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Tracking with Hashing

First year PhD Seminar

23th June 2023

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Outline

1. Thesis context: ITk building

2. ATLAS Qualification Task

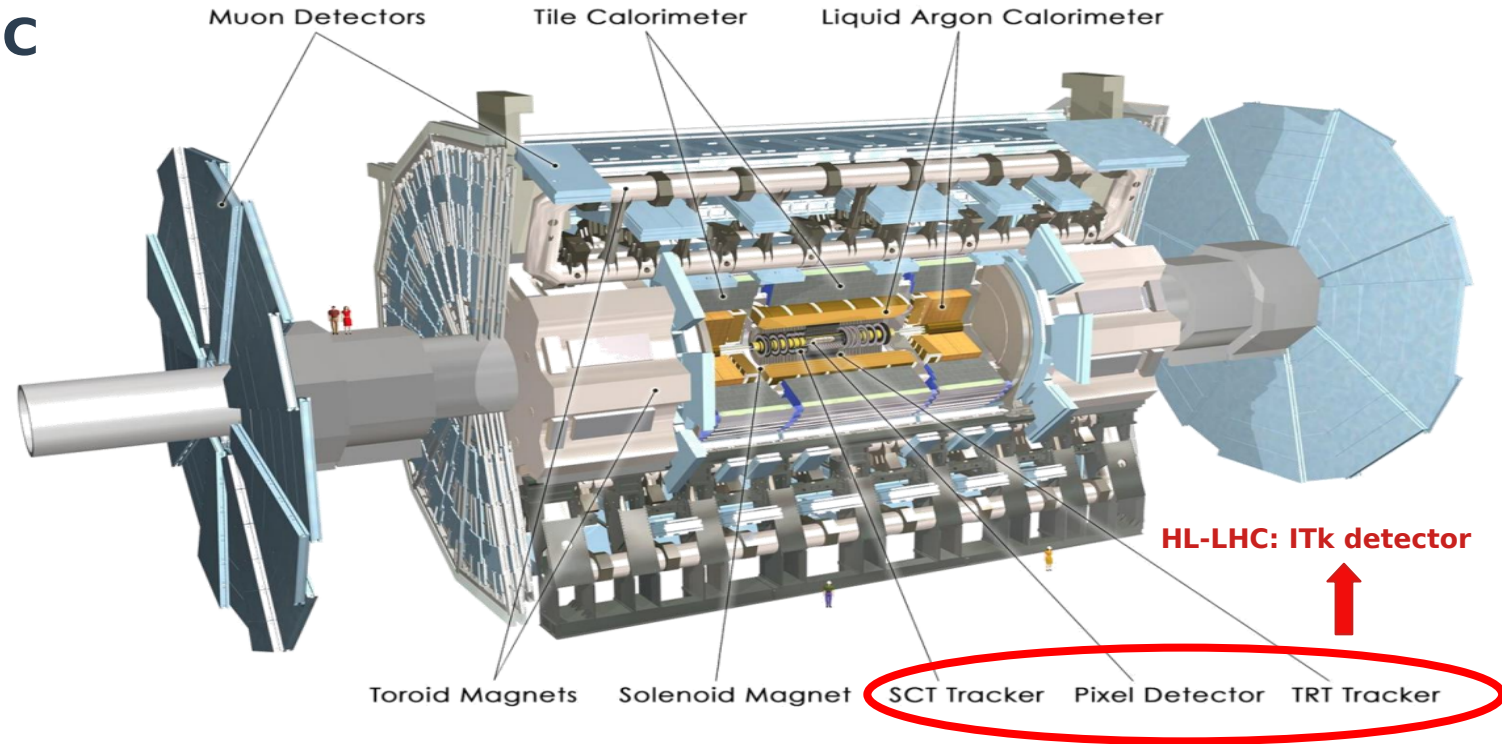
3. Thesis work:

- Baseline
- Method
- Results
- Where are we going?

1. Thesis context: ATLAS detector for HL-LHC

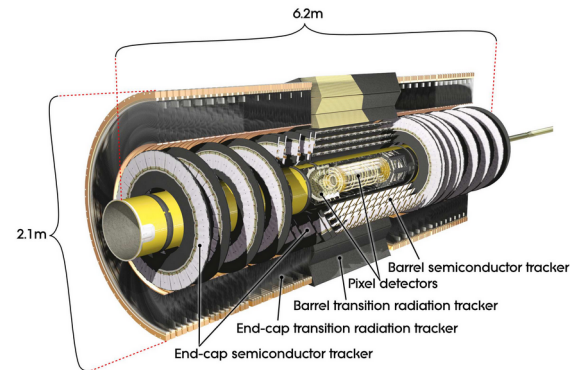
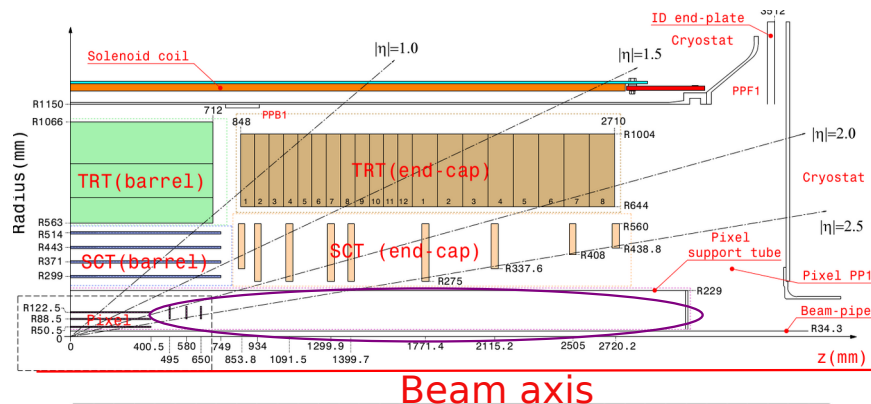
High Luminosity-LHC (HL-LHC):

- Expected in 2029
- Increase of luminosity
 - Luminosity: $\sim$ number of collisions per seconds

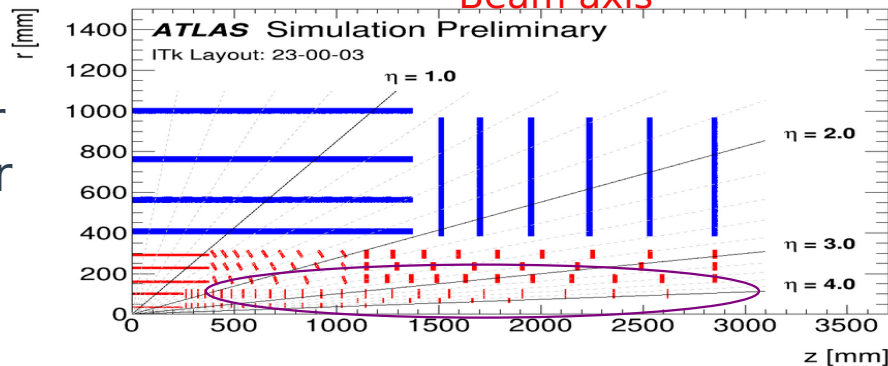


Inner Detector Overview

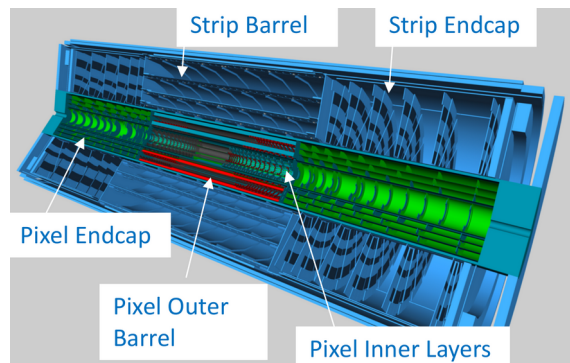
Current
Inner detector



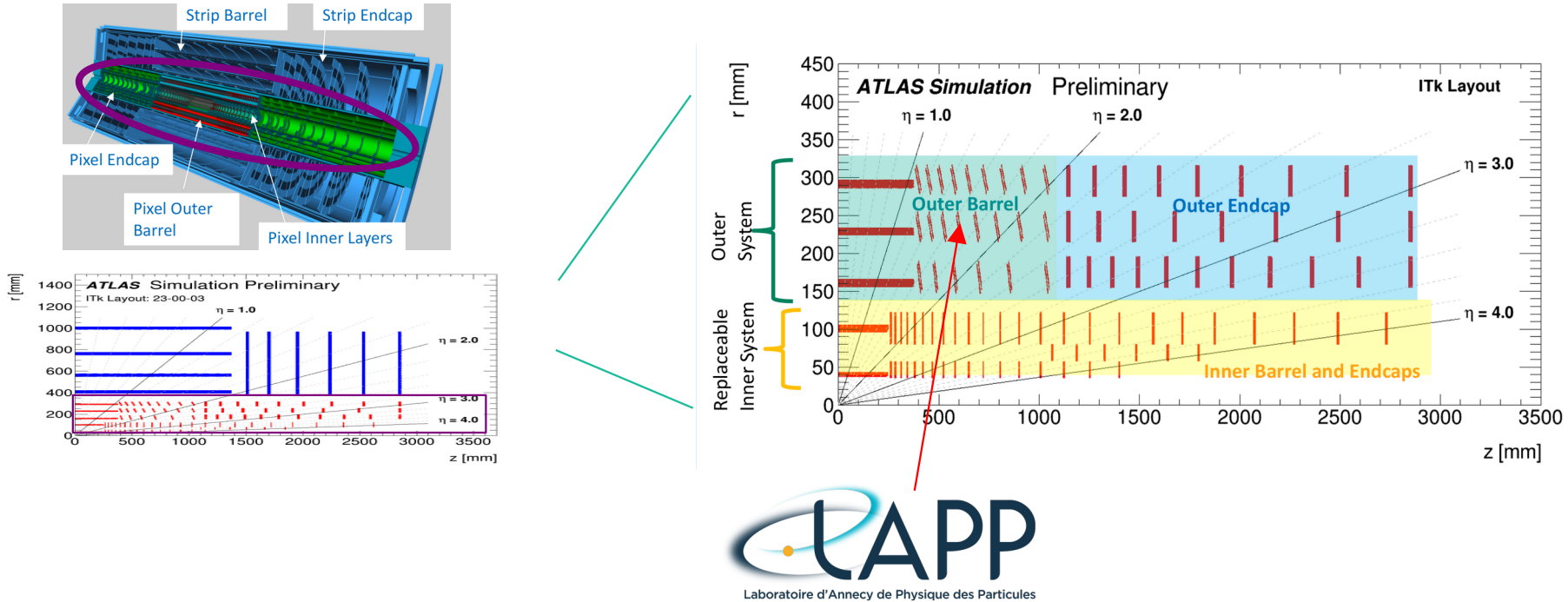
Inner Tracker
(ITk) detector



ITk: Wider coverage: $|\eta| < 4$
Higher granularity



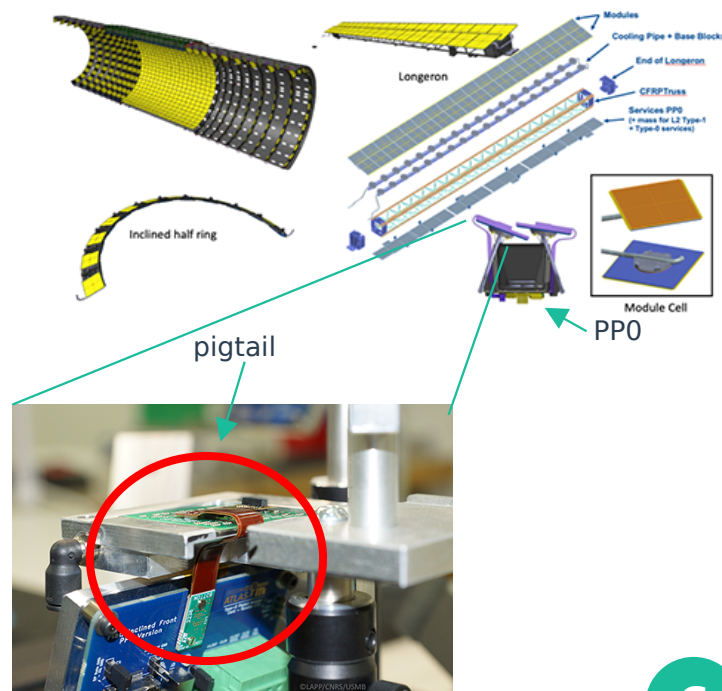
ITk Pixel Detector Overview



LAPP is building components of the Outer Barrel (OB)

LAPP ITk Pixel Detector Overview

- **LAPP is producing 75% of the OB Types 0 (5000 pigtails, 400 PP0 boards) and will be integrating 25% of the local supports (with LPSC and CPPM)**
 - Qualification Task on Production database for Types 0 (pigtails and PP0)
 - In 2024, participation to the electrical testing of modules at CERN or LAPP
- **Types 0: Components directly on the detector**
- **Pigtails: Power supply, monitoring of the cell and transmit data from the module cell**
- **Patch Panel 0 (PP0): Distribute power supply and aggregate data**



Pictures taken from: <https://lapp.in2p3.fr/spip.php?article3307>

Inner Tracker (ITk) for HL-LHC

- **ITk:**

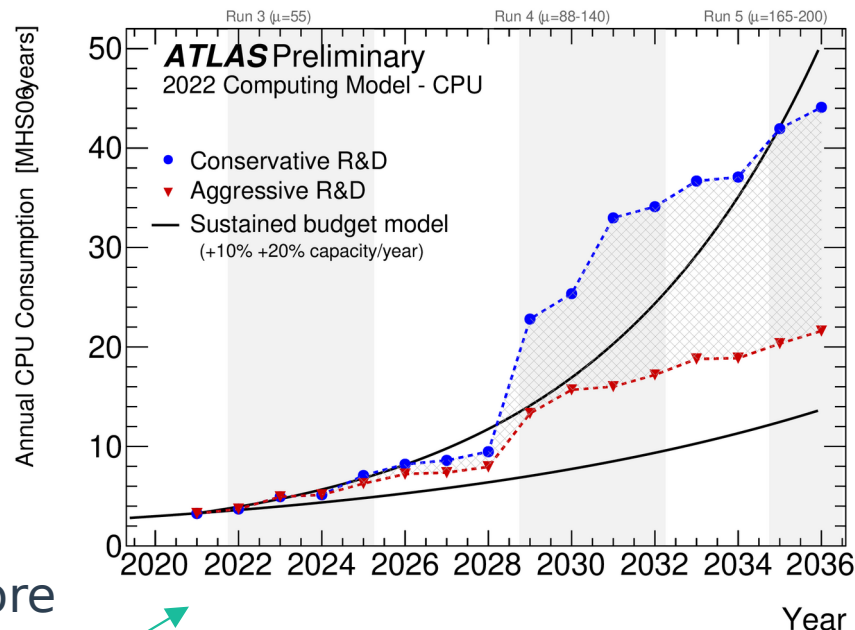
- Wider coverage: $|\eta| < 4$
- Higher granularity

More
particles
detected

- **High Luminosity-LHC (HL-LHC):**

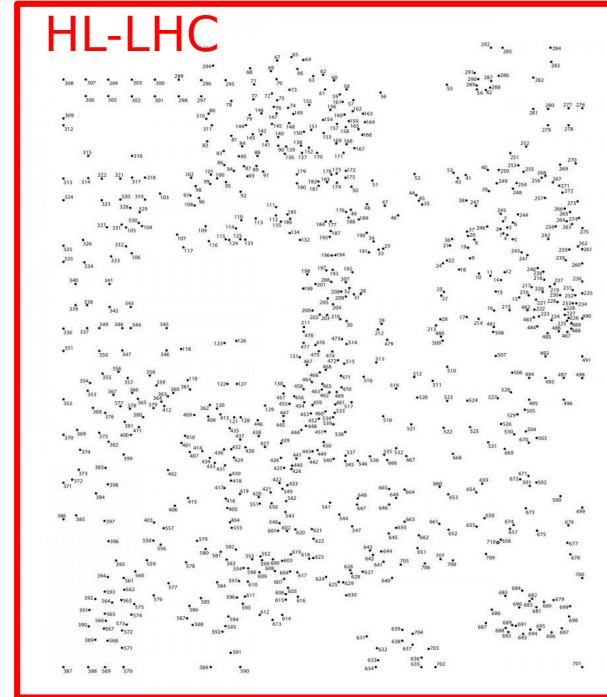
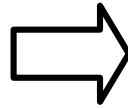
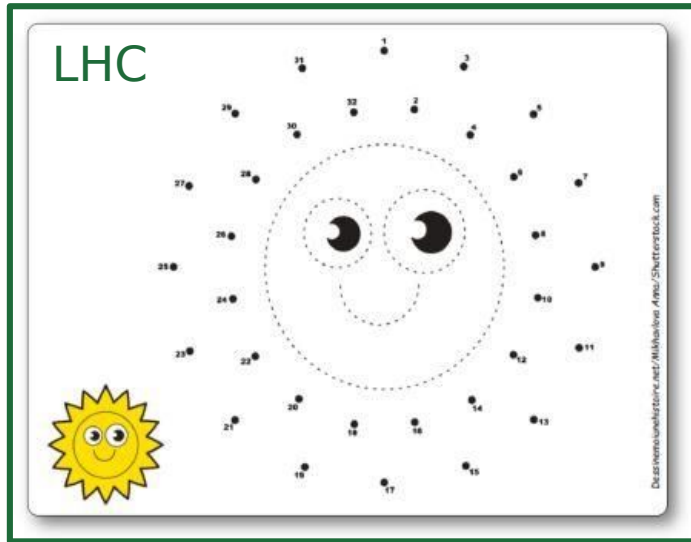
- Between now and 2029: Luminosity x2.5
 - Collisions piles up in an event
 - Pile-up ($\langle\mu\rangle$): average number of collisions in an event

