

Targeting Multi-Loop Integrals with Neural Networks

Ramon Winterhalder

March 30, 2022

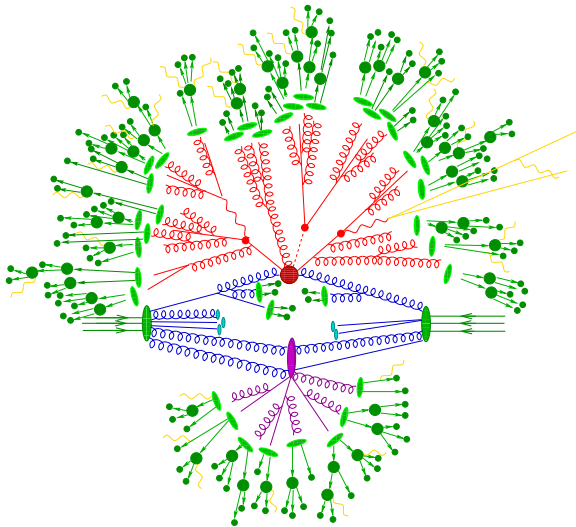
CP3, UC Louvain

Based on [\[2112.09145\]](#)

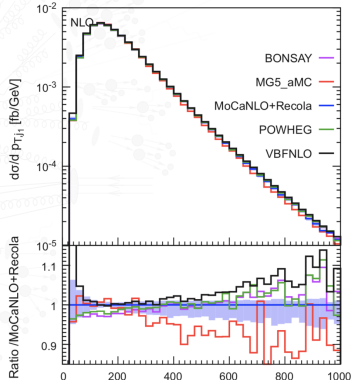
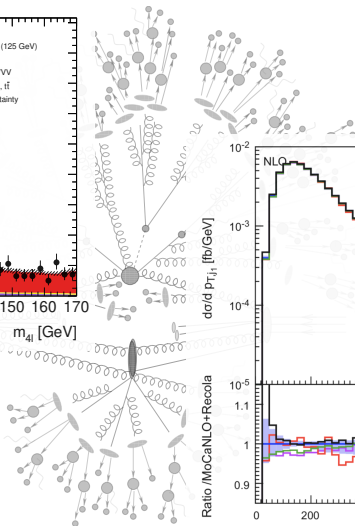
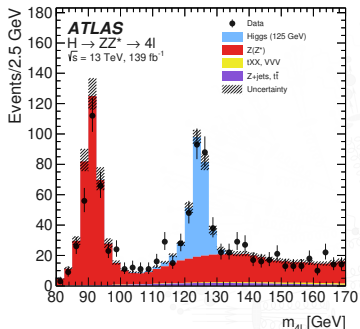
in collaboration with Vitaly Magerya, Emilio Villa, Stephen P. Jones, Matthias Kerner, Anja Butter, Gudrun Heinrich, and Tilman Plehn



Data analysis in HEP



Data analysis in HEP



Theory predictions in HEP

$$\mathcal{A}_{n,\{\lambda,c,\dots\}}(p_a, p_b | p_1, \dots, p_n) : \mathbb{M} \rightarrow \mathbb{C}$$

Quantum numbers:
spin, colour charge etc.

Kinematics:
Momenta in Minkowski space, masses, etc.

$$\sigma_n = \frac{1}{\text{flux}} \sum_{a,b} \int dx_a dx_b f(x_a) f(x_b) \int d\Phi_n \langle |\mathcal{A}_n(p_a, p_b | p_1, \dots, p_n)|^2 \rangle$$

Cross section:
more generally,
differential
observables*

PDFs:
convolution over
all possible initial
state configurations

**Phase-space
integral:**
over final state
kinematics

**Squared
amplitude:**
summed over final
states, averaged over
initial states

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Outline for today

→ Focus on **calculation of amplitudes**

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Towards high precision

Fixed-order perturbative expansion

$$\mathcal{A}_n = \mathcal{A}_n^{(0)} + \underbrace{\mathcal{A}_n^{(1)} + \mathcal{A}_n^{(2)} + \dots}_{\text{loop amplitudes}}$$

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→ Focus on **calculation of amplitudes**

Quantum numbers:
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Kinematics:

Calculation of loop amplitudes

For complicated integrals
analytic solution impractical
→ **numerical evaluation** in
Feynman parameter space

Towards high precision

Fixed-order perturbative expansion

$$\mathcal{A}_n = \mathcal{A}_n^{(0)} + \underbrace{\mathcal{A}_n^{(1)} + \mathcal{A}_n^{(2)} + \dots}_{\text{loop amplitudes}}$$

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Multi-loop integral

Feynman parametrization

$$G = \frac{(-1)^\nu \Gamma(\nu - LD/2)}{\prod_{j=1}^N \Gamma(\nu_j)} \int \left(\prod_{j=1}^N dx_j x_j^{\nu_j-1} \right) \delta\left(1 - \sum_{l=1}^N x_l\right) \frac{\mathcal{U}^{\nu-(L+1)D/2}}{\mathcal{F}^{\nu-LD/2}}$$

with $\mathcal{U} \equiv \det(M)$ and $\mathcal{F} \equiv \det(M) \left[\sum_{i,j=1}^L Q_i (M^{-1})_{ij} Q_j - J - i\delta \right]$

\mathcal{U} : polynomial in x_i only

\mathcal{F} : depends on x_i and kinematic invariants s_{ij}, m_i^2

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Caveats

Careful treatment of **singularities** in the integral!

