# Loop amplitudes at the precision frontier 

## Simon Badger (University ofTurin)

based on work with:<br>Aylett-Bullock, Brønnum-Hansen, Becchetti, Butter, Chaubey, Hartanto, Luchmann, Marcoli, Marzucca, Moodie, Peraro, Pitz, Plehn, Chicherin, Gehrmann, Henn, Zoia

ACAT, Bari

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# motivation: where is the precision frontier? 

# part I: <br> finite field arithmetic and two-loop amplitudes 

part II:<br>amplitude neural networks

from theory to experiment

from theory to experiment

from theory to experiment:


## from theory to experiment: precison frontier $\rightarrow$ I \%




## loop frontier



N3LO 2 $\rightarrow 2$, N4LO $2 \rightarrow 1(\mathrm{gg} \rightarrow \mathrm{H})$

## IR frontier



N3LO splitting functions, analytic resummation, SCET, beam functions etc.

## the precision wishlist

latest update LH202 I Huss, Huston, Jones, Pellen [2207.02 I 22]


precision
measurements
computations at the precision frontier


## bare amplitudes

$$
A^{(L), 4-2 \epsilon}=\sum_{i} c_{i}(\epsilon,\{p\}) \mathcal{F}_{i}(\epsilon,\{p\})
$$

rational functions integrals/special functions

$$
F^{(L)}=A^{(L), 4-2 \epsilon}-\sum_{k=1}^{L} I^{(k), 4-2 \epsilon} A^{(L-k), 4-2 \epsilon}
$$

universal IR/UV poles
[Catani (1998)][Becher, Neubert (2009)]
[Magnea, Gardi (2009)]

## computational toolbox

## numerical unitarity

## on-shell methods

hidden simplicity and underlying geometry

## momentum twistors

rational kinematics

## integrand reduction <br> algebraic reduction

## syzygy relations

optimising systems of IBP identities
recursion relations
reusing common blocks to evaluate diagrams efficiently

## computational toolbox



# part I: <br> finite fields arithmetic and two-loop amplitudes 

with: Brønnum-Hansen, Becchetti, Chaubey, Hartanto, Marcoli, Marzucca, Moodie, Peraro, Chicherin, Gehrmann, Henn, Zoia

## finite field arithmetic

## not a new idea - used in many computer algebra systems

## solving IBP systems: e.g. FINRED [von Manteuffe], <br> KIRA+FIREFLY [Maierhoefer, Usovitsch, Uwer, Klappert, Lange]

## framework for amplitude computations: FINITEFLOW [Peraro (2019)]

```
(* take some, reasonably large, prime number *)
FFPrimeNo[1]
(* all quantities evaluated modulo a prime number *
Mod[-3,FFPrimeNo[1]]
Mod[87+FFPrimeNo[1],FFPrimeNo[1]]
Solve[b*3==87,Modulus }->\mathrm{ FFPrimeNo[1]][[1]]
(* already implemented in Mathematica *)
```

Mod[87/3+FFPrimeNo[1], FFPrimeNo [1]]
9223372036854775643
9223372036854775640
87
$\{b \rightarrow 29\}$
29
extremely efficient solutions
to linear algebra systems
other talks at this year's ACAT
De Laurentis
Moodie
Usovitsch

$$
A^{(L), 4-2 \epsilon}=\sum_{i} c_{i}(\epsilon,\{p\}) \mathcal{F}_{i}(\epsilon,\{p\})
$$

## rational functions

## multiple numerical (mod prime) evaluations can used to reconstruct complete analytic information

> Newton (polynomial) and Thiele (rational) interpolation

Rational external kinematics: e.g. Momentum Twistors (Hodges)

```
* implement the Newton interpolation algorithm *
NewtonReconstruct[z_, zvalues_List, fvalues_List, primeno_]:=Module[{res,maxdegree,aa,eqs,sol},
maxdegree = Length[zvalues]-1;
res = Sum[aa[r]*Product[(z-zvalues[[i+1]]),{i,0,r-1}],{r,0,maxdegree}];
eqs = Equal@@@Transpose[{res /. ({Rule[z,#]}&/@zvalues),fvalues}];
sol = Solve[eqs,Table[aa[i],{i,0,Length[fvalues]-1}],{Modulus->primeno}];
Return[res/. sol[[1]]];
]
fff[\mp@subsup{z}{-}{\prime}]:=15/2*z+119/6* *^2;
values = {19,44,78};
FFRatMod[fff/@values,FFPrimeNo[0]]
test = NewtonReconstruct[z,values,%,FFPrimeNo[0]]
Collect[%,z,FFRatRec[#,FFPrimeNo[0]]&]
{6148914691236524491, 6148914691236555 916, 121 251}
6148914691236524491 + 1257 (-19 + z) + 1537228672809129317 (-44 + z)(-19 + z)
\frac{15z}{2}+\frac{119\mp@subsup{z}{}{2}}{6}
```


## Trivial parallelisation of sample points

## finite fields for amplitudes

## useful features:

- reconstruct exact results using chinese remainder theorem
- extremely efficient solutions to large linear systems
- reconstruct rational functions using Newton/Thiele interpolation
- modular approach in FiniteFlow allows us to link different algorithms and avoid large intermediate steps



## amplitudes $\rightarrow d \sigma$

$$
A^{(L), 4-2 \epsilon}=\sum_{i} c_{i}\left(\epsilon,\{p\} \quad F_{i}(\epsilon,\{p\})\right.
$$

pentagon functions. now with one off-shell leg [Chicherin, Sotnikov, Zoia '2I]

## $\mathrm{d} \sigma \mathrm{gg} \rightarrow \mathrm{YYg}$

[SB et al. 2 I 09.1 2003]




