

Feynman Parameter Integrals with Positive Integrands

Thomas Stone

TUM-MPP Collider Phenomenology Seminar, **Munich**
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Stephen Jones, Anton Olsson, **TS**

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Alex Bennett, Bakar Chargeishvili, Stephen Jones, Anton Olsson, **TS** +
Gudrun Heinrich, Matthias Kerner, Vitaly Magerya, Johannes Schlenk [*WIP*]



Outline

- 1 Preliminaries
 - Feynman Parameter Integrals
 - Positive Integrands
- 2 Univariate Bisectable Integrals
 - Algorithm
 - Massless Examples
- 3 Beyond UB Integrals
 - Massive Examples
- 4 Outlook
 - Generic Cylindrical Algebraic Decomposition

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Various methods to approximate/evaluate them numerically

ODE/PDE

Series Solutions

[Slide style inspired by Stephen Jones]

~Monte Carlo
Integration

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Big challenges in modern amplitude calculations:

- *Reduction* to master integrals
- *Evaluation* of master integrals (our focus)

Feynman integrals can be difficult to compute analytically

Various methods to approximate/evaluate them numerically

Numerical differential equations

ODE/PDE

Series solutions of differential equations (AMFlow, DiffExp, Seasyde)

Series Solutions

[Slide style inspired by Stephen Jones]

Taylor expansion in Feynman parameters (TayInt)

Numerical Mellin-Barnes (MB, Ambre)

Tropical sampling (Feyntrop)

~Monte Carlo
Integration

Numerical Loop-Tree Duality (cLTD, Lotty)

Sector decomposition (Sector_decomposition, FIESTA, pySecDec)

↪ *This talk*: avoiding contour deformation in the context of **sector decomposition**

Minkowski Regime

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Minkowski regime $\Rightarrow$ poles on the contour of integration

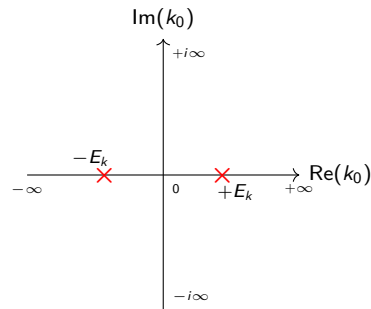


Figure: Tadpole k_0 integration

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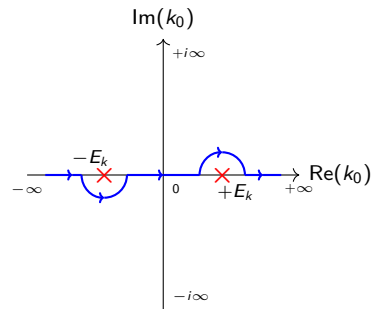


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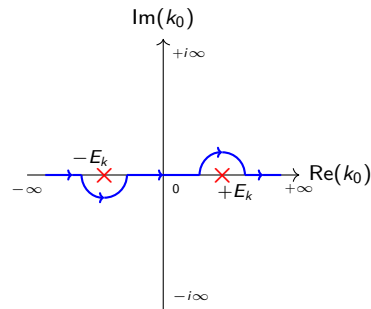


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Methods being explored to remove the need for contour deformation in momentum space

[Buchta, Chachamis, Draggiotis, Rodrigo; Anastasiou, Haindl, Sterman, Yang, Zeng; Aguilera-Verdugo, Hernandez-Pinto, Sborlini, Torres Bobadilla; Capatti, Hirschi, Kermanschah, Pelloni, Ruijl...]

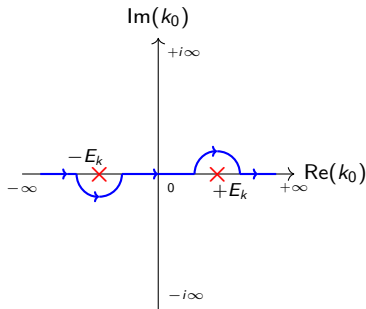


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Can we do the same in **Feynman parameter** space?

Feynman Parameter Integrals

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Loop integral in Feynman parameter space

$$J(\mathbf{s}) = \frac{(-1)^\nu \Gamma(\nu - LD/2)}{\prod_{i=1}^N \Gamma(\nu_i)} \lim_{\delta \rightarrow 0^+} \int_{\mathbb{R}_{\geq 0}^N} \prod_{i=1}^N dx_i x_i^{\nu_i-1} \frac{\mathcal{U}(\mathbf{x})^{\nu-(L+1)D/2}}{(\mathcal{F}(\mathbf{x}; \mathbf{s}) - i\delta)^{\nu-LD/2}} \delta(1 - \alpha(\mathbf{x}))$$

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$$\mathcal{U}(\mathbf{x}) = \sum_{T^1} \prod_{e \notin T^1} x_e \geq 0, \quad \text{independent of kinematics, degree of homogeneity in } \mathbf{x}: L$$

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$$\mathcal{F}(\mathbf{x}; \mathbf{s}) = \sum_{T^2} (-s_{T^2}) \prod_{e \notin T^2} x_e + \mathcal{U}(\mathbf{x}) \sum_e m_e^2 x_e$$

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We distinguish between 3 types of kinematic regime:

- $\forall \mathbf{s} \in \mathbf{s}_R : \forall \mathbf{x} \in \mathbb{R}_{>0}^N : \mathcal{F}(\mathbf{x}; \mathbf{s}) > 0 \quad \Rightarrow \quad \mathbf{s}_R \text{ is a } \textit{positive same-sign} \text{ regime}$

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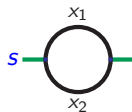
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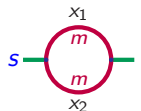
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Landau equations in Feynman parameter space

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↪ a) and b) satisfied $\longrightarrow$ further analysis required

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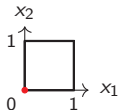
$$\text{e.g. } I = \int_0^1 dx x^{-1+a\epsilon} \mathcal{I}(x, \epsilon) = \frac{\mathcal{I}(0, \epsilon)}{a\epsilon} + \underbrace{\int_0^1 dx x^{-1+a\epsilon} [\mathcal{I}(x, \epsilon) - \mathcal{I}(0, \epsilon)]}_{\text{no problem as } x \rightarrow 0, \text{ finite!}}$$

Sector Decomposition

- “Overlapping” singularities $\rightarrow$ mapped to “factorised” type by **sector decomposition**

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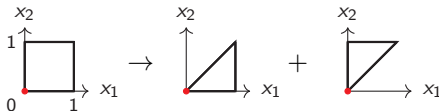
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e.g. $I = \int_{\square} dx_1 dx_2 \frac{1}{x_1^{1+a\epsilon} x_2^{b\epsilon} [x_1 + (1-x_1)x_2]}$

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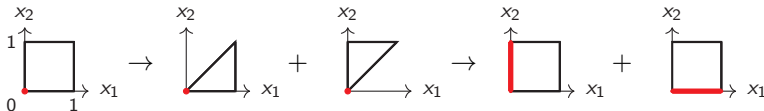


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Numerous algorithms to resolve “**overlapping**” singularities by performing **sector decomposition** (e.g. geometric approaches based on Newton polytopes, iterative approaches)

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TAKE HOME MESSAGE

Sector decomposition resolves issues on the *boundary* and can be considered ‘**solved**’*

*for our purposes at least!

Contour Deformation in Parameter Space

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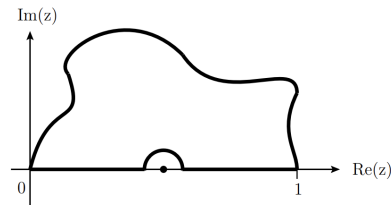
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Explicit contour deformation in parameter space

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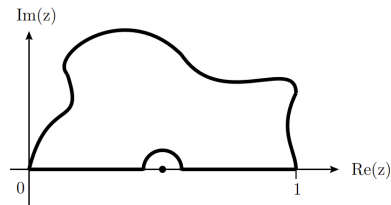
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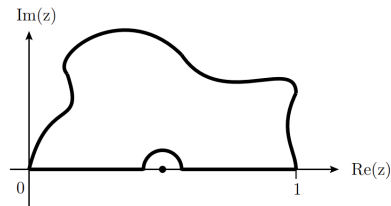
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Try to optimise the deformation e.g. train $\lambda_j \rightarrow \lambda_j(\mathbf{x})$ with a [neural network](#)

[Winterhalder, Magerya, Villa, Jones, Kerner, Butter, Heinrich, Plehn; ...]

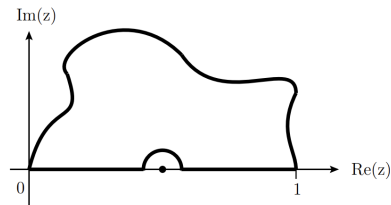
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If this works, why try to **avoid** it?

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Contour deformation reduces precision, can be **very** slow and can break down entirely!

