



ICP based Software Alignment at Lumi

November 2018 Progress Report

Roman Klasen | roklasen@uni-mainz.de



Software Tasks

Misalignment Determination

Determination of component positions with:

- measurement data
- cosmic rays
- etc.

This is done with PndLmdSensorAligner.

Software Alignment

Use align matrices for particle reconstruction

This is done with the new FairRoot AlignManager class presented in the computing session.



Software Alignment

We distinguish two related but very different concepts:

During Simulation / Pre-Experiment

Generate mc tracks (or similar) using a “wrong” geometry just like a real detector would produce. The tracks will be off w.r.t. to their “real” position. Use this to study how your analysis software handles a realistic, misaligned geometry. This can be done two different ways, see slides 8 and 9.

For Real Measurement Data / During Experiment

Once built, use the alignment parameters obtained from survey etc on *real measurement data*. This accounts for misaligned detector parts and produces reconstructed tracks that are closer to the real tracks than without alignment. ***This is the main goal of software alignment.***



Shift Detector vs. Shift Data

Shift Detector

- Realistic Detector Acceptance
- Realistic scenario for Track Finder, Fitter etc.
- Reco Macro need to only load Geometry from TGeoManager (which handles Align.)
- But need to generate MC Data again (esp. If you want multiple misaligned geometries)

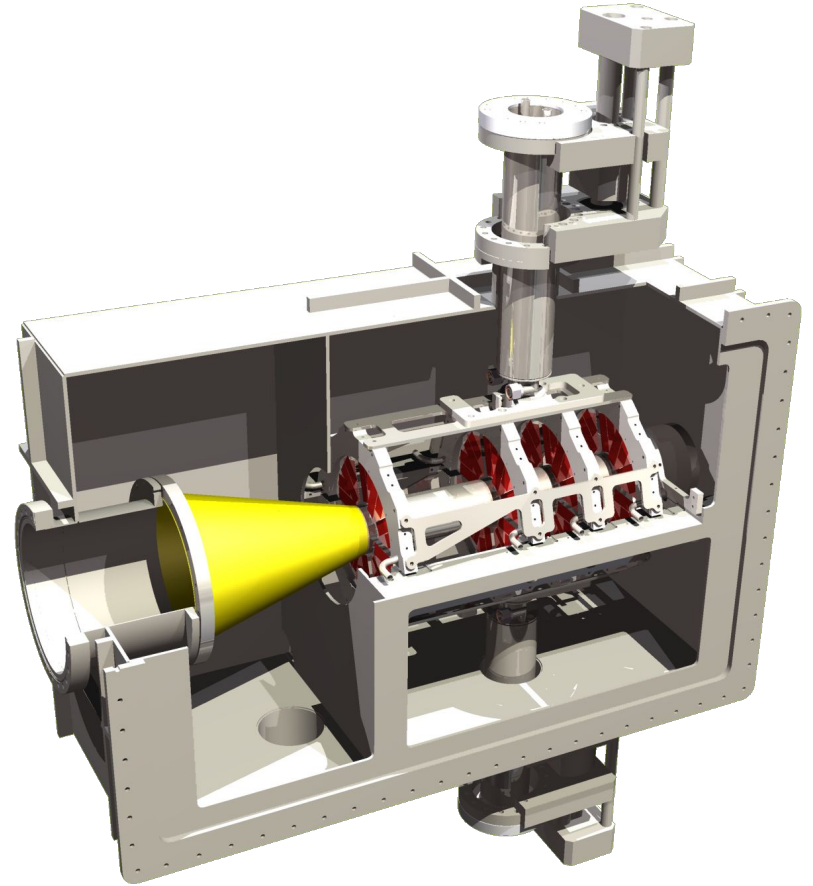
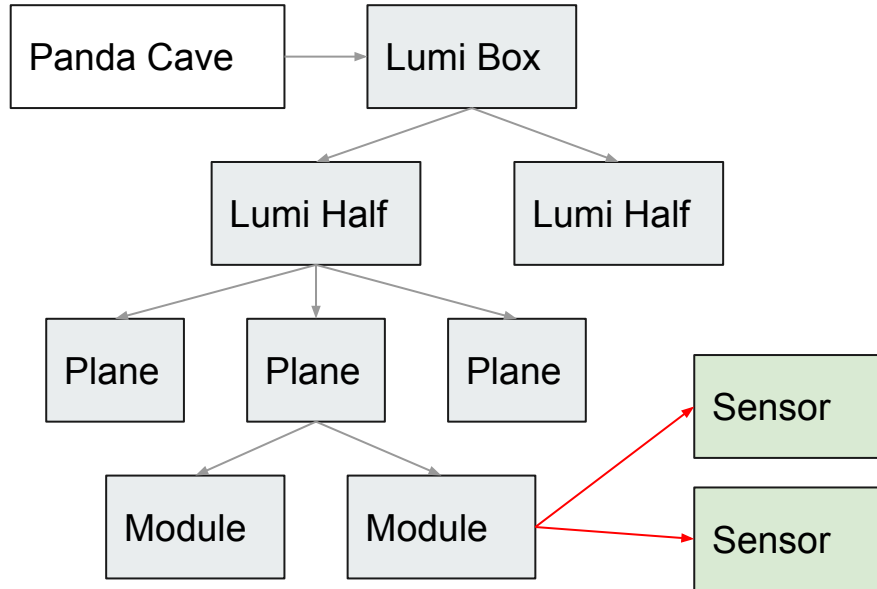
Shift Data

