

# Comments on e-by-e Sensitivity

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## •GENERAL DEFINITIONS

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

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

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

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

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

$$\sigma_{\bar{x}}^2 = \frac{\sigma_x^2}{n} \text{ for stat. ind. emission}$$

## •Two Measures of Sensitivity—a Theorem

1)

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

2)

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

I prefer 2, but-a simple theorem

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

## Event by Event Distribution

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

### 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 for central Pb+Pb collisions for NA49. The result is in excellent agreement with the NA49 measurement.

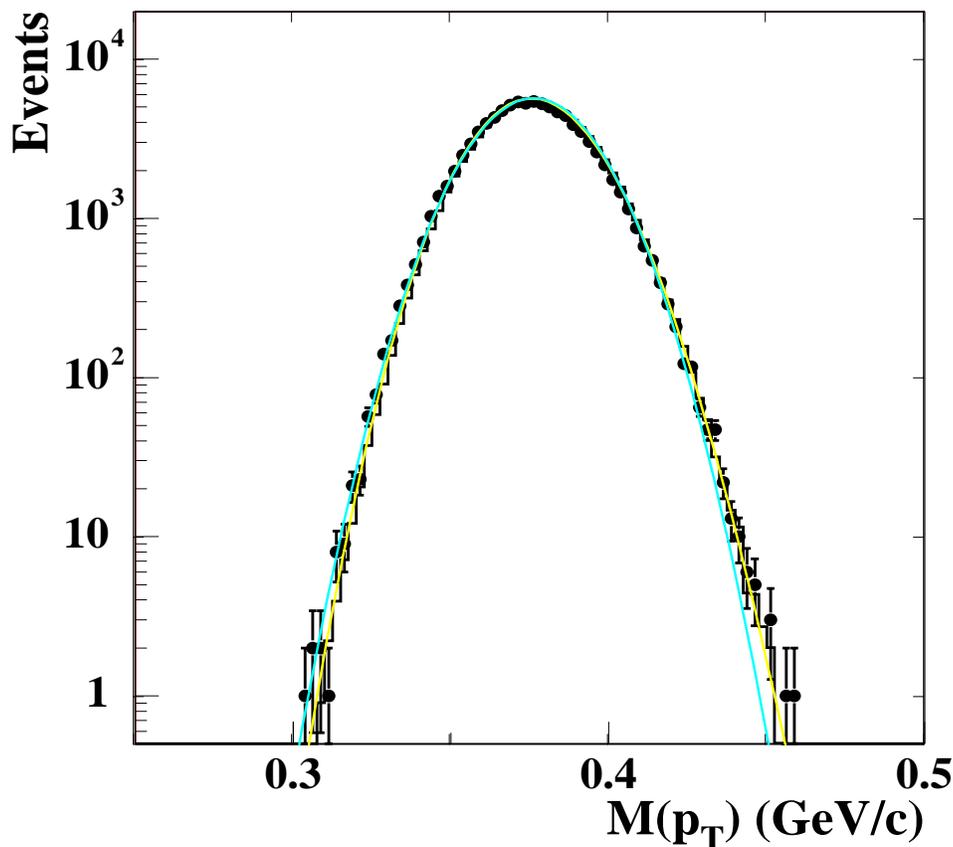


Figure 1: Full distribution in  $M_{p_T}$  (light line) compared to NA49 measurement (filled points) and mixed event distribution (histogram).

**Hint, it's a Gamma Distribution**  
see M. J. Tannenbaum, Phys. Lett. B498, 29 (2001)

## • The $\Phi$ Issue

$$\Phi = F \times \sigma_x$$

In words,  $\Phi$  multiplies the fractional difference between the e-by-e  $\sigma$  from random by the INCLUSIVE  $\sigma$ .

Suppose the e-by-e  $\sigma$  is  $\sim 1\%$  of the inclusive  $\sigma$  and you find a 1% difference in the e-by-e  $\sigma$  from random, why would you claim that this is equivalent to  $\Phi = 1\%$  of the inclusive  $\sigma$ , which in this example equals the ENTIRE e-by-e  $\sigma$  ???

• **What the e-by-e average tells you that you can't learn from the INCLUSIVE average**

♡ e-by-e separates two or several classes of events with different  $\mu$  and  $\sigma$

A) Same  $\mu$  different  $\sigma$

B) Different  $\mu$  same  $\sigma$

B') **Continuously varying  $T$**  From a recent paper [ R. Korus, *et al.* Phys. Rev. C64, 054908 (2001) ]  $T$  varies with a mean  $\langle T \rangle$  and standard deviation  $\sigma_T$ .

