



JETS FOR THE LHC AND BEYOND

SIMONE MARZANI UNIVERSITÀ DI GENOVA & INFN GENOVA

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Outline

Introduction to jet physics
 Jet substructure to look for boosted-objects

 a) example of a groomer: soft drop
 b) example of a tagger: N-subjettiness

 Measuring jet substructure

 a) soft drop observables
 b) extraction of fundamental parameters

 Final remarks

Introduction to jet physics



Jet definitions

- jet algorithms: sets of (simple) rules to cluster particles together
- implementable in experimental analyses and in theoretical calculations
- must yield to finite cross sections
 first example:

To study jets, we consider the partial cross section	
$\sigma(E,\theta,\Omega,\varepsilon,\delta)$ for e ⁺ e ⁻ hadron production events, in which all bu	t
a fraction $\epsilon \ll 1$ of the total e ⁺ e ⁻ energy E is emitted within	
some pair of oppositely directed cones of half-angle $\delta <<1,$	
lying within two fixed cones of solid angle Ω (with $\pi\delta^2 << \Omega << 1$)
at an angle θ to the e ⁺ e ⁻ beam line. We expect this to be meas	ur-

Sterman and Weinberg, Phys. Rev. Lett. 39, 1436 (1977):

 Start with a list of particles, compute all distances d_{ij} and d_{iB}

• Find the minimum of all d_{ij} and d_{iB}

d_{ij} (weighted) distance between i j d_{iB} external parameter or distance from the beam ...

for a complete review see G. Salam, Towards jetography (2009)

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- Find the minimum of all d_{ij} and d_{iB}
- If the minimum is a d_{ij}, recombine i and j and iterate
- Otherwise call i a final-state jet, remove it from the list and iterate

d_{ij} (weighted) distance between i j d_{iB} external parameter or distance from the beam ...

Actual choice for the measure d_{ij} determines the jet algorithm

The generalised kt family

Actual choice for the measure d_{ij} determines the jet algorithm

$$d_{ij} = \min\left(p_{ti}^{2p}, p_{tj}^{2p}\right) \frac{\Delta R_{ij}^2}{R^2}$$

 $d_{iB} = p_{ti}^{2p}$

$$\mathbf{A}R_{ij}^2 = (y_i - y_j)^2 + (\phi_i - \phi_j)^2$$

p = I k_t algortihm (Catani et al., Ellis and Soper) **p** = **0** Cambridge / Aachen (Dokshitzer et al., Wobish and Wengler) **p** = **-** I anti-k_t algorithm (Cacciari, Salam, Soyez)

Different algorithms serve different purposes

Comparing clustering algorithms

 Anti-kt clusters around hard particles giving round jets (default choice for ATLAS and CMS)

- kt & C/A reflect the structure of QCD matrix elements
- Anti-k_t is less useful for substructure studies: often reclustering is done with C/A





Cacciari, Salam, Soyez (2008)

Jet substructure to look for boosted-objects

Searching for new particles: resolved analyses

• the heavy particle X decays into two partons, reconstructed as two jets



 look for bumps in the dijet invariant mass distribution



Searching for new particles: boosted analyses

- LHC energy (10⁴ GeV) \gg electro-weak scale (10² GeV)
- EW-scale particles (new physics, Z/W/H/top) are abundantly produced with a large boost



- their decay-products are then collimated
- if they decay into hadrons, we end up with localized deposition of energy in the hadronic calorimeter: a jet





CMS Experiment at LHC, CERN Run 133450 Event 16358963 Lumi section: 285 Sat Apr 17 2010, 12:25:05 CEST



JETS Vimated, energetic vs of particles

R



we want to look inside a jet

exploit jets' properties to distinguish signal jets from bkg jets

R

 $p_t > 2m/R$

16

h

 \boldsymbol{Q}

The jet invariant mass

- First jet-observable that comes to mind
- Signal jet should have a mass distribution peaked near the resonance



• However, that's a simple partonic picture

A useful cartoon

inspired by G. Salam



A useful cartoon

inspired by G. Salam



A useful cartoon

et

inspired by G. Salam

underlying event (multiple parton interactions)

hadronisation

pert. radiation (parton branching)

pile-up (multiple proton interactions)

Effect on jet masses

 In reality perturbative and non-pert emissions broadens and shift the signal peak

Underlying vent and the let mass (both signal and background)



Beyond the mass: substructure

- Let's have a closer look: background peaks in the EW region
- Need to go beyond the mass and exploit jet substructure
- Grooming and Tagging:
 - 1. clean the jets up by removing soft junk
 - 2. identify the features of hard decays and cut on them



ATLAS.