- Can use existing MC data
- Wrong detector acceptance may lead to implausible tracks:
- Don't see some tracks that should be there
- See tracks that can't be there
- Reco Macros must account for Misalignment

I'll be using both methods in this talk.

Lumi Example

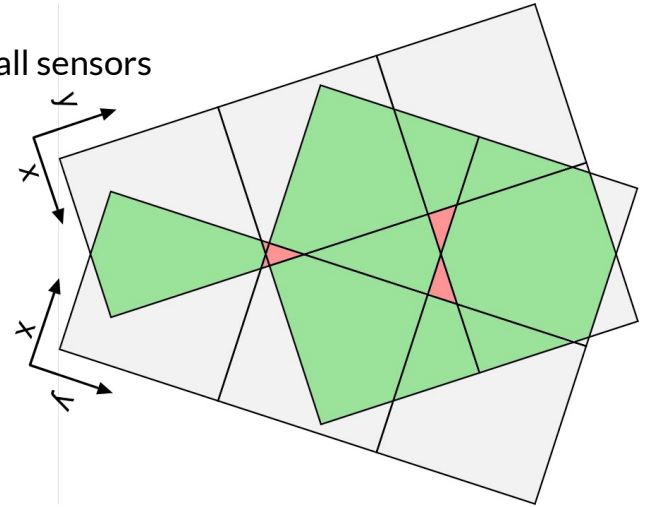
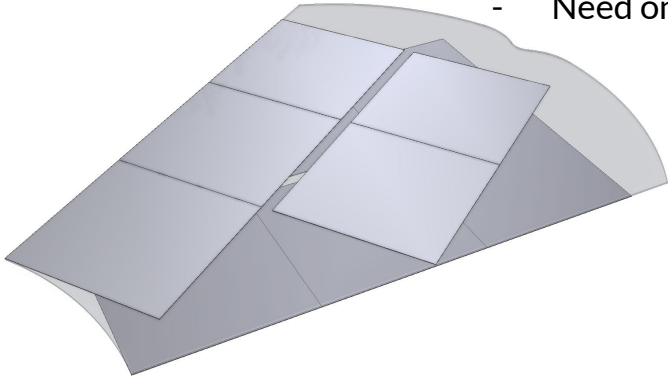
Alignable Components





Sensor Alignment with overlapping Areas

- Use overlaps to get matrix $s_1 \rightarrow s_2$
- Go from sensor to sensor to reach all sensors
- Need one sensor as reference





Point-cloud based Alignment with ICP

We treat the hits on the front and back sensors as point clouds.

After filtering and selection, two clouds with N elements remain, each point in cloud A corresponds to a point in cloud B.

The transformation from cloud A to cloud B is called M , and it's the transformation matrix from sensor A to sensor B.

We use an iterative closest point algorithm that finds the optimum transformation matrix.

Differences to the design matrices can stem from:

- Detector resolution
- Wrong pair filtering
- Wrong particle (or their angles an entry)