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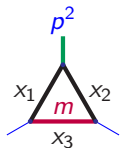
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Trivial analytic continuation in this representation: $\text{Im}(J)$ controlled by $(-1 - i\delta^+)^{LD/2-\nu} \times \mathbb{R}$

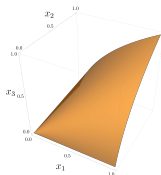
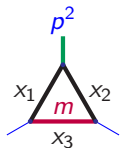
Constructing the Decomposition

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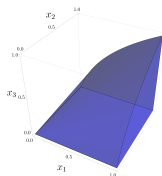
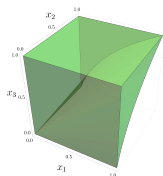
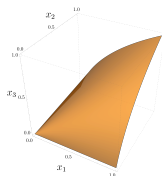
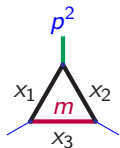
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Analyse the (projective) *variety of $\mathcal{F}$* : **codimension-1 hypersurface** defined by $\mathcal{F}(\mathbf{x}; \mathbf{s}) = 0$

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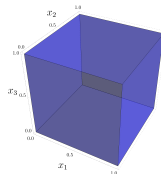
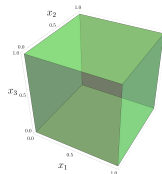
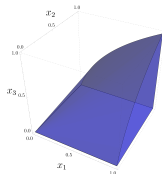
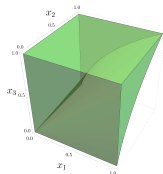
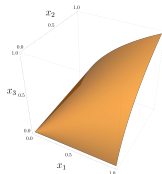
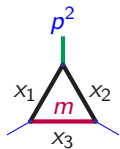


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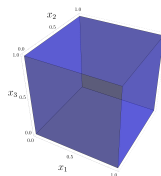
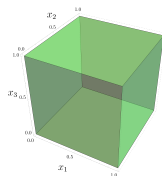
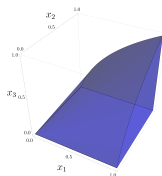
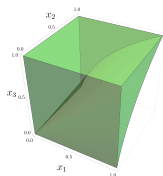
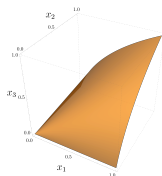
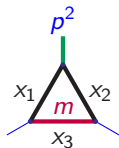
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Variety ($\mathcal{F} = 0$) mapped to *boundary* $\Rightarrow$ resolved by **sector decomposition**!

The Simplest Example

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$$\begin{array}{c} x_1 \\ \textcolor{red}{m} \\ \textcolor{blue}{s} \text{---} \bigcirc \text{---} \\ x_2 \end{array} = \Gamma(\epsilon) \int_{\mathbb{R}_{\geq 0}^2} dx_1 dx_2 \frac{(x_1 + x_2)^{-2+2\epsilon}}{(-\textcolor{blue}{s} x_1 x_2 + \textcolor{red}{m}^2 x_1 (x_1 + x_2) - i\delta^+)^{\epsilon}} \delta(1 - \alpha(\mathbf{x})), \quad \textcolor{blue}{s} > \textcolor{red}{m}^2 > 0$$

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$$\begin{aligned}
 \text{Diagram: a circle with a green line segment on the left labeled } s, \text{ a red arc on top labeled } m, \text{ and } x_1 \text{ above and } x_2 \text{ below.} &= \Gamma(\epsilon) \int_{\mathbb{R}_{\geq 0}^2} dx_1 dx_2 \frac{(x_1 + x_2)^{-2+2\epsilon}}{(-s x_1 x_2 + m^2 x_1 (x_1 + x_2) - i\delta^+)^{\epsilon}} \delta(1 - \alpha(\mathbf{x})), \quad s > m^2 > 0 \\
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The Simplest Example

$$\begin{aligned}
 \text{Diagram: } & \text{A circle with } x_1 \text{ at the top and } x_2 \text{ at the bottom. A green line segment labeled } s \text{ enters from the left, and a red arc labeled } m \text{ is on the top-left.} \\
 & = \Gamma(\epsilon) \int_{\mathbb{R}_{\geq 0}^2} dx_1 dx_2 \frac{(x_1 + x_2)^{-2+2\epsilon}}{(-s x_1 x_2 + m^2 x_1 (x_1 + x_2) - i\delta^+)^{\epsilon}} \delta(1 - \alpha(\mathbf{x})), \quad s > m^2 > 0 \\
 & = (-1 - i\delta^+)^{-\epsilon} \Gamma(\epsilon) (m^2)^{1-\epsilon} (s - m^2)^{1-2\epsilon} \int_{\mathbb{R}_{\geq 0}^2} dx_1 dx_2 \frac{x_1^{-\epsilon} x_2^{-\epsilon}}{(s x_1 + m^2 x_2)^{2-2\epsilon}} \delta(1 - \alpha^-(\mathbf{x})) \\
 & + \Gamma(\epsilon) (m^2)^{2-2\epsilon} \int_{\mathbb{R}_{\geq 0}^2} dx_1 dx_2 \frac{x_1^{-\epsilon} (m^2 x_1 + (s - m^2) x_2)^{-\epsilon}}{(s x_2 + m^2 x_1)^{2-2\epsilon}} \delta(1 - \alpha^+(\mathbf{x}))
 \end{aligned}$$

Univariate Bisectable Integrals

Is it always so easy to decompose the integration domain?

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Univariate bisectable integrals:

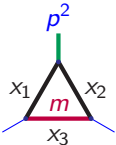
$\mathcal{F} = 0$ hypersurface **bisects** domain w.r.t. UB parameter

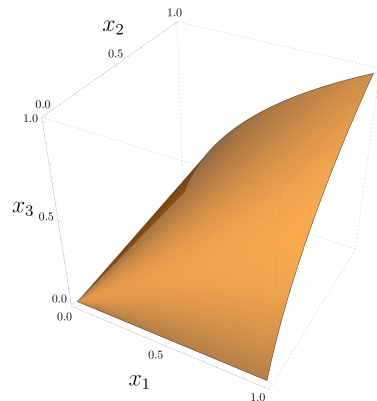
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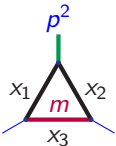


Univariate Bisectable Integrals

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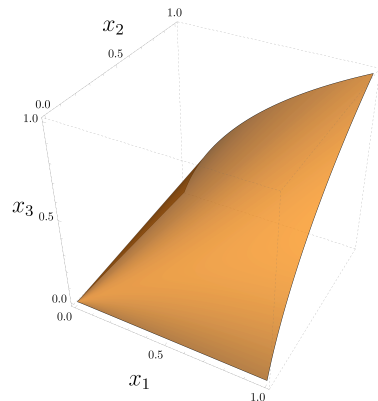
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Same-sign decomposition for UB integrals

$$J(\mathbf{s}) = J^+(\mathbf{s}) + (-1 - i\delta^+)^{LD/2-\nu} J^-(\mathbf{s})$$



UB Algorithm

Algorithm: Univariate Bisection (UB)

Input: $\mathcal{I}(\mathbf{x}; \mathbf{s}; \delta)$, $\mathbf{s}_R$

Output: $\mathcal{I}^+(\mathbf{x}; \mathbf{s})$, $\mathcal{I}^-(\mathbf{x}; \mathbf{s})$

foreach $x_i \in \mathbf{x}$ **do**

Let $r = \text{Reduce}[\{\mathcal{F}(\mathbf{x}; \mathbf{s}) < 0\} \cup \{0 < \mathbf{x}\} \cup \mathbf{s}_R, x_i]$;

if $r \sim \mathbf{A}$ **then**

Let $\mathcal{I}^-(\mathbf{x}; \mathbf{s}) = \mathcal{J}(\mathbf{x}_{\neq i}, y_i) \mathcal{I}(\mathbf{x}_{\neq i}, y_i; -\mathbf{s}; 0)$

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end

return $\neg \text{UB in } \mathbf{s}_R$

UB integrals $\rightarrow$ algorithmic procedure

$\mathbf{A} : \{0 < x_i < f(\mathbf{x}_{\neq i})\} \wedge \{0 < \mathbf{x}_{\neq i}\} \wedge \mathbf{s}_R$

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Essentially just *parameter transformations*:

$$x_i \rightarrow y_i = \frac{x_i}{x_i + x_j} f(\mathbf{x}_{\neq i}) \quad (1)$$

$$x_i \rightarrow y'_i = x_i + f(\mathbf{x}_{\neq i}) \quad (2)$$

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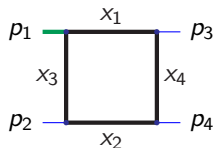
$$x_i \rightarrow y'_i = x_i + f(\mathbf{x}_{\neq i}) \quad (2)$$

(1) $\mathcal{F} = 0$ hypersurface $\rightarrow \infty$

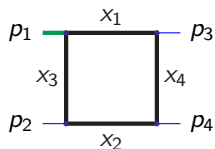
(2) $\mathcal{F} = 0$ hypersurface $\rightarrow 0$

1-Loop Box with One Off-Shell Leg

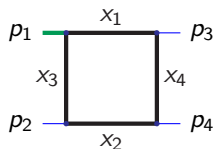
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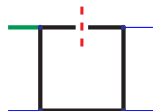
1-Loop Box with One Off-Shell Leg


 $\mathcal{U}(\mathbf{x})$

1-Loop Box with One Off-Shell Leg

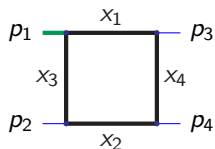


$$\mathcal{U}(\mathbf{x}) =$$



$$= x_1$$

1-Loop Box with One Off-Shell Leg

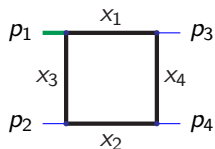


$$\mathcal{U}(\mathbf{x}) =$$

The first diagram shows the box with a vertical red dashed line on the top edge, representing a cut. The second diagram shows the box with a vertical red dashed line on the bottom edge, representing another cut.

$$= x_1 + x_2$$

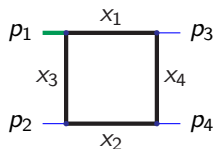
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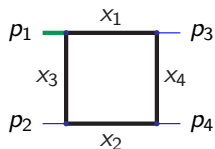
$$= x_1 + x_2 + x_3$$