More
particles
created



ATLAS CPU previsions: need to improve **tracking** performance significantly

Inner Tracker (ITk) for HL-LHC: in short



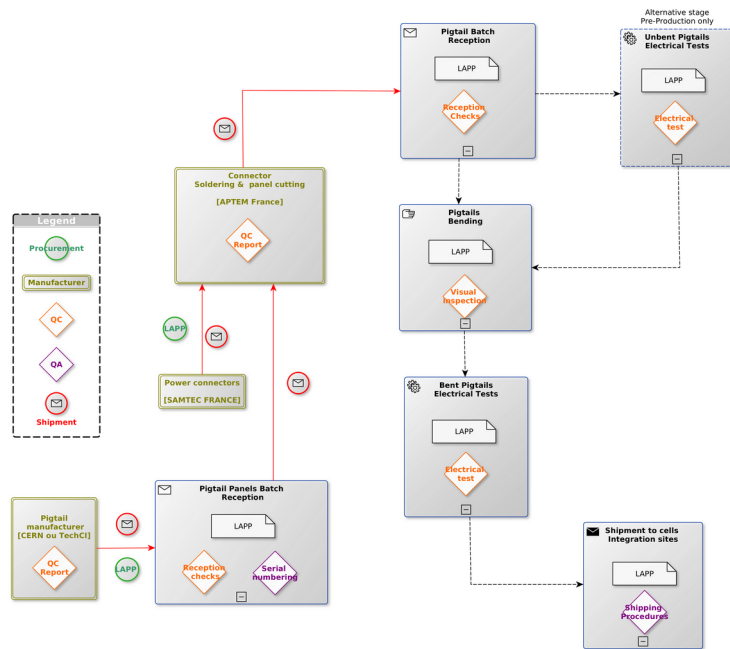
Tracking:



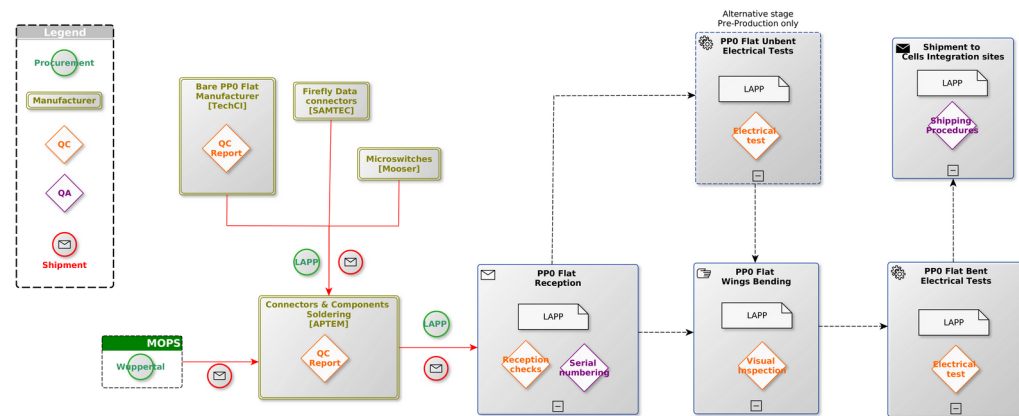
2. ATLAS Qualification Task: Production Database

- **ATLAS Qualification Task:**
 - 1 year, 6 months of work
 - Allows to publish as ATLAS member
- **ATLAS Production Database**
 - Create components, store quality control data, track shipping, API
- **Qualification Task:**
 - Creation of a dedicated “LAPP Types 0 Web app” to improve data registration in the database, robustness and scalability (**5000 pigtails**)
- **Help other laboratories: make the tool available for all the laboratories involved in Type 0 qualification**

ATLAS Qualification Task: Types 0 web app



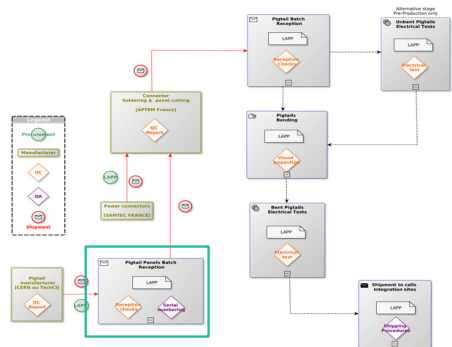
Pigtails production flow



PP0 production flow

Web app link: <https://itk-web-apps-pigtails.app.cern.ch/>

ATLAS Qualification Task: Types 0 web app example



Pigtails production flow

Example:
Registering 20 pigtails

ATLAS PDB interface
x20

The screenshot displays the ATLAS PDB interface, which consists of 20 vertical columns. Each column contains a list of data entries, likely representing pigtails. The interface is designed for managing and tracking the production and qualification of pigtails.

Web app
x1

Reception before cabling

Toggle debug for details

Refresh

Just testing

Reception date:

2023/06/23

Operator name:

QC operator name:

Humidity indicator (%)

0

Create a batch of panels

Select component sub-type for DB, PICTAL, PANEL

PANEL_PIG_INCLINED_BACK

Batch already exists

Production version

Prototype

Manufacturer of the panels

CERN

Batch serial number

11

Batch name: Panel_PIG_Inclined_Back_CERN_1

Number of panels to create in the batch

3

Number of pigtails per panel

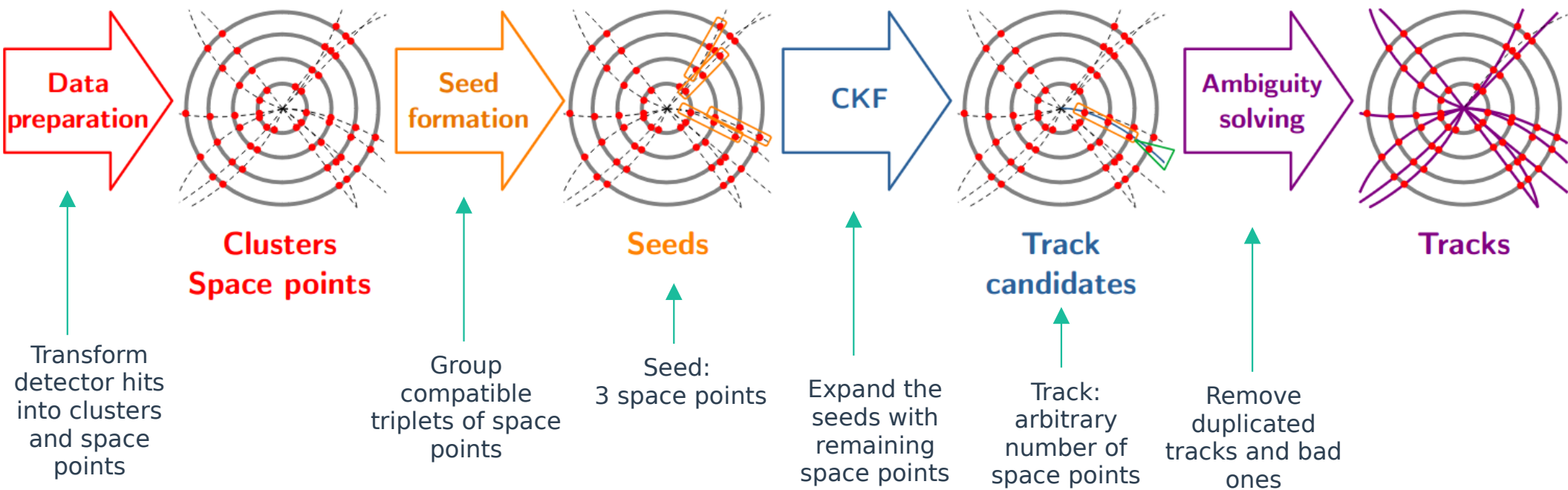
10

Create panels

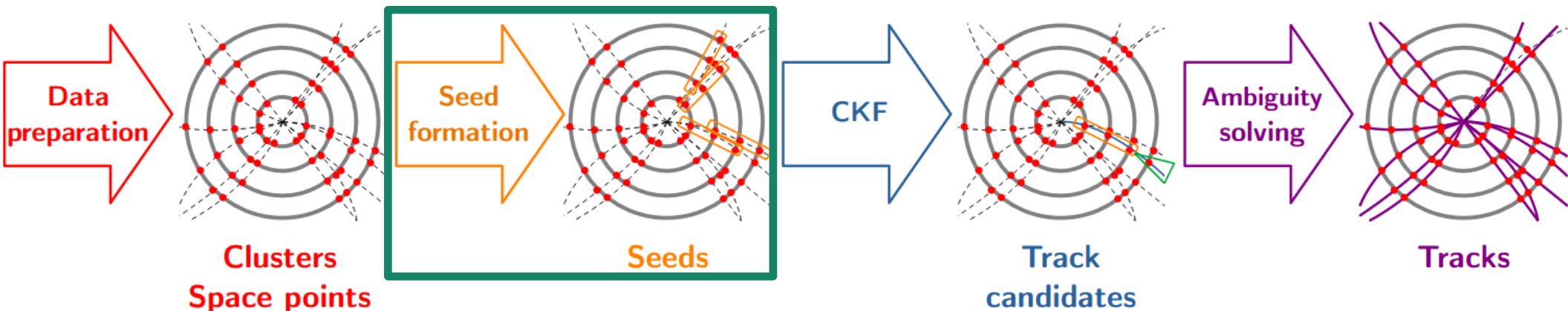
No panels created

Web app link: <https://itk-web-apps-pigtails.app.cern.ch/>

4. ATLAS Tracking



ATLAS Tracking Focus on Seeding



- **What do we hope to improve?**
 - Seeds' efficiency: reconstruct at least one seed per track
 - Seeds' purity (fake rate): reconstruct only tracks' seeds
 - Seeds' redundancy (duplication rate): reconstruct just enough seeds per track
- **How? Group similar Space Points? Might also improve timing?**

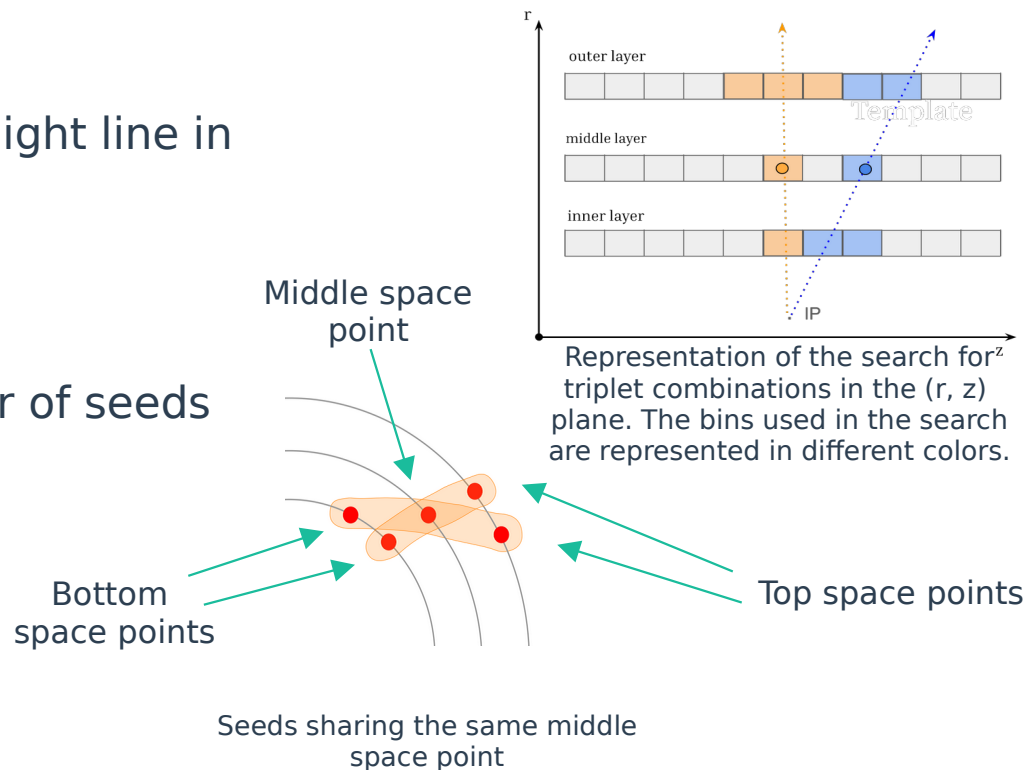
Default Seeding Algorithm description

1. Seed Finder

- Check if the triplet forms a nearly straight line in the (r, z) plane

2. Seed Filter

- MaxSeedPerSpM cut limits the number of seeds to speed up the tracking
- **Possible improvement:**
 - MaxSeedPerSpM: Non physical cut
- **Can we remove it?**



Baseline: Setup

ACTS: A Common Tracking Software
(generic tool) (experiment independent)



Run 4:
 $\langle \mu \rangle = 140$

Sample: 100 tt events; $|\eta| \leq 4$; $p_T > 1\text{GeV}$; $\mu = 50, 100, 150$

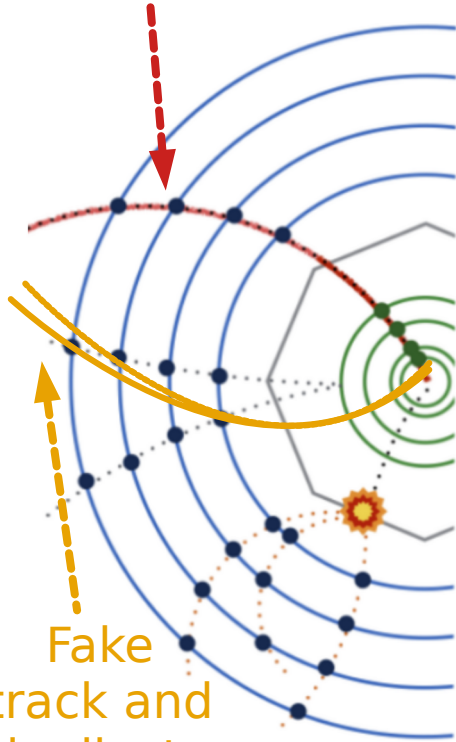
Generic detector (default detector, ideal detector without material)

Softwares: Pythia8 (simulation), FATRAS (propagation)

But not using Geant4: No secondary particles (particles created from the interaction with the detector)

Evaluation

Truth track

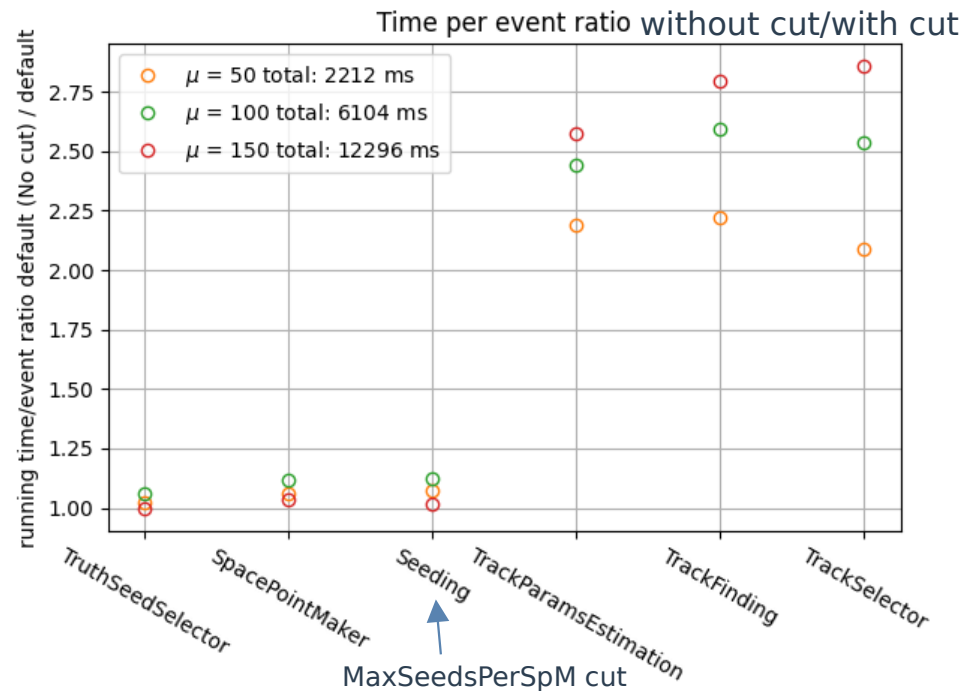
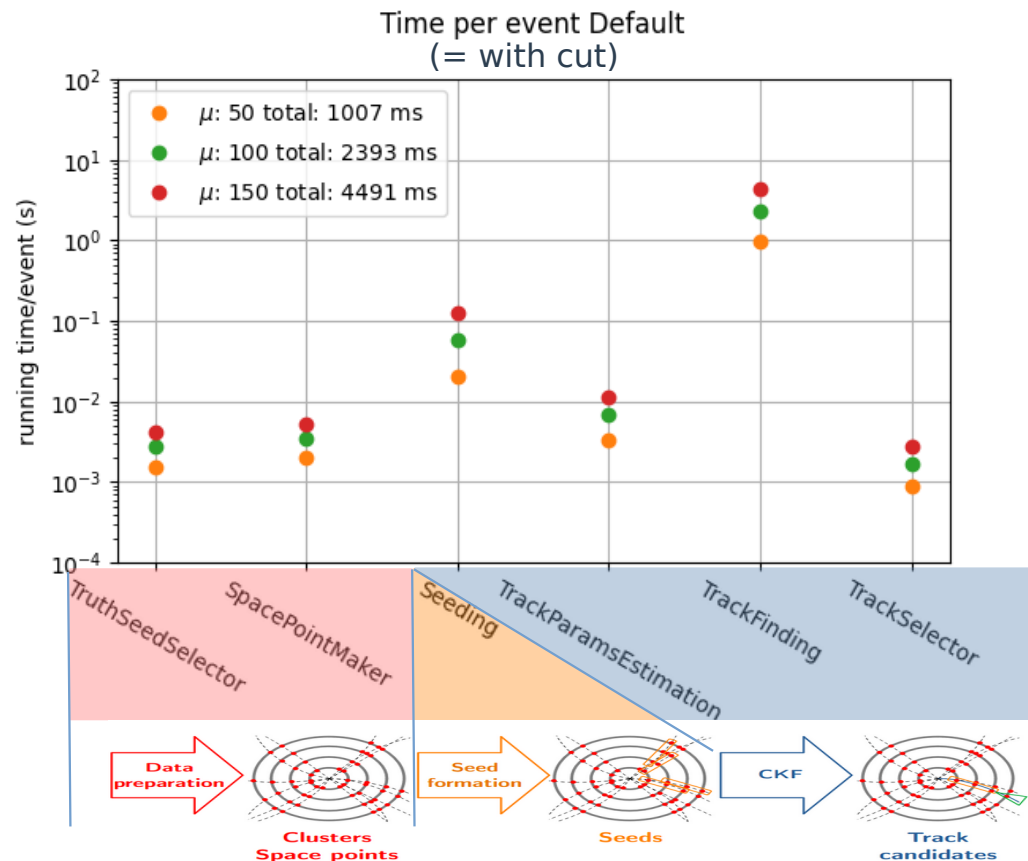