Note that in all cases the mean and standard deviation of the compound distribution must equal the measured inclusive values

♡ **For B and B' the e-by-e effect is seen at the level of the moments**

$$\frac{\sigma_{\bar{x}_T}^2}{\mu_T^2} = \frac{1}{n} \frac{\sigma_{x_T}^2}{\mu_T^2} + \left(1 - \frac{1}{n}\right) \frac{\sigma_T^2}{\langle T \rangle^2} \quad . \quad (1)$$

If you divide by random, you get an additional factor of  $n$ :

$$F = \frac{p}{2} (\langle n \rangle - 1) \frac{\sigma_T^2}{\langle T \rangle^2} \quad , \quad (2)$$

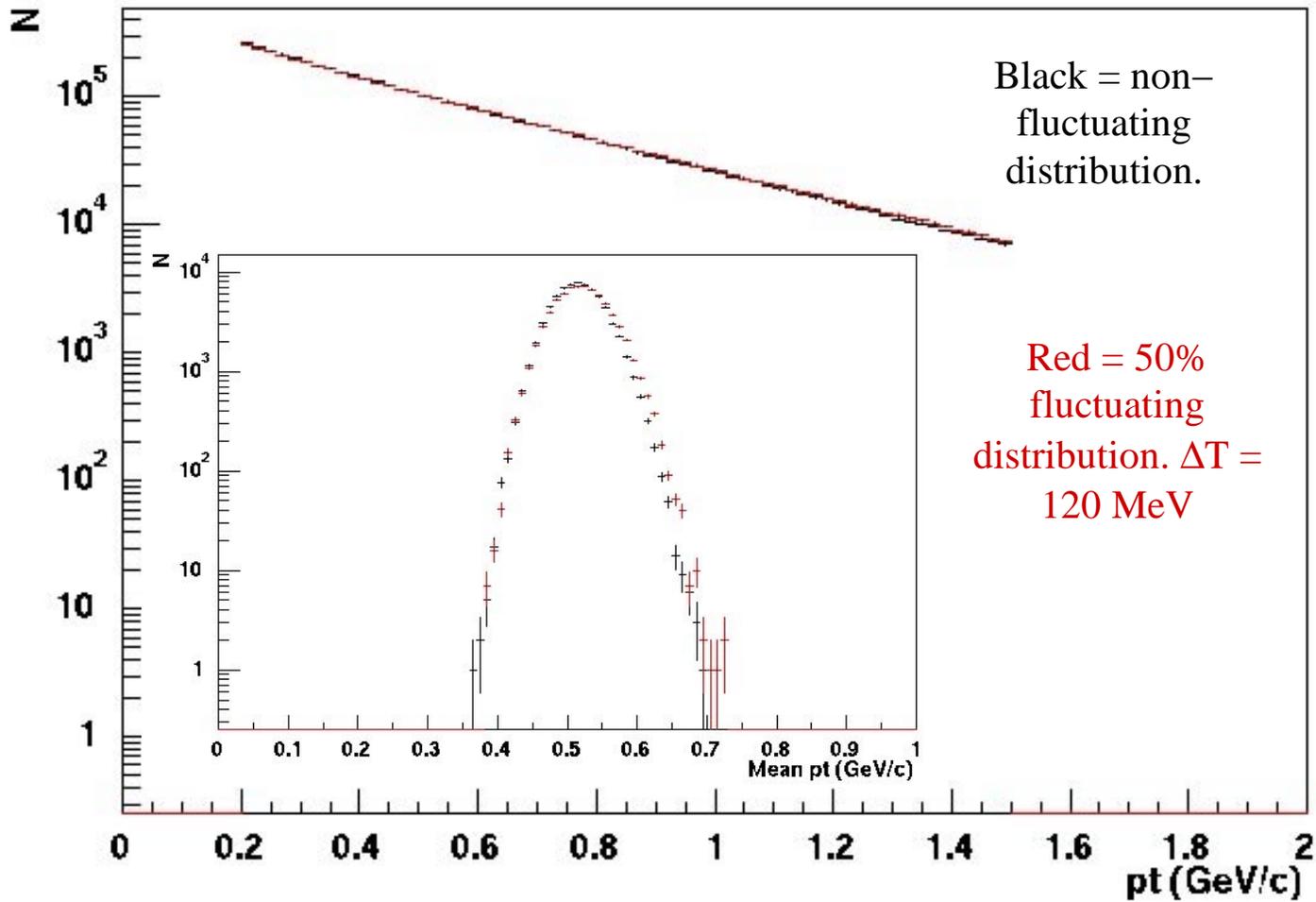
where  $p$  is the parameter of the inclusive  $\Gamma$  distribution.

♡ **For A the e-by-e effect cannot be seen at the level of the moments** The standard deviation  $\sigma_{\bar{x}_c}$ , of the compound distribution scales as  $1/n$ , exactly like the simple distribution, so that the e-by-e effect can not be seen from the variance. However the detailed shape of the e-by-e distributions are different for the simple and compound cases.

$$\frac{\sigma_{\bar{x}_c}^2}{\mu^2} = \frac{1}{n} \frac{\sigma_{x_c}^2}{\mu^2}$$

A) Distribution for same  $\mu$  different  $\sigma$

“Same mean, different variance” model demo



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