

EIC-ECCE Detector Design Optimization with AI

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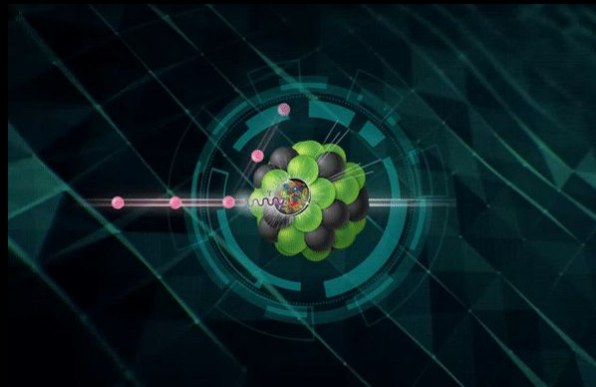
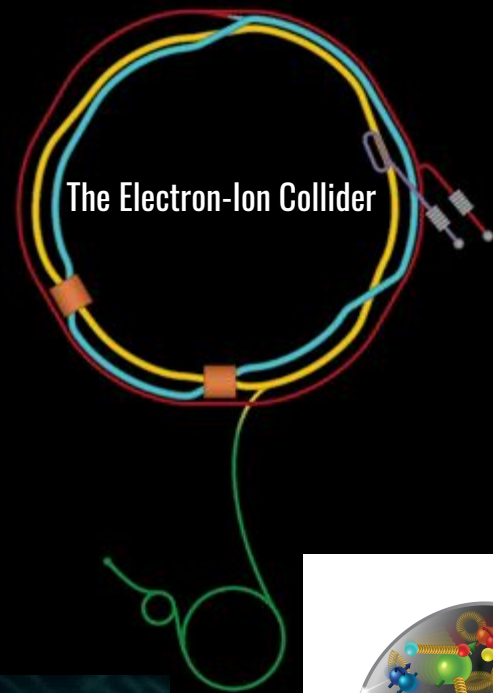


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Outline

- ❑ EIC- Electron Ion Collider
- ❑ “Generic” Detector system for EIC
- ❑ Inner tracking detector in ECCE
- ❑ Multi Objective Optimisation (MOO) and Results
- ❑ Summary and next steps

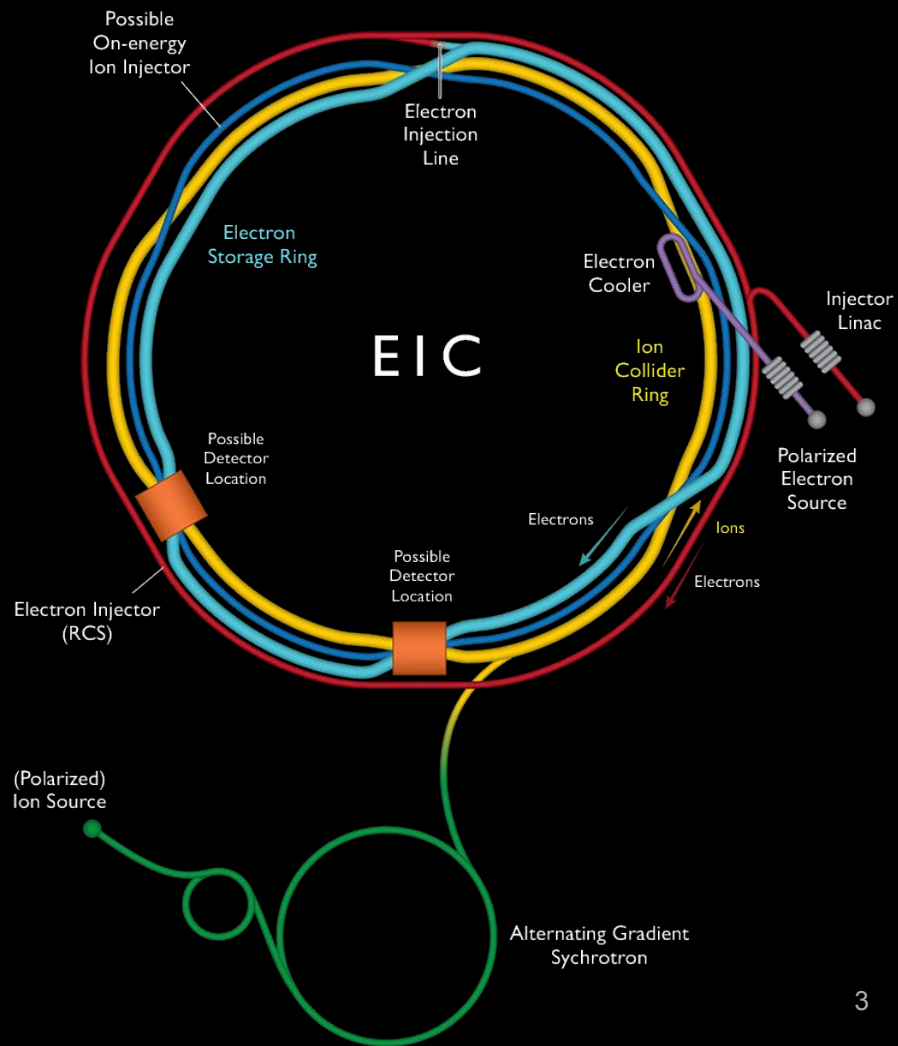


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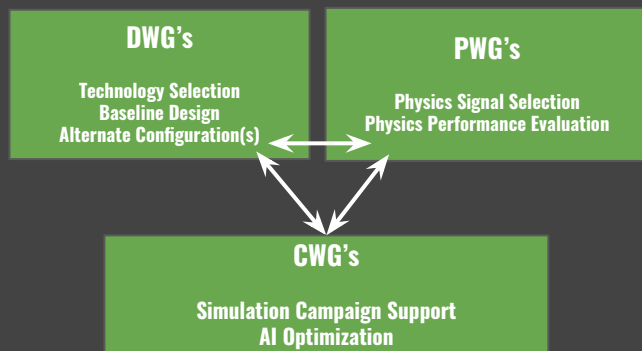
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Electron-Ion Collider (EIC)

