

High multiplicity processes with BlackHat and Sherpa



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In collaboration with

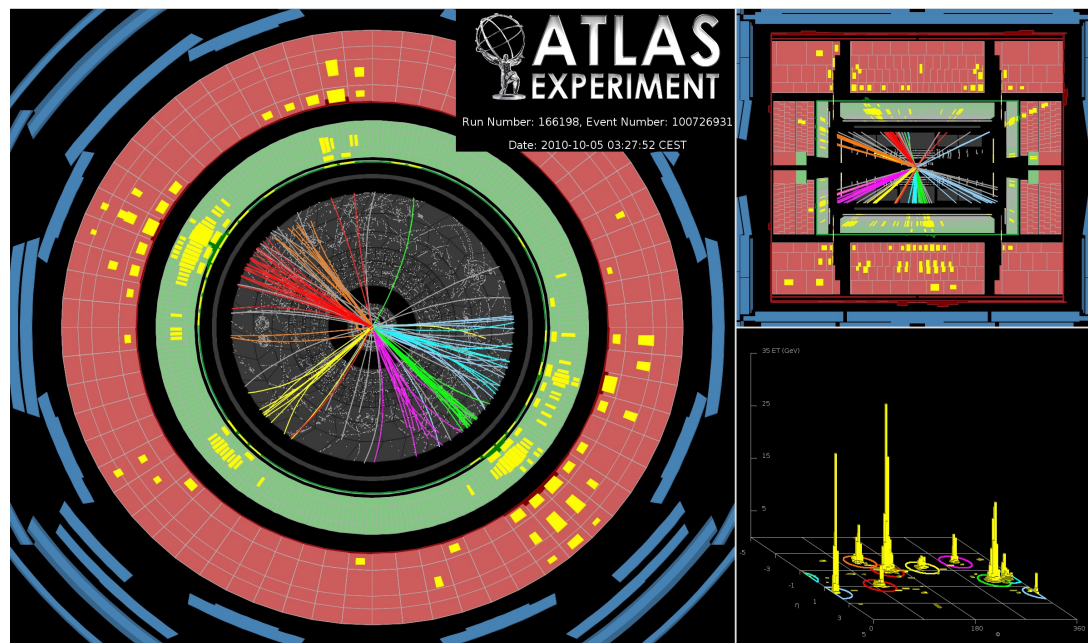
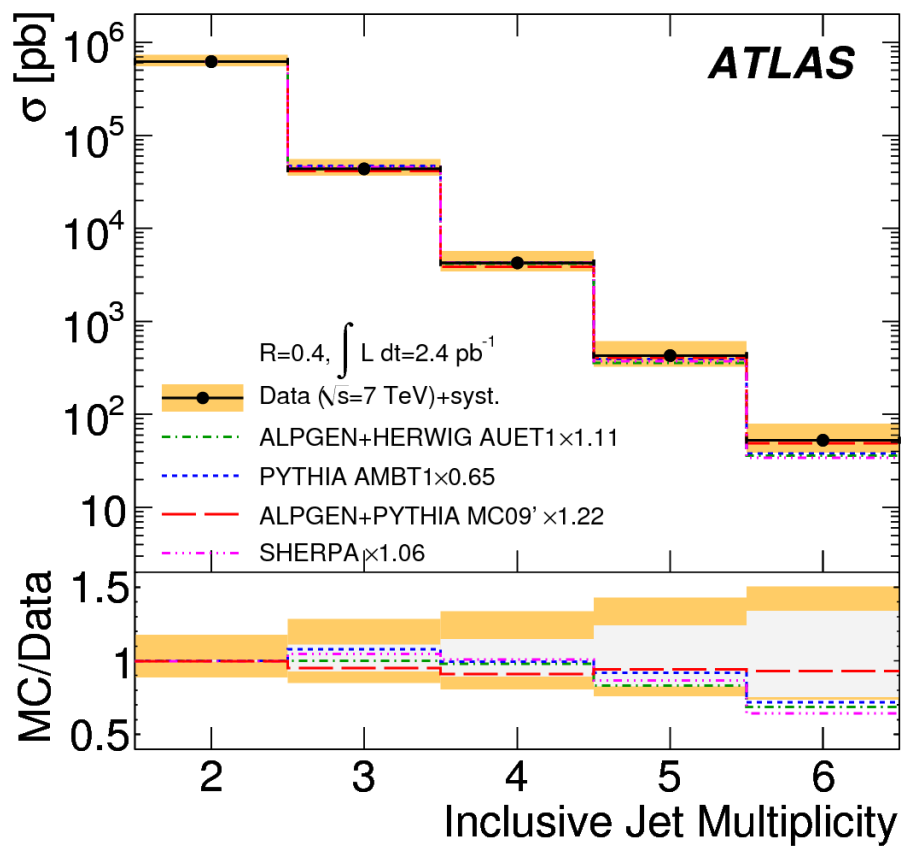
Zvi Bern, L. Dixon, F. Febres Cordero, S. Höche, H. Ita,
D. Kosower, K. Ozeren, N. A. Lo Presti

Outline

- BlackHat+Sherpa n-Tuple files and library
- Exclusive sums
- Universal behaviour in W+jets production