JHEP 1309

(2013) 076

Beyond the mass: substructure

- Let's have a closer look: background peaks in the EW region
- Need to go beyond the mass and exploit jet substructure
- Grooming and Tagging:
 - 1. clean the jets up by removing soft junk
 - 2. identify the features of hard decays and cut on them
- Grooming provides a handle on UE and pile-up



Example of a groomer / prong-finder: Soft Drop

Soft Drop

Larkoski, SM, Soyez and Thaler (2014)



Lund plane



- originally introduced to describe parton shower phase space
- very helpful to understand logarithmic structure for resummation

$$d\sigma = \alpha_s(k_t)C_i\frac{dk_t}{k_t}\frac{d\theta}{\theta}$$

 more recently employed as input for for machinemachine-learning algorithms

Dryer, Salam, Soyez (2018)



- Soft Drop kinematic plane as a function of the angular exponent
- the region below the green line is groomed away
- this understanding can be easily translated into <u>analytic</u> resummation formulae

Groomed jet properties



- smooth distributions
- flatness in bkg can be achieved for $\beta=0$
- now the standard choice for CMS

Soft drop at NNLL



 $\frac{\min[p_{Ti}, p_{Tj}]}{p_{Ti} + p_{Tj}} > z_{cut}^{\text{Frye}} \left(\frac{R_{ijk}}{R} \right)^{\beta} \text{ski, Schwartz, Yan (2016)}$

- soft-drop mass: something we can calculate
- reduced sensitivity to non-pert effects
- going to NNLL reduces scale variation but small changes in the shape
- for $\beta=0$ LL is zero, so state-of-the art NNLL is actually NLL

Example of a tagger: jet shapes

Tagging W boso

- jet shapes measure the distribution of radiation inside a signal from background
- standard analyses typically use a two-step approach

 $\tau_{21} = \frac{\tau_2^{(\beta)}(\text{jet; axes})}{\tau_1^{(\beta)}(\text{jet; axes})} = \frac{\sum_{i \in \text{constits}} z_i \min(\theta_{i, a_{2,1}}^{\beta}, \theta_{i, a_{2,2}}^{\beta})}{\sum_{i \in \text{constits}} z_i \theta_{i, a_{2,1}}^{\beta}}$

- first a mass window is identified (e.g. around mw)
- a cut on a shape is imposed
- N-subjettiness is widely used for this purpose



0.8



- axes?
- angular exponent ($\beta = 1 \text{ vs } \beta \neq 2$)?
- plain or groomed jets
 (or both: dichroic ratios)?

Salam, Schunk, Soyez (2016)



Kinematics of T₂₁



- note that Lund diagrams become more complicated because we are sensitive to two emissions (mass and shape)
- the shape can be set by emission off the leading parton or by a splitting of the primary emission (which generates the extra fin)
- also in this case we can translate the above into analytic expressions
- we can also add grooming and consider different shapes

Analytics vs parton shower

shape distributions - dijets - Pythia 0.9 (B=2)Solid: SoftDrop ($\beta = 2, z_{cut} = 0.05$) τ21 (dichroic) 0.8 Dashed: mMDT (z_{cut}=0.1) τ21 $D_2^{(\beta=2)}$ Dichroic: ρ_{mMDT} , $\tau_{2,SD}/\tau_{1,mMDT}$ 0.7 $p_t > 2$ TeV, $m = m_W$ 0.6 parton level σ/ν dσ/dν 0.5 0.4 Pythia8. 0.3 0.2 Tev, 4 0.1 0 0.001 0.01 0.1 10 100 1 V

shape distributions - WW - Pythia



shape distributions - QCD - analytic



shape distributions - W - analytic 0.5 $\tau_{21}^{(\beta=2)}$ Solid: SoftDrop ($\beta = 2, z_{cut} = 0.05$) (dichroic) Dashed: mMDT (z_{cut}=0.1) τ21 $D_2^{(\beta=2)}$ – Dichroic: ρ_{mMDT} , $\tau_{2,SD}/\tau_{1,mMDT}$ 0.4 pt=2 TeV, m=mW 0.3 v/a da/dv 0.2 0.1 0 0.001 0.01 0.1 1 10 100 V

region of large v difficult to model in resummation but recent progress has been made

Napoletano, Soyez (2018)