Singularity structure

UV and IR singularities

show up as poles $1/\epsilon^\alpha$

(1a) Overall UV poles: $\Gamma(\nu - LD/2) \propto \Gamma(n\epsilon)$

(1b) UV subdivergencies: arise from $\mathcal{U}(\vec{x}) = 0$ for some $x_i = 0$.

(2) IR divergencies (soft and collinear):

arise from $\mathcal{F}(x, s_{ij}, m_i^2) = 0$ for some $x_i = 0$.

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Threshold-type singularities

- (3) $\mathcal{F}(x, s_{ij}, m_i^2) = 0$ **inside** integration region

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Sector decomposition

algorithm to **isolate and subtract** poles

→ public tool **pySECDEC** [Heinrich et al, '10,'17,'19, '21]

and **expand** integral

$$G = \sum_{j=-2L}^n C_j \epsilon^j$$

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numerical integration of finite C_j

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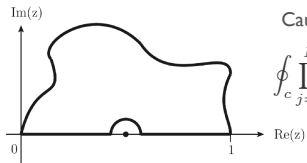
Parameter integrals

numerical integration of finite C_j
→ needs procedure to **avoid poles!**

Contour Deformation

Neural Contour Deformation

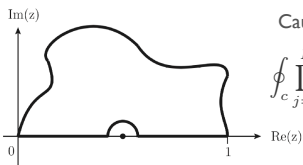
Standard contour deformation



Cauchy theorem:

$$\oint_c \prod_{j=1}^N dz_j \mathcal{I}(\vec{z}) = \int_0^1 \prod_{j=1}^N dx_j \mathcal{I}(\vec{x}) + \int_\gamma \prod_{j=1}^N dz_j \mathcal{I}(\vec{z})$$

Standard contour deformation



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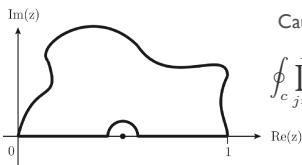
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Transformation: $x_j \rightarrow z_j(\vec{x}) = x_j - i\tau_j(\vec{x})$ Deformation parameter: λ_j

$$\tau_j = \lambda_j x_j (1 - x_j) \frac{\partial \mathcal{F}(\vec{x})}{\partial x_j}$$

$$\begin{aligned} \mathcal{F}(\vec{z}) = \mathcal{F}(\vec{x}) &- i \sum_j \lambda_j x_j (1 - x_j) \left(\frac{\partial \mathcal{F}(\vec{x})}{\partial x_j} \right)^2 - \frac{1}{2} \sum_{j,k} \tau_j \tau_k \frac{\partial^2 \mathcal{F}(\vec{x})}{\partial x_j \partial x_k} \\ &+ \frac{i}{6} \sum_{j,k,l} \tau_j \tau_k \tau_l \frac{\partial^3 \mathcal{F}(\vec{x})}{\partial x_j \partial x_k \partial x_l} + \dots \end{aligned}$$

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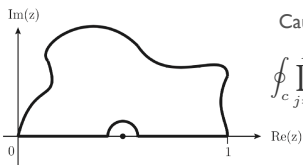
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always positive

Standard contour deformation



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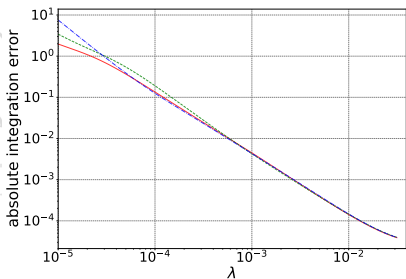
Does **not** spoil sign
for small enough λ

Standard contour deformation

$\text{Im}(z)$



Transformation



$$\int_{\gamma} \prod_{j=1}^N dz_j \mathcal{I}(\vec{z})$$

parameter: λ_j

$$\mathcal{F}(\vec{z}) = \mathcal{F}(\vec{x}) - i \sum_j \lambda_j x_j (1 - x_j) \left(\frac{\partial \mathcal{F}(\vec{x})}{\partial x_j} \right)^2 - \frac{1}{2} \sum_{j,k} \tau_j \tau_k \frac{\partial^2 \mathcal{F}(\vec{x})}{\partial x_j \partial x_k}$$