High multiplicities

- At the LHC events with many jets are common place

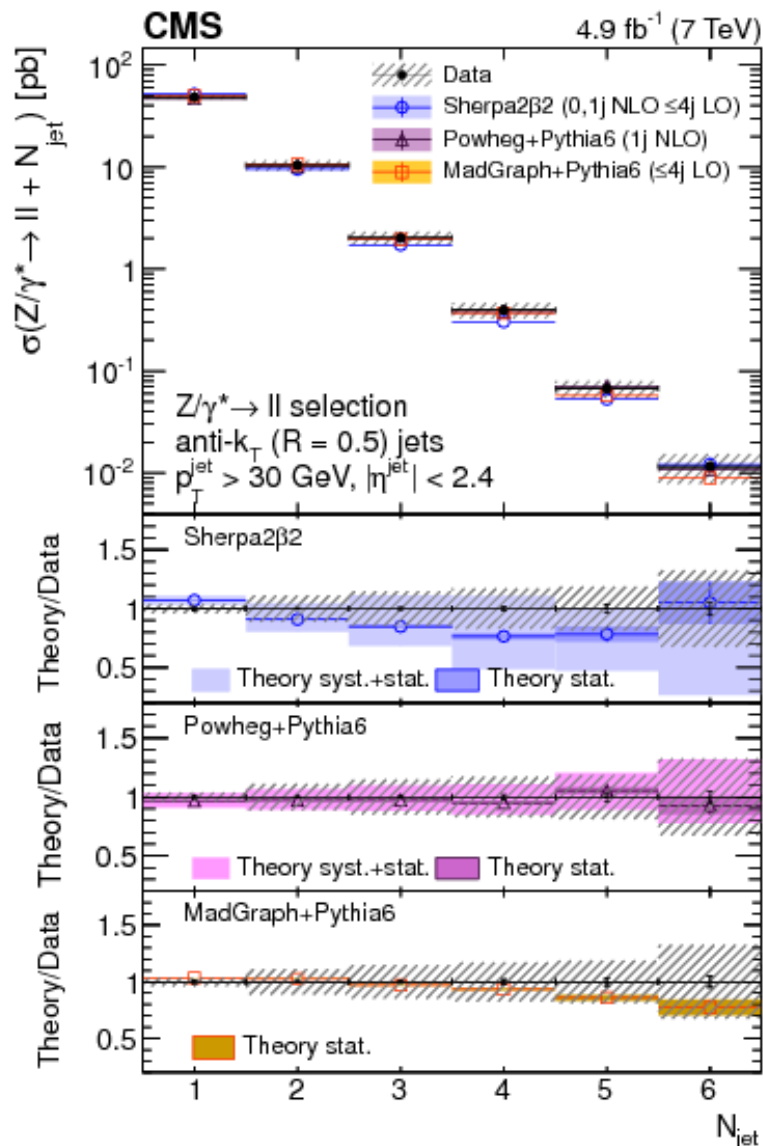


8-jets event at ATLAS

ArXiv:1202.3548

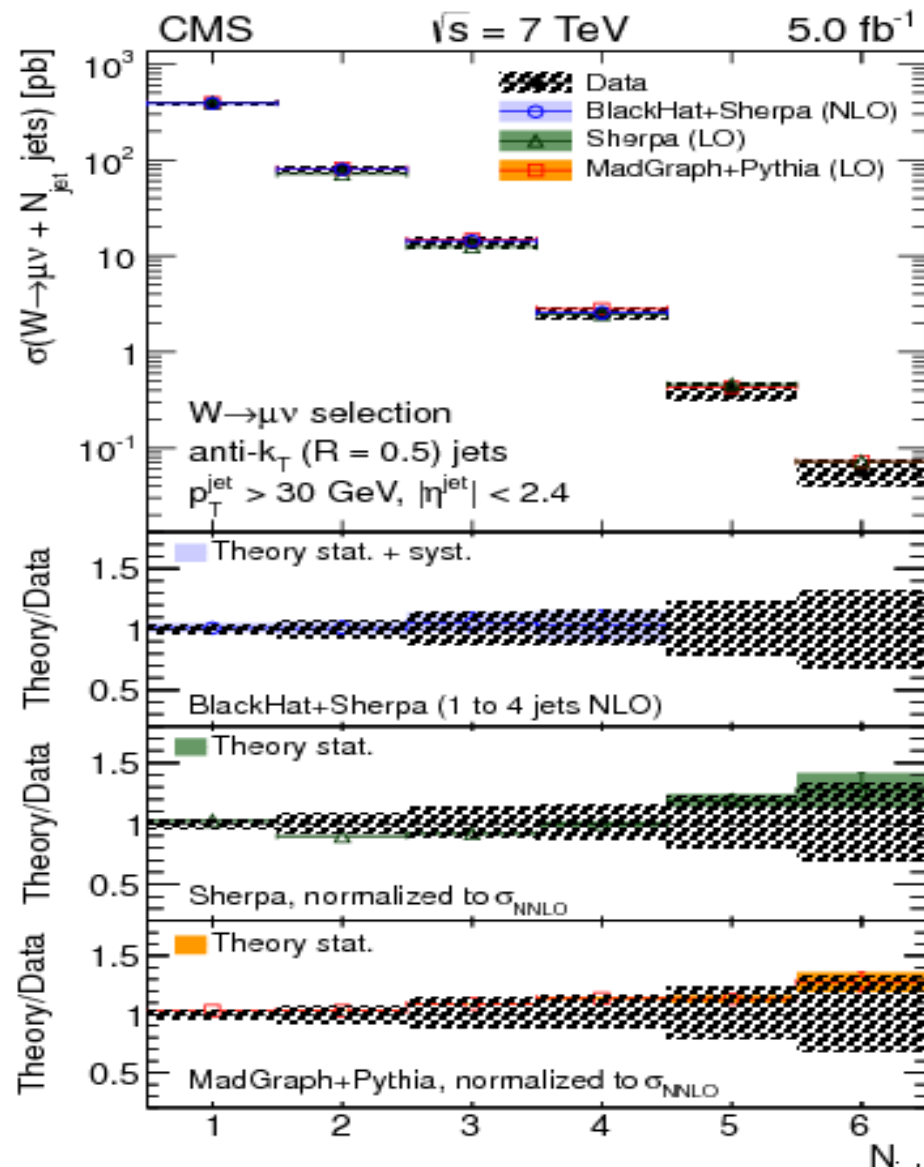
High multiplicities

- Z+jets



arXiv:1408.3104

- W+jets



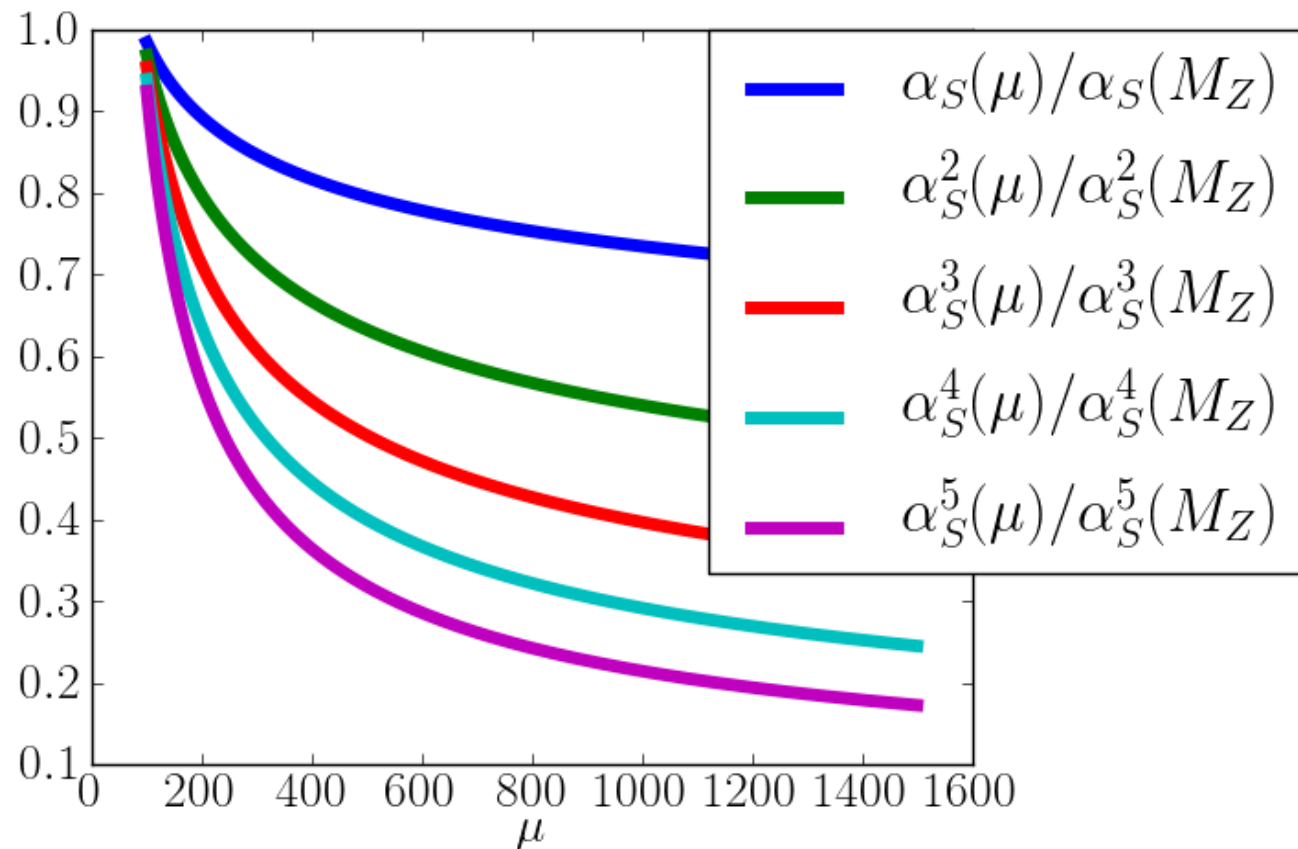
arXiv:1406.7533

Precise predictions

- Precise predictions are needed
 - Signal
 - Background
 - Also for data-driven methods
 - Extrapolation from control to signal region
 - Transfer of information from one process to another
- NLO improves
 - Absolute normalisation
 - Shapes of distributions
 - Scale dependence

NLO scale dependence

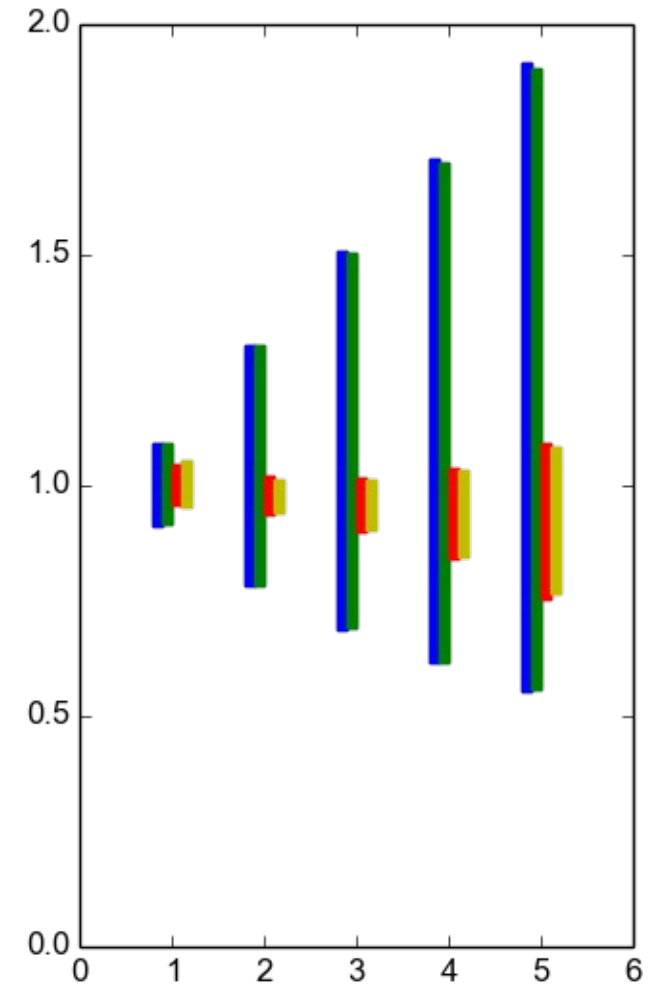
- The theory prediction depends on two unphysical scales, the renormalisation and factorisation scale
- The strong coupling constant is a function of the renormalisation scale



Scale variation

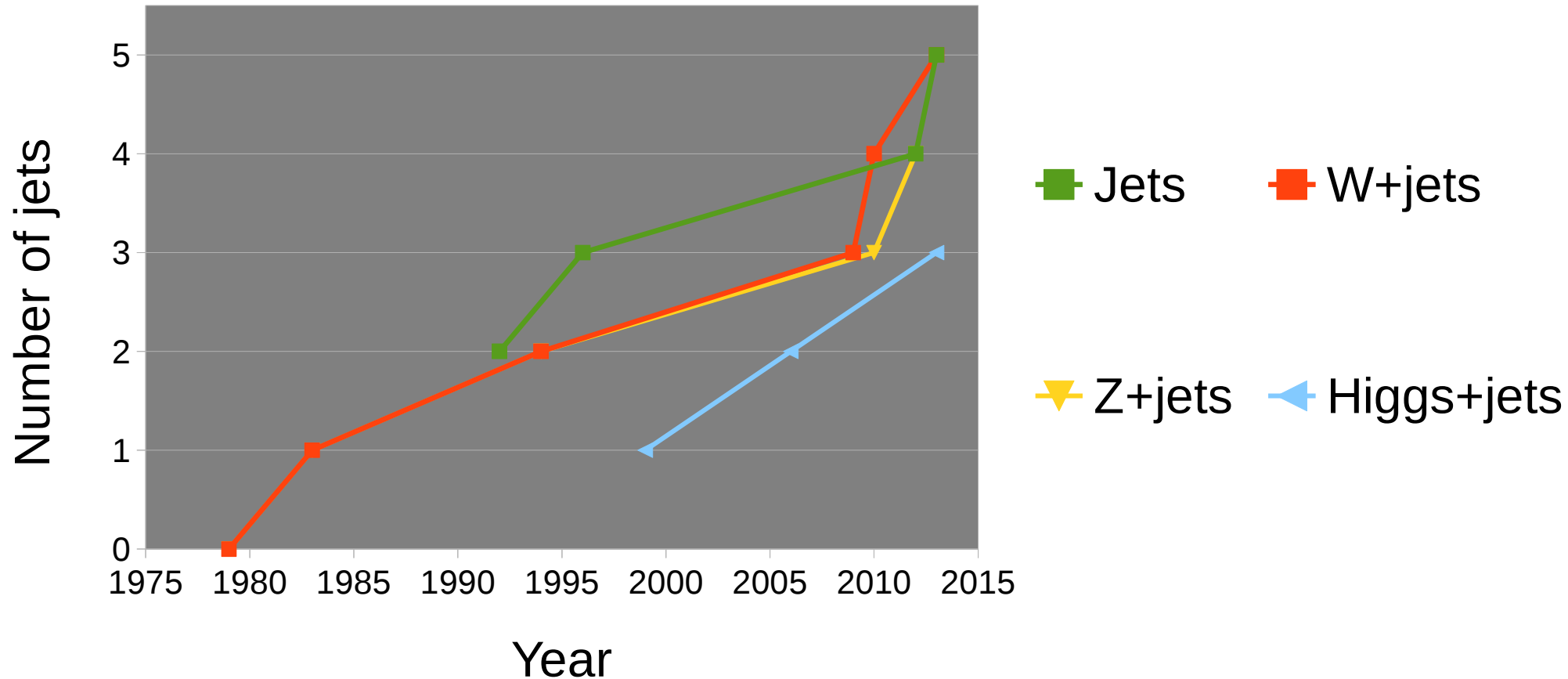
- Scale variation is largely reduced at NLO
- More and more important as multiplicity increases
- $W+n$ jets cross sections:

Jets	W^- LO	W^- NLO	W^+ LO	W^+ NLO
1	$284.0(0.1)^{+26.2}_{-24.6}$	$351.2(0.9)^{+16.8}_{-14.0}$	$416.8(0.6)^{+38.0}_{-35.5}$	$516(3)^{+29.}_{-23}$
2	$83.76(0.09)^{+25.45}_{-18.20}$	$83.5(0.3)^{+1.6}_{-5.2}$	$130.0(0.1)^{+39.3}_{-28.1}$	$125.1(0.8)^{+1.8}_{-7.4}$
3	$21.03(0.03)^{+10.66}_{-6.55}$	$18.3(0.1)^{+0.3}_{-1.8}$	$34.72(0.05)^{+17.44}_{-10.75}$	$29.5(0.2)^{+0.4}_{-2.8}$
4	$4.93(0.02)^{+3.49}_{-1.90}$	$3.87(0.06)^{+0.14}_{-0.62}$	$8.65(0.01)^{+6.06}_{-3.31}$	$6.63(0.07)^{+0.21}_{-1.03}$
5	$1.076(0.003)^{+0.985}_{-0.480}$	$0.77(0.02)^{+0.07}_{-0.19}$	$2.005(0.006)^{+1.815}_{-0.888}$	$1.45(0.04)^{+0.12}_{-0.34}$



Recent progress

- Number of jets in addition to the vector boson



- Has been called 'NLO revolution'

BlackHat+Sherpa

- BlackHat is a C++ library for virtual one-loop matrix elements
- Recent calculations with Sherpa:
 - W/Z+4 jets
 - 2,3,4 jets
 - Z/gamma ratios with up to 3 jets
 - W+5 jets
 - diphoton+2 jets

n-Tuple files [arXiv:1310.7439]

- High multiplicity NLO calculations are computationally intensive
- It would not be possible to rerun a high multiplicity W+jet calculation every time a new interesting observable comes up
- Matrix elements are expensive, while
 - Jet clustering
 - Observables
 - PDF evaluationare relatively cheap
- Store each matrix element, PS point and the information necessary to change the factorisation and renormalisation scales in large files we call n-Tuple files
- We use ROOT file as storage

n-Tuple files

- Goodies
 - One can change the analysis cuts, add observables
 - Scale variation
 - PDF errors (otherwise extremely expensive)
 - Easy communication between theorists and experimenters
 - No need for specific know-how of the tool which produced the NLO calculation
 - Easier to “endorse” an event file than a program
- Price to pay
 - Large files
 - Generation cuts need to be loose enough to accommodate many analysis → efficiency cost

n-Tuple availability

- The n-Tuple files are available
 - On the grid
 - On castor at CERN
- For a range of processes

Process	Pathname	Energy	Jet cut
W+ + 1,2,3,4 jets	Wp<n>j	7TeV	25GeV
W+ + 1,2,3 jets	Wp<n>j	8TeV	20GeV
W- + 1,2,3,4 jets	Wm<n>j	7TeV	25GeV
W- + 1,2,3 jets	Wm<n>j	8TeV	20GeV
Z/gamma* + 1,2 jets	Zee<n>j	7TeV	25GeV
Z/gamma* + 3,4 jets	Zee<n>j	7TeV	20GeV
Z/gamma* + 1,2,3 jets	Zee<n>j	8TeV	20GeV
2,3,4 jets	PureQCD<n>j	7TeV,8TeV	40GeV

From <http://blackhat.hepforge.org/trac/wiki/Availability>

nTupleReader library

- We provide a C++ library to facilitate the use of the n-Tuple files
- Allows:
 - Change of factorisation and renormalisation scales
 - Change of pdf (from LHAPDF set), including error sets
- Has a Python interface
- Template for a customised implementation
- Available on [hepforge](#)

nTupleReader library

- Example

```
import nTupleReader as NR
r=NR.nTupleReader()
r.addFile('sample.root')

r.setPDF("CT10nlo.LHgrid")
r.setPDFmember(12)

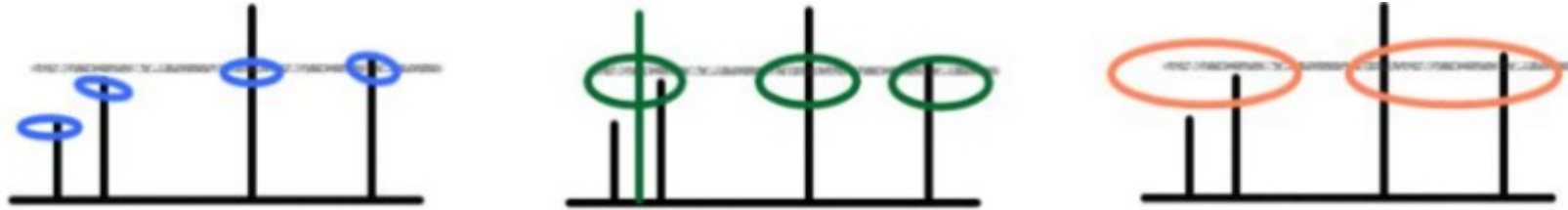
while r.nextEntry():
    # compute new scales
    RenScale = ....
    FacScale = ....
    newWeight=r.computeWeight(FacScale, RenScale)
    // use this weight in the analysis
    ...
```

Pitfalls

- There can be negative weights
- For the real part the matrix elements and the subtraction terms are highly anti-correlated (by construction)
- Some common operations have to be modified to take this into account:
 - statistical error calculation
 - Rebinning, cumulative distributions

NLO event files

- Issues with NLO event files
 - Only fixed number of radii are supported



- Negative weights
- Highly anti-correlated weights (by construction!) affect
 - Integration error estimate
 - Rebinning and cumulative distributions

NLO events

- Standard formula for integration error:

$$\begin{aligned}\sigma^2 &= \langle (\omega - \langle \omega \rangle)^2 \rangle = \langle \omega^2 \rangle - (\langle \omega \rangle)^2 \\ &\simeq \frac{\sum \omega_i^2}{N} - \left(\frac{\sum \omega_i}{N} \right)^2 = \frac{1}{N} \left(\sum \omega_i^2 - \frac{(\sum \omega_i)^2}{N} \right)\end{aligned}$$

- If we have only pairs of large anti-correlated weights

$$\begin{aligned}\omega_+ &\simeq L, \quad \omega_- \simeq -L, \quad \omega_+ - \omega_- \simeq s \\ \sigma^2 &\simeq \frac{1}{N} \left(NL^2 + \frac{(Ns)^2}{N} \right) = L^2 - s^2 \simeq L^2\end{aligned}$$

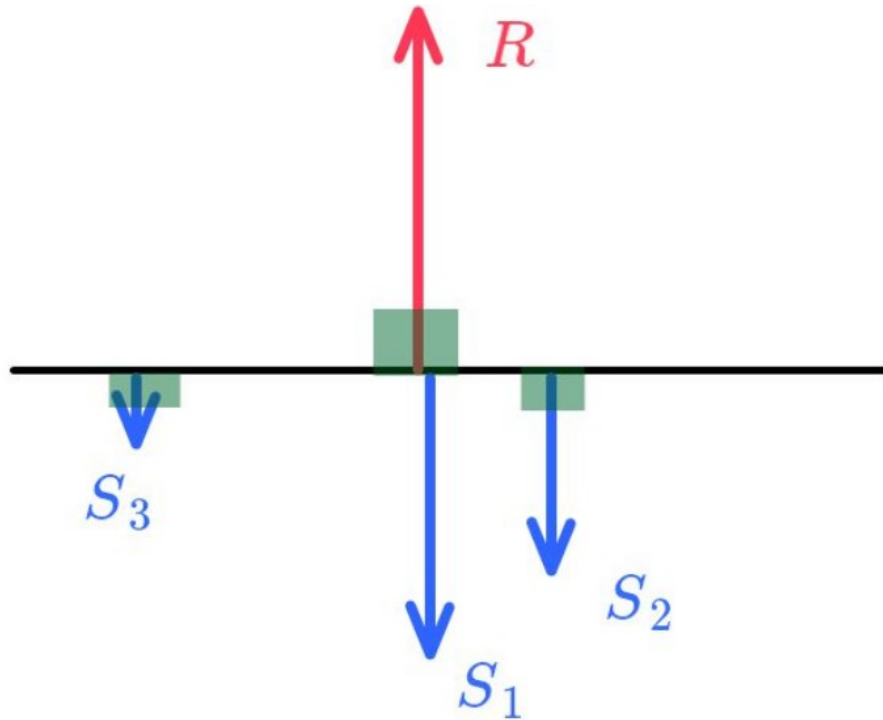
- This is far too large, one should use

instead

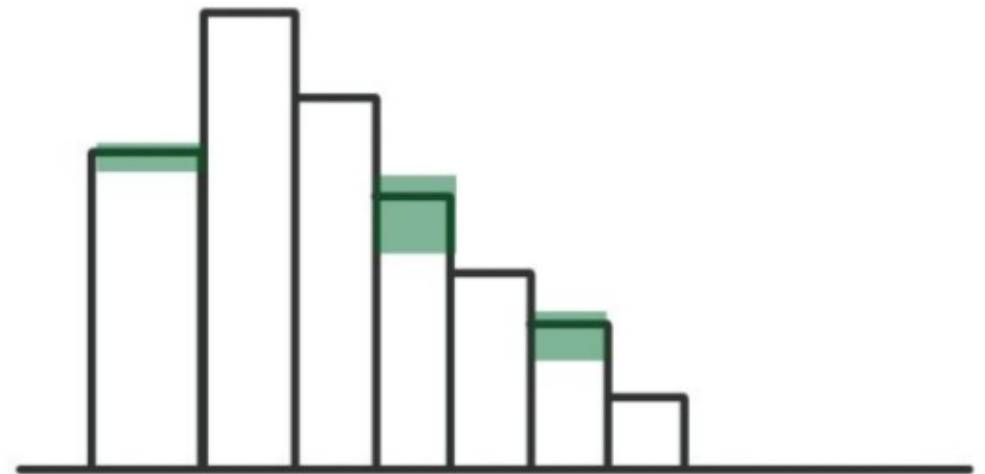
$$\sigma^2 \simeq \frac{1}{N} \left(\sum (\omega_{i,+} + \omega_{i,-})^2 - \frac{(\sum \omega_i)^2}{N} \right)$$

