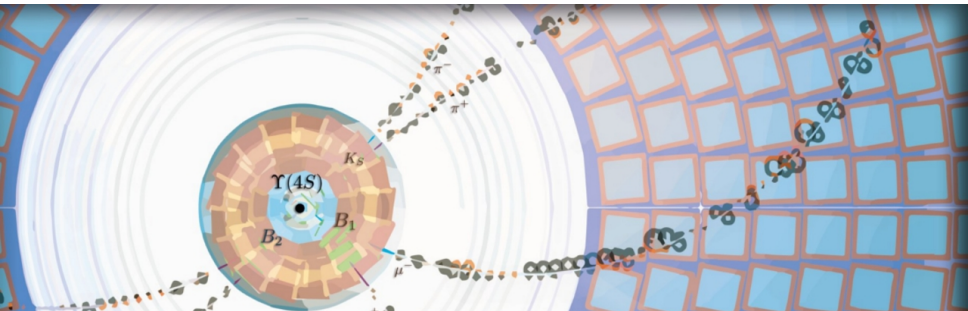


# Tales of Riemann.

## F2F tracking meeting - Pisa



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Deutsches Elektronensynchrotron

29th September 2014



- > Two dimensional fitting
- > Three dimensional fitting
- > Planes

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## Distance measure - fits to points

$$\epsilon = n_0 + n_1 \cdot x + n_2 \cdot y + n_3 \cdot r^2$$

(Frühwirth)

## Distance measure - fits to drift circles of radius $l$

$$\epsilon = n_0 + n_1 \cdot x + n_2 \cdot y + n_3 \cdot r^2 - a \cdot l$$

This requires some prior knowledge of the right left passage hypotheses  $a$  as seen by the particle.

## CDC measurements are drift circles

This difference is not neglectable especially for rather short segments. Only using the drift length and the correct right left passage hypotheses yields accurate results.

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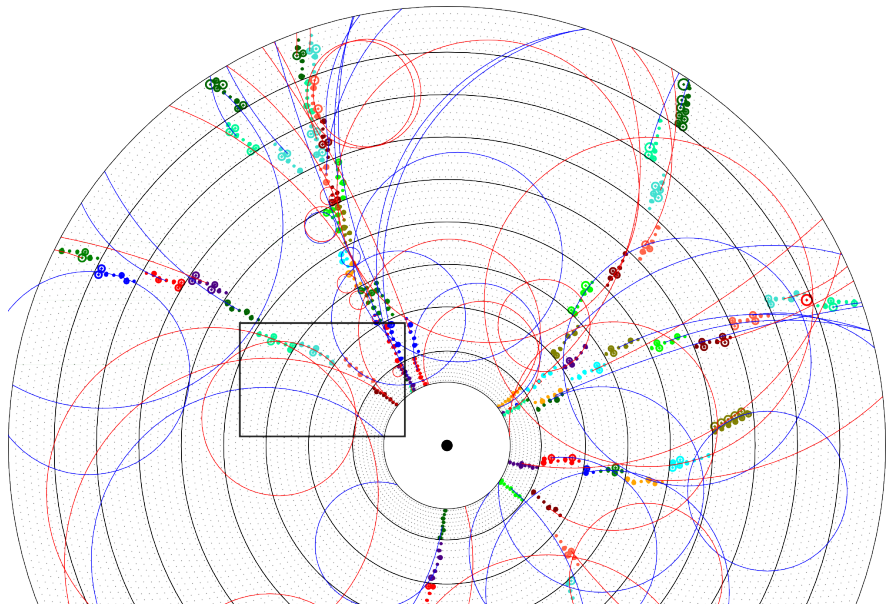
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# Segment fits to wire position - display



