

EXPONENTIAL DEPTH-HIERARCHY FOR

CONSTANT-DEPTH MULTILINEAR CIRCUITS

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CSE, IITB

MATH, IITB

Multilinear Polynomials

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Multilinear monomial

$$\prod_{i \in S} x_i \quad (S \subseteq [n])$$

$$x^S$$

Equivalent

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Multilinear polynomials: $\sum_S \alpha_S x^S$

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Multilinear monomial

$$\prod_{i \in S} x_i \quad (S \subseteq [n])$$

Equivalent $\sum_S x^S$

Multilinear polynomials: $\sum_S \alpha_S x^S$

E.g.: Determinant, Permanent

Constant-depth Algebraic formulas

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$$P(x_1, \dots, x_n) = \sum_{S \subseteq [n]} \alpha_S x^S$$

$\Sigma \Pi$ formula for P

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$\Sigma\Pi$ formula for P

+ Easy to analyze. Each poly. has a unique $\Sigma\Pi$ formula.

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$\Sigma\Pi$ formula for P

- + Easy to analyze. Each poly. has a unique $\Sigma\Pi$ formula.
- Not succinct.

Constant-depth Algebraic formulas

$$P(x_1, \dots, x_n) = \sum_{i=1}^s \prod_{j=1}^{t_i} L_{ij}$$

linear poly.

Constant-depth Algebraic formulas

$$P(x_1, \dots, x_n) =$$

$$\sum \prod$$

$$\sum_{i=1}^s \prod_{j=1}^{t_i} L_{ij}$$

linear poly.

formula

Constant-depth Algebraic formulas

$$P(x_1, \dots, x_n) =$$

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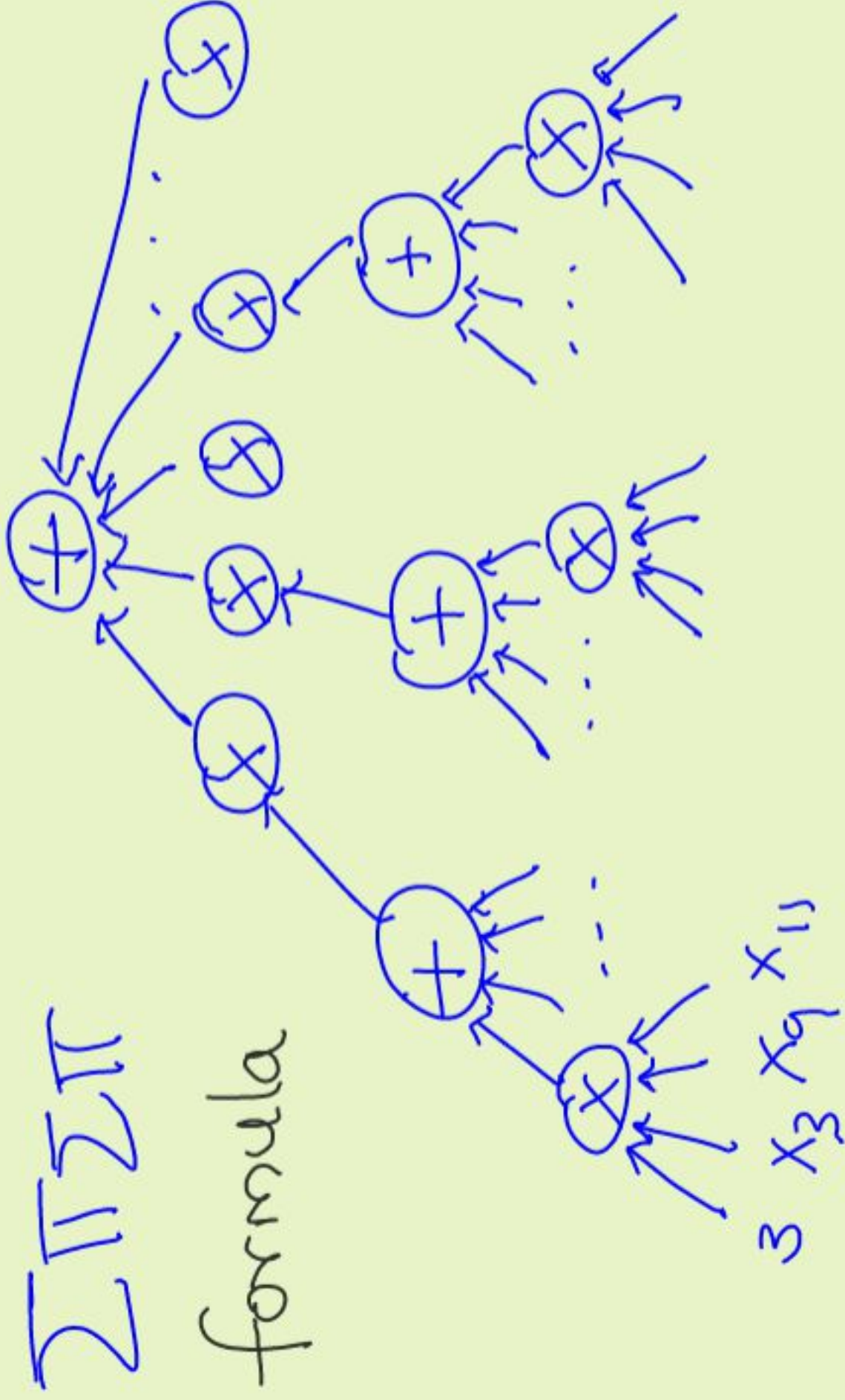
$\sum \prod \sum$