Software Parameters

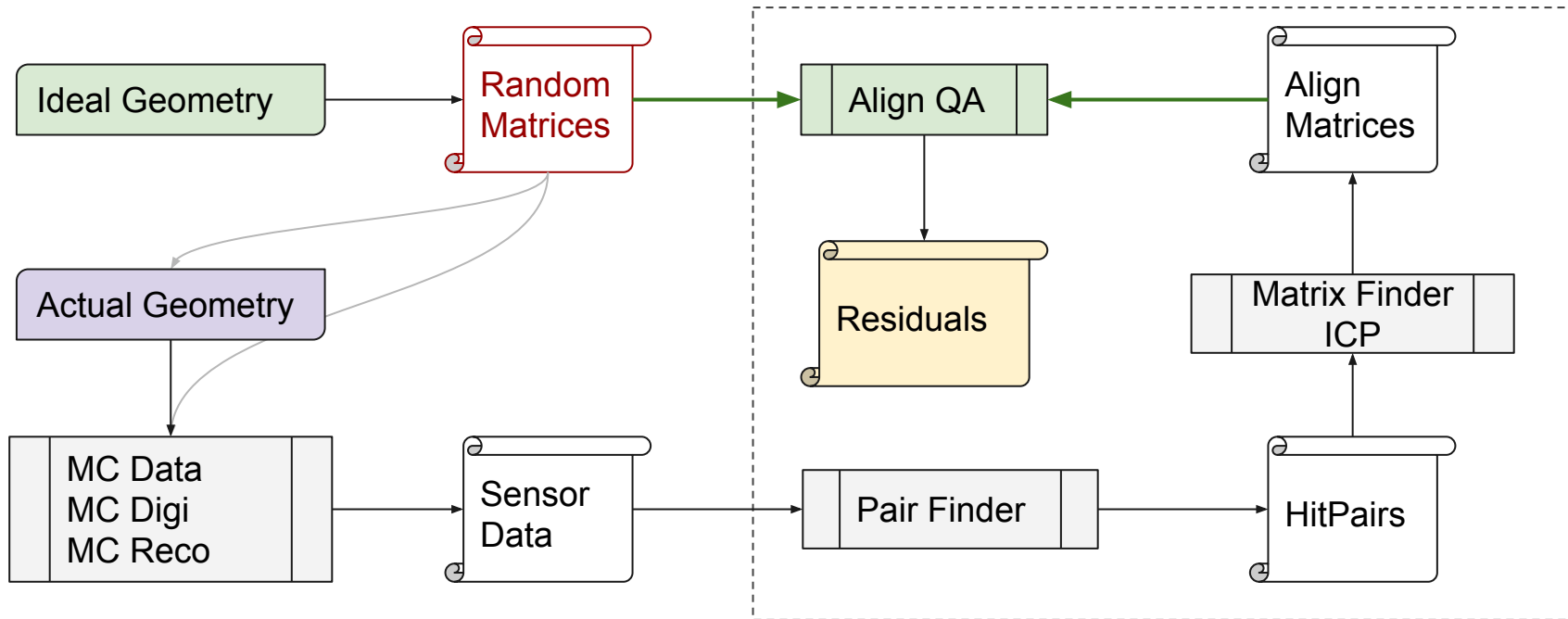
- We used multiple Geometries
- We Misalign Sensors only
- We allow XY Shift and z-Rotation only
- Enough data for $\sim 10^5$ pairs/area

We can reach all sensors just by stepping from sensor to sensor by their overlapping areas

We compare the found alignment matrices with the ones provided to the simulation

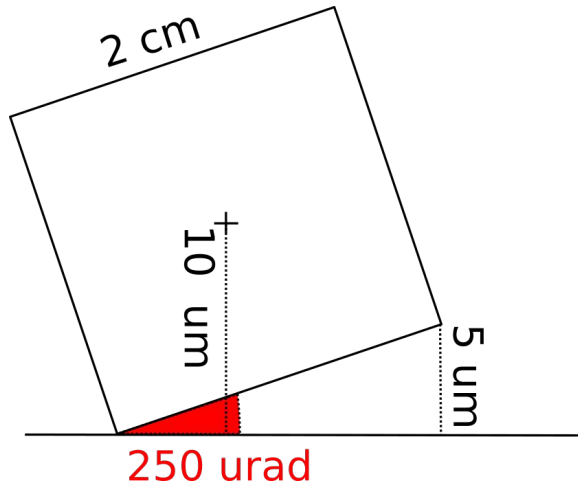
Estimate Quality of Alignment

PndLmdSensorAligner



Multiple Sets of random Matrices

Use a fixed relation shift \rightarrow rotation



We'll use these shorthands

- 0μ : perfect geometry
- 10μ : $10\mu\text{m}$ shift, $250\mu\text{rad}$ rot
- 50μ : $50\mu\text{m}$ shift, 1.25mrad rot
- 100μ : $100\mu\text{m}$ shift, 2.5 mrad rot
- 200μ : $200\mu\text{m}$ shift, 5.0 mrad rot
- 250μ : $250\mu\text{m}$ shift, 7.5 mrad rot



