

# Automatic threshold resummation in MG5\_aMC

Paolo Torrielli

Università di Torino

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In collaboration with P. Artoisenet, E. Laenen, F. Maltoni, L. Vernazza

# Outline

- ▶ Motivations
- ▶ Theoretical framework
- ▶ Implementation
- ▶ Validation
- ▶ Preliminary results
- ▶ Status and future directions

## Motivations

- ▶ Important to resum soft-gluon effects for precision phenomenology of a large class of processes.
- ▶ Automation of this resummation possible: achieving this in MG5\_aMC could significantly extend the set of processes for which these threshold effects are known.
- ▶ A step towards including analytic resummation among the predictions offered by MG5\_aMC.
- ▶ Could be a basis/template for the achievement of other kinds of resummation in the same framework and with similar methods.

## Theoretical framework: setup

- ▶ Processes with no massless final-state QCD particles.
- ▶ Kinematics:  $z = M^2/s$ ,  $\tau = M^2/S$ ,  $M =$  Born-system invariant mass,  $S =$  hadronic,  $s =$  partonic CM energy squared.
- ▶ Resum dominant  $\ln(1 - z)$  in the threshold limit  $z \rightarrow 1$ : extra radiation soft with energy  $\sim \sqrt{s}(1 - z)$ . So called ‘pair’-invariant-mass (PIM) kinematics.
- ▶ Use SCET framework to perform the resummation in momentum space.

## Theoretical framework: factorisation

- ▶ QCD factorisation theorem (with  $\mu_r = \mu_f$ ):

$$d\sigma = \frac{1}{2S} \sum_{ij} \int_{\tau_{\min}}^1 \frac{d\tau}{\tau} \int_{\tau}^1 \frac{dz}{z} \mathcal{L}_{ij}\left(\frac{\tau}{z}, \mu_f\right) C_{ij}(z, M, \mu_f) d\Phi_n,$$

$\mathcal{L}$  = luminosity, and partonic cross section  $d\hat{\sigma}_{ij} \propto \int dM C_{ij} d\Phi_n$ .

- ▶ At threshold, kernel  $C_{ij}$  factorises as

$$C(z, M, \mu_f) = \text{Tr}[\mathbf{H}(M, \mu_f) \mathbf{S}(\sqrt{s}(1-z), \mu_f)] + \mathcal{O}(1-z).$$

- ▶  $\mathbf{H}$  and  $\mathbf{S}$  matrices in colour space:  $\mathbf{H}_{AB} = \frac{\langle c_A | \mathbf{H} | c_B \rangle}{\langle c_A | c_A \rangle \langle c_B | c_B \rangle}$ .
- ▶  $|c_A\rangle$  = orthogonal basis of possible colour structures of the process.  
Example: in  $q\bar{q} \rightarrow Q\bar{Q}$ ,  $|c_1\rangle = \delta_{i_2 i_1} \delta_{i_3 i_4}$ ,  $|c_2\rangle = t_{i_2 i_1}^a t_{i_3 i_4}^a$ .
- ▶  $\mathbf{H}$  ( $\mathbf{S}$ ) includes virtual (real) contributions in the threshold limit.