linear poly.

formula

$$\sum \prod \sum \prod, \sum \prod \sum \prod \sum, \dots$$

Constant-depth Algebraic formulas



Constant-depth Algebraic formulas

$\Sigma\Pi$, $\Sigma\Pi\Sigma$, $\Sigma\Pi\Sigma\Pi$, $\Sigma\Pi\Sigma\Pi\Sigma$, \dots

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More nesting \Rightarrow Smaller expressions

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More nesting \Rightarrow Smaller expressions

Nesting measured by Product-Depth

Product-Depth = # of Π 's in
name of formula
class.

Constant-depth Algebraic formulas

$\Sigma \Pi, \Sigma \Pi \Sigma, \Sigma \Pi \Sigma \Pi, \Sigma \Pi \Sigma \Pi \Sigma, \dots$
Product depth 1 Product depth 2

More nesting \Rightarrow Smaller expressions

Nesting measured by Product-Depth

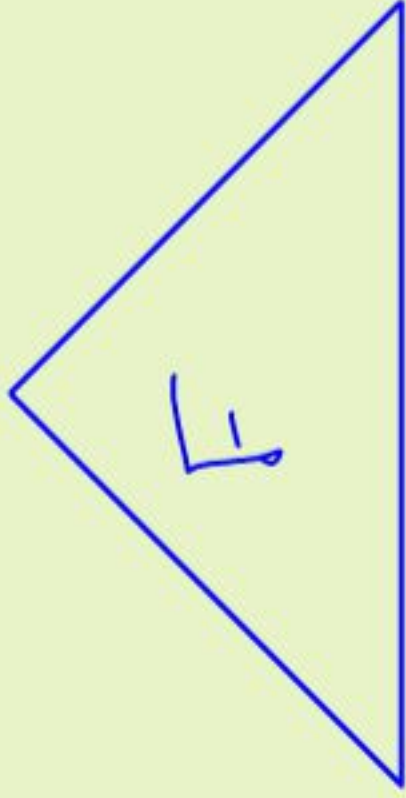
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The question

How much succinctness is gained by increase in product - depth?

