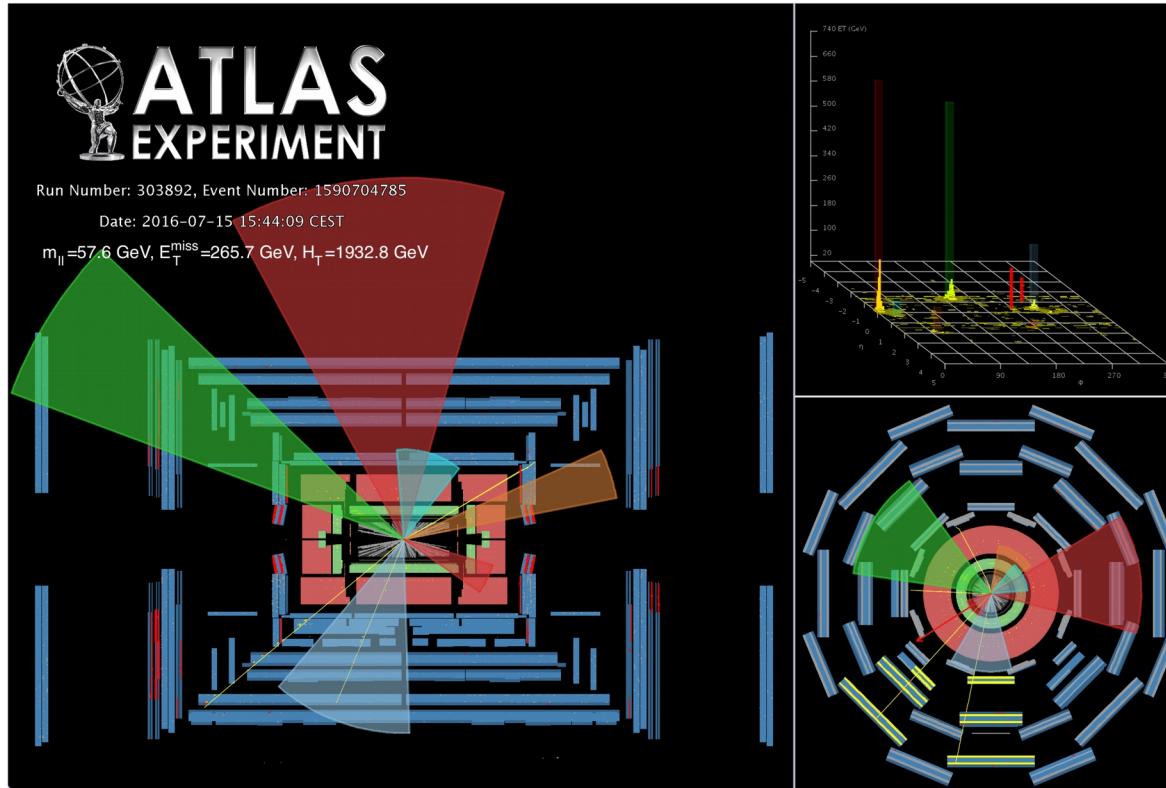


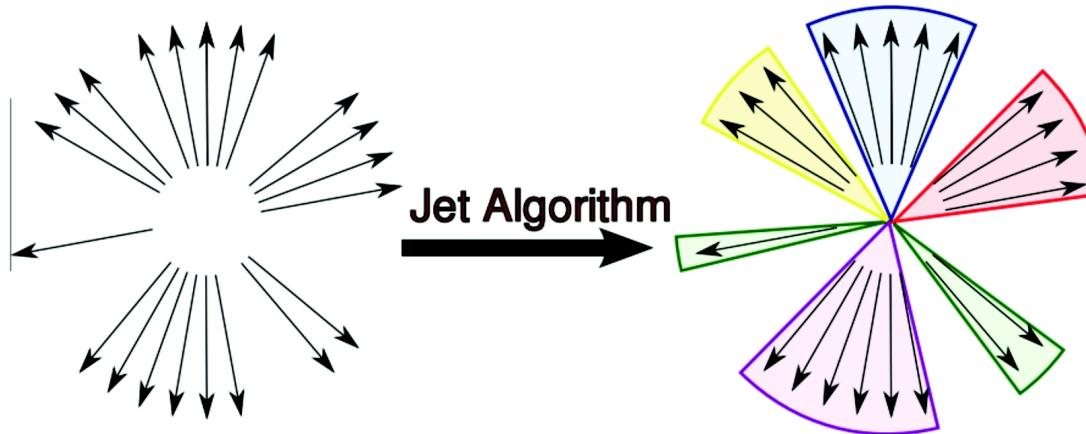
Jet Reconstruction and Inputs

Chris Young



What are Jets?