- To be built at BNL ([Brookhaven National Laboratory](#)) using existing infrastructure of RHIC.
- **Physics Goal** : Structure and dynamics of matter at high luminosity and energy using polarised beams. Wide range of nuclei [[arXiv:1212.1701](#)]
- The machine will be capable to perform
 - High luminosity measurements ($10^{33} \text{ cm}^{-2} \text{ s}^{-1} - 10^{34} \text{ cm}^{-2} \text{ s}^{-1}$)
 - Flexible center-of-mass energy range. $\sqrt{s} = \sqrt{4E_e E_p}$ (20-140 GeV)
 - Deliver highly polarised electron and proton/ light ion
 - Almost a 4π detector to measure particles scattering in all directions and at wide range of energies



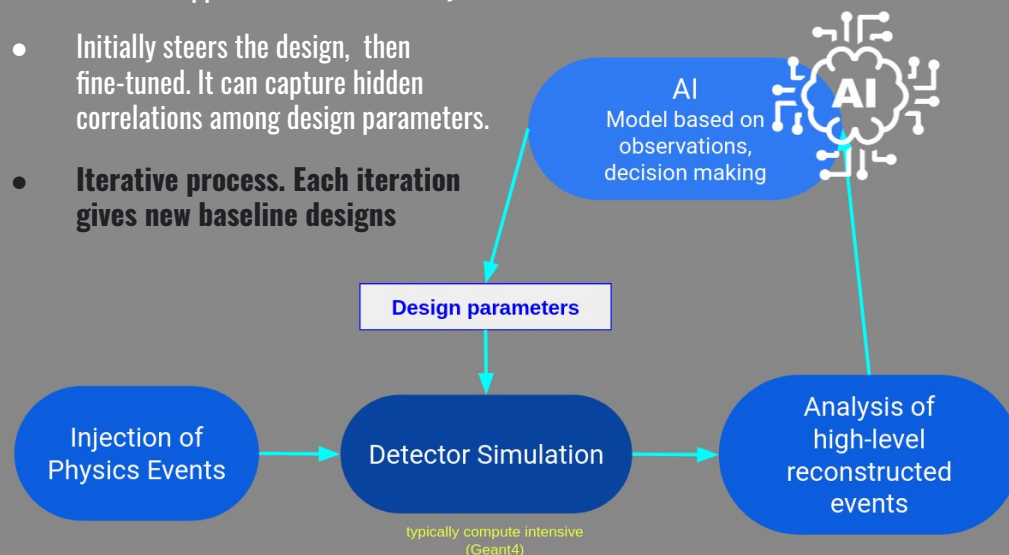
Role of AI



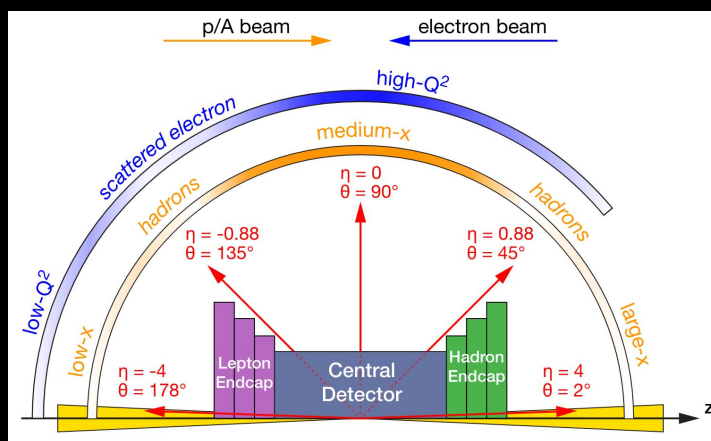
- **Detector Design optimisation, challenging due to dimensionality & constraints**
- **AI deployed at almost all stages of the project**
- **AI4EIC workshop**

- AI can assist in designing more efficiently detectors (performance, costs).
- Potential applications - Medical Physics
- Initially steers the design, then fine-tuned. It can capture hidden correlations among design parameters.
- **Iterative process. Each iteration gives new baseline designs**