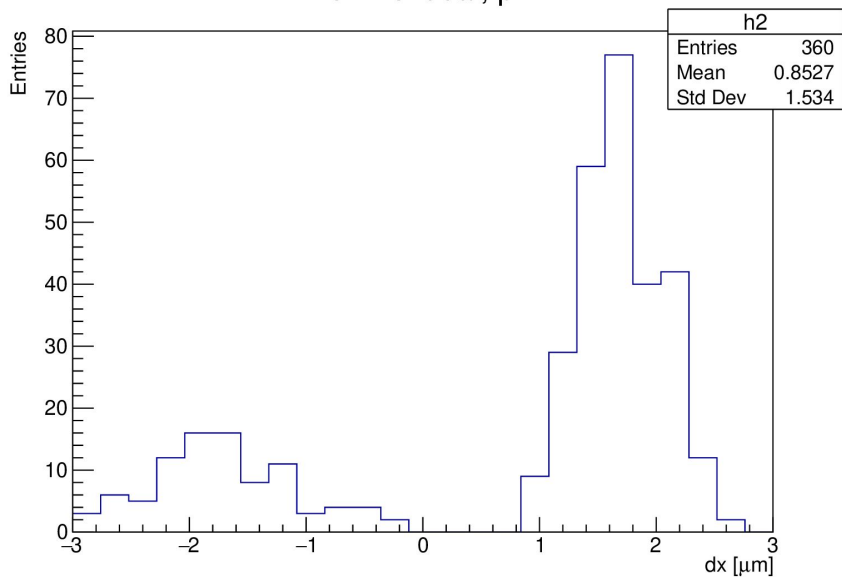
Comparison: Individual Sensor Matrices

For every overlapping area we determine the overlap matrix and compare it to the design misalignment matrix

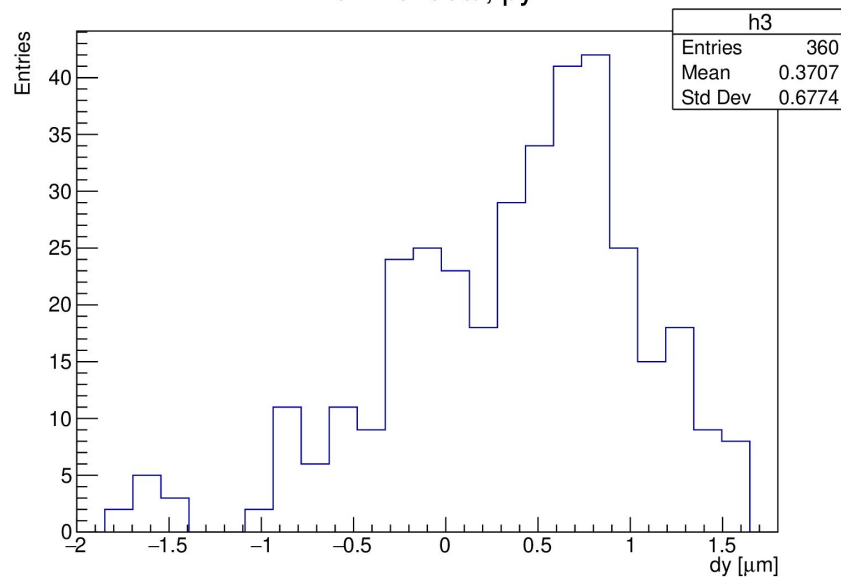


0μ - perfect geometry

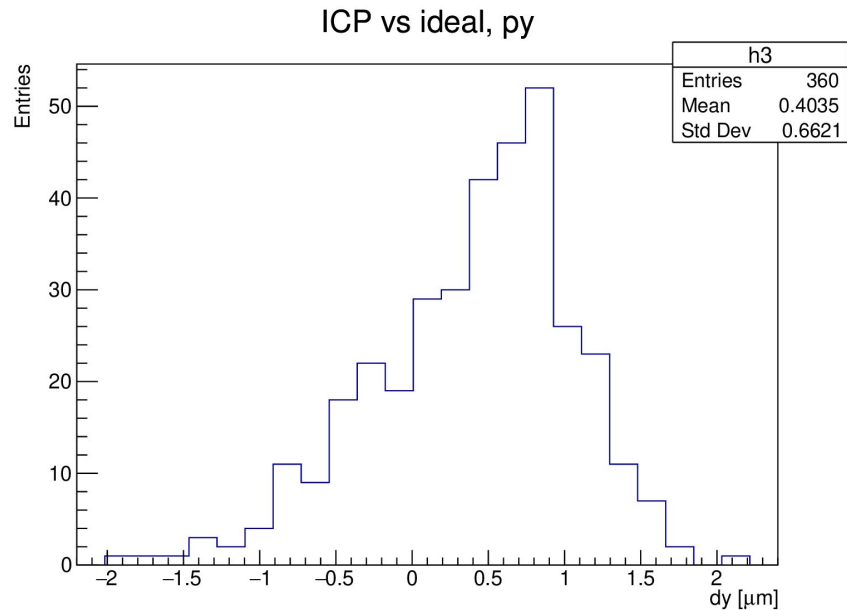
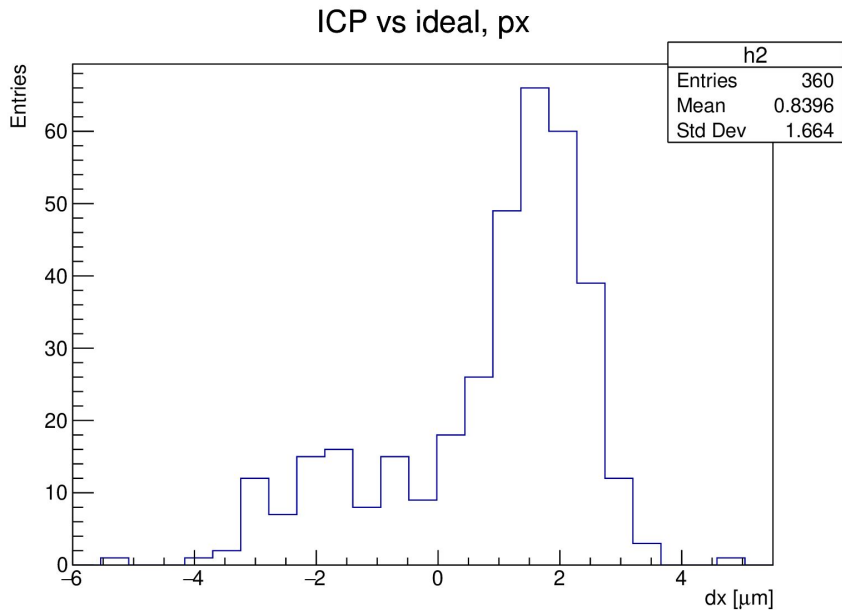
ICP vs ideal, px