1-Loop Box with One Off-Shell Leg



$$\begin{aligned}
 \mathcal{U}(\mathbf{x}) = & \text{[Box with top edge cut]} + \text{[Box with bottom edge cut]} + \text{[Box with left edge cut]} + \text{[Box with right edge cut]} \\
 = & x_1 + x_2 + x_3 + x_4
 \end{aligned}$$

1-Loop Box with One Off-Shell Leg



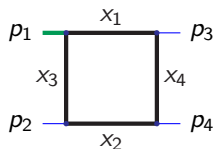
$$\mathcal{U}(\mathbf{x}) = \text{[Diagram 1]} + \text{[Diagram 2]} + \text{[Diagram 3]} + \text{[Diagram 4]}$$

$$= x_1 + x_2 + x_3 + x_4$$

The four diagrams represent the four edges of the box with the off-shell condition (red dashed line) applied to each edge in turn: top, bottom, left, and right.

$$\mathcal{F}(\mathbf{x}; \mathbf{s})$$

1-Loop Box with One Off-Shell Leg



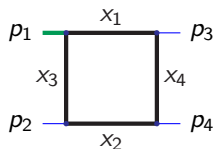
$$\mathcal{U}(\mathbf{x}) =$$

$$= x_1 + x_2 + x_3 + x_4$$

$$\mathcal{F}(\mathbf{x}; \mathbf{s}) = -s$$

$$= -s x_1 x_2$$

1-Loop Box with One Off-Shell Leg



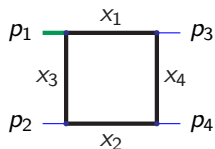
$$\mathcal{U}(\mathbf{x}) =$$

$$= x_1 + x_2 + x_3 + x_4$$

$$\mathcal{F}(\mathbf{x}; \mathbf{s}) = -s$$

$$= -s x_1 x_2 - t x_3 x_4$$

1-Loop Box with One Off-Shell Leg



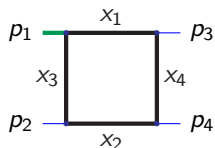
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$$= -s x_1 x_2 - t x_3 x_4 - p_1^2 x_1 x_3$$

1-Loop Box with One Off-Shell Leg



$$\mathcal{U}(\mathbf{x}) = \text{[Diagram 1]} + \text{[Diagram 2]} + \text{[Diagram 3]} + \text{[Diagram 4]}$$

$$= x_1 + x_2 + x_3 + x_4$$

Minkowski $\mathbf{s}_R$:

$$0 < p_1^2 < s$$

$$-s < t < 0$$

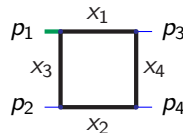
$$\mathcal{F}(\mathbf{x}; \mathbf{s}) = -s \text{[Diagram 1]} - t \text{[Diagram 2]} - p_1^2 \text{[Diagram 3]}$$

$$= -s x_1 x_2 - t x_3 x_4 - p_1^2 x_1 x_3$$

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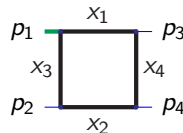
1-Loop Box with One Off-Shell Leg

$$J(\mathbf{s}) = \Gamma(2+\epsilon) \int_{\mathbb{R}_{\geq 0}^4} \prod_{i=1}^4 dx_i \frac{(x_1 + x_2 + x_3 + x_4)^{2\epsilon}}{(-s x_1 x_2 - t x_3 x_4 - p_1^2 x_1 x_3 - i\delta^+)^{2+\epsilon}} \delta(1 - \alpha(\mathbf{x}))$$



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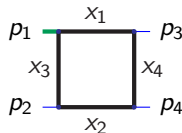
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Algorithm $\rightarrow$ UB in $\mathbf{s}_R = \{0 < p_1^2 < s, -s < t < 0\}$ w.r.t x_1

1-Loop Box with One Off-Shell Leg

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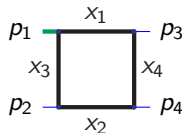


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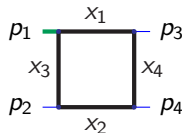
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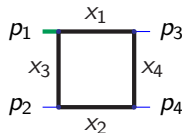
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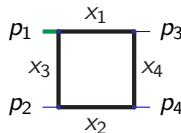
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Algorithm $\rightarrow$ UB in $\mathbf{s}_R = \{0 < p_1^2 < s, -s < t < 0\}$ w.r.t x_1

$$x_1 \rightarrow y_1' = x_1 + \left(\frac{-t x_3 x_4}{s x_2 + p_1^2 x_3} \right)$$

$$\mathcal{I}^- = \frac{[(s x_2 + p_1^2 x_3)(x_1 + x_2 + x_3 + x_4) - t x_3 x_4]^{2\epsilon}}{x_1^{2+\epsilon} (s x_2 + p_1^2 x_3)^{2+3\epsilon}}$$

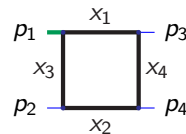
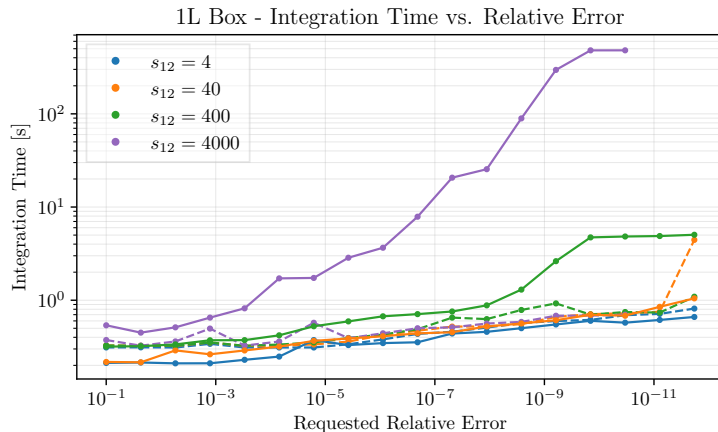
$$x_1 \rightarrow y_1 = \frac{x_1}{x_1 + x_4} \left(\frac{-t x_3 x_4}{s x_2 + p_1^2 x_3} \right)$$

$$\mathcal{I}^+ = \frac{[(s x_2 + p_1^2 x_3)(x_1 + x_4)(x_2 + x_3 + x_4) - t x_1 x_3 x_4]^{2\epsilon}}{x_4^{2+2\epsilon} (x_1 + x_4)^\epsilon (-t x_3)^{1+\epsilon} (s x_2 + p_1^2 x_3)^{1+2\epsilon}}$$

$$J(\mathbf{s}) = J^+(\mathbf{s}) + (-1 - i\delta^+)^{-2-\epsilon} J^-(\mathbf{s})$$

1-Loop Box with One Off-Shell Leg

1-Loop Box with One Off-Shell Leg



pySecDec

[Heinrich, Jones, Kerner, Magerya, Olsson, Schlenk]

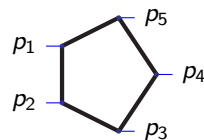
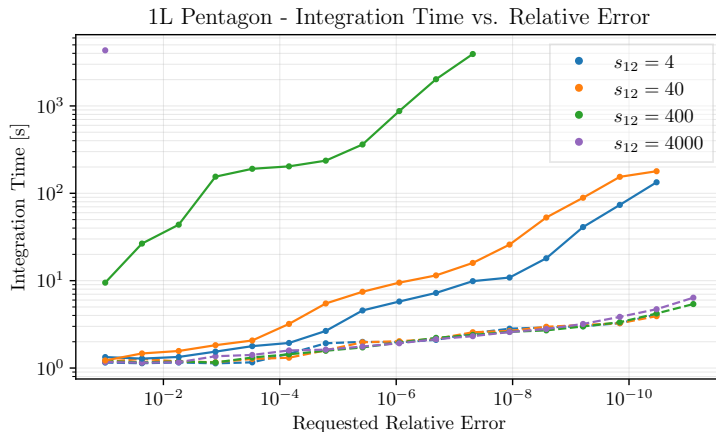
— CD
 - - - NOCD

$s = s_{12}$

$t = -1$

$p_1^2 = 2$

1-Loop Pentagon



pySecDec

— CD

- - - NOCD

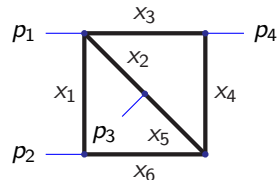
$$s_{23} = -3.0, \quad s_{34} = 2.5$$

$$s_{45} = -3.0, \quad s_{51} = 5.0$$

2-Loop Non-Planar 6 Propagator Box (BNP6)

2-Loop Non-Planar 6 Propagator Box (BNP6)

$$J(\mathbf{s}) = \Gamma(2+2\epsilon) \int_{\mathbb{R}_{\geq 0}^6} \prod_{i=1}^6 dx_i \frac{\mathcal{U}(\mathbf{x})^{3\epsilon}}{(\mathcal{F}(\mathbf{x}; \mathbf{s}) - i\delta^+)^{2+2\epsilon}} \delta(1 - \alpha(\mathbf{x}))$$

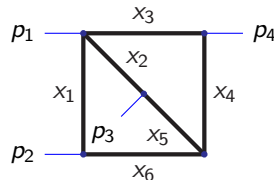


2-Loop Non-Planar 6 Propagator Box (BNP6)

$$J(\mathbf{s}) = \Gamma(2+2\epsilon) \int_{\mathbb{R}_{\geq 0}^6} \prod_{i=1}^6 dx_i \frac{\mathcal{U}(\mathbf{x})^{3\epsilon}}{(\mathcal{F}(\mathbf{x}; \mathbf{s}) - i\delta^+)^{2+2\epsilon}} \delta(1 - \alpha(\mathbf{x}))$$

$$\mathcal{U}(\mathbf{x}) = x_1 x_2 + x_1 x_3 + x_1 x_4 + x_1 x_5 + x_2 x_3 + x_2 x_4 + x_2 x_6 + \\ x_3 x_5 + x_3 x_6 + x_4 x_5 + x_4 x_6 + x_5 x_6$$

$$\mathcal{F}(\mathbf{x}; \mathbf{s}) = -s_{12} x_2 x_3 x_6 - s_{23} x_1 x_2 x_4 + (s_{12} + s_{23}) x_1 x_3 x_5$$

