

Alignment

The alignment problem

- Estimate 3 shifts + 3 rotation angles per detector module

- Vector \mathbf{a}_k of alignment

parameters for N modules, $N \sim 10^4$

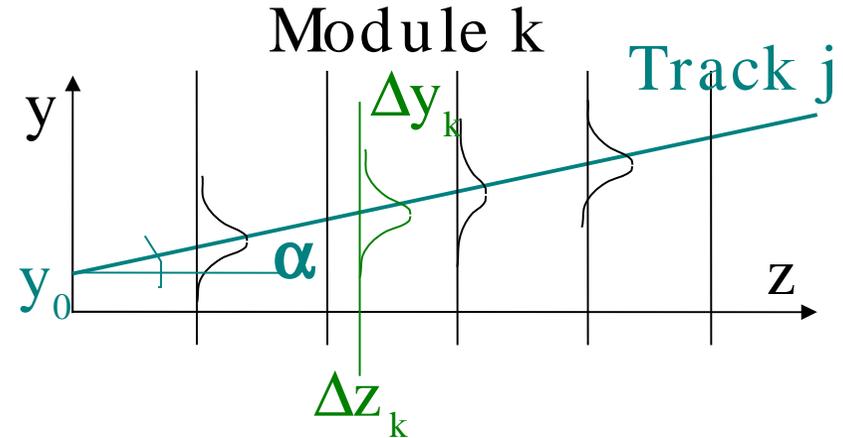
- Precision to achieve:

- \ll detector resolution
 - $\ll\ll$ assembly precision

→ **Track-based alignment**

- Principle of track-based alignment

- Fit simultaneously parameters of M tracks, and $6 \times N$ alignment parameters (identical for all tracks)



- $M = 100,000$; 10 points per track = 10^6 measurements

- $5M + 6N \sim 56 \cdot 10^4$ unknowns

- Measurement equation for track j in module k:

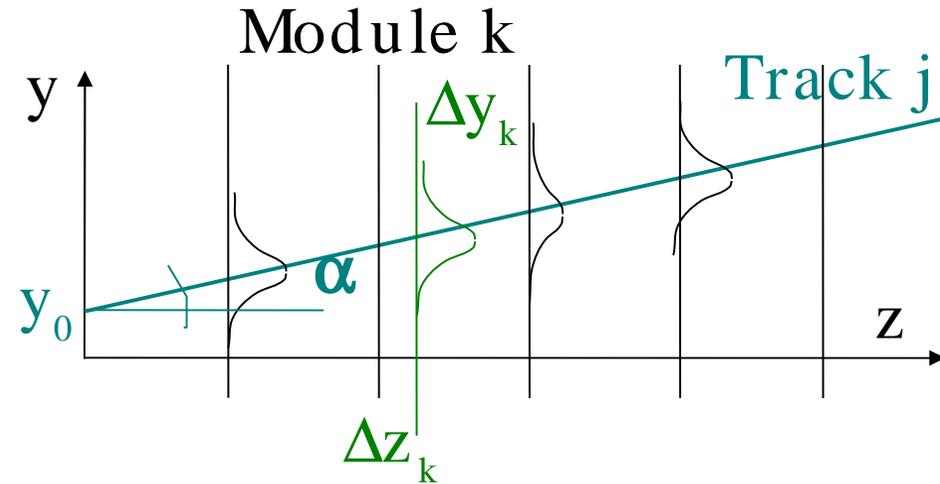
$$\mathbf{m}_{j,k} = h_k(\mathbf{p}_j, \mathbf{a}_k)$$

$$\mathbf{m}_{j,k} = p_j^1 + p_j^2(z_k + \Delta z_k) + \Delta y_k$$

- Linearize, minimize $\sum_{j,k} \mathbf{r}^T \mathbf{R}^{-1} \mathbf{r}$
 - **Linear system of $56 \cdot 10^4$ eq.**