- Jet reconstruction is an organizational principle where we group sets of particles to form jets.
- Several jet algorithms could be run on the same event to get different “views” - none is “correct”.
- The jet algorithm defines how the particles are clustered into these jets – we can apply it to calorimeter objects, tracks, combinations of the two, and truth particles.
- We tend to use the same algorithm on stable truth particles as the reference for performance or measurements – note a jet is not the same as a parton!

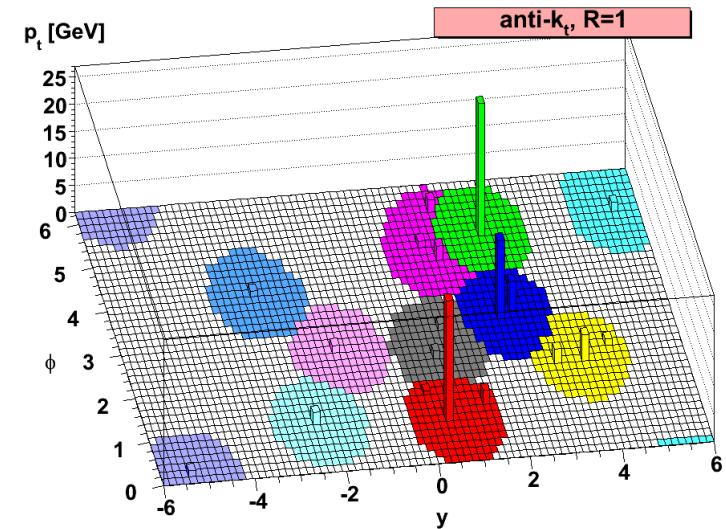
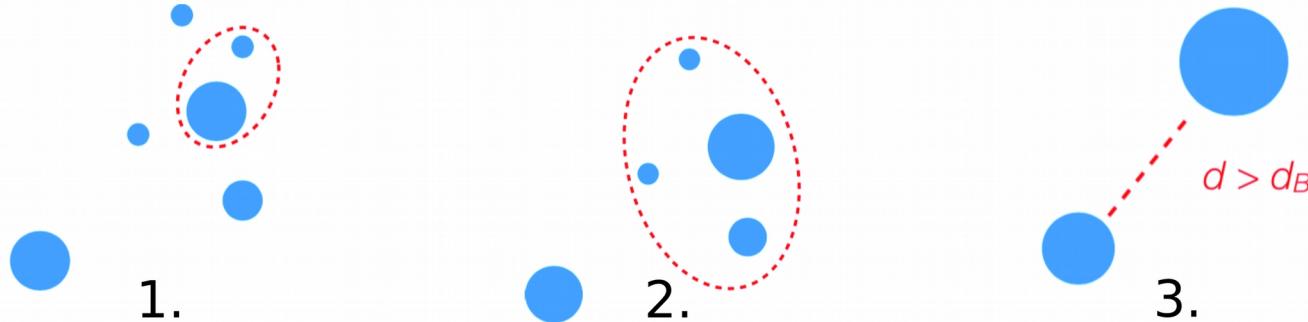


The Anti- k_t Algorithm

- Most jets in ATLAS use the Anti- k_t algorithm.
- This is a sequential combination algorithm defined by a “radius parameter”, R.
- We define a distance measure d_{ij} , and then combine the two particles with smallest d.
- Then repeat combining pairs with the lowest d_{ij} until the smallest d_{ij} is $> d_B$
- This results in reasonably round jets in the η, ϕ space.