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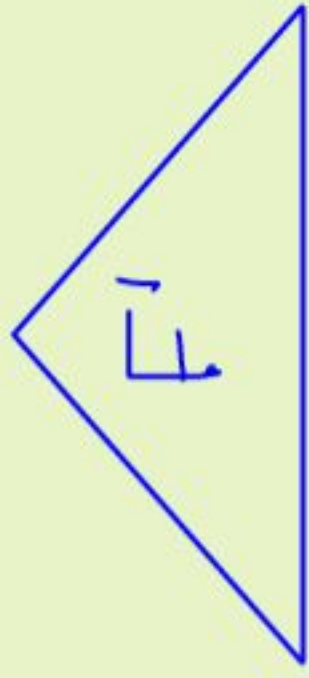
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P. depth Δt formula
of size s

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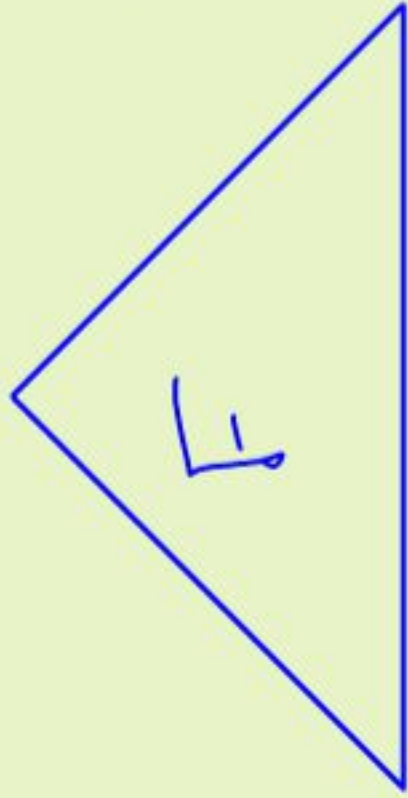
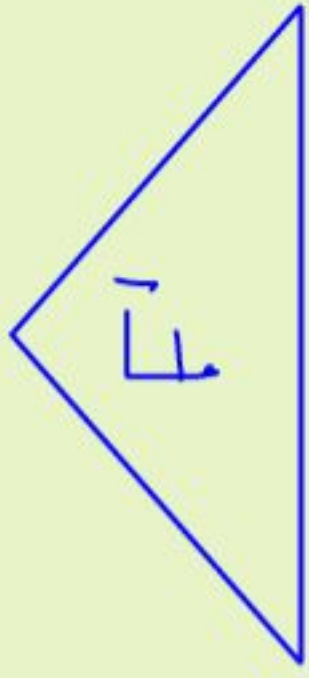
P.depth Δ formula
of size $s' = f(s)$



P.depth $\Delta + 1$ formula
of size s

The question

How much succinctness is gained by increase in product-depth?



P. depth Δ formula
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P. depth $\Delta + 1$ formula
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Worst-case blowup?

Reducing P. depth by 1

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Distribute X over t .

$$\prod_{i=1}^t \left(\sum_j T_{ij} \right) \xrightarrow[\text{blowup}]{\text{exp}(x)} \sum_{j_1, \dots, j_t} T_{1j_1} T_{2j_2} \dots T_{tj_t}$$

Reducing P. depth by 1

Distribute X over t .

$$\prod_{i=1}^{\ell} (\sum_j T_{ij}) \xrightarrow{\text{blowup}} \sum_{j_1, \dots, j_{\ell}} T_{1j_1} T_{2j_2} \dots T_{\ell j_{\ell}}$$

$\text{exp}(x)$

$$\text{Depth } \Delta+1: \sum \Pi \sum \dots \underbrace{\sum \Pi \sum \Pi \dots \sum \Pi \sum}$$