Fake
track and
duplicate

Evaluate on tracks:

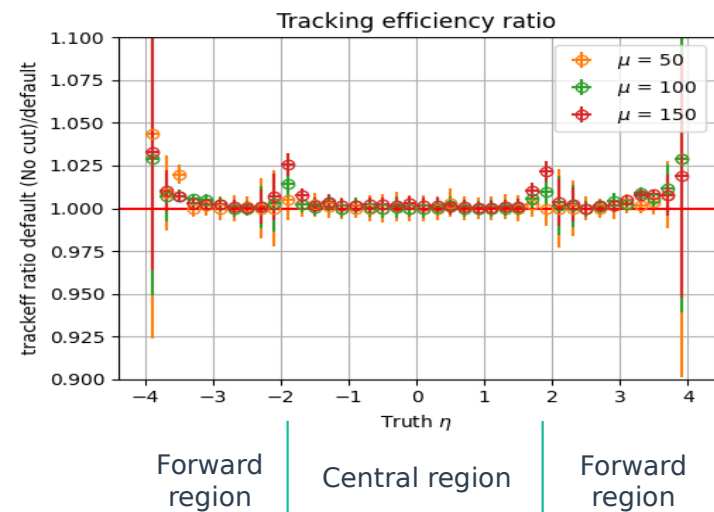
1. Efficiency: Reconstruct as much “truth” tracks as possible
2. Fake rate: Reconstruct as low “fake” tracks as possible
3. Duplication rate: Avoid to duplicate tracks
4. Running time: Going as fast as possible

ACTS performance: Timing/event

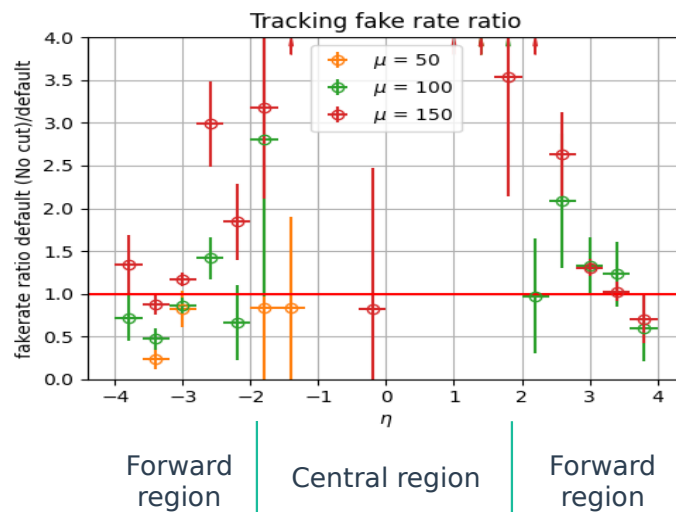


Timing multiplied by $\sim 2-3$ without the cut

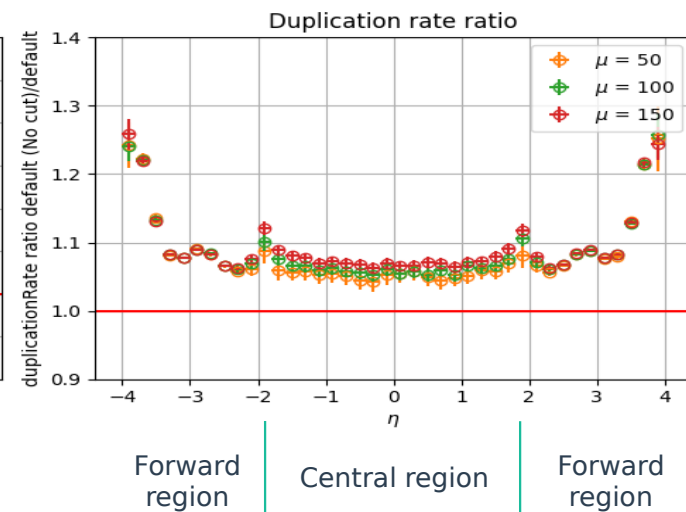
ACTS performance: Physics



Same in central region,
Better in forward region



Worst in central region,
Same in forward region



Worst everywhere,
Even worse in forward region

**MaxSeedPerSpM cut decreases the performance in forward region
But improves in central region**

Target Goal

- **Without the cut: improve performance but timing is crucial**
- **Goal: Improve performance with same timing**
 - Keep the cut but try to bypass it

A new method: Machine Learning/Hashing in the Seeding

Hashing:

1. Group similar space points into buckets
2. Do the seeding on each bucket

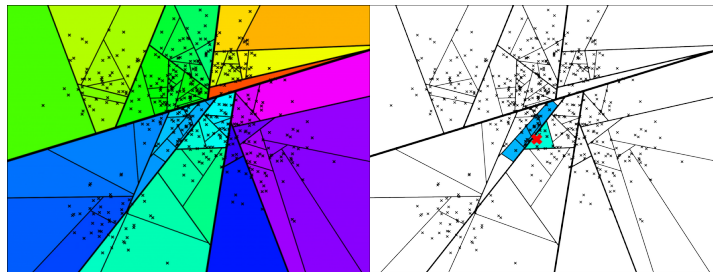
Algorithm used:

Approximate Nearest Neighbors Oh Yeah (**Annoy**)

→ Used by Spotify

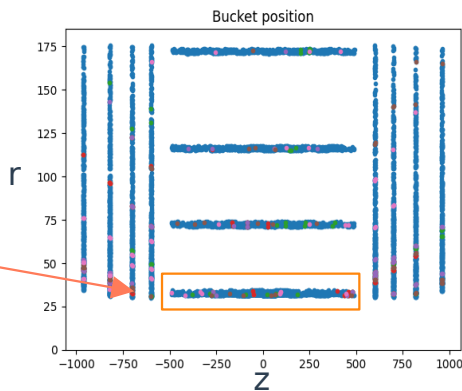
- Machine Learning algorithm type:
 - k Nearest Neighbors (unsupervised)
 - Random based
- Constraint: Starting from layer 0
- Parameter: Number of Neighbors (bucket size)
- Use the distance between the points
→ **need to define a (relevant) metric**

Annoy:



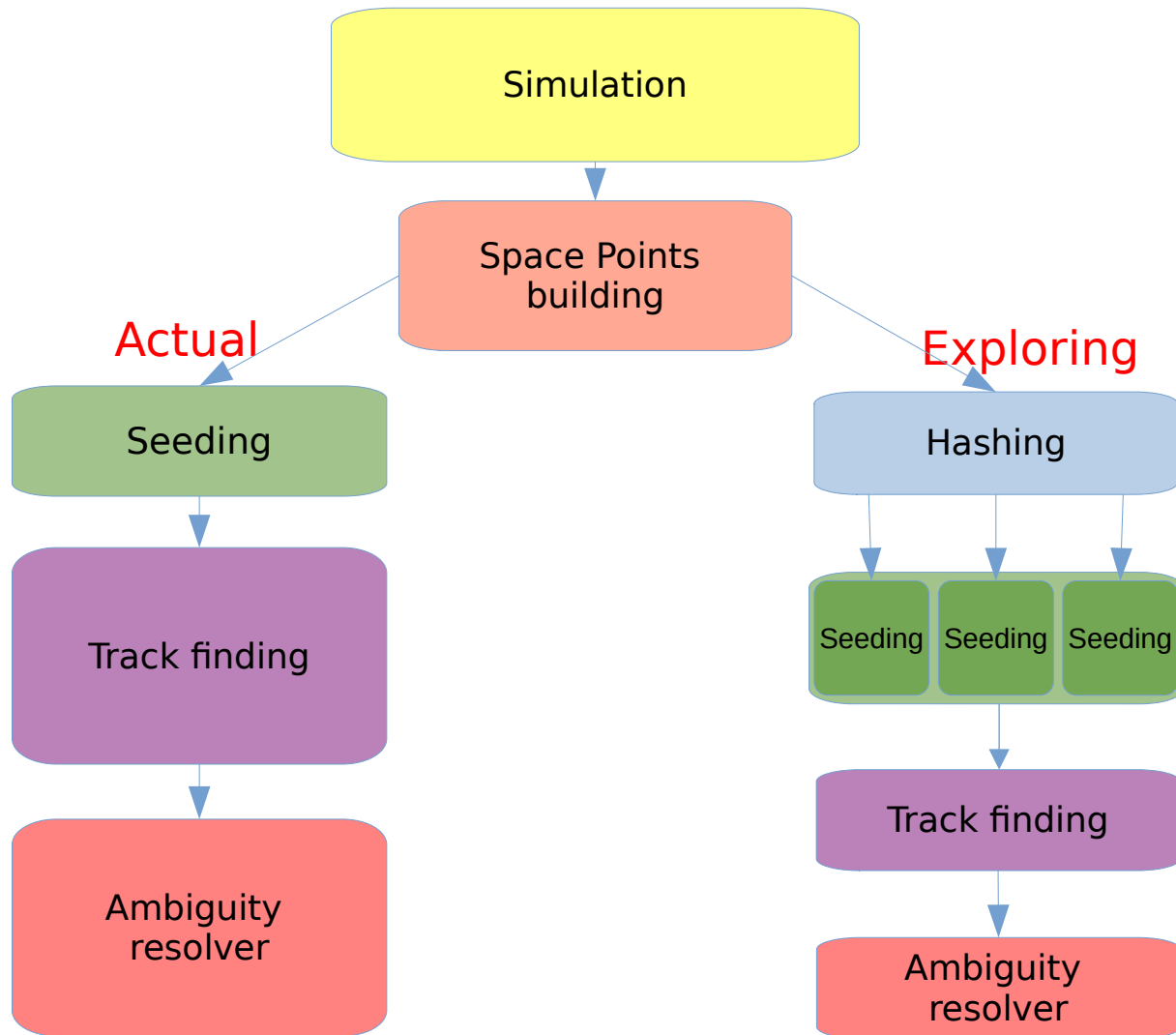
Space separation

Look for neighbors in the
closest regions



1 space point in layer 0 → 1 bucket

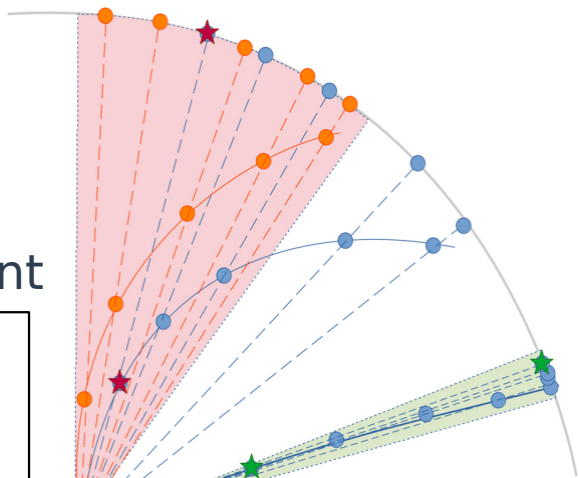
Approaches



- Seeding parallelization
- Hashing groups space points into buckets
- Hashing reduces the number of space points at a time (focus on relevant space points) → less seeds per bucket

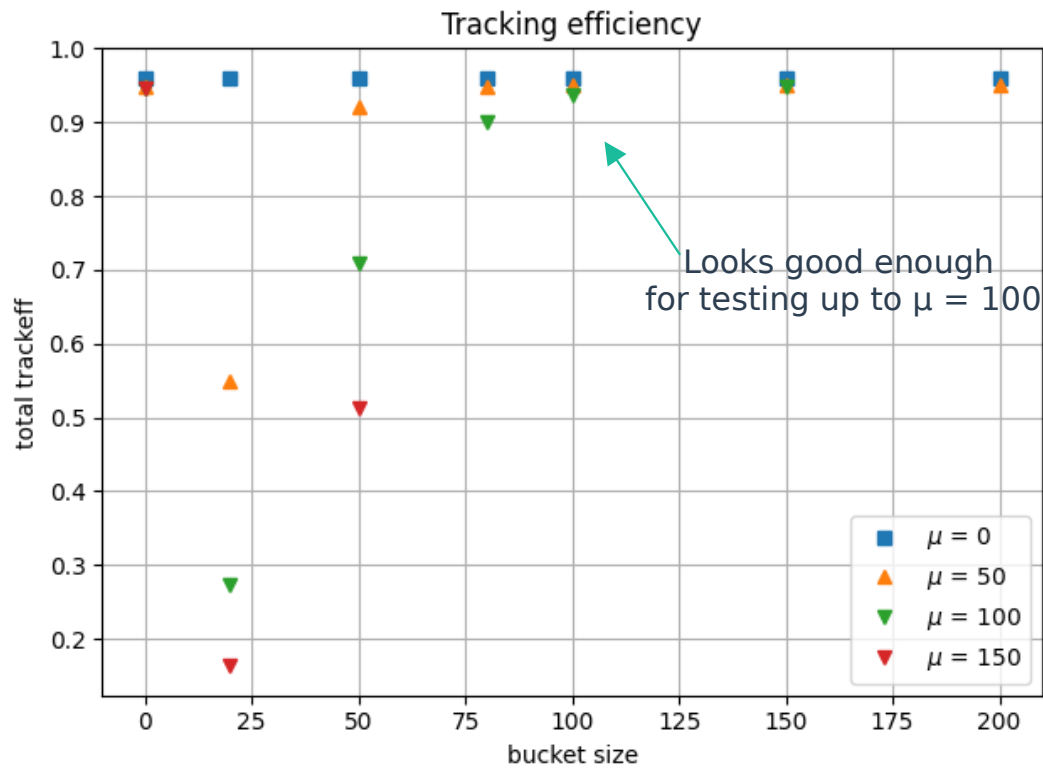
Metric and bucket size

Metric: $\Delta\phi$



SP: Space Point

- ☆ Layer 0 SP
- SP
- ▭ Bucket
- Projection
- Track



Fix the bucket size to 100

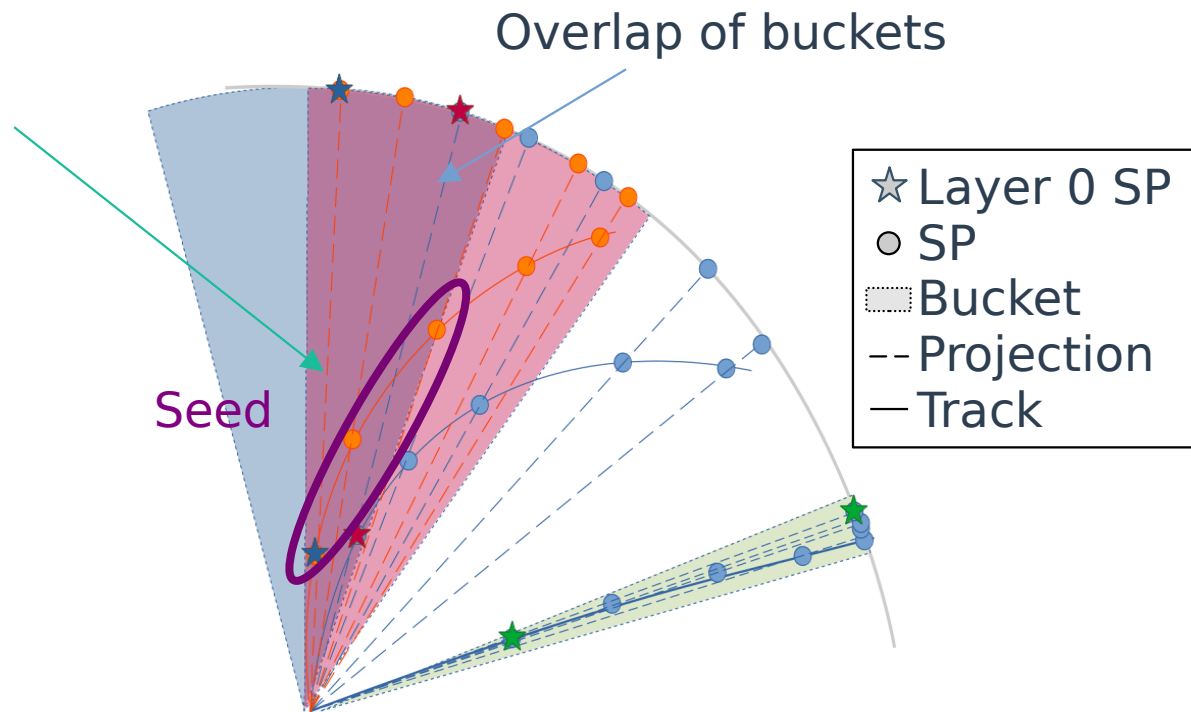
Hashing and overlap

- **Hashing introduces overlaps:**

- The same seed can be reconstructed in several buckets (14 times in average)
- Need some overlap to bypass the cut