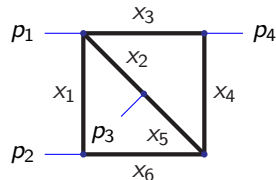


2-Loop Non-Planar 6 Propagator Box (BNP6)

$$J(\mathbf{s}) = \Gamma(2+2\epsilon) \int_{\mathbb{R}_{\geq 0}^6} \prod_{i=1}^6 dx_i \frac{\mathcal{U}(\mathbf{x})^{3\epsilon}}{(\mathcal{F}(\mathbf{x}; \mathbf{s}) - i\delta^+)^{2+2\epsilon}} \delta(1 - \alpha(\mathbf{x}))$$

$$\mathcal{U}(\mathbf{x}) = x_1 x_2 + x_1 x_3 + x_1 x_4 + x_1 x_5 + x_2 x_3 + x_2 x_4 + x_2 x_6 + \\ x_3 x_5 + x_3 x_6 + x_4 x_5 + x_4 x_6 + x_5 x_6$$

$$\mathcal{F}(\mathbf{x}; \mathbf{s}) = -s_{12} x_2 x_3 x_6 - s_{23} x_1 x_2 x_4 + (s_{12} + s_{23}) x_1 x_3 x_5 \quad \nexists \text{ same-sign regime!}$$



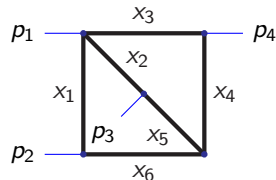
2-Loop Non-Planar 6 Propagator Box (BNP6)

$$J(\mathbf{s}) = \Gamma(2+2\epsilon) \int_{\mathbb{R}_{\geq 0}^6} \prod_{i=1}^6 dx_i \frac{\mathcal{U}(\mathbf{x})^{3\epsilon}}{(\mathcal{F}(\mathbf{x}; \mathbf{s}) - i\delta^+)^{2+2\epsilon}} \delta(1 - \alpha(\mathbf{x}))$$

$$\mathcal{U}(\mathbf{x}) = x_1 x_2 + x_1 x_3 + x_1 x_4 + x_1 x_5 + x_2 x_3 + x_2 x_4 + x_2 x_6 + \\ x_3 x_5 + x_3 x_6 + x_4 x_5 + x_4 x_6 + x_5 x_6$$

$$\mathcal{F}(\mathbf{x}; \mathbf{s}) = -s_{12} x_2 x_3 x_6 - s_{23} x_1 x_2 x_4 + (s_{12} + s_{23}) x_1 x_3 x_5 \quad \nexists \text{ same-sign regime!}$$

Algorithm $\rightarrow$ UB in $\mathbf{s}_R = \{0 < s_{12}, -s_{12} < s_{23} < 0\}$ w.r.t x_1



2-Loop Non-Planar 6 Propagator Box (BNP6)

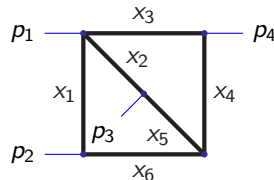
$$J(\mathbf{s}) = \Gamma(2+2\epsilon) \int_{\mathbb{R}_{\geq 0}^6} \prod_{i=1}^6 dx_i \frac{\mathcal{U}(\mathbf{x})^{3\epsilon}}{(\mathcal{F}(\mathbf{x}; \mathbf{s}) - i\delta^+)^{2+2\epsilon}} \delta(1 - \alpha(\mathbf{x}))$$

$$\mathcal{U}(\mathbf{x}) = x_1 x_2 + x_1 x_3 + x_1 x_4 + x_1 x_5 + x_2 x_3 + x_2 x_4 + x_2 x_6 + \\ x_3 x_5 + x_3 x_6 + x_4 x_5 + x_4 x_6 + x_5 x_6$$

$$\mathcal{F}(\mathbf{x}; \mathbf{s}) = -s_{12} x_2 x_3 x_6 - s_{23} x_1 x_2 x_4 + (s_{12} + s_{23}) x_1 x_3 x_5 \quad \nexists \text{ same-sign regime!}$$

Algorithm $\rightarrow$ UB in $\mathbf{s}_R = \{0 < s_{12}, -s_{12} < s_{23} < 0\}$ w.r.t x_1

$$f(\mathbf{x}_{\neq 1}) = \frac{s_{12} x_2 x_3 x_6}{(s_{12} + s_{23}) x_3 x_5 - s_{23} x_2 x_4}$$



2-Loop Non-Planar 6 Propagator Box (BNP6)

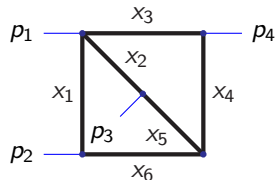
$$J(\mathbf{s}) = \Gamma(2+2\epsilon) \int_{\mathbb{R}_{\geq 0}^6} \prod_{i=1}^6 dx_i \frac{\mathcal{U}(\mathbf{x})^{3\epsilon}}{(\mathcal{F}(\mathbf{x}; \mathbf{s}) - i\delta^+)^{2+2\epsilon}} \delta(1 - \alpha(\mathbf{x}))$$

$$\mathcal{U}(\mathbf{x}) = x_1 x_2 + x_1 x_3 + x_1 x_4 + x_1 x_5 + x_2 x_3 + x_2 x_4 + x_2 x_6 + \\ x_3 x_5 + x_3 x_6 + x_4 x_5 + x_4 x_6 + x_5 x_6$$

$$\mathcal{F}(\mathbf{x}; \mathbf{s}) = -s_{12} x_2 x_3 x_6 - s_{23} x_1 x_2 x_4 + (s_{12} + s_{23}) x_1 x_3 x_5 \quad \nexists \text{ same-sign regime!}$$

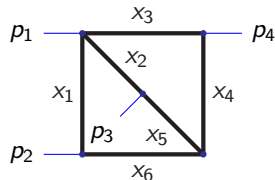
Algorithm $\rightarrow$ UB in $\mathbf{s}_R = \{0 < s_{12}, -s_{12} < s_{23} < 0\}$ w.r.t x_1

$$f(\mathbf{x}_{\neq 1}) = \frac{s_{12} x_2 x_3 x_6}{(s_{12} + s_{23}) x_3 x_5 - s_{23} x_2 x_4} \quad x_1 \rightarrow y_1 = \frac{x_1}{x_1 + x_6} f(\mathbf{x}_{\neq 1})$$



2-Loop Non-Planar 6 Propagator Box (BNP6)

$$J(\mathbf{s}) = \Gamma(2+2\epsilon) \int_{\mathbb{R}_{\geq 0}^6} \prod_{i=1}^6 dx_i \frac{\mathcal{U}(\mathbf{x})^{3\epsilon}}{(\mathcal{F}(\mathbf{x}; \mathbf{s}) - i\delta^+)^{2+2\epsilon}} \delta(1 - \alpha(\mathbf{x}))$$



$$\mathcal{U}(\mathbf{x}) = x_1 x_2 + x_1 x_3 + x_1 x_4 + x_1 x_5 + x_2 x_3 + x_2 x_4 + x_2 x_6 + x_3 x_5 + x_3 x_6 + x_4 x_5 + x_4 x_6 + x_5 x_6$$

$$\mathcal{F}(\mathbf{x}; \mathbf{s}) = -s_{12} x_2 x_3 x_6 - s_{23} x_1 x_2 x_4 + (s_{12} + s_{23}) x_1 x_3 x_5 \quad \nexists \text{ same-sign regime!}$$

Algorithm $\rightarrow$ UB in $\mathbf{s}_R = \{0 < s_{12}, -s_{12} < s_{23} < 0\}$ w.r.t x_1

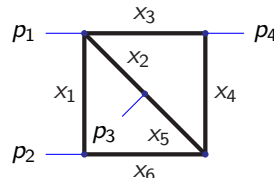
$$f(\mathbf{x}_{\neq 1}) = \frac{s_{12} x_2 x_3 x_6}{(s_{12} + s_{23}) x_3 x_5 - s_{23} x_2 x_4}$$

$$x_1 \rightarrow y_1 = \frac{x_1}{x_1 + x_6} f(\mathbf{x}_{\neq 1})$$

$$x_1 \rightarrow y'_1 = x_1 + f(\mathbf{x}_{\neq 1})$$

2-Loop Non-Planar 6 Propagator Box (BNP6)

$$J(\mathbf{s}) = \Gamma(2+2\epsilon) \int_{\mathbb{R}_{\geq 0}^6} \prod_{i=1}^6 dx_i \frac{\mathcal{U}(\mathbf{x})^{3\epsilon}}{(\mathcal{F}(\mathbf{x}; \mathbf{s}) - i\delta^+)^{2+2\epsilon}} \delta(1 - \alpha(\mathbf{x}))$$



$$\mathcal{U}(\mathbf{x}) = x_1 x_2 + x_1 x_3 + x_1 x_4 + x_1 x_5 + x_2 x_3 + x_2 x_4 + x_2 x_6 + x_3 x_5 + x_3 x_6 + x_4 x_5 + x_4 x_6 + x_5 x_6$$

$$\mathcal{F}(\mathbf{x}; \mathbf{s}) = -s_{12} x_2 x_3 x_6 - s_{23} x_1 x_2 x_4 + (s_{12} + s_{23}) x_1 x_3 x_5 \quad \nexists \text{ same-sign regime!}$$

