### $\alpha_s \rightarrow 0$ : easier said than done

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## The challenge: idea



Structure of NLL:

 $\Sigma_{\mathsf{NLL}}(\alpha_{\mathsf{s}}\mathcal{L}, \alpha_{\mathsf{s}}) = e^{g_1(\alpha_{\mathsf{s}}\mathcal{L})\mathcal{L} + g_2(\alpha_{\mathsf{s}}\mathcal{L}) + \dots}$ 

NLL accuracy test:

$$\frac{\Sigma_{\mathsf{MC}}}{\Sigma_{\mathsf{NLL}}} \stackrel{\alpha_s \to 0}{\to} f(\lambda = \alpha_s L) \stackrel{?}{=} 1$$

Here: discuss challenges associated with  $\alpha_s \rightarrow 0$ 

Say we take  $\lambda = \alpha_s L = 0.5$ 

$\alpha_s$	L	
0.04	12.5	
0.02	25	
0.01	50	
0.005	100	

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  - $\Rightarrow$  Shower cut should be smaller

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Deal with numbers over large numerical range  $\Rightarrow$  precision impaired

### Challenge 2:

$$g_1(\alpha_s L)L \gg 1 \Rightarrow \Sigma(\lambda, \alpha_s) \ll 1$$
  
 $\Rightarrow$  no events with standard "unweighted" techniques

(e.g. 
$$\lambda = 0.5$$
,  $\alpha_s = 0.005 \Rightarrow \Sigma_{y_{23}}(L) \sim 10^{-29}$ )

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# Sudakov suppression: $\beta_{shower} = \beta_{obs}$ (easy case)

Estimate contribution  $e^{L_{approx}}$  to obs PanScales:  $\begin{cases} L_{approx} = \ln k_t - \beta_{obs} |\eta|, \\ \eta = \overline{\eta} + \frac{1}{\beta} \ln \rho \end{cases}$ 

consider bins in  $L_{approx}$  focus on a single bin



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#### Element 1: weights

- Generate first emission in given bin
- Weight corresponding to Sudakov above the considered bin

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Element 2: dynamic cut-off

cut shower at factor  $e^{-\delta L}$  below largest contribution

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# Sudakov suppression: $\beta_{shower} \neq \beta_{obs}$ (harder case)



Estimate contribution 
$$e^{L_{approx}}$$
 to obs  
PanScales: 
$$\begin{cases} L_{approx} = \ln k_t - \beta_{obs} |\eta|, \\ \eta = \bar{\eta} + \frac{1}{\beta} \ln \rho \end{cases}$$

#### Element 1: weights+event veto

- Generate first emission in given bin
- Weight corresponding to Sudakov above the considered bin

• Veto event if 
$$L_{approx} > L_{approx}^{(min)}$$

Element 2: emission veto

veto emissions at factor  $e^{-\delta L}$ below largest contribution

# Numerical range

• For most shapes,  $\sqrt{\epsilon} = \sqrt{\epsilon_{\text{double}}} \approx 10^{-8} \ (\delta L \approx 18)$  does the job (covered range between  $\sqrt{\epsilon}L$  and L)

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- For  $k_t$  subjet multiplicity, need full clustering  $\Rightarrow$  need full kinematic range
  - $\Rightarrow$  rely on higher-precision types

type	precision		slow-down
double	$\epsilon pprox 10^{-16}$	$L \approx 37$	1
ddreal	$\epsilon pprox 10^{-31}$	$L \approx 71$	10
qdreal	$\epsilon pprox 10^{-64}$	$L \approx 147$	141

Just OK for our needs but non-negligible time cost (Note that multiplicity is a series in  $\sqrt{\alpha_s}L$ )

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Just OK for our needs but non-negligible time cost (Note that multiplicity is a series in  $\sqrt{\alpha_s}L$ )

• For  $E_t$  in a slice, naive  $\delta L$  does not work; use angular cut-off instead.

- From n values of α<sub>S</sub> (typically n = 3), determine a polynomial of degree n − 1
- The constant term is the extrapolation up to  $\mathcal{O}(\alpha_s^n)$
- Statistical uncertainties easily propagated
- $\bullet$  Can try to get an estimate of "systematic" uncertainties using different sets of  $\alpha_{s}$  values

(mostly helpful for multiplicity where convergence was slower)

## Other considerations

Computation of angles

- computing " $1 \cos \theta$ " has precision  $\sqrt{\epsilon}$  in collinear limit.
- ullet switch to a cross product at small angles to bring down to  $\epsilon$
- $\bullet$  alignment along the z axis  $\Rightarrow$  angles down to  $\epsilon \theta_{\max}$
- benefit from precision at various stages, e.g. precise determination of the thrust axis

Multiplicities can get large

- $N \propto \exp(\sqrt{\alpha_s L^2}) \propto \exp(\sqrt{\lambda L})$  blows up if  $L \gg 1$  at fixed  $\lambda$
- currently limited by previous prescriptions but speed useful nonetheless
- adopted a "Roulette Wheel" strategy for dipole choice
- fast observable calculations (e.g. clustering)