



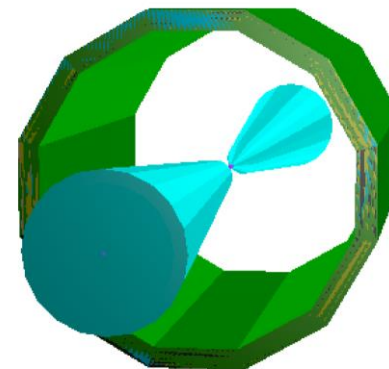
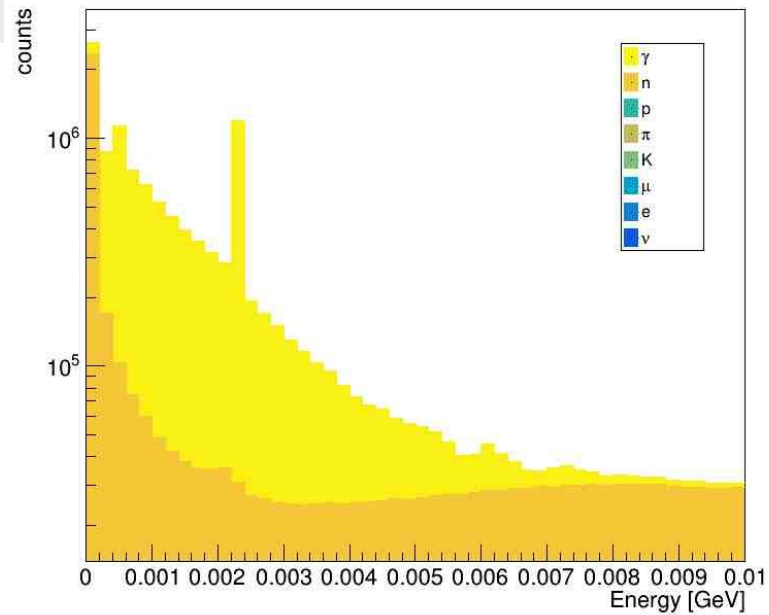
Towards design studies of a Muon Collider ECal

Federico Nardi, Tommaso Dorigo, Julien Donini



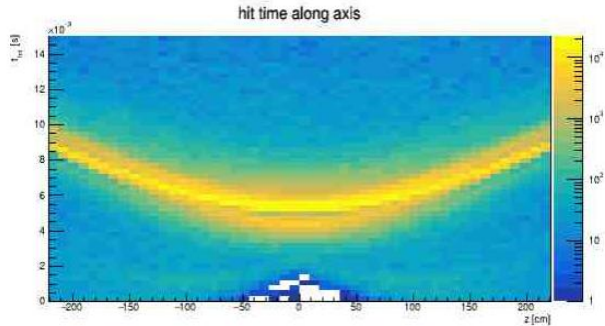
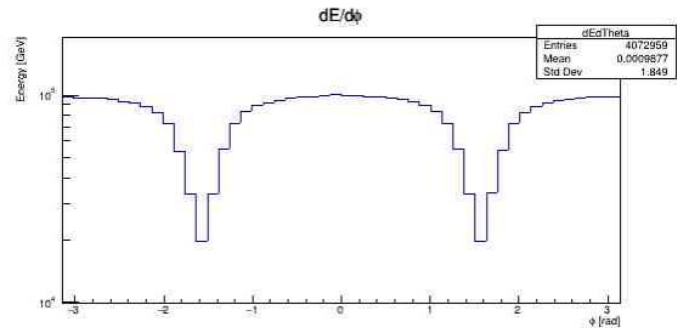
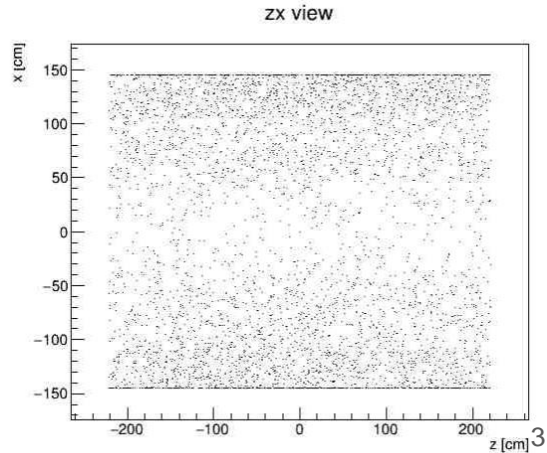
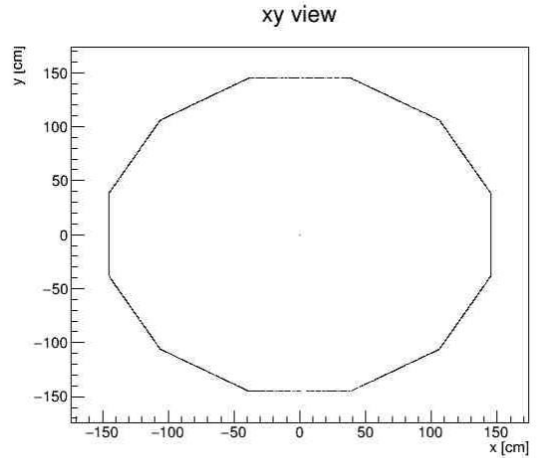
BIB - photons