Systematic effects

- Unconstrained parameters
 - Global scale factors, rotations, shearings
 - In-situ survey of a few reference modules (Laser alignment systems etc.)
- Correlated displacements of modules due to mechanical structure
 - Added to sum of squared residuals with Lagrange multipliers



- Correlations bw. Alignment parameters of different modules, mediated by the tracks that cross both
 - Essentially 2 recent developments
 - Millepede algorithm [1,2]
 - KF- based method [3]

■ Millepede algorithm [2]

- Takes advantage of special structure of system of eq.
- First fit each track separately
- Then modify system of $6N$ equations accounting for track fit results (Schur complement matrix of $\sum_k C_k^{global}$)
 - NB: here k is track index
- A system of $6N$ equations remains to be solved
 - Advanced iterative methods needed

$$\begin{array}{c}
 \text{6N} \quad \text{6N} \quad \text{M blocks of 5x6N} \\
 \left(\begin{array}{c|c|c|c}
 \sum_k C_k^{global} & \dots & H_k^{global-local} & \dots \\
 \hline
 \vdots & \ddots & 0 & 0 \\
 \hline
 (H_k^{global-local})^T & 0 & C_k^{local} & 0 \\
 \hline
 \vdots & 0 & 0 & \ddots
 \end{array} \right) \times \\
 \left(\begin{array}{c}
 \Delta p^{global} \\
 \hline
 \vdots \\
 \hline
 \Delta q_k^{local} \\
 \hline
 \vdots
 \end{array} \right) = \left(\begin{array}{c}
 \sum_k b_k^{global} \\
 \hline
 \vdots \\
 \hline
 b_k^{local} \\
 \hline
 \vdots
 \end{array} \right) .
 \end{array}$$

- [1] V.Blobel, “Software alignment for tracking detectors”, Nucl. Instr. and Meth. A566 (2006) 5
- [2] V.Blobel, “Alignment algorithms”, 1st LHC Detector Alignment Workshop, CERN, Geneva, Switzerland, 4 - 6 Sep 2006, pp.5- 12 (CERN Yellow Report)
- [3] R.Fruhworth, E.Widl, “Track- based alignment using a Kalman filter technique”, 1st LHC Detector Alignment Workshop, CERN, Geneva, Switzerland, 4 - 6 Sep 2006, pp.13- 19 (CERN Yellow Report)
- [4] P.Avery, lectures on tracking (on line)

Vertex fitting

Set of N tracks
(+ beam constraint)

Vertex fitting

3D point and N momentum vectors, most compatible with input tracks (and beam)

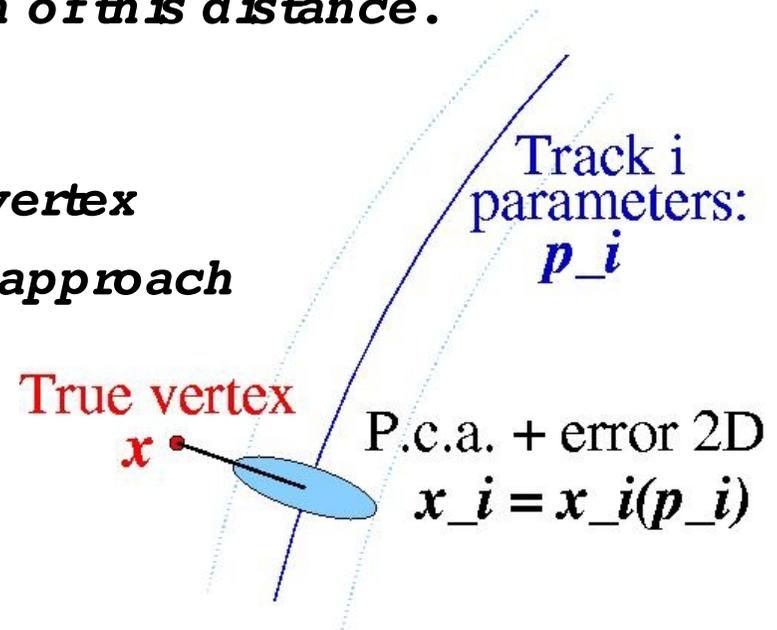
Minimization problem :

- Compatibility defined by a distance F between the vertex parameters x and the vertex "measurements", i.e. the track parameters p_i (more precisely, a function f of them).
- "Most compatible" implies minimization of this distance.