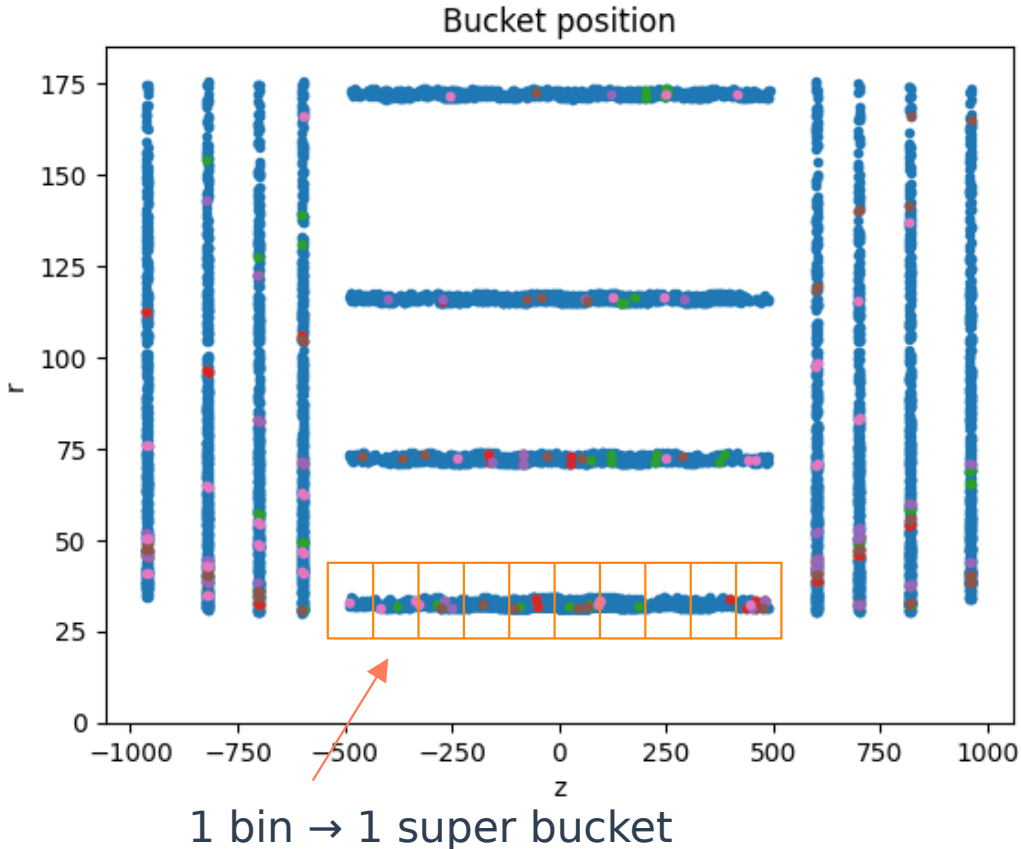
$\mu = 150$	Timing/ event (ms)
Without Hashing	4491
With Hashing	7909

Hashing made
timing x2

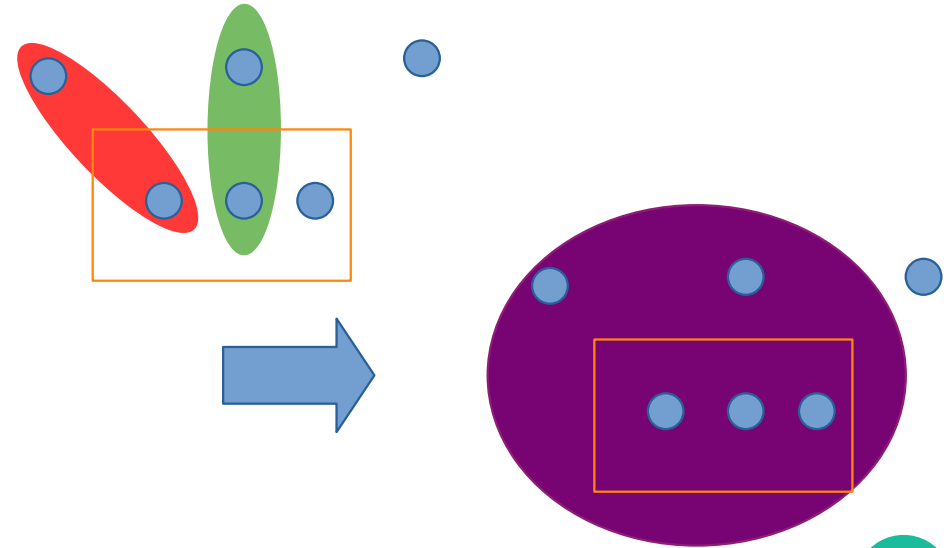


**New idea: Group buckets
→ less overlap**

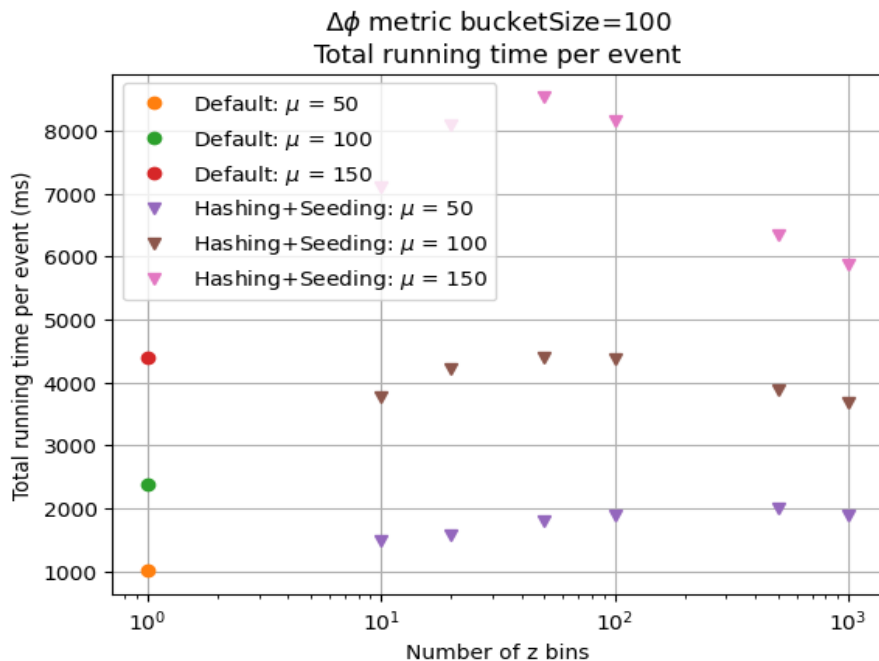
Super buckets and binning



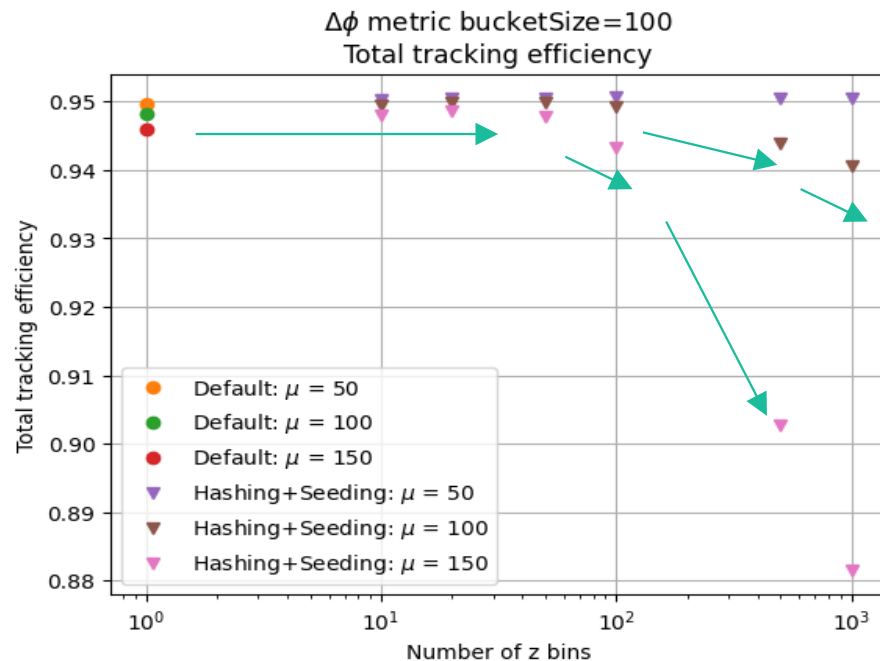
Super bucket:
Merging of the buckets created from
the space points inside the z bin



Hashing performance: Timing and efficiency



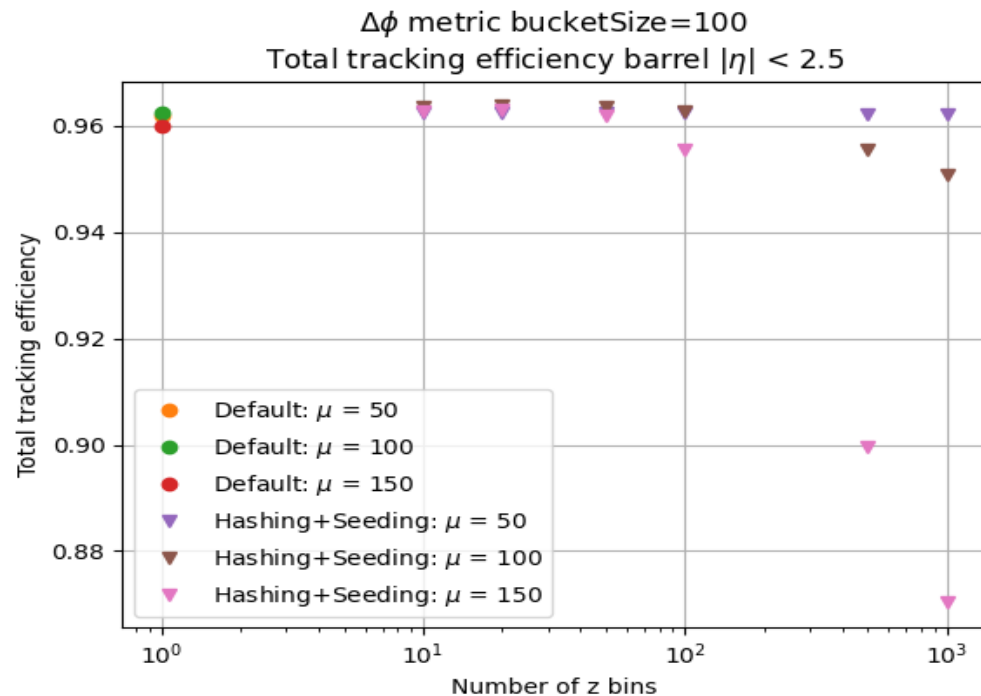
Running time $\sim x2$



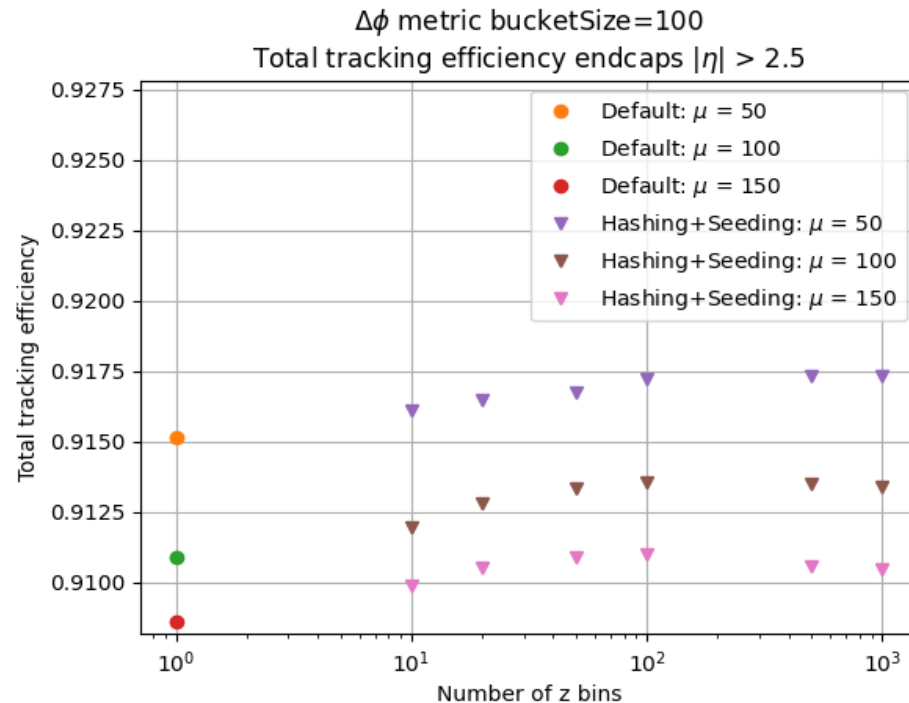
Improvement for small number of bins

Slower but better

Hashing performance: Efficiency (detailed)