- Data from MARS15 simulation: interactions with nozzle
- Focus on photons component
- Using Crilin design (v1)
 - 1x1x4 granular cells (neglect electronics)



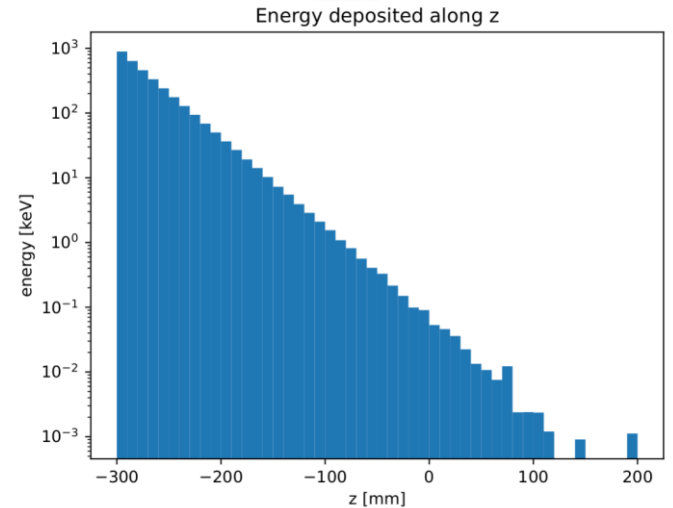
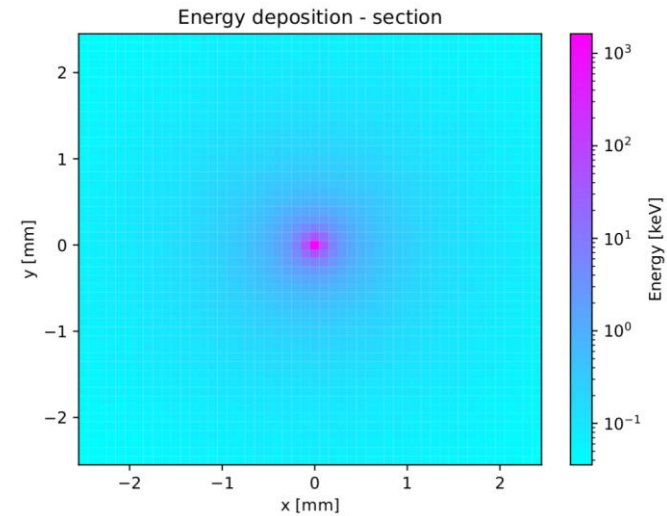
Photons II

- Hit distribution, propagated to barrel surface
- Symmetry in $\phi=0$ plane, less hits on top and bottom edges
 - (Explanation still unsure)
 - Non-homogeneous -> optimization might give interesting results
- Well defined hit time distribution -> interesting for later studies (i.e. optimal placement of timing layers)