NLO events

- First bin all contribution from the event in auxiliary histogram



- Then transfer to the true histogram



fastNLO

- A direct analysis using n-Tuple files can still be too slow for some applications
 - PDF4LHC-type error require several PDF sets with errors
 - alphas determination
 - PDF fitting
- We can use the n-Tuple files to create fastNLO tables for very quick evaluation
 - Very fast
 - Need one table for each measured distribution
- Similar to the APPLgrid and other projects

fastNLO

- Replace PDF functions with

where the the extrapolation kernels $E(x)$ have the properties

$$f_a(x) \simeq \sum_i f_a(x_i) E(x)$$

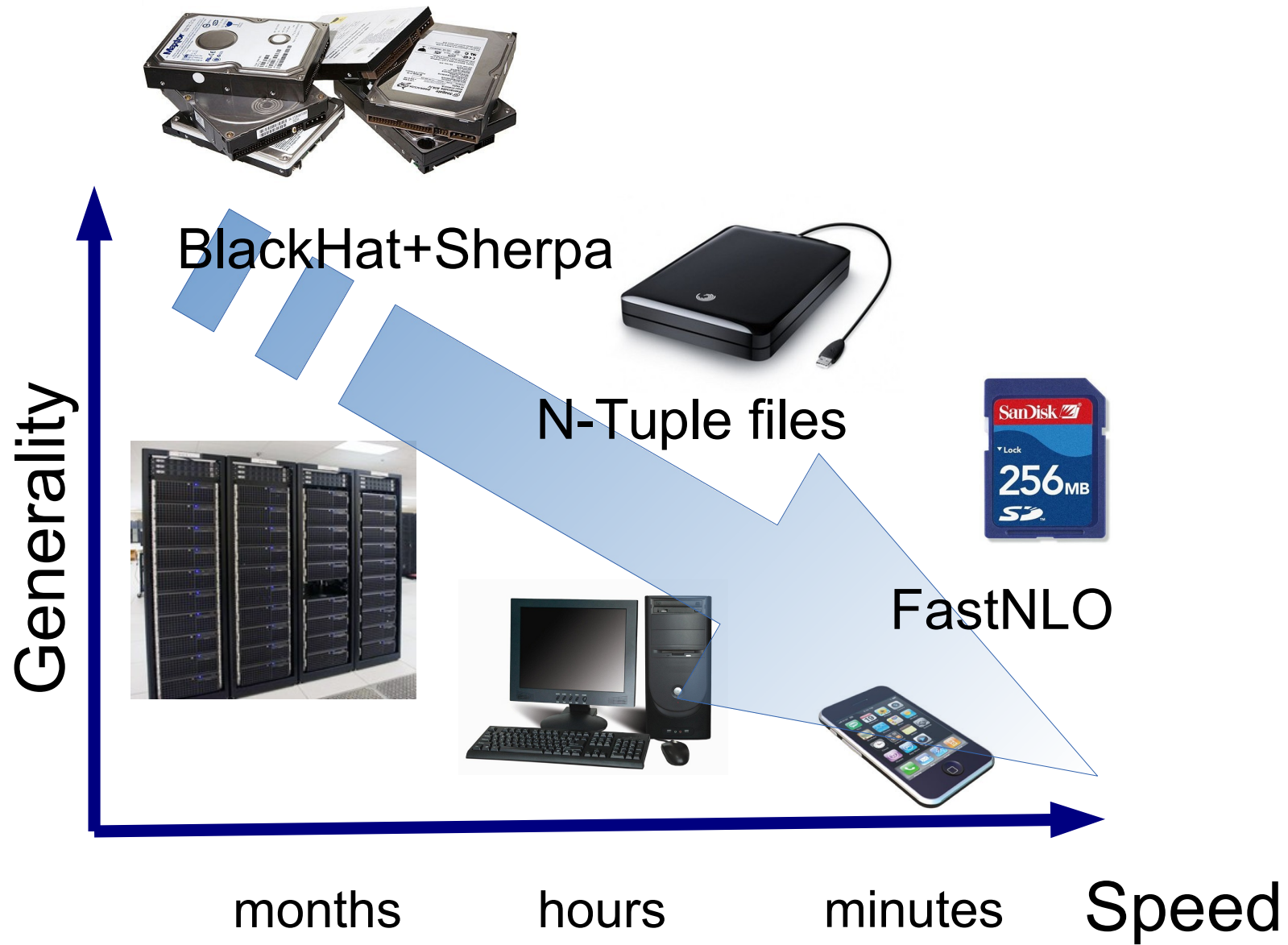
- The cross section can be written as

$$\sum_i E_i(x) = 1, \quad E_i(x_j) = \begin{cases} 1 & \text{if } i = j \\ 0 & \text{otherwise} \end{cases}$$

$$\sigma = \sum_{a,b,i,j} f_a(x_i) f_b(x_j) \underbrace{\int dx_1 dx_2 E_i(x_1) E_j(x_2) \int d\phi \hat{\sigma}(x_1, x_2; \phi)}_{c_{a,b;i,j}}$$

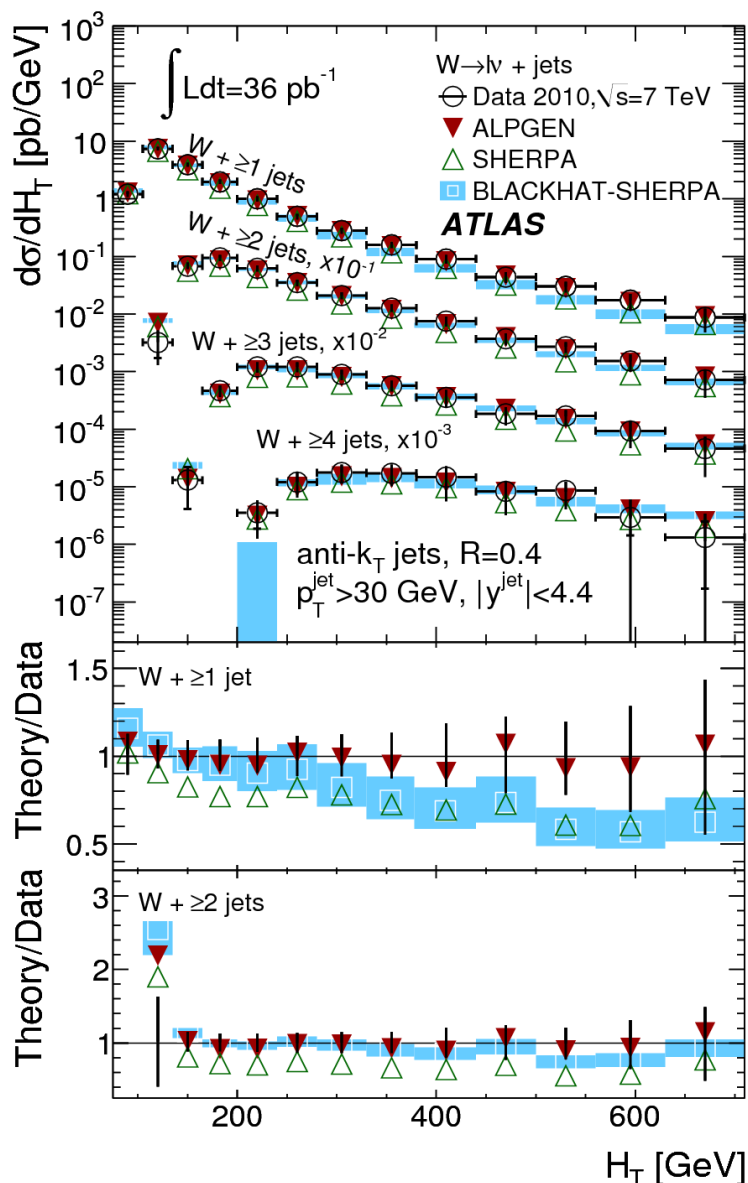
Can compute this with n-Tuples!