Detector Optimisation Workflow

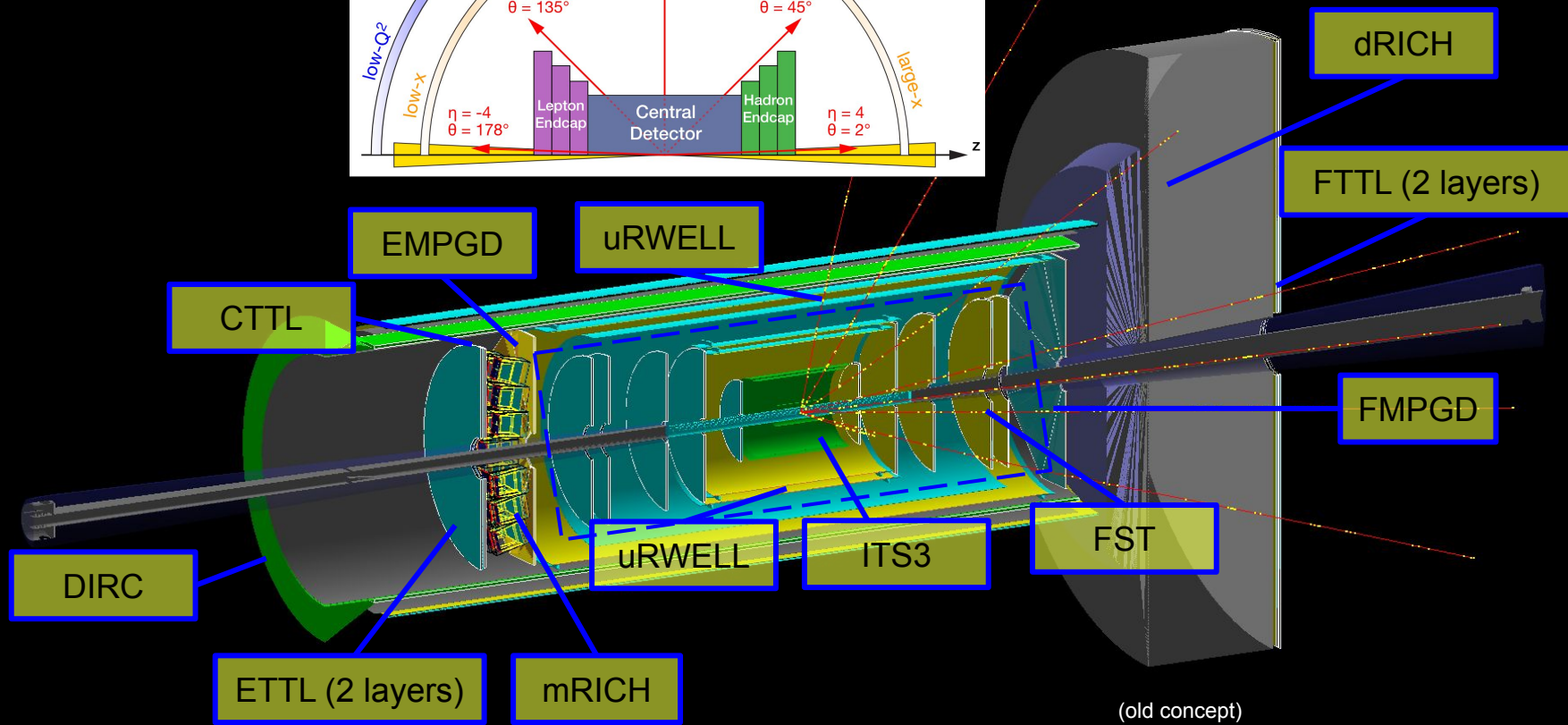


The ECCE Tracking System optimisation



Pseudorapidity

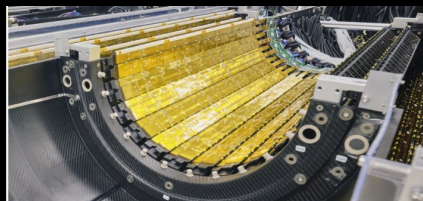
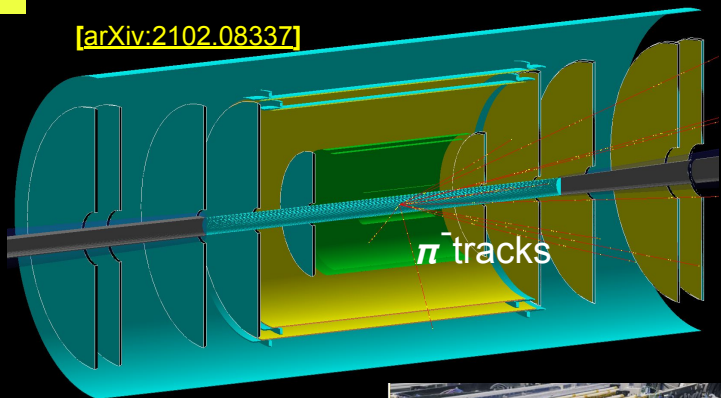
$$\eta \equiv -\ln \left[\tan \left(\frac{\theta}{2} \right) \right]$$



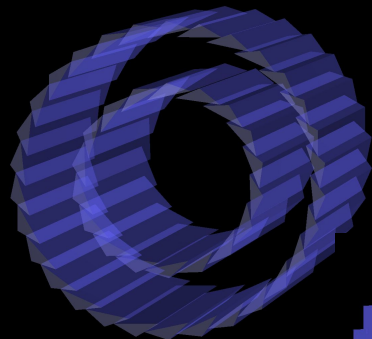
(old concept)

Start with inner tracker

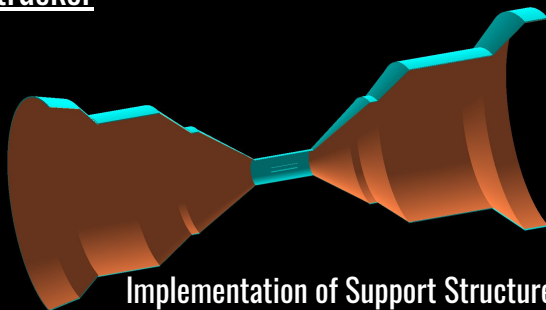
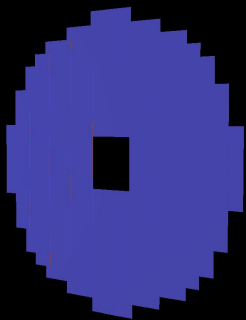
[arXiv:2102.08337]