## Theoretical framework: resummation through RG evolution

- ▶  $\text{Tr}[\mathbf{H}(\mu_f)\mathbf{S}(\mu_f)]$  = threshold approximation of fixed-order cross section (at lowest order = full Born since Born lives at  $z = 1$ ).
- ▶  $\mathbf{H}$  satisfies RG equation  $\frac{d}{d \log \mu} \mathbf{H}(\mu) = \mathbf{\Gamma}_H(\mu)\mathbf{H}(\mu) + \mathbf{H}(\mu)\mathbf{\Gamma}_H^\dagger(\mu)$ .
- ▶ Analogous (but more complicated) equation for  $\mathbf{S}$ , deduced imposing RG invariance of total cross section and using DGLAP.
- ▶ All-order solutions:  $\mathbf{H}(\mu) = \mathbf{U}_h(\mu_h, \mu)\mathbf{H}(\mu_h)\mathbf{U}_h^\dagger(\mu_h, \mu)$ ,  
 $\tilde{\mathbf{s}}(\mu) = \mathbf{U}_s(\mu_s, \mu)\tilde{\mathbf{s}}(\mu_s)\mathbf{U}_s^\dagger(\mu_s, \mu)$ .
- ▶  $\mathbf{U}_{h/s}(\mu_1, \mu_2) = \mathcal{P} \exp \int_{\mu_1}^{\mu_2} \frac{d\mu'}{\mu'} \mathbf{\Gamma}_{H/S}(\mu')$ : diagonalise  $\mathbf{\Gamma}_{H/S}$  to solve RG evolution.
- ▶ Evolution matrix  $\mathbf{U}$  contains threshold logarithms.  
Resummation in direct space (SCET inspired): replace  $\mathbf{H}^{\text{fo}}(\mu_f)$  with  $\mathbf{U}_h(\mu_h, \mu_f)\mathbf{H}^{\text{fo}}(\mu_h)\mathbf{U}_h^\dagger(\mu_h, \mu_f)$ , and similarly for  $\mathbf{S}$ .

## Theoretical framework: resummation formula

- ▶ Perturbative accuracy of  $\mathbf{H}$  and  $\mathbf{S}$  determines log accuracy of the resummation: (N)LO  $\mathbf{S}$  and  $\mathbf{H}$  give (N)NLL resummation.
- ▶ NLL resummation formula

$$C_{ij}^{\text{NLL}} = \exp[2(a_i^0 + a_j^0)] \text{Tr}[\mathbf{U}^0(\mu_h, \mu_s) \mathbf{H}_{ij}^0(\mu_h) \mathbf{U}^{0\dagger}(\mu_h, \mu_s) \tilde{\mathbf{s}}^0(\mu_s)] \frac{e^{-2\gamma_E \eta}}{\Gamma(2\eta)} \frac{z^{-\eta}}{(1-z)^{1-2\eta}}$$

- ▶ All quantities evaluated at lowest order,  $\mathbf{H}^0$  obtained through Born information,  $\tilde{\mathbf{s}}$  = Laplace transform of  $\mathbf{S}$ :

$$(H_{ij}^0)_{AB} = \frac{1}{4} \frac{\langle c_A | \mathcal{M}_b \rangle \langle \mathcal{M}_b | c_B \rangle}{\langle c_A | c_A \rangle \langle c_B | c_B \rangle} \quad \tilde{s}_{AB}^0 = \langle c_A | c_B \rangle$$

- ▶  $a_k$  and  $\eta$  = colour factors times integrated anomalous dimensions

## The method

- ▶ Process dependence encoded in **H**: can **automate** this resummation at the same level as automatic fixed-order code.
- ▶ Need to build a general structure to handle the process-independent parts.
- ▶ We chose the **reweighting** way.  
Produce LO (for NLL resummation) events, reweight each through  $C_{ij}^{\text{NLL}}$  evaluated at the specific event kinematics, and obtain the resummed  $d\sigma/dM$ .
- ▶ Strengths: **automatic**, **fast** (if evaluation and integration of  $C_{ij}^{\text{NLL}}$  is), **parallelisable**, **modular** (different reweighting modules to achieve different resummations), **accurate** (accuracy given by size of original LO sample).
- ▶ Having kinematic configurations one could in principle obtain as many observables as wanted, as with a shower. Question: logarithmic accuracy for those?