Energy deposition in PbF₂

- Geant4 simulation of photons through a PbF₂ block
- Particle energies selected using BIB spectrum
 - Successively scaled depending on position inside the detector
 - Can do it if flux is high enough
- Radial symmetry with respect to beam axis
 - Can obtain a model of energy deposition $f(r|z)$ dependent only on z -coordinate and distance from z -axis



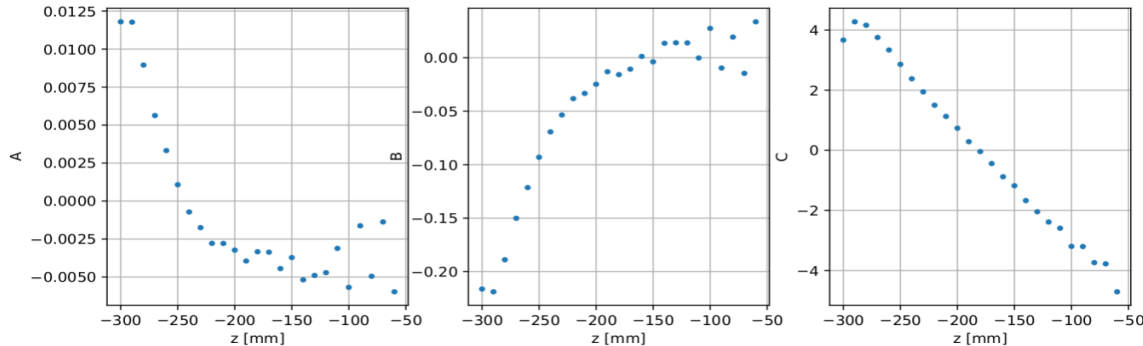
Parametrization of energy deposition

- 25 bins with z-values
- Radial distribution obtained by fitting on each bin

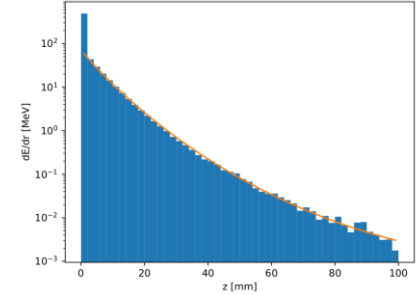
$$f(r) = A \times \mathcal{N}(b, c)$$

- Z-dependence of fit parameters allows to define a continuous and differentiable parametrization to run optimization cycles

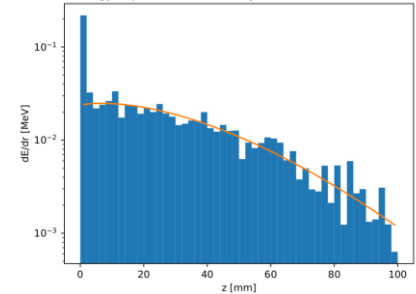
Fit Parameters



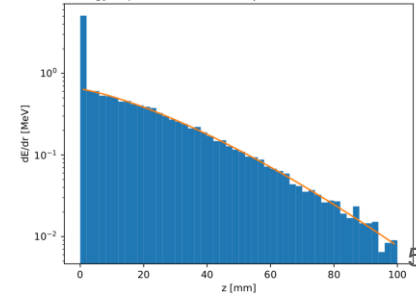
Radial energy deposition - Z = -290.0 | R2 score: 0.9937261435714658