Example:

- $f(p_i) = x_i$ point of closest approach to vertex
- C_i covariance matrix of point of closest approach

- $F = \sum_{i=1}^N (x - x_i)^T C_i^{-1} (x - x_i)$
sum of squared reduced residuals



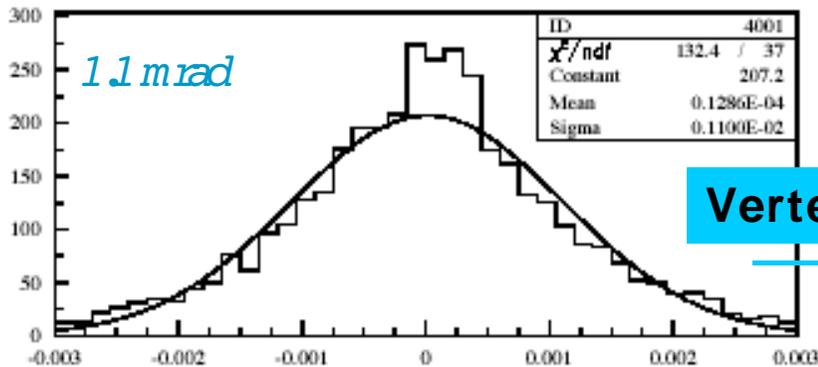
$F =$ sum of squared reduced residuals

Algorithm chosen: **Kalm an vertex fitting**

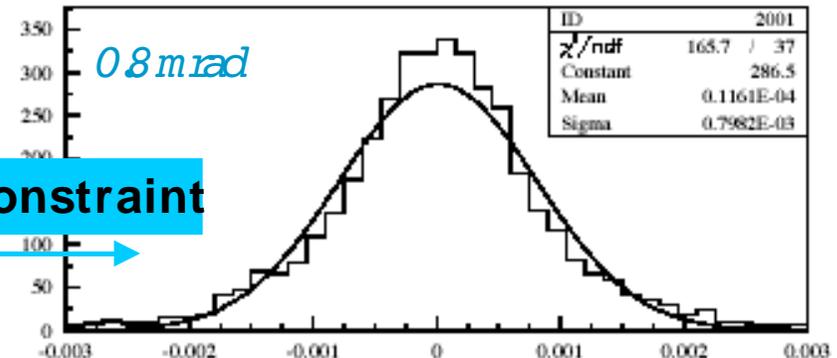
- vertex parameters $x \equiv$ state vector, in Kalm an parlance
- $f \equiv$ reciprocal of measurement function h , non-linear in presence of B-field
- problem is linearized:
 - h replaced by 1st order Taylor expansion
 - minimum of F is solution of a linear system of equations
 - explicit expression, no need for numerical minimization
- tracks added sequentially to the vertex (state vector update)
- problem is factorizable:
 - first fit vertex position, then constrain track momenta at vertex

Residuals of track parameters at vertex
 $B_s \rightarrow J/\psi \phi \rightarrow \mu^+\mu^- K^+K^-$
 Full CMS tracker simulation and reconstruction