## Implementation: colour basis

$$(H_{ij}^0)_{AB} = \frac{1}{4} \frac{\langle c_A | \mathcal{M}_b \rangle \langle \mathcal{M}_b | c_B \rangle}{\langle c_A | c_A \rangle \langle c_B | c_B \rangle} \quad \tilde{s}_{AB}^0 = \langle c_A | c_B \rangle$$

- ▶ Implemented all colour bases  $|c_A\rangle$  for 2 coloured  $\rightarrow$  0,1,2 coloured +  $N$  colourless
- ▶ The interface automatically projects the Born matrix elements of `matrix.f` on the basis, to get  $\mathbf{H}^0$ .
- ▶  $\mathbf{H}^0$  validated against literature for  $3 \otimes \bar{3} \rightarrow 1$  (Drell-Yan) and for all  $2 \rightarrow 2$  colour combinations (but  $8 \otimes 8 \rightarrow 8 \otimes 8$ , still to fully test).
- ▶  $\tilde{s}^0$  automatically follows from the implementation of the basis.

## Implementation: RG evolution

$$\mathbf{U}(\mu_h, \mu) = \mathcal{P} \exp \int_{\mu_h}^{\mu} \frac{d\mu'}{\mu'} \mathbf{\Gamma}_H(\mu')$$

- ▶ Need to diagonalise  $\mathbf{\Gamma}_H$  in evolution matrix  $\mathbf{U}$  to solve RG.
- ▶ Efficient and fast **numerical diagonalisation** of complex squared matrices using dedicated routines from the **LAPACK** package.
- ▶ **Excellent speed performances** even for high-rank matrices.

## Implementation: reweighting

- ▶ Born weight  $\mathcal{P}_{\text{LO}}$  for kinematics  $\mathbf{X}$ :

$$d\sigma_{\text{LO}} = \frac{1}{2S} \sum_{ij} \int_{\tau_{\text{min}}}^1 \frac{d\tau}{\tau} d\Phi_n \mathcal{P}_{\text{LO}}(\mathbf{X}) = \sum_{i=1}^N e_i(\mathbf{X}_i)$$

- ▶ Resummed weight for kinematics  $\mathbf{X}$

$$d\sigma_{\text{NLL}} = \frac{1}{2S} \sum_{ij} \int_{\tau_{\text{min}}}^1 \frac{d\tau}{\tau} d\Phi_n \mathcal{P}_{\text{NLL}}(\mathbf{X}),$$

with  $\mathcal{P}_{\text{NLL}}(\mathbf{X}) = \int_{\tau}^1 \frac{dz}{z} \mathcal{L}_{ij}(\frac{\tau}{z}, \mu_f) C_{ij}^{\text{NLL}}$ .

- ▶  $d\sigma_{\text{NLL}}$  obtained reweighting LO events by  $\mathcal{P}_{\text{NLL}}$ :

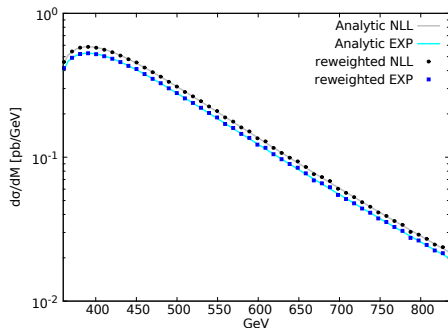
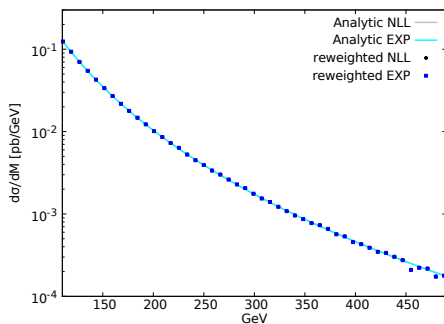
$$d\sigma_{\text{NLL}} = \sum_{i=1}^N e_i(\mathbf{X}_i) \times \frac{\mathcal{P}_{\text{NLL}}(\mathbf{X}_i)}{\mathcal{P}_{\text{LO}}(\mathbf{X}_i)}.$$

- ▶  $\mathcal{P}_{\text{NLL}}(\mathbf{X})$ : 2D integration by quadrature methods, to maximise speed.

