

# Fast Feynman integration with tropical geometry

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Based on [arXiv:2008.12310](https://arxiv.org/abs/2008.12310)

# Motivation

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# Quantum field theory

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perturbative expansions

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$$\text{where } I_G = \prod_{\ell} \int d^D \mathbf{k}_{\ell} \prod_{e \in E} \frac{1}{D_e(\{\mathbf{k}\}, \{\mathbf{p}\}, m_e)}.$$

# Questions

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What is the value of  $A_0, A_1, A_2, A_3, \dots$ ?

How can we calculate them effectively?

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## Practical question

What is the value of  $A_0, A_1, A_2, A_3, \dots$ ?

How can we calculate them effectively?

## Associated abstract question: Computability

Is there an algorithm to compute  $A_n$ ?

What is *the fastest* algorithm to compute  $A_n$ ?

What is its runtime?



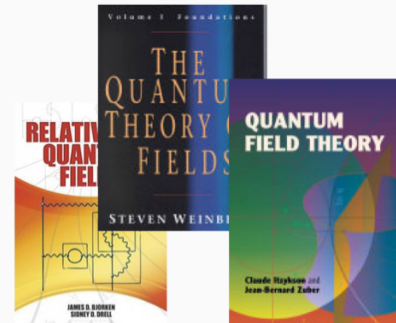
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$$\mathcal{O}(\hbar) = \sum_{n \geq 0} A_n \hbar^n$$

YES



# What is its runtime?

$$\mathcal{O}(\hbar) = \sum_{n \geq 0} A_n \hbar^n = \sum_{\text{graphs } G} \frac{I_G}{|\text{Aut } G|} \hbar^{L_G}$$

Runtime to compute  $A_n$  for  $n$  large:

$$\mathcal{O}\left(\alpha^n \Gamma(n + \beta)\right) \times F(n)$$

→ number of Feynman graphs  
with  $n$  loops for  $n \rightarrow \infty$   
( $\alpha$  and  $\beta$  depend on theory  
and observable)

→ time it takes to  
evaluate a single  
Feynman integral  
of order  $n$

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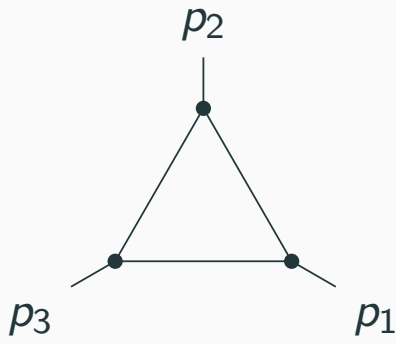
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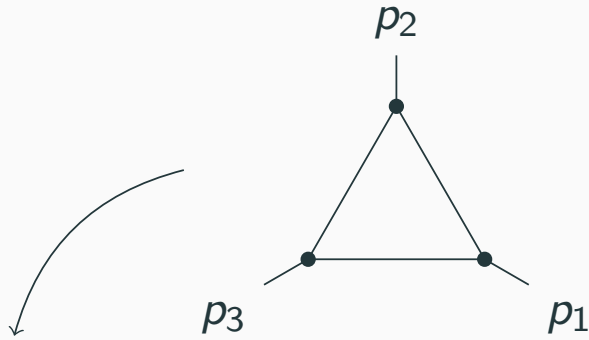
⇒ Tough to phrase runtime question for analytic computations.

# Analytic vs numerical evaluation



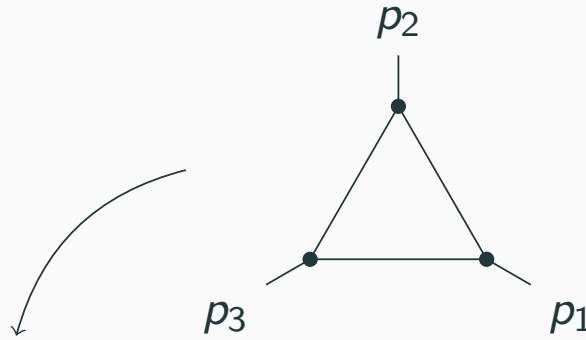


# Analytic vs numerical evaluation



$$\int \frac{d^4 k}{\pi^2} \frac{1}{k^2 (k+p_1)^2 (k+p_1+p_2)^2}$$

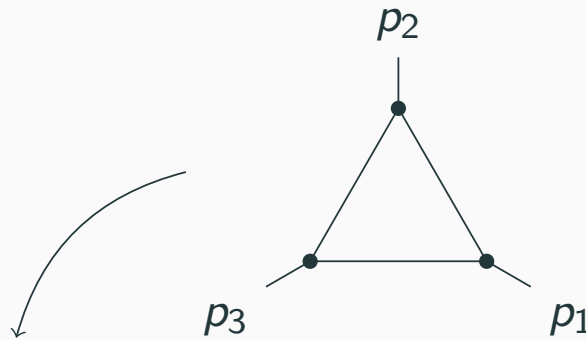
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$$\int \frac{d^4 k}{\pi^2} \frac{1}{k^2 (k+p_1)^2 (k+p_1+p_2)^2} = \frac{D(z, \bar{z})}{\sqrt{-\lambda(p_{12}^2, p_{13}^2, p_{23}^2)}}$$

The diagram above is connected to the integral by a curved arrow pointing from the diagram to the integral. A second curved arrow points from the integral to the right-hand side of the equation.