ϕ Reconstructed

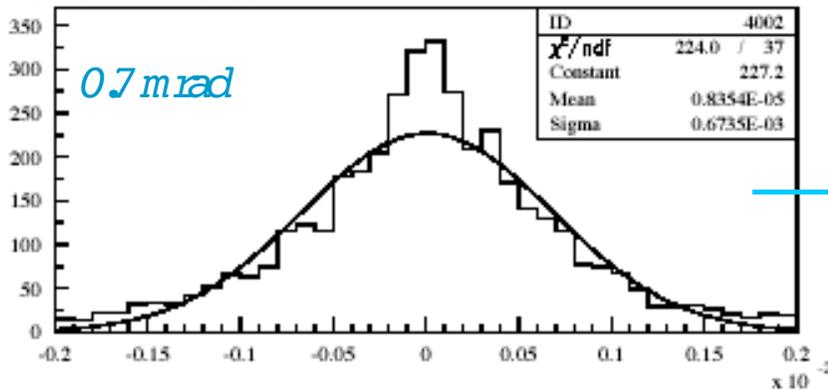


ϕ Refitted

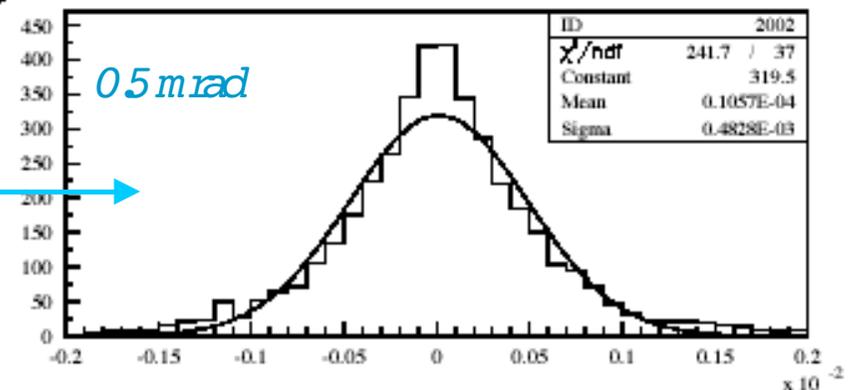


Vertex constraint

θ Reconstructed



θ Refitted



Expansion of h only valid in vicinity of a given point:

- needs *initial guess* of vertex position
- *iterations* if fitted vertex too far from *initial guess*
- ..but convergence is quick (1 or 2 iterations for least-squares vertex fits)
- track = helical constraint

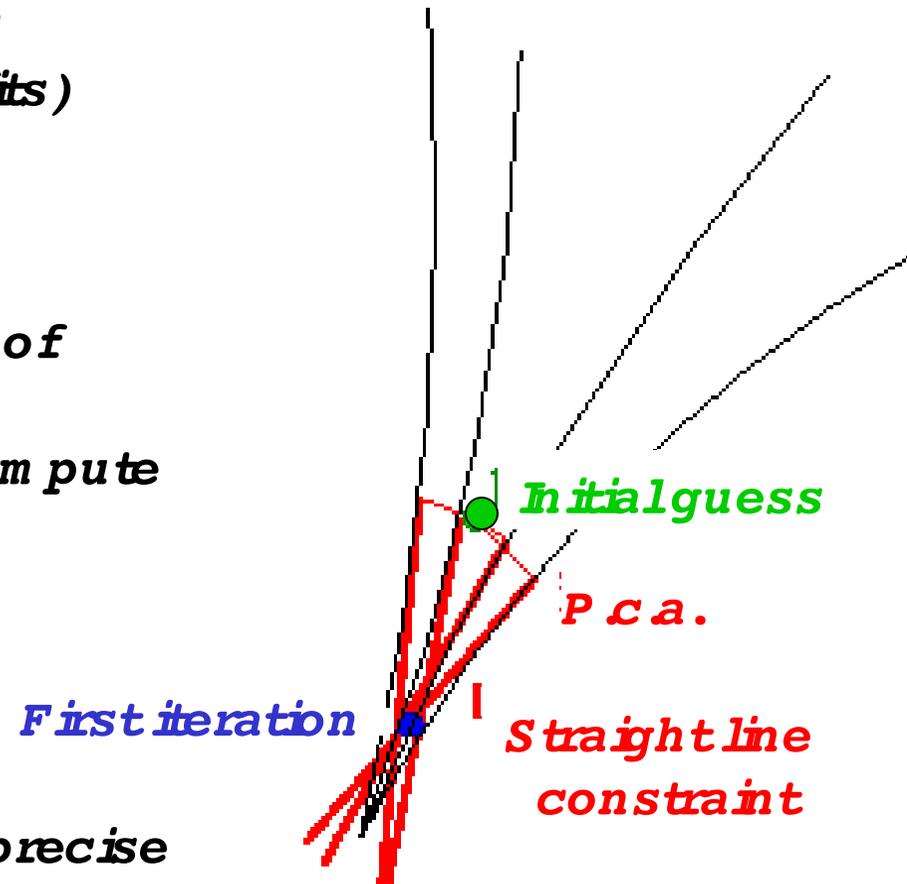
Initial guess:

- computed from average of points of closest approach of track pairs
- fast numerical algorithm to compute p.c.a.'s of two helices
- made robust by a Half-Sample Mode finder

Linearization:

- Perigee parametrization fast and precise

Illustration of vertex fit convergence when tracks are approximated by straight lines



Least squares estimators are efficient and unbiased if:

- track parameter reduced residuals are $N(0,1)$
 - Gaussian, unbiased, covariance matrix perfectly known
- all tracks originate from fitted vertex

In real experiments and in detailed simulations:

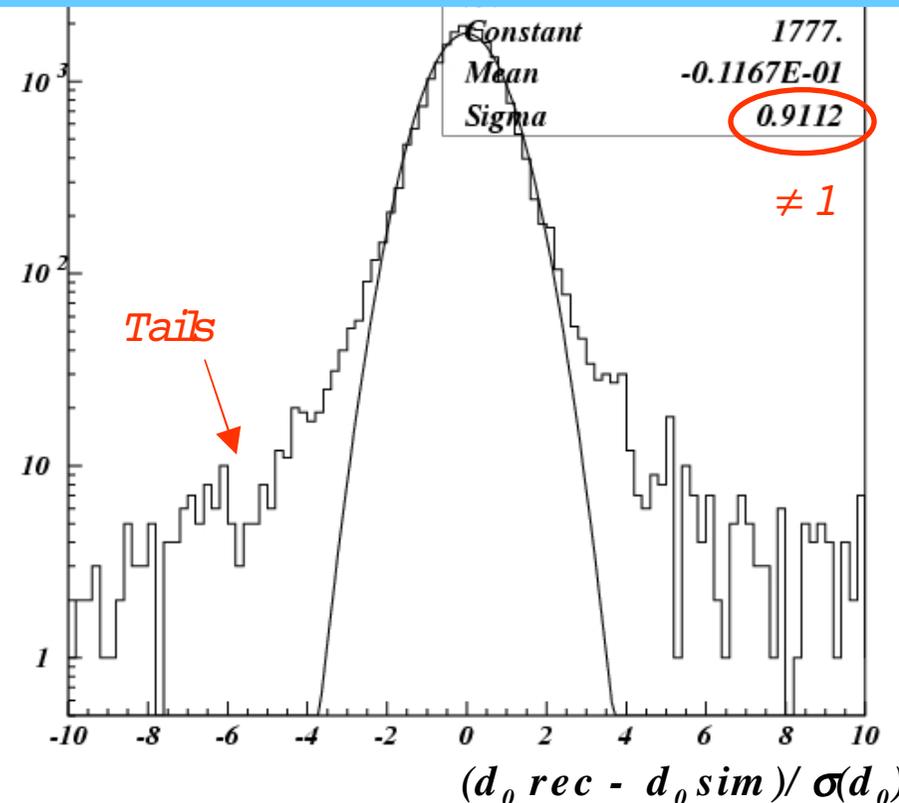
- detector resolutions not Gaussian
- distribution of dead material in perfectly known
- non-Gaussian multiple scattering
- track finding problems
- vertices not well separated ...
- ..difficult to get it all right!