$$d = \min(p_{T,i}^{-1}, p_{T,j}^{-1}) \frac{\Delta_{i,j}^2}{R^2}$$

$$d_B = p_{T,i}^{-2}$$

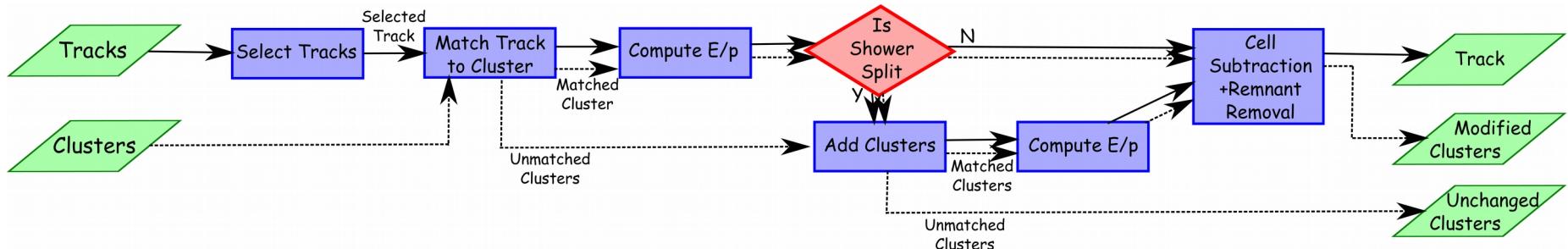


Topoclusters as Inputs

- The simplest jets use calorimeter topoclusters as inputs.
- These can either be at the electro-magnetic scale (EMTopo) or local calibration scale (LCTopo).
- Topoclusters were covered in the Calo talk of the tutorial – reminder;
 1. Take cells with $|E| > 4 \sigma(E)$ (note σ includes pile-up noise)
 2. Add adjacent cells with $|E| > 2 \sigma(E)$
 3. Add a final layer without any $|E|/\sigma(E)$ requirement.
 4. Split clusters based on multiple local maxima.
- The LC calibration accounts of differences in the response of hadrons and EM-particles, energy of the shower falling outside of clusters, and energy deposited in dead material.
- Historically small-R jets in Run II have been primarily EMTopo jets (but this is no longer what is recommended). Large-R jets often use LCTopo inputs – these provide some cluster level calibration for forming sub-structure variables from the jet inputs.