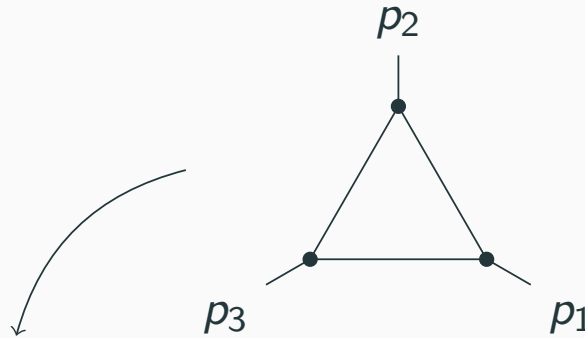
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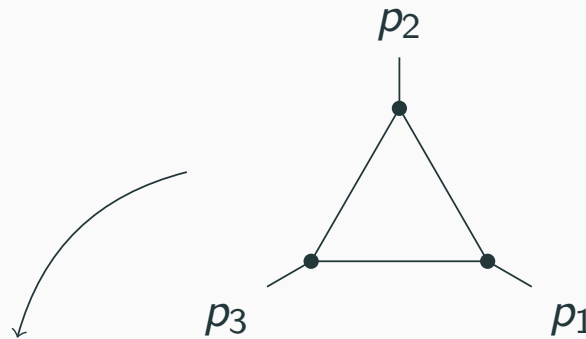
$$z\bar{z} = \frac{p_{13}^2}{p_{12}^2}, \quad (1-z)(1-\bar{z}) = \frac{p_{23}^2}{p_{12}^2}$$

$$D(z, \bar{z}) = \text{Im}(\text{Li}_2(z) + \log(1-z) \log|z|)$$

$$\lambda(x, y, z) = x^2 + y^2 + z^2 - 2xy - 2xz - 2yz$$

$$\text{Li}_2(z) = \sum_{n=1}^{\infty} \frac{z^n}{n^2}$$

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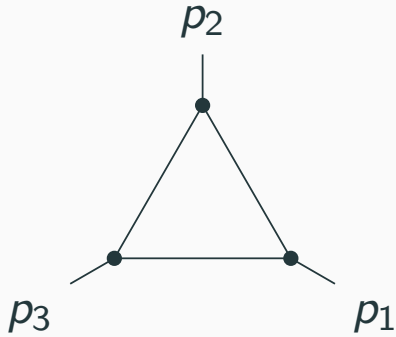
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slow 'brute-force' evaluation

fast evaluation

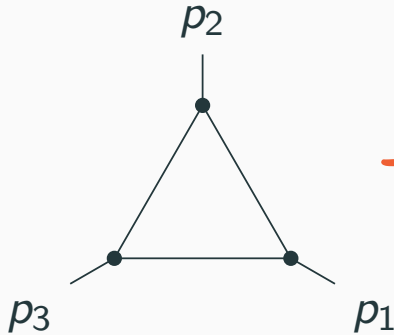
# Approach here: slow but general



$$\int \frac{d^4 k}{\pi^2} \frac{1}{k^2 (k+p_1)^2 (k+p_1+p_2)^2}$$

Numerical  
answer

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Numerical  
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Strategy here:

Start with slow-but-general approach,  
then improve on it.

# Direct evaluation

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# Algebraic geometric perspective

$$\mathcal{O}(\hbar) = \sum_{n \geq 0} A_n \hbar^n = \sum_{\text{graphs } G} \frac{I_G}{|\text{Aut } G|} \hbar^{L_G}$$

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rewrite via Schwinger trick/Feynman parameters:

$$I_G = \Gamma(\omega_G) \int_{\mathbb{P}_{>0}^{E-1}} \frac{\Omega}{\Psi_G(\mathbf{x})^{D/2}} \left( \frac{\Psi_G(\mathbf{x})}{\Phi_G(\mathbf{x})} \right)^{\omega_G}$$

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$$\int_{\mathbb{P}_{>0}^{E-1}} \frac{\Omega}{\Psi_G(\mathbf{x})^{D/2}} \left( \frac{\Psi_G(\mathbf{x})}{\Phi_G(\mathbf{x})} \right)^{\omega_G}$$

where

- $\Omega$  is the standard volume form on  $\mathbb{P}^{E-1}$ :  

$$\Omega = \sum_{k=1}^E (-1)^k dx_1 \wedge \dots \wedge \widehat{dx_k} \wedge \dots \wedge dx_E.$$
- $\Psi_G = \sum_T \prod_{e \notin T} x_e$  (sum over spanning trees)
- $\Phi_G = \sum_F \|\mathbf{p}(F)\|^2 \prod_{e \notin F} x_e + \Psi_G \sum_e m_e^2 x_e$  (sum over 2-forests)
- $\Psi_G$  and  $\Phi_G$  are **homogeneous** polynomials in  $x_1, \dots, x_E$ .
- We assume that the integral exists.

$$\int_{\mathbb{P}_{>0}^{E-1}} \frac{\Omega}{\Psi_G^{D/2}(\mathbf{x})} \left( \frac{\Psi_G(\mathbf{x})}{\Phi_G(\mathbf{x})} \right)^{\omega_G}$$

$\Psi_G(\mathbf{x})$  and  $\Phi_G(\mathbf{x})$  exhibit complicated geometric structures.

$\Rightarrow$  These integrals are hard to evaluate

$\Rightarrow$  These integrals are very interesting

$$\int_{\mathbb{P}_{>0}^{E-1}} \frac{\Omega}{\Psi_G^{D/2}(\mathbf{x})} \left( \frac{\Psi_G(\mathbf{x})}{\Phi_G(\mathbf{x})} \right)^{\omega_G}$$

### Obstruction for direct numerical evaluation

Integrand has singularities on the boundary of  $\mathbb{P}_{>0}^{E-1}$ .

I.e. vanishing locus of  $\Psi_G$  and  $\Phi_G$  meets the boundary of  $\mathbb{P}_{>0}^{E-1}$ .

⇒ Singularities have to be blown up first

# Obstructions to numerical evaluation: Stereotypical example

$$\int_0^1 \int_0^1 \frac{dx dy}{x + y}$$

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naive Monte Carlo evaluation

where  $X_n, Y_n \in [0, 1]$  are i.i.d. uniform random variables.



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$$\mathbf{Var} \approx \int_0^1 \int_0^1 \frac{dx dy}{(x+y)^2} = \infty$$

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$$\mathbf{Var} \approx \int_0^1 \int_0^1 \frac{dx dy}{(x + y)^2} = \infty \rightarrow \text{naive Monte Carlo fails}$$

where  $X_n, Y_n \in [0, 1]$  are i.i.d. uniform random variables.