Radial energy deposition - Z = -80.0 | R2 score: 0.9010992315125068

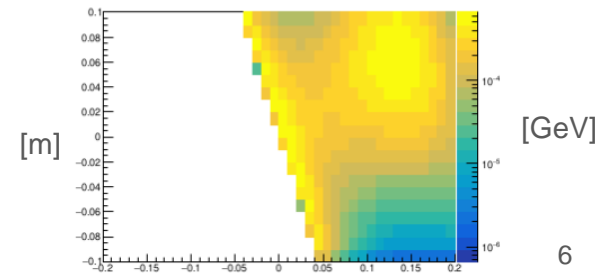
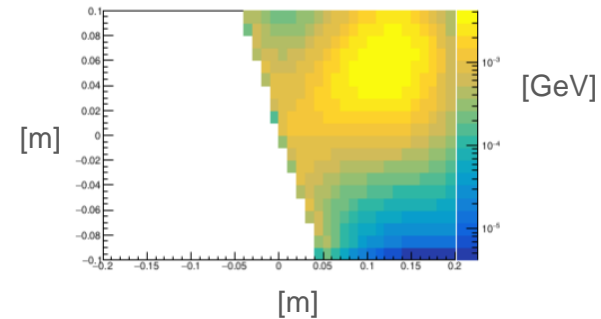
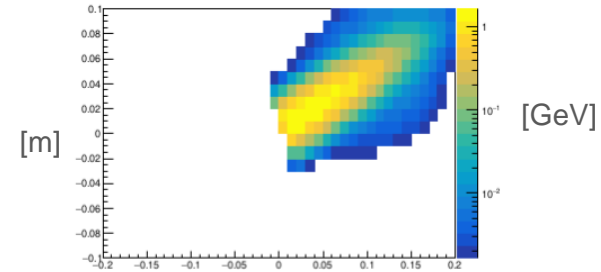
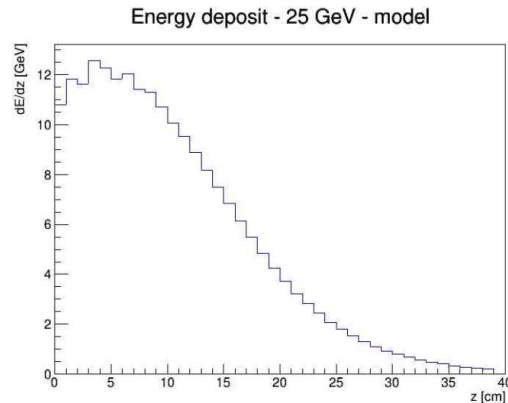
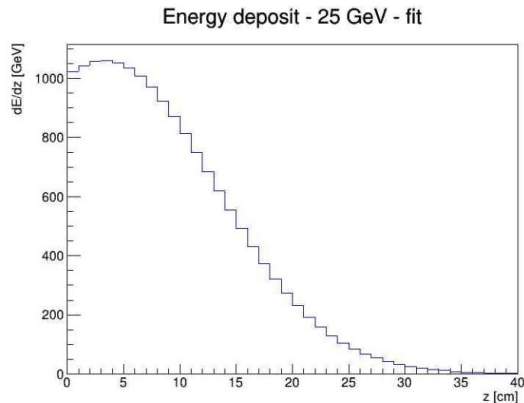


Radial energy deposition - Z = -170.0 | R2 score: 0.997403449668706



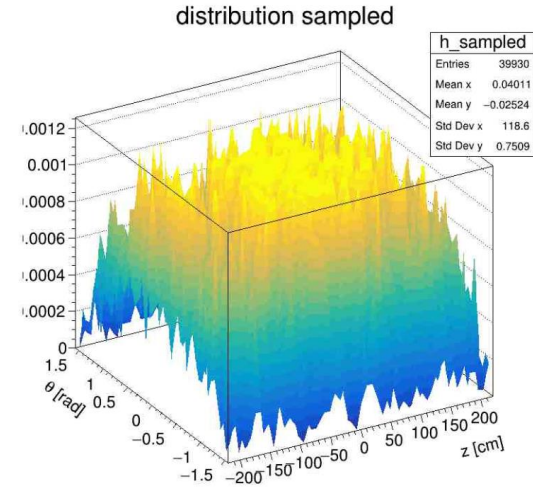
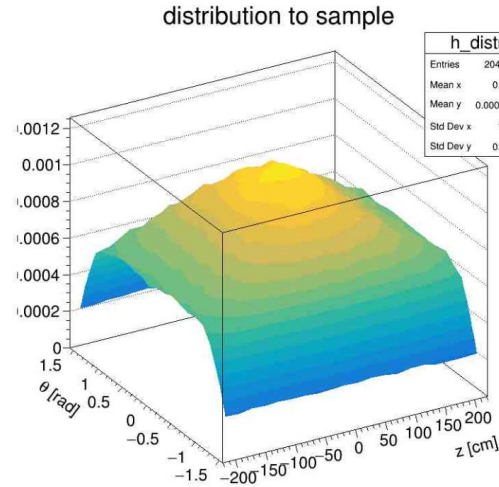
Simulation of detector layers