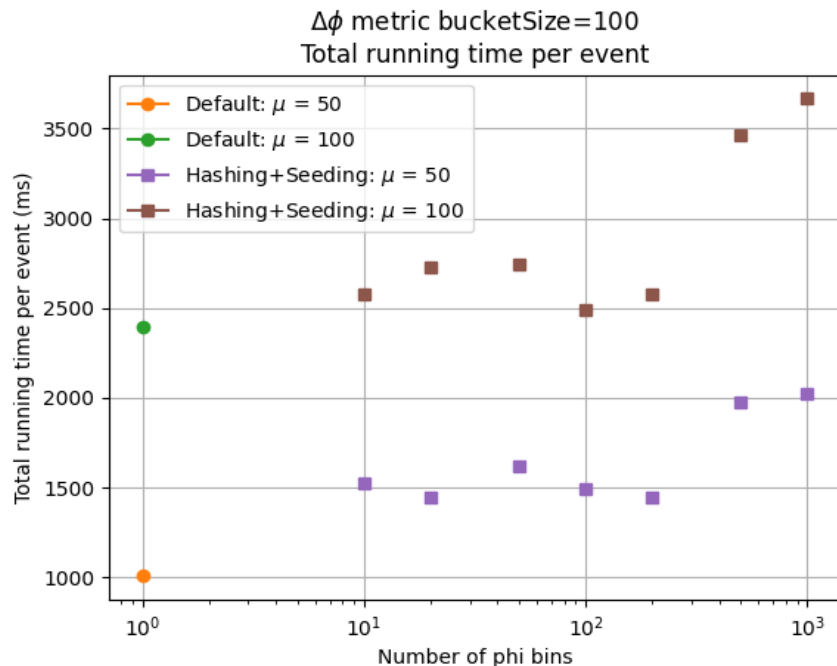
Improves then drops



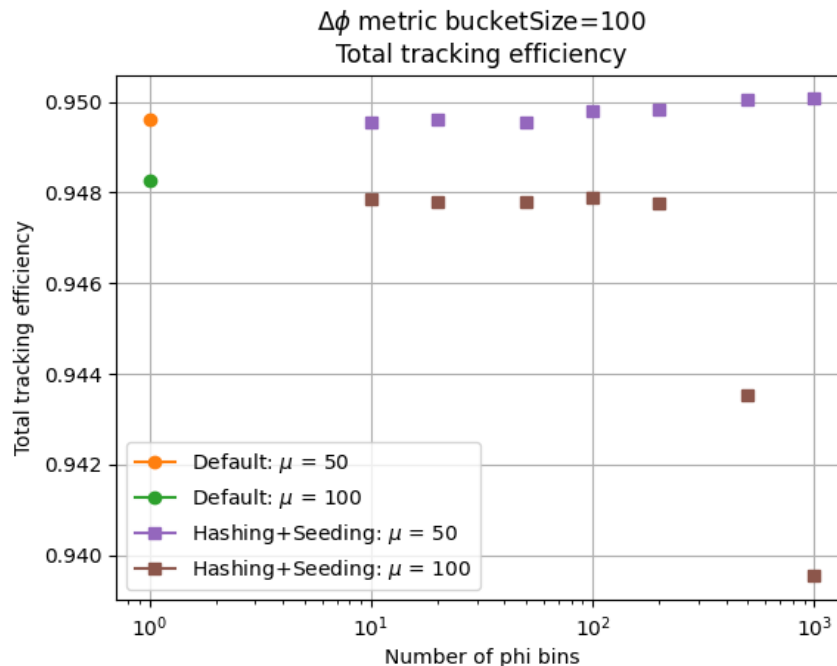
Always improve

Can be better in every regions

Hashing ϕ bins: Timing and efficiency



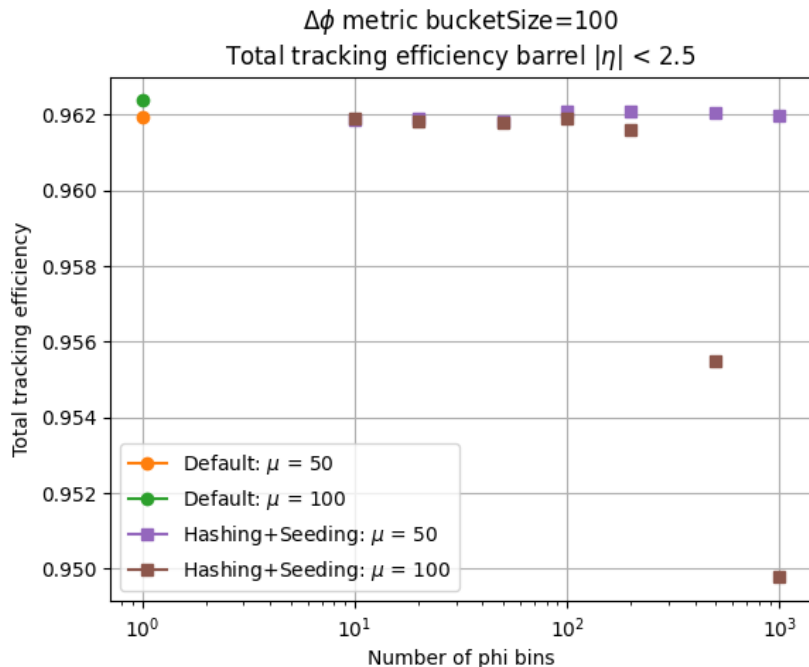
Similar running times



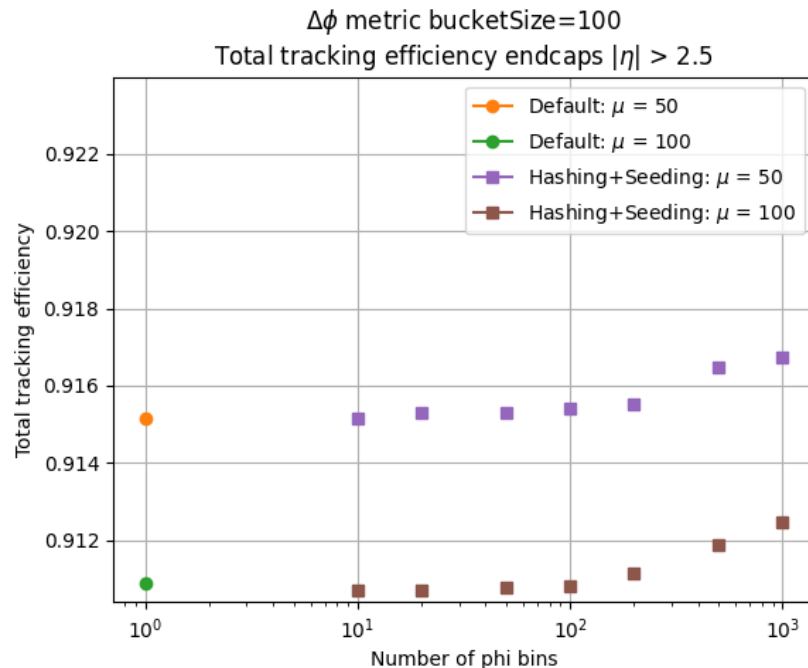
Small loss of efficiency

Not as good but as fast

Hashing ϕ bins: Efficiency (detailed)



Drop of efficiency in the barrel

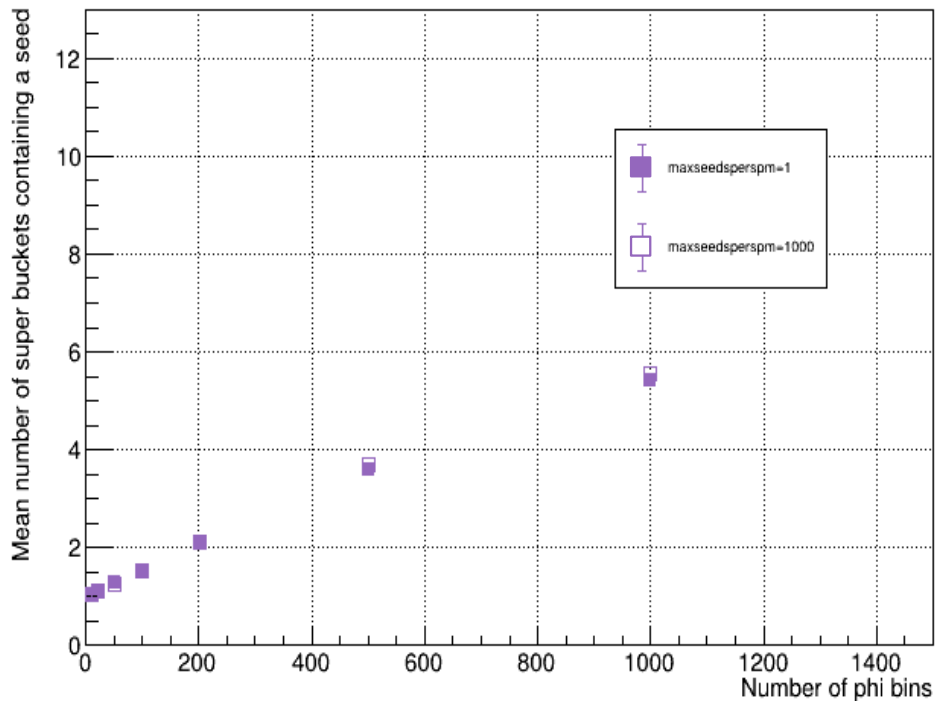


Better efficiency in the endcaps

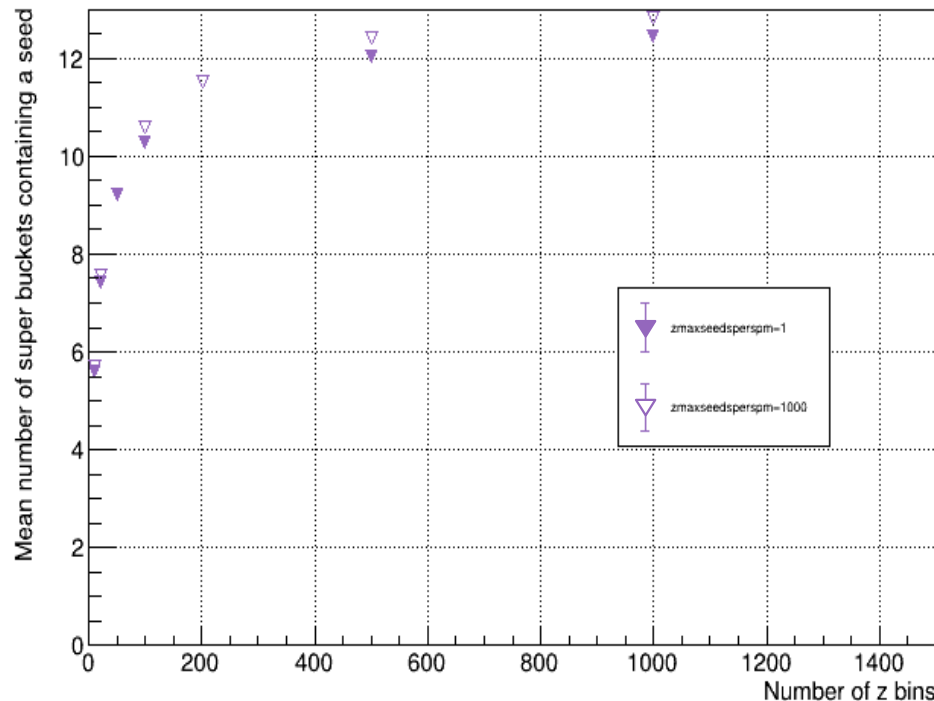
Better in forward region

Overlap in buckets

Overlap in buckets $\langle \mu \rangle = 50$ $\Delta\phi$ metric



Overlap in buckets $\langle \mu \rangle = 50$ $\Delta\phi$ metric



Less overlaps between buckets with ϕ binning

Where are we going?

- **Current state:**

- + Comparable performance with the baseline (slight improvement in the forward region!)
- + Trade off between timing and performance (Z binning vs ϕ binning)
- Not better timing than baseline for now

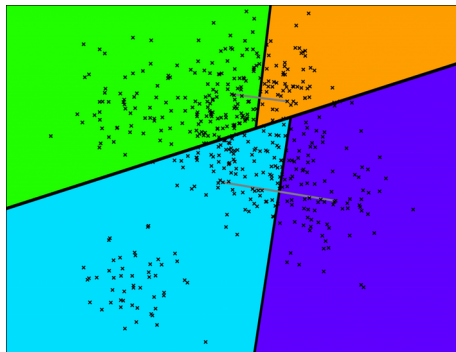
Next:

- **Full sim with Geant4 (secondaries), ITk, metric learning, binning in (z, ϕ), changing bucket size with detector region, Cluster shape?**
- **Summer School: HEP and ML. [Link](#)**
- **September: Quark/Gluon Tagging with ML**

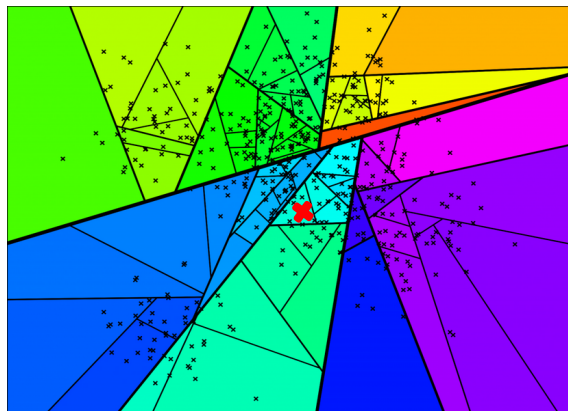
Backup

Annoy training

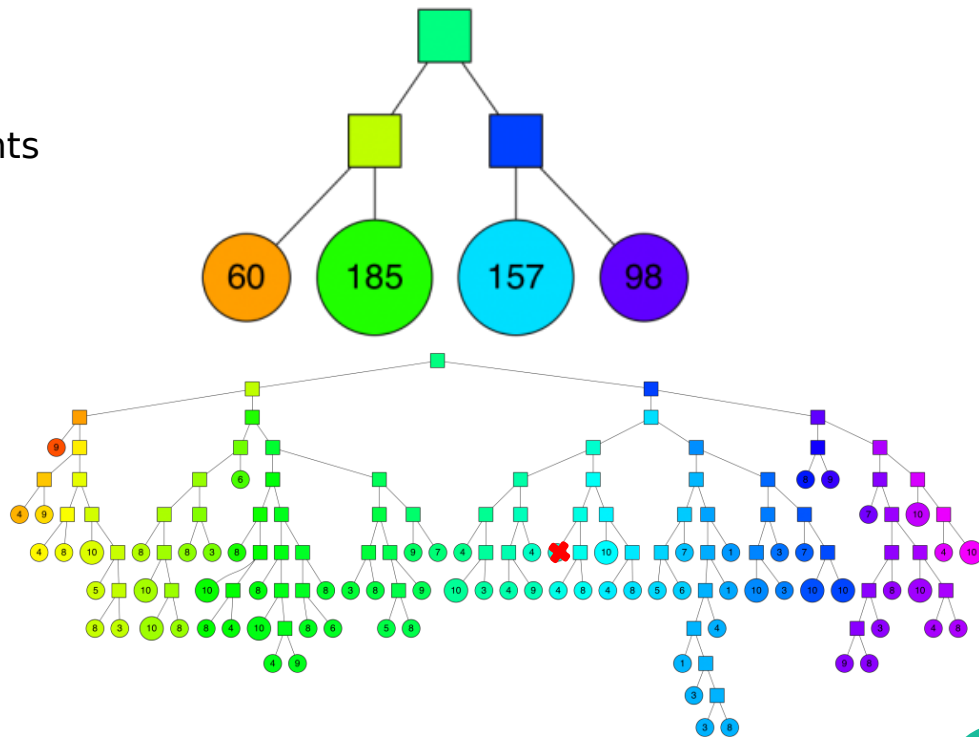
Space separation



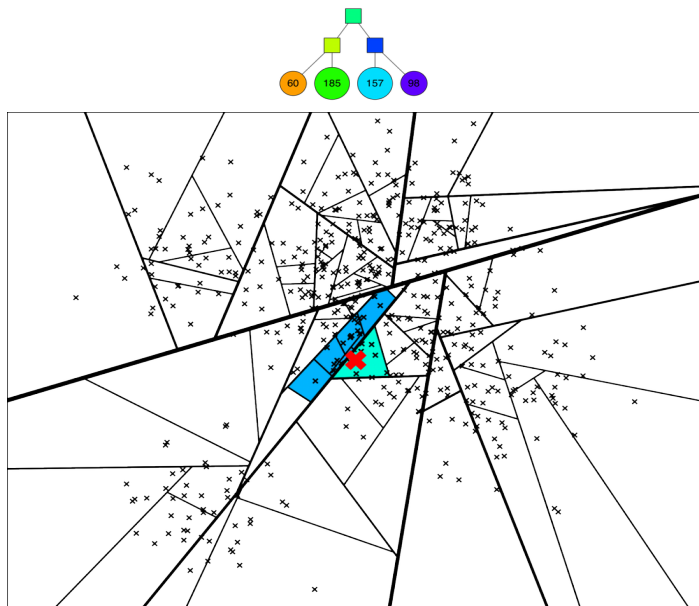
Takes two
random points
iteratively



Corresponding binary tree



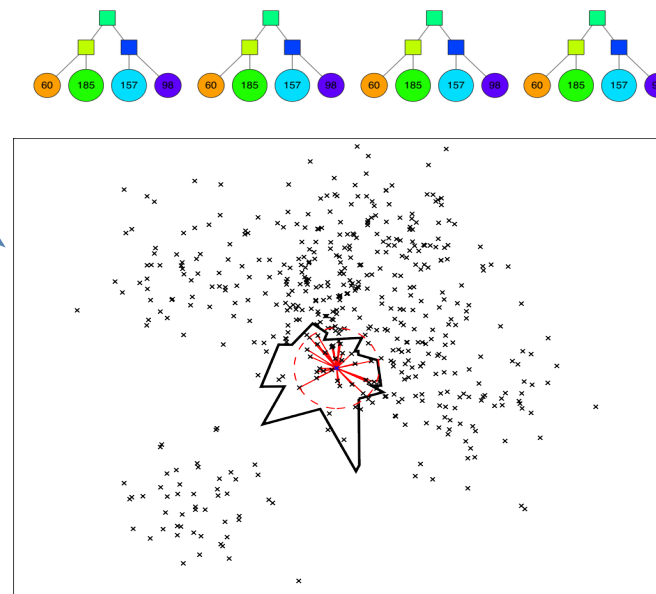
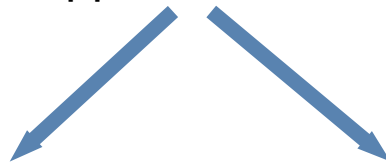
Annoy query



Merge neighbor subspaces

- Annoy tuning parameters: number of neighbors, number of trees, metric used, features used, number of subspace to look at

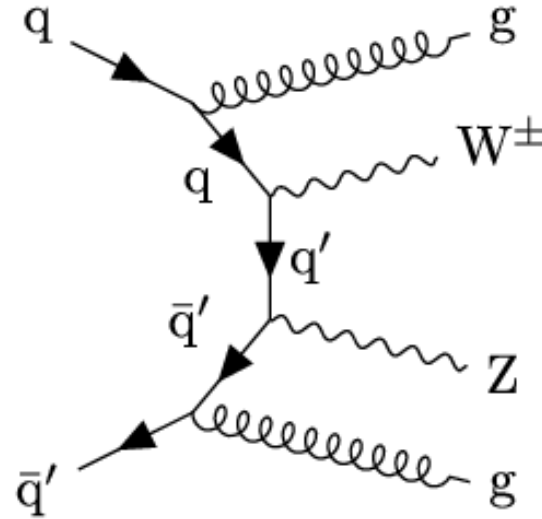
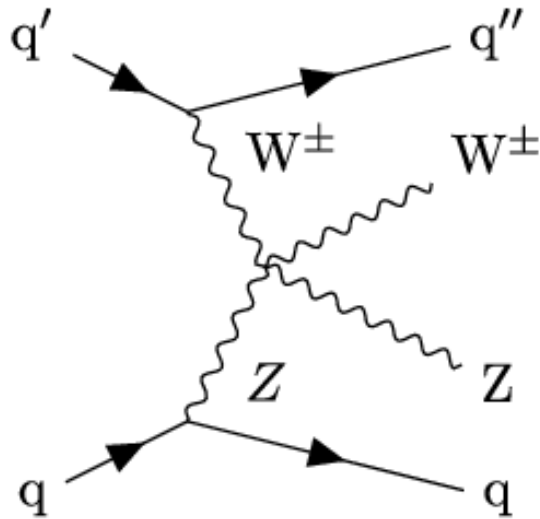
Approximation