→ **Robust estimators:** insensitive to m is-measured or m is-associated tracks

d_0 pull distribution

$H(m_H = 300 \text{ GeV}/c^2) \rightarrow ZZ \rightarrow e^+e^-\mu^+\mu^-$

CMS full simulation and reconstruction - ORCA_7_6_0



Discard (trimming) or down-weight (adaptive) distant tracks

Implemented as re-weighted least-squares fits:

$$F = \sum_{i=1}^N w_i r_i \quad r_i = (x - x_i)^T C_i^{-1} (x - x_i)$$

Weight $w_i \equiv$ probability of association of track i to vertex

Usually, time-consuming but fast approached algorithms exist (tried in CMS, ATLAS, H1, D0)

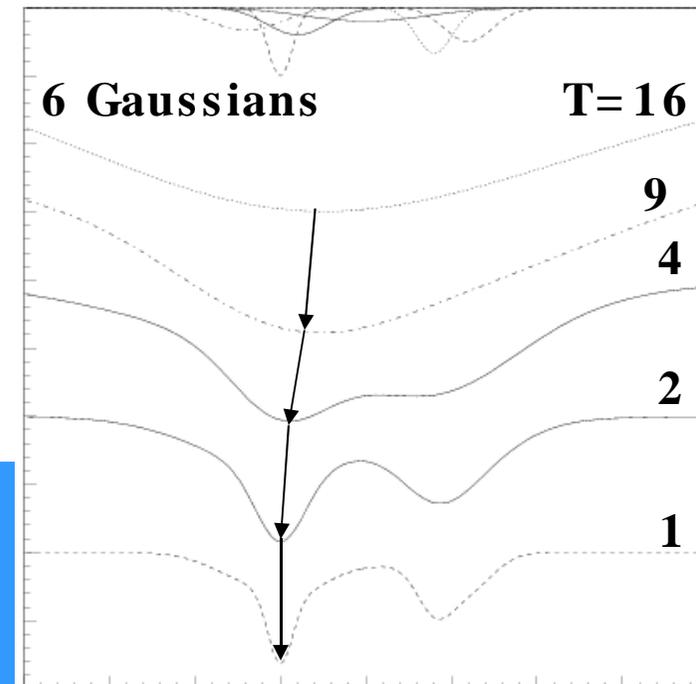
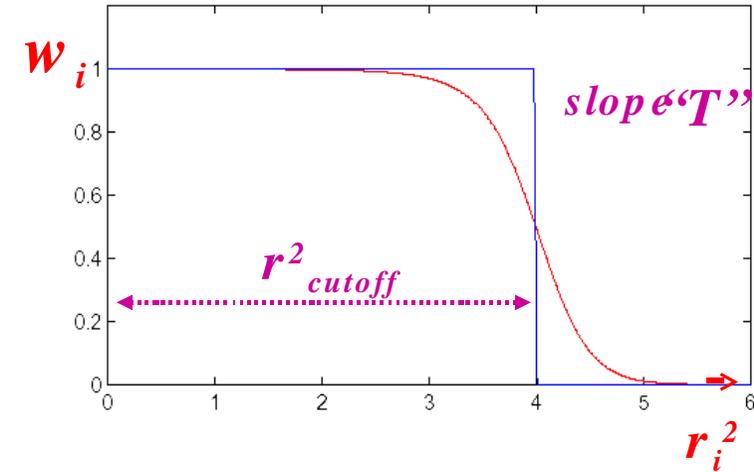
Trimming: $w_i = 0$ or 1

- **Hard assignment**
- **Fixed trimming fraction: k out of N tracks are discarded**
 - **Fast Least-Trimmed-Squares**
 - *P. Rousseeuw et al., "Computing LTS regression for large data sets", Technical report, University of Antwerp, 1999.*
- **Fixed compatibility: tracks with probability of compatibility to the vertex below some cut discarded**

Robust estimators (2)

Adaptive: $w_i = \{1 + \exp[(r_i^2 - r_{cutoff}^2) / 2T]\}^{-1}$
 M.Ohlsson et al., *Comput. Phys. Commun.* 71 (1992) 77
 R.Frühwirth and A.Strandlie, *Comput. Phys. Commun.* 120 (1999) 197

