

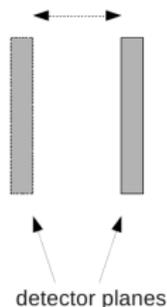
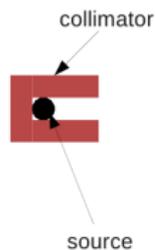
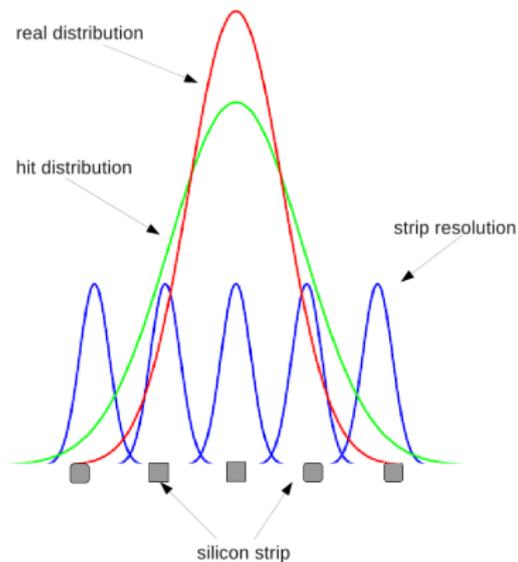
# Update on hardware activities

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# spatial resolution

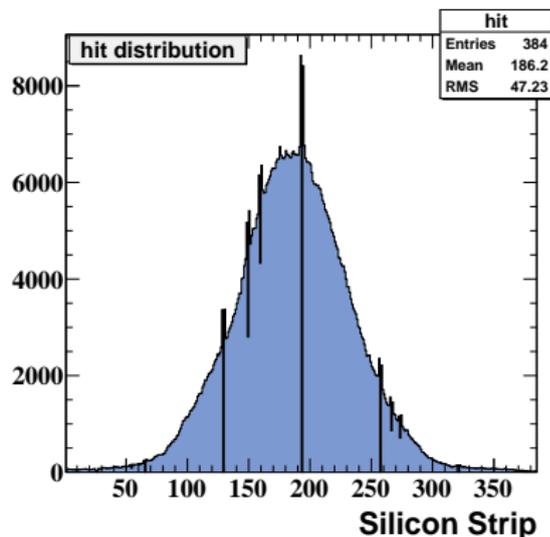
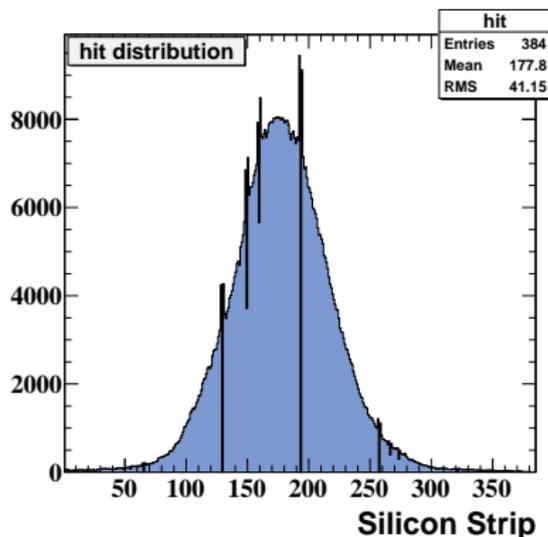


- ▶ deconvolute hit distribution for different displacements of detector and source  $\Rightarrow$  spatial resolution

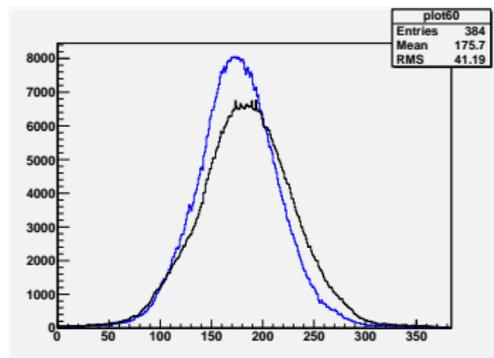
Distance source to sensor:

$d = 60$  mm

$d = 85$  mm



# simple deconvolution



- ▶ smoothing of both distributions
- ▶ Unfold detector resolution
$$\vec{x}_{true} = R^{-1} \vec{x}_{meas}$$
- ▶ Rescale and center one distribution
$$\vec{y} = S \vec{x}_{true}$$
- ▶ fold distribution with detector resolution again  $\vec{x}_{fin} = R \vec{y}$
- ▶ find minimum of  $\chi^2(\sigma_D, w, \mu)$

- ▶ resolution matrix: Use random number generator to dice initial channel position and add gaussian for smearing

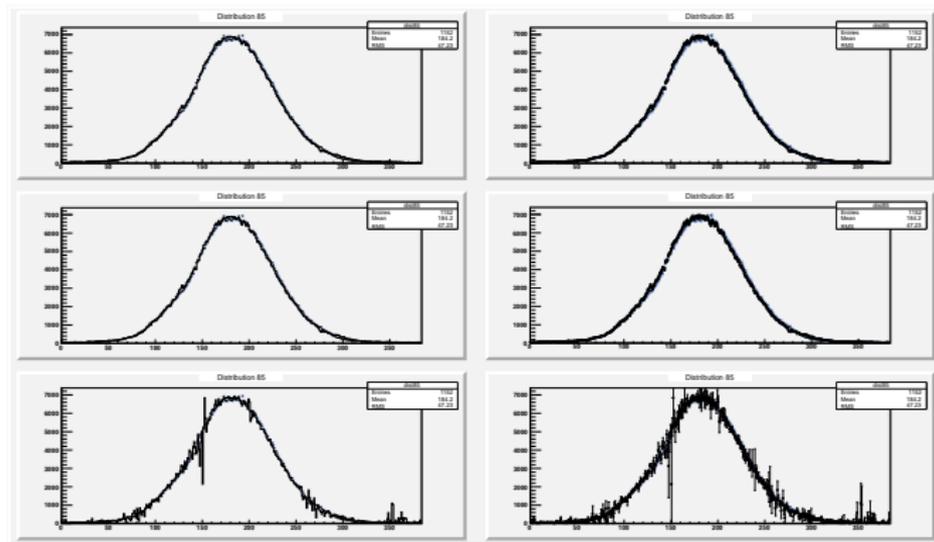
$$f_{ini} = \begin{pmatrix} 1000 \\ 1000 \\ \vdots \\ 1000 \end{pmatrix} \quad R = \begin{pmatrix} 680 & 150 & 5 & \dots & 0 & 0 & 0 \\ 150 & 680 & 150 & \dots & 0 & 0 & 0 \\ \vdots & \vdots & \vdots & \ddots & \vdots & \vdots & \vdots \\ 0 & 0 & 0 & \dots & 5 & 150 & 680 \end{pmatrix}$$