Union of trees' subspace

Quark/Gluon tagging: Physics motivation 1

Vector Boson Scattering: Quartic Gauge coupling measurement

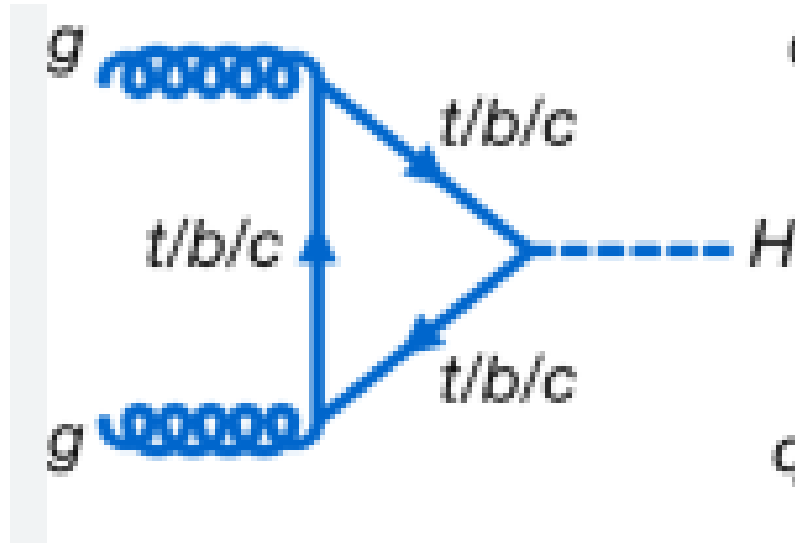


SIGNAL: Electroweak process
Forward quark jets

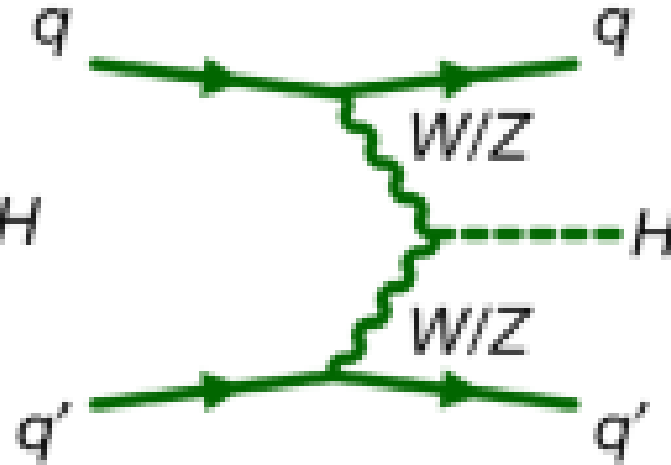
Background: QCD Process
randomly distributed gluon jets

Quark/Gluon tagging: Physics motivation 2

Higgs Boson production: separate cross-section measurements in various modes



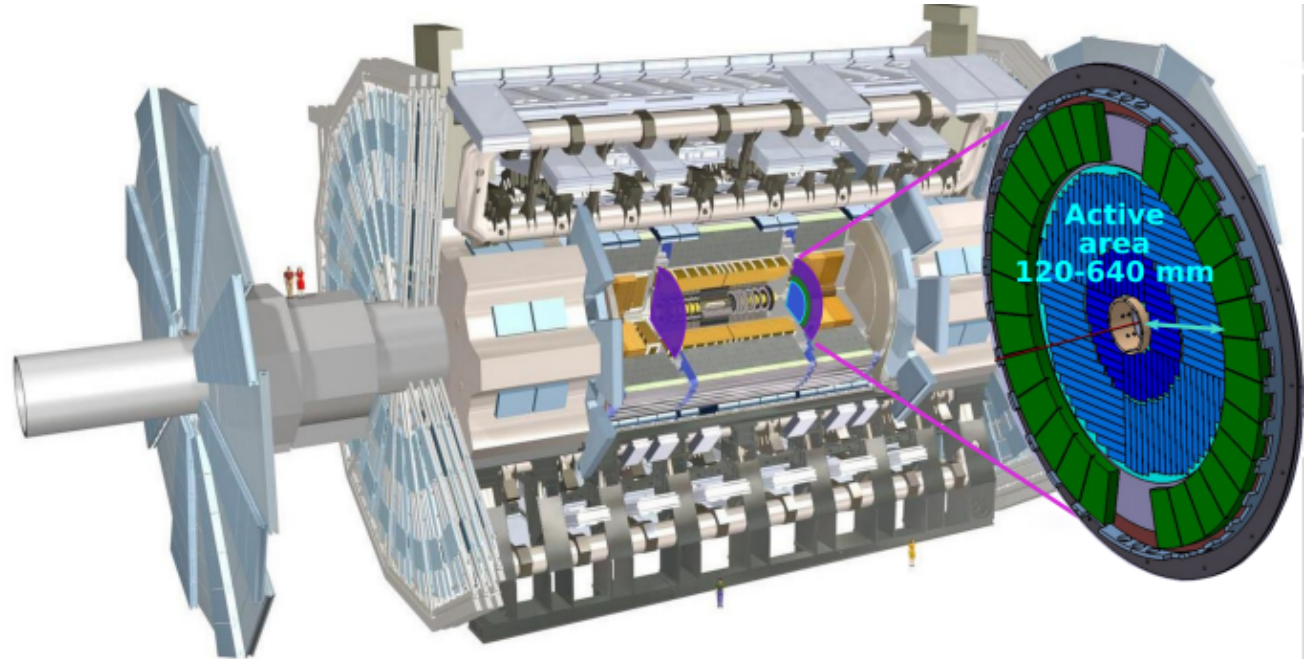
QCD Process: randomly distributed gluons jets



VBF process: forward quark jets

Quark/Gluon tagging: ATLAS forward region@HL-LHC

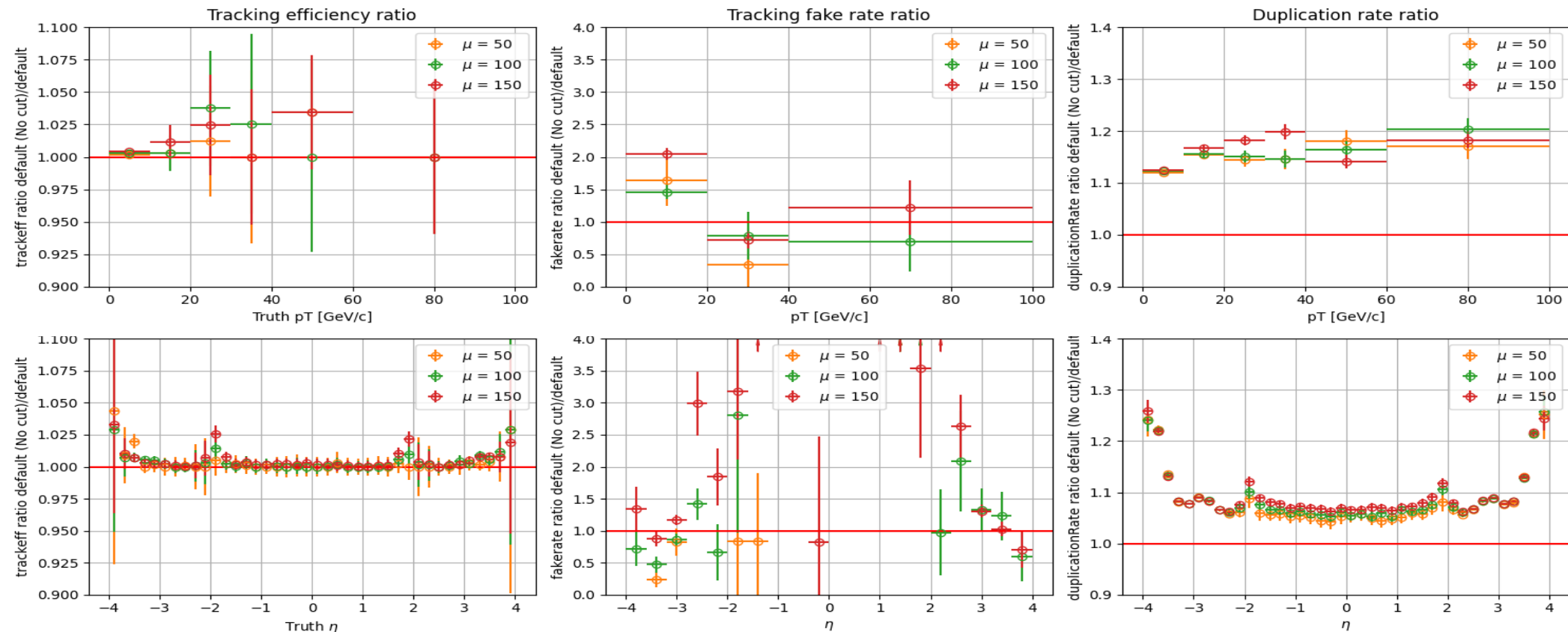
Nouveau détecteur ITk +
HGTD:



Formations

- **Doctoral School (UGA):**
 - Requires 120 hours: 1/3 Scientific, 1/3 Professional, 1/3 Transversal
- **Professional:**
 - “S'ADAPTER A SON ENVIRONNEMENT DE TRAVAIL” (10 hours)
- **Scientific:**
 - Workshops: ATLAS ML, ITk Tracking, ATLAS Induction Day and Software Tutorial (44 hours)
- **Transversal:**
 - Opened Science and HAL (4 hours)
 - “JOURNEE DE RENTREE DES DOCTORANTS 2022” (10 hours)
 - Planned: **Mooc** on ethics (15 hours)
- **BONUS: Summer School@ MITP (HEP and ML). [Link](#)**

ACTS performance: Physics



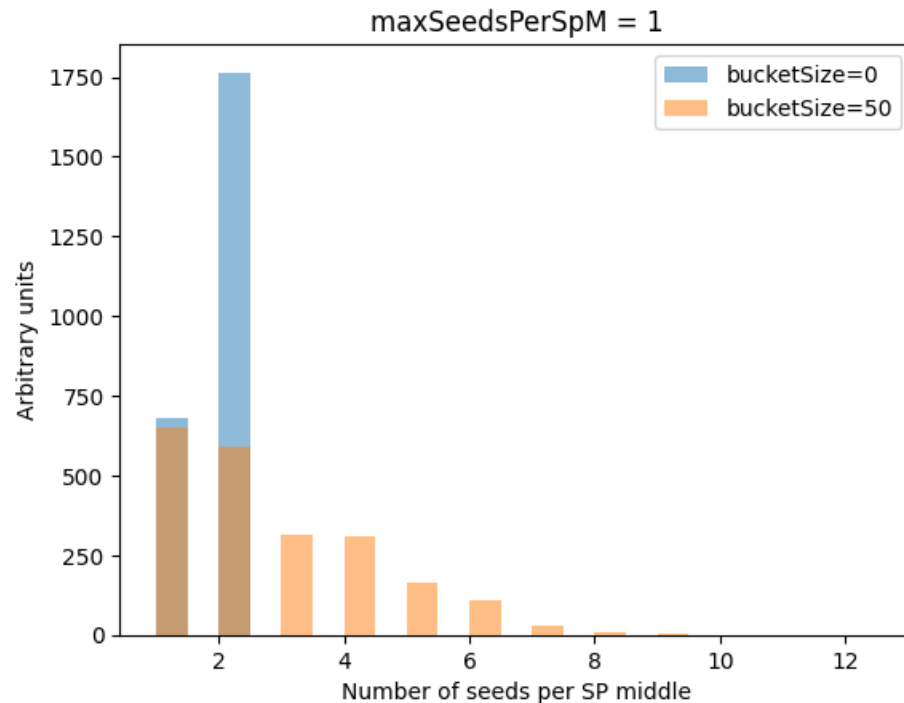
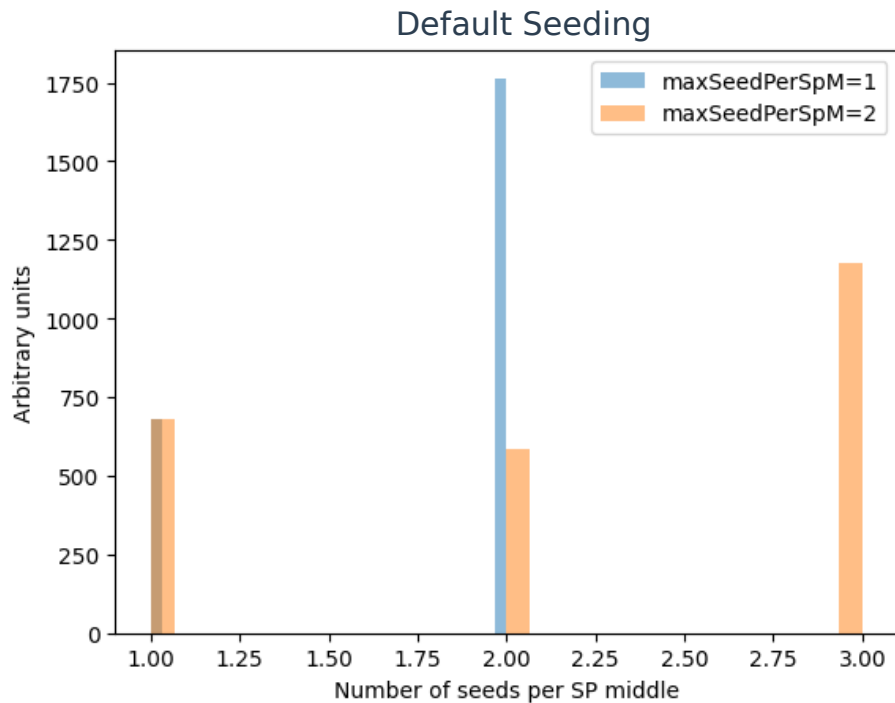
Seed finder configuration

```
SeedfinderConfigArg = SeedfinderConfigArg(  
    r=(None, 200 * u.mm), # rMin=default, 33mm  
    deltaR=(1 * u.mm, 60 * u.mm),  
    collisionRegion=(-250 * u.mm, 250 * u.mm),  
    z=(-2000 * u.mm, 2000 * u.mm),  
    maxSeedsPerSpM=1,  
    sigmaScattering=5,  
    radLengthPerSeed=0.1,  
    minPt=500 * u.MeV,  
    bFieldInZ=1.99724 * u.T,  
    impactMax=3 * u.mm,  
    cotThetaMax=cotThetaMax # =1/tan(2×atan(e-eta))  
)
```

MaxSeedsPerSpM cut

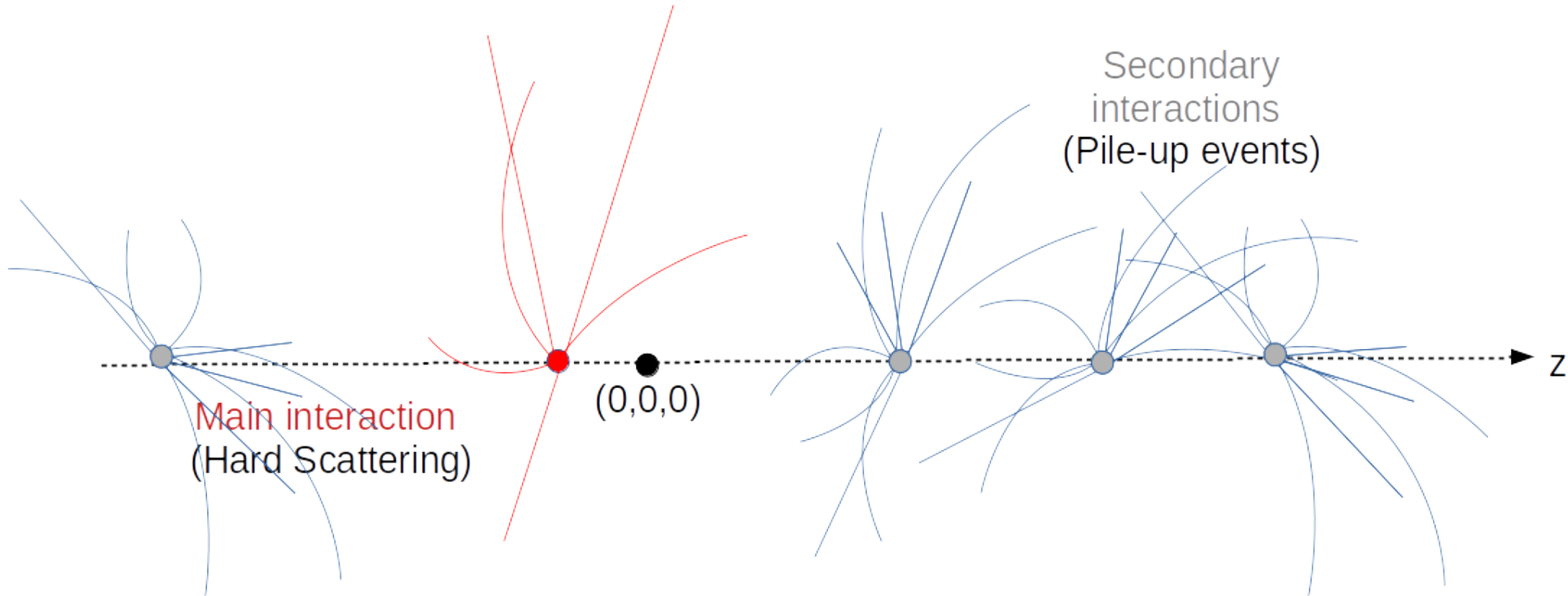
- **Purpose:**
 - Reduce the number of seeds to expand to speedup the track finding
- **Idea:**
 - Only keep at most $\text{MaxSeedsPerSpM}+1$ seeds sharing the same middle space point
- **Implementation:**
 - Uses a score to compare the seeds
 - The score is related to how close the impact parameter is to 0
- **Benefit:**
 - speedup and less memory used
- **Consequence:**
 - Loss of efficiency