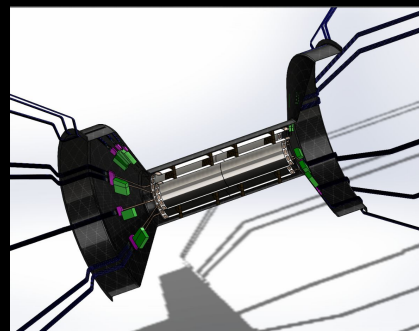
ALICE Si Vertex tracker



Fine grained implementation



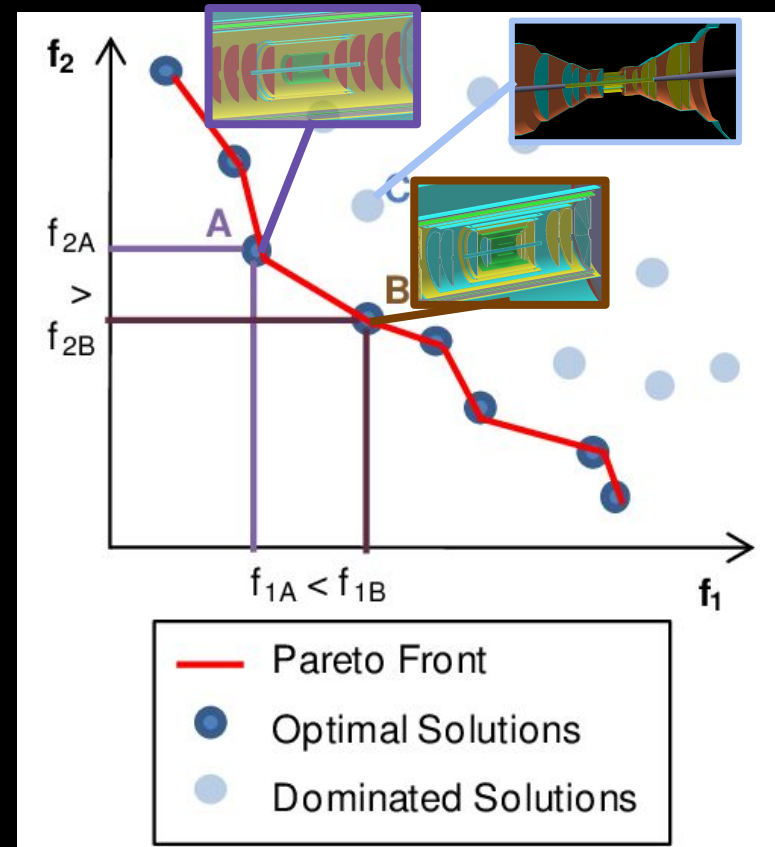
Implementation of Support Structures with realistic material Budgets.



- The performance of tracker characterised by detector's response (eg. resolution, reconstruction efficiency for the tracks). Often more than one metric.
 - Geometric/Design parameters have significant impact in the performance of the tracker.
 - Optimisation is a continuous and iterative process. Each time add more subsystems when available. 11 parameters in this example.
- | | | | | | | | | | | |
|-------|-------|-------|-------|-------|-------|-------|-------|-------|----------|-------|
| B_1 | B_2 | B_3 | B_4 | B_5 | B_6 | B_7 | B_8 | B_9 | B_{10} | B_M |
|-------|-------|-------|-------|-------|-------|-------|-------|-------|----------|-------|
- Efficient parameterisation of the detector to reduce dimensionality of design parameters.
 - Encode different geometric and mechanical constraints; ITS3 (ITS2) constrained due to fixed strip length

Multi Objective Optimization

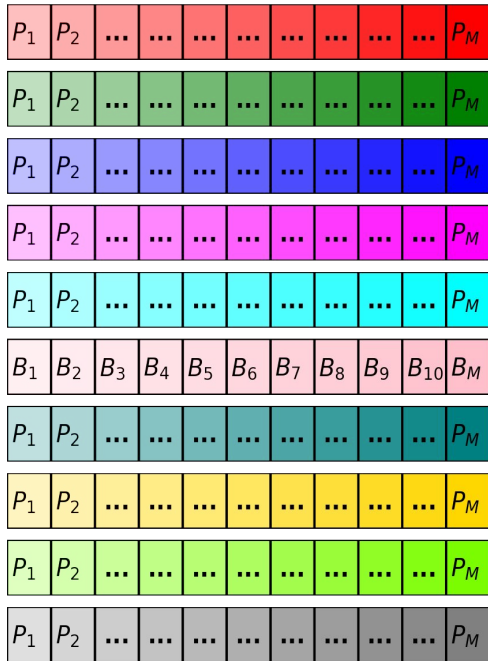
- The performance of tracker determined by multiple “objectives”, e.g., weighted avg momentum resolution, θ resolution, KF efficiency, projected θ resolution at PID location. Objectives could be conflicting.
- In the figure, f_1 and f_2 are two objectives. Points corresponds to one set of design parameters. The objectives are to be minimised.
- In solving such problems, with or without constraints, yields trade-off optimal solutions, popularly known as Pareto-optimal solutions. Locus of points in Objective Space which are non-dominating to one another.
- Due to multiplicity in solutions, Evolutionary Algorithm (EA) is preferred since it uses a population based approach to converge.
- Developed a pipeline for optimisation with [pymoo \(MOGA\)](#) to optimise and “[Fun4All](#)” (Geant4 based framework) to simulate and analyse the detector response.



- N_vars (Design params M) ≥ 11
- N_gen (calls) = 200
- N_pop = 100
- Offspring ≥ 30
- $N_Objs \geq 3$