ICP vs ideal, py

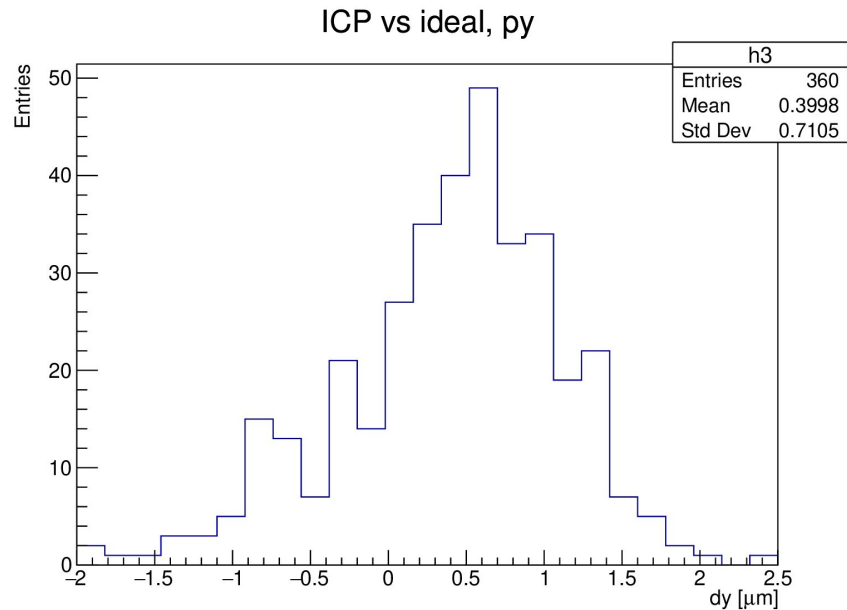
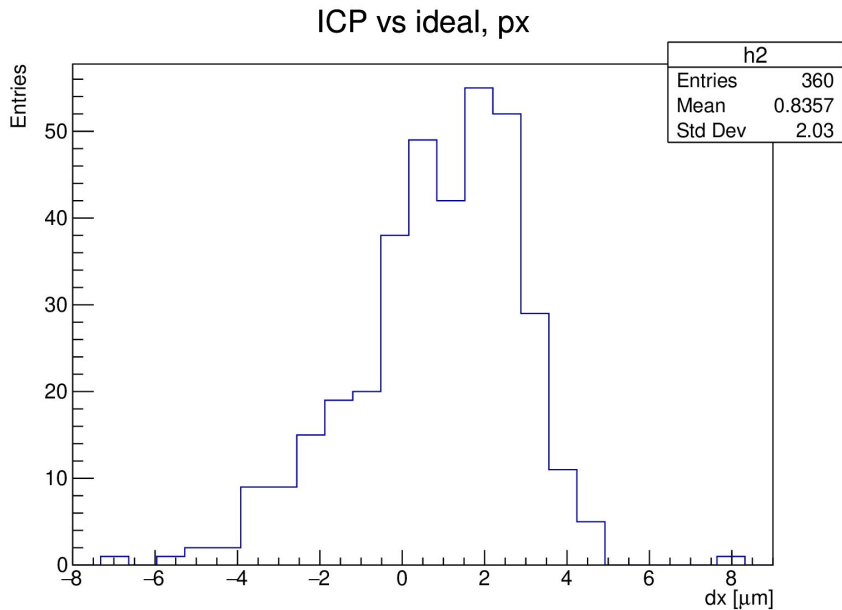


