

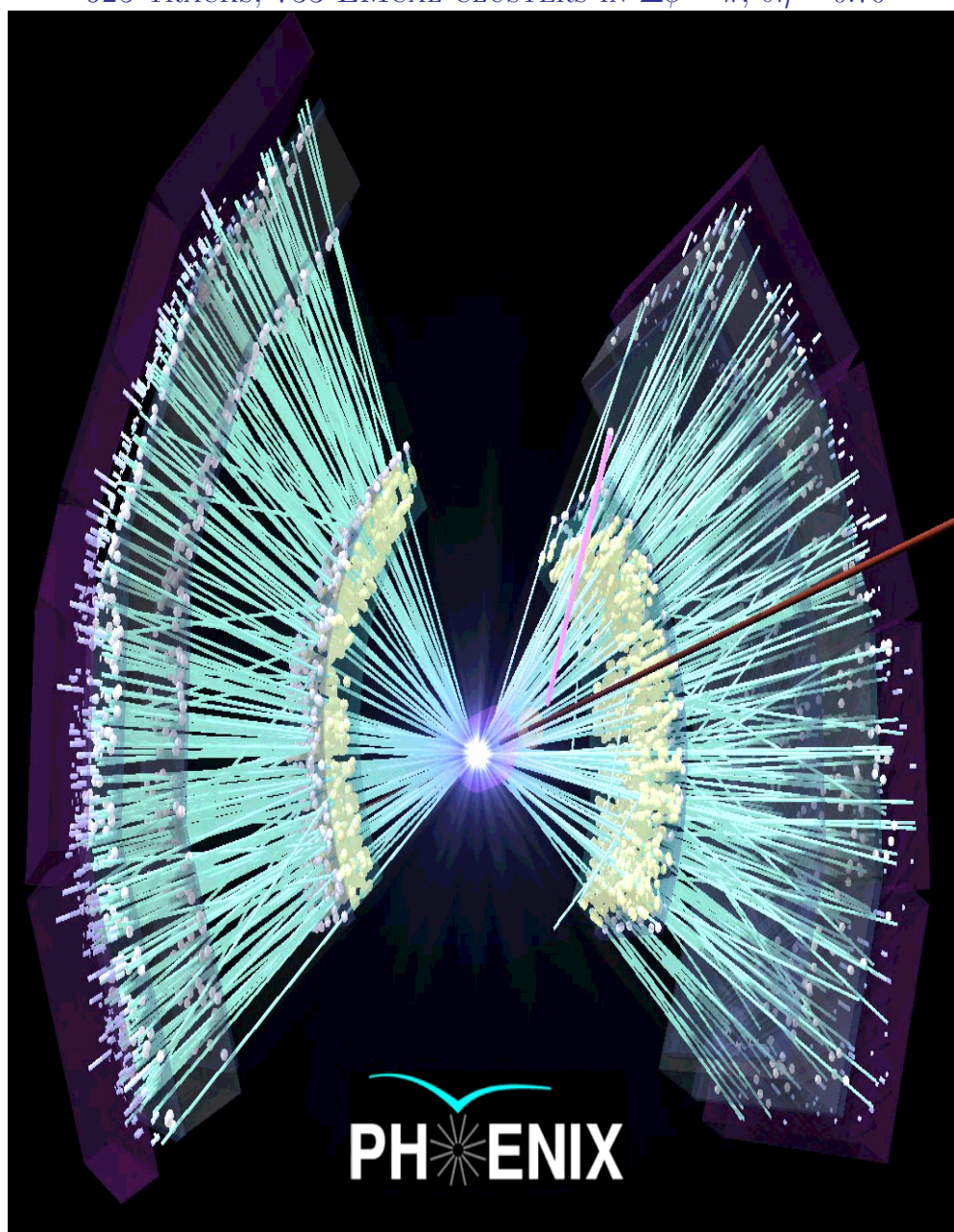
# $E_T$ Distributions and other Event-by-Event Fluctuations

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625 TRACKS, 733 EMCAL CLUSTERS IN  $\Delta\phi = \pi$ ,  $\delta\eta = 0.70$



CURRENT AND FUTURE DIRECTIONS AT RHIC, AUGUST 8, 2002

## Some Favorite Statistics

• In the theory of probability and statistics, a statistic is a quantity computed entirely from the sample, i.e. a statistic is any function of the observed sample values.

