

*A new method for the reconstruction of  
rational functions*

**Xiao Liu**

University of Oxford

Based on [Phys. Lett. B 850 \(2024\) 138491](#) and work in preparation

Loops and Legs in Quantum Field Theory  
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# Outline

**I. Introduction**

**II. The method**

**III. Examples**

**IV. Summary and outlook**

# High precision particle physics

## ➤ Multiloop scattering amplitudes

- Construct the amplitude

$$A = \sum c_i I_i$$

- $I_i$ : scalar Feynman integrals in **dimensional regularization**

$$I(\vec{\nu}) = \int \prod_{i=1}^L \frac{d^D \ell_i}{i\pi^{D/2}} \frac{\mathcal{D}_{K+1}^{-\nu_{K+1}} \cdots \mathcal{D}_N^{-\nu_N}}{(\mathcal{D}_1 + i0)^{\nu_1} \cdots (\mathcal{D}_K + i0)^{\nu_K}}$$

- Compute the scalar integrals: **reduction + computation**

- reduction: express all the scalar integrals in terms of a smaller set of independent integrals (**master integrals**)

$$I_i = \sum_j b_{ij} M_j$$

- computation: compute the master integrals as expansions in the dimensional regulator  $\epsilon = (4 - D)/2$

$$M_j = \sum_{l=-2L} d_{jk} \epsilon^k$$

# Feynman integrals

- **Integration-by-parts reduction** [Chetyrkin and Tkachov, Nucl. Phys. B, 1981] [Laporta, Int. J. Mod. Phys. A, 2000]
  - **AIR** [Anastasiou and Lazopoulos, JHEP, 2004]
  - **FIRE** [Smirnov, JHEP, 2008] [Smirnov, Smirnov, Comput. Phys. Commun., 2013] [Smirnov, Comput. Phys. Commun., 2015] [Smirnov and Chuharev, Comput. Phys. Commun., 2020]
  - **Reduze** [Studerus, Comput. Phys. Commun., 2010] [Manteuffel and Studerus, e-Print: 1201.4330]
  - **Kira** [Maierhofer, Usovitsch and Uwer, Comput. Phys. Commun., 2018] [Klappert, Lange, Maierhofer and Usovitsch, Comput. Phys. Commun., 2021]
  - **LiteRed** [Lee, 2012] [Lee, 2014]
  - **NeatIBP** [Wu, Boehm, Ma, et al, Comput. Phys. Commun., 2024]
- **Finite-field reconstruction** [Manteuffel and Schabinger, Phys. Lett. B, 2015] [Peraro, JHEP, 2016]
  - **FiniteFlow** [Peraro, JHEP, 2019]
  - **FireFly** [Klappert and Lange, Comput. Phys. Commun., 2020]

# Summary

$$\text{time} = \frac{\text{time for a single sample} \times \text{number of samples}}{\text{number of CPUs}}$$

Refined IBP systems:

**syzygy equations** [Gluza, Kajda and Kosower, Phys. Rev. D, 2011] [Larsen and Zhang, Phys. Rev. D, 2016]

**block-triangular systems** [Guan, XL, Ma, Chin.Phys.C, 2020]

**new version of Kira** → talks by Matteo Fael, Fabian Lange

More powerful linear solver:

**RATRACER** [Magerya, e-Print: 2211.03572]

**better interpolation methods** [Klappert and Lange, Comput.Phys.Commun. 2020] [Belitsky, Smirnov, Yakovlev, 2023.02511]

**more compact ansatz** [Badger, Hansen, Chicherin, et al, JHEP 2021][Laurentis, Page, JHEP 2022][Abreu, Laurentis, Ita, et al, 2305.17056]

**Q-linear relations** → talk by Giuseppe De Laurentis

**P-adic reconstruction** → talk by Herschel Chawdhry  
the method in this talk

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# Motivation

## ➤ A simple observation

- Traditional strategy: reconstructing functions individually & neglecting common structures
- Example

$$f_i(x) = \left( \frac{1+x}{1-x} \right)^{i-1}, \quad i \in [1, 100]$$

- approximately 200 samples using Thiele's interpolation formula
- a system of relations

$$(1-x)f_{i+1}(x) - (1+x)f_i(x) = 0, \quad i \in [1, 99]$$

- ansatz + linear fit → 4 samples

$$(a_i + b_i x)f_{i+1}(x) + (c_i + d_i x)f_i(x) = 0$$

- Linear relations → common structures utilized → number of samples reduced

# The method

## ➤ General description

- Relations among nonzero functions  $f_1(\vec{x}), \dots, f_n(\vec{x})$

$$Q_1(\vec{x})f_1(\vec{x}) + \dots + Q_n(\vec{x})f_n(\vec{x}) = 0$$

- Comments:
  - Number of independent relations is  $n - 1$ 
    - not  $n$ , otherwise  $f_i(\vec{x}) = 0$
    - no less than  $n - 2$ , otherwise the basis contains at least two functions  $f_i$

and  $f_j$ . However,  $f_i = \frac{f_i}{f_j} \times f_j$

- Polynomial-coefficients  $Q(\vec{x})$  exist
- Goal: **find  $n - 1$  independent relations with polynomial-coefficients**