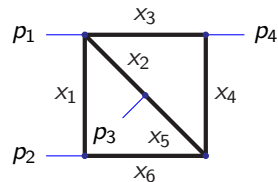
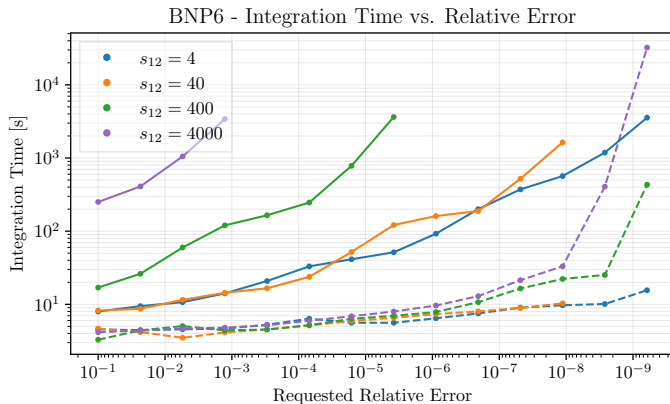
Algorithm $\rightarrow$ UB in $\mathbf{s}_R = \{0 < s_{12}, -s_{12} < s_{23} < 0\}$ w.r.t x_1

$$f(\mathbf{x}_{\neq 1}) = \frac{s_{12} x_2 x_3 x_6}{(s_{12} + s_{23}) x_3 x_5 - s_{23} x_2 x_4} \quad x_1 \rightarrow y_1 = \frac{x_1}{x_1 + x_6} f(\mathbf{x}_{\neq 1}) \quad x_1 \rightarrow y'_1 = x_1 + f(\mathbf{x}_{\neq 1})$$

$$J(\mathbf{s}) = J^+(\mathbf{s}) + (-1 - i\delta^+)^{-2-2\epsilon} J^-(\mathbf{s})$$

2-Loop Non-Planar 6 Propagator Box (BNP6)

2-Loop Non-Planar 6 Propagator Box (BNP6)

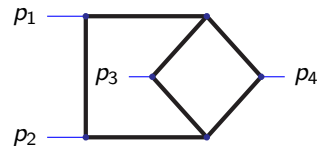
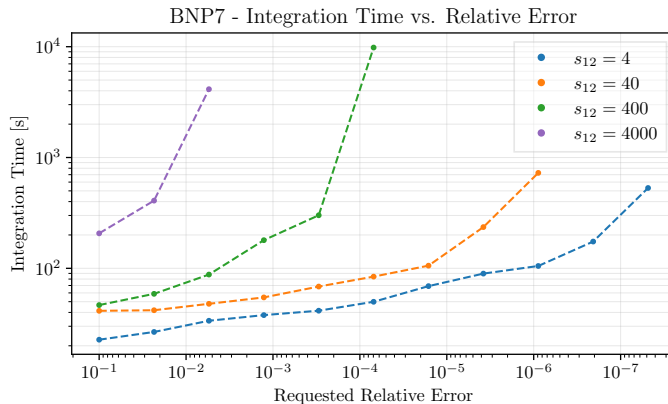


pySecDec

— CD
 - - - NOCD

$$s_{23} = -1$$

2-Loop Non-Planar 7 Propagator Box (BNP7)



pySecDec

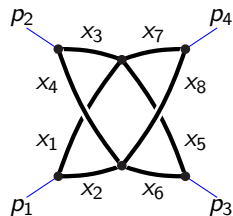
— CD
 - - - NOCD

$$s_{23} = -1$$

3-Loop Non-Planar Box (Crown)

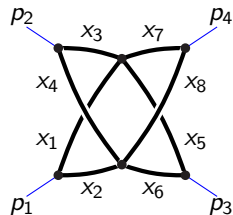
3-Loop Non-Planar Box (Crown)

$$J(\mathbf{s}) = \Gamma(2+3\epsilon) \int_{\mathbb{R}_{\geq 0}^8} \prod_{i=1}^8 dx_i \frac{\mathcal{U}(\mathbf{x})^{4\epsilon}}{(\mathcal{F}(\mathbf{x}; \mathbf{s}) - i\delta^+)^{2+3\epsilon}} \delta(1 - \alpha(\mathbf{x}))$$



3-Loop Non-Planar Box (Crown)

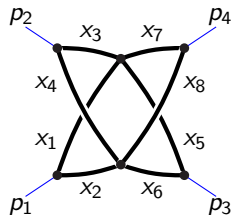
$$J(\mathbf{s}) = \Gamma(2+3\epsilon) \int_{\mathbb{R}_{\geq 0}^8} \prod_{i=1}^8 dx_i \frac{\mathcal{U}(\mathbf{x})^{4\epsilon}}{(\mathcal{F}(\mathbf{x}; \mathbf{s}) - i\delta^+)^{2+3\epsilon}} \delta(1 - \alpha(\mathbf{x}))$$



$$\mathcal{F}(\mathbf{x}; \mathbf{s}) = -s_{12} (x_2 x_5 - x_1 x_6) (x_4 x_7 - x_3 x_8) - s_{13} (x_2 x_3 - x_1 x_4) (x_6 x_7 - x_5 x_8)$$

3-Loop Non-Planar Box (Crown)

$$J(\mathbf{s}) = \Gamma(2+3\epsilon) \int_{\mathbb{R}_{\geq 0}^8} \prod_{i=1}^8 dx_i \frac{\mathcal{U}(\mathbf{x})^{4\epsilon}}{(\mathcal{F}(\mathbf{x}; \mathbf{s}) - i\delta^+)^{2+3\epsilon}} \delta(1 - \alpha(\mathbf{x}))$$

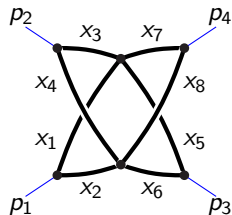


$$\mathcal{F}(\mathbf{x}; \mathbf{s}) = -s_{12} (x_2 x_5 - x_1 x_6) (x_4 x_7 - x_3 x_8) - s_{13} (x_2 x_3 - x_1 x_4) (x_6 x_7 - x_5 x_8)$$

Leading Landau singularity for **arbitrary kinematics** $\rightarrow$ CD breaks down!

3-Loop Non-Planar Box (Crown)

$$J(\mathbf{s}) = \Gamma(2+3\epsilon) \int_{\mathbb{R}_{\geq 0}^8} \prod_{i=1}^8 dx_i \frac{\mathcal{U}(\mathbf{x})^{4\epsilon}}{(\mathcal{F}(\mathbf{x}; \mathbf{s}) - i\delta^+)^{2+3\epsilon}} \delta(1 - \alpha(\mathbf{x}))$$



$$\mathcal{F}(\mathbf{x}; \mathbf{s}) = -s_{12} (x_2 x_5 - x_1 x_6) (x_4 x_7 - x_3 x_8) - s_{13} (x_2 x_3 - x_1 x_4) (x_6 x_7 - x_5 x_8)$$

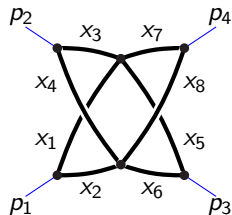
Leading Landau singularity for **arbitrary kinematics** $\rightarrow$ CD breaks down!

Dissect integration domain on *Landau locus*

[Gardi, Herzog, Jones, Ma]

3-Loop Non-Planar Box (Crown)

$$J(\mathbf{s}) = \Gamma(2+3\epsilon) \int_{\mathbb{R}_{\geq 0}^8} \prod_{i=1}^8 dx_i \frac{\mathcal{U}(\mathbf{x})^{4\epsilon}}{(\mathcal{F}(\mathbf{x}; \mathbf{s}) - i\delta^+)^{2+3\epsilon}} \delta(1 - \alpha(\mathbf{x}))$$



$$\mathcal{F}(\mathbf{x}; \mathbf{s}) = -s_{12} (x_2 x_5 - x_1 x_6) (x_4 x_7 - x_3 x_8) - s_{13} (x_2 x_3 - x_1 x_4) (x_6 x_7 - x_5 x_8)$$

Leading Landau singularity for **arbitrary kinematics** → CD breaks down!