⇒ even direct 'brute-force' numerical Feynman integration  
is a non-trivial problem

# Traditional solution

$$\int_{\mathbb{P}_{>0}^{E-1}} \frac{\Omega}{\Psi_G^{D/2}(\mathbf{x})} \left( \frac{\Psi_G(\mathbf{x})}{\Phi_G(\mathbf{x})} \right)^{\omega_G} \rightarrow \int_{\mathbb{P}_{>0}^{n-1}} \frac{\prod_i p_i(\mathbf{x})^{\mu_i}}{\prod_j q_j(\mathbf{x})^{\nu_j}} \Omega$$

Feynman integral

generalization

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Feynman integral

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## Sector decomposition approach

- Algorithms to perform blowups in the general case:  
**Binoth, Heinrich '03; Bogner, Weinzierl '07; (Hironaka 1964)**
- Simple geometric interpretation:  
**Kaneko, Ueda '09**

(No use of rich structure of Feynman integrals)

# Projective algebraic integrals

$$\int_{\mathbb{P}_{>0}^{E-1}} \frac{\Omega}{\Psi_G^{D/2}(\mathbf{x})} \left( \frac{\Psi_G(\mathbf{x})}{\Phi_G(\mathbf{x})} \right)^{\omega_G} \rightarrow \int_{\mathbb{P}_{>0}^{n-1}} \frac{\prod_i p_i(\mathbf{x})^{\mu_i}}{\prod_j q_j(\mathbf{x})^{\nu_j}} \Omega$$

Feynman integral

generalization

The general right hand side

- is essentially a ‘Stringy integral’  
**Arkani-Hamed, He, Lam 2019.**
- is also a generalized Mellin transform **Nilsson, Passare 2010;**  
**Berkesch, Forsgård, Passare 2011.**
- can be interpreted in terms of toric geometry/varieties  
**Schultka 2018.**

# Approximate runtime of the sector decomposition approach

$$\int_{\mathbb{F}_{>0}^{n-1}} \frac{\prod_i p_i(\mathbf{x})^{\mu_i}}{\prod_j q_j(\mathbf{x})^{\nu_j}} \Omega$$

Numerical evaluation using sector decomposition for blowups:



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Numerical evaluation using sector decomposition for blowups:

- Runtime to evaluate the integral up to  $\delta$ -accuracy

$$\approx \mathcal{O}(V^2 \cdot \delta^{-2})$$

where  $V$  is the number of monomials in  $\frac{\prod_i p_i(\mathbf{x})^{\mu_i}}{\prod_j q_j(\mathbf{x})^{\nu_j}}$ .

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- For Feynman integrals  $V$  grows  $\approx$  exponentially with  $n$ .

4  
# of edges

# New results

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2. The general (oblivious) approach can be accelerated:

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3. Euclidean Feynman integration can be accelerated extremely:

$$\mathcal{O}(V^2 \cdot \delta^{-2}) \approx \mathcal{O}(2^{cn} \cdot \delta^{-2}) \quad \rightarrow \quad \mathcal{O}(n2^n + n^4\delta^{-2})$$

with  $c \gg 1$  where  $n$  is the number of edges of the graph.

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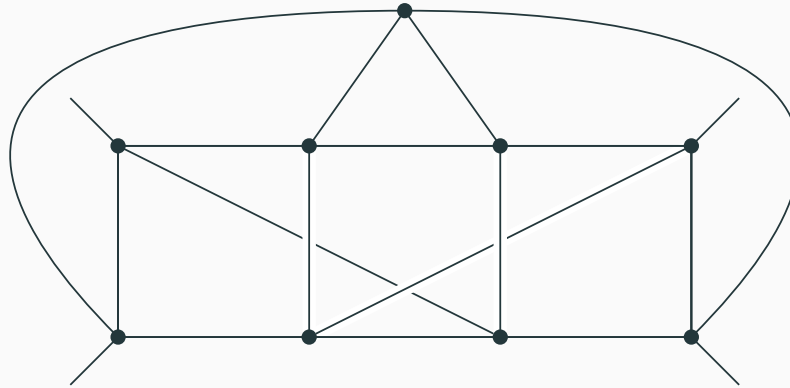
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- New:  $\approx 17$  loops possible (with basic implementation).

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- 3 loops is already a tough challenge for existing programs.
- New:  $\approx 17$  loops possible (with basic implementation).
- Caveat: Only Euclidean - no Minkowski regime (so far).

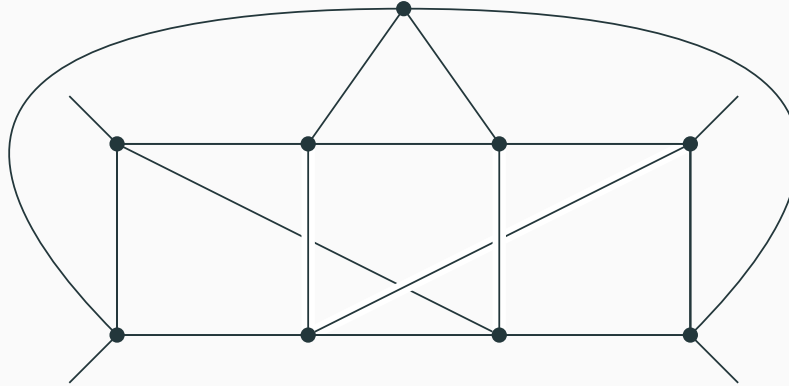
# Interesting example



(Brown-Schwarz)

**Figure 1:** A non-generalized polylog/non-MZV 8-loop  $\varphi^4$ -graph.

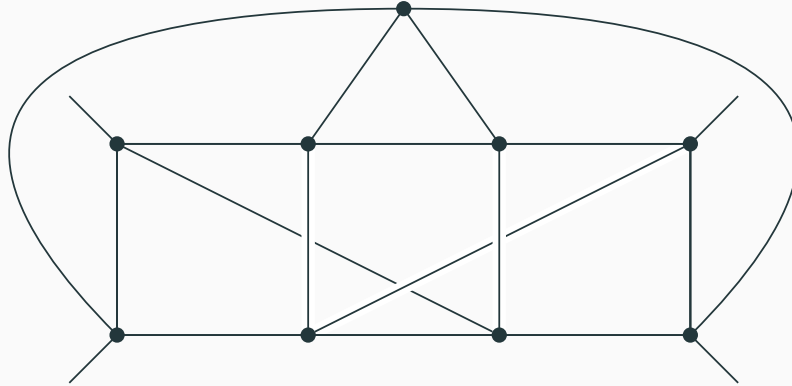
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$$\Gamma(\varepsilon) \int_{\mathbb{P}_{>0}^{E-1}} \frac{1}{\Psi_G(\mathbf{x})^{2-\varepsilon}} \left( \frac{\Psi_G}{\Phi_G} \right)^\varepsilon \Omega \approx \frac{1}{\varepsilon} 422.9610 \cdot (1 \pm 10^{-6}) + \dots$$

