be checked by changing the truncation order like L^{th} , $L+1^{\text{th}}$, $L+2^{\text{th}}$ order.

Additional comments

- ✓ **In the case of net-particle distribution, 2-D fitting is necessary.**
- ✓ **We assumed $P_{MC}(N) = P(N)$, which is not the case in the real experiment, since we don't know true $P(N)$.**
 - Correct results were obtained by using a Gauss distribution $C_1=40$ and $C_2=\sqrt{40}$ for $P_{MC}(N)$ with respect to the Poisson distribution for $P(N)$ with $C_1=40$.
- ✓ **We should try several models to determine $R(n;N)$, which should be included as systematic uncertainties**
 - HIJING+GEANT?? Or embedding simulation??
- ✓ **It would be better if data driven method will be developed in the future.**

Summary

- General procedures to correct non-binomial detector effects were presented.
- Truncation will be necessary if efficiency depends on the multiplicity.
- Systematic studies using different models would be needed.

Thank you!

Back up

Cumulants of binomial distribution

✓ Cumulants of binomial distribution is proportional to N .

M. Asakawa and M. Kitazawa, Prog. Part. Nucl. Phys. 90, 299 (2016), arXiv:1512.05038

$$B_{p,N}(m) = {}_N C_m p^m (1-p)^{N-m}, \quad {}_N C_m = \frac{N!}{m!(N-m)!},$$

$$\langle m^n \rangle_c = \xi_n N,$$

$$\xi_1 = p, \quad \xi_2 = p(1-p), \quad \xi_3 = p(1-p)(1-2p), \quad \xi_4 = p(1-p)(1-6p+6p^2).$$

Cumulants of non-binomial distributions

$$\langle n \rangle_c^{\text{HG}} = Np, \quad (\text{D4})$$

$$\langle n^2 \rangle_c^{\text{HG}} = \frac{N(N - Y)(-1 + p)p}{-1 + Y}, \quad (\text{D5})$$

$$\langle n^3 \rangle_c^{\text{HG}} = \frac{N(2N^2 - 3NY + Y^2)(-1 + p)p(-1 + 2p)}{(-2 + Y)(-1 + Y)}, \quad (\text{D6})$$

$$\begin{aligned} \langle n^4 \rangle_c^{\text{HG}} = & [N(N - Y)(-1 + p)p(6N^2(-1 - 6(-1 + p)p + Y(1 + 5(-1 + p)p)) \\ & - 6NY(-1 - 6(-1 + p)p + Y(1 + 5(-1 + p)p)) \\ & + (-1 + Y)Y(1 + Y(1 + 6(-1 + p)p)))] / [(-3 + Y)(-2 + Y)(-1 + Y)^2], \end{aligned} \quad (\text{D7})$$

$$\langle n \rangle_c^\beta = Np, \quad (\text{D8})$$

$$\langle n \rangle_c^\beta = -\frac{N(N + a + b)(-1 + p)p}{1 + a + b}, \quad (\text{D9})$$

$$\langle n \rangle_c^\beta = \frac{N(N + a + b)(2N + a + b)(-1 + p)p(-1 + 2p)}{(1 + a + b)(2 + a + b)}, \quad (\text{D10})$$

$$\begin{aligned} \langle n \rangle_c^\beta = & -((N(N + a + b)(-1 + p)p(6N^2(1 + 6(-1 + p)p + (a + b)(1 + 5(-1 + p)p)) \\ & + 6N(a + b)(1 + 6(-1 + p)p + (a + b)(1 + 5(-1 + p)p)) + (a + b)(1 + a + b) \\ & (-1 + (a + b)(1 + 6(-1 + p)p)))) / ((1 + a + b)^2(2 + a + b)(3 + a + b)). \end{aligned} \quad (\text{D11})$$

Non-binomial efficiency