100 μ - misaligned geometry





200 μ - misaligned geometry





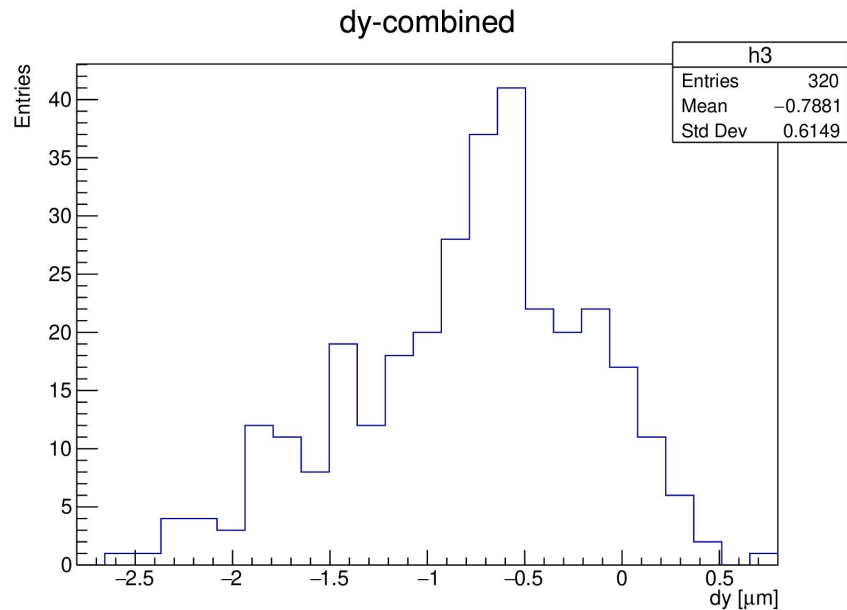
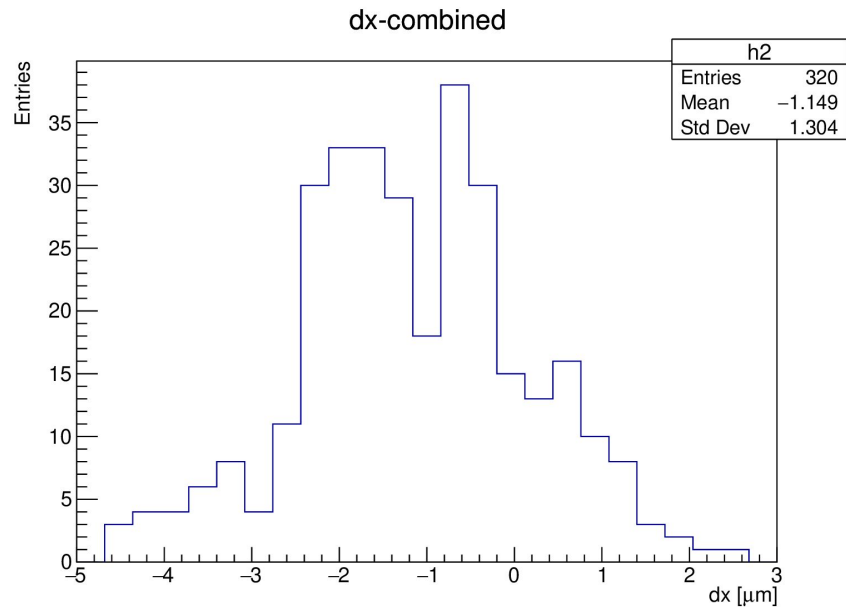
Comparison: Combined Matrices

For all modules, we calculate the resultant matrices from our reference sensor to all other sensors on that module.

That means there are fewer combined matrices than individual matrices, and each combined matrix consists of several individual matrices.

