



# Some thoughts on alignment

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Universität Heidelberg

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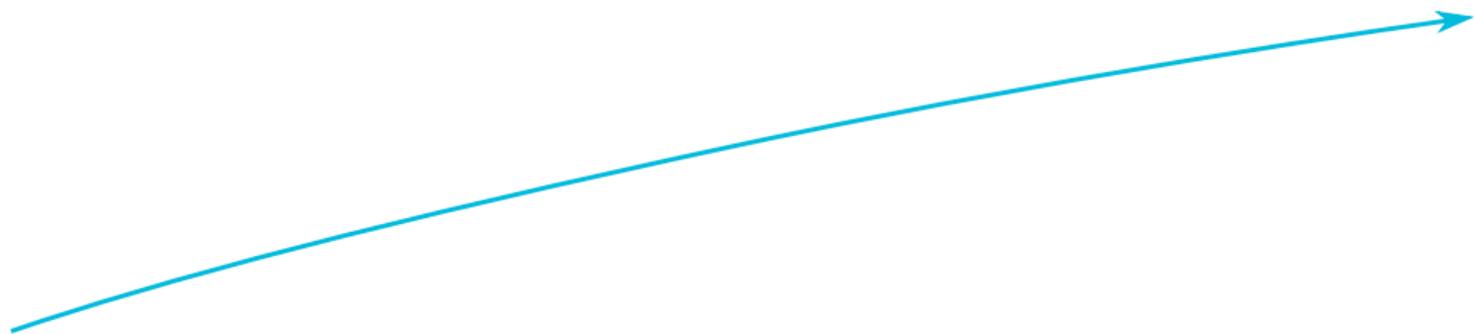
# Preface

- ▶ This is a collection of slides – I will breeze over some of them but am happy to revisit
- ▶ Just for starting the discussion
- ▶ Although I'm with Mu3e today, I used to be „Alignment Convener“ with CMS
- ▶ I'm an outsider, so please apologise for being mostly a „linear collider agnostic“.

Disclaimer: Material shown here has a strong CMS bias and was shown and public occasions back during my time with CMS.



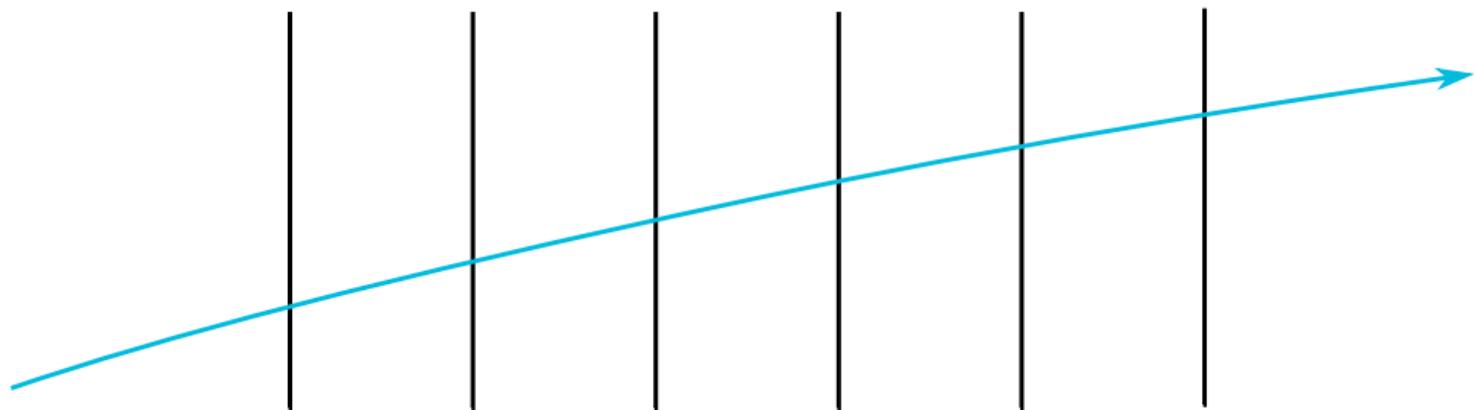
## Alignment of a simple toy tracker



You want to track a charged particle in a magnetic field



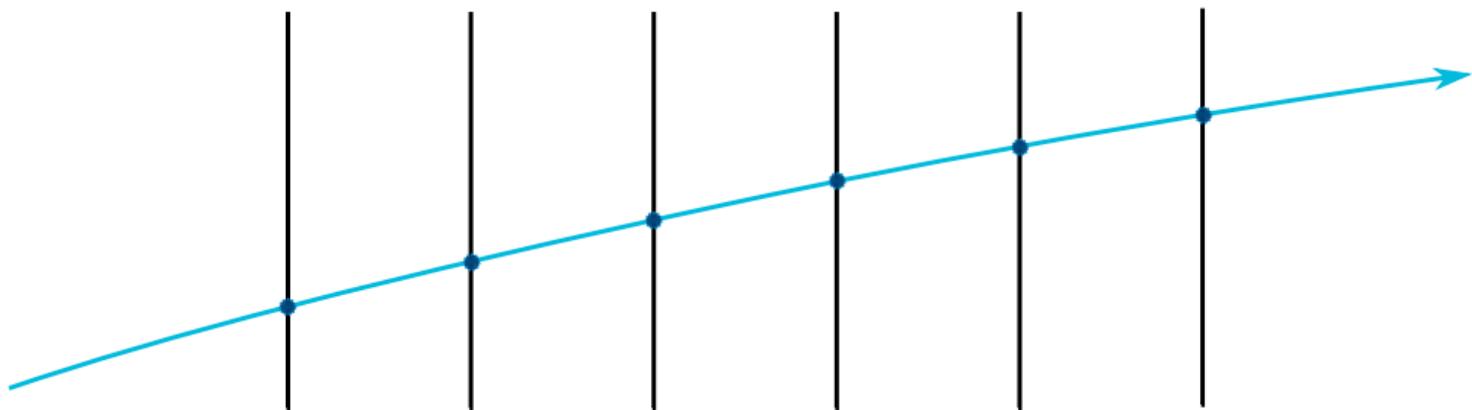
## Alignment of a simple toy tracker



So you take six modules of your high-precision tracking detectors



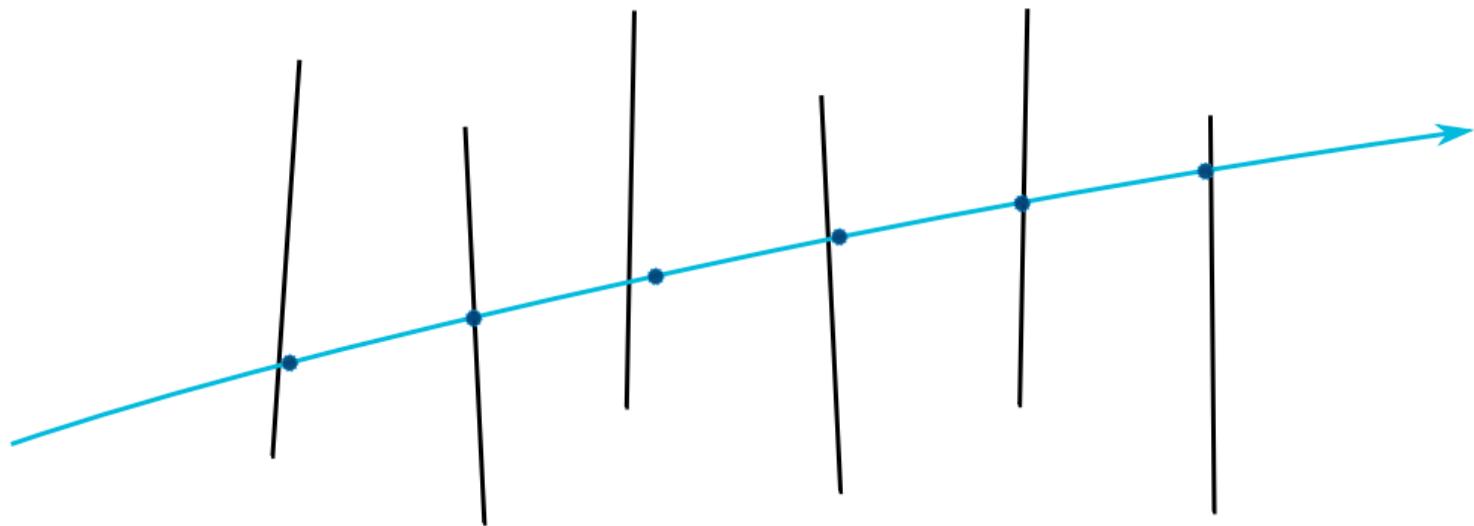
## Alignment of a simple toy tracker



and you think you have recorded the right signal.



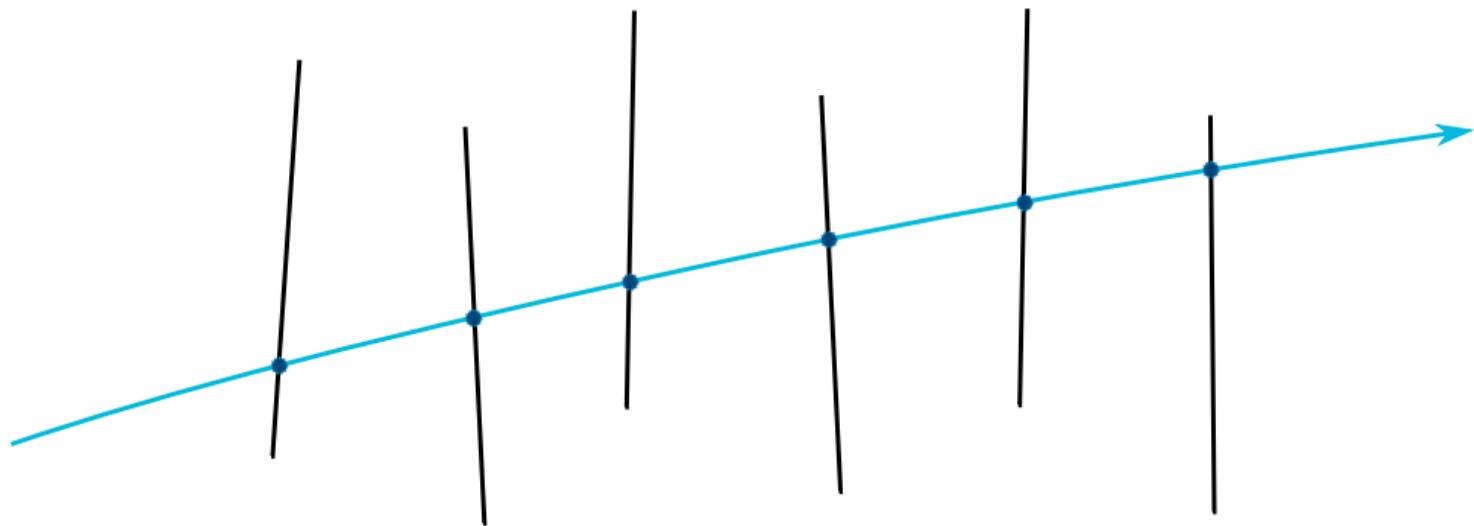
## Alignment of a simple toy tracker



But unfortunately the modules were not mounted precisely where needed



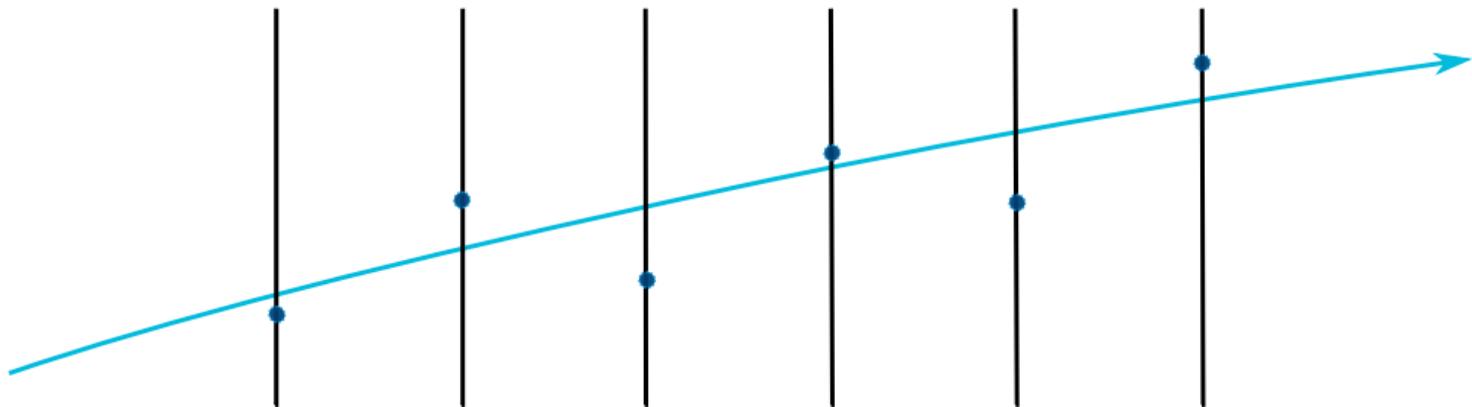
## Alignment of a simple toy tracker



and your hits are not where you thought



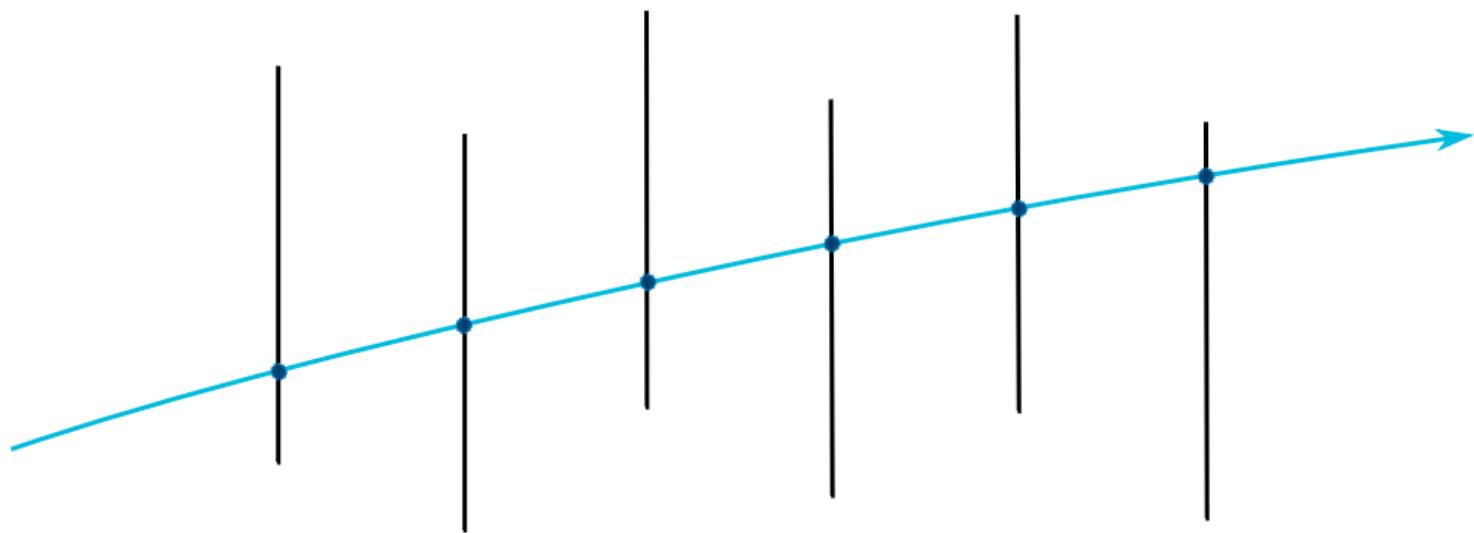
## Alignment of a simple toy tracker



You still assume an ideal tracker and you know that the particle follows a smooth trajectory



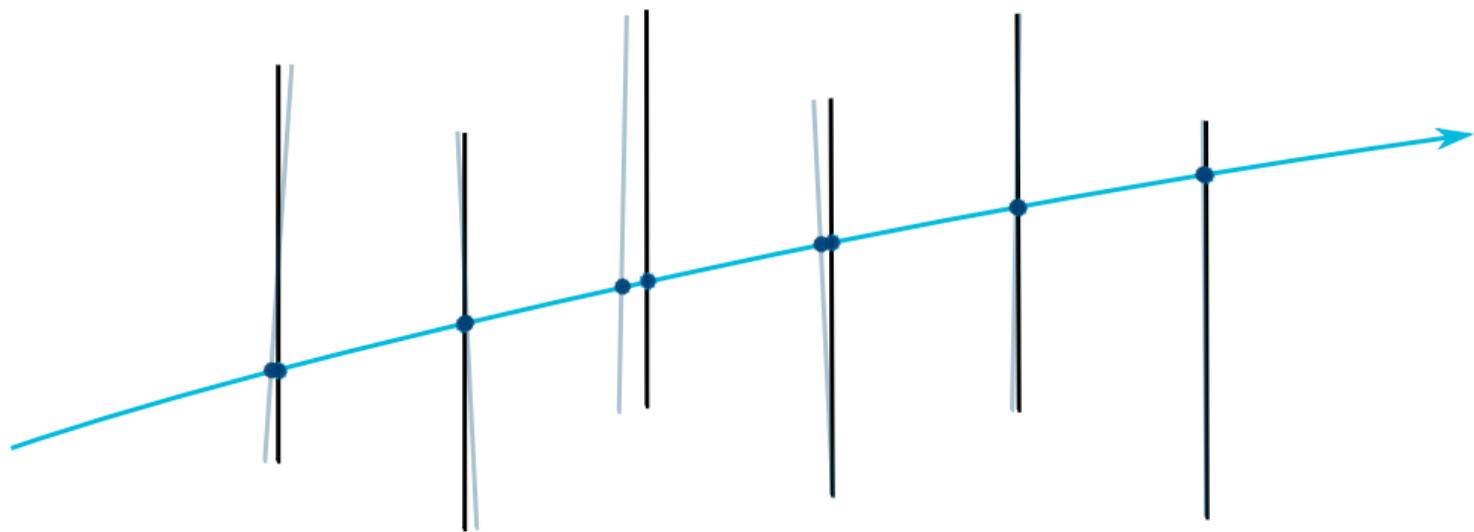
## Alignment of a simple toy tracker



so you correct the modules' positions.