# The method

- Algorithm from [Guan, XL, Ma, Chin.Phys.C, 2020]
  1. start with  $n = 0$
  2. make a degree- $n$  ansatz for the relations
  3. generate numerical samples of functions
  4. determine the unknown parameters in the ansatz through a linear fit
  5. count the number of independent relations: if sufficient, end; otherwise increase  $n$  by 1 and go to step 2

# The method

$$f_1(x) = 1, \quad f_2(x) = \frac{1+x}{1-x}$$

- degree-0 ansatz:  $Q_1 f_1(x) + Q_2 f_2(x) = 0$ 
  - $x = 100 \rightarrow f_1 = 1, f_2 = 101/99 \rightarrow Q_1 + 101/99 Q_2 = 0$
  - $x = 101 \rightarrow f_1 = 1, f_2 = 102/100 \rightarrow Q_1 + 102/100 Q_2 = 0$
  - no solution  $\rightarrow$  go to degree-1 ansatz
- degree-1 ansatz:  $(Q_{10} + Q_{11}x)f_1(x) + (Q_{20} + Q_{21}x)f_2(x) = 0$ 
  - $x = 100 \rightarrow f_1 = 1, f_2 = \frac{101}{99} \rightarrow (Q_{10} + 100Q_{11}) + \frac{(Q_{20} + Q_{21}100)101}{99} = 0$
  - $x = 101 \rightarrow f_1 = 1, f_2 = \frac{102}{100} \rightarrow (Q_{10} + 101Q_{11}) + \frac{(Q_{20} + Q_{21}101)102}{100} = 0$
  - $x = 102 \rightarrow f_1 = 1, f_2 = \frac{103}{101} \rightarrow (Q_{10} + 102Q_{11}) + \frac{(Q_{20} + Q_{21}102)103}{101} = 0$
  - $x = 103 \rightarrow f_1 = 1, f_2 = \frac{104}{102} \rightarrow (Q_{10} + 103Q_{11}) + \frac{(Q_{20} + Q_{21}103)104}{102} = 0$
  - One solution:  $Q_{10} = Q_{11} = Q_{20} = -1, Q_{21} = 1$

$$(1+x)f_1(x) + (1-x)f_2(x) = 0$$

# The method

- Summary
  - generator of samples
    - e.g., IBP system + linear solver over finite fields
  - linear relations
    - make various ansatz for  $Q_1(\vec{x})f_1(\vec{x}) + \cdots + Q_n(\vec{x})f_n(\vec{x}) = 0$
    - generate samples
    - linear fit (dense linear system over finite fields)
- About the obtained relations
  - can be easily solved by e.g.
    - traditional rational functions reconstruction strategy
    - additional finite fields + rational numbers reconstruction (Chinese Remainder Theorem + Wang's algorithm [Wang, 1981])