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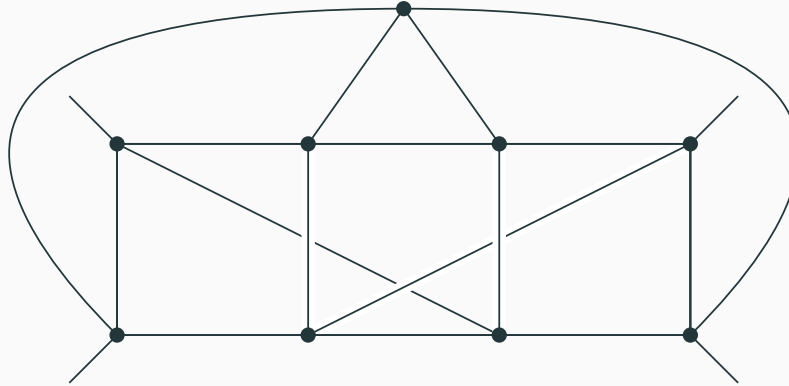


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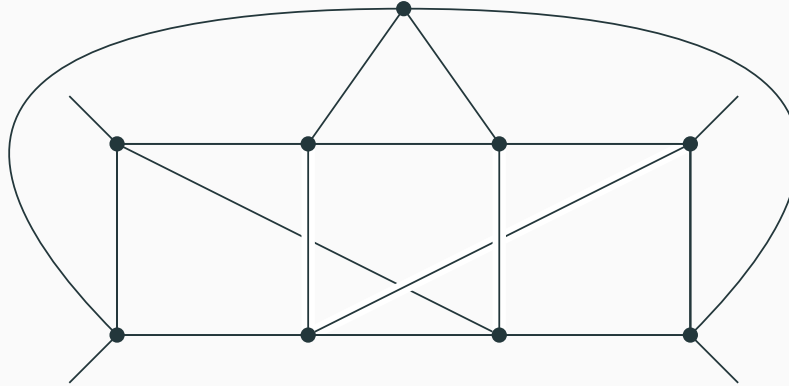


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- $\sim 10$  CPU secs to compute up to  $10^{-3}$ -accuracy at 8 loops.
- $\sim 30$  CPU days to compute up to  $10^{-6}$ -accuracy at 8 loops.
- Higher orders in  $\varepsilon$  can also be computed.



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Minimal runtime to evaluate a Euclidean Feynman integral with  $n$  edges up to  $\delta$ -accuracy is *at most*  $\mathcal{O}(n2^n + n^4\delta^{-2})$

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⇒ Improvements likely!

# Tropicalized Feynman integrals

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# Algebraic geometric motivation

$$\int_{\mathbb{P}_{>0}^{E-1}} \frac{\Omega}{\Psi_G^{D/2}(\mathbf{x})} \left( \frac{\Psi_G(\mathbf{x})}{\Phi_G(\mathbf{x})} \right)^{\omega_G}$$

# Algebraic geometric motivation

$$\int_{\mathbb{P}_{>0}^{E-1}} \frac{\Omega}{\Psi_G^{D/2}(\mathbf{x})} \left( \frac{\Psi_G(\mathbf{x})}{\Phi_G(\mathbf{x})} \right)^{\omega_G}$$

Problem:      Complicated geometry obstructs integration

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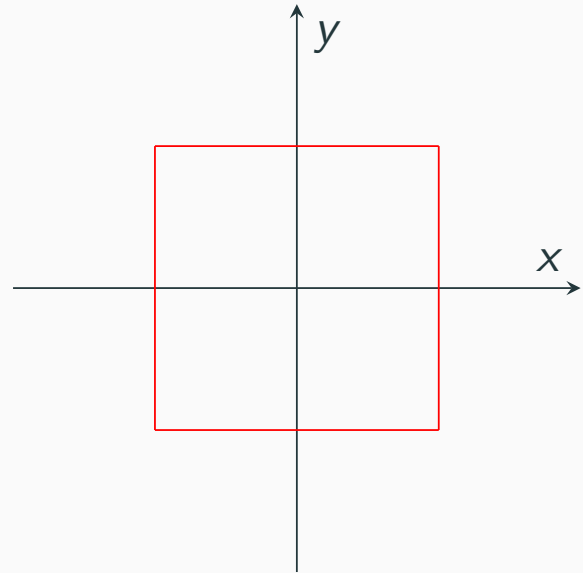
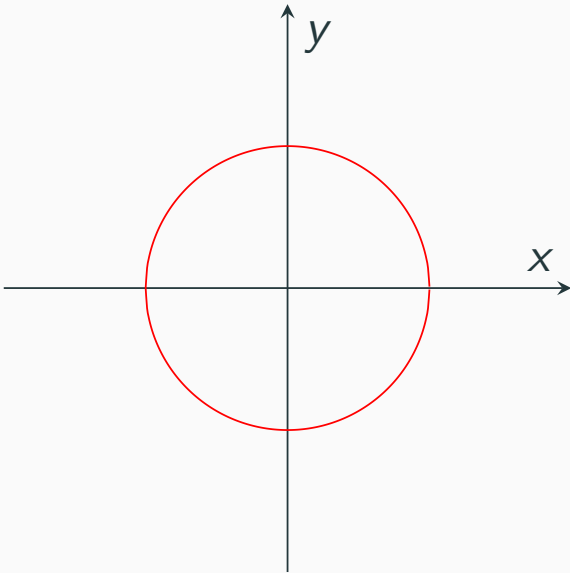
Solution: Simplify the geometry

## Philosophy

Deform geometry to sacrifice smoothness for simplicity.