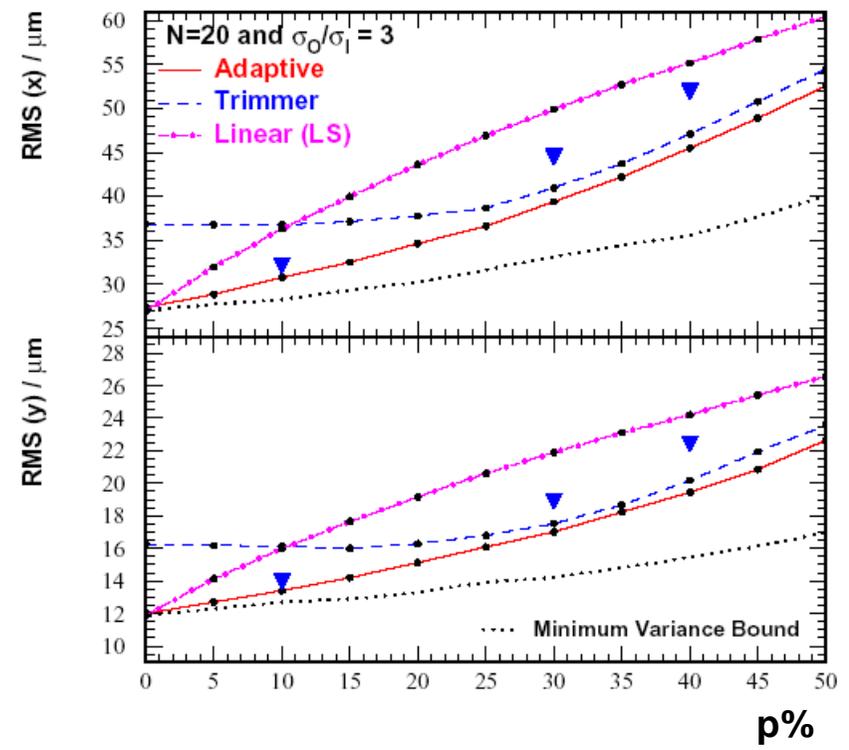
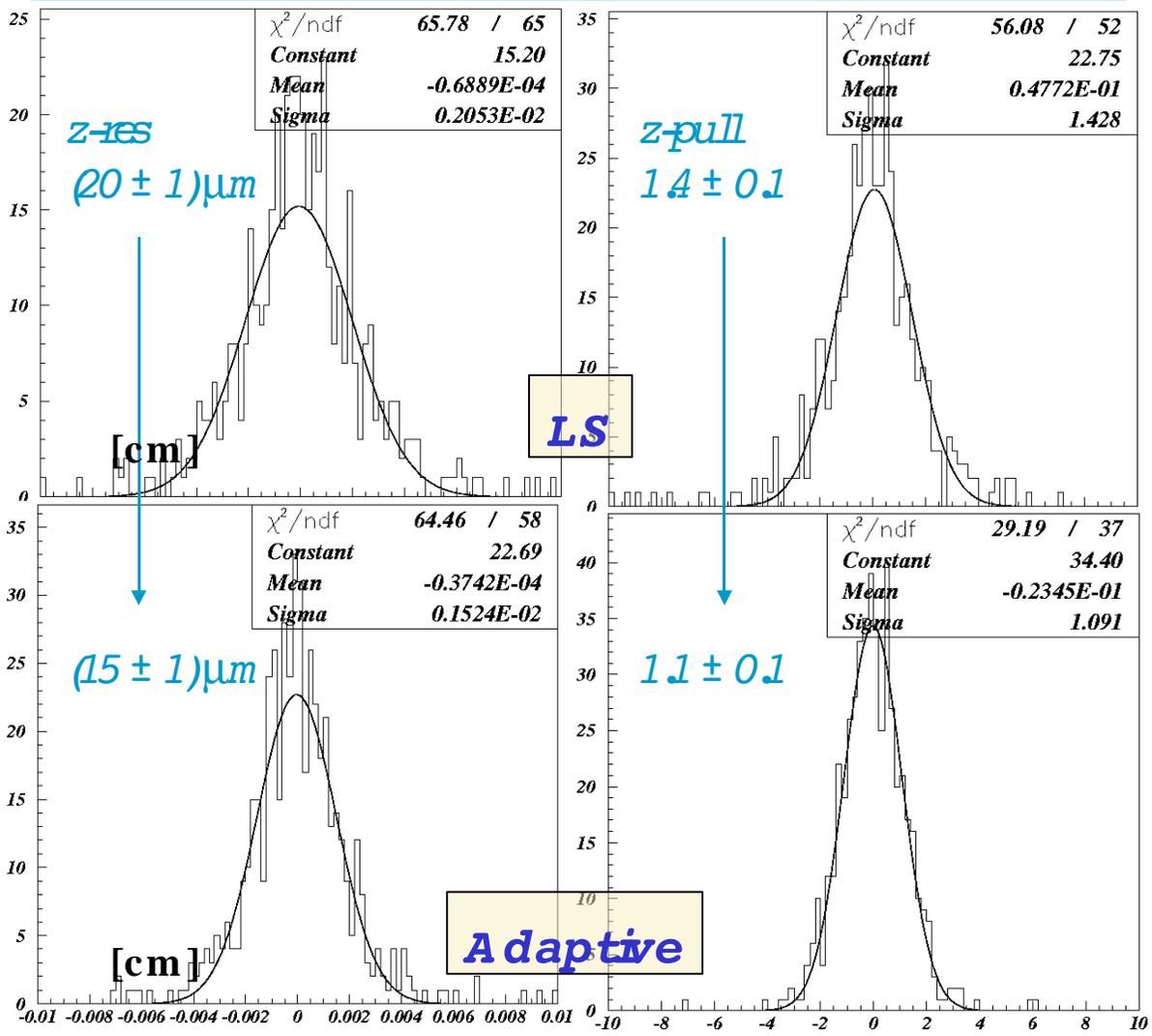
- **Soft assignment: association probabilities are fractional**
 - algorithm converges to optimal weight for each track
- **Annealing:**
 - avoids local minima
 - T starts high
 - decreases at each iteration
 - according to well-chosen schedule