- QCD jet shapes significantly affected by grooming, while signal ones less so
- Grooming does clean the jet up but tend to decrease separation, i.e. performance

Performance and resilience

- resilience measures a tagger's robustness against non-perturbative effects (hadronisation and UE)
- it is defined in terms of signal/background efficiencies with/without non-pert. contributions

/2

$$\zeta = \left(\frac{\Delta \epsilon_S^2}{\langle \epsilon \rangle_S^2} + \frac{\Delta \epsilon_B^2}{\langle \epsilon \rangle_B^2}\right)^{-1}$$

$$\Delta \epsilon_{S,B} = \epsilon_{S,B} - \epsilon'_{S,B},$$
$$\langle \epsilon \rangle_{S,B} = \frac{1}{2} \left(\epsilon_{S,B} + \epsilon'_{S,B} \right)$$



Measuring jet substructure

Soft Drop observables



- what's the impact of finite z_c contributions (formally LL)?
- what's the impact of logs of z_c (formally N^kLL)?
- conclusions will change if we move away from z_c=0.1

SM, Schunk, Soyez (2017)

- let's start wit the simplest observable: jet mass
- large range of masses where NP corrections are small and we can trust resummation



and the DATA!



CMS-PAS-16-010

- CMS & ATLAS measurements
- NNLL is a small correction
- importance of FO for the tail
- ATLAS did β survey



Why unfolded measurements ?

What is the value of SM measurements and their comparison to theory, especially for "discovery" tools?

- understanding systematics (e.g. kinks and bumps)
- where non-perturbative corrections are small, test perturbative showers in MCs
- at low mass, hadronisation is large but UE is small: TUNE!





Extraction of Standard Model parameters

Strong coupling

- current precision below 1%, dominated by lattice extractions
- LEP event shapes also very precise (5%), however they are in tension with the world average
- Thrust (and C parameter) known with outstanding accuracy
- Strong correlations with non-pert. parameter







Soft-drop thrust





Baron, SM, Theeuwes (2018)

 noticeable reduction of non-pert. corrections may allow to disentangle the degeneracy

can we compute it at the same accuracy as standard event shapes?

NNLO calculations recently performed

Kardos, Somogyi, Trocsanyi (2018)

α_s with soft-drop thrust

SM, Reichelt, Schumann, Soyez and Theeuwes, (soon to appear)



- fits to pseudo-data generated by SHERPA
- preliminary results shows reduced dependence on non-pert. corrections
- subleading effects are under investigation
- general question: is there a natural way to define soft-drop event shapes?
- bottom-up soft drop allows one to groom an entire event

Dreyer, Necib, Soyez, Thaler (2018); Baron (in preparation)

mt with soft-drop jets

- determination of other fundamental parameters may benefit from grooming, e.g. the top quark mass
- in the context of e+e- collisions SCET factorisation theorems allow for a precision-determination of the top-jet mass
- the picture at pp collisions is polluted by wide-angle soft radiation grooming "turms" $\sum_i |\vec{n} \cdot \vec{p_i}|$ observables into e+e- ones
- $\tau \stackrel{\tau \to 0}{\approx} \frac{M_1^2 + M_2^2}{O^2}$





Final remarks

Summary & Outlook

- importance of substructure studies
- soft drop: theoretical status and physics opportunities
- Open questions
 - 1. higher-order corrections (i.e. beyond NLO) and grooming?
 - 2. in the boosted regime electro-weak corrections are significant
 - 3. in the opposite direction: non-perturbative physics and hadronisation in particular. Is "standard"? and what does standard even mean?

Humans vs machines

Jet physics (and particle physics!) undergoing a revolution
ideas / techniques from machine (deep) learning continuously poured into the field

 I had to make a choice and concentrated on humans for this talk



• Food for thoughts: • what are the machinelearning ideas best suited for particle physics? (images, language...) • are we scared of black boxes? (should we?) • can we make black boxes more transparent?

Thank you !

Looking inside jets: an introduction to jet substructure and boosted-object phenomenology

Simone Marzani¹, Gregory Soyez², and Michael Spannowsky³

 ¹Dipartimento di Fisica, Università di Genova and INFN, Sezione di Genova, Via Dodecaneso 33, 16146, Italy
 ²IPhT, CNRS, CEA Saclay, Université Paris-Saclay, F-91191 Gif-sur-Yvette, France
 ³Institute of Particle Physics Phenomenology, Physics Department, Durham University, Durham DH1 3LE, UK