$A^{(2)} p p \rightarrow W 2 j$ Lcv
[Abreu et al. $2 \mid 10.0754$ I]

$$
d \sigma p p \rightarrow 3 j \text { Lcv }
$$

[Czakon et al. 2 106.0533 I]


LCV = Leading Colour
Double Virtual

$$
\text { d } \sigma \text { gg } \rightarrow 3 \mathrm{~g} \text { LCV }
$$

[Chen et al. 2203.13531]

## $d \sigma p p \rightarrow W b b$ Lcv <br> [Hartanto et al. 2205.01687]


$A^{(2)} \mathrm{pp} \rightarrow$ WYj Lcv
[SB et al. 220I.04075]
$\mathrm{d} \sigma \mathrm{Pp} \rightarrow \mathrm{YYj}$ LCV
[Czakon et al. 2 105.06940]


## differential equations for $\mathrm{pp} \rightarrow \mathrm{tt}+\mathrm{j}$


[SB, Becchetti, Chaubey, Marzucca (to appear)]
|st steps towards pp $\rightarrow$ tt+j @ NNLO in QCD
see also IL pp $\rightarrow \mathrm{tt+j} \mathrm{O}\left(\varepsilon^{2}\right)$ [SB, Becchetti, Chaubey, Marzucca]
canonical form [Henn 'I3] DE of 88 master integrals

$$
\begin{aligned}
& d \overrightarrow{\mathcal{I}}(\vec{x}, \varepsilon)=\varepsilon d A(\vec{x}) \overrightarrow{\mathcal{I}}(\vec{x}, \varepsilon) \\
& A(\vec{x})=\sum c_{i} \log \left(w_{i}(\vec{x})\right) .
\end{aligned}
$$

IBP reduction and reconstruction
over finite fields: good basis leads to simple reconstruction
dLog alphabet of 7 I letters
numerical evaluation with generalised series expansions (DiffExp)

$$
\begin{aligned}
& \mathcal{I}_{1}=\epsilon^{4} 8 d_{23} d_{45}\left(d_{12}+m_{t}^{2}\right) I_{1,1,1,1,1,1,1,1,-1,0,0}, \\
& \mathcal{I}_{2}=\epsilon^{4} \frac{d_{45}}{2 \operatorname{tr}_{5}} I_{1,1,1,1,1,1,1,1,0,0,0}^{[11]}, \\
& \mathcal{I}_{3}=\epsilon^{4} \frac{d_{45}}{2 \operatorname{tr}_{5}} I_{1,1,1,1,1,1,1,0,0,0}^{[12]} .
\end{aligned}
$$

high precision boundary values
(AMFlow)

## two-loop five-point processes in NJET

| Channel | f64/f64 |  | Evaluation strategy |  |
| :---: | :---: | :---: | :---: | :---: |
|  | Time (s) | $f(\%)$ | Time (s) | $f(\%)$ |
| $g g \rightarrow g g g$ | 1.39 | 69 | 1.89 | 77 |
| $g g \rightarrow \bar{q} q g$ | 1.35 | 91 | 1.37 | 91 |
| $q g \rightarrow q g g$ | 1.34 | 92 | 1.57 | 93 |
| $q \bar{q} \rightarrow g g g$ | 1.34 | 93 | 1.38 | 93 |
| $\bar{q} Q \rightarrow Q \bar{q} g$ | 1.14 | 99 | 1.16 | 99 |
| $\bar{q} \bar{Q} \rightarrow \bar{q} \bar{Q} g$ | 1.36 | 99 | 1.39 | 99 |
| $\bar{q} g \rightarrow \bar{q} Q \bar{Q}$ | 1.36 | 99 | 1.39 | 99 |
| $\bar{q} q \rightarrow Q \bar{Q} g$ | 1.14 | 99 | 1.14 | 99 |
| $\bar{q} g \rightarrow \bar{q} q \bar{q}$ | 1.84 | 99 | 1.90 | 99 |
| $\bar{q} \bar{q} \rightarrow \bar{q} \bar{q} g$ | 1.82 | 99 | 1.94 | 99 |
| $\bar{q} q \rightarrow q \bar{q} g$ | 1.71 | 99 | 1.77 | 99 |
| $g g \rightarrow \gamma \gamma g *$ | 9 | 99 | 26 | 99 |


https://bitbucket.org/njet/njet
from Ryan Moodie's slides


## part II: amplitude neural networks

with: Aylett-Bullock, Butter, Luchmann, Moodie, Pitz, Plehn
other talks and new results!
Bothmann, Butter, Truong, Janssen

## optimising simulations

how can we speed up simulations with expensive amplitude calls?

SB, (Aylett-)Bullock [2002.075 16] Aylett-Bullock, SB, Moodie [2I 06.09474]


- Single NN does badly
- Understanding IR sectors via FKS improves reliablity (ensemble of networks)
- Error estimates by varying model initialisation
- Various tests suggest single run speed improvements at least $\times 10$
factorisation aware approach looks to be working nicely!


Maitre,Truong [2| 07.06625]


## Bayesian Networks

SB, Butter, Luchmann, Pitz, Plehn [2206. I 483I]
another experiment with [See talk by Butter] loop amplitudes
gg $\rightarrow$ YYg, $g g \rightarrow Y Y g g$ @(IL)^2

- better defined error estimates
- improved training via loss and performance boosting




## outlook

- new theory techniques are essential to meet the precision requirements at the LHC
- finite field arithmetic is making a dent in the $2 \mathrm{~L} 2 \rightarrow 3$ wishlist
- some progress for amplitudes with internal masses (ttj)
- amplitudes neural networks look to be a promising way to significantly optimise MC simulations