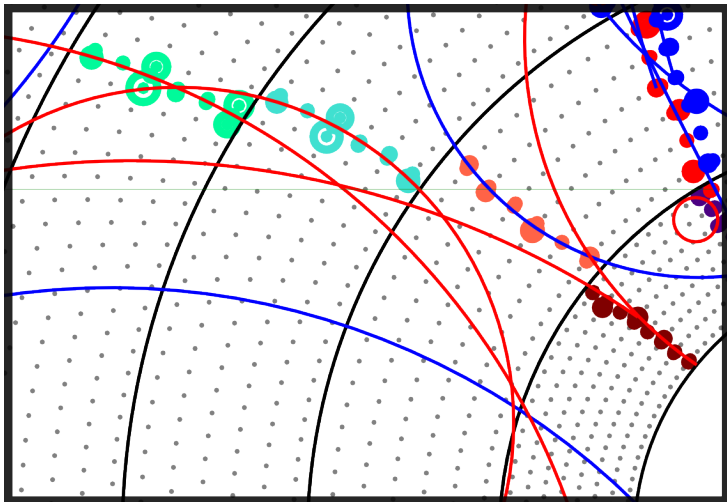
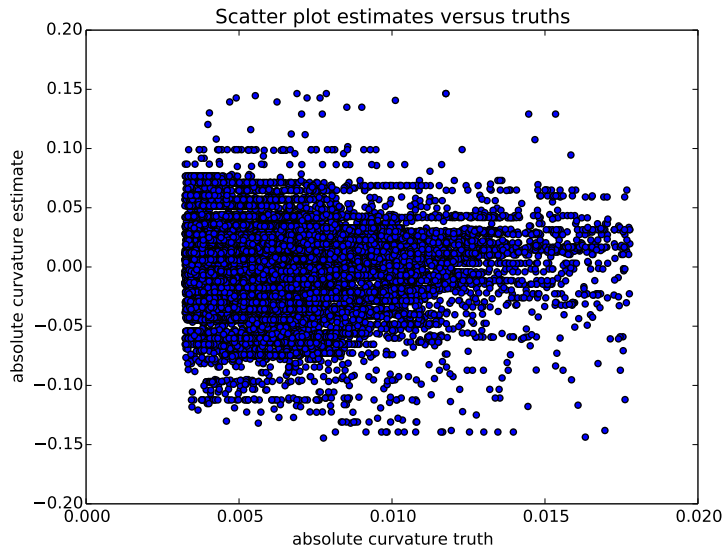
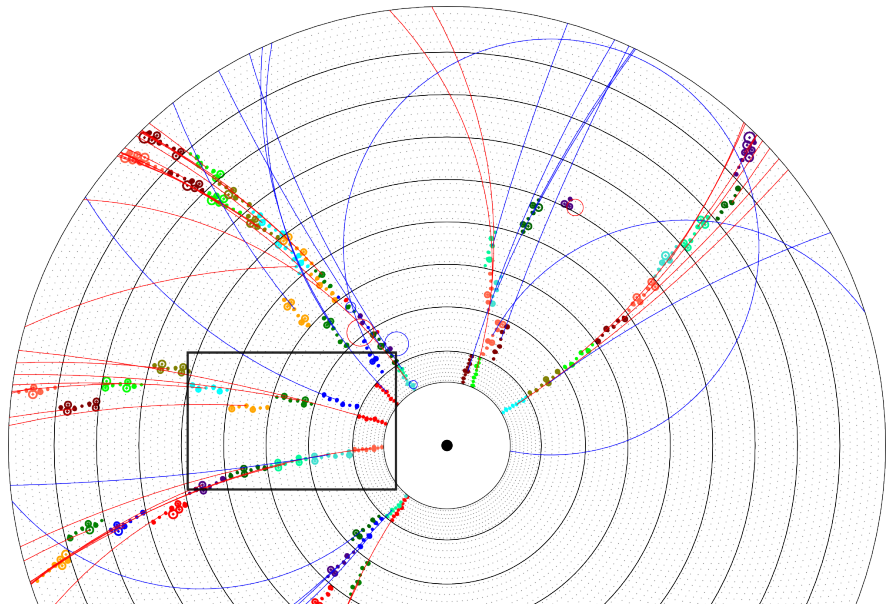


Figure 1: Segment fits have too high curvature, more information needed to straighten trajectories.