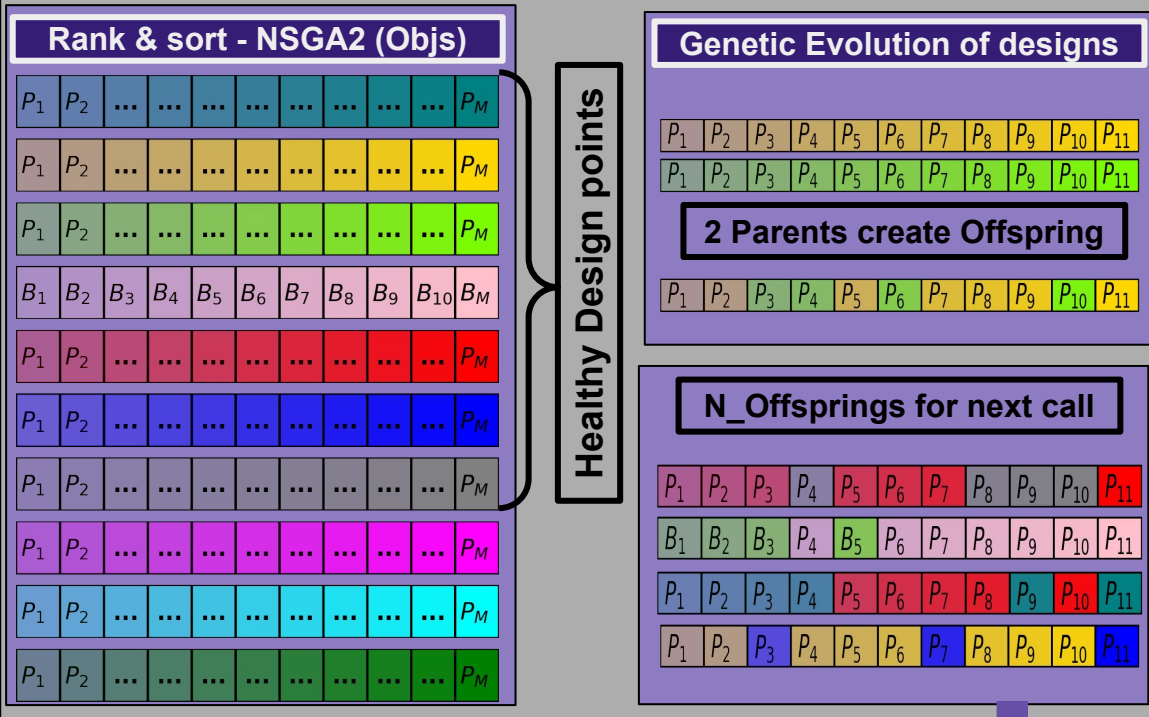
The Summary of MOGA Pipeline

Initial population creation (N_{pop})



Inject baseline Genes
Faster convergence

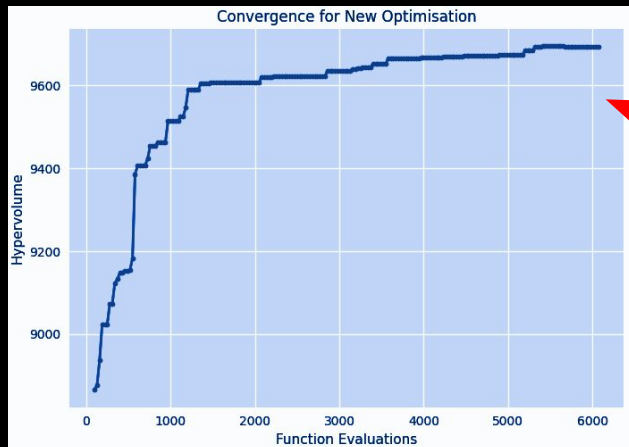
The Evolution Cycle



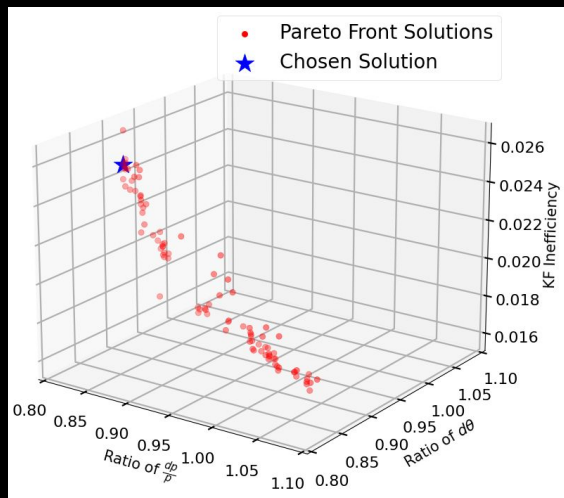
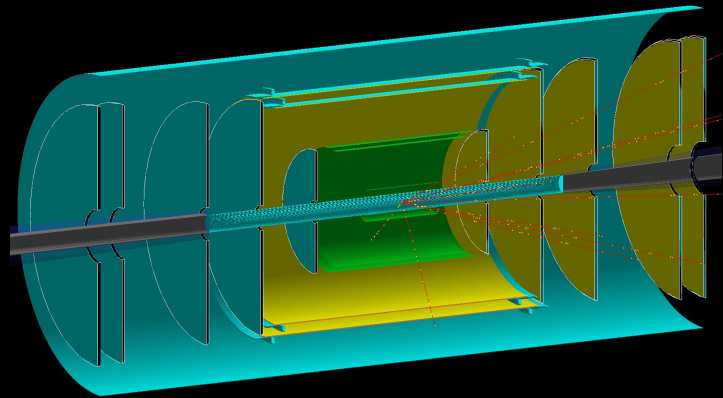
Fun4All Geant4 Simulations