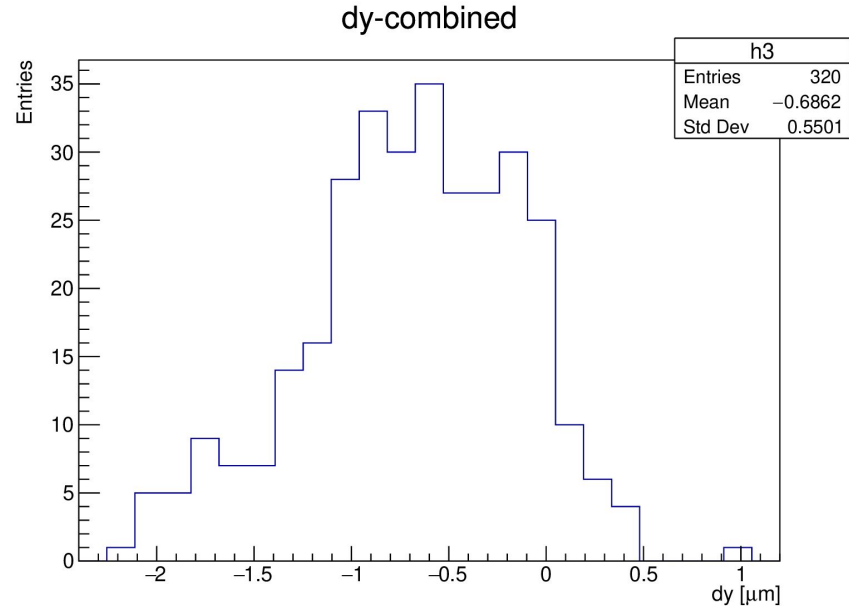
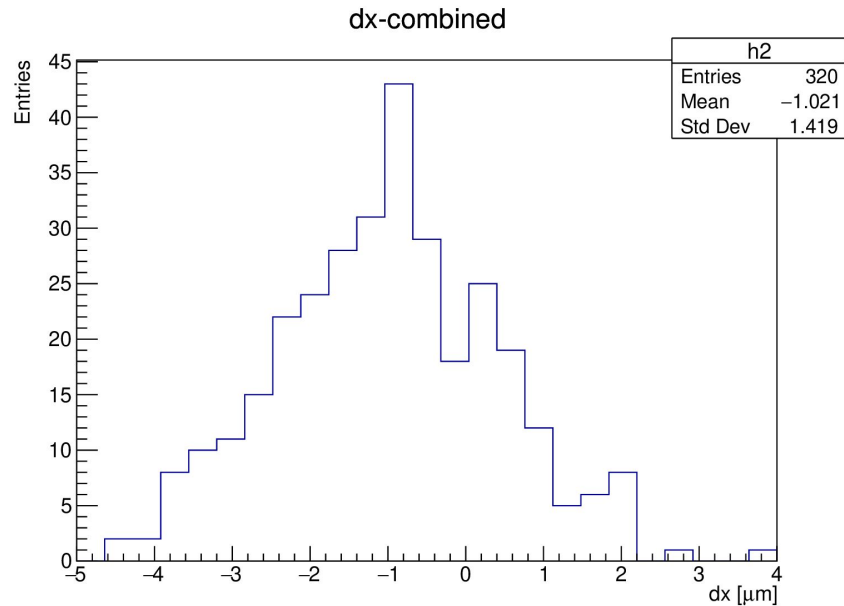


0μ - perfect geometry

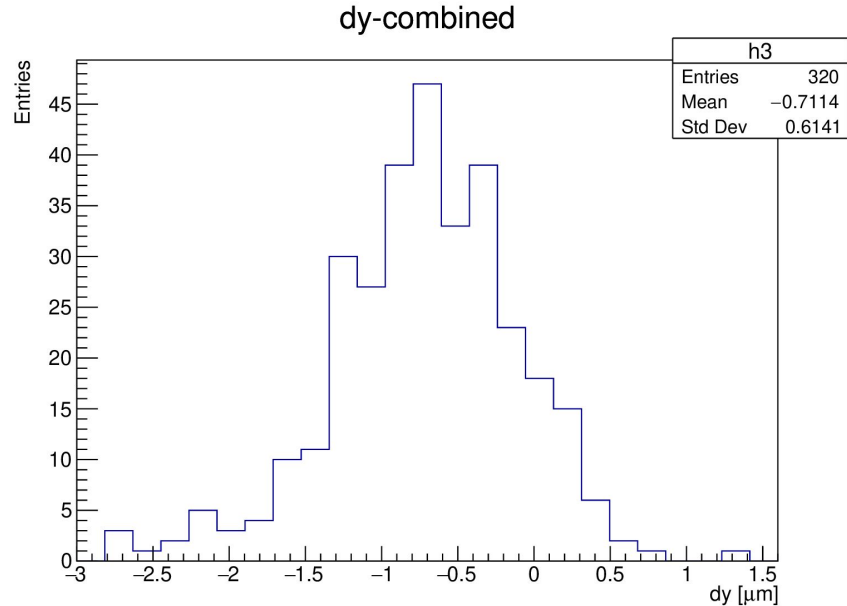
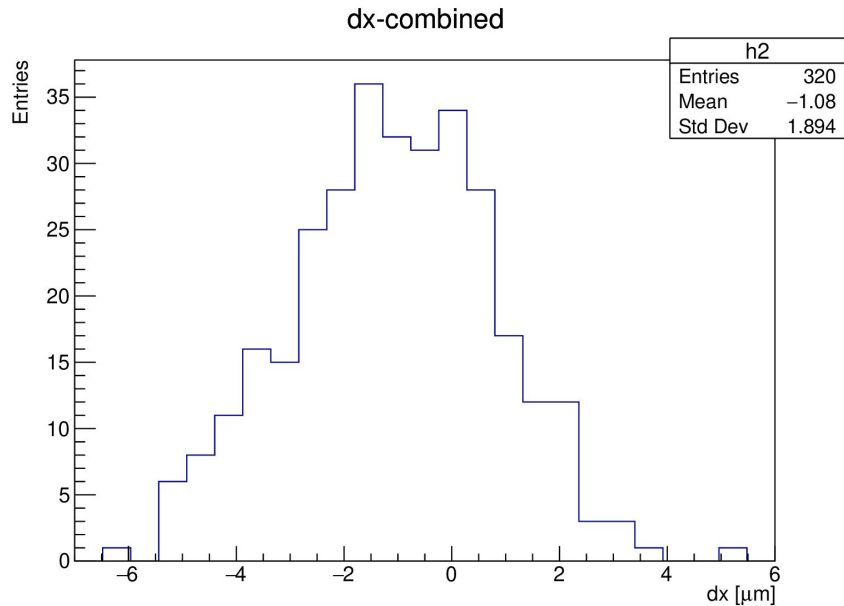