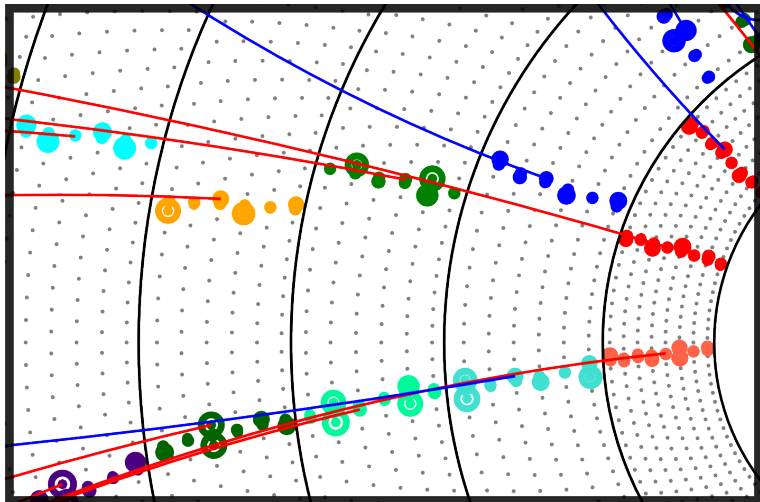
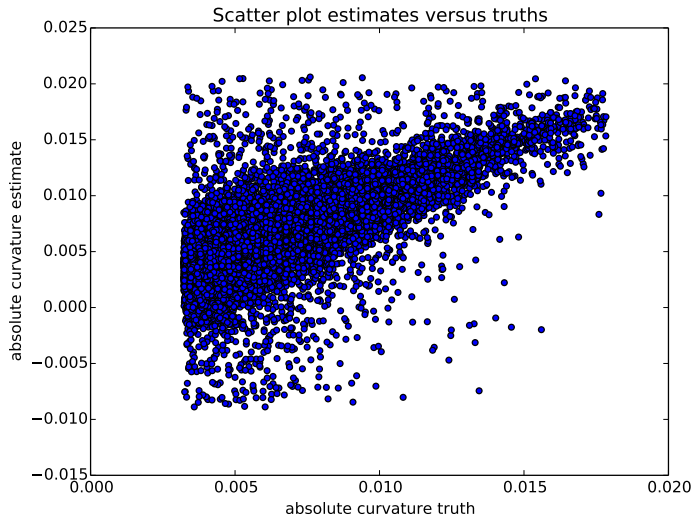


Figure 2: Using drift circles fits to axial segment yields right curvature. Fits to stereo segments have differing curvature. Here the curvature sign is opposite (blue vs. red).

# Axial segment fit to drift circles - estimation quality



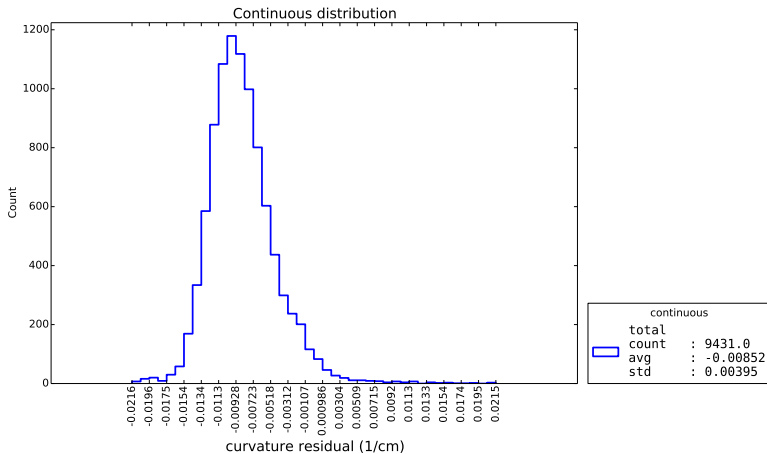
## Axial layers

- > Projection carried out by the wires is the xy projection.
- > Curvature in xy projection is the real curvature.
- > Unbiased fit in projection