• Two of the most popular statistics are the sum and the average:

$$S_n \equiv \sum_{i=1}^n x_i \quad (1)$$

$$\bar{x}_{(n)} \equiv \frac{1}{n} \sum_{i=1}^n x_i = \frac{1}{n} S_n \quad (2)$$

where the  $x_i$  are  $n$  samples from a the same population or probability density function,  $f(x)$ .

• From the theory of mathematical statistics, the probability distribution of a random variable  $S_n$ , which is itself the sum of  $n$  independent random variables with a common distribution  $f(x)$ :

$$S_n = x_1 + x_2 + \cdots + x_n \quad (3)$$

is given by  $f_n(x)$ , the  $n$ -fold convolution of the distribution  $f(x)$ :

$$f_n(x) = \int_0^x dy f(y) f_{n-1}(x-y) \quad . \quad (4)$$

The mean,  $\mu_n = \langle S_n \rangle$ , and standard deviation,  $\sigma_n$ , of the  $n$ -fold convolution obey the familiar rule

$$\mu_n = n\mu \quad \sigma_n = \sigma\sqrt{n} \quad , \quad (5)$$

where  $\mu$  and  $\sigma$  are the mean and standard deviation of the distribution  $f(x)$ .

• A complementary case is that of a random variable  $Z_n$ , which is the sum of  $n$  random variables with a common distribution  $f(x)$ —which are themselves 100% correlated—for example:

$$Z_n = x + x + \cdots + x = nx \quad . \quad (6)$$

This is just a scale transformation. The behavior of the mean and the standard deviation for a scale transformation is  $\mu \rightarrow n\mu$ ,  $\sigma \rightarrow n\sigma$ , which is quite different than the behavior of the standard deviation under convolution (Eq. 5).

## Some Favorite Functions

**The Gamma distribution** is an example of a probability density function (pdf) which has particularly simple properties under convolutions and scale transformations. The Gamma distribution is a function of a continuous variable  $x$  and has parameters  $p$  and  $b$

$$f(x) = f_{\Gamma}(x, p, b) = \frac{b}{\Gamma(p)} (bx)^{p-1} e^{-bx} \quad (7)$$

where

$$p > 0, \quad b > 0, \quad 0 \leq x \leq \infty$$

$\Gamma(p) = (p-1)!$  if  $p$  is an integer, and  $f(x)$  is normalized,  $\int_0^{\infty} f(x) dx = 1$ . The mean and standard deviation of the distribution are

$$\mu \equiv \langle x \rangle = \frac{p}{b} \quad \sigma \equiv \sqrt{\langle x^2 \rangle - \langle x \rangle^2} = \frac{\sqrt{p}}{b} \quad \frac{\sigma^2}{\mu^2} = \frac{1}{p} \quad . \quad (8)$$

The  $n$ -fold convolution of the Gamma distribution (Eq. 7) is simply given by the function

$$f_n(x) = \frac{b}{\Gamma(np)} (bx)^{np-1} e^{-bx} = f_{\Gamma}(x, np, b) \quad (9)$$

i.e.  $p \rightarrow np$  and  $b$  remains unchanged. Note that the mean and standard deviation of Eq. 9

$$\mu_n = \frac{np}{b} \quad \sigma_n = \frac{\sqrt{np}}{b} \quad \frac{\sigma_n}{\mu_n} = \frac{1}{\sqrt{np}} \quad (10)$$

when compared to Eq. 8 explicitly obey Eq. 5. The result of a scale transformation  $x \rightarrow nx$  for a Gamma distribution (Eq. 7) is simply  $b \rightarrow b/n$ , with  $p$  remaining unchanged. **To summarize, the  $n$ -th convolution of the Gamma distribution  $f_{\Gamma}(x, p, b)$  is  $f_{\Gamma}(x, np, b)$ ; the scale transformation  $x \rightarrow nx$  of  $f_{\Gamma}(x, p, b)$  is  $f_{\Gamma}(x, p, b/n)$ .**

## The $p_T$ distribution is a Gamma Distribution

The principal advantage of the Gamma distribution for the present problem is that it is one of the standard representations of the inclusive single particle  $p_T$  distribution:

$$\frac{d\sigma}{p_T dp_T} = b^2 e^{-bp_T} \quad (11)$$

$$\frac{d\sigma}{dp_T} = b^2 p_T e^{-bp_T} \quad . \quad (12)$$

Clearly, Eq.s 11, 12 represent a Gamma distribution with  $p = 2$ ,  $\langle p_T \rangle = 2/b$ , where typically  $b = 6 \text{ (GeV/c)}^{-1}$  for p-p collisions. The ‘inverse slope parameter’  $1/b$  is sometimes referred to as the ‘Temperature parameter’.