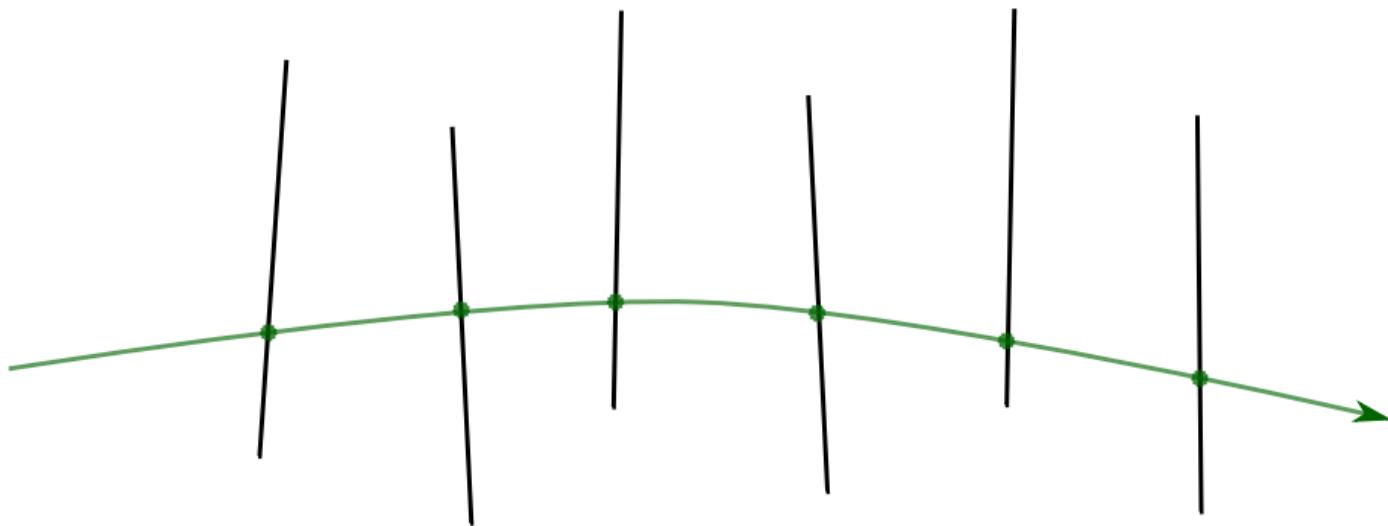
## Alignment of a simple toy tracker



But compared to the reality, you are still off.



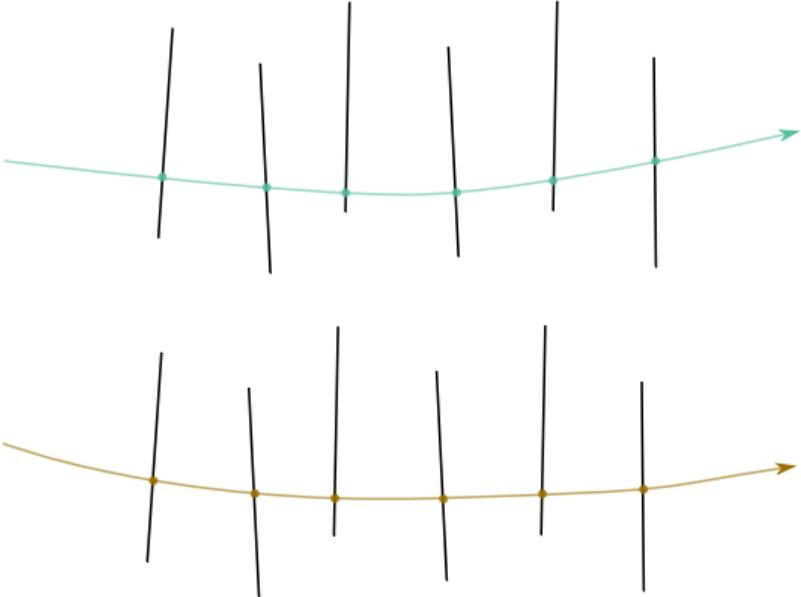
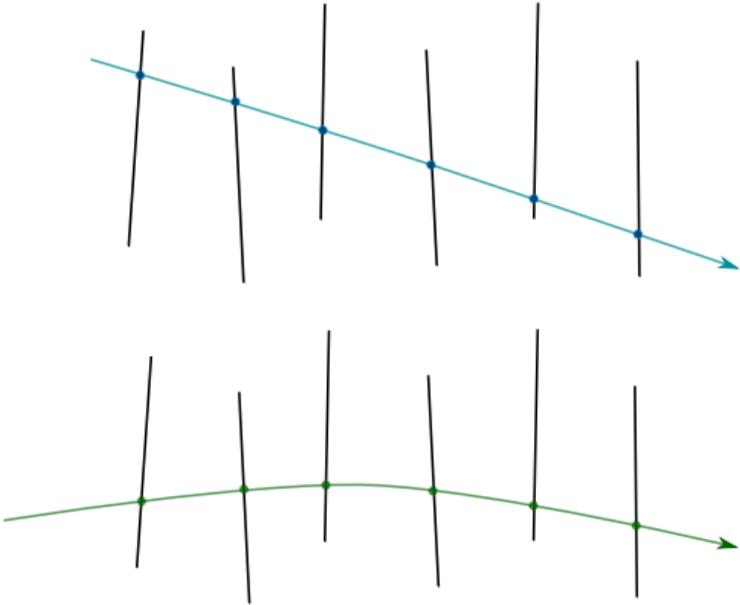
## Alignment of a simple toy tracker



So you record more tracks



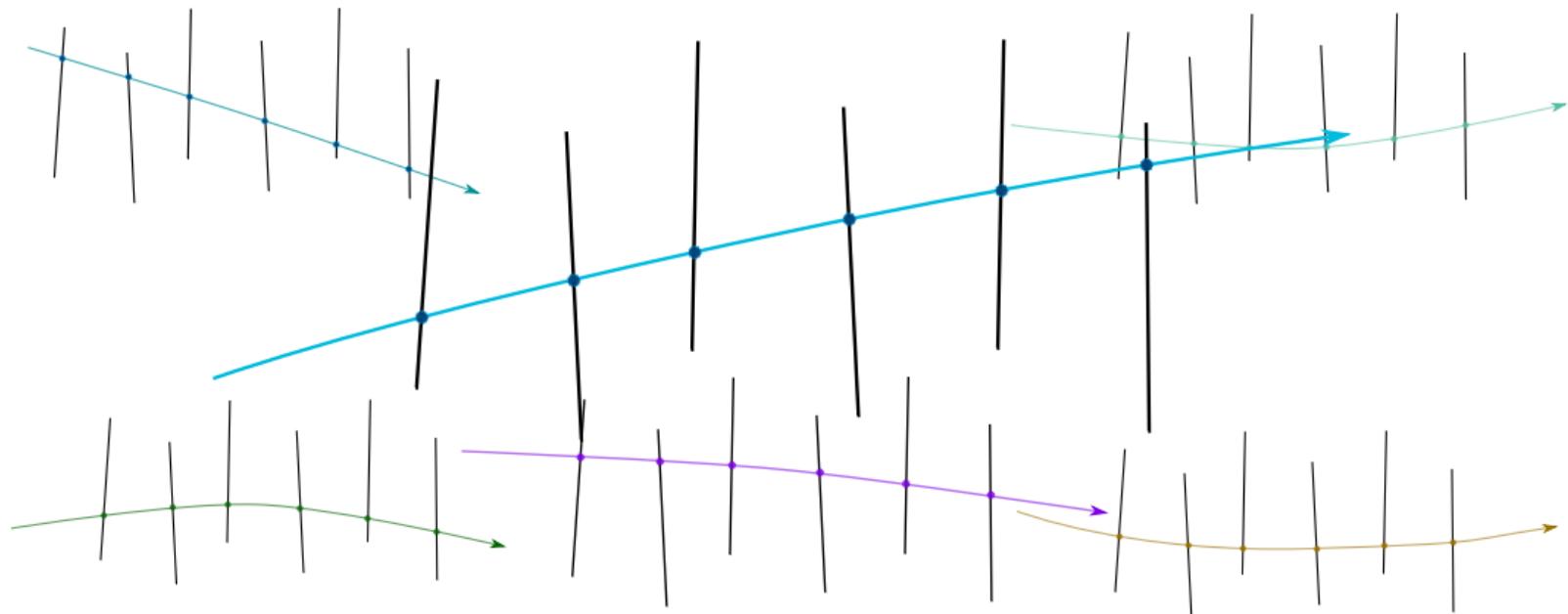
# Alignment of a simple toy tracker



and more



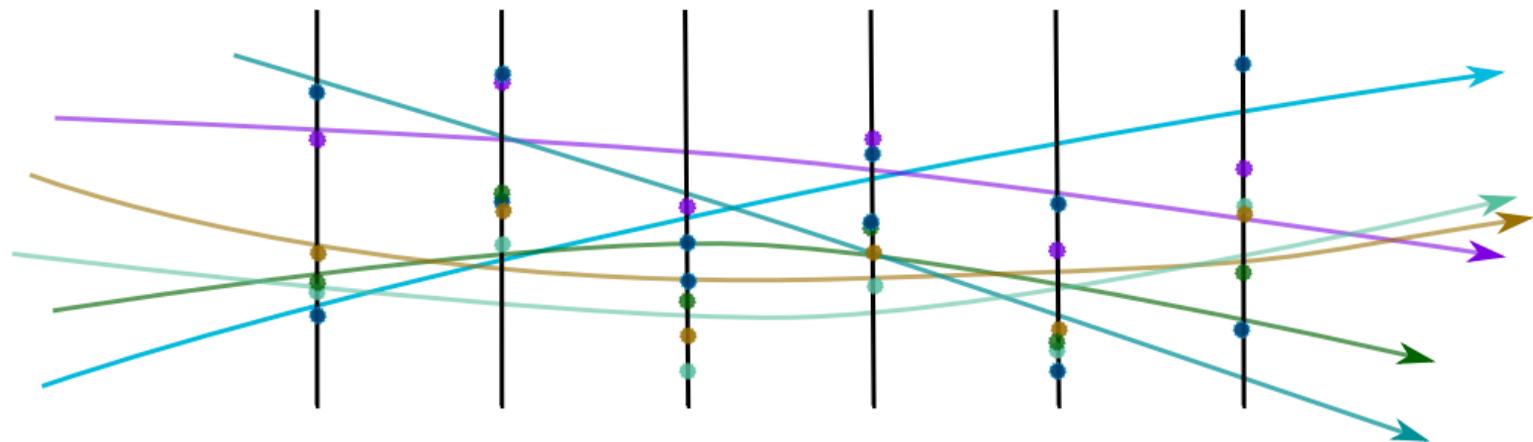
# Alignment of a simple toy tracker



and even more



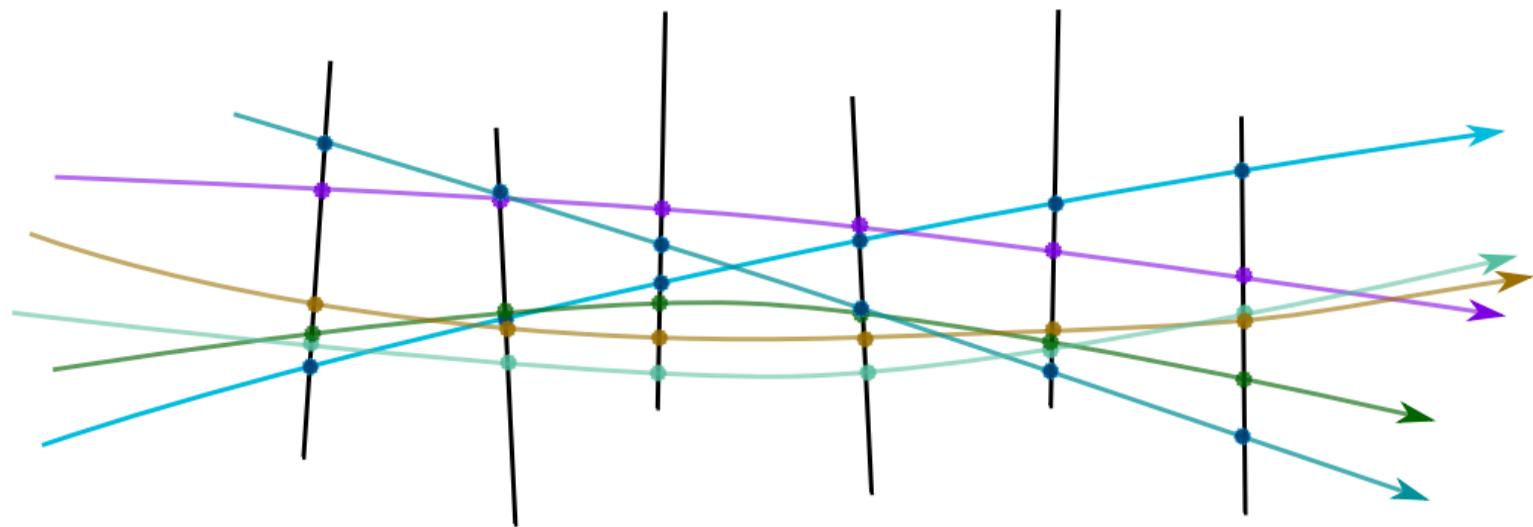
## Alignment of a simple toy tracker



you take them all into account and deduce the true position



## Alignment of a simple toy tracker



until you get it. This is what alignment has to do.