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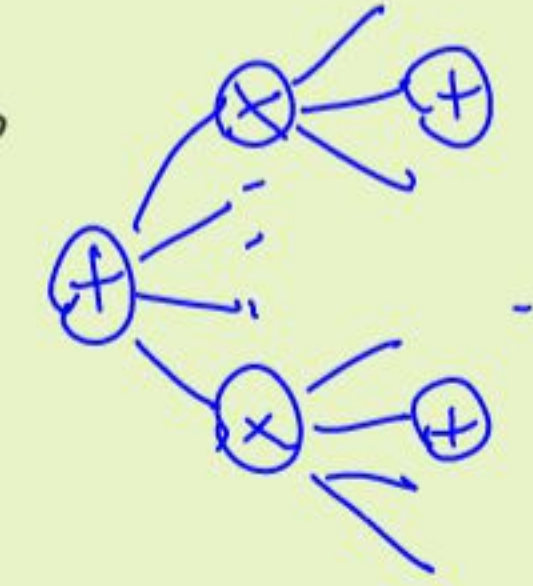
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Reducing P. depth by 1

Products of small fan-in?

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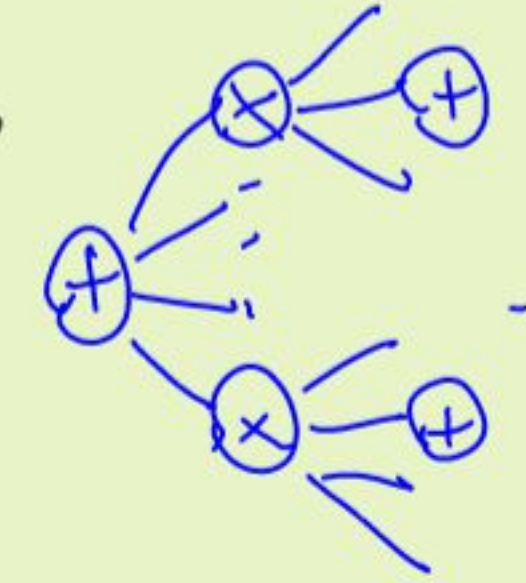


vertices
& p. depth $\Delta+1$



Reducing p. depth by 1

Products of small fan-in?



\approx vertices
& p. depth $\Delta+1$
 \Downarrow
 X_s of fan-in
 $\leq \approx^{1/\Delta}$



Reducing P. depth by 1

Fact: P. depth $\Delta + 1$, size n
↓
P. depths Δ , size $\exp(\frac{1}{\Delta} + o(1))$

Reducing P. depth by 1

Fact: P. depth $\Delta+1$, size s
P. depth Δ , size $\exp(s^{1/\Delta+o(1)})$

Δ	Blowup
1	$\exp(s)$
2	$\exp(\sqrt{s})$
3	$\exp(s^{1/3})$
\vdots	\vdots

Reducing P. depth by 1

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Is this exponential
blowup necessary?

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p. depth 1

Multilinear formulas

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All subexpressions compute multilinear polynomials.

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Eg:

$$(x_1 + x_2 + x_3)^2 - x_1^2 - x_2^2 - x_3^2 = 2x_1x_2 + 2x_2x_3 + 2x_3x_1$$

Non-multilinear $\Sigma \Pi$ formula

Multilinear

$\Sigma \Pi$ formula

Multi linear formulas

Why multi linear?

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→ Natural restriction on expressions
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- Wealth of techniques, starting with work of Raz (2004).

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Qn: Multilinear separation plus depths Δ & $\Delta + 1$?

Previous work on multilinear computation

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- Raz-Yehudayoff (2009): Exponential lower bounds for constant p -depth
- Raz (2005), Raz-Shpilka-Yehudayoff (2007), Hruběš-Yehudayoff (2009), Div et al. (2012), Kayal-Nair-Saha (2015) Kumar-Volk (2017)...

Raz-Yehudayoff

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of P is $S_\Delta = \exp(\Theta(n^{1/\Delta}))$.

Δ	S_Δ
1	$\exp(n)$
2	$\exp(\sqrt{n})$
3	$\exp(n^{1/3})$
⋮	⋮

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→ Quasipolynomial depth hierarchy.