## Stereo layers

- > Projection carried out by the wires changes with z position.
- > A particle trajectory can gradually pick up additional displacement as it is moving forward/backward.
- > Curvature in the projection and  $\tan \lambda$  are intertwined.
- > This generally leads to a bias.

## Curvature residuals in super layer 1

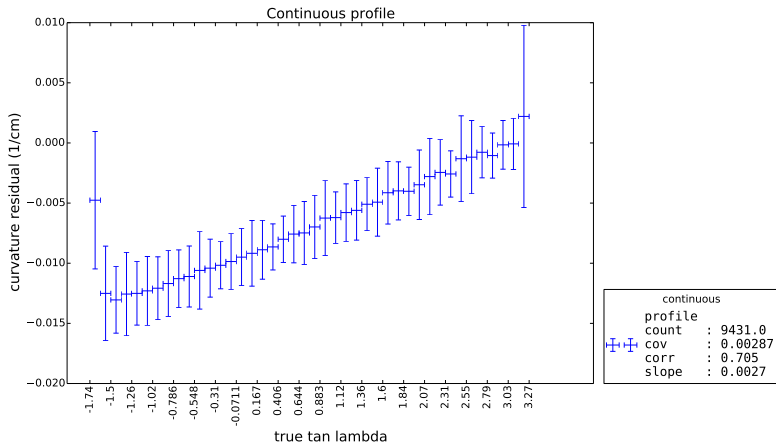


# Cause of the bias



## Stereo effect influences curvature in the xy-projection

Curvature residual versus tan lambda in super layer 1





## Covariance estimation

$$V_{ij}^{-1} = \sum_k w_k \frac{d\epsilon_k}{dn_i} \frac{d\epsilon_k}{dn_j}$$

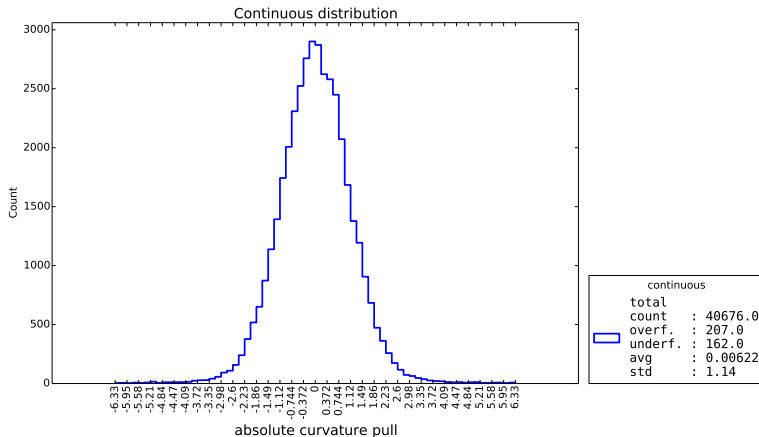
(Karimaki)

## Application to Riemann fit

- > Simple in normal parameters  $n$ , since they enter the distance  $\epsilon$  linearly.
- > Slightly complicating translation from four  $n$  parameters to three perigee parameters.



## Distribution of absolute curvature pull (clipped)





- > Fitting methods interfacing to other code easily do able.
- > Transport / extrapolation of perigee covariances to new reference point.
- > Helix class also doing closest approaches to points.