100 μ - misaligned geometry



200 μ - misaligned geometry





Comparison: Shift Data vs. Shift Geometry

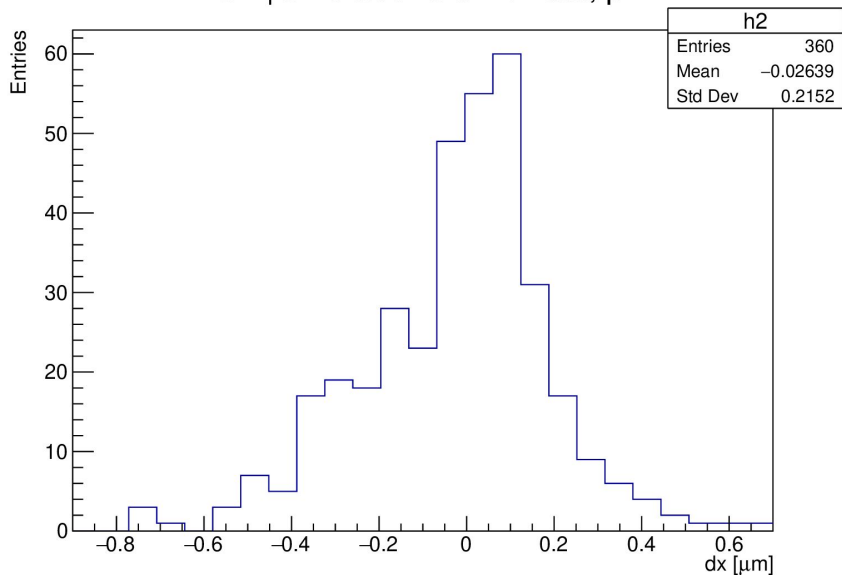
For the single case of 100u misalignment, I compare the resultant individual overlap matrices obtained with shifted geometry with the matrices from shift data.

Due to time constraints, I will not show other misalignments or combined matrices.

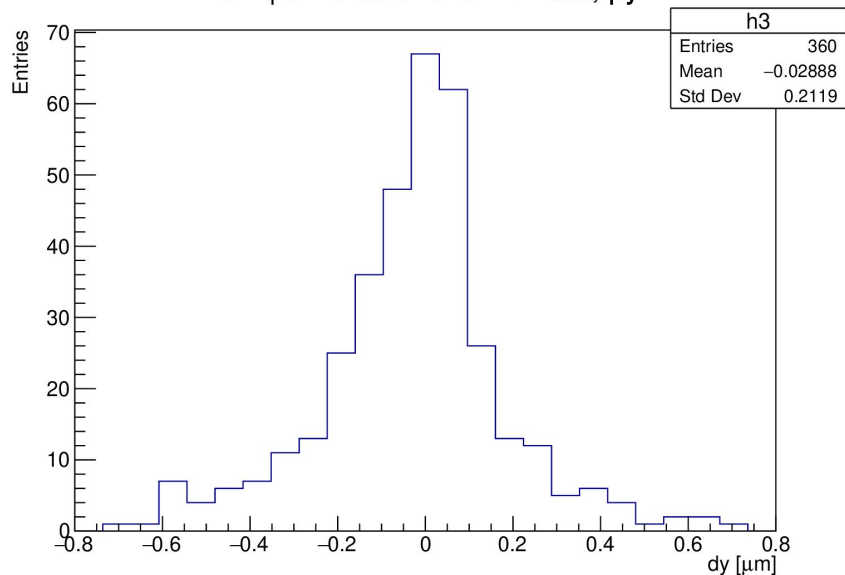


100 μ - Comparison Shift Geo vs Shift Data

ICP | shift Geo vs Shift Data, px



ICP | shift Geo vs Shift Data, py





Conclusion

Software alignment using the overlapping areas of two sensors using an ICP algorithm works.

The results are consistent even when the misalignment between two sensors is large.

Both methods (shift geometry and shift data) yield similar results.

Thank you for your attention!