Dissect integration domain on *Landau locus*

[Gardi, Herzog, Jones, Ma]

$$J(\mathbf{s}) = 4 \Gamma(2+3\epsilon) \sum_K \int_{\mathbb{R}_{\geq 0}^8} \prod_{i=1}^8 dx_i \frac{\mathcal{U}_K(\mathbf{x})^{4\epsilon}}{(x_2 x_4 x_6)^{1+3\epsilon} x_8^{1+9\epsilon} (\mathcal{F}_K(\mathbf{x}; \mathbf{s}) - i\delta^+)^{2+3\epsilon}} \delta(1 - \alpha(\mathbf{x}))$$

3-Loop Non-Planar Box (Crown)

3-Loop Non-Planar Box (Crown)

$$\mathbf{s}_R = \{0 < s_{12}, -s_{12} < s_{13} < 0\}$$

$$\mathcal{F}_A(\mathbf{x}; \mathbf{s}) = -[s_{12}x_3(x_1 + x_3 + x_5) + (s_{12} + s_{13})x_1x_5]$$

$$\mathcal{F}_B(\mathbf{x}; \mathbf{s}) = -[(s_{12} + s_{13})x_1x_3 + s_{13}x_5(x_1 + x_3 + x_5)]$$

$$\mathcal{F}_C(\mathbf{x}; \mathbf{s}) = -[s_{12}x_1(x_1 + x_3 + x_5) - s_{13}x_3x_5]$$

$$\mathcal{F}_D(\mathbf{x}; \mathbf{s}) = +[(s_{12} + s_{13})x_5(x_1 + x_3 + x_5) + s_{13}x_1x_3]$$

$$\mathcal{F}_E(\mathbf{x}; \mathbf{s}) = +[s_{12}x_3x_5 - s_{13}x_1(x_1 + x_3 + x_5)]$$

$$\mathcal{F}_F(\mathbf{x}; \mathbf{s}) = +[s_{12}x_1x_5 + (s_{12} + s_{13})x_3(x_1 + x_3 + x_5)]$$

3-Loop Non-Planar Box (Crown)

$$\mathbf{s}_R = \{0 < s_{12}, -s_{12} < s_{13} < 0\} \quad (\Rightarrow \quad 0 < s_{12} + s_{13})$$

$$\mathcal{F}_A(\mathbf{x}; \mathbf{s}) = -[s_{12}x_3(x_1 + x_3 + x_5) + (s_{12} + s_{13})x_1x_5]$$

$$\mathcal{F}_B(\mathbf{x}; \mathbf{s}) = -[(s_{12} + s_{13})x_1x_3 + s_{13}x_5(x_1 + x_3 + x_5)]$$

$$\mathcal{F}_C(\mathbf{x}; \mathbf{s}) = -[s_{12}x_1(x_1 + x_3 + x_5) - s_{13}x_3x_5]$$

$$\mathcal{F}_D(\mathbf{x}; \mathbf{s}) = +[(s_{12} + s_{13})x_5(x_1 + x_3 + x_5) + s_{13}x_1x_3]$$

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$$\mathcal{F}_F(\mathbf{x}; \mathbf{s}) = +[s_{12}x_1x_5 + (s_{12} + s_{13})x_3(x_1 + x_3 + x_5)]$$

3-Loop Non-Planar Box (Crown)

$$\mathbf{s}_R = \{0 < s_{12}, -s_{12} < s_{13} < 0\} \quad (\Rightarrow \quad 0 < s_{12} + s_{13})$$

$$\mathcal{F}_A(\mathbf{x}; \mathbf{s}) = -[s_{12}x_3(x_1 + x_3 + x_5) + (s_{12} + s_{13})x_1x_5]$$

$$\mathcal{F}_B(\mathbf{x}; \mathbf{s}) = -[(s_{12} + s_{13})x_1x_3 + s_{13}x_5(x_1 + x_3 + x_5)] \quad \longrightarrow \quad \mathbf{s}_R \text{ is a } \textit{Minkowski regime}$$

$$\mathcal{F}_C(\mathbf{x}; \mathbf{s}) = -[s_{12}x_1(x_1 + x_3 + x_5) - s_{13}x_3x_5]$$

$$\mathcal{F}_D(\mathbf{x}; \mathbf{s}) = +[(s_{12} + s_{13})x_5(x_1 + x_3 + x_5) + s_{13}x_1x_3] \quad \longrightarrow \quad \mathbf{s}_R \text{ is a } \textit{Minkowski regime}$$

$$\mathcal{F}_E(\mathbf{x}; \mathbf{s}) = +[s_{12}x_3x_5 - s_{13}x_1(x_1 + x_3 + x_5)]$$

$$\mathcal{F}_F(\mathbf{x}; \mathbf{s}) = +[s_{12}x_1x_5 + (s_{12} + s_{13})x_3(x_1 + x_3 + x_5)]$$

3-Loop Non-Planar Box (Crown)

3-Loop Non-Planar Box (Crown)

$$J(\mathbf{s}) = 4 \left[\sum_{n_+=1}^3 J_B^{+,n_+} + \sum_{n_+=1}^3 J_D^{+,n_+} + J_E^+ + J_F^+ + (-1 - i\delta^+)^{-2-3\epsilon} [J_A^- + J_B^- + J_C^- + J_D^-] \right]$$

3-Loop Non-Planar Box (Crown)

$$J(\mathbf{s}) = 4 \left[\sum_{n_+=1}^3 J_B^{+,n_+} + \sum_{n_+=1}^3 J_D^{+,n_+} + J_E^+ + J_F^+ + (-1 - i\delta^+)^{-2-3\epsilon} [J_A^- + J_B^- + J_C^- + J_D^-] \right]$$

Verified numerical result against known analytic result ✓

[Henn, Mistlberger, Smirnov, Wasser 20; Bargiela, Caola, von Manteuffel, Tancredi 21]

3-Loop Non-Planar Box (Crown)

$$J(\mathbf{s}) = 4 \left[\sum_{n_+=1}^3 J_B^{+,n_+} + \sum_{n_+=1}^3 J_D^{+,n_+} + J_E^+ + J_F^+ + (-1 - i\delta^+)^{-2-3\epsilon} [J_A^- + J_B^- + J_C^- + J_D^-] \right]$$

Verified numerical result against known analytic result ✓

[Henn, Mistlberger, Smirnov, Wasser 20; Bargiela, Caola, von Manteuffel, Tancredi 21]

$J(s_{12} = 1, s_{13} = -1/5)$ after $\mathcal{O}(\text{mins})$ with pySecDec

3-Loop Non-Planar Box (Crown)

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Verified numerical result against known analytic result ✓

[Henn, Mistlberger, Smirnov, Wasser 20; Bargiela, Caola, von Manteuffel, Tancredi 21]

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How to handle integrals which are **not** UB?

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Case-by-case (⚠ sub-optimal!!) treatment given here $\rightarrow$ **proof-of-concept**

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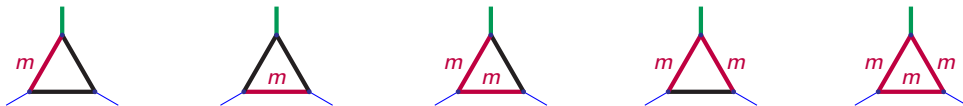


Figure: *Independent* equal-mass triangles with an off-shell leg

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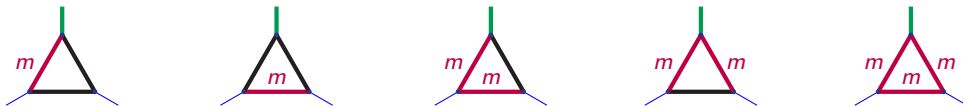


Figure: Independent equal-mass triangles with an off-shell leg

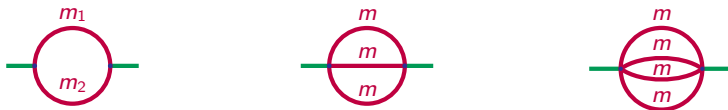
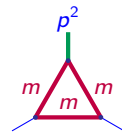
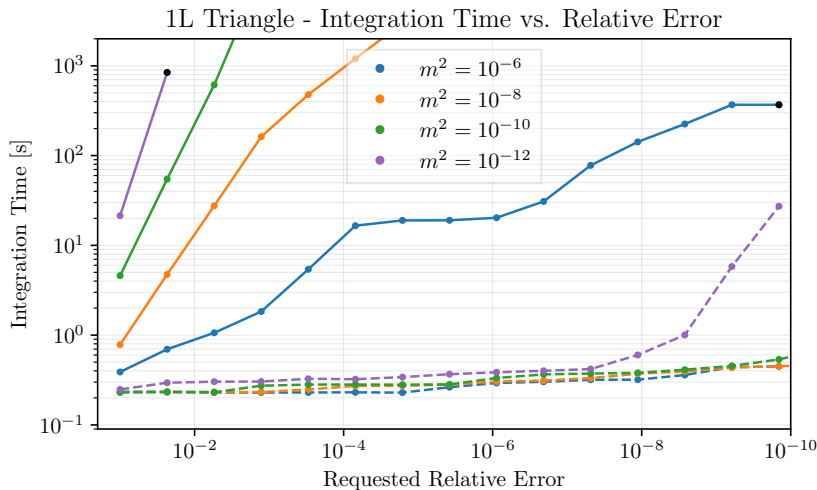


Figure: Banana-type integrals at 1, 2, and 3 loops

1-Loop Massive Triangle with One Off-Shell Leg



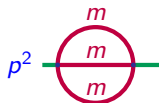
pySecDec

— CD

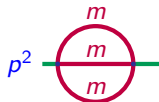
- - - NOCD

$$p^2 = 1$$

Massive Elliptic Sunrise

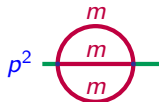


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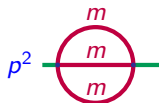
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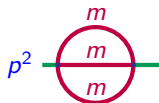
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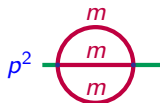
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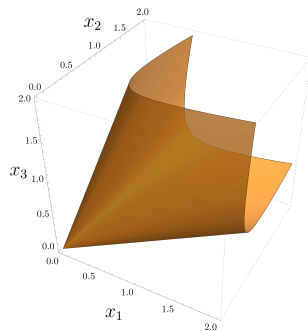
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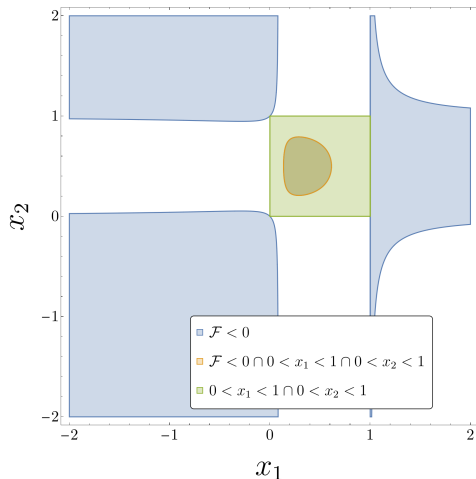