# Algorithms

- ▶ Alignment can be done as a variant of a *linear least squares* problem



# Algorithms

- ▶ Alignment can be done as a variant of a *linear least squares* problem
- ▶ The expression to be minimized is

$$\chi^2(\mathbf{p}, \mathbf{q}) = \sum_j^{\text{tracks}} \sum_i^{\text{hits}} \mathbf{r}_{ij}^T(\mathbf{p}, \mathbf{q}_j) \mathbf{V}_{ij}^{-1} \mathbf{r}_{ij}(\mathbf{p}, \mathbf{q}_j) \quad (1)$$

where

- ▶  $\mathbf{p}$ : alignment parameters describing the actual geometry
- ▶  $\mathbf{q}_j$ : track parameters of the  $j^{\text{th}}$  track
- ▶  $\mathbf{r}_{ij}$ : residual, distance between measured hit and position predicted by track fit
- ▶  $\mathbf{V}_{ij}^{-1}$ : the inverse covariance matrix, diagonal if measurements are uncorrelated

$\mathbf{V}_{ii}^{-1} = 1/\sigma_i^2$  with  $\sigma_i$  the Gaussian error of the measurement



## Algorithms

- ▶ Alignment can be done as a variant of a *linear least squares* problem
- ▶ The expression to be minimized is

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- ▶ Follows a  $\chi^2$  distribution for a given *number of degrees of freedom* (ndof), i.e.

$$\left\langle \frac{\chi^2(\mathbf{p}, \mathbf{q})}{\text{ndof}} \right\rangle = 1$$

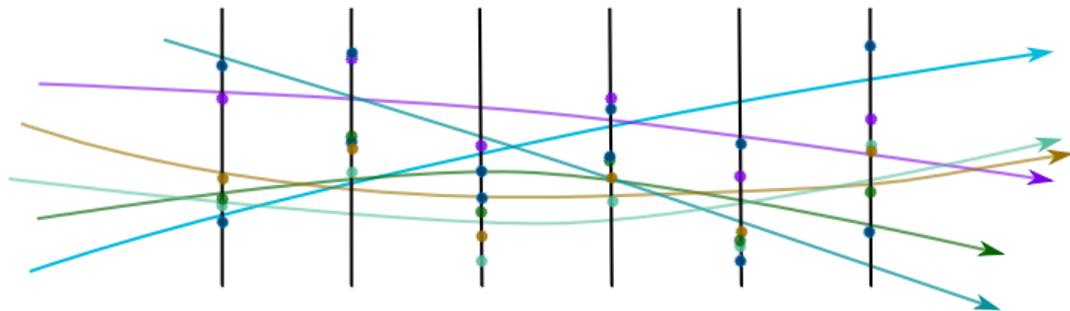
$$\langle \text{prob}(\chi^2, \text{ndof}) \rangle = 1/2$$



# Algorithms

CMS uses two competing algorithms:

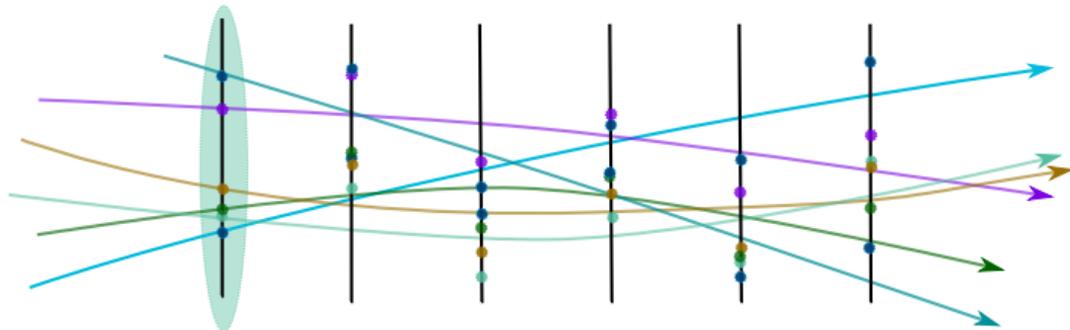
- ▶ one that works iteratively: HIP



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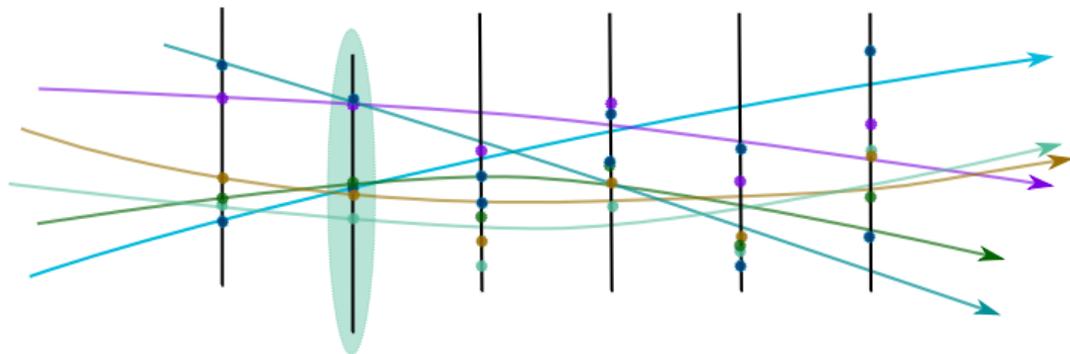
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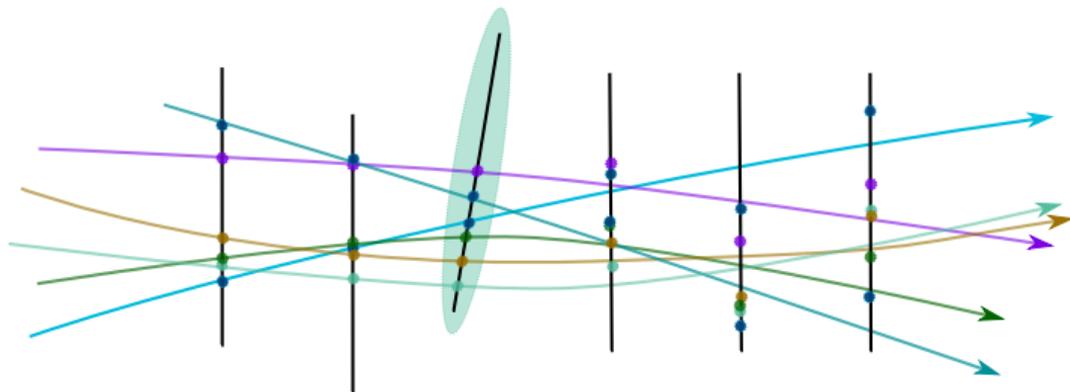
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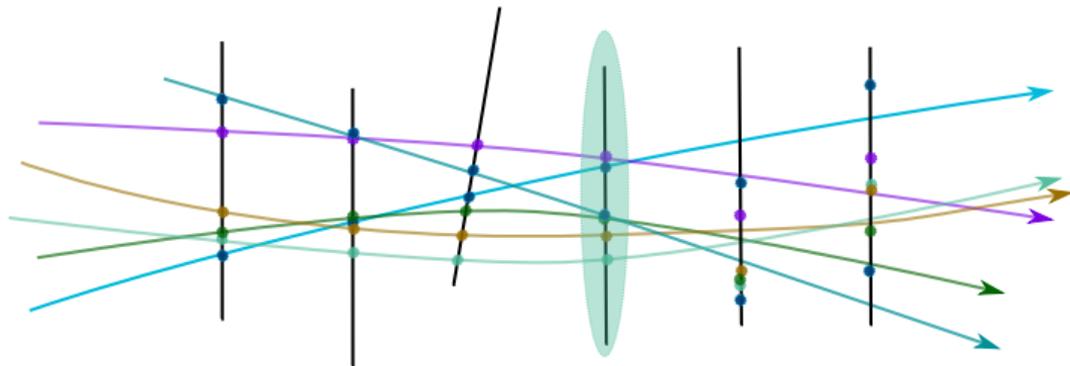
- ▶ one that works iteratively: HIP



# Algorithms

CMS uses two competing algorithms:

- ▶ one that works iteratively: HIP



i.e. the sum in the  $\chi^2$ -equation (1) runs over all tracks passing through one module, optimizes the parameters  $\mathbf{p}$  of this module.

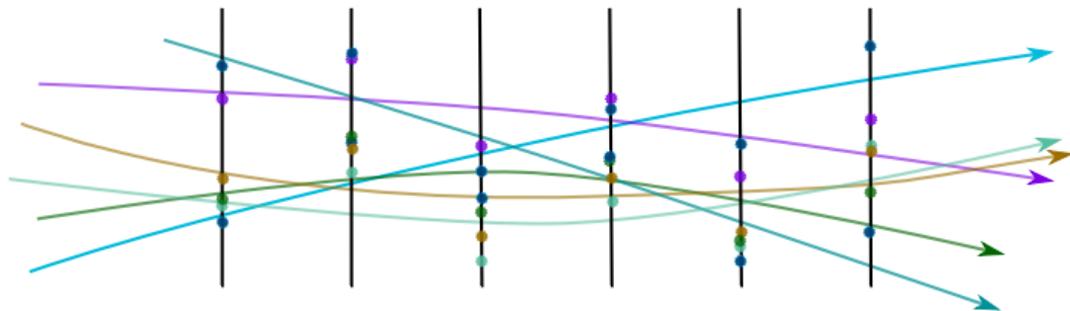
- ▶ **Benefit:** small problems to solve in each step
- ▶ **Advantage:** only local residuals contribute to  $\chi^2$ , tf. insensitive to large misalignment in rest of tracker  $\rightarrow$  startup alignment
- ▶ **Disadvantage:** a lot of iterations needed (recall: 16 588 modules with 5 to 6 d.o.f.)



# Algorithms

CMS uses two competing algorithms:

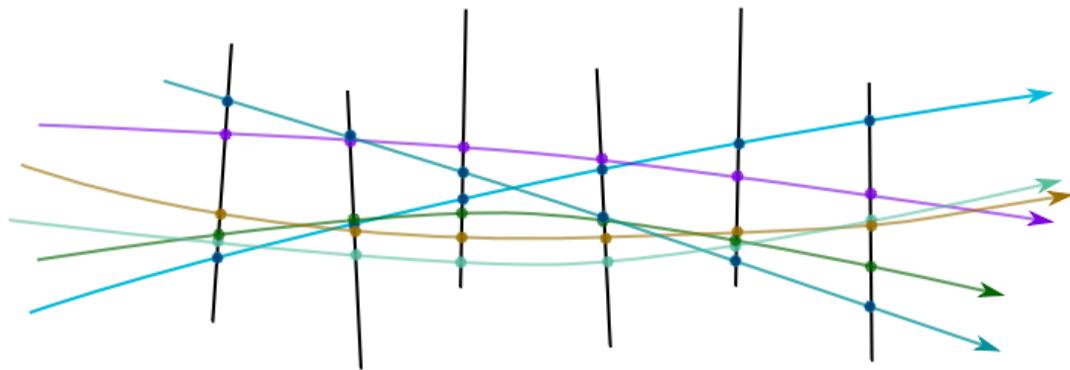
- ▶ one that works iteratively: HIP
- ▶ and one that does it in one step: Millepede-II



# Algorithms

CMS uses two competing algorithms:

- ▶ one that works iteratively: HIP
- ▶ and one that does it in one step: Millepede-II



i.e. the sum in the  $\chi^2$ -equation (1) runs over all tracks and modules

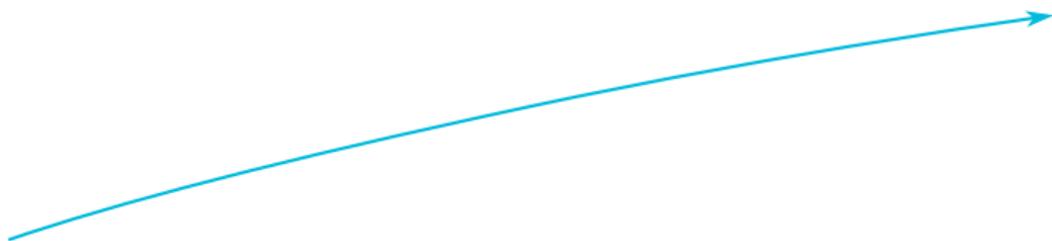
- ▶ **Benefit:** everything in one go
- ▶ **Disadvantage:** huge problem to solve  
(16 588 modules, 5 to 6 d.o.f.  $\Rightarrow$  80 000  $\times$  80 000-matrix to handle)



## Material effects

Two effects we have to take into account:

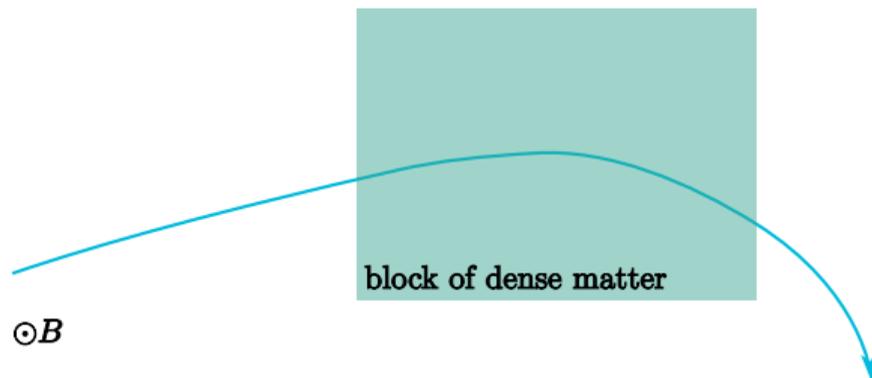
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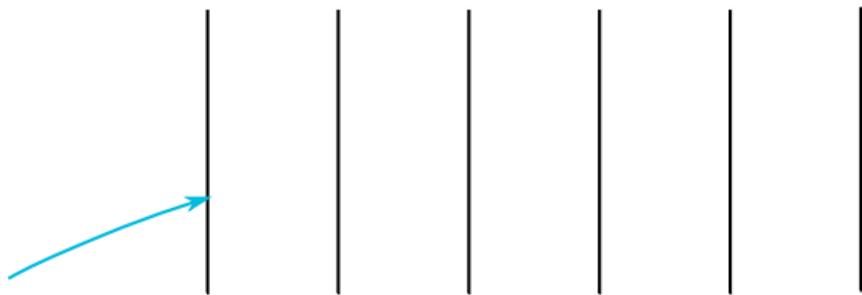
The charged track loses energy while passing through matter  
 $\Rightarrow$  particle loses momentum



## Material effects

Two effects we have to take into account:

- ▶ Energy loss
- ▶ Multiple scattering



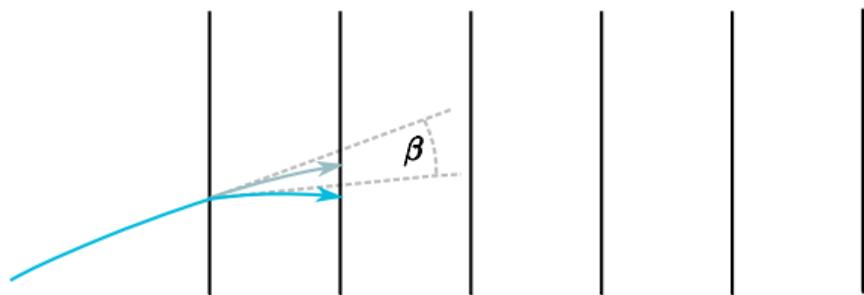
A charged particle hits material



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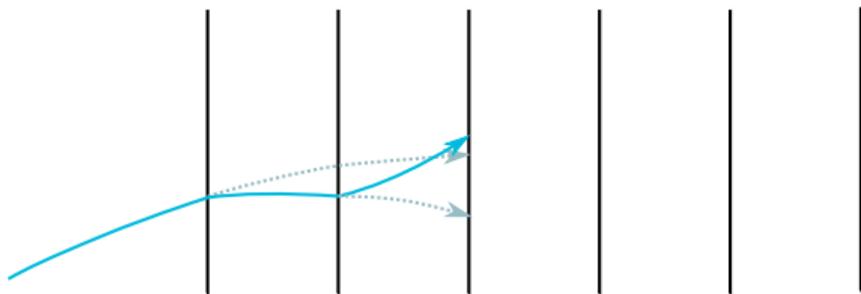
A charged particle hits material and scatters off some atoms in the material



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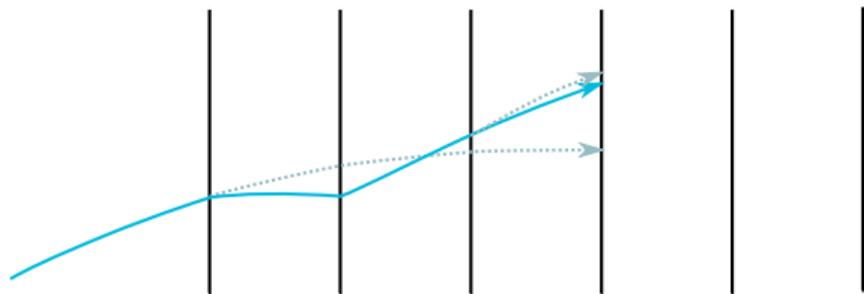
A charged particle hits material and scatters off some atoms in the material which changes its direction



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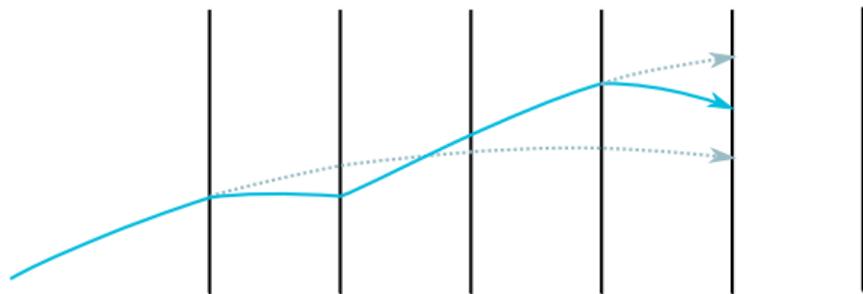
A charged particle hits material and scatters off some atoms in the material which changes its direction in a random fashion



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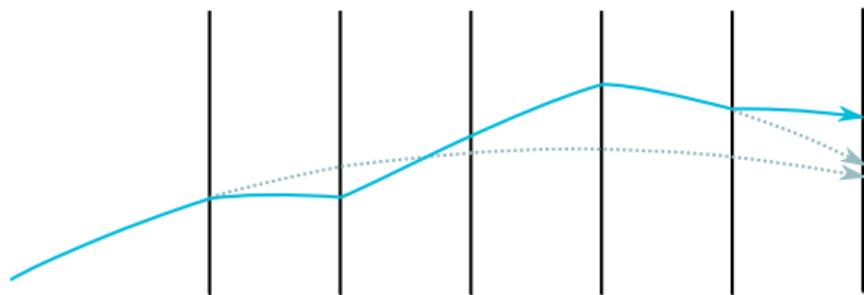
A charged particle hits material and scatters off some atoms in the material which changes its direction in a random fashion at every plane made of matter



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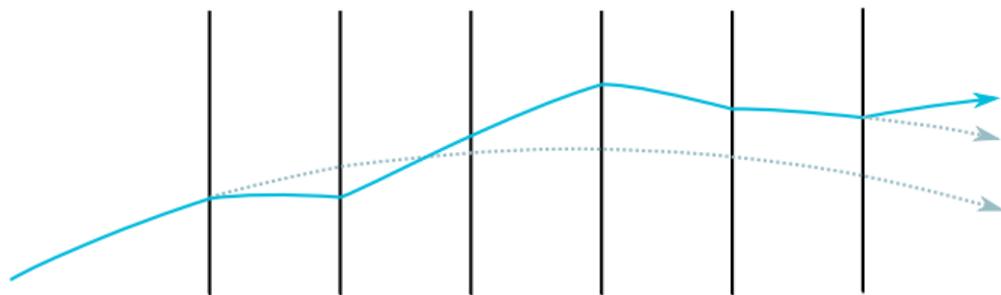
A charged particle hits material and scatters off some atoms in the material which changes its direction in a random fashion at every plane made of matter and ends up



## Material effects

Two effects we have to take into account:

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A charged particle hits material and scatters off some atoms in the material which changes its direction in a random fashion at every plane made of matter and ends up at a significantly different place and direction.