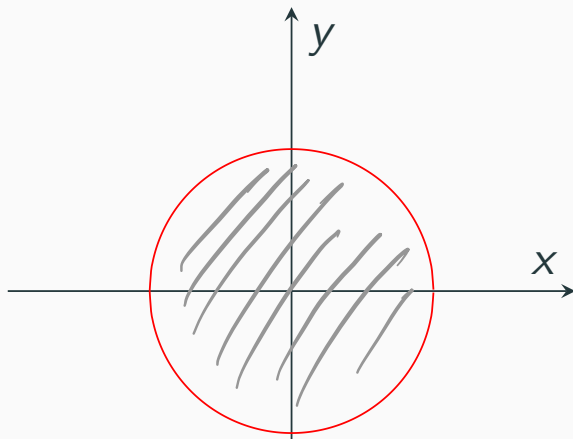
Various applications in algebraic geometry.

$$1 = x^2 + y^2 \quad \rightarrow \quad 1 = (x^2 + y^2)^{\text{tr}} = \max\{x^2, y^2\}$$

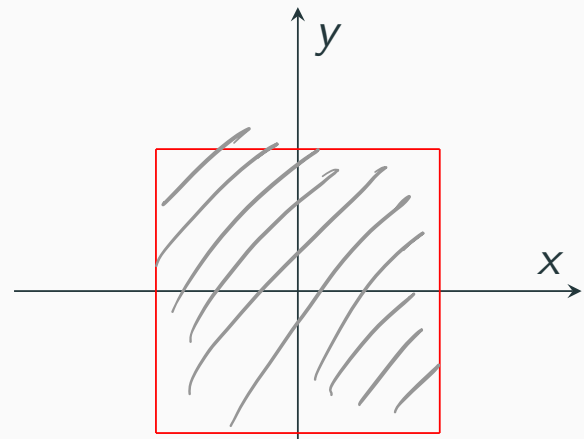




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$$V = \pi$$



$$V = 4$$

rationalization

# Tropical approximation of a polynomial

Let  $p$  be a homogeneous polynomial in  $n$  variables:

$$p(x_1, \dots, x_n) = \sum_{k \in \mathbb{N}_0^n} a_k x^k$$

$\rightarrow \prod_{i=1}^n x_i^{k_i}$

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$$p(x_1, \dots, x_n) = \sum_{k \in \mathbb{N}_0^n} a_k x^k$$

'tropicalize'



$$p^{\text{tr}}(x_1, \dots, x_n) = \max_{\substack{k \in \mathbb{N}_0^n \\ \text{s.t. } a_k \neq 0}} x^k$$

$p^{\text{tr}}$  is the **tropical approximation** of  $p$ .

# The approximation property of the tropicalization

$$p(x_1, \dots, x_n) = \sum_{k \in \mathbb{N}_0^n} a_k x^k; \quad p^{\text{tr}}(x_1, \dots, x_n) = \max_{\substack{k \in \mathbb{N}_0^n \\ \text{s.t. } a_k \neq 0}} x^k$$

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## Theorem (MB 2020)

If  $p(\mathbf{x})$  is completely non-vanishing on  $\mathbb{P}_{>0}^n$ , then

$$C_1 p^{\text{tr}}(\mathbf{x}) \leq |p(\mathbf{x})| \leq C_2 p^{\text{tr}}(\mathbf{x}) \text{ for all } \mathbf{x} \in \mathbb{P}_{>0}^n$$

for some positive constants  $C_1, C_2 > 0$ .

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Proof:

Obvious if  $p(\mathbf{x})$  has only positive coefficients. Otherwise not...

# Application to Feynman graph polynomials

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$$\begin{aligned}\Psi_G &= \sum_T \prod_{e \notin T} x_e & \Rightarrow & \Psi_G^{\text{tr}} = \max_T \prod_{e \notin T} x_e \\ \Phi_G &= \sum_F \|p(F)\|^2 \prod_{e \notin F} x_e & \Rightarrow & \Phi_G^{\text{tr}} = \max_{\substack{F \\ \text{s.t. } \|p(F)\|^2 \neq 0}} \prod_{e \notin F} x_e\end{aligned}$$



Feynman integral:

$$I_G = \int_{\mathbb{P}_{>0}^{E-1}} \frac{\Omega}{(\Psi_G)^{D/2}} \left( \frac{\Psi_G}{\Phi_G} \right)^{\omega_G}$$

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## QFT tropicalization

Replace all instances of  $\Psi$  and  $\Phi$  with their tropicalized versions.

# Approximation property

There are constants  $C_1, C_2 > 0$ , such that

$$C_1 \frac{1}{(\Psi_G^{\text{tr}})^{D/2}} \left( \frac{\Psi_G^{\text{tr}}}{\Phi_G^{\text{tr}}} \right)^{\omega_G} \leq \left| \frac{1}{(\Psi_G)^{D/2}} \left( \frac{\Psi_G}{\Phi_G} \right)^{\omega_G} \right| \leq C_2 \frac{1}{(\Psi_G^{\text{tr}})^{D/2}} \left( \frac{\Psi_G^{\text{tr}}}{\Phi_G^{\text{tr}}} \right)^{\omega_G}$$

for all  $\mathbf{x} \in \mathbb{P}_{>0}^{E-1}$ , if  $\Psi_G$  and  $\Phi_G$  are *completely non-vanishing*.

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for all  $\mathbf{x} \in \mathbb{P}_{>0}^{E-1}$ , if  $\Psi_G$  and  $\Phi_G$  are *completely non-vanishing*.

$$\Rightarrow C_1 I_G^{\text{tr}} \leq |I_G| \leq C_2 I_G^{\text{tr}}$$

$\Rightarrow$  The tropicalized integral gives **both** an upper and a lower bound

# Tropical approach

Evaluating tropicalized Feynman integrals is easy

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- At low order ( $n = 1, \dots, 20$ ):

Tropicalized Feynman integrals are easily calculated exactly

All observables are **rational** numbers/functions.