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Reduce the dimension of the problem from 3 variables to 2 by integrating out x_3 using $\delta(1 - x_1 - x_2 - x_3)$ in the integrand

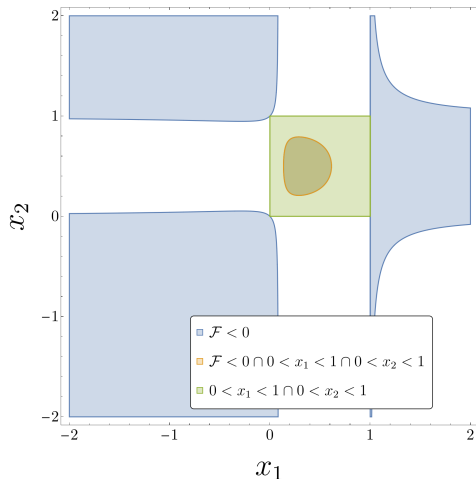


Massive Elliptic Sunrise



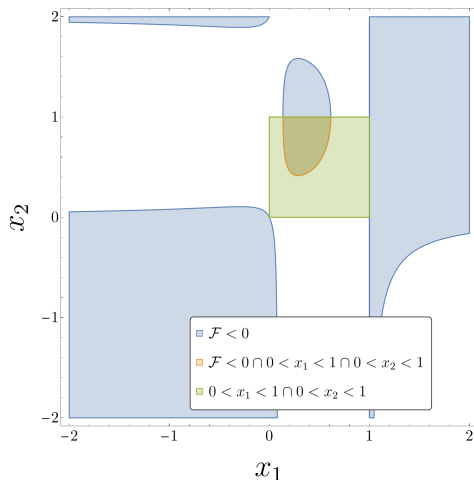
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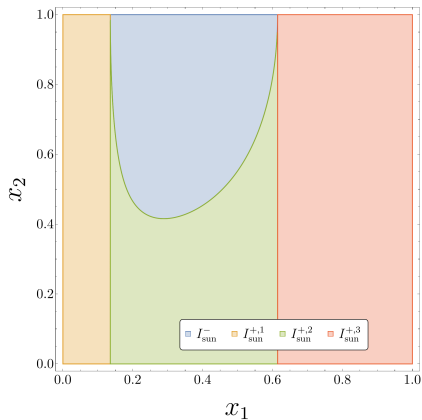
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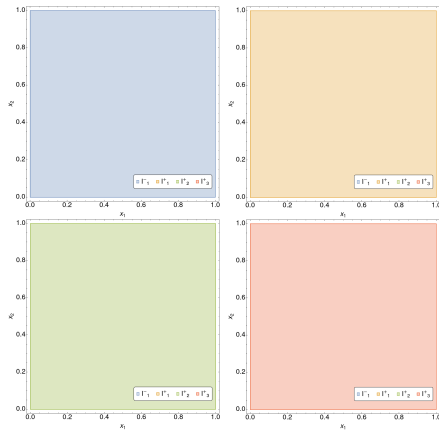
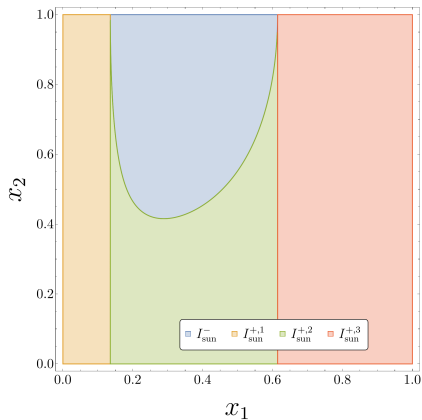
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Thomas Stone

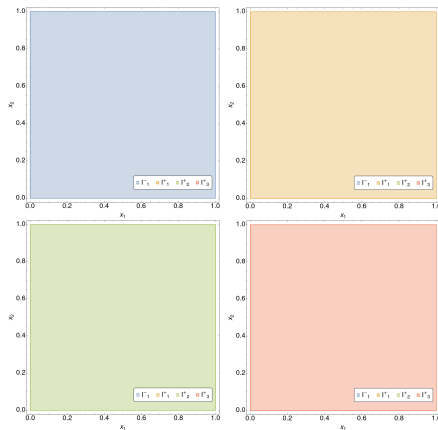
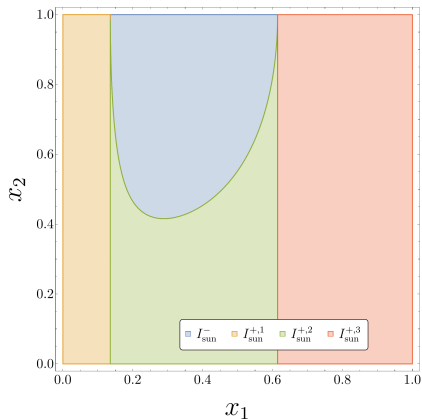
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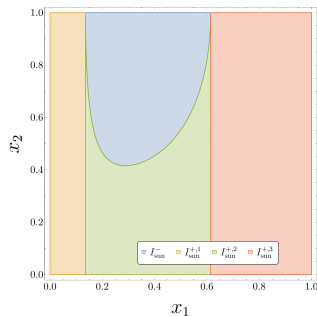


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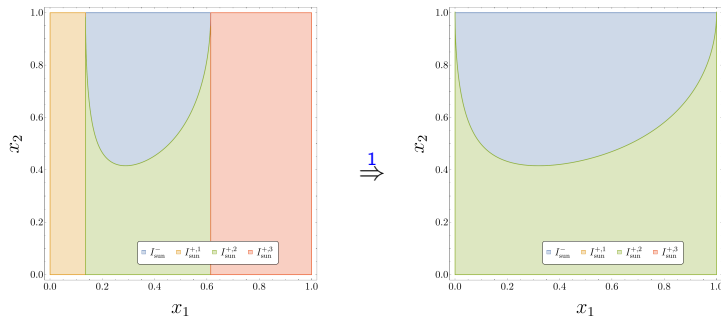


Let's consider the transformations for the **negative** contribution, I_1^-

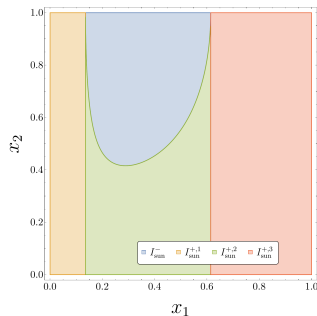
Massive Elliptic Sunrise (Negative Contribution)



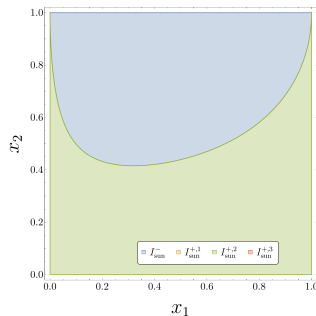
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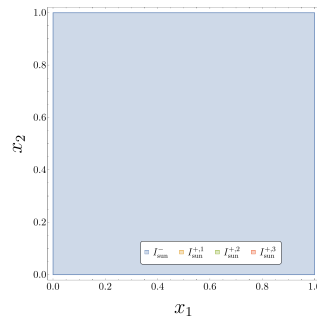
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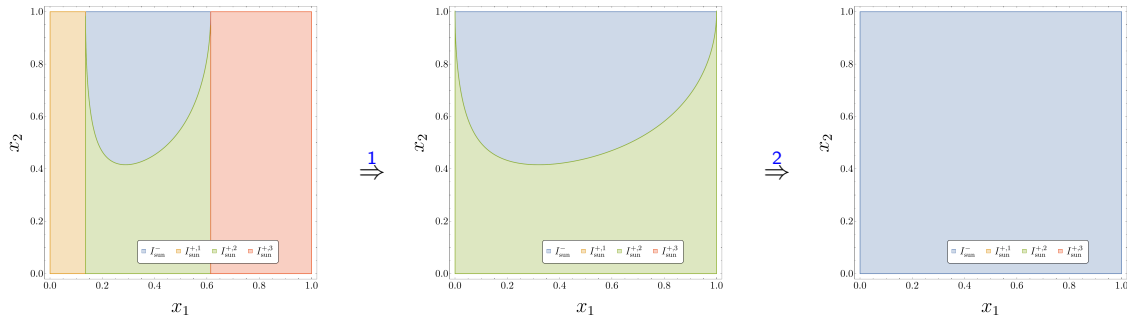
1
⇒