- Parametrization for both BIB and signal
 - Monochromatic photons with random angle from IP
- Normalization enforced to match Geant4 deposition



Full detector simulation

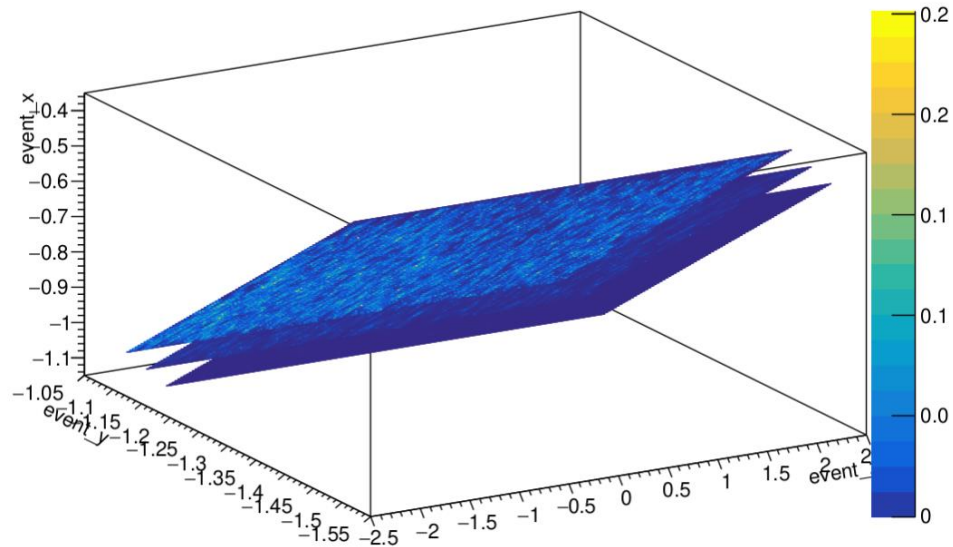
- Basic MC sampling of hit position on surface
- $4e5$ photons generated (correct by a factor 10x)
- Signal area stored on root ntuple



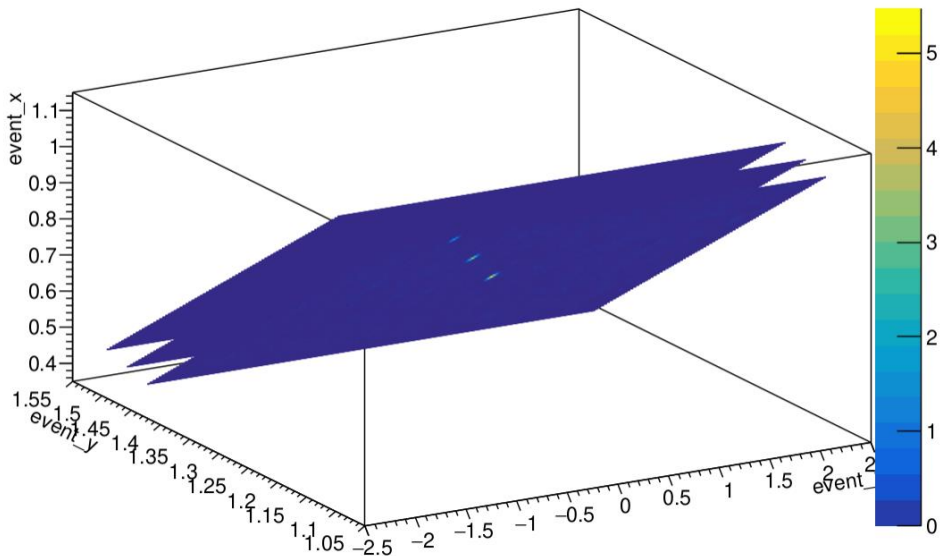


Event visualizations

event_x:event_y:event_z:event_dE



event_x:event_y:event_z:event_dE





Still some issues

- Strong dependence of deposition on incidence angle
 - Improves with denser voxels
- BIB generation slow -> $O(\text{hours})$ per single signal event
 - Keep stochastic nature of BIB
 - Switch to a energy flux modelling instead of event hits