**Panzer 2019; MB 2020**

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Tropicalized Feynman integrals are easily calculated exactly

All observables are **rational** numbers/functions.

**Panzer 2019; MB 2020**

- When the tropical version is known exactly, numerical integration of the original integrals is just an extra step.

**MB 2020**



# Tropical Numerical integration

---

It is convenient to generalize first (and specify again later):

$$\int_{\mathbb{P}_{>0}^{E-1}} \frac{\Omega}{\Psi_G^{D/2}(\mathbf{x})} \left( \frac{\Psi_G(\mathbf{x})}{\Phi_G(\mathbf{x})} \right)^{\omega_G} \rightarrow \int_{\mathbb{P}_{>0}^{n-1}} \frac{\prod_i p_i(\mathbf{x})^{\mu_i}}{\prod_j q_j(\mathbf{x})^{\nu_j}} \Omega$$

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## Warm-Up

If  $q_j$  are all completely non-vanishing, then

$$\int_{\mathbb{P}_{>0}^{n-1}} \frac{\prod_i p_i^{\text{tr}}(\mathbf{x})^{\mu_i}}{\prod_j q_j^{\text{tr}}(\mathbf{x})^{\nu_j}} \Omega < \infty \iff \int_{\mathbb{P}_{>0}^{n-1}} \frac{\prod_i p_i(\mathbf{x})^{\mu_i}}{\prod_j q_j(\mathbf{x})^{\nu_j}} \Omega < \infty,$$

i.e. tropical convergence is equivalent to ordinary convergence.

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## Strategy

Treat the exact integral as perturbation around the tropical one.

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Tropicalization solves the blowup problem!

The form  $\tilde{\mu}^{\text{tr}}$  is the canonical measure on the common refinement of the normal fans of the Newton polytopes of the  $p_i, q_j$ .



## The tropical form

$$\tilde{\mu}^{\text{tr}} = \frac{\prod_i p_i^{\text{tr}}(\mathbf{x})}{\prod_j q_j^{\text{tr}}(\mathbf{x})} \Omega$$

Sampling from this measure allows numerical integration.

## Theorem (MB 2020)

If the Newton polytopes of  $p_i$  and  $q_j$  are ‘not too complicated’, then there is a (reasonably) fast algorithm to sample the measure

$$\mu^{\text{tr}} = \frac{1}{Z} \frac{\prod_i p_i^{\text{tr}}(\mathbf{x})}{\prod_j q_j^{\text{tr}}(\mathbf{x})} \Omega \quad \text{on } \mathbb{P}_{>0}^{n-1}$$

with  $Z$  chosen such that  $\int_{\mathbb{P}_{>0}^{n-1}} \mu^{\text{tr}} = 1$ .

$$\int_{\mathbb{P}_{>0}^{n-1}} \frac{\prod_i p_i(\mathbf{x})}{\prod_j q_j(\mathbf{x})} \Omega = Z \int_{\mathbb{P}_{>0}^{n-1}} \mu^{\text{tr}} \frac{\prod_i \frac{p_i(\mathbf{x})}{p_i^{\text{tr}}(\mathbf{x})}}{\prod_j \frac{q_j(\mathbf{x})}{q_j^{\text{tr}}(\mathbf{x})}}$$

⇒ Can be evaluated by sampling from the measure  $\mu^{\text{tr}}$ .

(Monte Carlo)

runtime depends on the shape of the polytopes:  $\mathcal{O}(V^2 + \mathcal{V}\delta^{-2})$ .

# Generalized permutahedra

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# Back to Feynman integrals and QFT

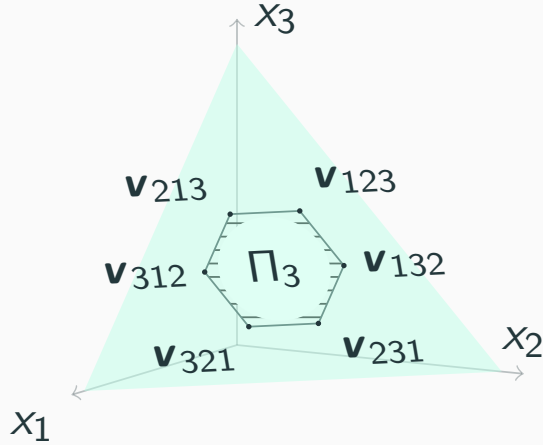
The relevant polytopes that appear in QFT have a special shape.

They are **generalized permutahedra Postnikov 2008**.

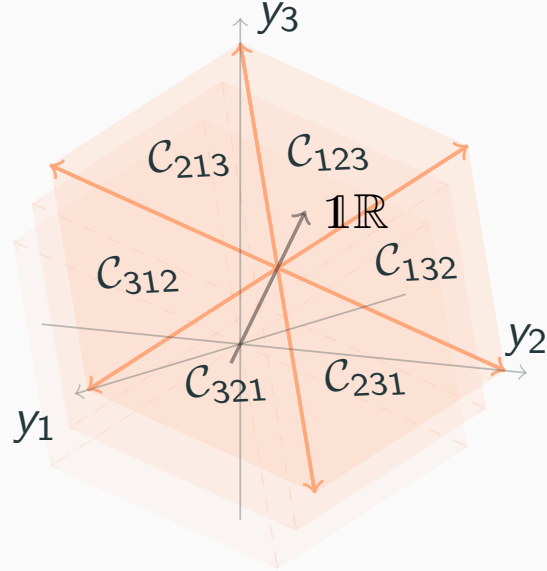
**Theorem (Schultka 2018) using results from (Brown 2015)**

The Newton polytopes of the graph polynomials  $\Psi_G$  and  $\Phi_G$  are **generalized permutahedra**.

( $\Phi_G$  only if the momenta are Euclidean.)



(a) The permutahedron  $\Pi_3 \subset \mathbb{R}^3$ .



(b) Dual of  $\Pi_3$ : The corresponding braid arrangement fan.

This structure follows from the factorization properties of  $\Psi$  and  $\Phi$ .