## Multiple scattering

The mean of the deflection angle is  $\langle\beta\rangle = 0$ , but the  $\sigma(\beta)$  can be approximated by<sup>1</sup>:

$$\sigma(\beta) = \frac{13.6 \text{ MeV}}{vp} z \sqrt{x/X_0} [1 + 0.038 \ln(x/X_0)] \quad (2)$$

- ▶  $v = \beta c$  velocity of particle (here  $\beta$  is the relativistic velocity factor)
- ▶  $p$  its momentum
- ▶  $z$  its charge
- ▶  $x/X_0$  thickness of traversed material in radiation lengths



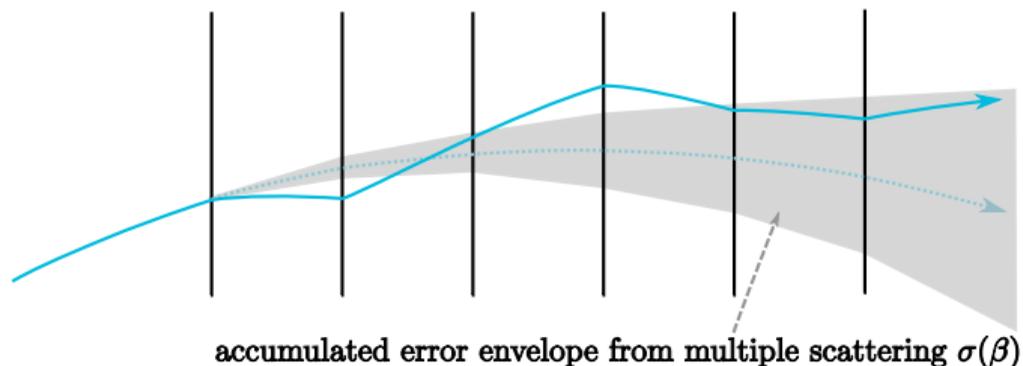
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<sup>1</sup>formula as found in PDG booklet

## Multiple scattering

How to take this into account?

- ▶ First attempt: Error envelope around whole track without multiple scattering



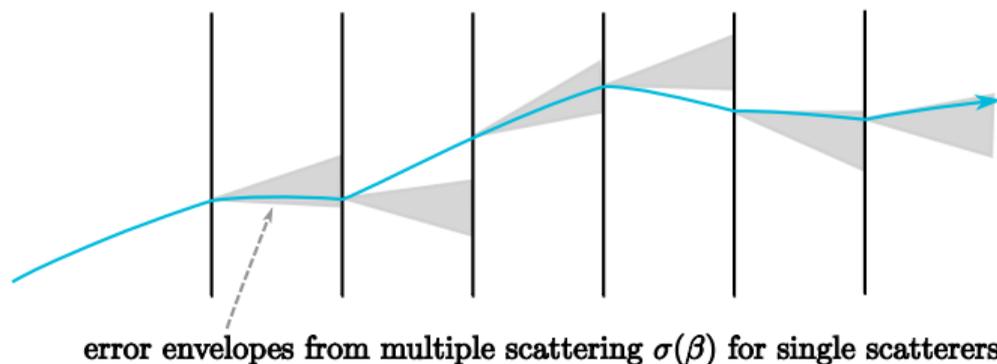
**Disadvantage:**  $\sigma$  of hits larger than expected, tf.  $\langle \chi^2 / \text{ndof} \rangle < 1$   
Difficult to judge result as  $\chi^2$ -distribution will be distorted.  
This approach would work if correlations introduced by multiple scattering are taken into account. This is not implemented in Millepede-II.



## Multiple scattering

How to take this into account?

- ▶ First attempt: Error envelope around whole track without multiple scattering
- ▶ Improvement: **Broken lines**  
Allow for kinks within error envelope for each track segment and use  $\langle \beta \rangle = 0$



**Advantage:** Track model matches reality,  $\langle \chi^2/\text{ndof} \rangle = 1$



## Multiple scattering

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- ▶ First attempt: Error envelope around whole track without multiple scattering
- ▶ Improvement: **Broken lines**  
Allow for kinks within error envelope for each track segment and use  $\langle \beta \rangle = 0$   
**Advantage:** Track model matches reality,  $\langle \chi^2 / \text{ndof} \rangle = 1$
- ▶ Have to pay a price for proper implementation of multiple scattering: correlated hits or more track parameters.  
Choice of broken lines leads to bordered band matrix structure for the local fit  $\Rightarrow$  fast solution by Cholesky decomposition.



## Results

With this improved track model available in CMS-software, we did alignment using Millepede-II and got

- ▶ nice results using track data from **collisions**:

$$\langle \chi^2/\text{ndof} \rangle \approx 1.04$$

This is good when compared to simple envelope where  
 $\langle \chi^2/\text{ndof} \rangle \approx 0.34$



## Results

With this improved track model available in CMS-software, we did alignment using Millepede-II and got

- ▶ nice results using track data from **collisions**:

$$\langle \chi^2/\text{ndof} \rangle \approx 1.04$$

- ▶ unusable results using track data from **cosmic rays**:

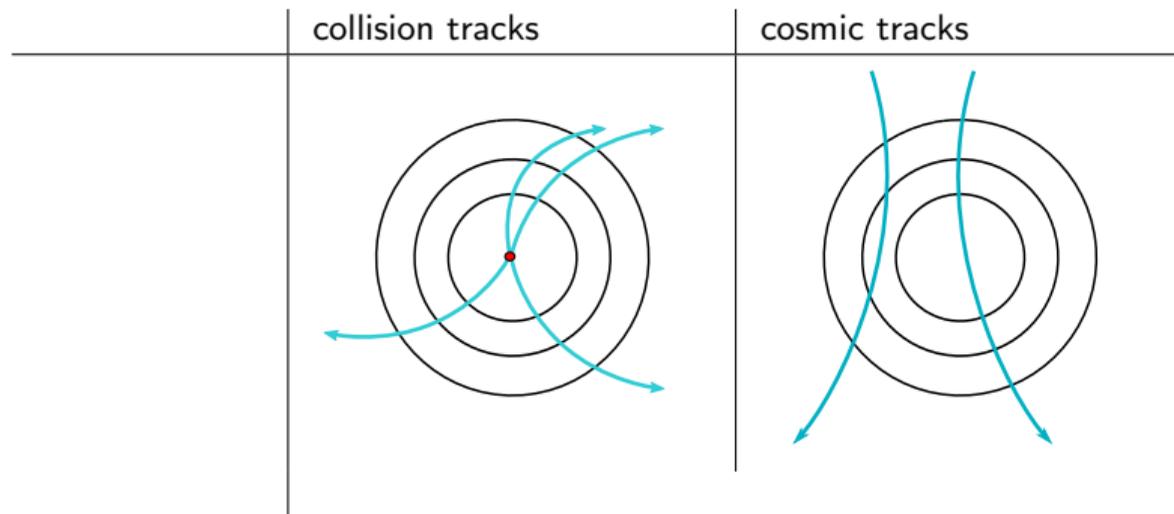
$$\langle \chi^2/\text{ndof} \rangle > 1.6$$

Millepede-II outlier treatment requires fraction of outliers  $< 1/3$ . In this case, it aborted before finish.



# Results

What are the differences between the two classes of tracks?



## Results

What are the differences between the two classes of tracks?<sup>2</sup>

	collision tracks	cosmic tracks
Momentum	small, larger contribution of multiple scattering	medium
Track angle (in barrel)	medium, driven by $q/p \cdot R_{\text{hit}}$	large, driven by $d_0/R_{\text{hit}}$

- ▶  $q/p$ : charge signed inverse momentum
- ▶  $R_{\text{hit}}$ : radial distance of hit to the tracker center
- ▶  $d_0$ : distance of closest approach of track to tracker center



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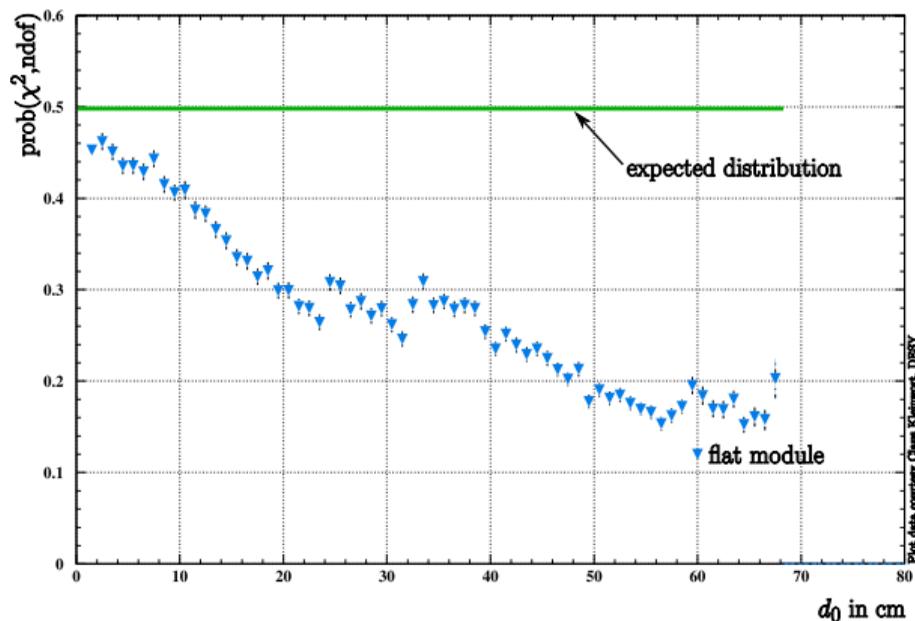
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Note: Plots shown on following slides are old but have educational qualities.



# Results

Distribution of the probability of the  $\chi^2$  vs.  $d_0$ :



Individual tracks from cosmic rays (CRAFT09) were fitted and binned. Shown are averages (markers) and RMS (error bars).



## Results

We clearly have a problem.

- ▶ First suspect: calculation of the hit errors at large angles.



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But this was not the solution. Just shifted/distorted the pattern.  
But this allowed for a hot fix anyway.



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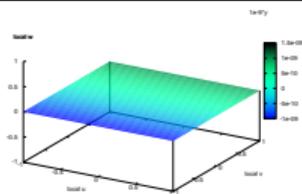
- ▶ First suspect: calculation of the hit errors at large angles.  
But this was not the solution. Just shifted/distorted the pattern.  
But this allowed for a hot fix anyway.
- ▶ May it be that we have a more complicated surface than flat sensors?



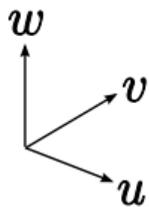
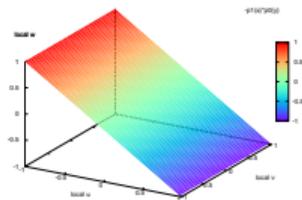
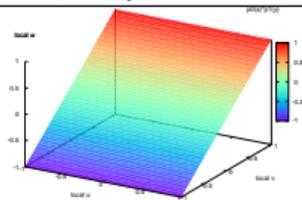
# More complicated surface distortions

The movements in alignment are superpositions of

Shifts along the axes



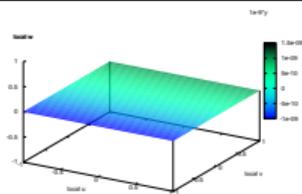
Slopes (i.e. rotations)



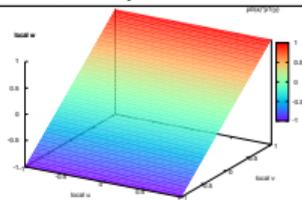
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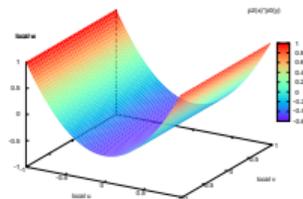
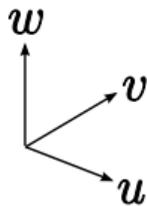
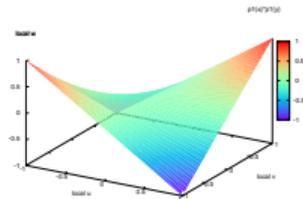
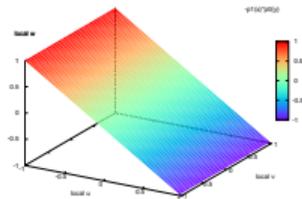
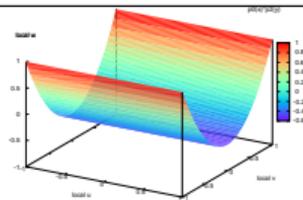
Shifts along the axes



Slopes (i.e. rotations)



bowed surfaces

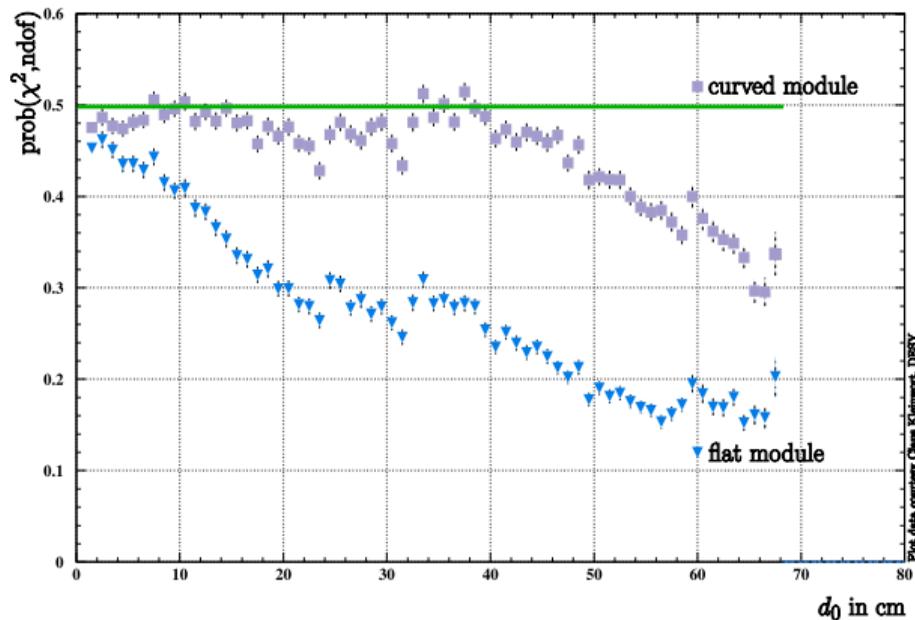


This adds 3 more parameters per module. Does this help?