Yields Performance of the design.
Objectives that decide evolution

Optimal Detector Design Solutions

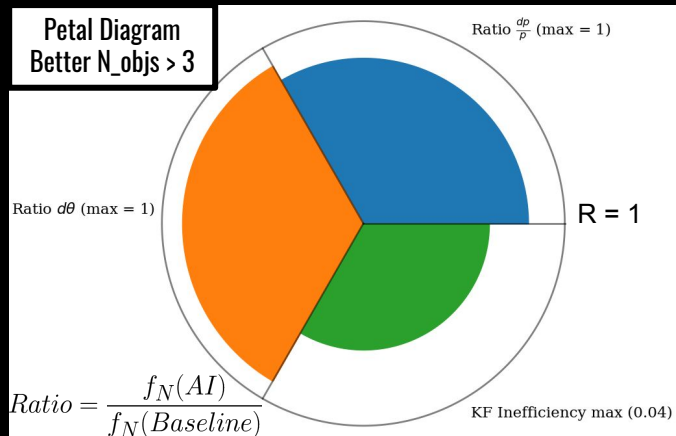


Inferring solutions at any stage of optimisation

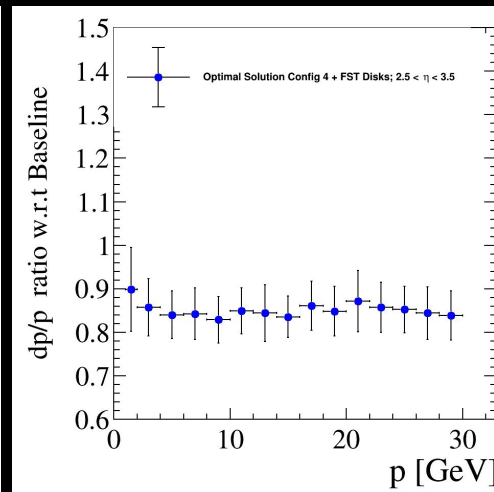
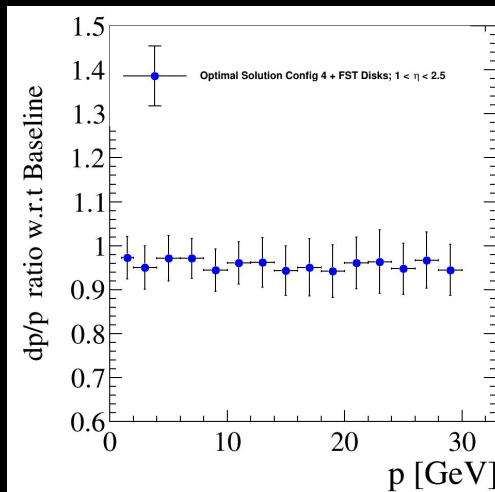
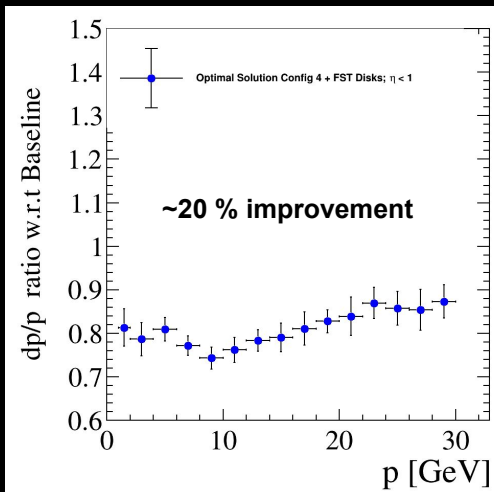
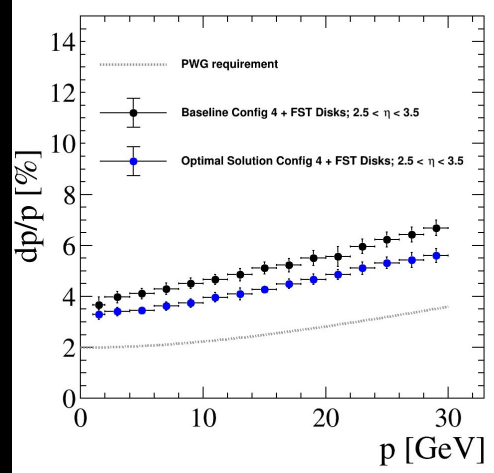
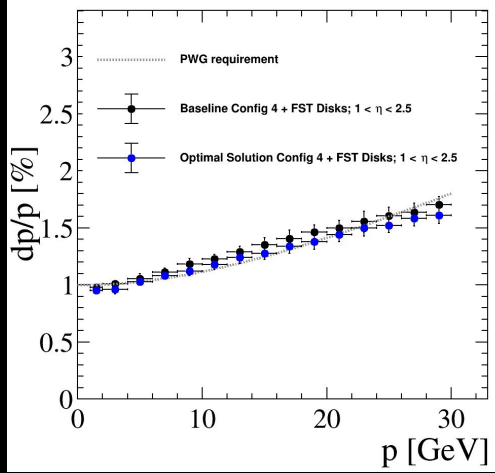
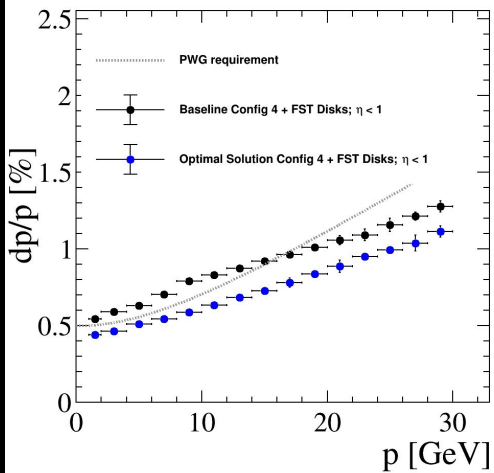