# A direct and simple proof

Recall that

$$\Psi_G = \sum_T \prod_{e \notin T} x_e \quad (\text{sum over spanning trees})$$

$$\Phi_G = \sum_F \|p(F)\|^2 \prod_{e \notin F} x_e \quad (\text{sum over 2-forests})$$

Classical combinatorial arguments show that,

a given ordering of the edges fixes

a maximal spanning tree  $\Rightarrow \Psi_G$  is a gen. permutahedron

a maximal 2-forest\*  $\Rightarrow \Phi_G$  is a gen. permutahedron

(Also slightly more general than **Brown** and **Schultka**)

# Consequences of the generalized permutahedron property

## Theorem (MB 2020)

If the Newton polytopes of  $p_i$  and  $q_j$  are gen. permutahedra, then there is a (very) fast algorithm to sample from the measure

$$\mu^{\text{tr}} = \frac{1}{Z} \frac{\prod_i p_i^{\text{tr}}(\mathbf{x})}{\prod_j q_j^{\text{tr}}(\mathbf{x})} \Omega \quad \text{on } \mathbb{P}_{>0}^{n-1},$$

with  $Z$  chosen such that  $\int_{\mathbb{P}_{>0}^{n-1}} \mu^{\text{tr}} = 1$ .

Makes heavy use of tools from [Aguiar, Ardila 2017](#).



# Generalized permutahedron sampling algorithm

---

**Algorithm 4** to generate a sample from  $\mu^{\text{tr}}$  for generalized permutahedra

---

Set  $A = [n]$  and  $\kappa = 1$

**while**  $A \neq \emptyset$  **do**

Pick a random  $e \in A$  with probability  $p_e = \frac{1}{J_r(A)} \frac{J_r(A \setminus e)}{r(A \setminus e)}$ .

Remove  $e$  from  $A$ , i.e. set  $A \leftarrow A \setminus e$ .

Set  $\sigma(|A|) = e$ .

Set  $x_e = \kappa$ .

Pick a uniformly distributed random number  $\xi \in [0, 1]$ .

Set  $\kappa \leftarrow \kappa \xi^{1/r(A)}$ .

**end while**

Return  $\mathbf{x} = [x_1, \dots, x_n] \in \text{Exp}(\mathcal{C}_\sigma) \subset \mathbb{P}_{>0}^{n-1}$  and  $\sigma = (\sigma(1), \dots, \sigma(n)) \in S_n$ .

---

*→ easy to implement*

*→ code available on my website*

# Outlook

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# Minkowski singularities

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But the approximation property still works.

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- The generalized permutahedron structure breaks down at singular momentum configurations (IR singularities).
- $\Phi_G$  can vanish in the integration domain  
( $\Rightarrow$  analytic continuation is necessary).

But the approximation property still works.

- Vanishing locus of  $\Phi_G$  for complex  $x$  important.

Q: What topological geometries do appear?

## Further outlook: Amplitudes in Outer space

The  $n$ -th order correction to the scattering amplitude is given by,

$$A_n = \sum_{\substack{\text{graphs } G \\ L_G = n}} \frac{I_G}{|\text{Aut } G|}$$

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$$= \sum_{\substack{\text{graphs } G \\ L_G=n}} \frac{1}{|\text{Aut } G|} \int_{\mathcal{T}_G} \mu_G^{\text{tr}} \left( \frac{\Psi_G^{\text{tr}}}{\Psi_G} \right)^{2-\omega_G} \left( \frac{\Phi_G^{\text{tr}}}{\Phi_G} \right)^{\omega_G}$$



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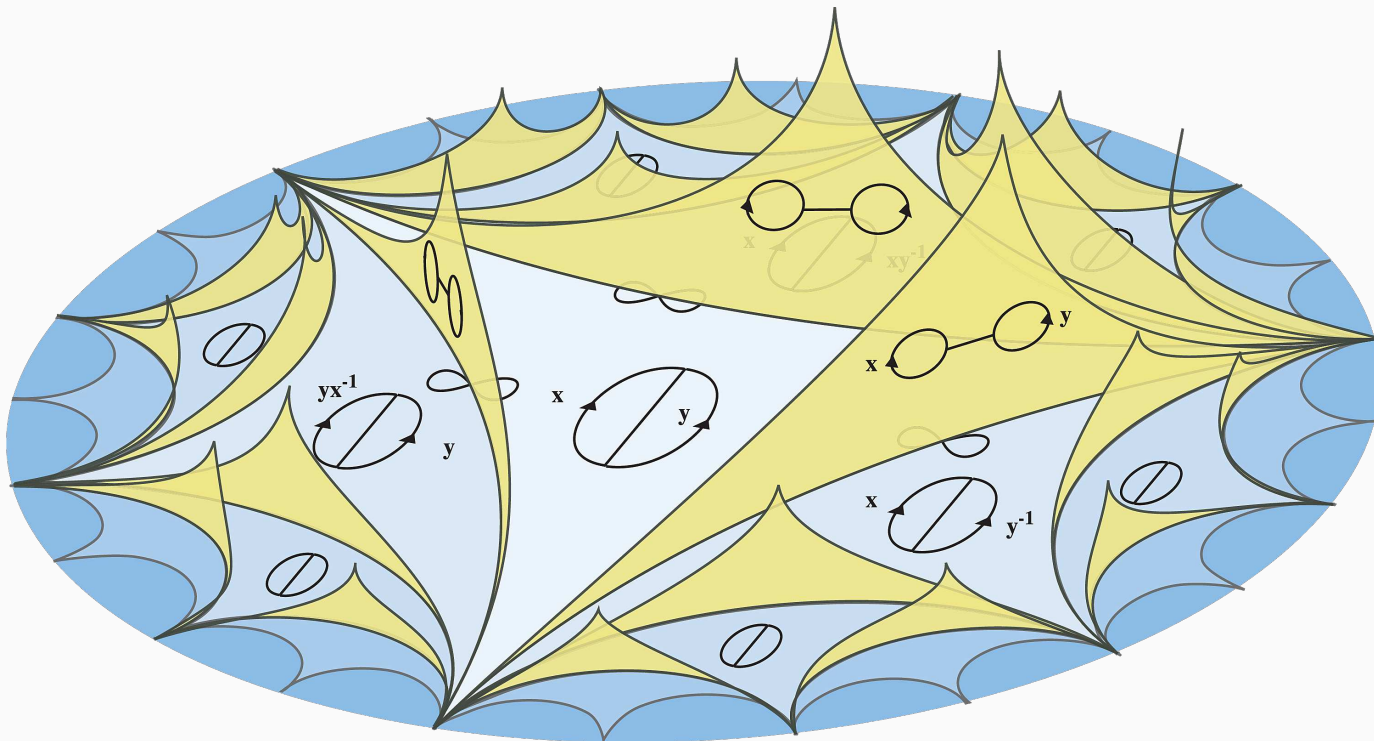
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Integral over version of Outer space/tropical moduli space