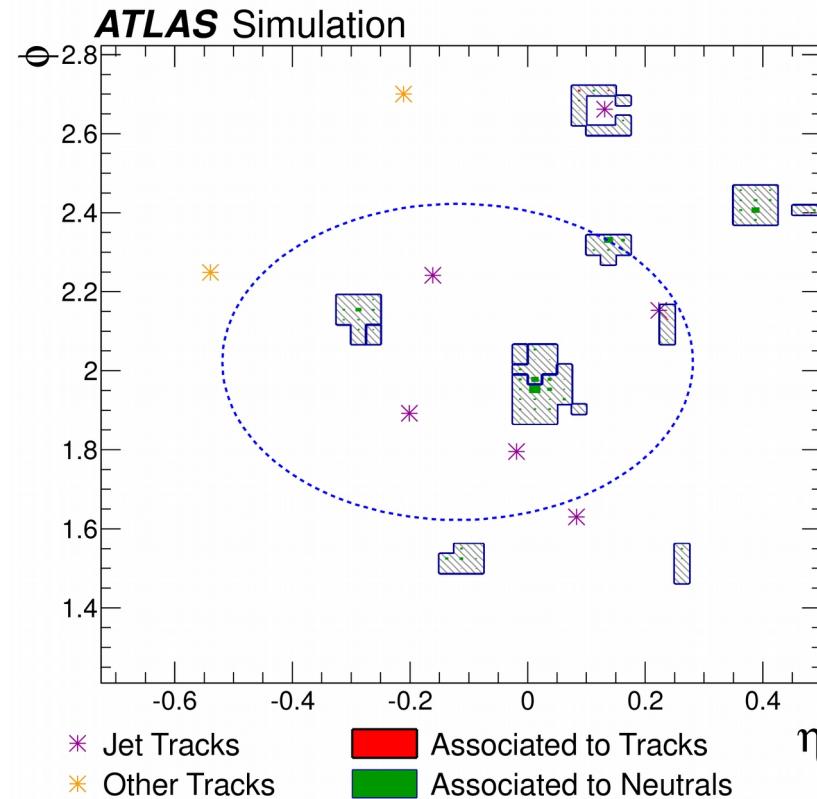
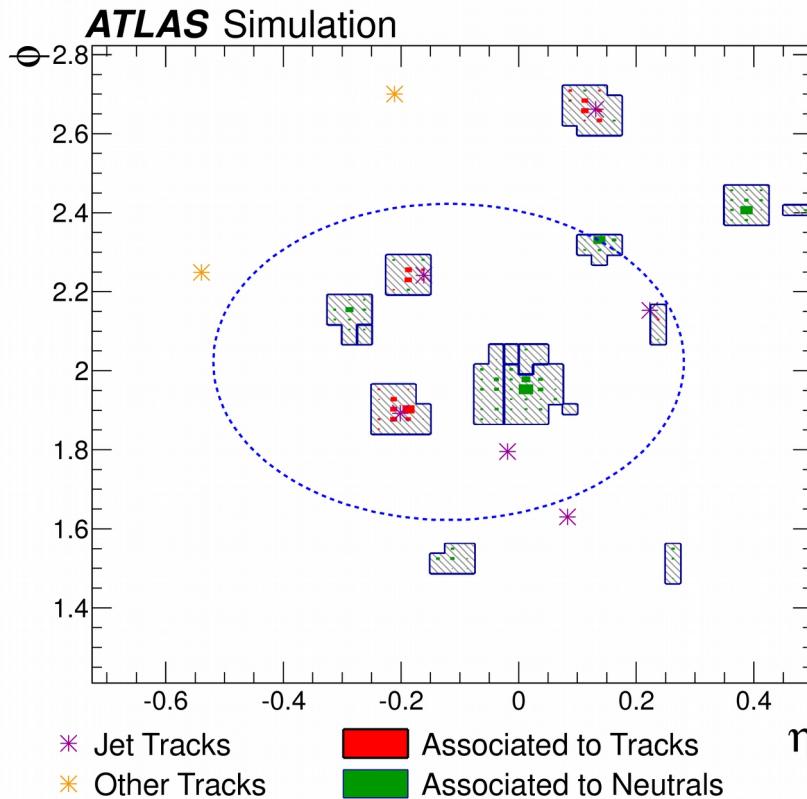
Particle Flow in ATLAS

- Particle Flow aims to combine tracker and calorimeter measurements to improve performance.
- Pflow in ATLAS is primarily for jets – we do not try to do the event globally like CMS.
- The tracker advantages are; better p_T resolution at low p_T , better angular resolution, can distinguish pile-up.
- The calorimeter advantages are; better p_T resolution at high p_T , can reconstruct neutral particles.
- It is optimized to improve the resolution of low p_T jets, and maintain the resolution of high p_T jets by gradually turning off.
- For a given track we decide if the tracker will give a significantly better measurement, if so, then we subtract the cells from the calorimeter that are associated with this particle's shower (note this is sometimes imperfect).