Example of solution

Performance of the chosen Solution



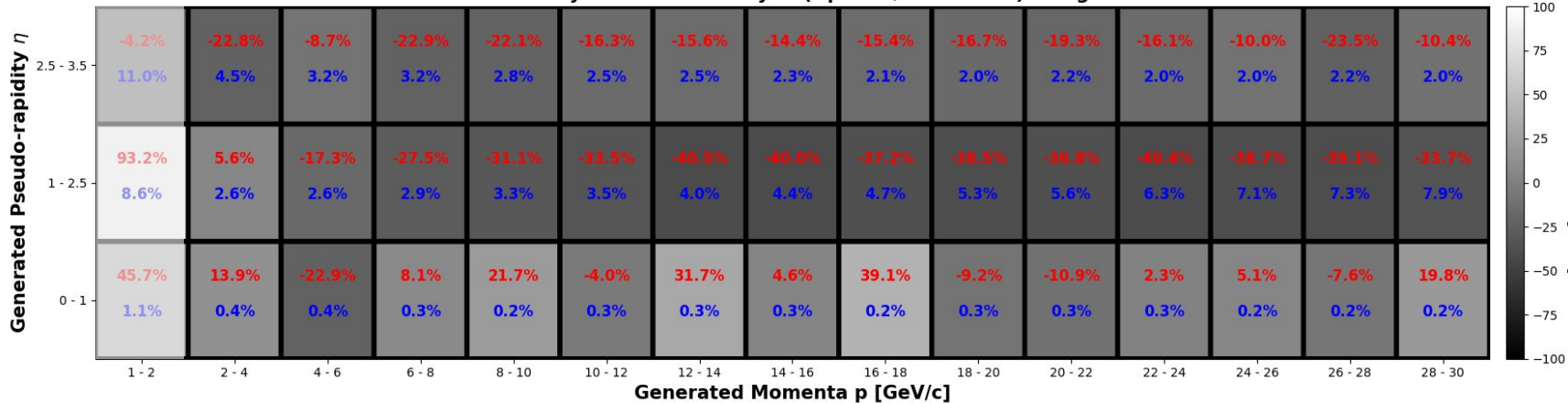
Momentum Resolution



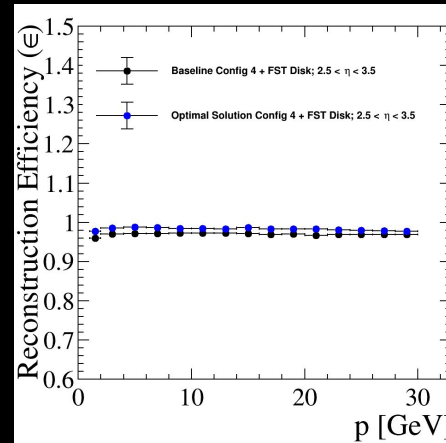
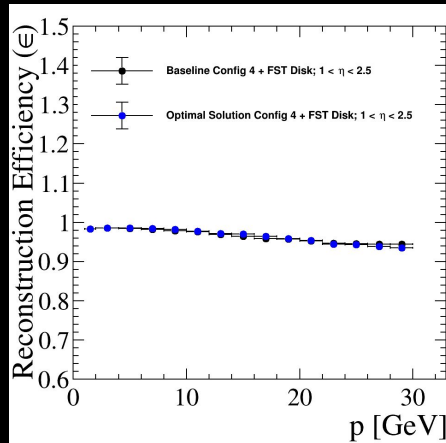
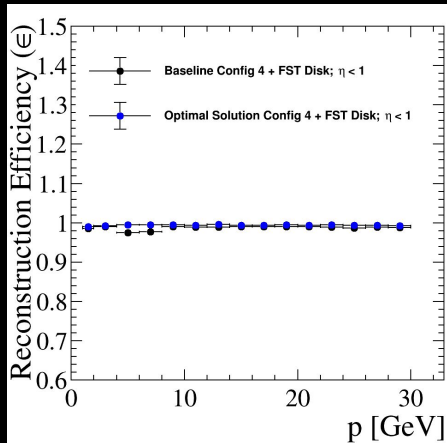
KF Inefficiency Improvement