## More complicated surface – more results

The same distribution of the probability of the  $\chi^2$  vs.  $d_0$  looks far better now:



But for  $d_0 \gtrsim 40$  cm still something happens



## Results

The topologies of the strip modules differ in the regions of the tracker:

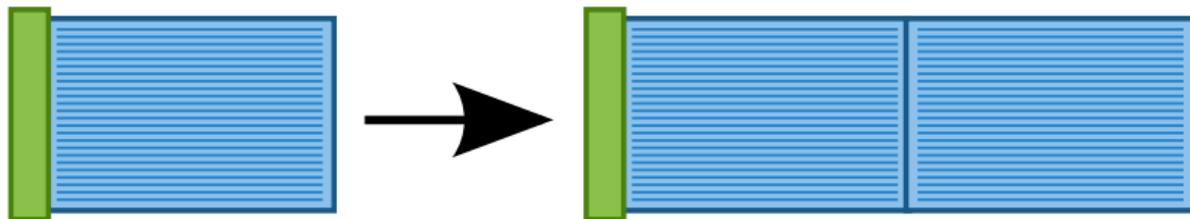


In the inner regions, the modules consist of one sensor



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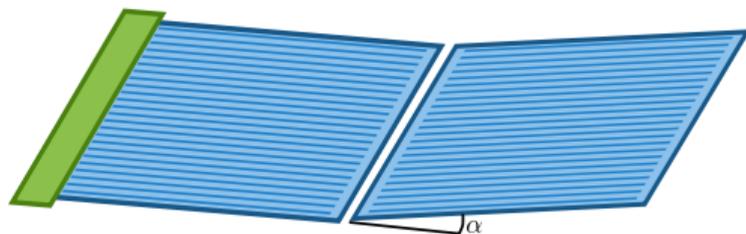


In the outer regions, the modules are made of two sensors, daisy-chained to the read-out electronics



## Results

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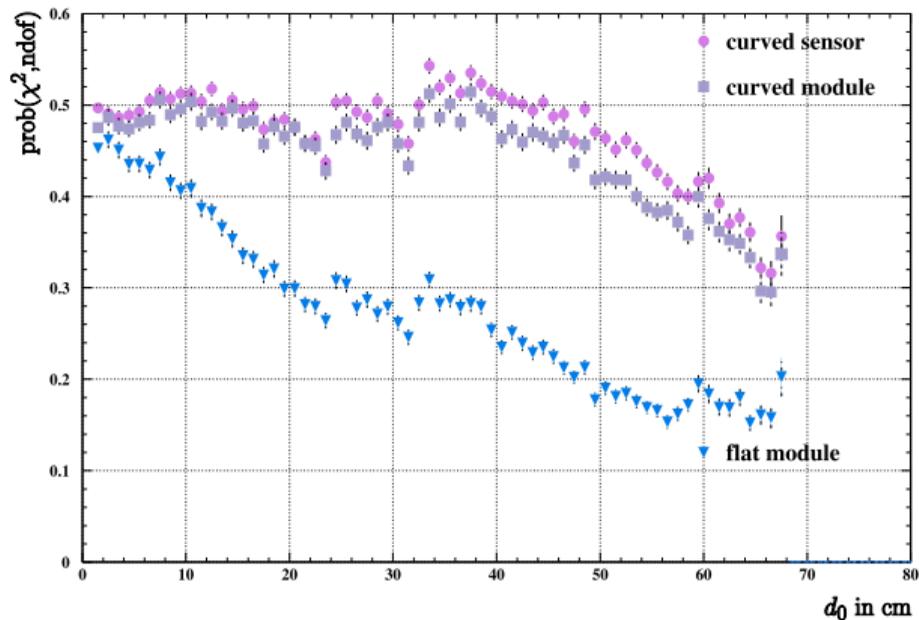


These composite modules may have a kink in between



## More complicated surface – more results

Splitting the composite sensors in two gives an improvement:

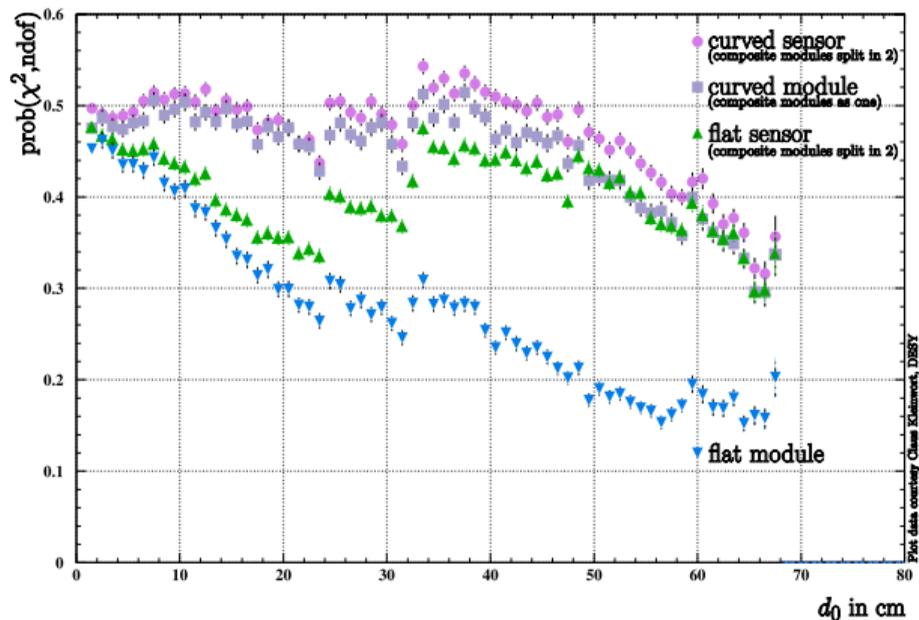


Observe that the track- $\chi^2$  depends on the full track, not only the nearest hit.



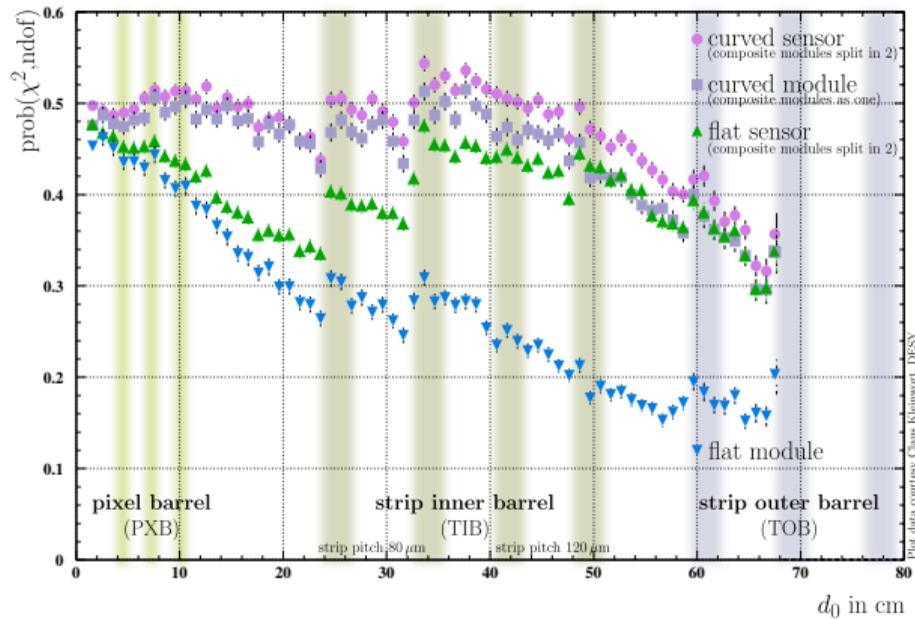
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Here we have the full picture: Allow for bows in the sensor surface and for kinks between composite modules:



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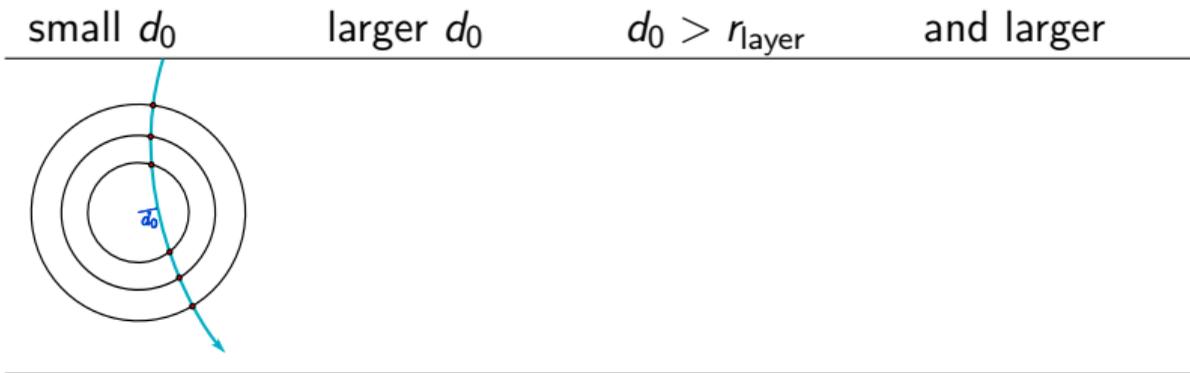


Adding the boundaries of the barrel layers is instructive and shows that an explanation for the jumps.



## More complicated surface – more results

To understand these jumps, we need to see what happens as  $d_0$  grows:

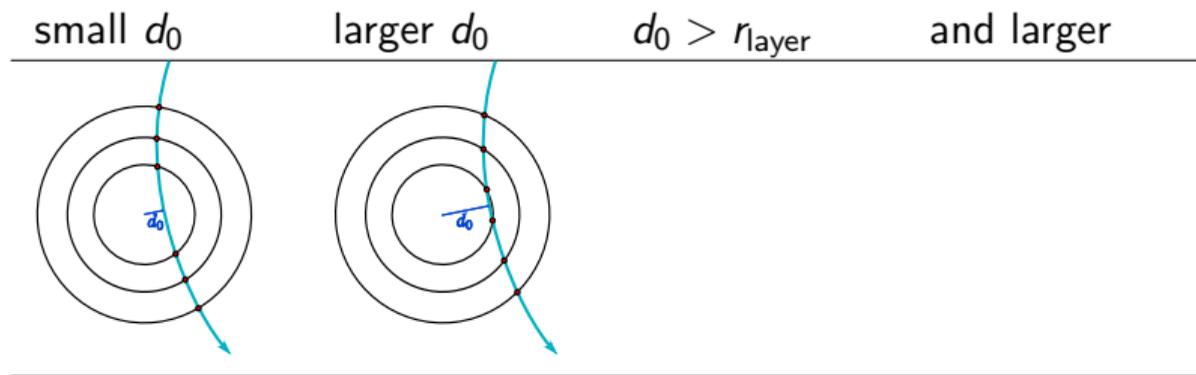


Such a track shows moderate track angles at every hit



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To understand these jumps, we need to see what happens as  $d_0$  grows:

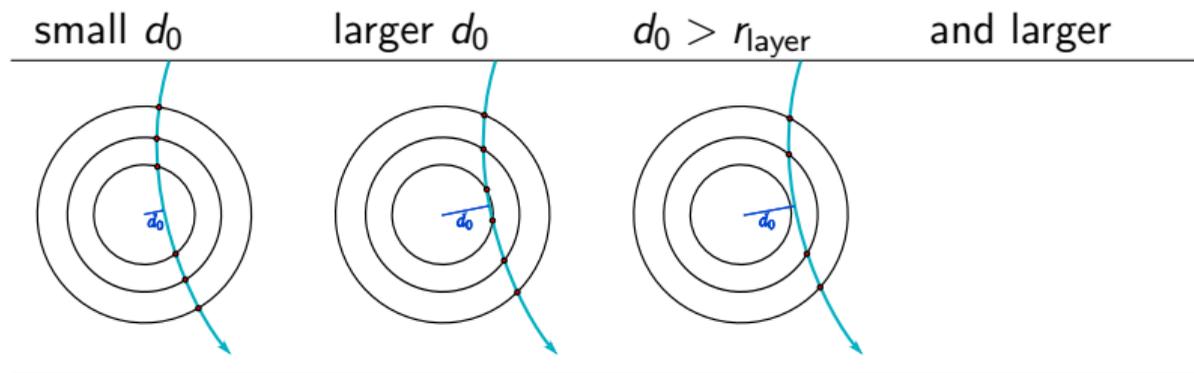


In this case, the track angles at the innermost hits are larger, the  $\chi^2$  gets deteriorated by a wrong surface description



## More complicated surface – more results

To understand these jumps, we need to see what happens as  $d_0$  grows:

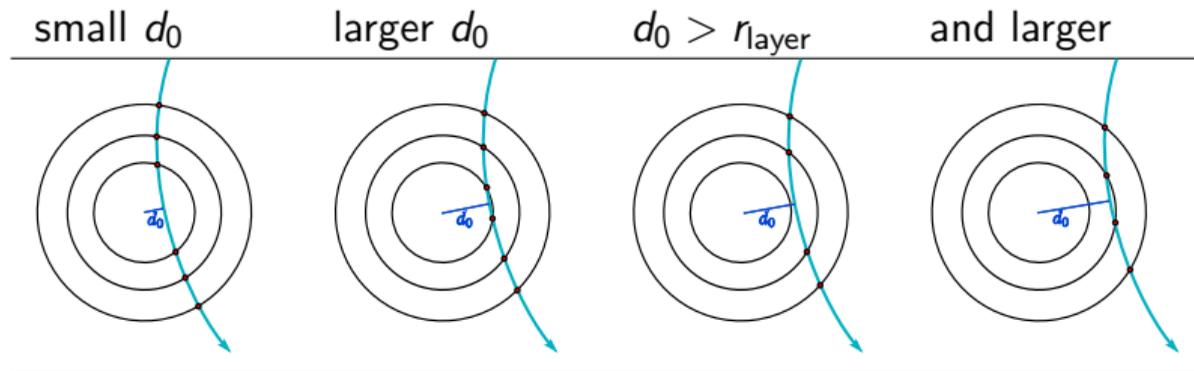


The innermost hits are lost, the  $\chi^2$  jumps to a better, though still lower than ideal, value



## More complicated surface – more results

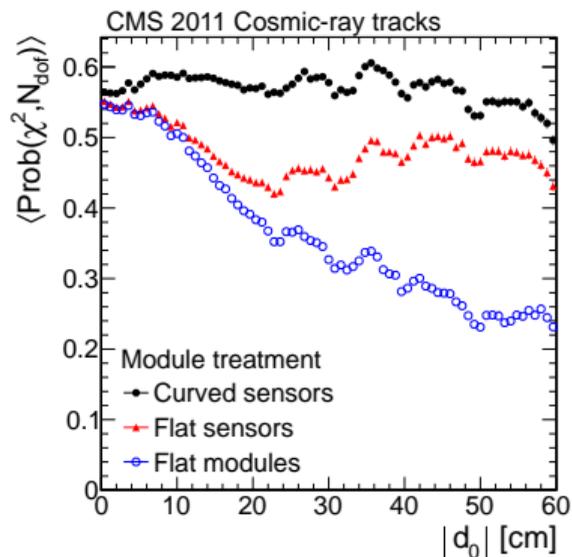
To understand these jumps, we need to see what happens as  $d_0$  grows:



And the game starts over again



# Results



And this is the plot from the paper TRK-11-002, recently published.

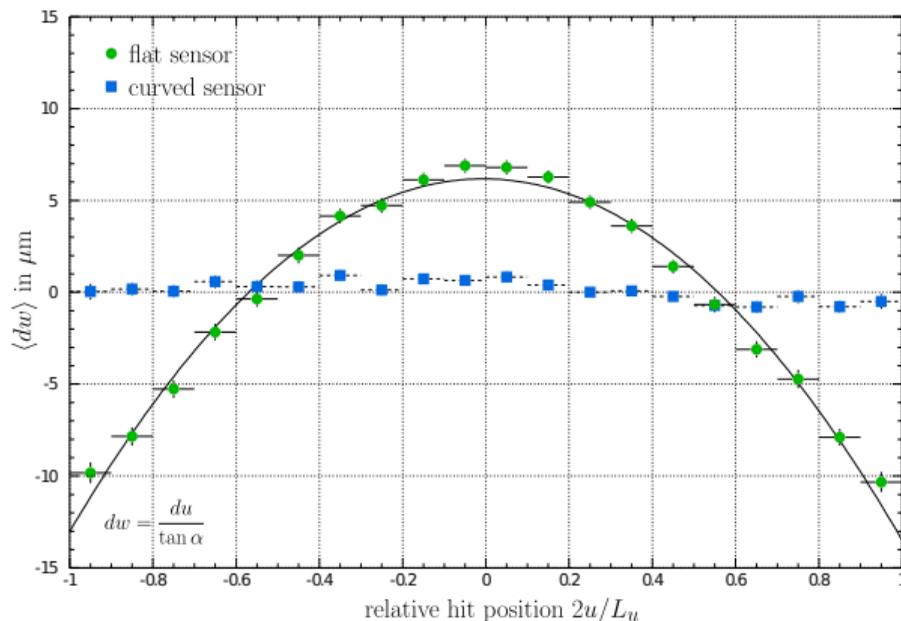
<http://arxiv.org/abs/1403.2286>

With strong DESY involvement – big THANKS on behalf of the CMS TkAlignment conveners!



# Results

Another look at the same problem:



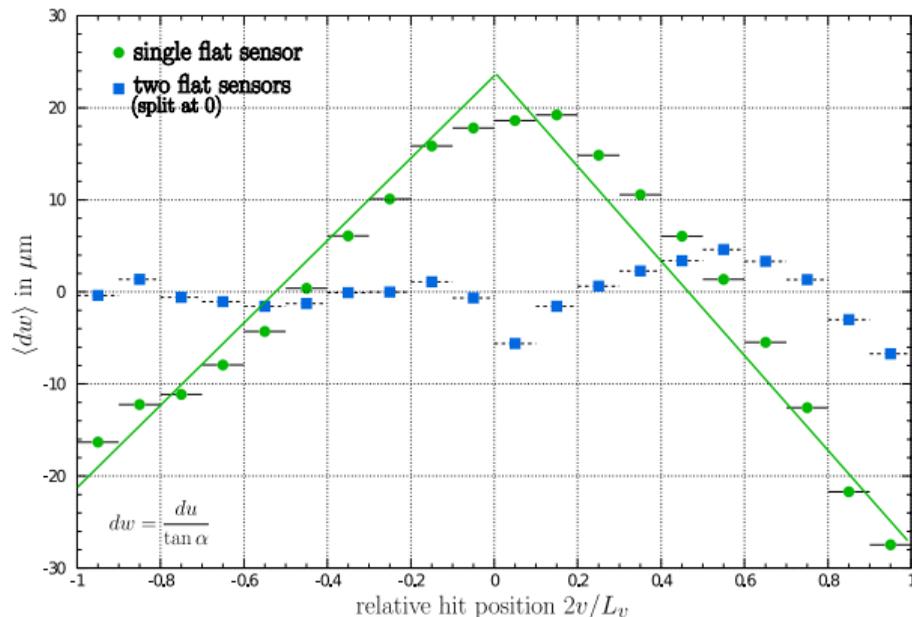
This shows the residuals when aligning for flat and curved sensor. Observe: The bowed curve belongs to the flat sensor assumption.