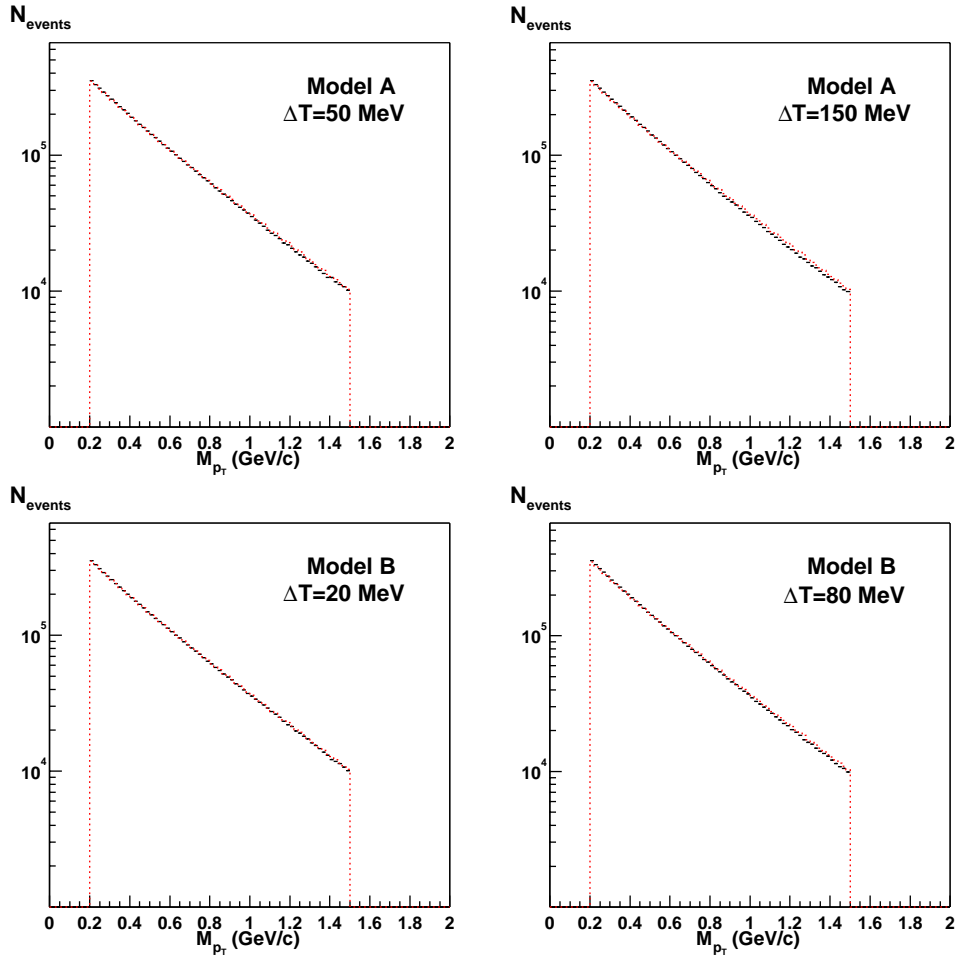


Figure 1: PHENIX inclusive  $p_T$  **Not**  $M_{p_T}$  distribution for 5% most central collisions(black). **Red** are 2-component models to be discussed later. Note the limited range of  $p_T$  used.

## Event by Event Distribution

### It's a not a Gaussian, it's a Gamma Distribution

$$M_{p_T} = \overline{p_T(n)} = \frac{1}{n} \sum_{i=1}^n p_{T_i} = \frac{1}{n} E_{Tc}$$

$$E_{Tc} = \sum_{i=1}^n p_{T_i} \quad E_T = \sum_{i=1}^n e_{T_i}$$

### Analytical formula for statistically independent emission

For statistical independent emission an analytical formula for the distribution in  $M_{p_T}$  can be obtained. It depends on the 4 semi-inclusive parameters  $\langle n \rangle$ ,  $1/k$ ,  $b$  and  $p$  which are derived from the quoted means and standard deviations of the semi-inclusive  $p_T$  and multiplicity distributions. The result is in excellent agreement with the NA49 Pb+Pb-central measurement.