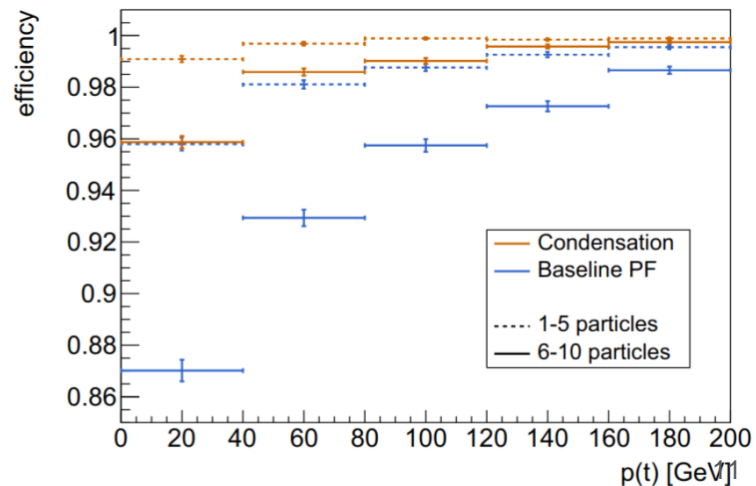
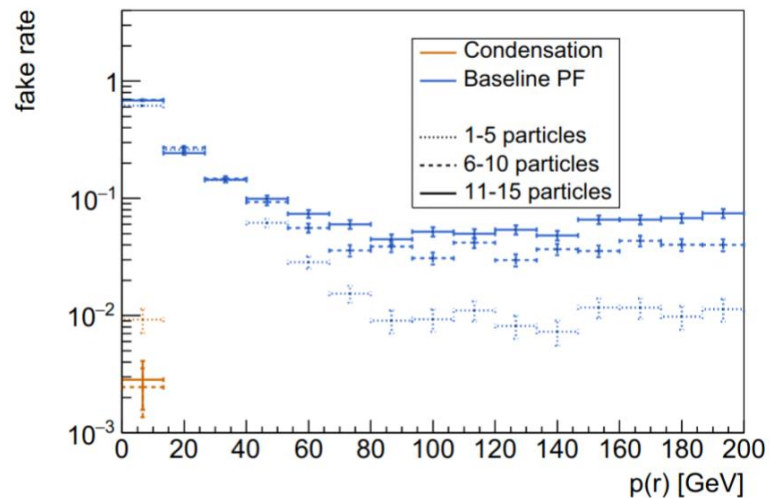
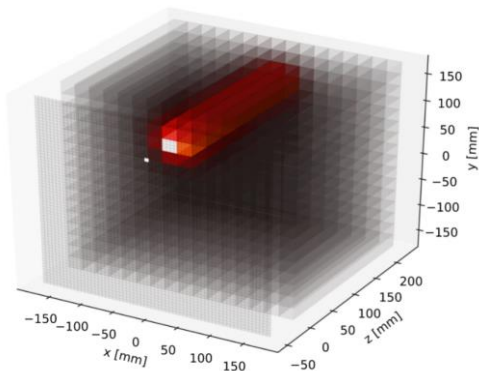


Object Condensation for reconstruction

- Need a differentiable reconstruction engine
- Keep (or improve) original efficiencies
- Use Deep Learning techniques, analogous to image recognition methods
- DeepJetCore -> Library developed for jets at CMS HGCal
 - J. Kieseler, *Object condensation: one-stage grid-free multi-object reconstruction in physics detectors, graph, and image data*, EPJC **80** 866 (2020)