- ▶  $\alpha$ : track angle to the normal
- ▶  $du$ : residual of measured and predicted position (track fit)



# Results

And for composite modules:



Again this confirms our findings. The bow in the right halve of the sensor is visible.



# Thoughts

Allow for a few personal comments on alignment of silicon detectors:

- ▶ **Survey** comes labelled with an expiration date (but you can't read it).

Tracks in low-occupancy regimes – Cosmics – can be reconstructed even when heavily misaligned



# Thoughts

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Tracks in low-occupancy regimes – Cosmics – can be reconstructed even when heavily misaligned
- ▶ **Scale** comes to a large extent implicit when properly implemented  
Sensor pitch is a very precise ruler (photolithography)



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Movements slower than the time to collect tracks for alignment are perfectly ok.



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but can be handled by global alignment algorithms when **all** available information is used.



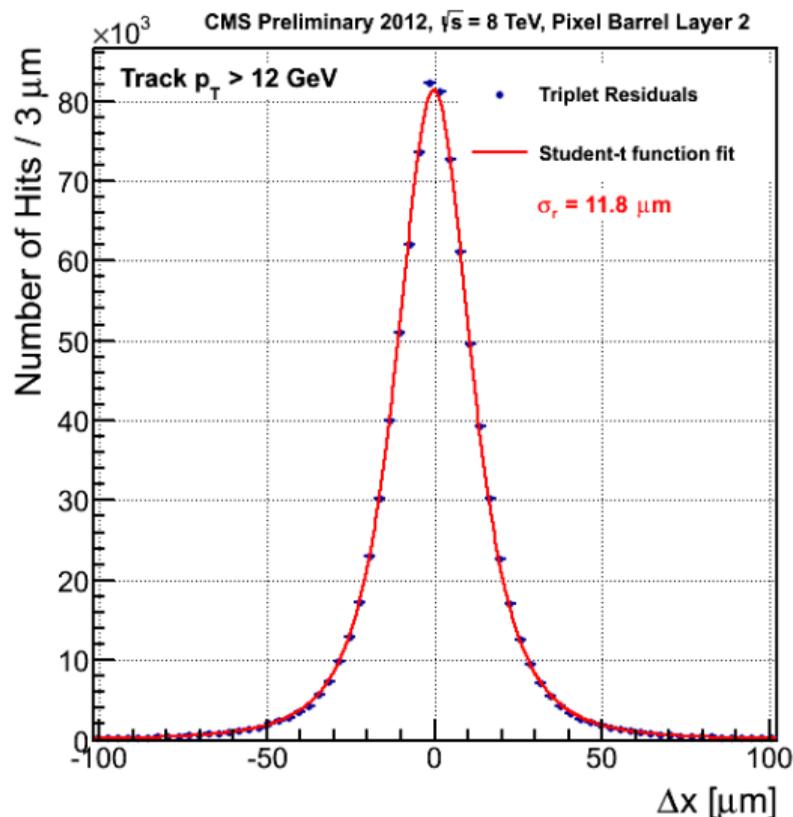
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Movements slower than the time to collect tracks for alignment are perfectly ok.
- ▶ **Weak modes** are a concern  
but can be handled by global alignment algorithms when **all** available information is used.
- ▶ **Non-standard events** are extremely helpful, e.g. resonances (CMS:  $\Upsilon, Z \rightarrow \mu\mu\beta_e$ )



# Achievable resolution



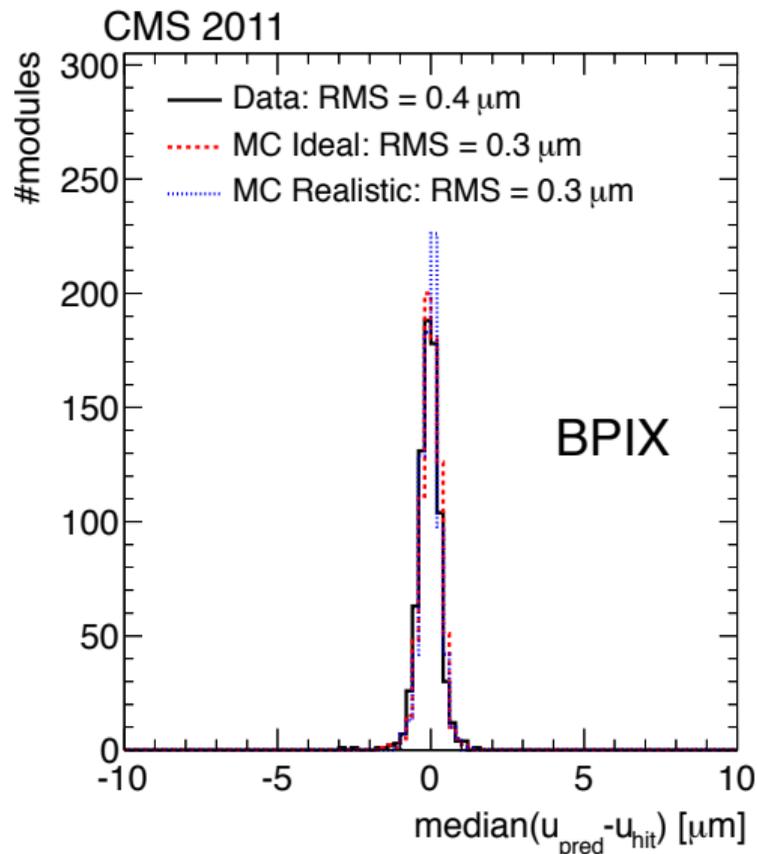
This shows the resolution measured for CMS pixel layer 2.

This demonstrates the power of pulse-height readout and make use of charge-sharing.

How precisely can you align this detector?



## Achievable resolution



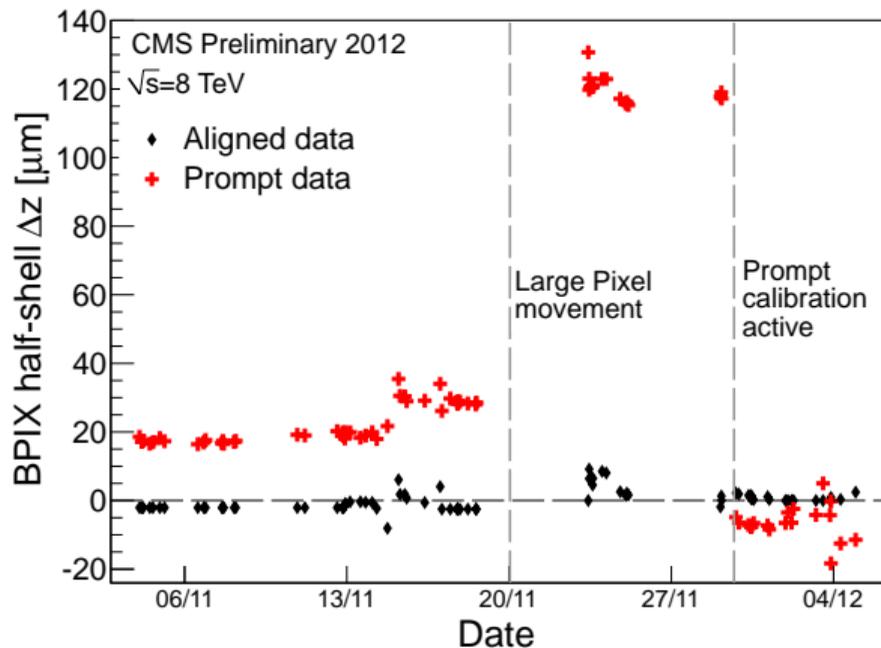
This is what you can reach.

We believed this to be a little overkill, but the heavy ion people loved this.

The precision comes from statistics and sound understanding of *all the details* of your detector.



## Example for prompt alignment



The CMS pixel barrels are not fixed, they are standing on rails.

Changes in temperature or magnetic field can trigger movements.

A few minutes of data is sufficient to align this (if triggers wouldn't be that prescaled, a few seconds would do...)



# Conclusions

- ▶ Introduction of an improved track model for use with the Millepede-II alignment algorithm in CMS lead to problems with Cosmics
- ▶ More complex models introduced and cover now:
  - ▶ bowed sensors
  - ▶ composite modules with two sensor surfaces
  - ▶ Lorentz angles
  - ▶ Backplane correction
- ▶ Quite some infrastructure and organization needed.



# Conclusions

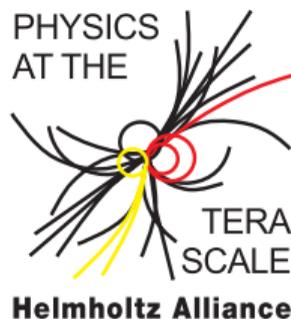
- ▶ Introduction of an improved track model for use with the Millepede-II alignment algorithm in CMS lead to problems with Cosmics
- ▶ More complex models introduced and cover now:
  - ▶ bowed sensors
  - ▶ composite modules with two sensor surfaces
  - ▶ Lorentz angles
  - ▶ Backplane correction
- ▶ Quite some infrastructure and organization needed.
- ▶ This has entered routine



# Acknowledgements

The success of CMS alignment is largely (!) based on work carried out at University of Hamburg and DESY (Volker Blobel, Claus Kleinwort, Gero Flucke, Rainer Mankel + students).

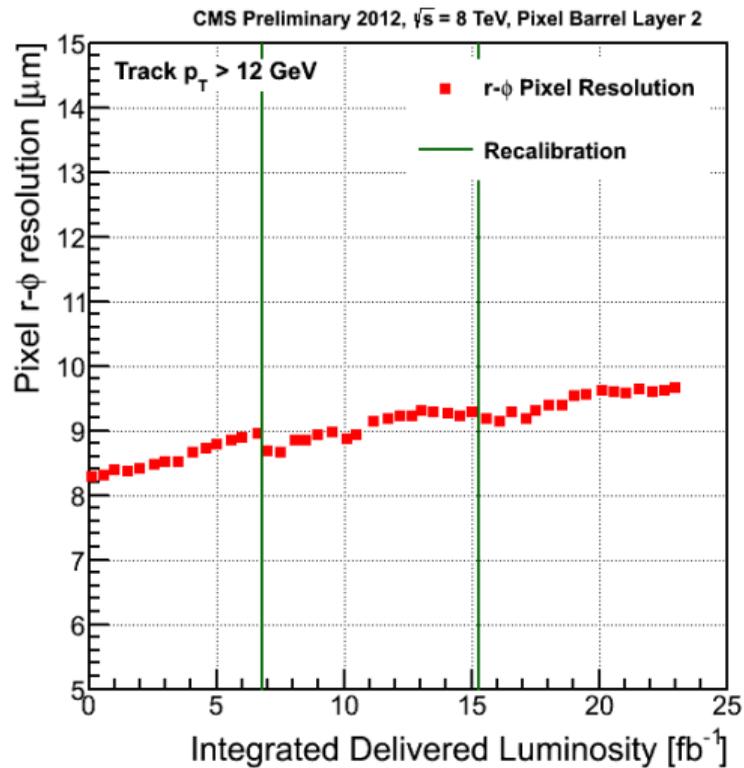
MP-II and GBL are supported by DESY/Terascale.



ENCORE



# Resolution degradation



## Backup: How the bowed surfaces are implemented

The bows were implemented using two-dimensional Legendre polynomials:

$$w(u, v) = \sum_{i=0}^N \sum_{j=0}^i c_{ij} L_j(u) L_{i-j}(v)$$

where

- ▶  $w(u, v)$  is the deviation from a plane at the origin in  $w$  direction as function of  $u, v$
- ▶  $N$  the maximal order of the Legendre polynomials.  
For  $N \rightarrow \infty$  every possible surface may be described.
- ▶  $c_{ij}$  coefficients
- ▶  $L_i(x)$  Legendre polynomial of  $i$ -th order



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Pro memoria:

- ▶ Legendre polynomials are orthogonal on  $x \in [-1, 1]$
- ▶ The first three Legendre polynomials are
  - ▶  $L_0(x) = 1$     $L_1(x) = x$     $L_2(x) = \frac{1}{2}(3x^2 - 1)$
- ▶  $N = 1$  is the same situation as in the past, just slopes instead of angles
- ▶  $N = 2$  allows for bends, the sagittae will be
  - ▶  $S_u = \frac{3}{2}c_{22}$     $S_{uv} = c_{21}$     $S_v = \frac{3}{2}c_{20}$



## Backup: Error estimate on parameters

How precise can we determine these bows? In principle, Millepede-II solves for  $\mathbf{x}$  like in<sup>3</sup>

$$\mathbf{M}\mathbf{x} = \mathbf{y}$$

Inversion of  $\mathbf{M}$  would give access to the covariance matrix, but inversion is of  $O(n^3)$  and  $n \approx 200\,000$ .



---

<sup>3</sup>see backup slides for more

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Observe that when solving for  $\mathbf{x}$  in

$$\mathbf{M}\mathbf{x} = \delta_i$$

where  $\delta_i$  is the Kronecker delta,  $\mathbf{x}$  will be the  $i$ -th column of  $\mathbf{M}^{-1}$ . Solving for  $\mathbf{x}$  is  $O(n^2)$ .



---

<sup>3</sup>see backup slides for more

## Backup: Error estimate on parameters

Calculation cost for one error:

- ▶ Basis: Alignment using 200 000 parameters (bows and composite modules)
- ▶ Memory footprint: 19 GB of RAM (grace to sparsity of the matrix)
- ▶ 12 minutes of wall-clock time (parallelized on 7 cores)
- ▶ 1.1 hrs of CPU time

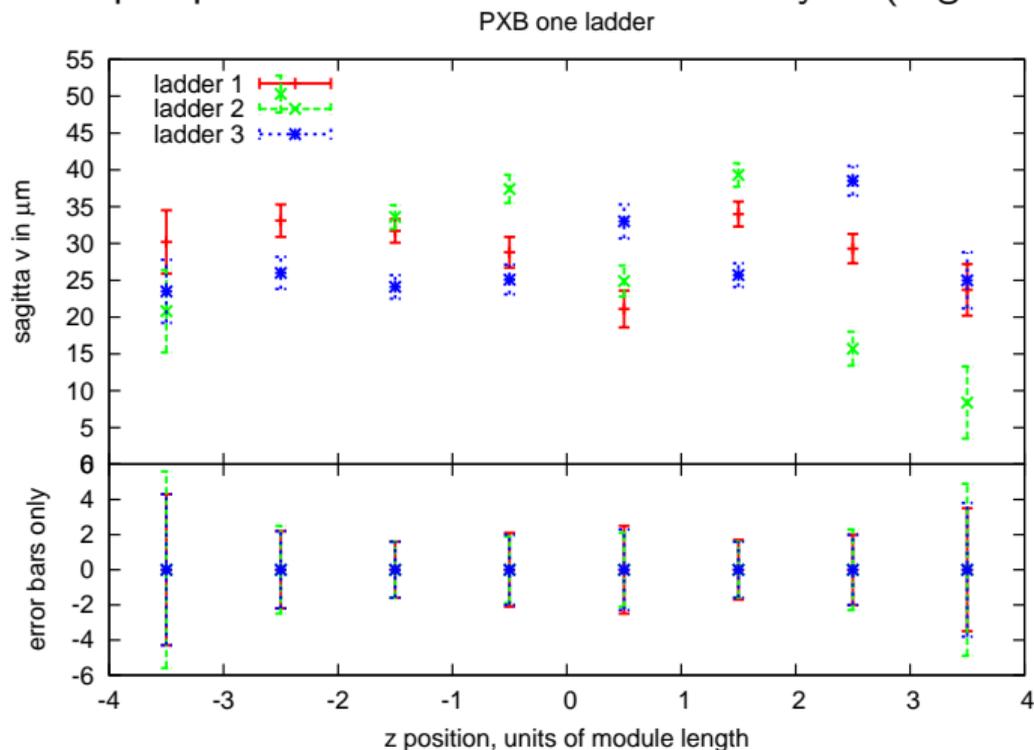
So determining the errors for all parameters would require more than 4 years. . .

Carried out on the Tier3 at PSI using one CPU with 8 cores (Intel Xeon Nehalem) and 24 GB of RAM.



# Backup: Error estimate on parameters

Example: pixel modules in the innermost layers (sagitta in  $v$ ):



## Backup: How Millepede works

Millepede-II is a general solver for *linear least squares* problems with a special structure typical for alignment problems.

The expression for the  $\chi^2$  to be minimized is

$$\chi^2(\mathbf{p}, \mathbf{q}) = \sum_j^{\text{tracks}} \sum_i^{\text{hits}} \mathbf{r}_{ij}^T(\mathbf{p}, \mathbf{q}_j) \mathbf{V}_{ij}^{-1} \mathbf{r}_{ij}(\mathbf{p}, \mathbf{q}_j) \quad (3)$$

where  $\mathbf{p}$  denotes the alignment parameters describing the actual geometry and  $\mathbf{q}_j$  denotes the track parameters of the  $j^{\text{th}}$  track.

