Jet Calibration in ATLAS Line Delagrange



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Strong interaction

- strong coupling constant $\alpha_{\rm S}$
- Quarks and gluons carry colour charge \rightarrow self-interaction



- They form bound colourless states (hadrons)
- Due to colour confinement, quarks and gluons shower and hadronise immediately into collimated bunches of particles \rightarrow Jets



Jets

Jets represent the shower produced by the hadronisation of a quark or gluon



Courtesy of Louis Ginabat

"Truth" jet

"Reco" jet

- Dominant production at the LHC
- Used either as signal or background in most analyses





Jet reconstruction

- Goal: Construct jets from the input 4-vectors - calorimeter hits and tracks (data) - simulated particles (mc truth)
 - simulated calorimeter hits and tracks (mc reco)
- Anti- k_T algorithm: sequential jet clustering algorithm of near-by entities, with $d_{ij} = \min(k_{ti}^{-2}, k_{tj}^{-2}) \frac{\Delta_{ij}}{R^2}$ and deduce the 4-vector of the associated jet





Jet calibration Goal

- To have the reconstructed 4-vector of the jet matching that of the true 4vector corresponding jet (in data and mc)
- Correct energy and direction of the jet for:
 - Energy lost in the upstream material
 - Energy lost in dead material
 - Non-compensating nature of ATLAS detector
 - Bending of the particles in the magnetic field
 - interaction (pile-up)



- Busy data taking environment resulting from the multiple proton-proton

Derive correction factors to be applied to reconstructed jets in mc and data



Jet calibration Principle

direction are calibrated.



muon-segment variables.

applied only to data.



Residual in situ Calibration

GOAL: Correct the residual differences between data and Monte Carlo, with in situ measurement

Principle : Use of the p_T balance between a jet (probe) and a reference object (ref) Correction factors derived in bins of p_T and η :



$$\frac{\mathcal{L}_{MC}}{\mathcal{L}_{data}} = \frac{\mathcal{R}_{data}}{\mathcal{R}_{MC}} = \left[\frac{(p_T^{probe})_{reco}}{p_T^{ref}}\right]_{data} / \left[\frac{(p_T^{probe})_{reco}}{p_T^{ref}}\right]_{M}$$

Intercalibration factors

Correction factors (to be applied to data)

 η -intercalibration \rightarrow homogeneity in η





η -intercalibration

Using di-jet events



Exemple of an asymmetry distribution. The central value of a gaussian fit is extracted



Dijet Selection for R=0.4: $\Delta \phi > 2.5$ $p_T^3 / p_T^{avg} < 0.25$ JVT < 0.25



Standard Method:

- probe = jet to calibrate
 - ref = jet in the reference region (central region)

• Asymmetry evaluated in bins of p_T^{avg} , η_{ref} and η_{probe}

$$= \langle \mathscr{R} \rangle = \frac{2 + \langle \mathscr{A} \rangle}{2 - \langle \mathscr{A} \rangle}$$
 intercalibration factors

Problem : low statistics









η -intercalibration

- χ^2 minimisation process



 $S(c_1, \dots, c_N) = \sum_{j=1}^N \sum_{i=1}^{j-1} \left(\frac{1}{\Delta \langle \mathcal{R}_{ij} \rangle} \left(c_i \langle \mathcal{R}_{ij} \rangle - c_j \right) \right)^2 + X(c_1, \dots, c_N)$

• In each bin of $p_T^{avg} \rightarrow$ **Over-constrained system** : *N intercalibration factors to determined* < $\sim \frac{N^2 - N}{2}$ constraints

The correction factors are the ratio of the inter-calibration factors (mc/data)



Global χ^2 /NDF

- The intercalibration factors are determined by a χ^2 minimisation process: $S(c_1, \ldots, c_N) = \sum_{j=1}^N \sum_{i=1}^{j-1} \left(\frac{1}{\Delta \langle \mathscr{R}_{ij} \rangle} (c_i \langle \mathscr{R}_{ij} \rangle c_j) \right)^2 + X(c_1, \ldots, c_N)$ A global χ^2 /NDF can be calculated for each p_T bin:
 - $\chi^{2} = \sum_{j=1}^{N} \sum_{i=1}^{j-1} \left(\frac{1}{\Delta \langle \mathcal{R}_{ij} \rangle} \left(c_{i} \langle \mathcal{R}_{ij} \rangle c_{j} \right) \right)^{2}$ $NDF = \frac{N_{\eta bins}^2 - N_{\eta bins}}{2} - N_{dropped} - N_{\eta bins}$ # of constraints

of intercalibration factors

The global χ^2 /NDF provides information on the compatibility between the constraints



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Global χ^2 /NDF

• The global χ^2 /NDF study allows to target the highly contributing bins for studies









Bootstrap Method

- Evaluate the statistical uncertainty of a measurement
- Using a set of replicas of the nominal dataset, derived by introducing Poisson perturbations
- Analysing each replica, the same way as the nominal dataset
- Extract the statistical uncertainty and correlations from the measurements
- The fluctuations that generate the bootstrap replicas are deterministic



 $85 < p_T < 115$, working with 100 replicas

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Bootstrap Method

- Correlation Matrix, determined here for the first time
- Very important for quantitative data/theory comparisons e.g. for jet cross-sections
- (Anti-)correlations: in the MM, an asymmetry bin constrains two intercalibration factors

 \rightarrow currently not taken into account in the computation of the uncertainties

 Could improve the calibration, useful for many studies involving jets







What's next Jet substructure and α_S

- Usually: Jet production cross section
- Lund Jet Plane: a modern way to explore the jet substructure, sensitive to α_{S}
- Re-clustering the jet, entering the "emission" coordinates in a $(\ln(k_T), \ln(1/\Delta))$ plane
- Broad range of scale covered to test the running of $\alpha_{\rm S}$, + Normalisation sensitive to α_{S}

scale with Run-2 data



From 1807.04758 Factorisation of QCD effects

GOAL : Evaluation of α_{S} and test of its running as a function of the energy



What's next FCC-ee

- FCC-ee : 91km of circumference, ~2040, e^+e^- collisions at 4 center of mass energies between 90 and 365 GeV. Very high statistics, very clean environment. \rightarrow **Contraints on the detectors** : minimising the systematics to take advantage of the high statistic
- Prospective studies of the Lund Jet Plane in a FCC-ee environment with mc simulations

Lund jet plane study, to improve the determination of α_S



FUTURE CIRCULAR COLLIDER

GOAL : Optimise the detector design (energy resolution, granularity, etc) for the



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Thank you for your attention!