# FastSim dE/dx Resolution

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## BaBar

- <u>BaBar NIM Paper:</u> 7.5% dE/dx resolution
- <u>Sasha Telnov's BAD 1500:</u> It's complicated, but around 7.5%
- Resolution defined as follows with measured and expected truncated means:

$$\frac{\left\langle \frac{dE}{dx} \right\rangle_{meas} - \left\langle \frac{dE}{dx} \right\rangle_{exp}}{\left\langle \frac{dE}{dx} \right\rangle_{exp}}$$

# SuperB

- We expect slightly worse dE/dx performance due to different geometry (~8%)
- We are still working toward a clustercounting technique

# SuperB DCH in FastSim

dE/dx measurements for each cell hit by a track are drawn from a Gaussian distribution with mean given by the Bethe formula and standard deviation given by:

A fraction of the individual measurements are used to form a truncated mean, which is then fed to the normal reconstruction code.

$$\mu_0 = \left[\frac{dE}{dx}\right]$$

Bethe formula result normalized to material density, per unit length

$$\sigma_0 = p_1 \left(\frac{\mu_0}{C}\right)^{p_2} L^{p_3}$$

Track segment length in cm (cell\_thickness/sin(theta) at large momentum

Bethe formula result for minimum ionizing particles in appropriate units (1.622e-3)

trunc\_frac="0.7" p1="0.00154" ▶ p2="1" p3="-0.34"

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# dE/dx Resolution in FastSim

For a track crossing M layers, with a T truncated mean fraction, you get k = MT samples. Model a straight uniform track, then the truncated mean will also be a Gaussian variable with the same mean, and reduced standard deviation

The resolution we use is the RMS/mean of the truncated mean measurements:

Given that the second parameter is identically 1, this reduces to a constant function of the track segment length:

Putting in the numerical factors (13.75mm cells):

 $\mu = \mu_0$ 

$$\sigma = \frac{\sigma_0}{\sqrt{k}}$$
$$\Gamma = p_1 \frac{\mu_0^{p_2 - 1}}{\sqrt{k}C^{p_2}} L^{p_3}$$

$$\Gamma = \frac{p_1 L^{p_3}}{\sqrt{kC}}$$

 $\Gamma = 0.161$ 

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### Confirmation in Simulation



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# Cluster Counting

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It would be difficult to implement microscopic simulation of cluster counting in FastSim, so instead we would use "parametrized" cluster counting. A certain amount of dE/dx could correspond to a cluster

The uncertainty in counting follows simple Poisson statistics, reflected in a choice of p2 = 1/2

Cluster-counting efficiency is reflected in the choice of p1. Expected efficiency is around 60%



From Garfield in He:Isobutane 90:10

### Deriving Parameters for Cluster Counting

$\mu_c = \begin{bmatrix} dE \\ dx \end{bmatrix}$ (1) the mean ionization per unit length normalized by gas density, given by the Bethe formula.	$C$ $p_1$ p	ionization per unit length of a minimally-ionizing particle parameter with dimension
$\mu_{c_{cell}}=\mu_{c}L$ (2) mean ionization in a cell of size L	$p_2$	dimensionless
$\sigma = m \left( \frac{\mu_c}{p_2} \right)^{p_2} I^{p_3}$	$p_3$	parameters
$O_c = P_1 \left( \frac{1}{C} \right)  L^2$ (3) standard deviation of the ion	nizatio	n per unit length
$\sigma_{c_{cell}} = \sigma_c L  \  \   \   \   \   \   \   \ $	lpha	average ionization corresponding to one cluster
$\mu_{N_{cell}} = lpha \epsilon \mu_{c_{cell}}$ (5) mean number of clusters counted in a cell, proportional to the total ionization in	$\epsilon$	cluster-counting efficiency
$\sigma_{N_{cell}} = \sqrt{\mu_{N_{cell}}}  \  \  \  \  \  \  \  \  \  \  \  \  \$	$N_{min}$	number of clusters created per cm by a minimally-ionizing particle
$\sigma_{N_{cell}} = \alpha \epsilon \sigma_{c_{cell}} \qquad (7) \text{ also the standard deviation of} \\ \text{the number of clusters counted in} \\ \text{a cell, following from equation 5} \end{cases}$	Eq substit	uate (6) and (7), ute (2),(3), and (5)
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#### Deriving Parameters for Cluster Counting (2)

$$\begin{split} \sqrt{\alpha \epsilon \mu_c L} &= \alpha \epsilon L p_1 \left(\frac{\mu_c}{C}\right)^{p_2} L^{p_3} \\ \end{split}$$
This holds for all cell sizes L and mean ionization densities, so we immediately get:
$$p_2 &= \frac{1}{2} \\ p_3 &= -\frac{1}{2} \\ \end{aligned}$$
The rest of the parameters are globbed up into p1:
$$p_1 &= \frac{C^{p_2}}{\sqrt{\alpha \epsilon}} = \sqrt{\frac{C}{\alpha \epsilon}} \\ N_{min} \text{ is the minimum number of clusters created, so} \\ N_{min} \alpha &= \frac{N_{min}}{C} \\ \end{split}$$

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#### Deriving Parameters for Cluster Counting (3)

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Numerical values for some constants (in appropriate units):

 $C = 1.622 \times 10^{-3}$   $\epsilon \approx 0.6$   $N_{min} \approx 13$  $p_1 = 5.808 \times 10^{-4}$ 

## Proposal

- For regular productions, change the first dE/dx parameter to 0.000765 (from 0.00154) to get ~8% dE/dx resolution (direct scaling of result to 8%)
- For cluster-counting simulation, change the first parameter to 0.0005808 (from 0.00154), change second parameter to 1/2 (from 1) and the third parameter to -1/2 (from -0.34) to reflect the cluster counting parametrization presented above

# Preliminary Results



# Preliminary Results



Tuesday, 20 March, 12