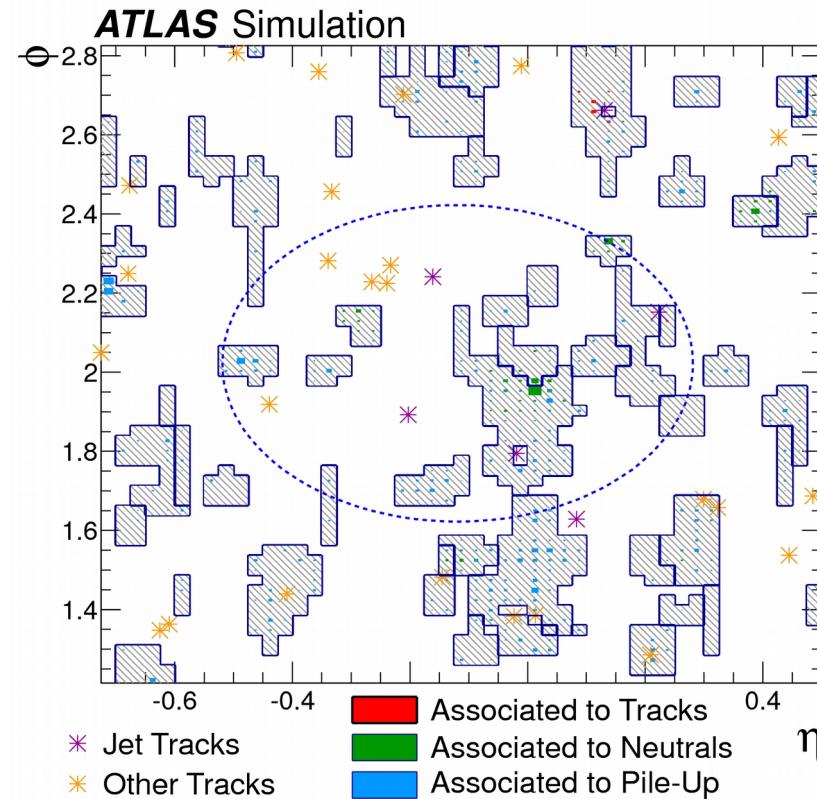
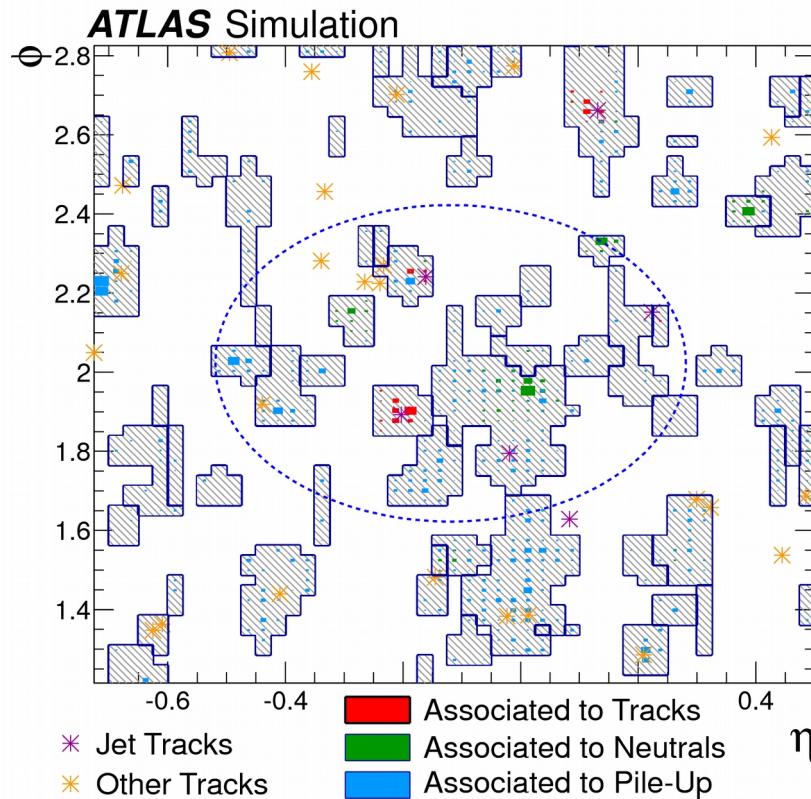
Particle Flow Visualization

- Below you see the effect (left to right) of the particle flow subtraction in an event in the 2nd layer of the EM calorimeter with no pile-up and a 30 GeV jet.



Particle Flow Visualization

- Below you see the effect (left to right) of the particle flow subtraction in an event in the 2nd layer of the EM calorimeter with 40 interactions per crossing pile-up and a 30 GeV jet.



Track-Calorimeter-Clusters (TCCs)

- At high momentum boosted hadronic objects like $W \rightarrow q\bar{q}$, $t \rightarrow b\bar{q}\bar{q}$ become very close together.
- This makes resolving their sub-structure difficult using the calorimeter due to the finite cell granularity.
- The track-calorimeter-cluster algorithm uses the better angular resolution of the tracker to split clusters, and in the case of a single match take the track spatial coordinates.