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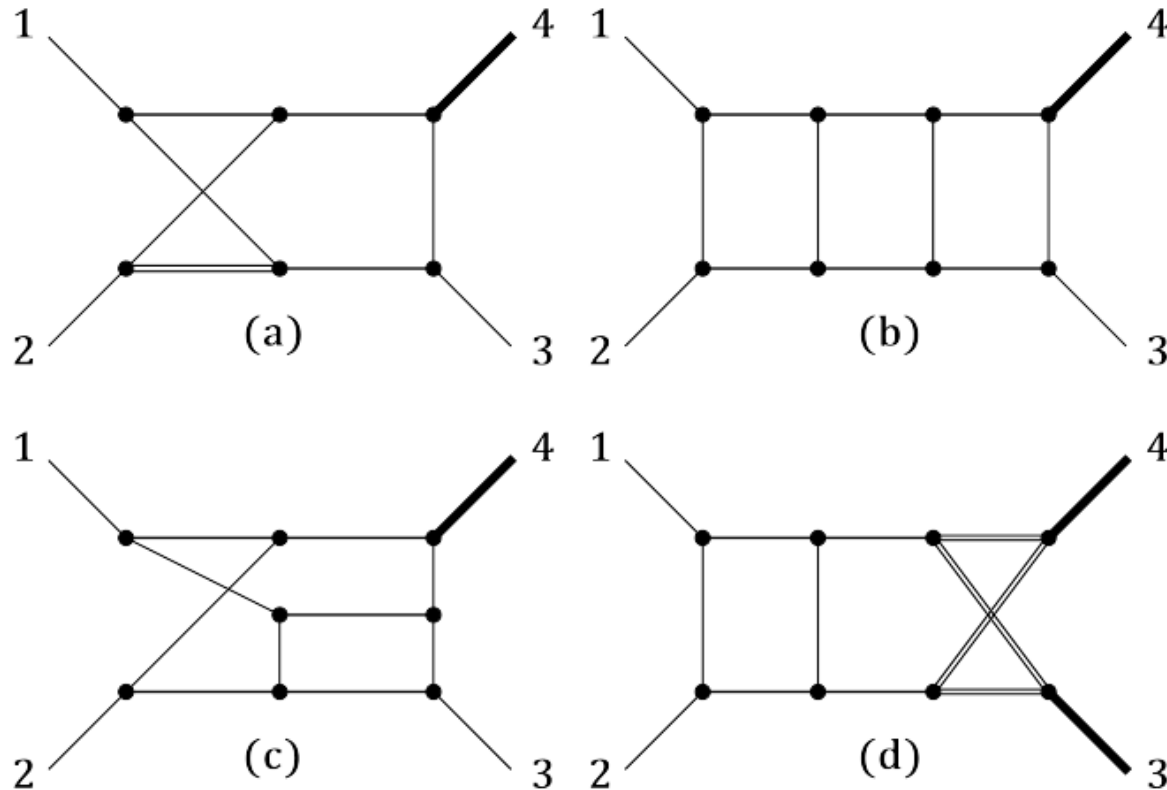
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# Examples

- Reduction coefficients of Feynman integrals or amplitudes

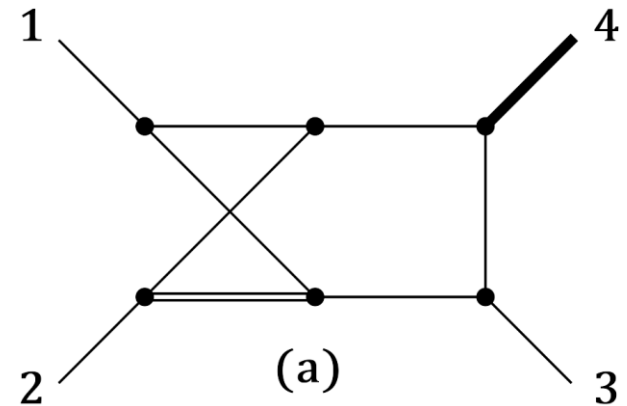
$$\mathcal{A} = f_1 \mathcal{M}_1 + \cdots + f_n \mathcal{M}_n$$

- a common set of denominators reflecting the singularities



# Examples

- Topology (a): two-loop amplitude of the mixed QCD-electroweak correction to  $pp \rightarrow Z + j$  [Bargiela, Caola, Chawdhry, XL, 2312.14145]
- Setup
  - $m_Z^2 = 1, m_W^2 = 7/9$
  - remaining:  $\{\epsilon, s_{12}, s_{13}\}$
  - 56 master integrals  $\Rightarrow$  56 rational functions
  - auxiliary function  $f_{57} = 1$
  - LiteRed + FiniteFlow
- Details
  - Completed with **degree-6** ansatz
  - Number of samples: from 54978 (FiniteFlow) to **5321**  $\rightarrow$  a factor of **10.3**
  - Computational cost: from 4.6h to **0.47h**  $\Rightarrow$  a factor of **9.8**
  - The computational cost is dominated by **samples generation.**