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Raz-Yehudayoff

→ Quasi-polynomial separation blue
p.depth Δ and p.depth $\Delta+1$

Raz- Yehudayoff

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- Kayal-Nair-Saha (2015):
Improvement to Exponential Separation
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Raz- Yehudayoff

- Quasi polynomial separation blue
- p.depth Δ and p.depth $\Delta+1$
- Kayal-Nair-Saha (2015):
Improvement to Exponential Separation
-tion when $\Delta=1$.
- Our result: Similar improvement
for all constant Δ .

Main theorem

Thm: For any const. Δ , there is an explicit polynomial $P_{\Delta+1} \in \mathbb{F}[x_1, \dots, x_n]$ s.t.:

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② Any multilinear F of p. depth $\leq \Delta$ for $P_{\Delta+1}$ must have size $\exp(n^{\alpha_\Delta})$.

$$\alpha_\Delta = \frac{1}{O(\Delta)}.$$

Digression into Boolean Setting

→ Boolean setting: $\Sigma \rightarrow V, \Pi \rightarrow \wedge$

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→ Hastad (1986): Exponential depth-hierarchy.

Constructing the hard

polynomial

Small formula of P -depth

$\Delta+1$

$P_{\Delta+1}$

Large formulas of P -depth

$\leq \Delta$

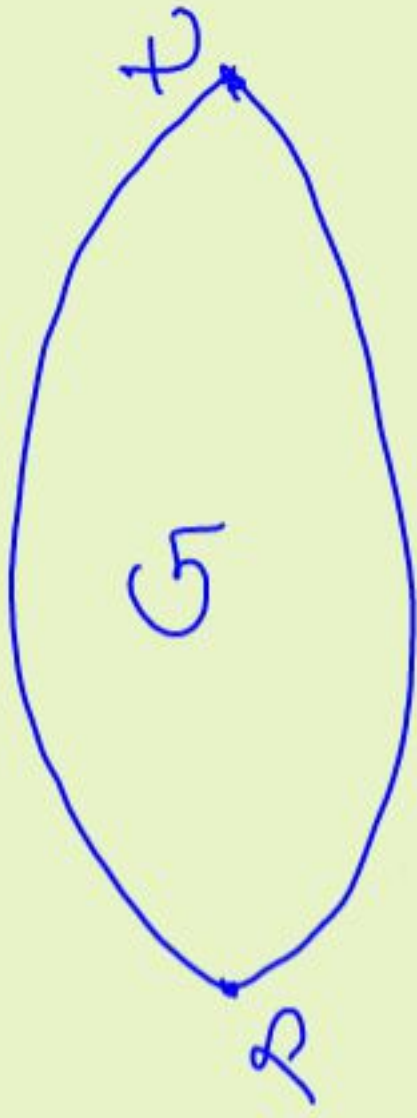
Graph Polynomials (ABPs)

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G - Directed Acyclic

Graph (DAG)