Speed vs Generality

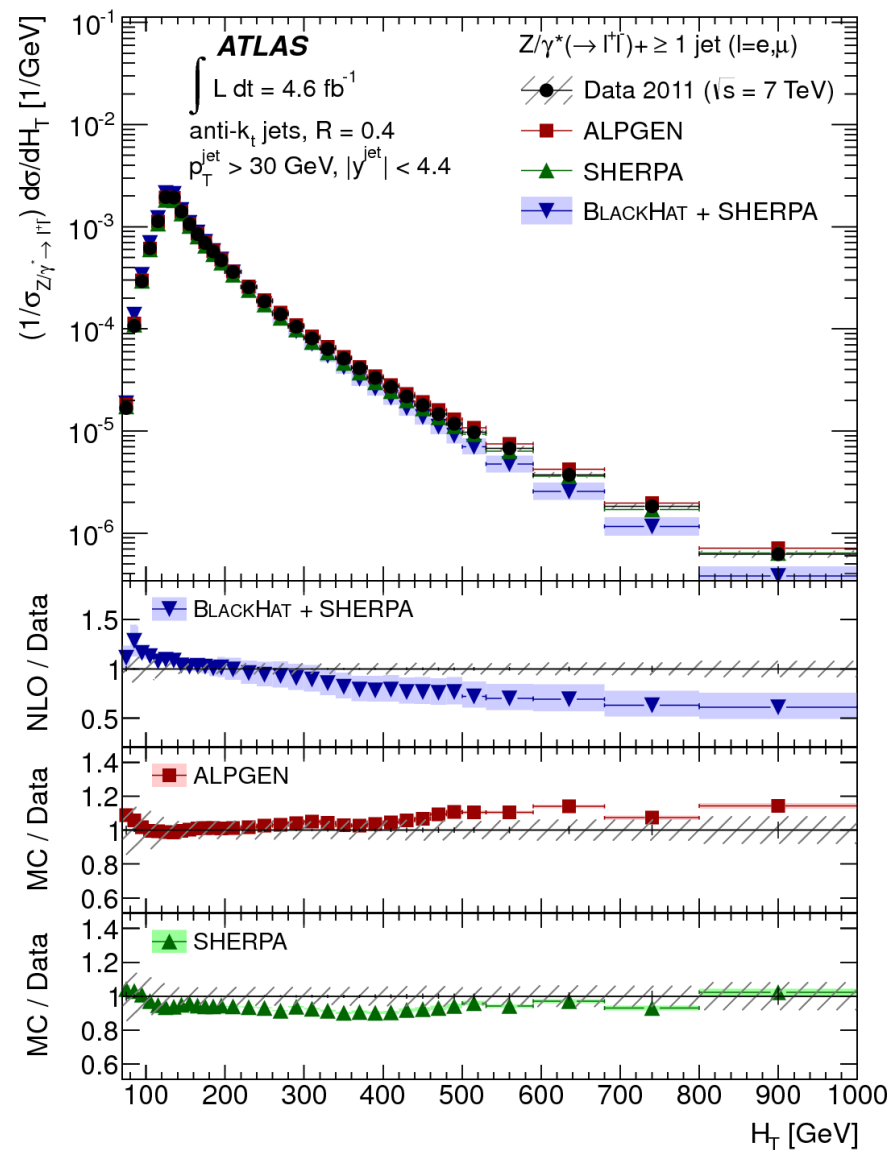


HT distribution

- W+jet



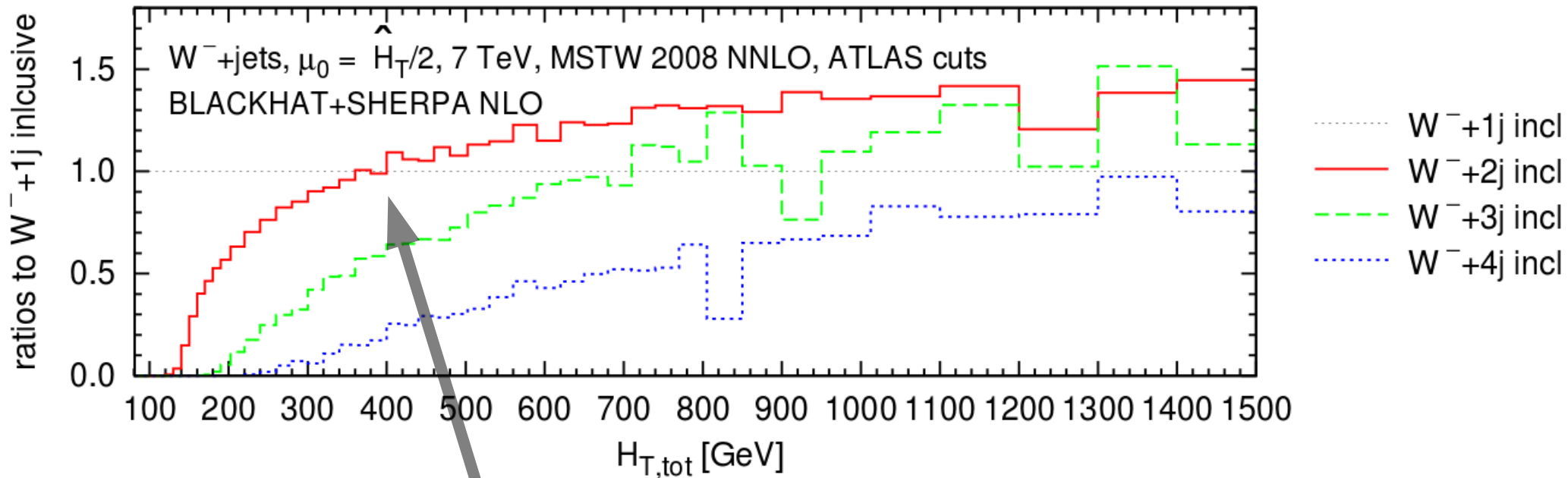
- Z+jets



*HT is basically the sum of all transverse energies of the particles in the event

HT distribution

- Already at quite moderate HT the higher multiplicity samples have sizeable contributions



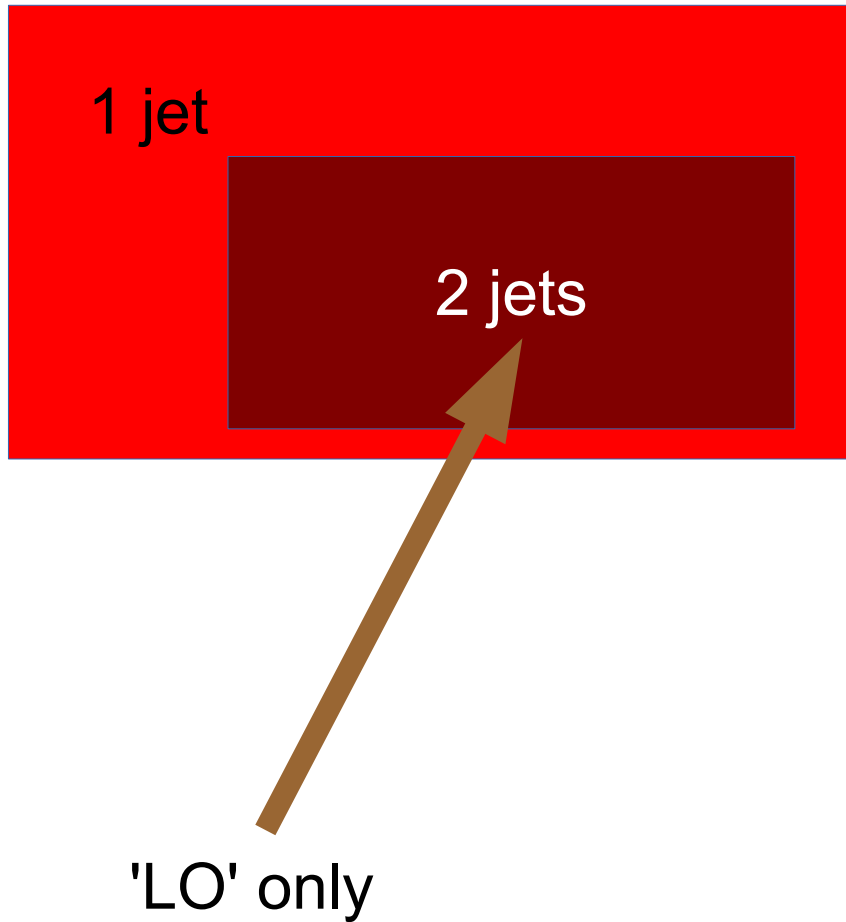
$W+2$ jet contribution “overtakes” $W+1$ jet

Exclusive sums

- Combine NLO event samples of different multiplicity
- Justified (if at all) for observables where higher multiplicities are important
- Avoid double counting by restricting the samples to a fixed multiplicity
- Formally not better than a NLO calculation
- No systematic study of uncertainties/stability
- Very crude method

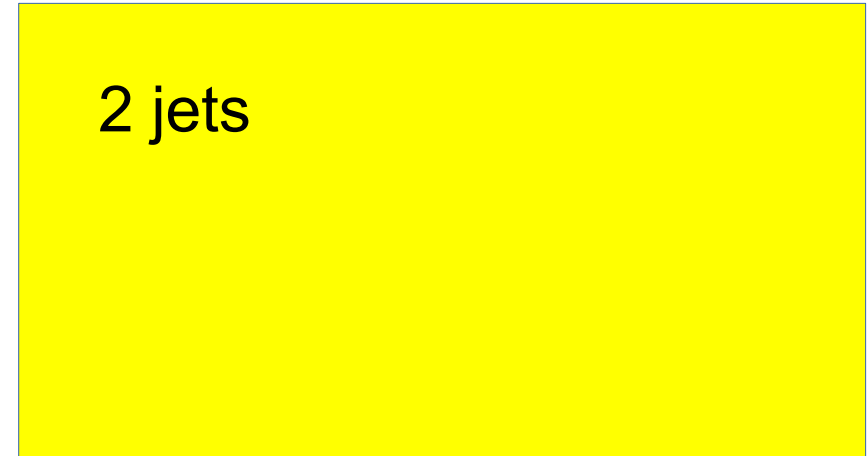
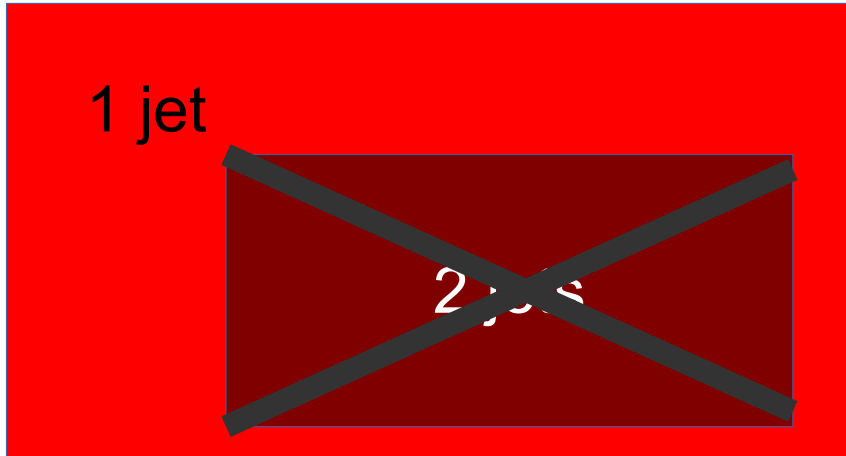
Exclusive sums