**Illustration:
minimization of sum
of 6 Gaussians with
annealing**

Primary vertex z-residuals and pulls
 $H (m_H = 300 \text{ GeV}/c^2) \rightarrow ZZ \rightarrow e^+e^-\mu^+\mu^-$
 Only primary tracks used - ORCA_7_6_0

High multiplicity vertex (20 tracks, p% with covariance underestimated by factor 3)
 x- and y-residuals - VertexGun



- [1] R.Fruhworth, Nucl. Instr. and Meth. A262 (1987) 444
- [2] P.Billoir, S.Qian, Nucl. Instr. and Meth. A311 (1992) 139
- [3] P.Avery, lectures on tracking (on line)
- [4] J. D'Hondt, R. Fruhwirth, P. Vanlaer and W. Waltenberger, "Sensitivity of Robust Vertex Fitting Algorithms", IEEE Transactions on Nuclear Science Vol.51 (2004), 2037
- [5] R. Fruhwirth, W. Waltenberger and P. Vanlaer, "Adaptive Vertex Fitting", CMS Note 2007/008

Conclusions

- Detector design studies needed in order to meet physics requirements
 - Cost- driven optimization, compromises...
 - Detailed detector simulations needed
- Track fitting; vertex fitting; alignment
 - Well- defined statistical problems
- Lot of recent developments and creativity in:
 - Handling of combinatorics
 - Handling of ambiguities, robustness
 - Application of known mathematical algorithms to high- energy physics
 - Sometimes in extreme cases like alignment of 10,000 modules