Source s , Sink t



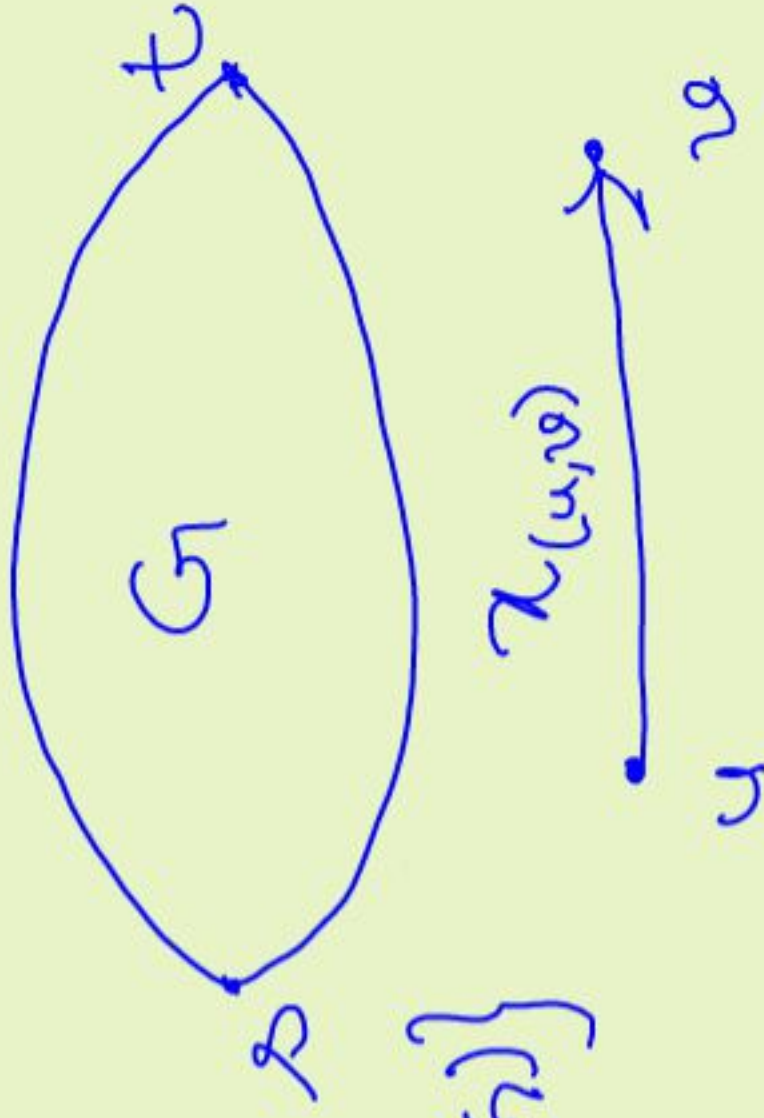
Graph Polynomials (ABPs)

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Variables $X = \{x_e \mid e \in E(G)\}$