- $W+1$ jet at NLO

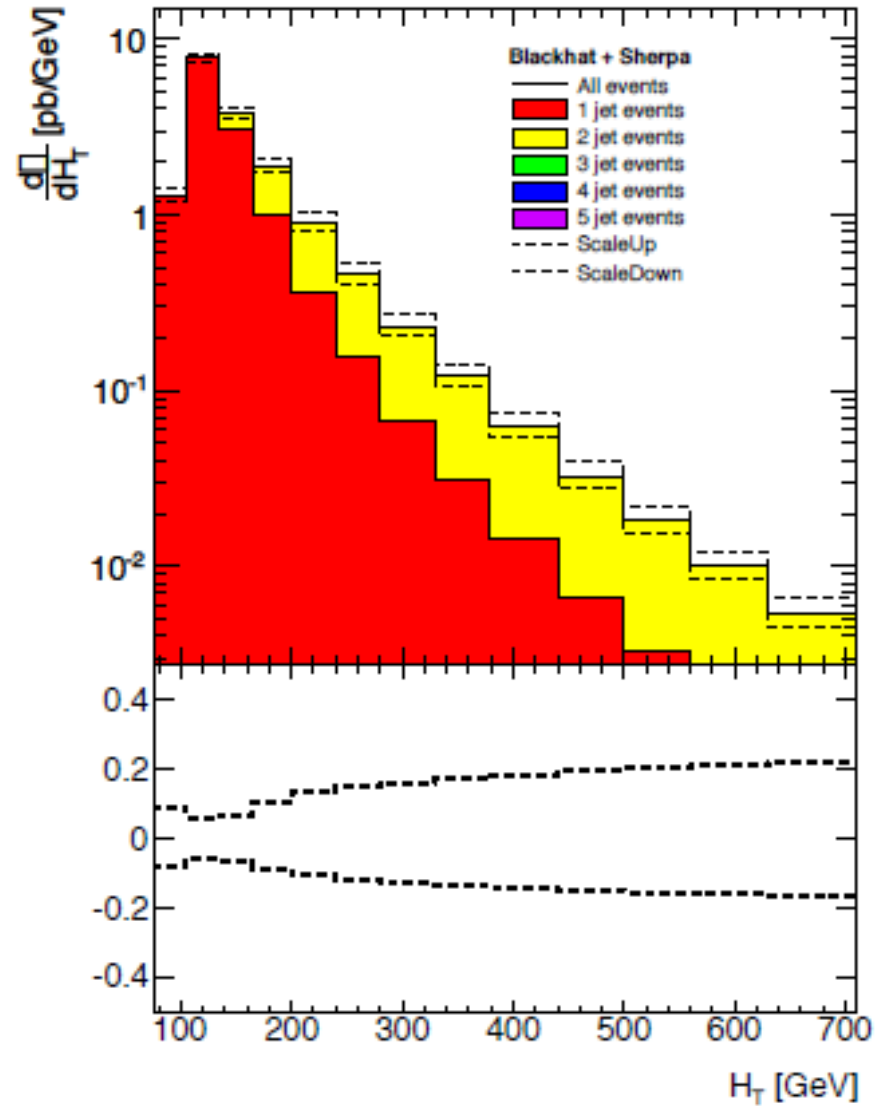


Exclusive sums

- W+1 jet at NLO W+2 jets at NLO

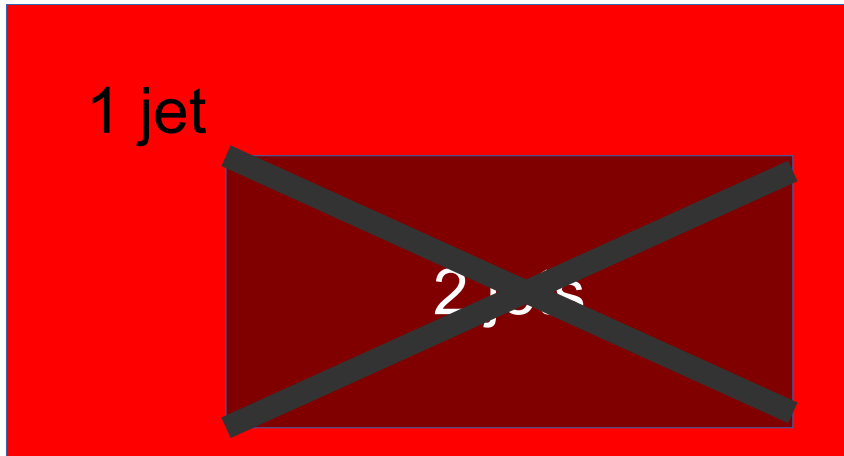


W+1 jet



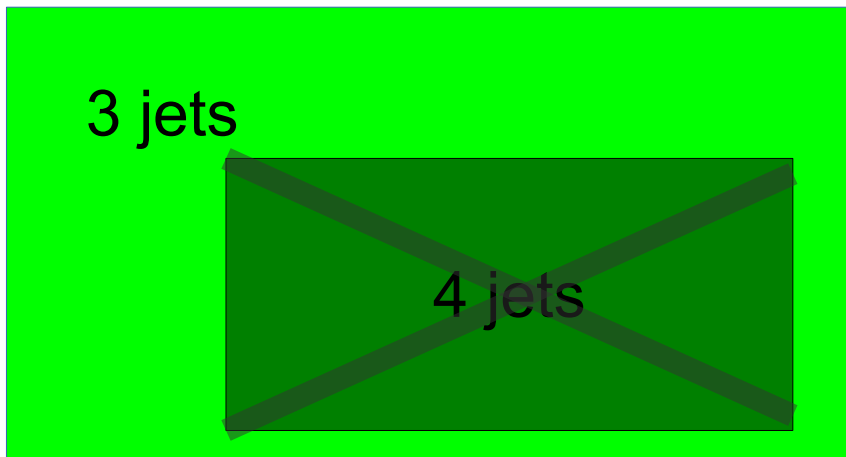
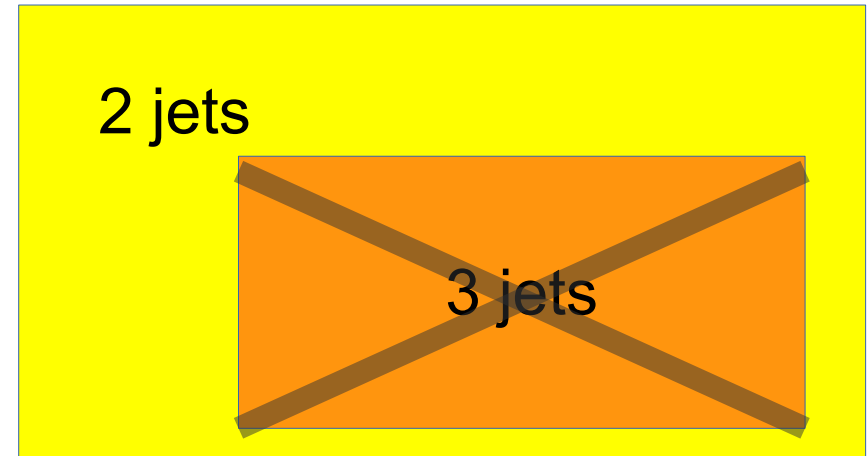
Exclusive sums

- W+1 jet at NLO W+2 jets at NLO



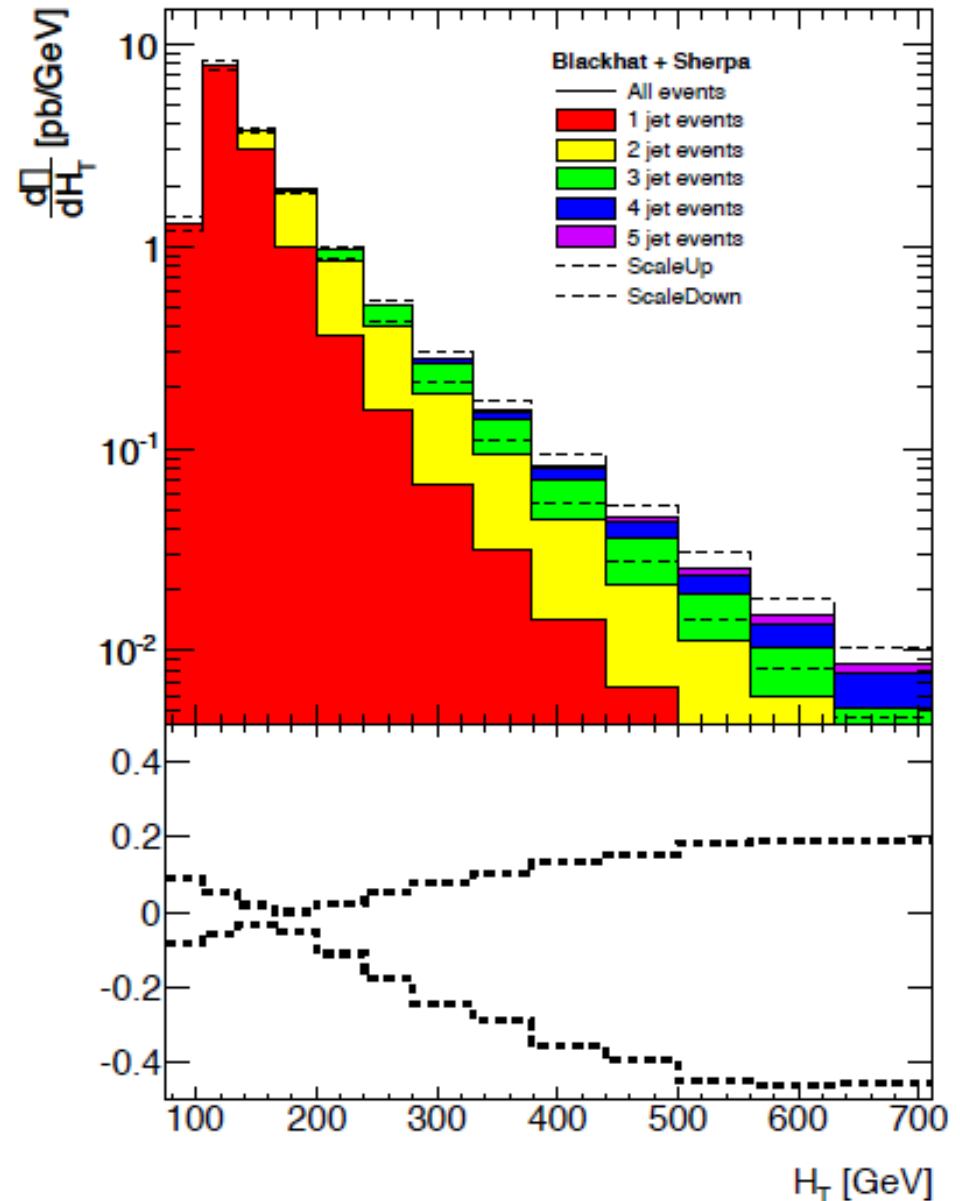
Exclusive sums

- W+1 jet at NLO W+2 jets at NLO



Exclusive sums

- Scale variation much larger than at NLO
- Need to be investigated more precisely
- Combination can be made 'official' using LoopSim [Rubin,Salam,Sapeta] (under investigation)
- Better : 'ME+PS'-type merging