## Implementation: general considerations

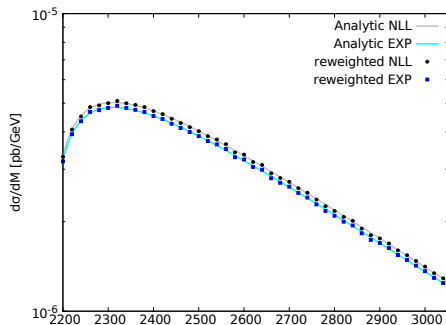
- ▶ Good speed and accuracy performance: **one event typically reweighted in few to few tens of milliseconds.**
- ▶ Possibility to include scale variations for all scales involved:  $\mu_f$ ,  $\mu_s$ , and  $\mu_h$  (but, yes, one run per variation).
- ▶ **Highly parallelised:** with 100 jobs, 1M events for a  $2 \rightarrow 3$  process is reweighted in few minutes per scale variation.
- ▶ Parallelisation may be interesting for MadSpin developers/users: by implementing the same job organisation there, that tool as well could be parallelised, **with huge performance improvements.**

## Validation: Drell-Yan and $t\bar{t}$



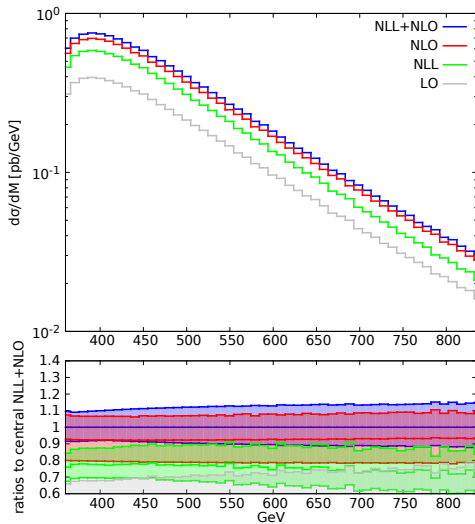
- ▶ Simplified DY (only  $\gamma^*$ , with  $m_{ee} > 100$  GeV – left) and  $t\bar{t}$  (right) @ LHC7.
- ▶  $M$  = Born system invariant mass.
- ▶ Lines = analytic codes, dots = reweighting interface.

## Validation: stop pair



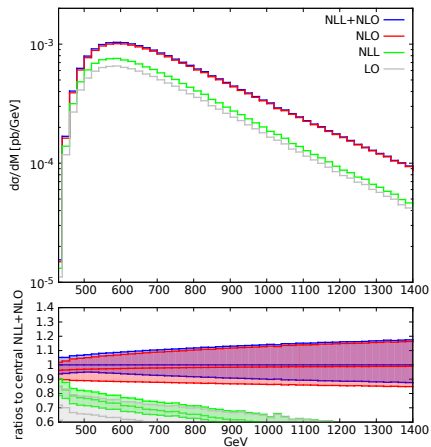
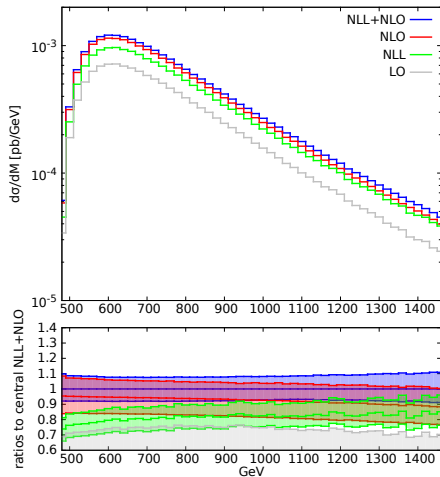
- ▶ Stop pair production @ LHC14.
- ▶  $M$  = Born system invariant mass.
- ▶ Lines = analytic codes, dots = reweighting interface.