- ▶ scaling matrix: Assume that particle trajectory is a straight line

$$x_2 = \underbrace{\frac{d_2}{d_1}}_{=w} \cdot x_1$$

- ▶ other effects like multiple scattering are not included

# simple deconvolution



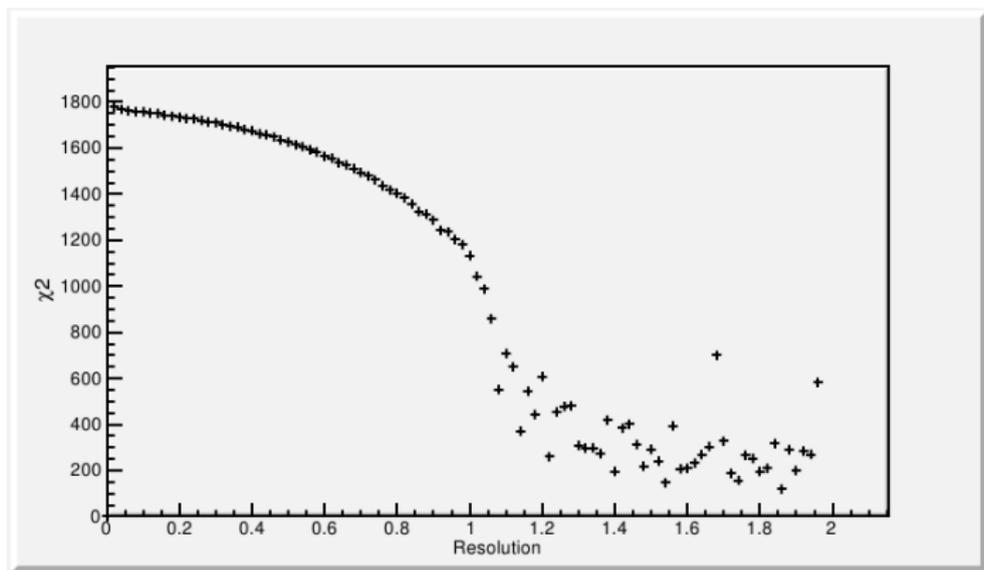
$$\sigma_D = 0.3$$

$$\sigma_D = 0.8$$

$$\sigma_D = 1.3$$

- ▶ for detector resolutions  $\sigma_D > 1$  distributions start to show oscillations  
⇒ possible minimum in  $\chi^2$  is smeared out

# simple deconvolution



- ▶ due to oscillations regularization is required

# deconvolution using SVD

- ▶ incorporate general assumptions about size or smoothness of unfolded distribution (second derivative)

$$x'' \approx x_{i-1} - 2x_i + x_{i+1}$$

$$C = \begin{pmatrix} 1 & -2 & 1 & 0 & \cdots & 0 & 0 & 0 \\ 0 & 1 & -2 & 1 & \cdots & 0 & 0 & 0 \\ \vdots & \vdots & \vdots & \vdots & \ddots & \vdots & \vdots & \vdots \\ 0 & 0 & 0 & 0 & \cdots & 1 & -2 & 1 \end{pmatrix}$$