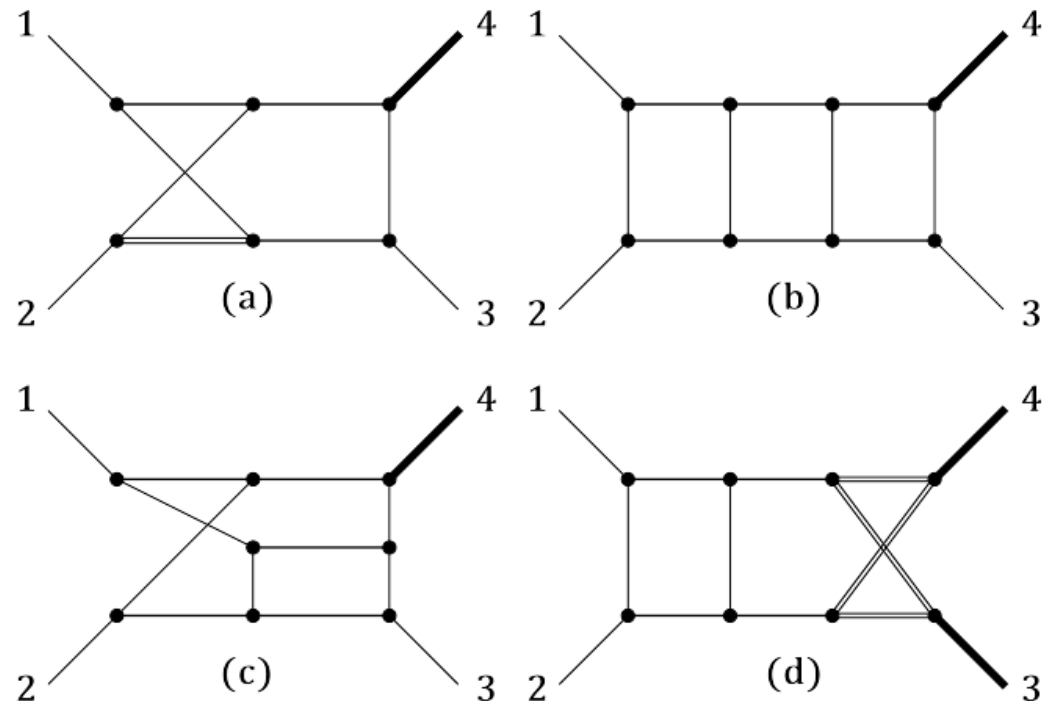




# Examples

topology	variables	generator	degree	samples	samples (old)	$R_{\text{samples}}$	cost	cost (old)	$R_{\text{cost}}$
(a), 56	$\epsilon, s_{12}, s_{13}$	LR+FF	6	5321	54978	10.3	0.47h	4.6h	9.8
(b), 83	$\epsilon, s_{12}, s_{13}$	NI+FF	8	15208	145722	9.6	8.2h	78.5h	9.6
(c), 280	$\epsilon, s_{12}, s_{13}$	LR+FF	8	43662	2351622	54	8549h	450728h	53
(d), 336	$\epsilon, \eta$	LR+FF	5	41350	473946	11.5	340h	2320h	9.5

- (a): amplitude
- (b): rank -6 integral
- (c): differential equations with respect to  $s_{12}, s_{13}$
- (d): differential equations with respect to  $\eta$  (AMFlow)



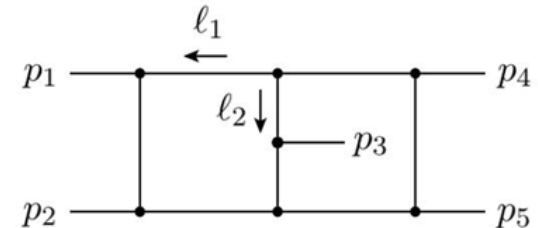
# Examples

- Discussion
  - samples generation + linear fit + explicit solutions
  - linear fit
    - **Undominated if the ansatz is not too big (less than 20000 parameters)**
    - For most problems with less than 3 variables, this holds true.
      - For topology (c): the biggest ansatz contains **6810** parameters
    - For problems with five-point kinematics, this can become a problem.
  - explicit solutions
    - **Undominated in most cases**
    - For topology (b): IBP system costs 1.3s per point; our linear relations cost 0.0024s per point → **540 times faster**

# Examples

- A step towards multivariate problems [in preparation]

- five variables:  $\epsilon, s_{23}, s_{34}, s_{45}, s_{51}$  ( $s_{12} = 1$ )
- full ansatz may contain  $O(100000)$  parameters
- work in four-dimensional slices ( $s_{23}, s_{34}, s_{45}, s_{51}$ )



- $\epsilon \rightarrow \epsilon_0$
- For fixed  $\epsilon$ , deal with the four-variate problem (ansatz size: 14000)

$$\frac{a_0 + a_1 s_{23} + a_2 s_{34} + a_3 s_{45} + \dots}{1 + b_1 s_{23} + a_2 s_{34} + b_3 s_{45} + \dots}$$

- $a$ 's and  $b$ 's are rational functions of  $\epsilon$  to be reconstructed
- 400000 samples in total to reconstruct the fully analytic expressions over a finite field
- FiniteFlow requires approximately  $O(10^7)$  points

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# Summary and Outlook

- A new method for the reconstruction of rational functions is proposed, which works by exploiting all the independent linear relations among the target functions.
- The method effectively reduces the number of sample points required in the framework of finite-field sampling.
- The method works well for problems with no more than three variables, i.e., problems with three-point or four-point kinematics.
- For problems with five-point kinematics, some preliminary results shows that the method can still work, but further studies are required.

***Thank you!***