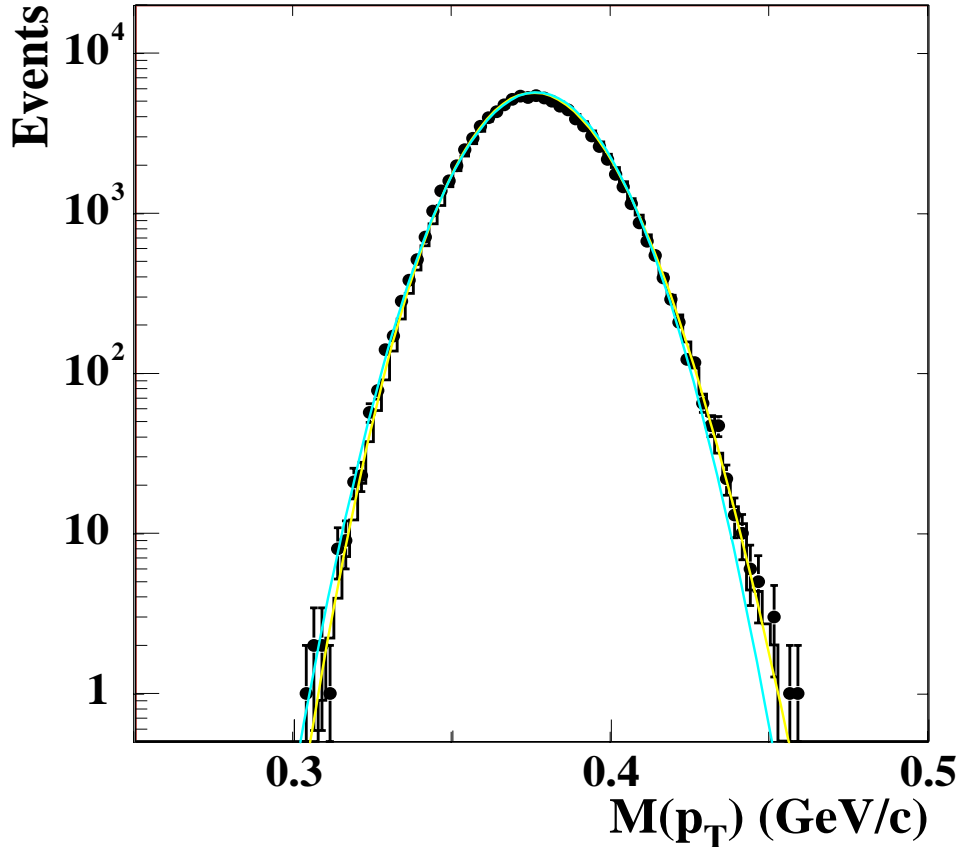


Figure 2: Gamma Distribution for  $M_{p_T}$  (yellow) compared Gaussian with same  $\mu$  and  $\sigma$  (light blue) for NA49 measurement (filled points) and mixed event distribution (histogram).

see M. J. Tannenbaum, Phys. Lett. B498, 29 (2001)