- > Parameter and covariance matrix estimation yield reasonable results for axial layers.
- > Fitting in stereo layers alone leads to a bias.
- > On segment level fitting the wire position is not enough.
- > **Active use of drift length is not optional.**
- > The latter **requires** the **correct right left passage hypotheses**.
- > Comparison of the right left passage to the Monte Carlo truth would be desirable.

- > Two dimensional fitting
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## Pairs instead of triples

- > Use pairs of segments as cells for the cellular automaton instead of triples

## Benefits

- > Needs only one filter in the creation step instead of two.
- > Tracks are allowed to end in stereo layers.
- > Fewer left over segments
- > Still enough information to construct a helix with from combination of axial and stereo information

## Kalmanesk combination of parameters

Combined helix covariance

$$V^{-1} = H_a^T \cdot V_a^{-1} \cdot H_a + H_s^T \cdot V_s^{-1} \cdot H_s$$

Combined helix parameters

$$x = V \cdot (H_a^T \cdot V_a^{-1} \cdot p_a + H_s^T \cdot V_s^{-1} \cdot p_s)$$

Residuals

$$r_{a/s} = p_{a/s} - H_{a/s} \cdot x$$

Combined chi square

$$\chi^2 = r_a^T \cdot V_a^{-1} \cdot r_a + r_s^T \cdot V_s^{-1} \cdot r_s$$

with perigee parameters  $p_{a/s}$ , perigee covariances  $V_{a/s}$  and segment ambiguity matrix  $H_{a/s}$  of axial and stereo segment respectively.

## For axial segments

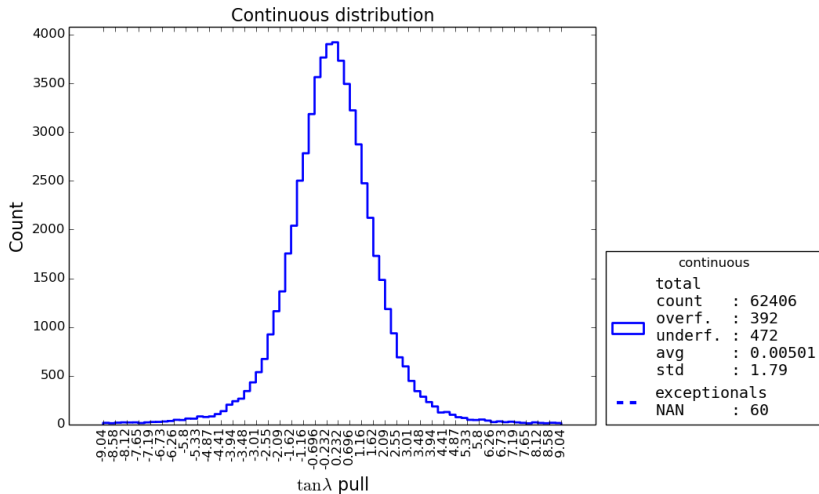
$$H_a = \frac{dp}{dx} = \begin{pmatrix} 1 & 0 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 & 0 \\ 0 & 0 & 1 & 0 & 0 \end{pmatrix}$$

## For stereo segments

$$H_s = \frac{dp}{dx} = \begin{pmatrix} 1 & 0 & 0 & 0 & 0 \\ 0 & 1 & 0 & \zeta & 0 \\ 0 & 0 & 1 & 0 & -\zeta \end{pmatrix}$$

where

$$\zeta = \frac{1}{\#hits} \sum_{hits} \frac{\text{wire vector}_{xy} \cdot \text{normal to trajectory}_{xy}}{\text{wire vector}_z}$$

Distribution of  $\tan\lambda$  pull (clipped)



- > Two dimensional fitting
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## Obvious

- > Eventually use the fit output in proper rejection of cells and neighborhoods.
- > Integrate the new track model (RecoTrack).

## Not so obvious

- > In superlayer segment merging
- > Break the forward backward symmetry.
- > Merge left over segments in superlayer 0 with VXD.
- > Reconstuct cycles in the cellular automaton as curlers.