MaxSeedsPerSpM cut vs Hashing

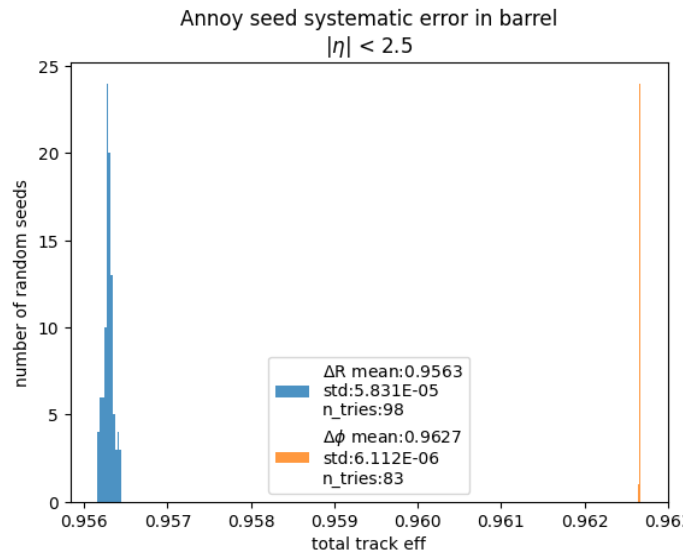
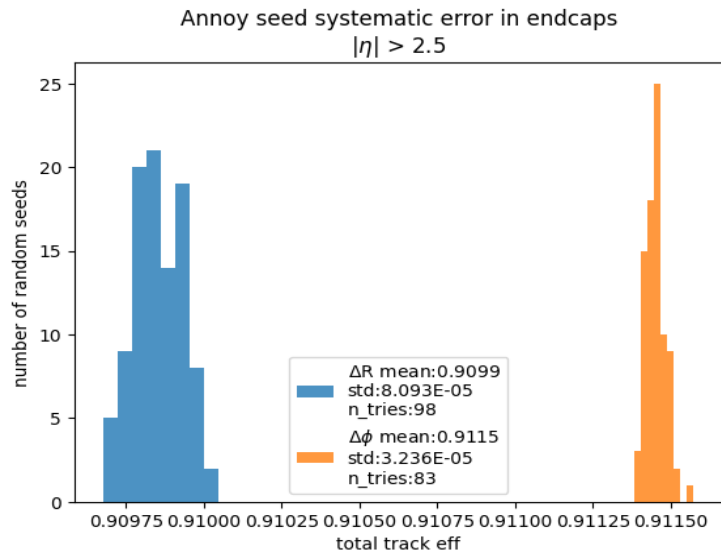
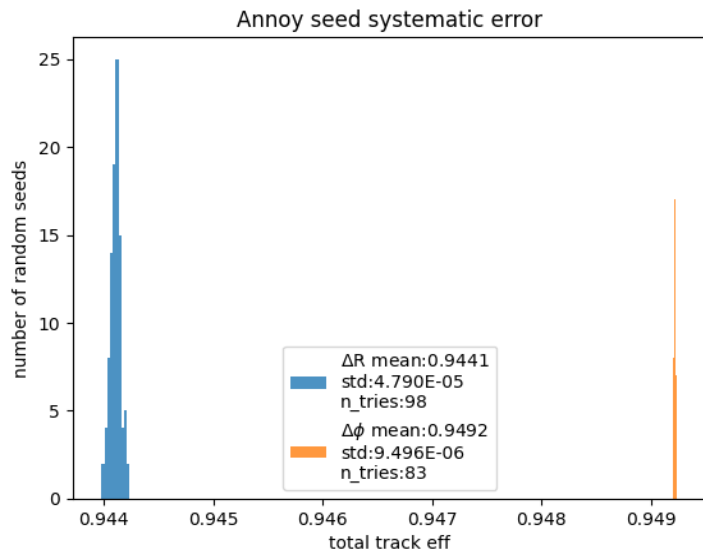


➡ Hashing get through the cut

Pile-up events



Annoy random seed systematic error



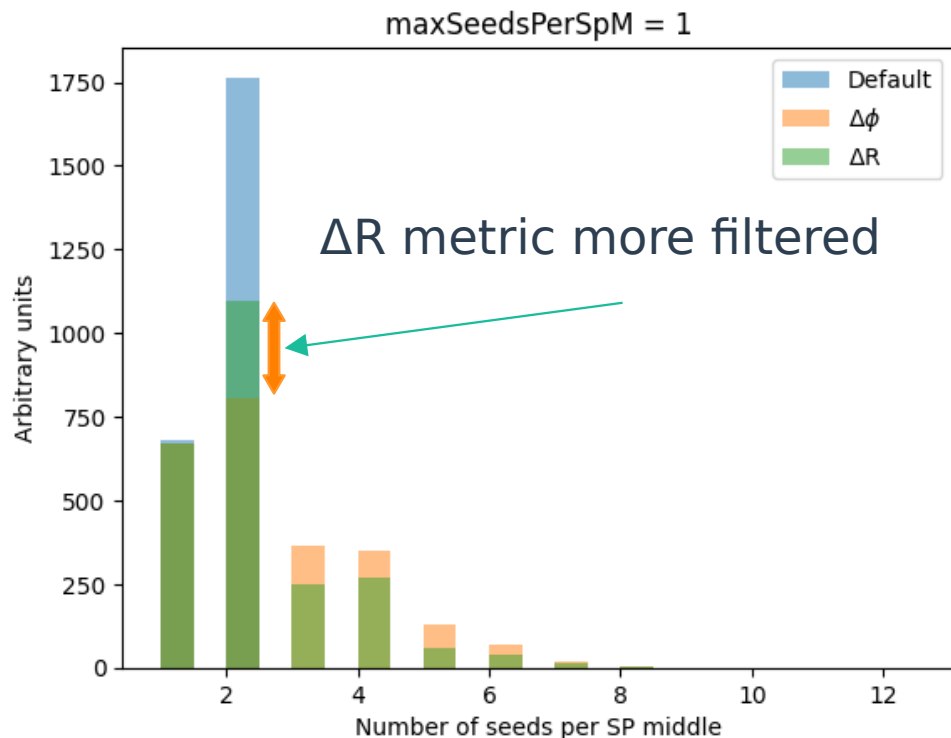
1000 events
in each try

BucketSize: 100
Mu: 50

$\Delta\phi$ is better

MaxSeedsPerSpM and ΔR metric

On 1 event:



Filtered Middle Space points are on the maxSeedsPerSpM bin

Some of the “Buckets shared Middle Space points” are on the bins after the maxSeedsPerSpM bin

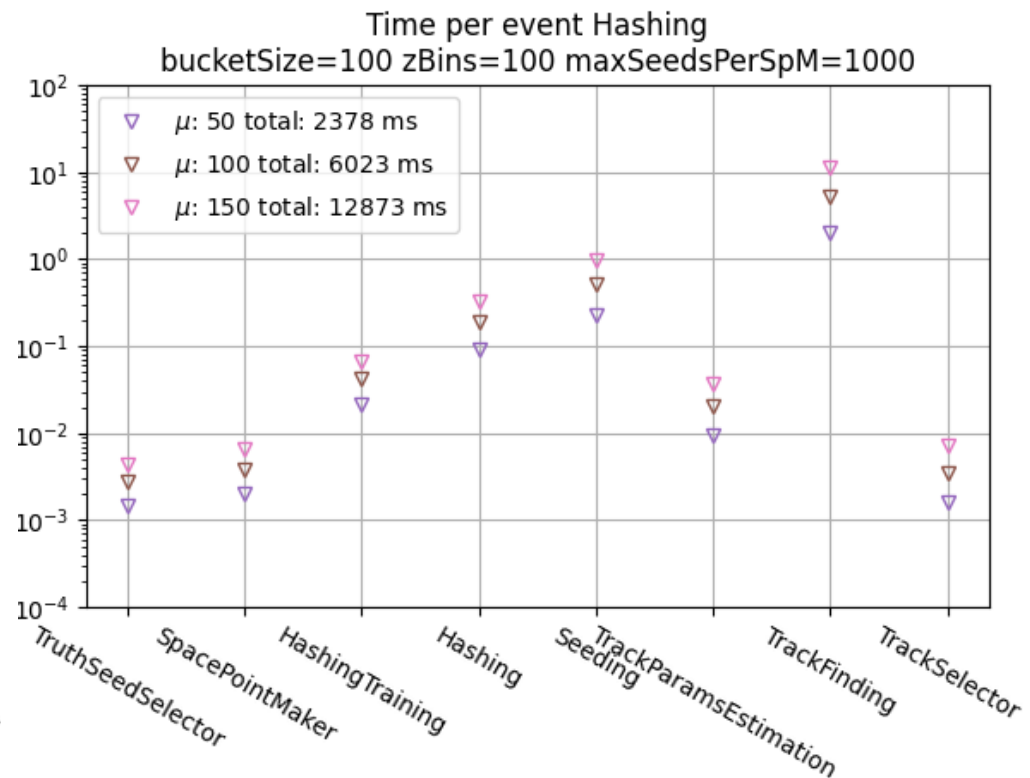
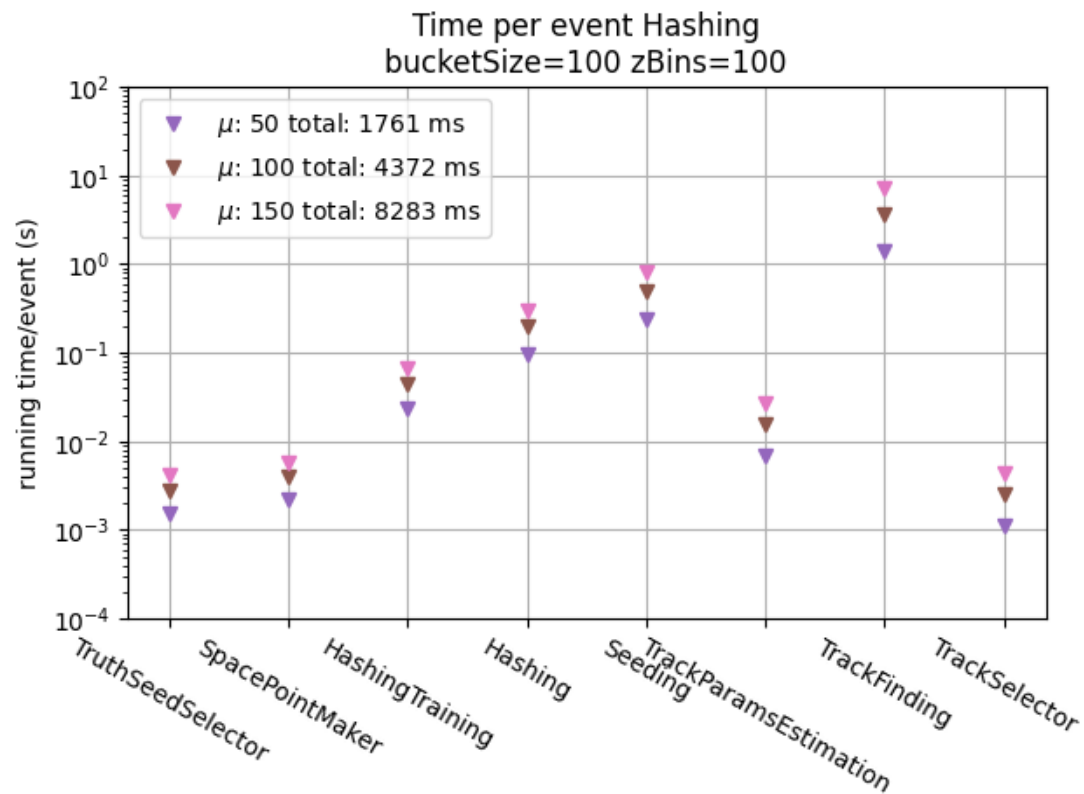
Differences in the bins before maxSeedsPerSpM correspond to lost seeds

Default nSeeds: 4208

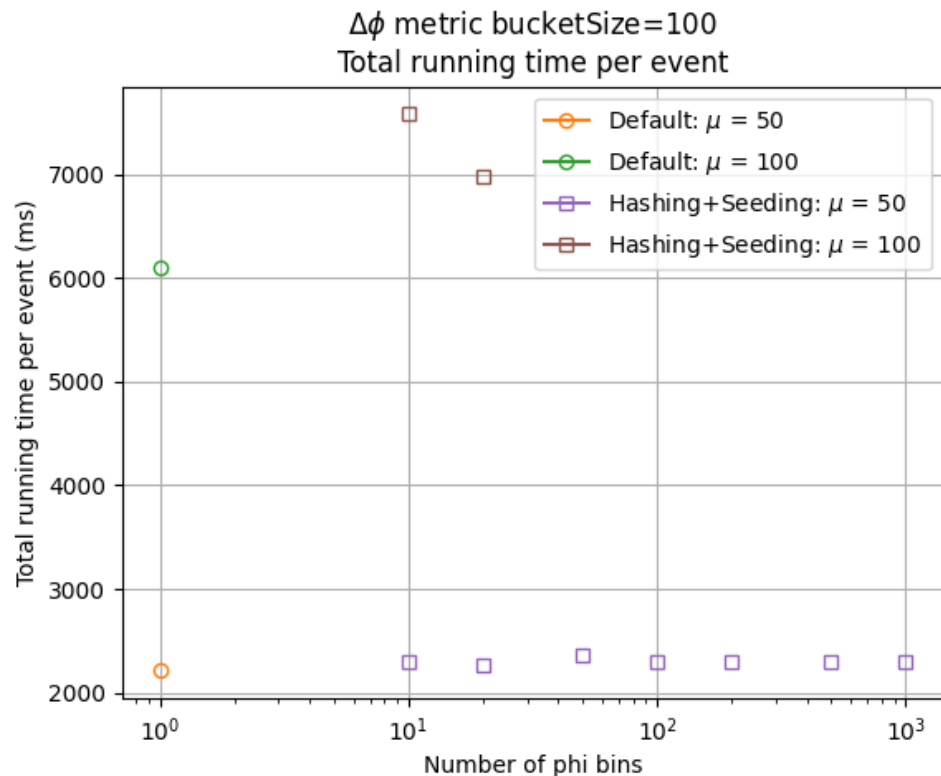
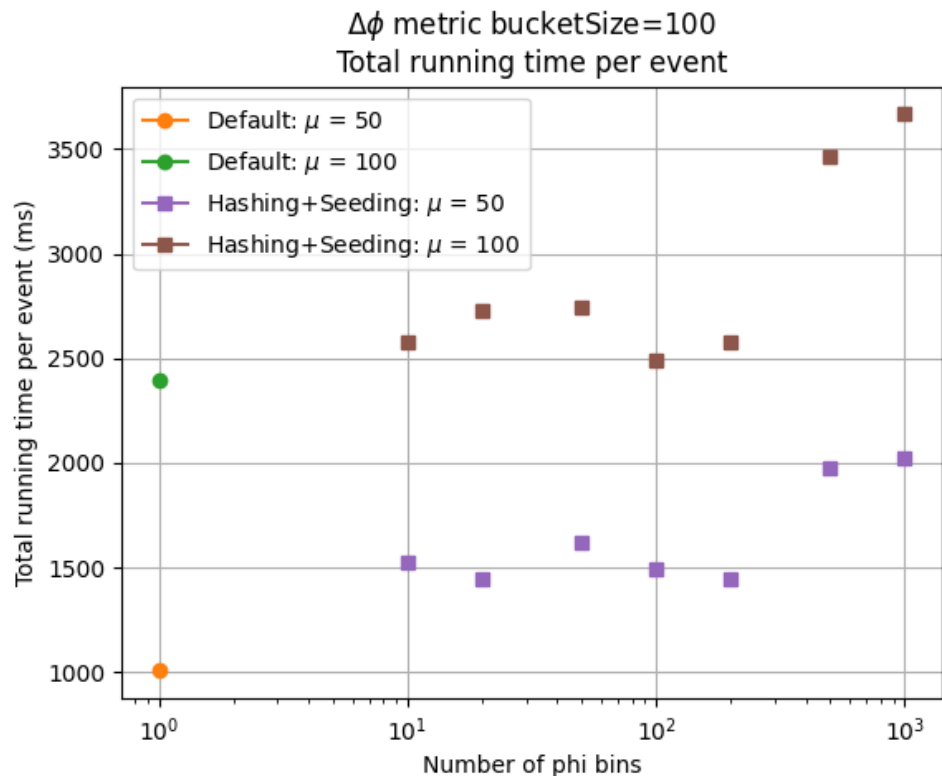
$\Delta\phi$ nSeeds: 6053