Allow for the following identification:

- ▶ Alignment parameters ( $\mathbf{p}$ )  $\mapsto$  global parameters
- ▶ Track parameters ( $\mathbf{q}_j$ )  $\mapsto$  local parameters



## Backup: How Millepede works

Nonlinear dependencies require local linearization:

$$\chi^2(\mathbf{p}, \mathbf{q}) = \sum_j^{\text{tracks}} \sum_i^{\text{hits}} \frac{1}{\sigma_{ij}^2} \left( \mathbf{m}_{ij} - \mathbf{f}_{ij}(\mathbf{p}_0, \mathbf{q}_{j0}) - \frac{\partial \mathbf{f}_{ij}}{\partial \mathbf{p}} \Delta \mathbf{p} - \frac{\partial \mathbf{f}_{ij}}{\partial \mathbf{q}_j} \Delta \mathbf{q}_j \right)^2 \quad (4)$$

Here,  $\mathbf{f}_{ij}$  is the hit position predicted by the track model from track reconstruction and  $\mathbf{m}_{ij}$  is the measured hit position. Assuming uncorrelated measurements allows to replace the inverse covariance matrix by  $\frac{1}{\sigma_{ij}^2}$ .

To minimize this expression one does the obvious: rewrite it as a matrix expression and then differentiate and you will get corrections to your parameters as a vector  $\Delta \mathbf{a}$ .



## Backup: How Millepede works

Doing this ends up with<sup>4</sup>

$$\Delta \mathbf{a} = (\mathbf{A}^T \mathbf{W} \mathbf{A})^{-1} \mathbf{A}^T \mathbf{W} \mathbf{r} \quad (5)$$

where  $\Delta \mathbf{a}$  are the estimated correction of parameters,  $\mathbf{A}$  the Jacobian,  $\mathbf{W}$  the inverse covariance matrix of the measurements and  $\mathbf{r}$  the residuals.

The estimate of the covariance matrix of the *parameters*  $\mathbf{V}[\Delta \mathbf{a}]$  is then

$$\mathbf{V}[\Delta \mathbf{a}] = (\mathbf{A}^T \mathbf{W} \mathbf{A})^{-1} \quad (6)$$

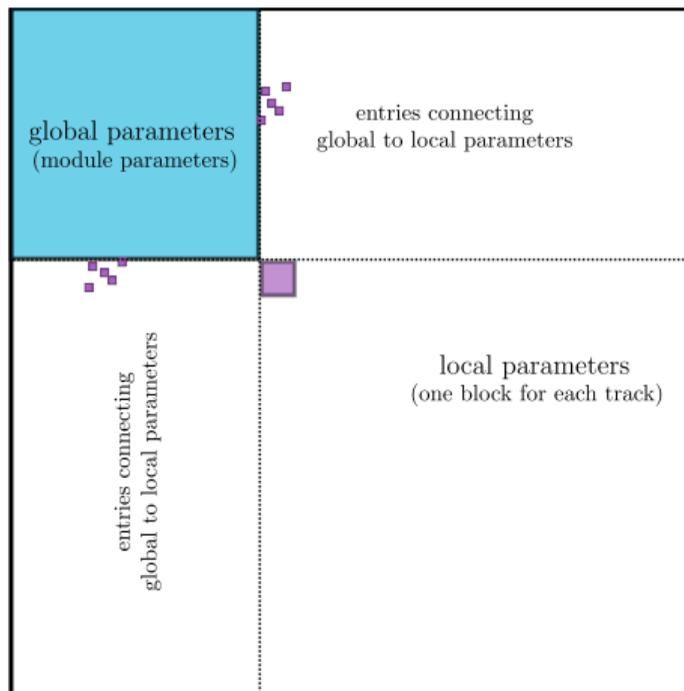
If inversion is feasible, the errors are available.



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<sup>4</sup>following the notation in Blobel/Lohrmann

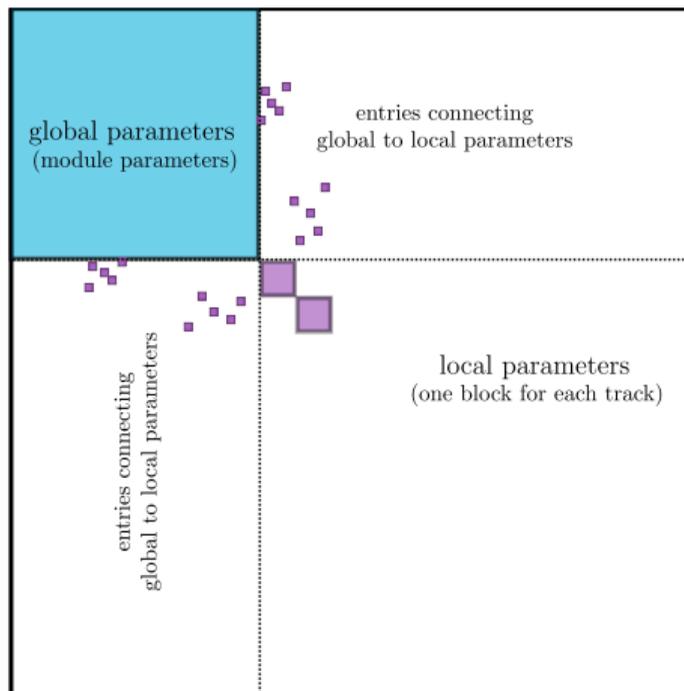
## Backup: How Millepede works



Arrange the matrix (and the vectors as well, but not shown here) in the following way: Put all global parameters at the begin, followed by the local parameters.



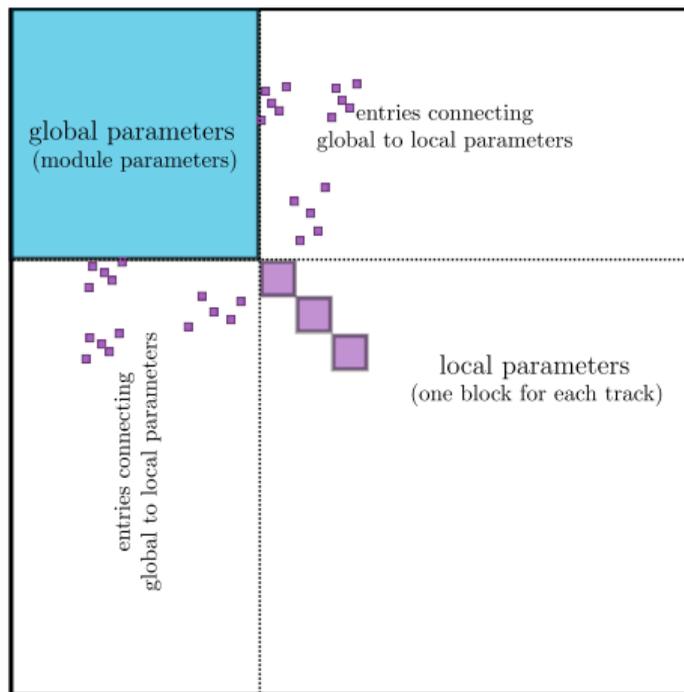
## Backup: How Millepede works



For each track a new block will be added. Entries in the global block are updated.



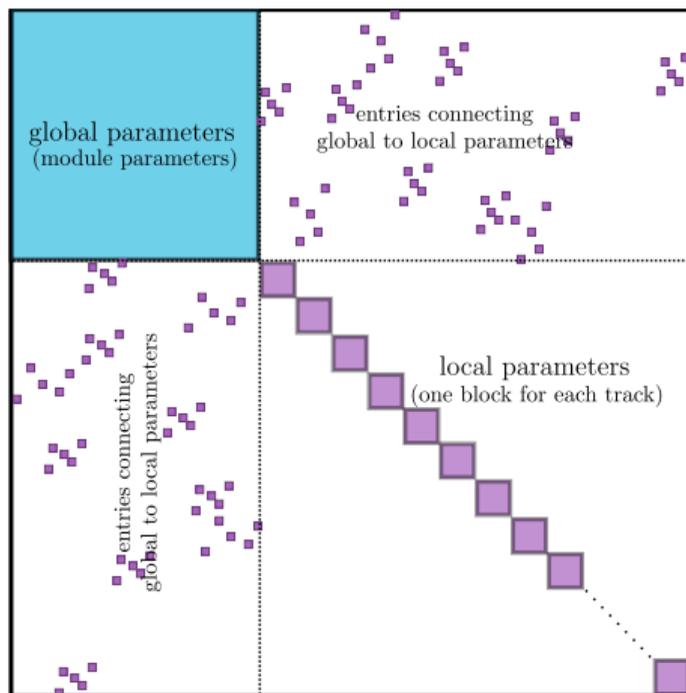
## Backup: How Millepede works



The entries for the local block are connected to the global parameters via the band parts of the matrix



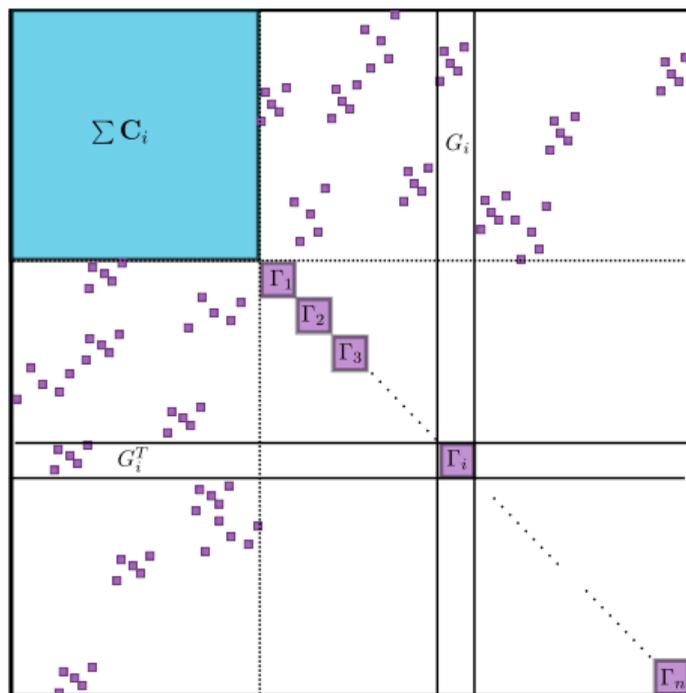
## Backup: How Millepede works



The pattern starts to appear:  
Each track contributes to the global and local parameters.  
The entries of the local parameters connect to the global parameters in the band outside



## Backup: How Millepede works



More formally the matrix consists of the following parts:

- ▶  $\Sigma \mathbf{C}_i$ : the block containing the sum of the contributions to the global parameters
- ▶  $\Gamma_i$ : small blocks for each track, local parameters, disjoint between measurements
- ▶  $G_i$ : band matrix connecting the local parameters of track  $i$  with the global parameters hit by the track



## Backup: How Millepede works

So you end up with this equation

$$\left( \begin{array}{c|ccc} \sum \mathbf{C}_i & \cdots & \mathbf{G}_i & \cdots \\ \hline \vdots & \ddots & \mathbf{0} & \mathbf{0} \\ \mathbf{G}_i^T & \mathbf{0} & \mathbf{\Gamma}_i & \mathbf{0} \\ \vdots & \mathbf{0} & \mathbf{0} & \ddots \end{array} \right) \cdot \begin{pmatrix} \mathbf{a} \\ \vdots \\ \alpha_i \\ \vdots \end{pmatrix} = \begin{pmatrix} \sum \mathbf{b}_i \\ \vdots \\ \beta_i \\ \vdots \end{pmatrix}$$

where the matrix has size

$$N = N_{\text{parameters}} + N_{\text{tracks}} \cdot N_{\text{track parameters}}$$

In alignment, you want to solve for the global parameters only, so there might be a possibility to reduce the problem to

$$N_{\text{parameters}}$$



## Backup: How Millepede works

Using some block matrix theorems (partitioning formulas for calculation of inverse matrices) you can reduce this problem to

$$\begin{pmatrix} \mathbf{c}' \end{pmatrix} \cdot \begin{pmatrix} \mathbf{a} \end{pmatrix} = \begin{pmatrix} \mathbf{b}' \end{pmatrix}$$

where a new matrix and a new vector of size  $N_{\text{parameters}}$  are used:

$$\mathbf{c}' = \sum \mathbf{c}_i - \sum \mathbf{G}_i \mathbf{\Gamma}_i^{-1} \mathbf{G}_i^T \quad \mathbf{b}' = \sum \mathbf{b}_i - \sum \mathbf{G}_i (\mathbf{\Gamma}_i^{-1} \beta_i)$$

where  $\mathbf{\Gamma}_i^{-1}$  is small and therefore can be calculated in reasonable time. Even though  $i$  might be large, the cost for solving the reduced equation is

$$N_{\text{pars}}^2 + N_{\text{tracks}} \cdot N_{\text{track pars}}^2 \ll (N_{\text{pars}} + N_{\text{tracks}} \cdot N_{\text{track pars}})^2$$



## Backup: How Millepede works

If you don't believe the impact, let us calculate:

$$N_{\text{pars}}^2 + N_{\text{tracks}} \cdot N_{\text{track pars}}^2 \ll (N_{\text{pars}} + N_{\text{tracks}} \cdot N_{\text{track pars}})^2$$

using the following typical values for an alignment in CMS:

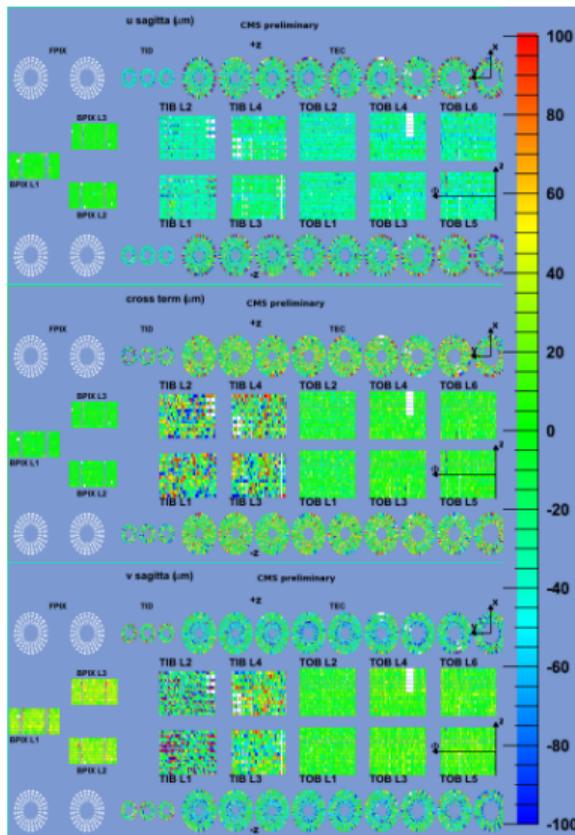
- ▶  $N_{\text{pars}} = 200\,000$
- ▶  $N_{\text{tracks}} = 10^7$
- ▶  $N_{\text{track pars}} = 30$

this gives

$$5 \cdot 10^{10} \approx 4 \cdot 10^{10} + 10^7 \cdot 9 \cdot 10^2 \ll (2 \cdot 10^5 + 10^7 \cdot 30)^2 \approx 10^{17}$$



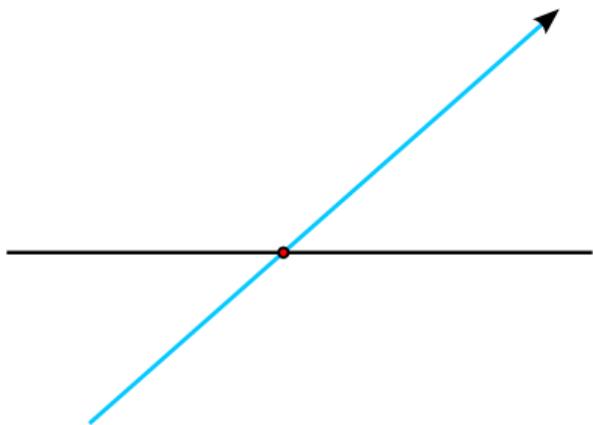
# Results



This is a rough overview of the bows determined using alignment.  
Shown are the sagittae in  $\mu\text{m}$ .



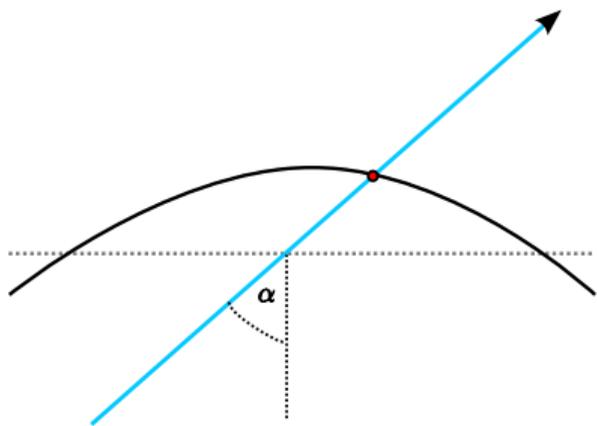
## What is the impact?