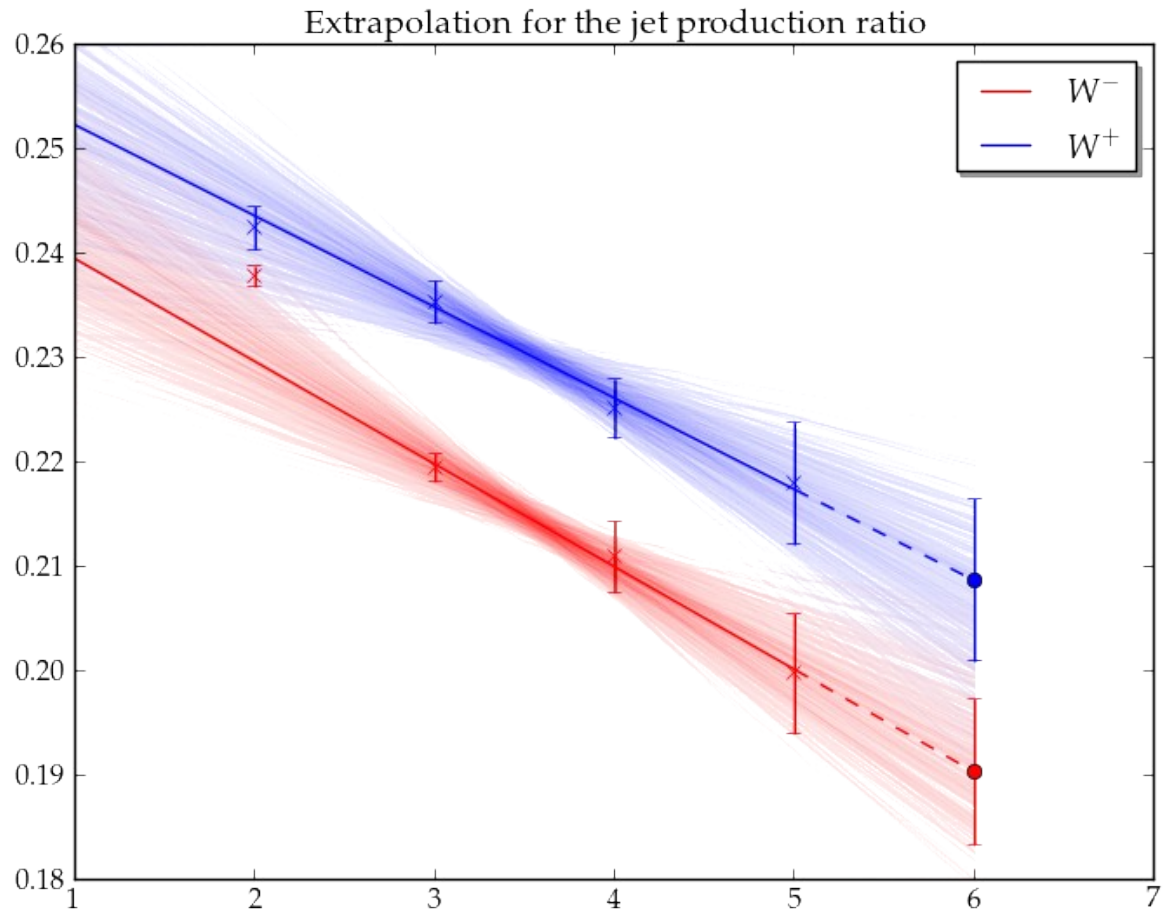


Towards higher multiplicities?

- We have a lot of prediction for high multiplicity processes at NLO
- We can try to find 'universal' properties/features
- Usually need to discard 0-jet and 1-jet because new partonic channels open
- Usually these features are more easily seen in ratios between multiplicities

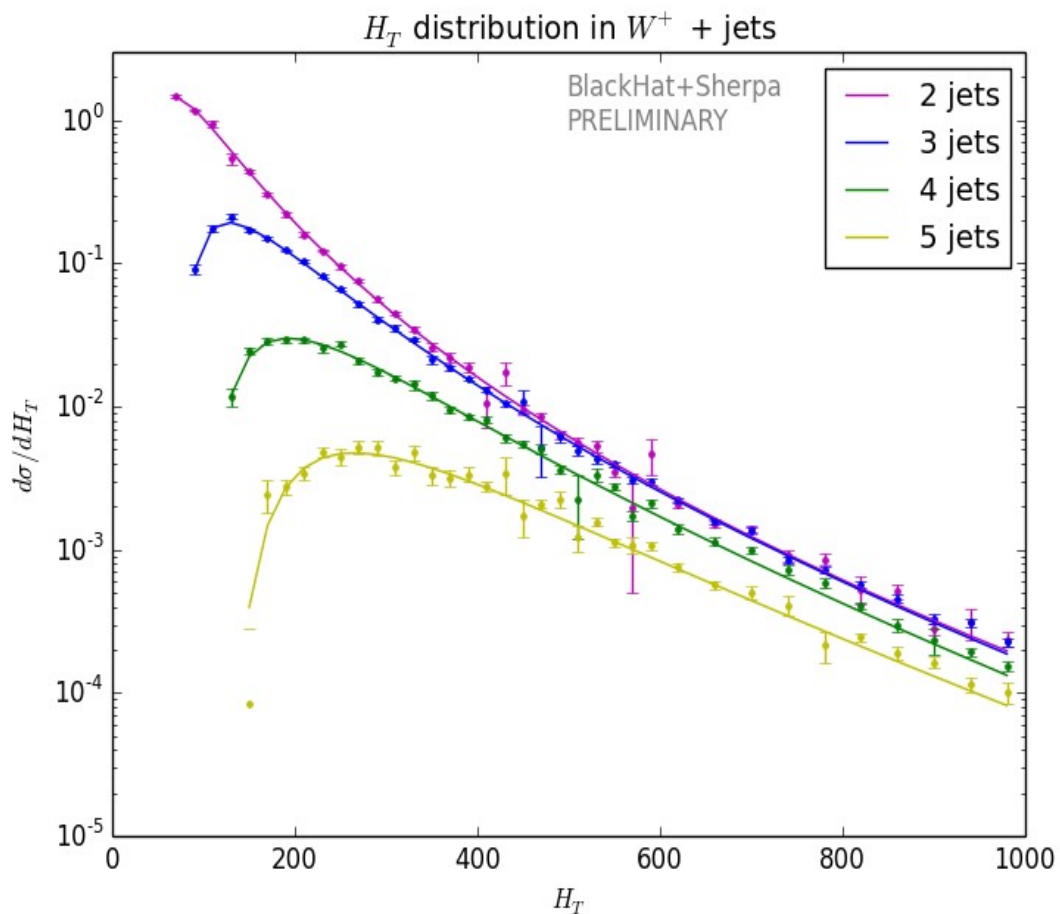
Extrapolation for ratios

- Ratio $V+n \text{ jets}/(V+n-1 \text{ jets})$
- Consistent with straight line for $n>2$
- Use extrapolation for 6 jets:
- W^- : $0.15 \pm 0.01 \text{ pb}$
- W^+ : $0.30 \pm 0.03 \text{ pb}$
- Consistent with extrapolation of charge asymmetry
- Error estimates through Monte Carlo method



Distributions

- What about distributions?
- Look at sum of transverse energies of the jets (H_T)
- Cannot extrapolate the value of each bin separately
 - Statistical errors are too large
 - Different thresholds
 - Different peak positions



Distributions

- Instead find a parametrisation and extrapolate the parameters of the parametrisation
- Ansatz for the HT distribution:

$$\frac{d\sigma_{V+n}}{dH_T} = \left(\frac{N_C \alpha_s}{2\pi} \right)^n f(H_T) \mathcal{N}_n \ln^{\tau_n} \rho_{H,n} \left(1 - H_T / H_T^{\max} \right)^{\gamma_n}$$

$$\rho_{H,n} = H_T / (np_T^{\min})$$

HT Distribution

- Instead find a parametrisation and extrapolate the parameters of the parametrisation
- Ansatz for the HT distribution:

$$\frac{d\sigma_{V+n}}{dH_T} = \left(\frac{N_C \alpha_s}{2\pi} \right)^n f(H_T) \mathcal{N}_n \ln^{\tau_n} \rho_{H,n} \left(1 - H_T / H_T^{\max} \right)^{\gamma_n}$$