$$TCC1 = (p_T^{c1}, \eta^{t1}, \phi^{t1}, 0)$$

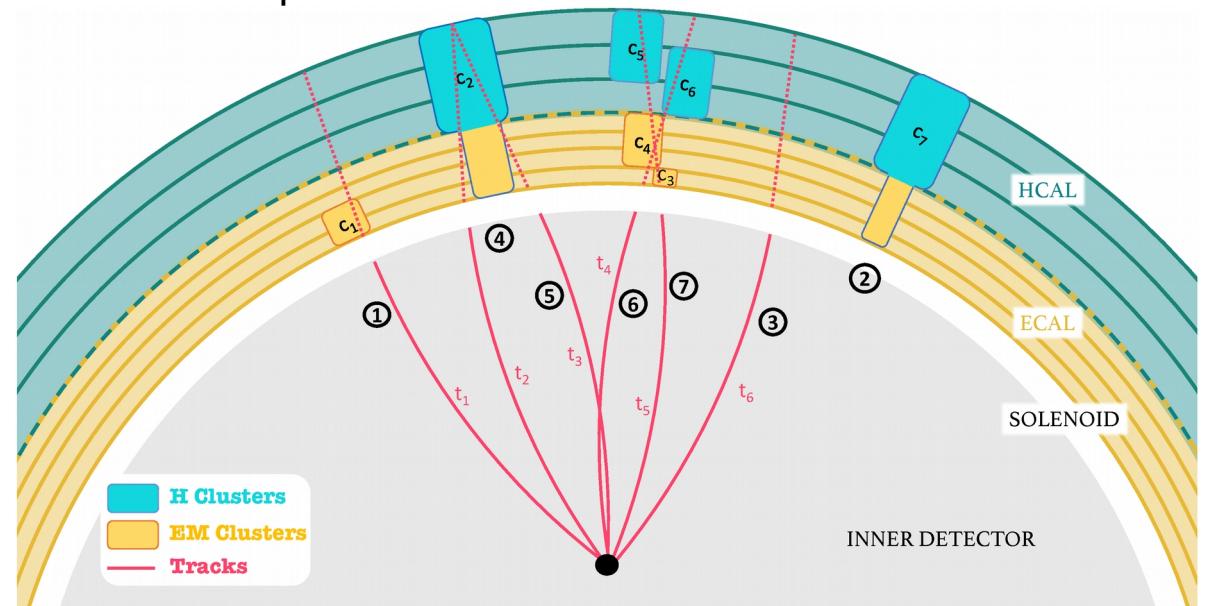
$$TCC2 = (p_T^{c7}, \eta^{c7}, \phi^{c7}, 0)$$

$$TCC3 = (p_T^{t6}, \eta^{t6}, \phi^{t6}, 0)$$

$$TCC4 = (p_T^{c2} \times (p_T^{t2}/p_T^{(t2+t3)}), \eta^{t2}, \phi^{t2}, 0)$$

$$TCC5 = (p_T^{c2} \times (p_T^{t3}/p_T^{(t2+t3)}), \eta^{t3}, \phi^{t3}, 0)$$

Etc...



When to use what and the EDM

- Pflow was optimized for small-R jets with particular attention to low p_T – it is the default collection for AntiKt4 jets.
 - If your signal has no tracks, is out-of-time, etc then you may still have to use calorimeter (EMTopo) jets.
 - For large-R jets most analyses currently use LCTopo jets – good performance across all p_T .
 - For very boosted searches eg. VV resonance significant gains come from TCC as they have best tagging performance at highest p_T .
 - We are working on releasing a combination of the Pflow and TCC algorithms (called Unified Flow Object (UFO)) which would provide the best large-R jet performance across all p_T .
-
- In the xAOD we store CaloCluster::CaloCalTopoClusters (EM+LC clusters), TrackParticle::InDetTrackParticles, PFO::JetETMissChargedParticleFlowObjects, PFO::JetETMissNeutralParticleFlowObjects.
 - TCCs have their own EDM and are built in the derivation step.
 - Most people don't need to use these objects as jets can easily be built from these objects in the derivation framework.

Conclusions

- Jets are built either at Tier-0 or in the derivation framework using the Anti- k_t algorithm.
- They can be built from several different input containers; CaloCluster, PFO, TCC.
- Depending on your use case you might prefer to use a different input collection to get the best performance – Pflow for small-R, LCTopo large-R, TCC large-R at high pT, (and in the future UFO!).
- Throughout this presentation you have seen how people can improve our object reconstruction making the most of our detector, and how this can influence many analyses at once – consider joining the JetEtmiss group to further optimize jet reconstruction!
- Have fun building jets – helper functions to build jets in the derivation framework can be found in [JetCommon.py](#) and [ExtendedJetCommon.py](#)