In a flat sensor, the hit is where it is.



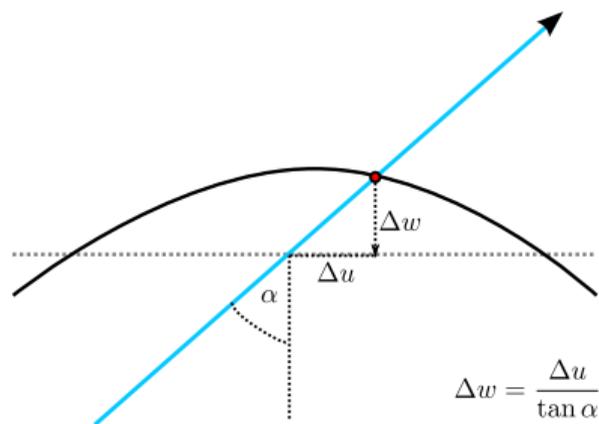
## What is the impact?



As soon as there is a bow, the track angle  $\alpha$  determines the shift due to a deviation from the anticipated position. For large  $\alpha$  typically observed in cosmic tracks, this gets substantial.



## What is the impact?

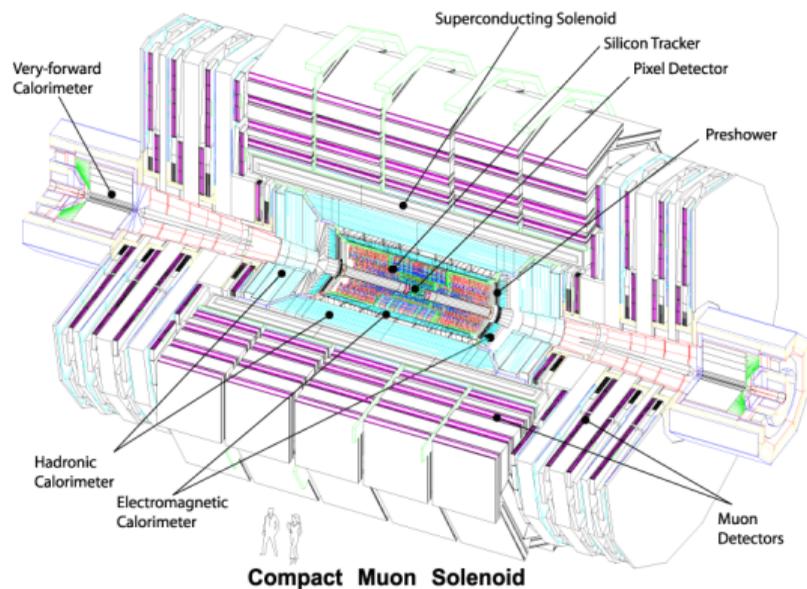


As soon as there is a bow, the track angle  $\alpha$  determines the shift due to a deviation from the anticipated position. For large  $\alpha$  typically observed in cosmic tracks, this gets substantial.

- ▶ The intrinsic resolution is as low as  $15 \mu\text{m}$  (pixel detector).
- ▶ If  $\Delta w$  is of the same order, there will be a systematic error in the tracking.
- ▶ **Worst case:** sagitta of  $30 \mu\text{m}$ ,  $\eta = 2.5 \Rightarrow \tan \alpha = 6$  (far end of pixel barrel, track from interaction point):  
 $du = 30 \mu\text{m} \cdot 6 = 180 \mu\text{m}$



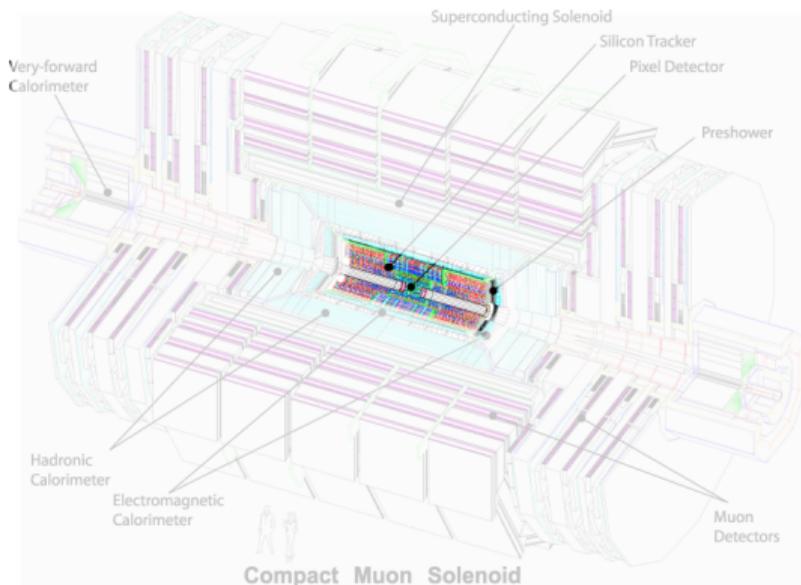
# A brief detour: The CMS tracker



CMS is one of the two multi purpose detector at CERN's Large Hadron Collider (LHC)



## A brief detour: The CMS tracker

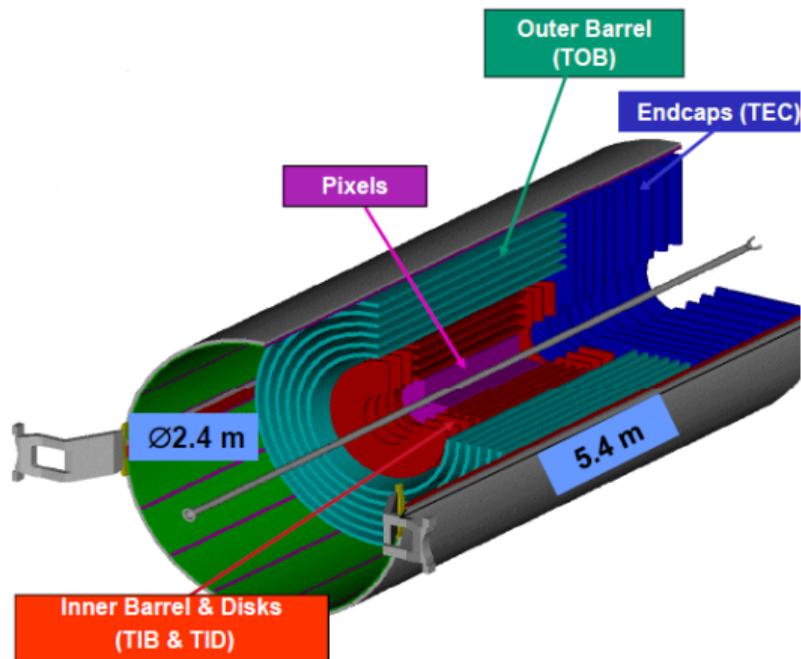


The silicon tracker is in the heart of CMS. It consists of

- ▶ 1440 silicon pixel modules
- ▶ 15 148 silicon strip modules



## A brief detour: The CMS tracker

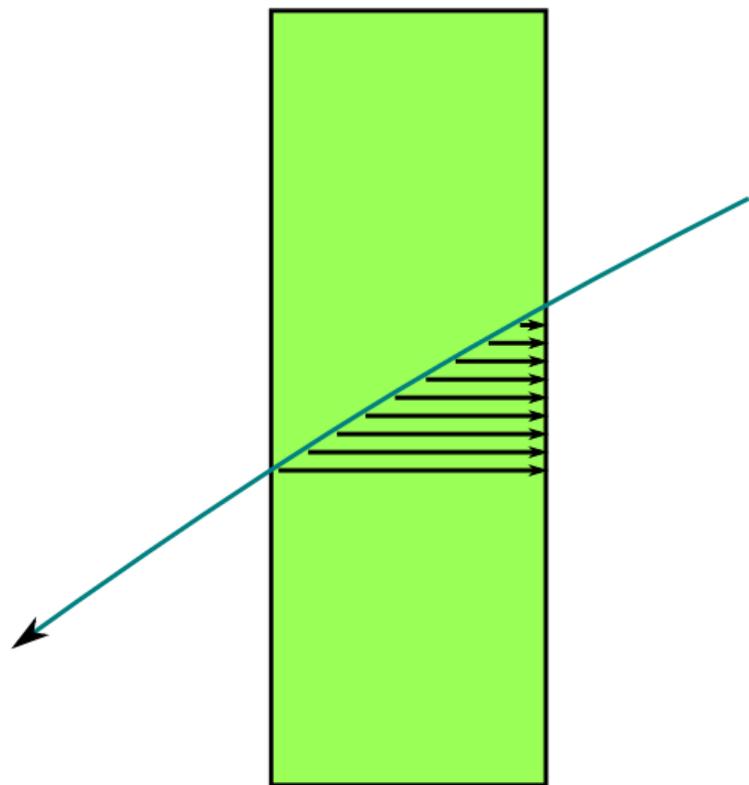


This is a closer view of the tracker. Observe the layerwise structure in the barrel and endcap.

Image courtesy Steven Lowette



## More parameters

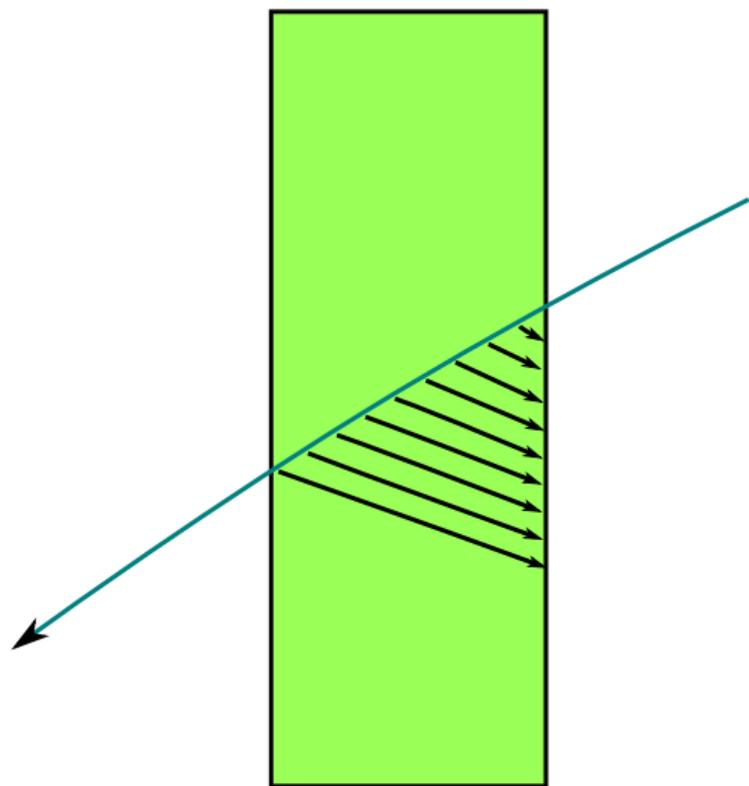


A sensor (pixel, strip) collects charge from an ionizing particle traversing the active volume.

Without any B-field, the charge propagates along the E-field and gets collected.



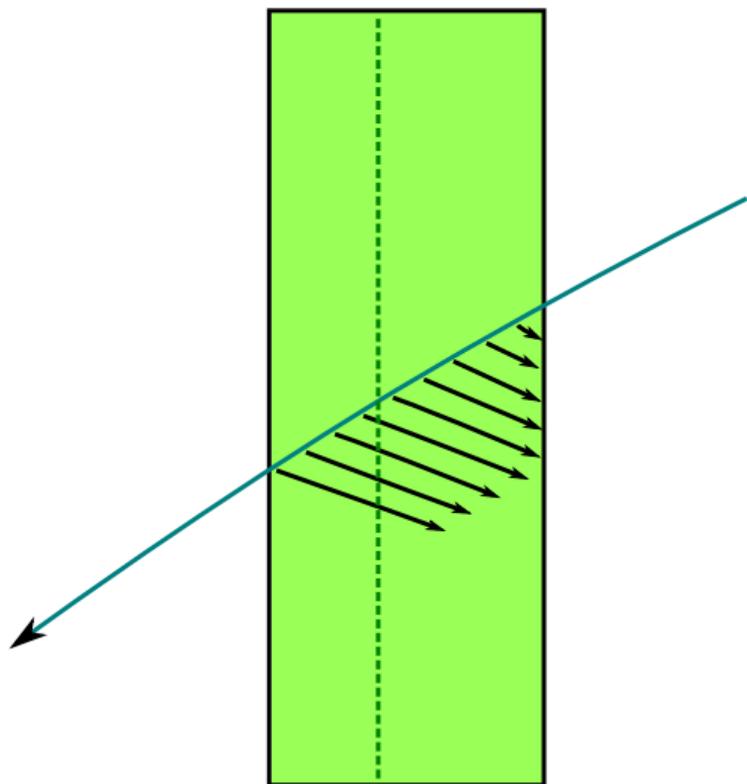
## More parameters



With B-field, another component is added to the drift. Usually described by a so called Lorentz angle (LA).



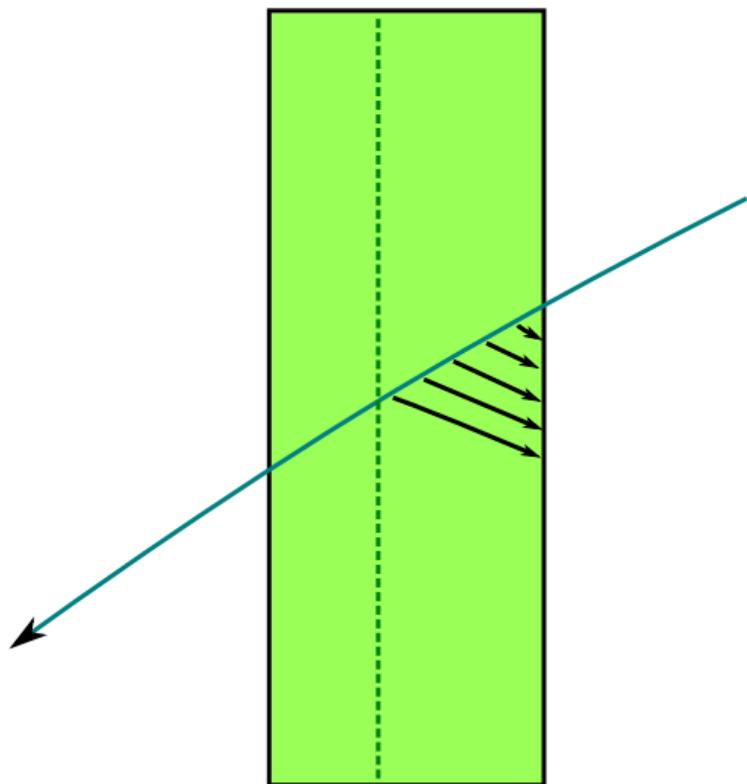
## More parameters



In case of an integration time shorter than the drift time, not all charge reaches the read-out amplifier.



## More parameters



This can be described by a (hypothetical) backplane inside the detector.

This is the case for strip sensors in CMS (when run in deconvolution mode).



# Computing challenges

What are the challenges in aligning the CMS tracker?

- ▶ Prompt calibration loop



# Computing challenges

What are the challenges in aligning the CMS tracker?

- ▶ Prompt calibration loop
  - ▶ Runs during datataking
  - ▶ Every about 10 000 good tracks a simple alignment is started
  - ▶ If movements are above threshold: check with larger sample, update conditions DB if manifest
  - ▶ Reason: Pixel barrel movements



# Computing challenges

What are the challenges in aligning the CMS tracker?

- ▶ Prompt calibration loop
- ▶ Full alignment



# Computing challenges

What are the challenges in aligning the CMS tracker?

- ▶ Prompt calibration loop
- ▶ Full alignment
  - ▶ Align for the full detector (16k objects)
  - ▶ 6 position parameters (3 coordinates and 3 angles)
  - ▶ 3 bow parameters
  - ▶ kink parameters for composite modules
  - ▶ end up with 200k parameters
  - ▶ requires 32 MB of memory



# Computing challenges

What are the challenges in aligning the CMS tracker?

- ▶ Prompt calibration loop
- ▶ Full alignment
- ▶ Lorentz angle and BP alignment



# Computing challenges

What are the challenges in aligning the CMS tracker?

- ▶ Prompt calibration loop
- ▶ Full alignment
- ▶ Lorentz angle and BP alignment
  - ▶ Even more parameters
  - ▶ Calls for even bigger machines (was done at DESY last year, now we have 256 MB machines at Cern)



## Computing challenges

200 000 parameters  $\rightarrow$  200 000  $\times$  200 000 matrix =  $40 \cdot 10^{12}$  numbers... How to handle this?

- ▶ Matrix is sparse (about 30% of off-diagonal elements are  $\neq 0$ )
- ▶ Apply compression
- ▶ Parallelize task using OpenMP
- ▶ Equation system solved using MINRES and not inversion
- ▶ Make use of recent hardware

One example: Full alignment using 22 Mio tracks runs for about 10 wallclock hours on an 8 core CPU with 48 GB of RAM memory