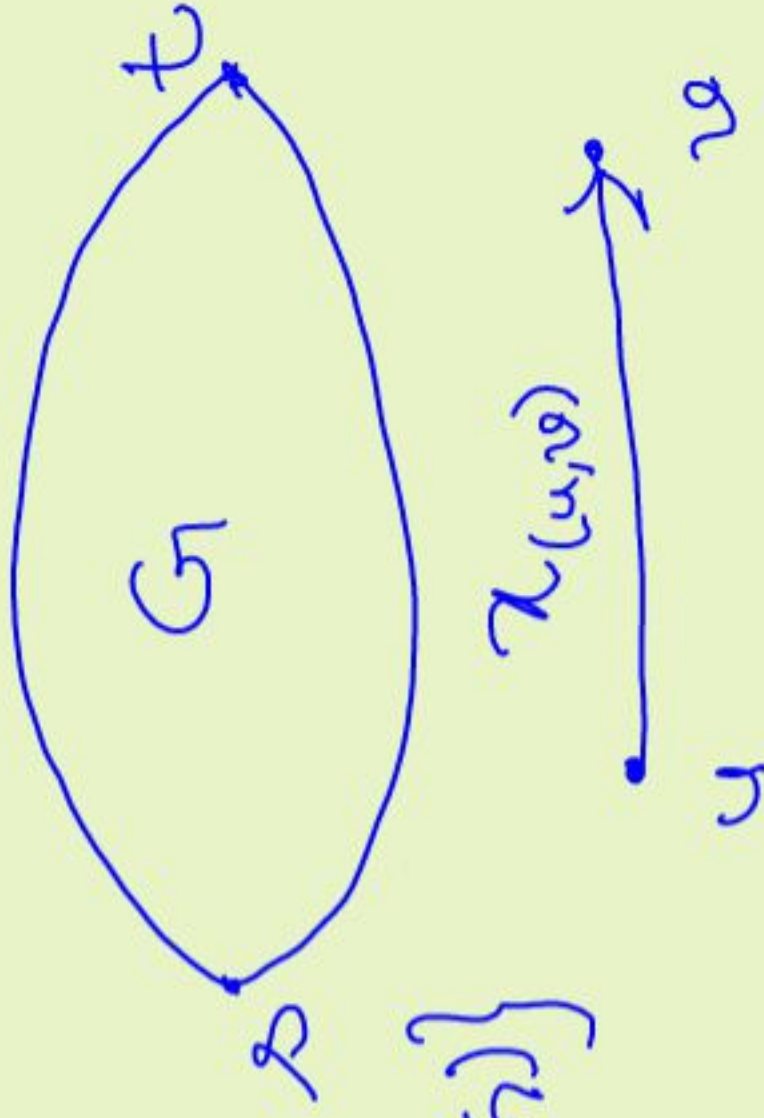


Graph Polynomials (ABPs)

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Source s , Sink t



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$$P_G = \sum_{\pi: s \rightsquigarrow t}$$

$\pi = (u_0 = s, \dots, u_l = t)$

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$$x_{u_0, u_1} \cdot x_{u_1, u_2} \cdot \dots \cdot x_{u_{l-1}, u_l}$$

Graph Polynomials (ABPs)

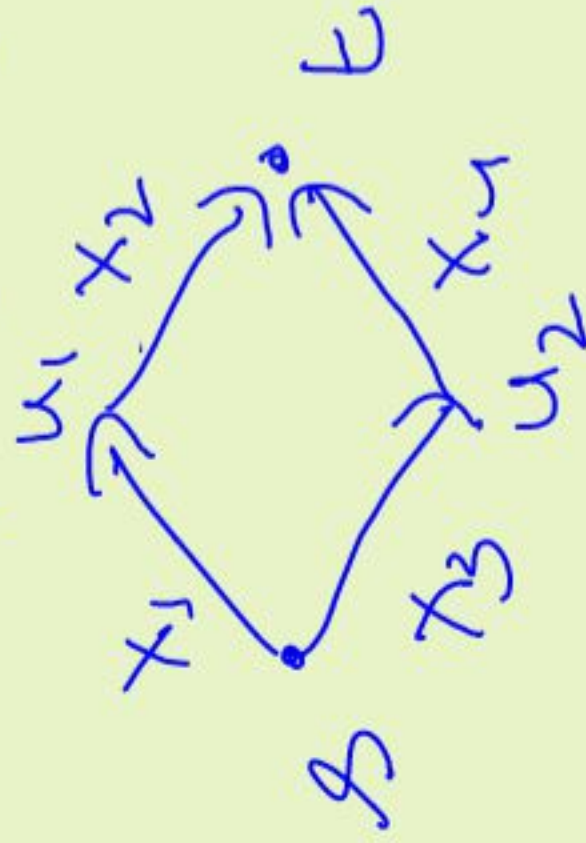


$$P_G = x_1 x_2 x_3 x_4$$

Graph Polynomials (ABPs)



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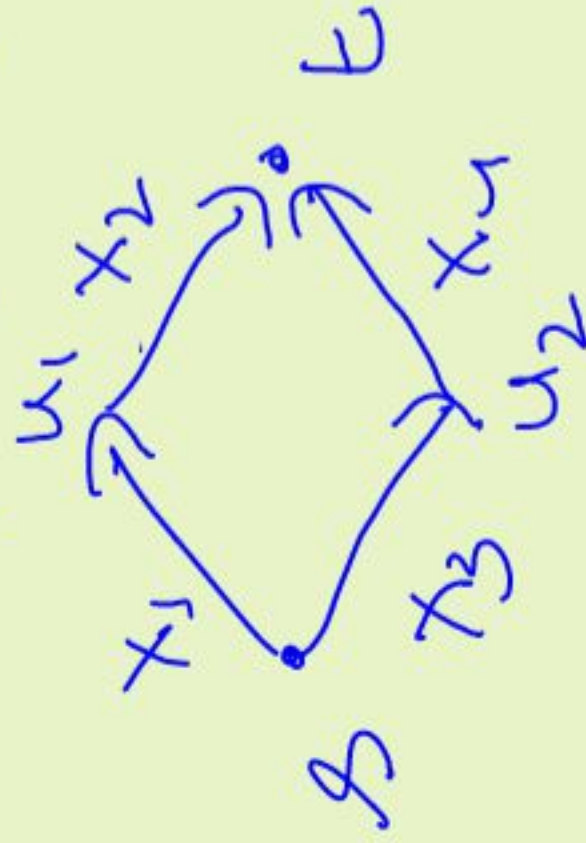