- Optimal/baseline -1
- Baseline Ineff

Summary of KF Inefficiency of (Optimal/Baseline -1) Design

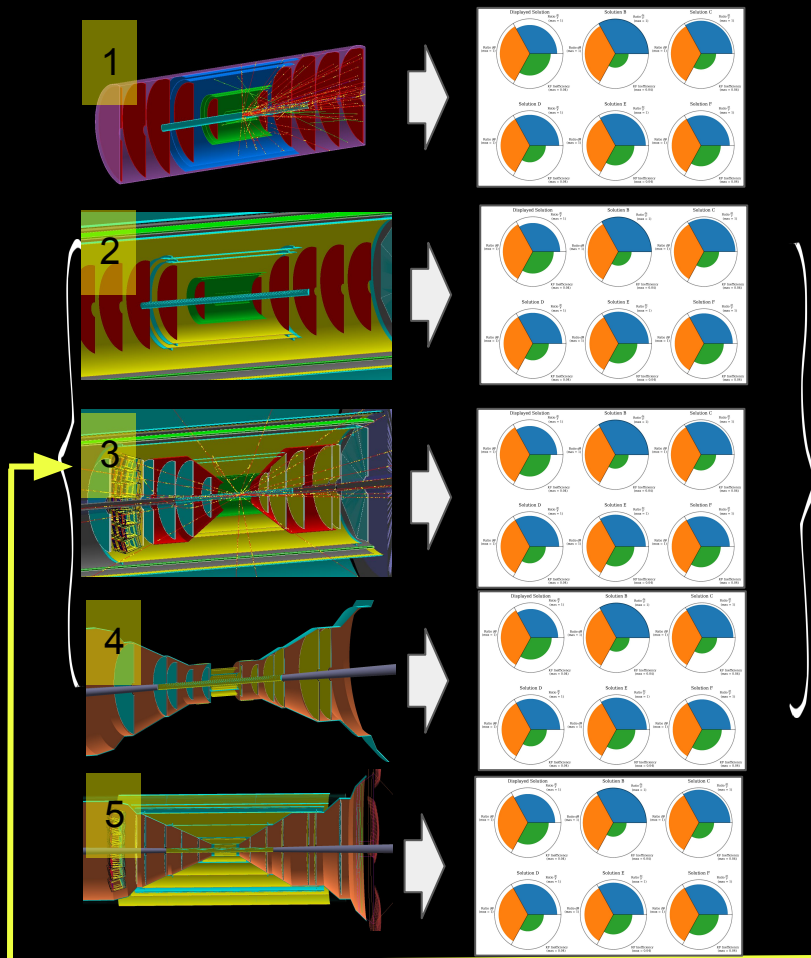


Validation Reconstruction Efficiency



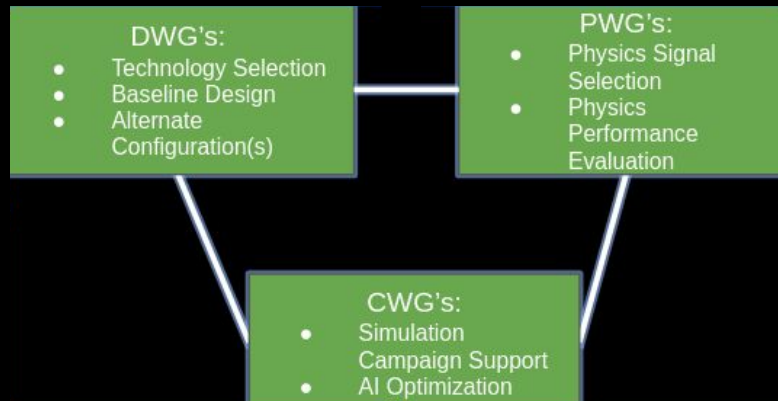
Phases of Optimisation

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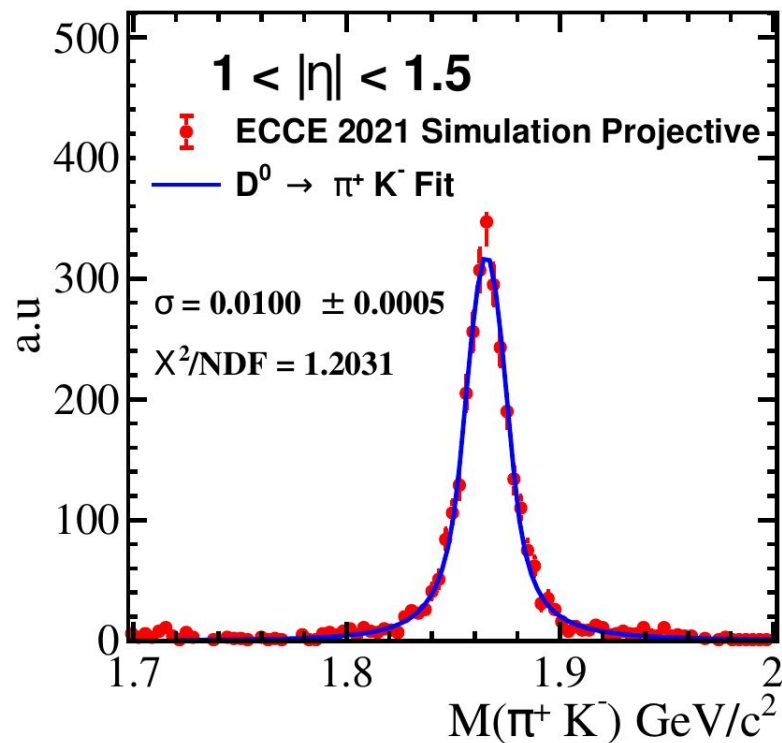
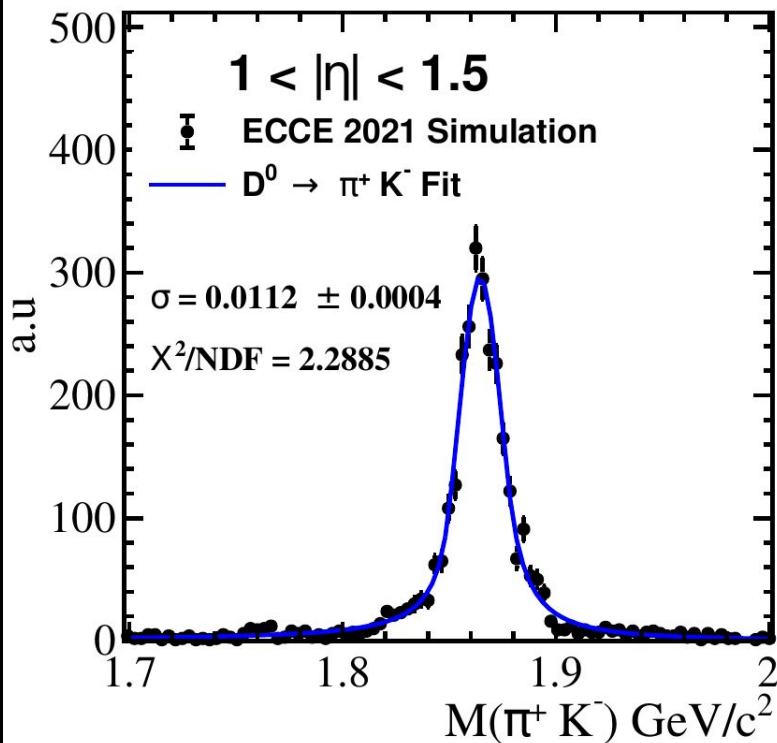


Tracker Optimisation timeline.

- 1: Optimisation of Barrel alone. Made technological choices.
- 2: Optimisation of Barrel+Disks. Without any support structures. Symmetric design
- 3. Optimisation of Barrel+Disks. With fixed support structures. Asymmetric design
- 4. Optimisation of Barrel+Disks and support structure. Asymmetric design
- 5. Full tracking system optimisation.



updated configurations with any additional requirements
Optimisation phases



The $\pi^+ K^-$ invariant mass obtained from the SIDIS events with updated baseline and recent optimised projective geometry. A region of eta that is sensitive due to considerable materials for support structure was also taken in to account for this optimisation.

Summary

- **EIC can be one of the first experiments to be designed with the support of AI**
- The current tracking system in proposed by ECCE for EIC is an AI-assisted one.
- For the “first” time a framework integrating the **full** detector design using Geant4-based simulation coupled to MOO has been developed. This framework can be massively parallelized.
 - The developed framework is modular enough it can be effectively used within ECCE EIC software stack.
 - The decision making from the Pareto solutions are intentionally manual to effectively explore the feasibility of the design based on qualitative factors like cost, engineering realisation etc.
- ECCE-EIC tracker optimisation is a continuous and an iterative process. AI had assisted in making technological choices for the tracker design and also had assisted in the current design of the ECCE-EIC tracker.
- Currently developing framework to include different optimisers (e.g. MOBO, MOEA.)



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Thank You