2
⇒

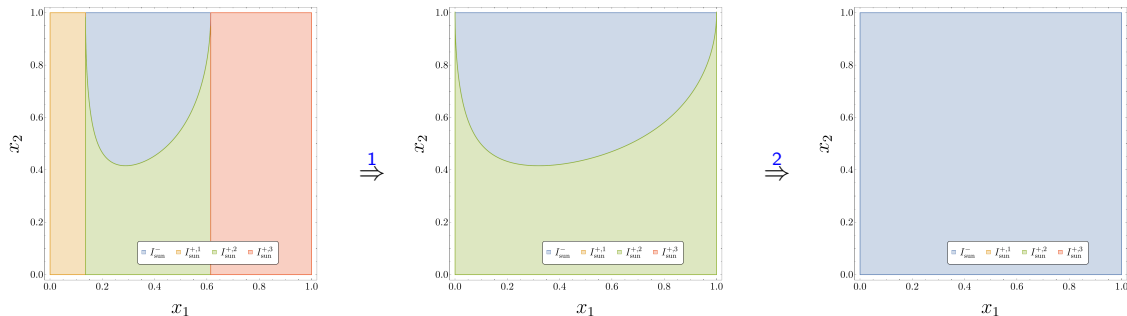


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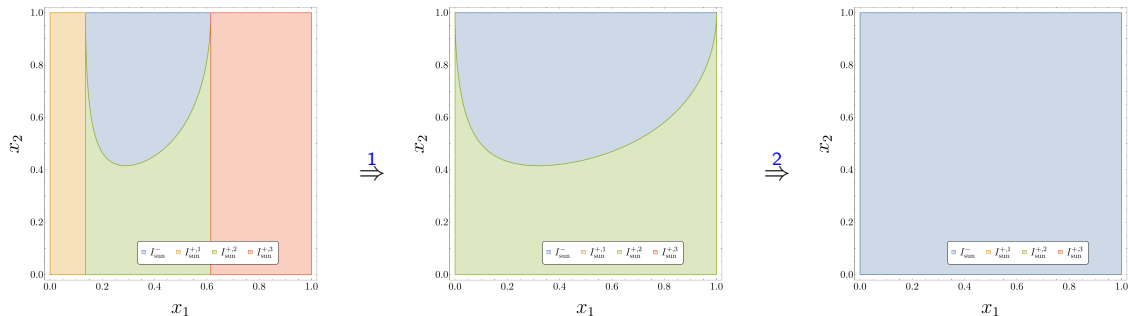
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Massive Elliptic Sunrise (Negative Contribution)



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 $\hookrightarrow \left\{ x_1 \xrightarrow{!} x_1^-(\beta) + [x_1^+(\beta) - x_1^-(\beta)] x_1, \quad x_2 \xrightarrow{!} x_2 \right\}$

Massive Elliptic Sunrise (Negative Contribution)

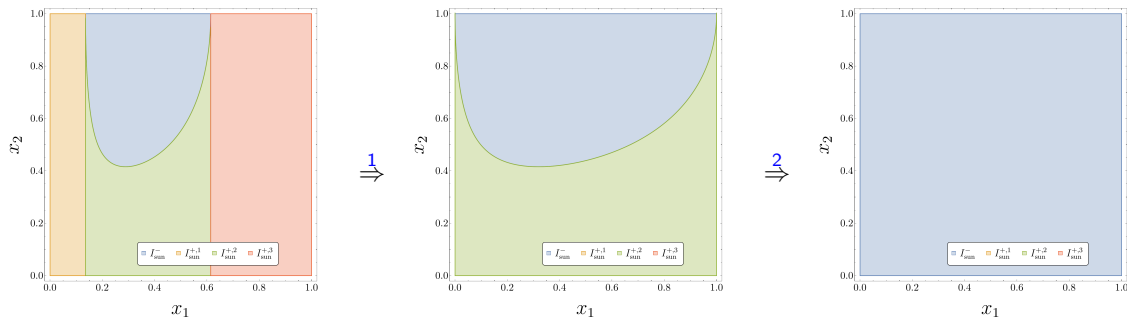


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Massive Elliptic Sunrise (Negative Contribution)



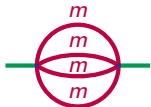
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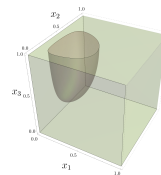
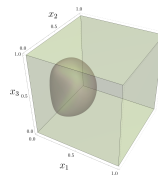
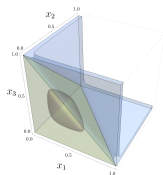
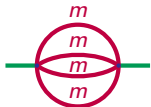
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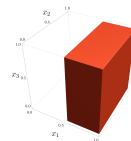
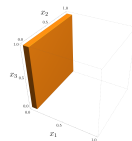
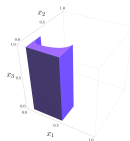
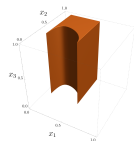
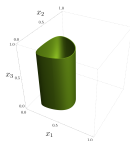
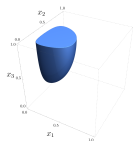
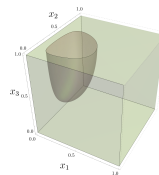
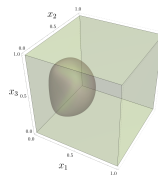
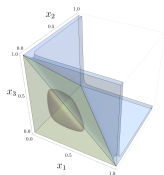
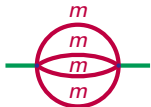
Massive 3-Loop Banana



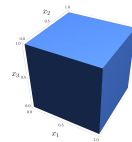
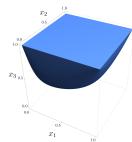
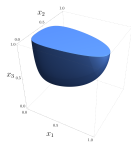
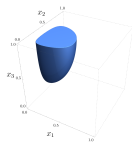
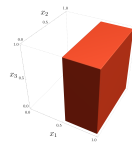
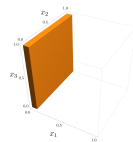
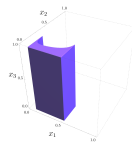
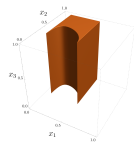
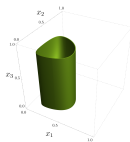
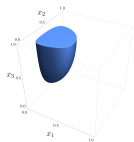
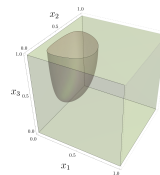
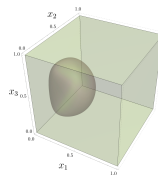
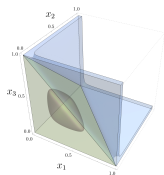
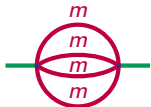
Massive 3-Loop Banana



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Generic Cylindrical Algebraic Decomposition [see also Jones, Loop Summit 2025]

Can we **always** perform this procedure?

Generic Cylindrical Algebraic Decomposition [\[see also Jones, Loop Summit 2025\]](#)

Can we **always** perform this procedure?

Sign-invariant decomposition of

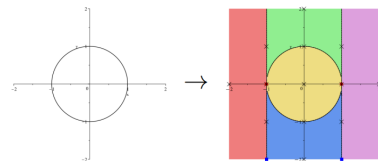
$$\{\mathcal{F}(\mathbf{x}, \mathbf{s}) \geq 0\} \wedge \{\mathbf{x} > 0\} \wedge \mathbf{s}_R$$

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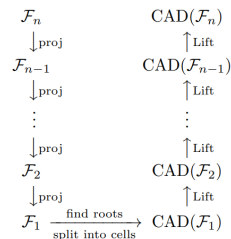
Sign-invariant decomposition of

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Lee, del Río, Rahkooy 25

General solution: **GCAD**

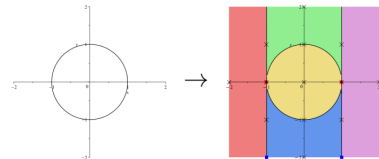


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Sign-invariant decomposition of

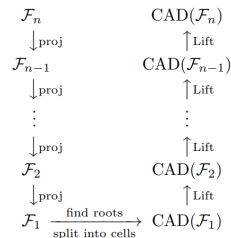
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Lee, del Río, Rahkooy 25

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✓ PROS: constructive algorithm, guaranteed to work

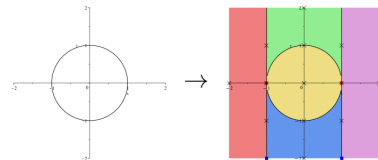


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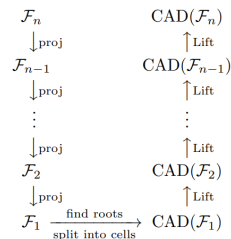


Lee, del Río, Rahkooy 25

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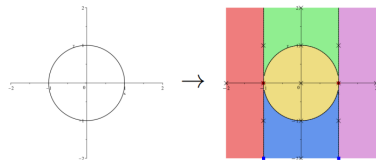


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Lee, del Río, Rahkooy 25

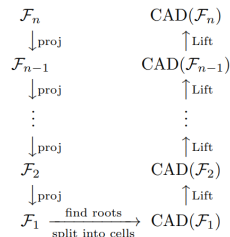
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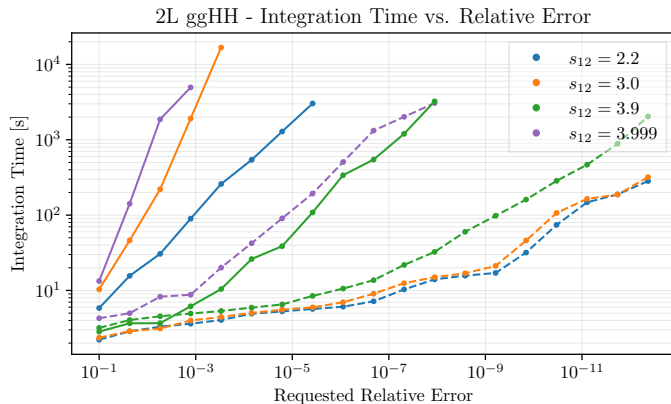
Leverage physics insight? [WIP]:

$\mathcal{F}$ multilinear/quadratic in $\mathbf{x}$, homogeneous



[PRELIMINARY] ggHH 2-Loop Box

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pySecDec

— CD
 - - NOCD

Outlook

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UB integrals $\rightarrow$ problem *solved*

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UB integrals $\rightarrow$ problem *solved* $\rightarrow$ implement in public codes

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Investigate **GCAD** to resolve pheno-relevant integrals

Outlook

UB integrals → problem *solved* → implement in [public codes](#)

Investigate **GCAD** to resolve [pheno-relevant integrals](#)

Connect to picture of **cuts/discontinuities** in [parameter space](#)

[\[Britto, Hannesdottir...\]](#)

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[Britto, Hannesdottir...]

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Thank you for listening!

Outline

5 Backup