**It's a not a Gaussian, it's a Gamma Distribution**

## PHENIX $M_{pT}$ vs. Centrality

### Now Gamma Distribution Shape is Obvious

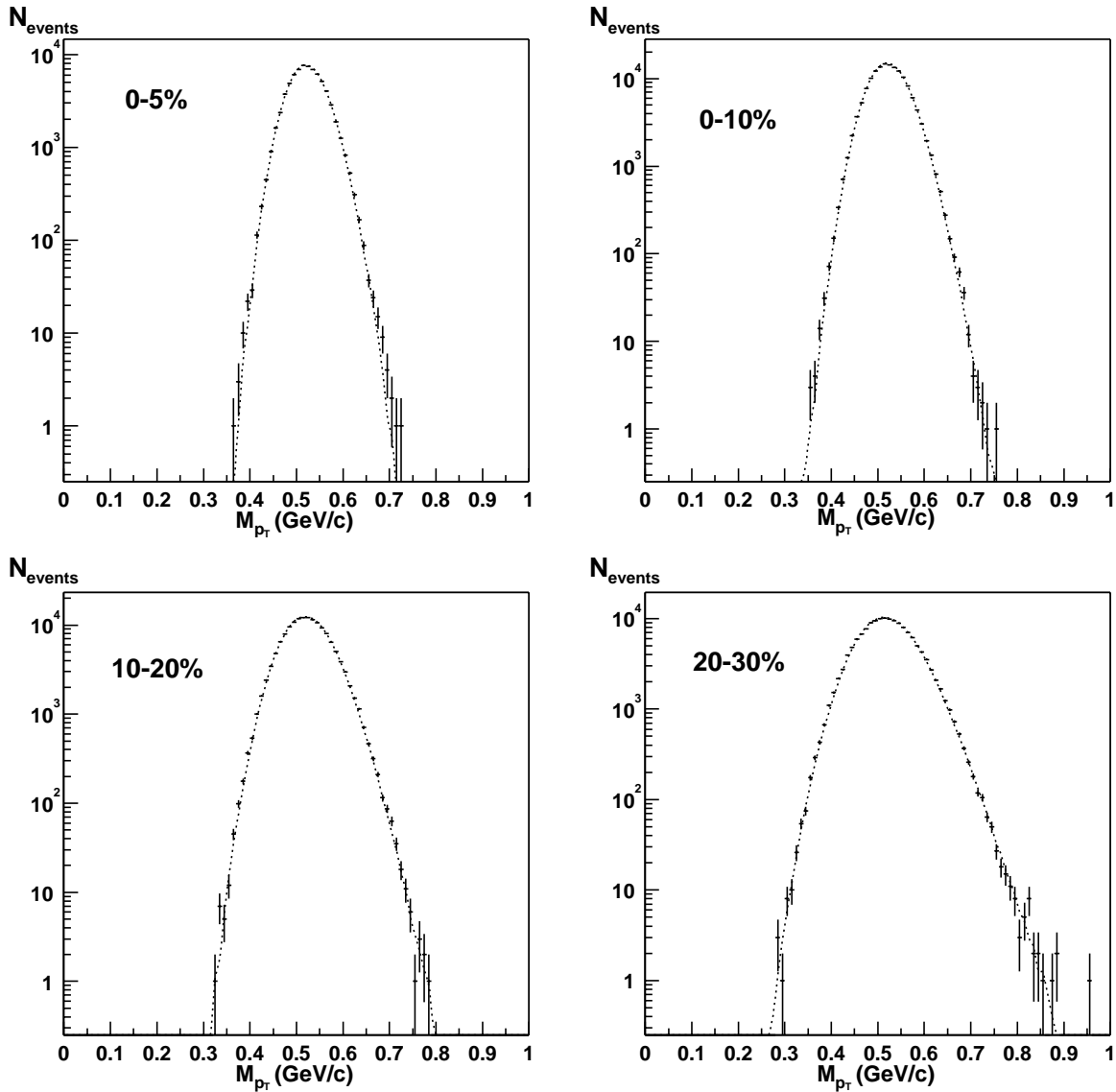


Figure 3: PHENIX  $M_{pT}$  data for Au+Au at  $\sqrt{s_{NN}} = 130$  GeV vs centrality. The dotted curves are mixed event distributions used as the random baseline.

- Use Mixed Events as a random baseline reference since analytical formula doesn't work in general.
- For Mixed Events **must use exactly the same  $n$  distribution as the data and match the inclusive  $\langle p_T \rangle$  to high precision.**
- Data indicate very small, if any, non random effect **How to Quantify?**

## Most Groups Use Moments (fortunately just $\mu$ and $\sigma$ )

$$\bar{x} = \frac{1}{n} \sum_{i=1}^n x_i$$

$$\langle \bar{x} \rangle = \langle x \rangle \equiv \mu$$

$$\sigma_x^2 = \langle x^2 \rangle - \langle x \rangle^2$$

$$\sigma_{\bar{x}}^2 = \langle \bar{x}^2 \rangle - \langle \bar{x} \rangle^2$$

### •For Statistically Independent Emission

$$\sigma_{\bar{x}}^2 = \left\langle \frac{\sigma_x^2}{n} \right\rangle$$

## Typical Measures of Sensitivity

$$\frac{\text{Measured} - \text{Random}}{\text{Random}}$$

$$\left( \frac{\sigma_{\bar{x}}^2}{\mu^2} - \frac{1}{n} \frac{\sigma_x^2}{\mu^2} \right) / \frac{1}{n} \frac{\sigma_x^2}{\mu^2}$$

$$\left( \frac{\sigma_{\bar{x}}}{\mu} - \frac{1}{\sqrt{n}} \frac{\sigma_x}{\mu} \right) / \frac{1}{\sqrt{n}} \frac{\sigma_x}{\mu} = F$$

### •For small effects these measures are equivalent

$$\frac{\Delta \sigma^2}{\sigma^2} = 2 \frac{\Delta \sigma}{\sigma} = 2F$$

•PHENIX uses Mixed Events as Random Baseline with exactly the same  $n$  distribution as the data and with the inclusive  $\langle p_T \rangle = \mu$  matched to high precision.

$$\left( \frac{\sigma_{\bar{x}}}{\mu} - \frac{\sigma_{\bar{x}-\text{mixed}}}{\mu} \right) / \frac{\sigma_{\bar{x}-\text{mixed}}}{\mu} = F$$