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always positive

Does **not** spoil sign
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Λ -Glob algorithm

Complex shift

$$x_j \rightarrow z_j(\vec{x}) = x_j - i\tau_j(\vec{x}),$$

$$\tau_j = \lambda_j x_j (1 - x_j) \frac{\partial F(\vec{x})}{\partial x_j}, \quad \rho(\vec{z}(\vec{x})) = \left| \frac{\partial \vec{z}(\vec{x})}{\partial \vec{x}} \right|^{-1}$$

$$I = \int_{\gamma} \prod_{j=1}^N dz_j(\vec{x}) \frac{\mathcal{I}(\vec{z})}{\rho(\vec{z})} = \frac{1}{N} \sum_{i=1}^N \frac{\mathcal{I}(\vec{x}_i)}{\rho(\vec{x}_i)} = \langle \mathcal{I} / \rho \rangle_x$$

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Loss function

$$L = \sigma_I^2 = \frac{\langle (\mathcal{I}/\rho)^2 \rangle_x - \langle \mathcal{I}/\rho \rangle_x^2}{N - 1}$$

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$$L = \sigma_I^2 = \frac{\langle (\mathcal{I}/\rho)^2 \rangle_x - \langle \mathcal{I}/\rho \rangle_x^2}{N-1} + L_{\text{sign}}$$

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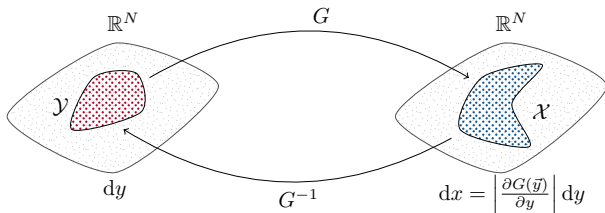
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Note: Making $\lambda_j \rightarrow \lambda_j(\vec{x})$ local has negligible effects! $\rightarrow \Lambda$ -Glob

Normalizing flow

Neural importance sampling Gao et al. [2001.05486, 2001.10028], Bothmann et al. [2001.05478]

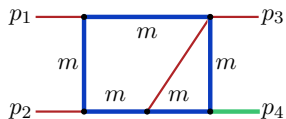
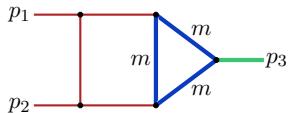


$$x_j \rightarrow x_j(\vec{y}) = G_j(\vec{y}), \quad g(\vec{x}(\vec{y})) = \left| \frac{\partial G(\vec{y})}{\partial \vec{y}} \right|^{-1}$$

$$I = \frac{1}{N} \sum_{i=1}^N \frac{\mathcal{I}(\vec{y}_i)}{g(\vec{y}_i) \rho(\vec{y}_i)} = \langle \mathcal{I} / (\rho g) \rangle_{\mathcal{Y}}$$

Results – Two-loop integrals

Example diagrams

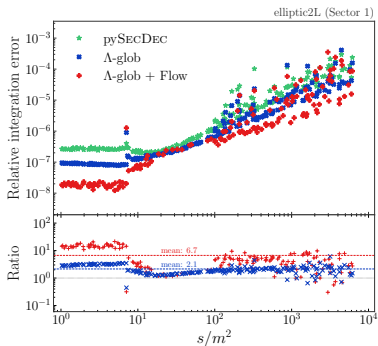
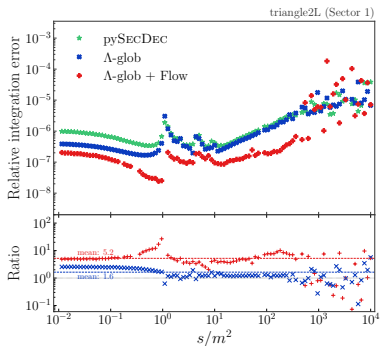
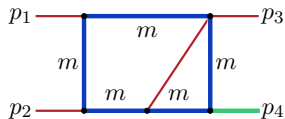
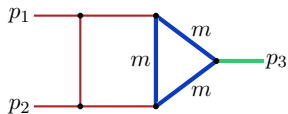


- NLO correction to $gg \rightarrow h$
- 5 feynman parameters
- 2 kinematic invariants:
 s, m^2

- $gg \rightarrow hj$ at two loops
- 5 feynman parameters
- 4 kinematic invariants:
 s, t, p_4^2, m^2

Results – Two-loop integrals

Example diagrams



Conclusion and Outlook

Summary

- **Improved** precision of numerical loop integrals!
- **Two-step** procedure:
 - (a) **Λ -Glob**: complex shift
 - (b) **Normalizing Flow**: importance sampling reals

Outlook

- Further investigation of **more complicated integrals**
- Possibly check integrals for which **standard pySECDEC fails**.