## Preliminary results: $t\bar{t}$



►  $t\bar{t}$  production @ LHC7 (LO with NLO PDF), additive matching.

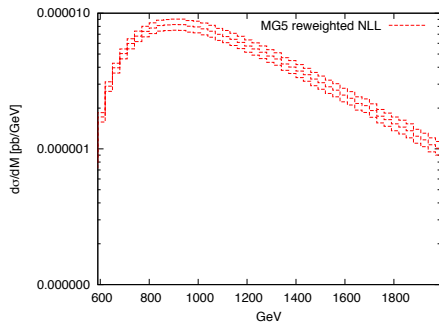
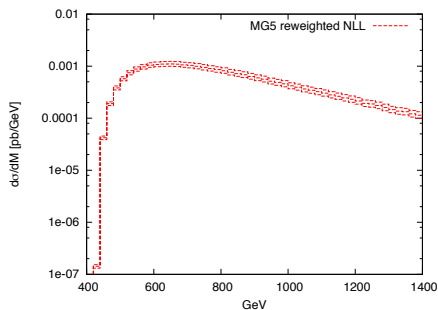
# Preliminary results: $t\bar{t}H$ and $t\bar{t}W$



- ▶  $t\bar{t}H$  (left) and  $t\bar{t}W^\pm$  (right) @ LHC13 (LO with NLO PDF), additive matching.



## Preliminary results: $t\bar{t}Z$ and $t\bar{t}W^+W^-$



- ▶  $t\bar{t}Z$  (left) and  $t\bar{t}W^+W^-$  (right) @ LHC13.
- ▶ Still to be matched to NLO.

## Status and future directions

General interface for threshold resummation in MG5\_aMC.

### Current limitations

- ▶ Only processes with no final-state massless QCD.
- ▶  $\mu_r = \mu_f$ .
- ▶ Only NLL for the moment.
- ▶ Only 2 coloured  $\rightarrow$  0,1,2 coloured +  $N$  colourless.

### Future directions

- ▶ Real goal is NNLL, through reweighting of NLO configurations.
- ▶ Overcome the above constraints.
- ▶ Consolidate and improve the interface.

Thank you for your attention

# Backup

## Integration of resummation kernel

- ▶  $\mathcal{P}_{\text{NLL}}(\mathbf{X}) \propto \int_{\tau}^1 dz \mathcal{L}_{ij}(\frac{\tau}{z}, \mu_f) z^{-1-\eta} (1-z)^{-1+2\eta} = \int_{-1}^1 du \int_{\tau}^1 dz \mathcal{I}(u, z) (1-z)^{-1+2\eta}$
- ▶ Bulk coming from  $z \sim 1$  for small  $\eta$ .
- ▶ Split  $z$ -integration in  $[\tau, z_0]$ , and  $[z_0, 1]$ :  $\int_{z_0}^1 = g(u, z_0) + h(u, z_0)$

$$g(u, z_0) = \int_{z_0}^1 dz [\mathcal{I}(u, z) - \sum_{n=0}^m \frac{1}{n!} (1-z)^n \partial_z^n \mathcal{I}(u, 1)] (1-z)^{-1+2\eta}$$

$$h(u, z_0) = \sum_{n=0}^m \frac{1}{n!} \partial_z^n \mathcal{I}(u, 1) \frac{(1-z_0)^{n+2\eta}}{(n+2\eta)}$$

- ▶  $\partial_z^n \mathcal{I}$  **analytic**: PDFs on a basis of Chebyshev polynomials with fitted coefficients.
- ▶ Subtractions: make  $g$  (numerical) small and  $h$  (analytic) large, to speed up.
- ▶ Subtraction **implements analytic continuation for negative  $\eta$** .
- ▶ Integrations by means of quadrature methods (need to be fast since one 2D integration per reweighted event).