**Figure 3:** Outer space  $\mathcal{O}_2$  which is a specific graph orbispace

# Conclusions

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Much left to explore:

- Generalize generalized permutahedra  
 $\Rightarrow$  IR singularities/Minkowski space
- Tropical amplitudes
- Relation to tropical moduli spaces/Outer space?
- Gauge theory?
- ...



# Factorization properties

Generalized permutahedron property follows from factorizations:

For some  $\gamma \subset \Gamma$ , let  $x'_e = \lambda_\gamma x_e$  if  $e \in \gamma$  and else  $x'_e = x_e$ :

$$\Psi'_\Gamma = \lambda_\gamma^{h_1(\gamma)} \Psi_\gamma \Psi_{\Gamma/\gamma} + \mathcal{O}(\lambda_\gamma^{h_1(\gamma)+1}) \text{ as } \lambda_\gamma \rightarrow 0$$

$$\Phi'_\Gamma = \lambda_\gamma^{h_1(\gamma)} \Psi_\gamma \Phi_{\Gamma/\gamma} + \mathcal{O}(\lambda_\gamma^{h_1(\gamma)+1}) \text{ as } \lambda_\gamma \rightarrow 0$$

Degenerate case **Brown 2015**: If  $\Phi_{\Gamma/\gamma} = 0$ , then

$$\Phi'_\Gamma = \lambda_\gamma^{h_1(\gamma)+1} \Phi_\gamma \Psi_{\Gamma/\gamma} + \mathcal{O}(\lambda_\gamma^{h_1(\gamma)+2}) \text{ as } \lambda_\gamma \rightarrow 0.$$

(only with Euclidean momenta)

## Open question

Generalized permutahedra are *universal* with respect to their Hopf monoid structure **Aguiar, Ardila 2017**.

How general are Feynman integrals within this class of polytopes?

# Structures at play

A gen. permutahedron has a *facet presentation* (**Postnikov 2008**):

$$\mathcal{G}_z = \left\{ \mathbf{v} \in \mathbb{R}^n : \sum_{i \in [n]} v_i = z([n]) \text{ and } \sum_{i \in I} v_i \geq z(I) \text{ for all } I \subset [n] \right\},$$

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Complete geometric data is encoded in the boolean function

$$z : \mathbf{2}^{[n]} \rightarrow \mathbb{R}$$

$\Rightarrow$  good control over (tropical) geometry.

## Important: Control over vertices

There is a surjective map from  $S_n$  to the vertices of  $\mathcal{G}_Z$ :

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There is a surjective map from  $S_n$  to the vertices of  $\mathcal{G}_z$ :

Given  $\sigma \in S_n$ , then  $\mathbf{w}^{(\sigma, z)} \in \text{Vert}_{\mathcal{G}_z}$  with

$$w_{\sigma(k)}^{(\sigma, z)} = z(A_k^\sigma) - z(A_{k-1}^\sigma) \text{ for all } k \in [n],$$

where  $A_k^\sigma = \{\sigma(1), \dots, \sigma(k)\} \subset [n] = \{1, \dots, n\}$ .

# Implicit recursive structures

## Computing the gen. permutahedral geometry (MB 2020)

For a boolean function  $r : \mathbf{2}^{[n]} \rightarrow \mathbb{R}$  with  $r(\emptyset) = 1$  and  $r(A) > 0$  for all non-empty  $A \subsetneq [n]$ , we define the boolean function  $J_r : \mathbf{2}^{[n]} \rightarrow \mathbb{R}_{>0}$  recursively as

$$J_r(A) = \sum_{e \in A} \frac{J_r(A \setminus e)}{r(A \setminus e)} \text{ for all non-empty } A \subset [n] \text{ where } J_r(\emptyset) = 1.$$

## Tropicalized integral is a special case (MB 2020)

If  $r(A) = z_{\mathcal{A}}(A) - z_{\mathcal{B}}(A)$  for all non-empty  $A \subsetneq [n]$  and  $r(\emptyset) = 1$ , then  $I^{\text{tr}} = J_r([n])$ .



# Feynman integrand evaluation

$$\frac{1}{\Psi_G(\mathbf{x})^{D/2}} \left( \frac{\Psi_G(\mathbf{x})}{\Phi_G(\mathbf{x})} \right)^{\omega_G}$$

$$\Psi_G(\mathbf{x}) = \sum_T \prod_{e \notin T} x_e$$

$$\Phi_G(\mathbf{x}) = \sum_F \mathbf{p}(F)^2 \prod_{e \notin F} x_e$$

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$$\begin{aligned} \Psi_G(\mathbf{x}) &= \sum_T \prod_{e \notin T} x_e &= \left( \prod_e x_e \right) \det(\tilde{L}) \\ \Phi_G(\mathbf{x}) &= \sum_F \mathbf{p}(F)^2 \prod_{e \notin F} x_e &= \Psi_G \left( \text{Tr}(P^T \tilde{L}^{-1} P) \right) \end{aligned}$$

where  $P_{v,\mu} = p_\mu^{(v)}$  and  $\tilde{L}$  is the **graph Laplacian**.

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where  $P_{v,\mu} = p_\mu^{(v)}$  and  $\tilde{L}$  is the **graph Laplacian**.

Very fast algorithms for such graph Laplacian computations exist (see **Spielman-Teng 2004**).