ΔR nSeeds: 5300

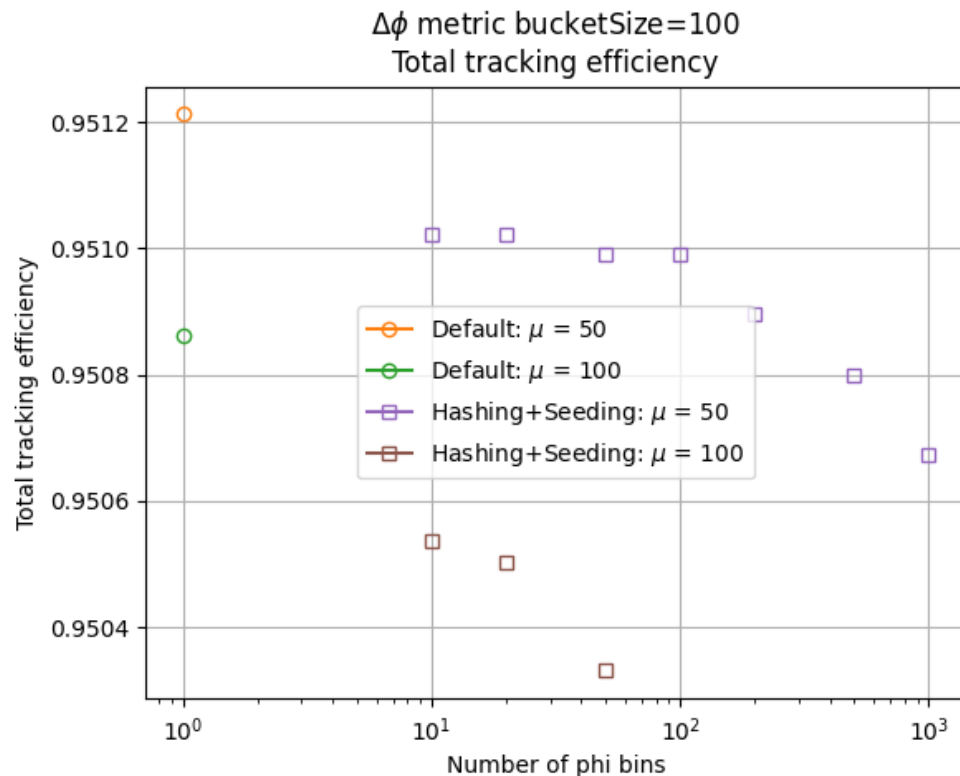
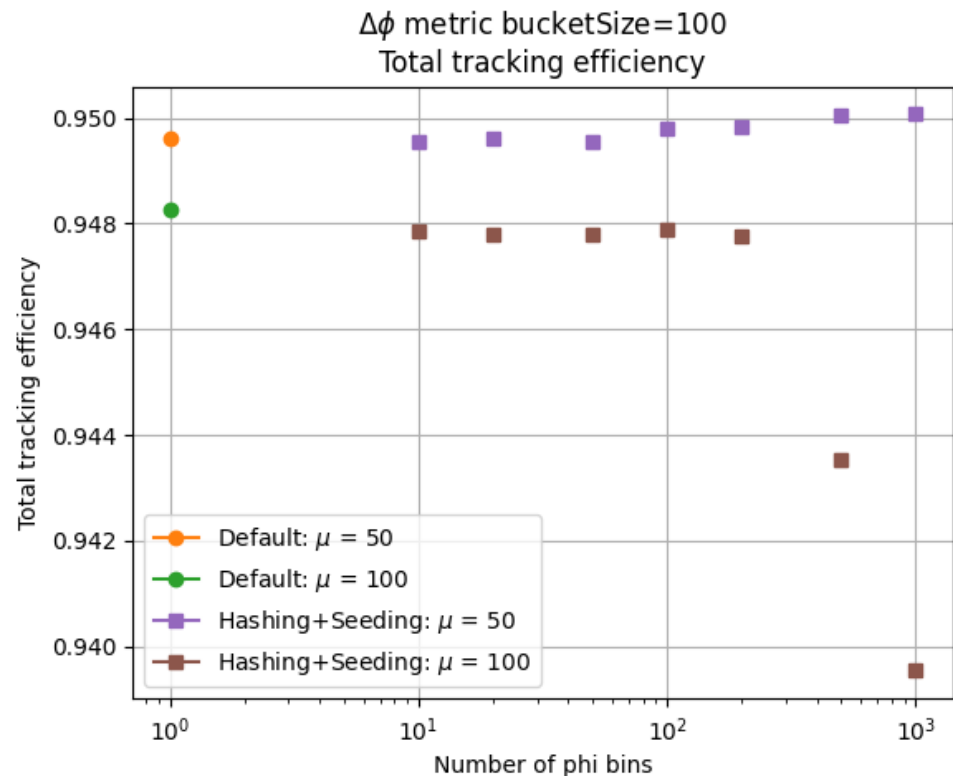
Running time no cut



Phi bins: Timing

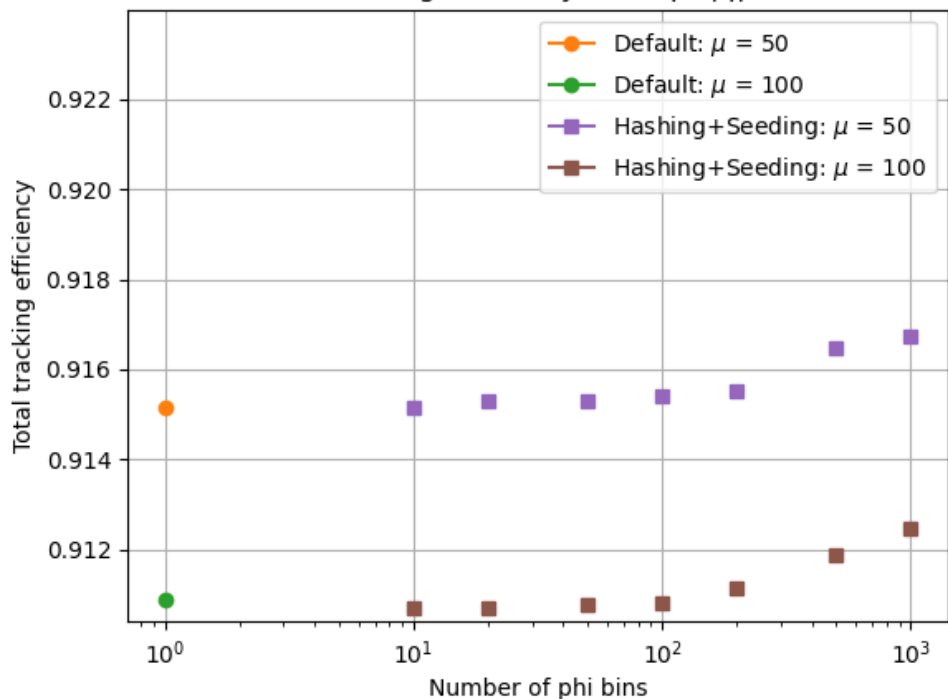


Phi bins: Tracking efficiency

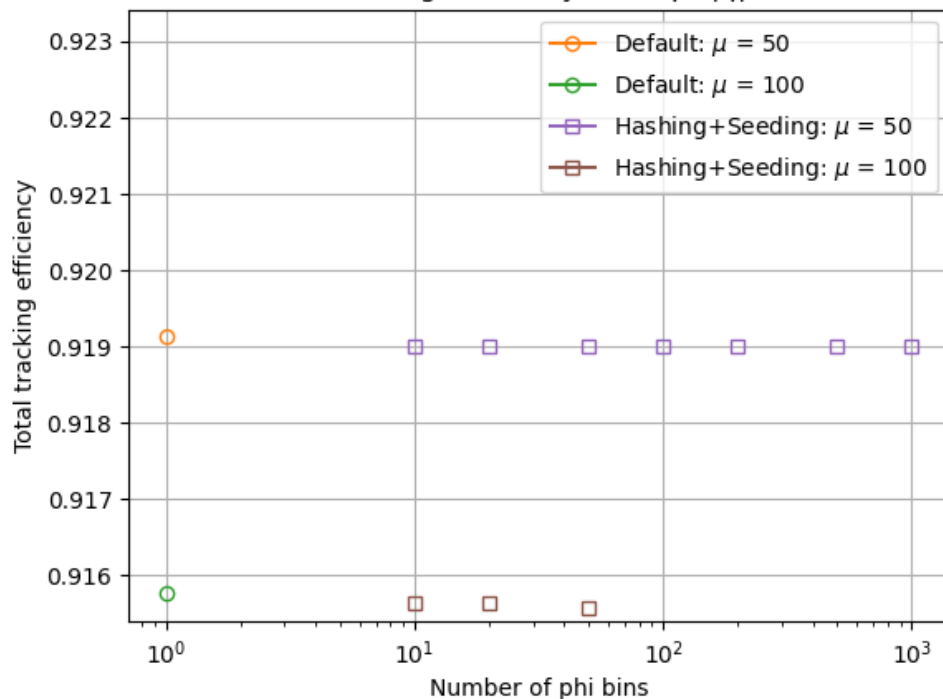


Phi bins: Tracking efficiency

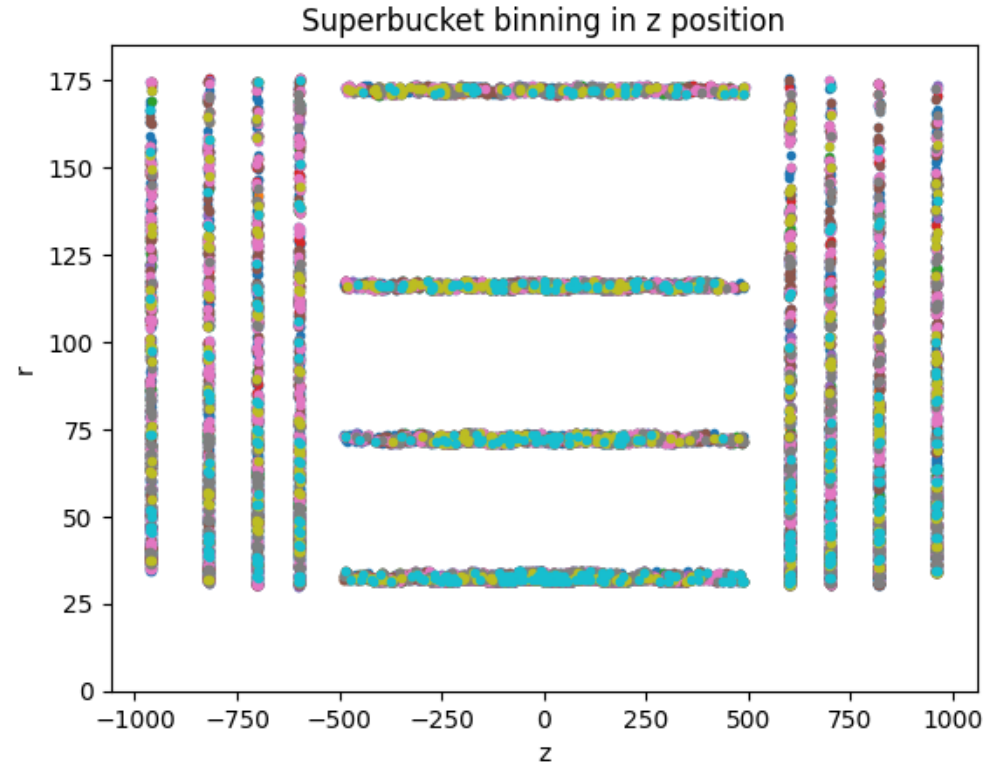
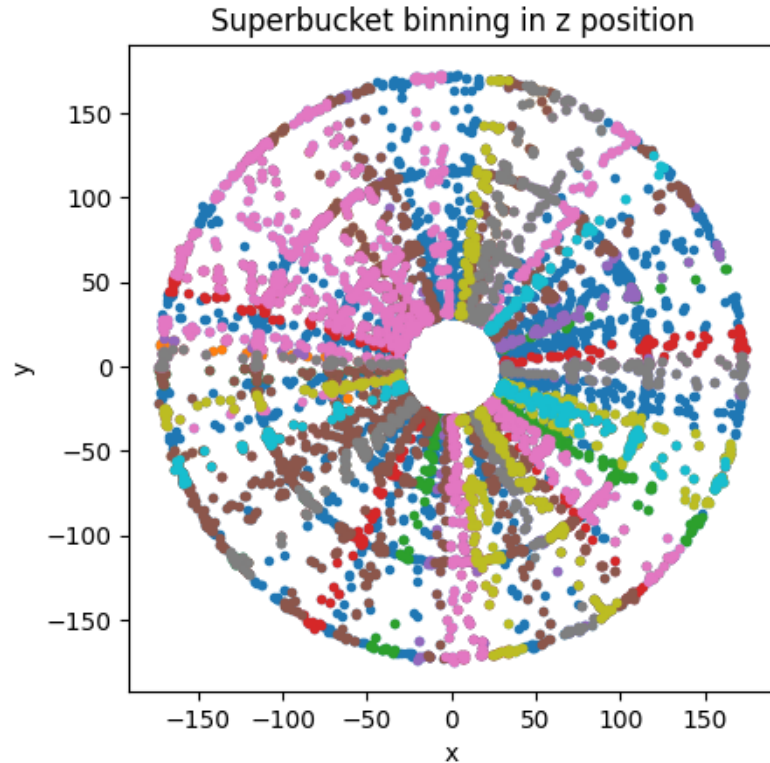
$\Delta\phi$ metric bucketSize=100
Total tracking efficiency endcaps $|\eta| > 2.5$



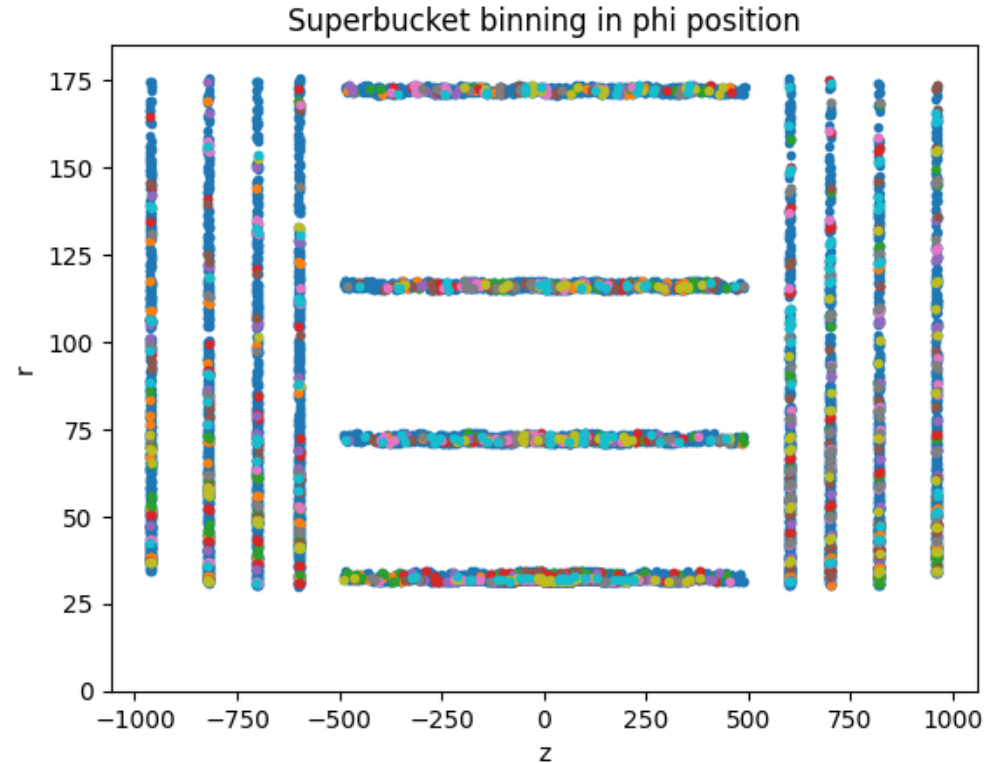
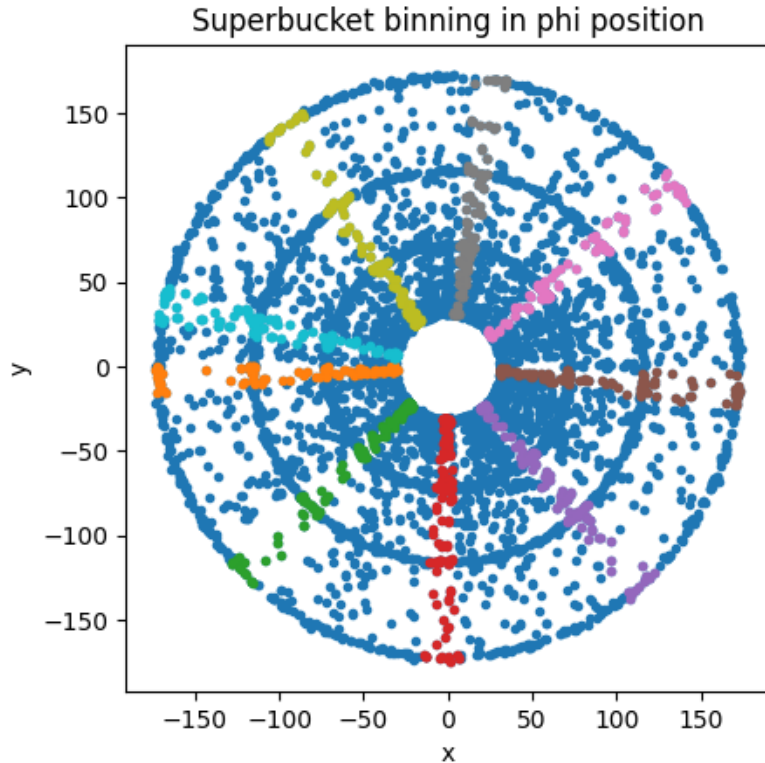
$\Delta\phi$ metric bucketSize=100
Total tracking efficiency endcaps $|\eta| > 2.5$



Superbucket binning in Z position



Superbucket binning in Phi position



Combinatorial problem

Combinatorial Kalman Filter:

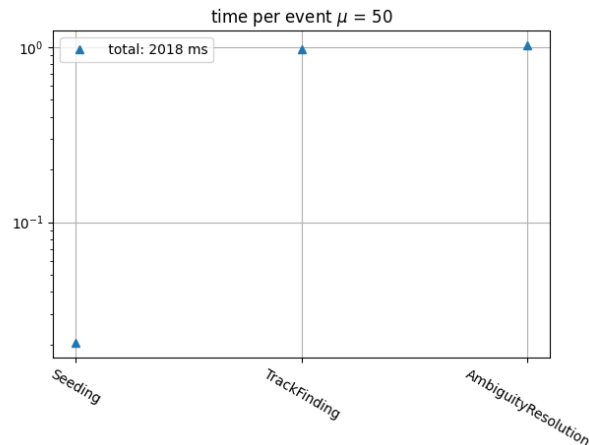
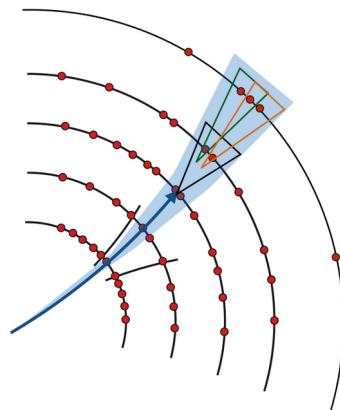
- Several possibilities of expanding the seeds at each layer → need to test them all
- Number of combinations increases exponentially with the number of layers

- **Every seed is expanded:**

- Less seeds → less tracks → less bad quality and duplicated tracks

How to get less seeds?

- Remove the bad ones!
 - How?
 - Current: Filter the seeds + detailed optimisation
 - My work: Build the seeds differently



ACTS Poor man's
Ambiguity resolver