- ▶ calculate  $AC^{-1}$  and use singular value decomposition

$$AC^{-1} = U\Sigma V^T$$

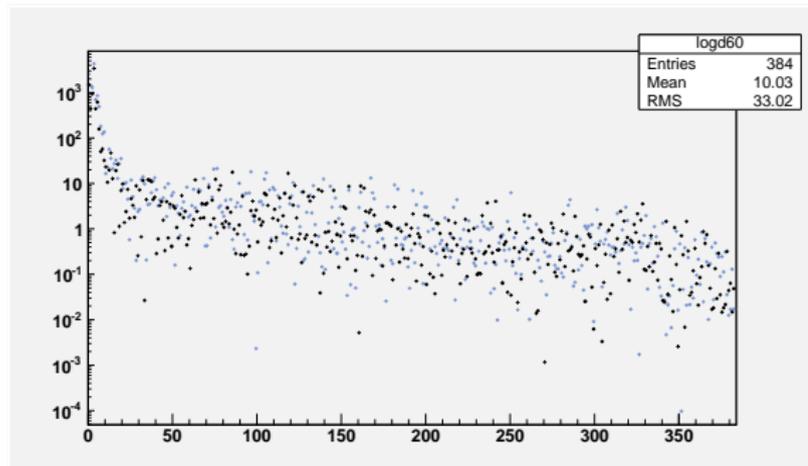
U, V contain eigenvectors of  $AC^{-1}$

$\Sigma$  is a diagonal matrix containing the singular values



# deconvolution using SVD

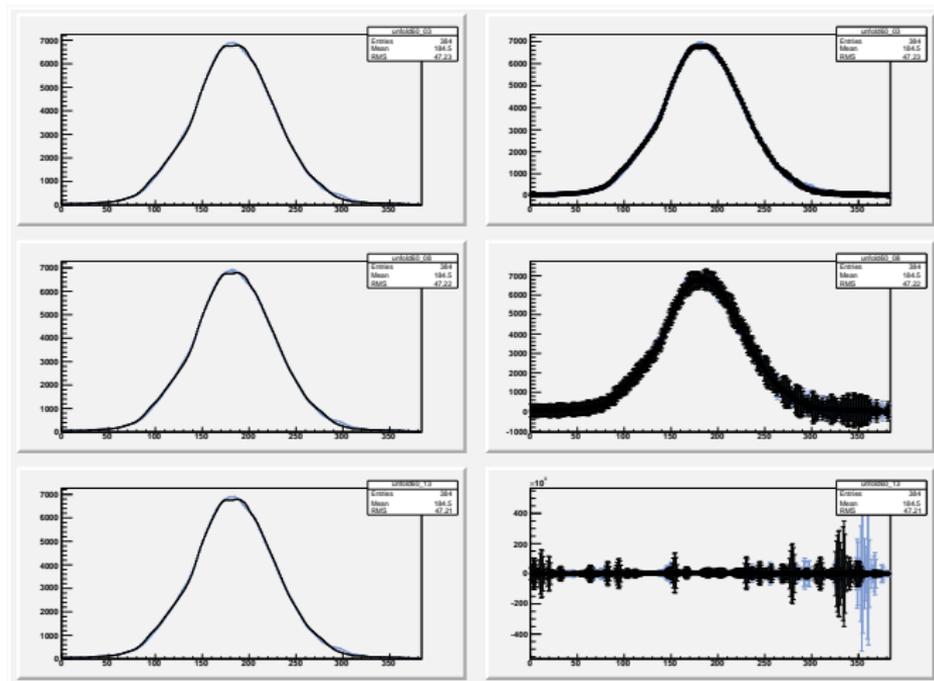
- In order to obtain the effective rank  $k$  of the response matrix and the regularisation parameter  $\alpha = \sigma_k^2$ , plot  $\log(|d|)$  with  $d = U^T \vec{x}$



- unfolded solution given by

$$w^{(\alpha)} = C^{-1} V \cdot \frac{d_i \sigma_i}{\sigma_i^2 + \alpha}$$

# deconvolution using SVD



$$\sigma_D = 0.3$$

$$\sigma_D = 0.8$$

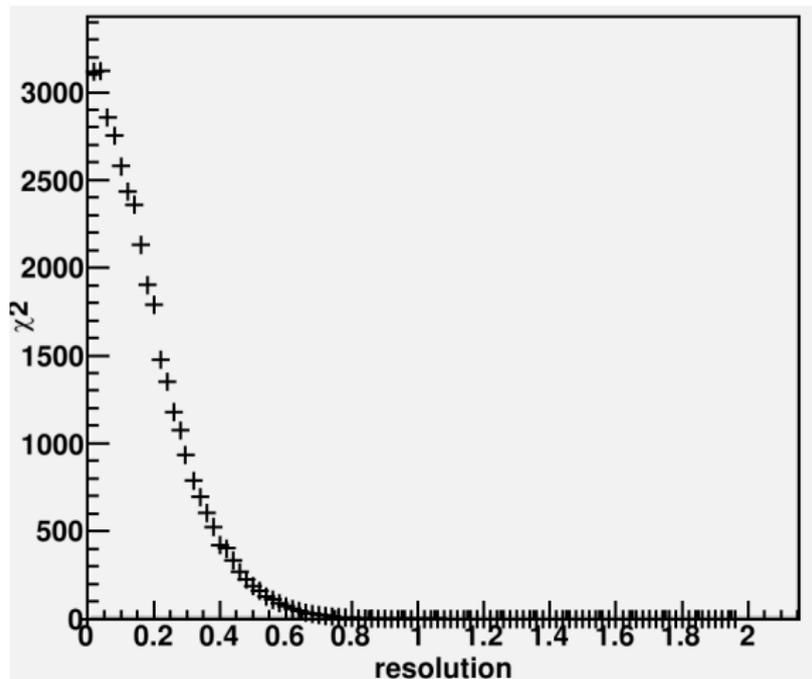
$$\sigma_D = 1.3$$

- ▶ oscillations disappear, but very high and oscillating errors at high resolutions



# deconvolution using SVD

- ▶ after regularisation still no minimum in  $\chi^2$



## Results so far:

- ▶ neither simple deconvolution nor deconvolution using SVD give a value for spatial resolution
- ▶ using *RooUnfold* without success

## Ideas:

- ▶ test of algorithms with simple distribution (gaussian)
- ▶ include multiple scattering