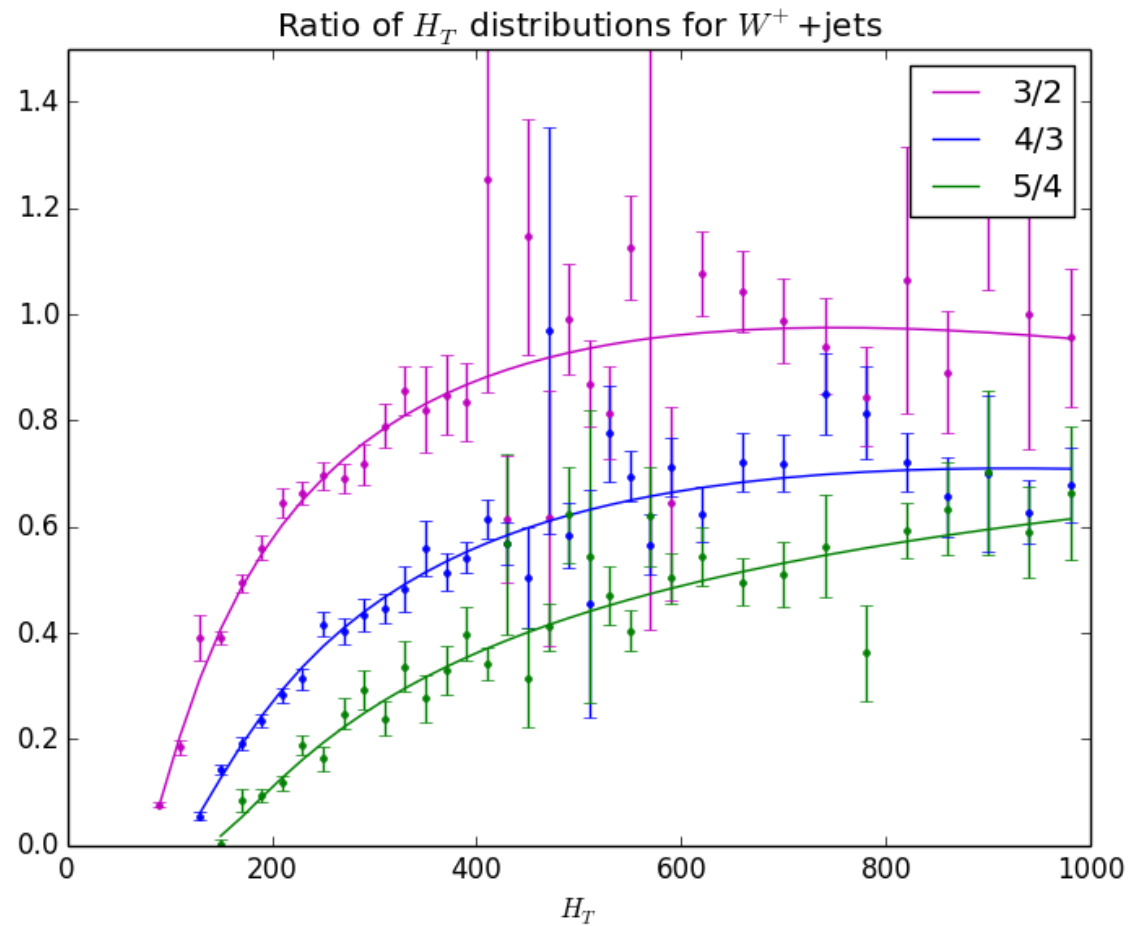
Independent of n

parameters

$$\rho_{H,n} = H_T / (np_T^{\min})$$

HT distribution

- Fit ratios to get the parameters



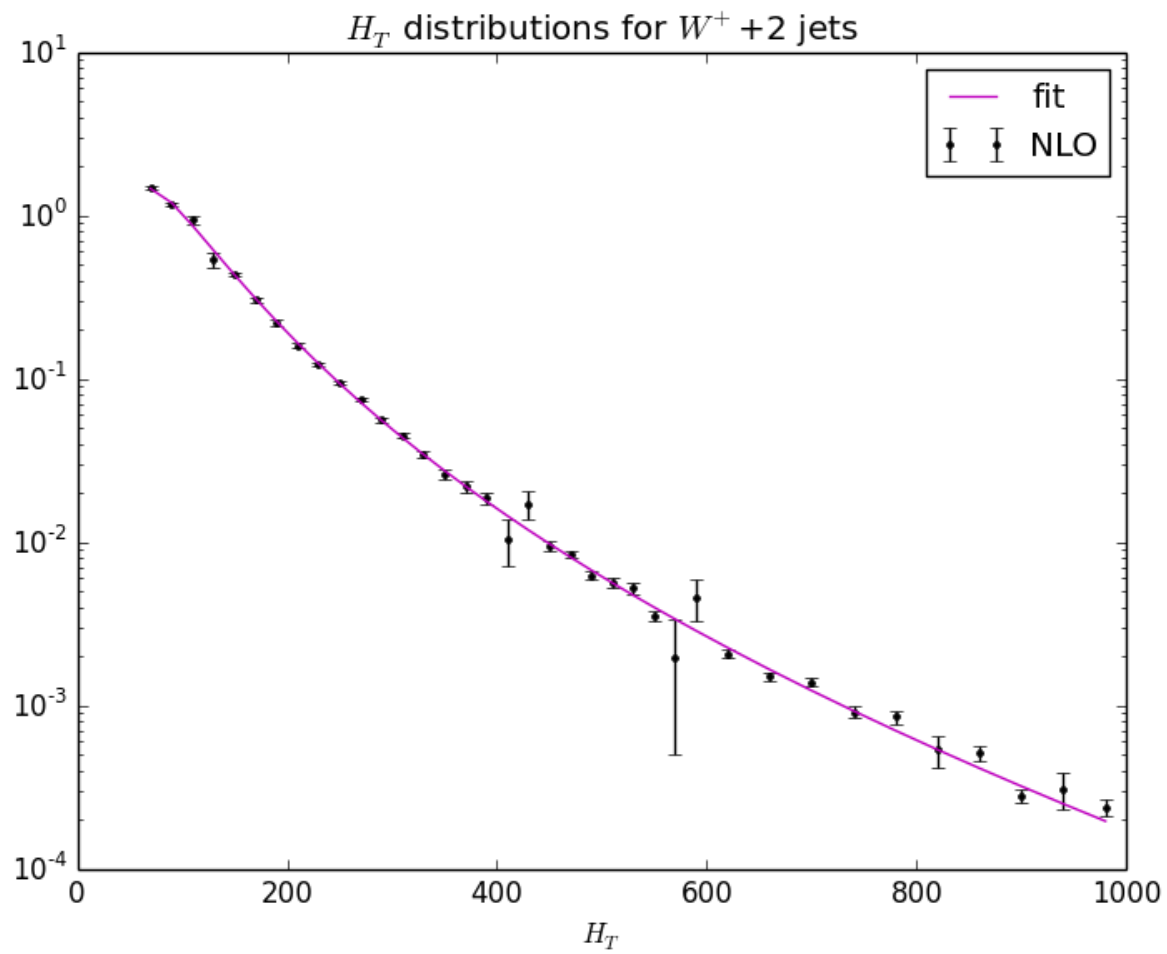
HT distribution

- With the parameters one can extract $f(H)$ from a distribution
- But it is more convenient to have an analytical form for it
- We can use the following form

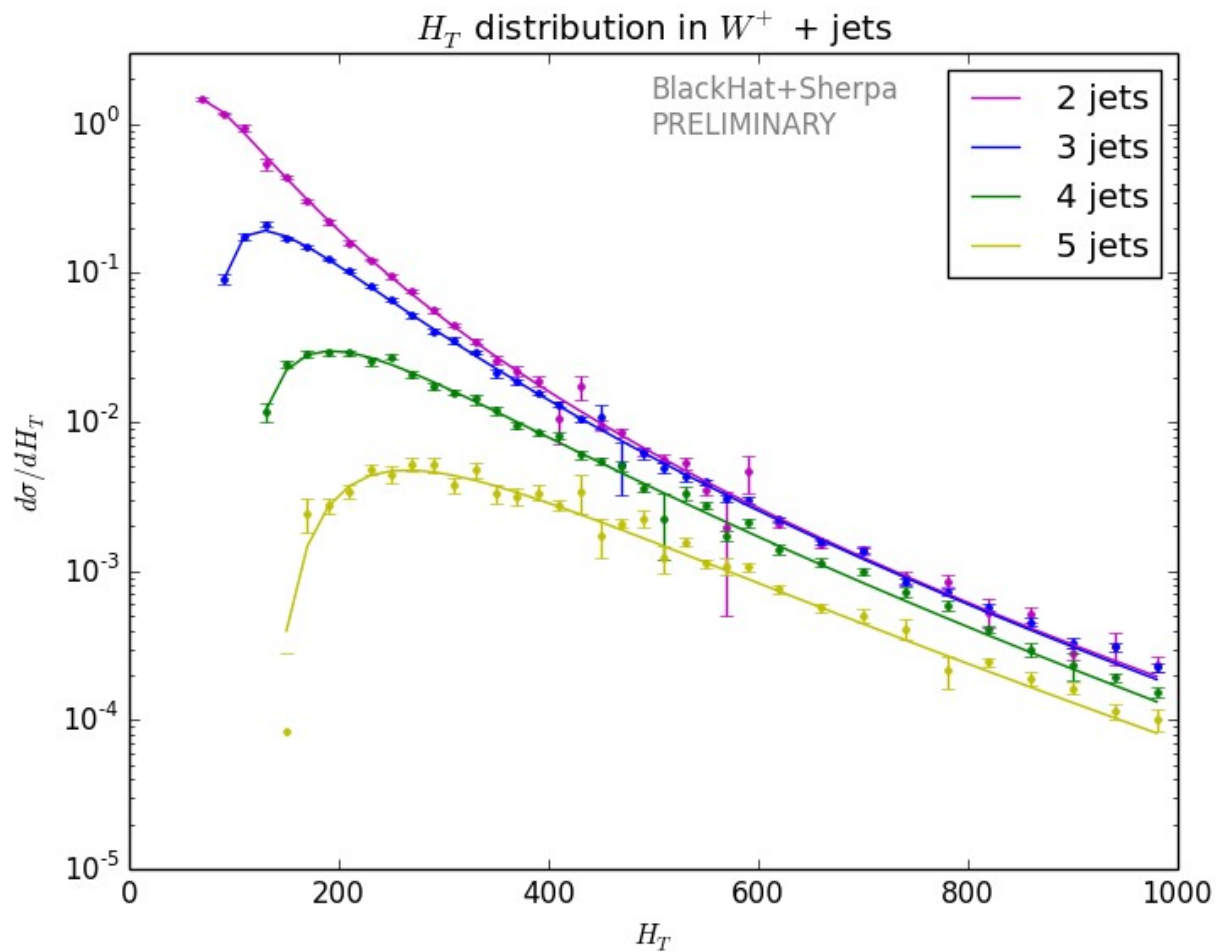
$$f(H) = c \ln^r (H/10) \left(\frac{H}{2p_T^{\min}} \right)^{\omega_2} e^{-h_* H},$$

and extract the parameters from the W+2 jets distribution

HT distribution



HT distribution



Distributions

- Extrapolated HT distribution
- Uncertainty bands are estimated using a MC method