DeepJetCore performance

- Electrons and photons from 1 to 200 GeV
- Granular bulk of PbWO4 cells, with tracker in the front
- Compared with PF algorithm





Status

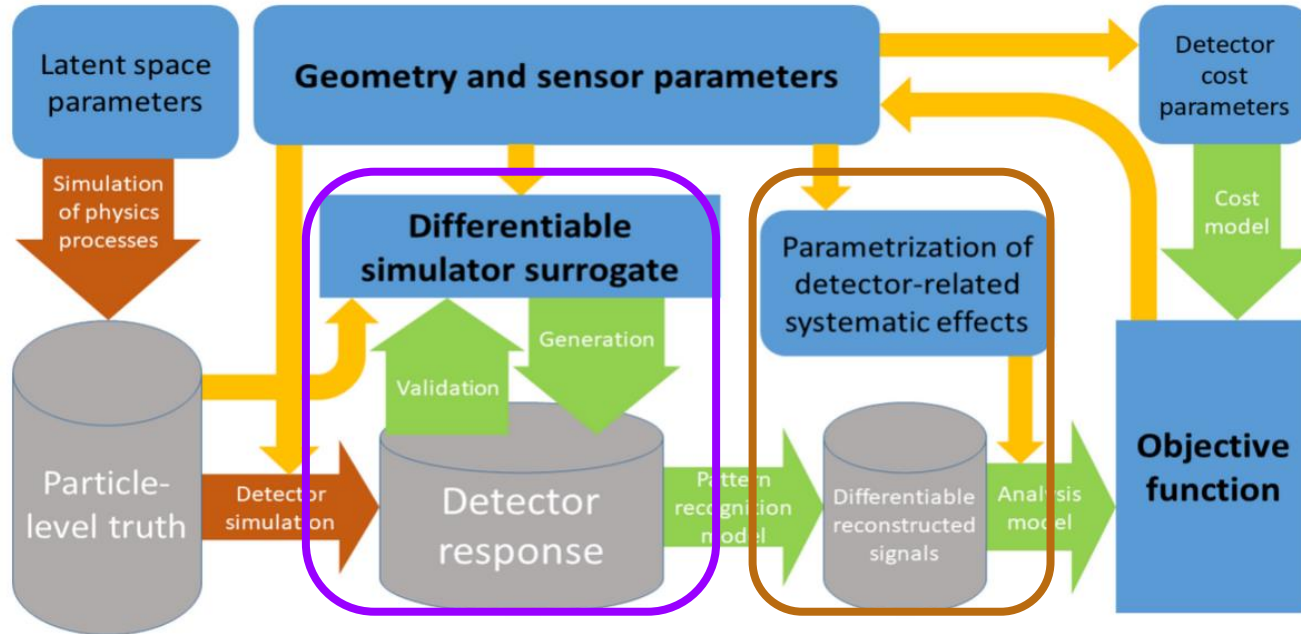
- Training ongoing for monochromatic photons in Crilin ECal
 - With and without BIB (3TeV)
- Aim: present first results in the next weekly meetings

Towards optimization of ECAL design

- MODE Collaboration: Machine-Learning-Optimized Design of Experiments
 - <https://mode-collaboration.github.io>
- Idea: Use automatic tools to come up with better solutions in experimental setups
 - Heuristics and intuition work great, but what if we can approach it in a more systematic way?
- 4 elements needed:
 - Event generator
 - Simulator of detector response
 - Object Reconstruction algorithm
 - Loss function
- Note: Every part needs to be differentiable, we need automatic differentiation to minimize the loss function!



Design optimization: how?





Backup

TMOPT - muon tomography framework

WORK IN PROGRESS!

From G.Strong's presentation at ICHEP2022

- Task is to infer presence of uranium block in container filled with scrap metal
 - Inference uses a dedicated summary statistic
 - The U block can be anywhere in the volume, so intuitively expect the detectors should be placed centrally in XY over the volume
- Detectors start in corner of volume and optimisation does indeed move them to cover the volume

- Optimised detector provides large improvement to ROC AUC