$$P_G = x_1 x_2 + x_3 x_4$$

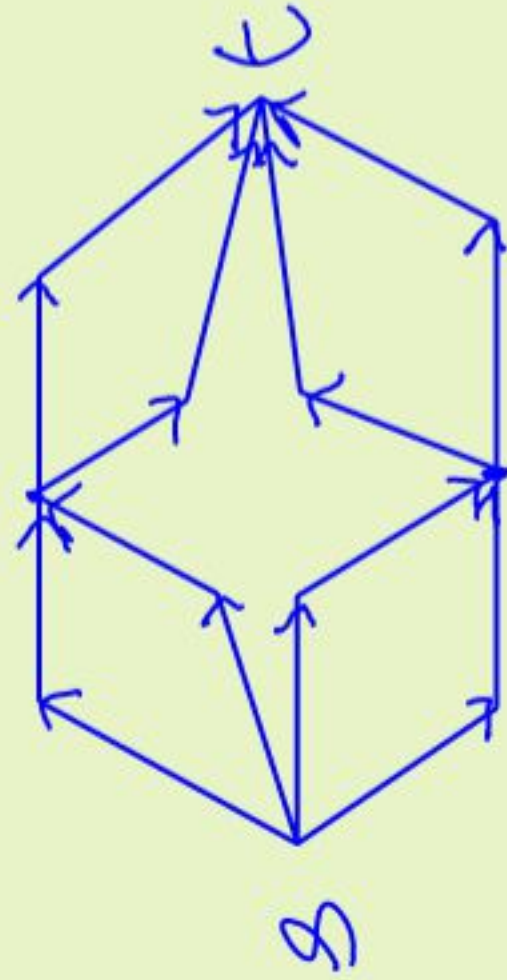
Graph Polynomials (ABPs)



$$P_G = x_1 x_2 x_3 x_4$$



$$P_G = x_1 x_2 + x_3 x_4$$



$$P_G = \text{sum over 8 paths of length 4}$$

Composing graphs

G_1, \dots, G_n - Labelled DAGs

Composing graphs

G_1, \dots, G_r - Labelled DAGs

Series composition



Composing graphs

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Series composition



Parallel composition



Composing graphs

G_1, \dots, G_r - Labelled DAGs

Series composition



$$P_{\text{ser}} = P_{G_1} \cdot P_{G_2} \cdot \dots \cdot P_{G_r}$$

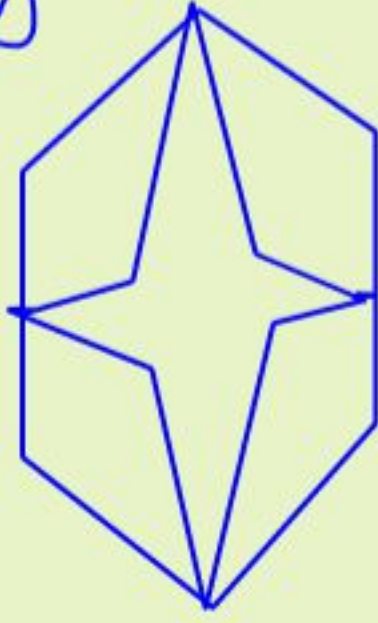
Parallel composition



The hard polysomials (Chen et al. 2016)

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G_1

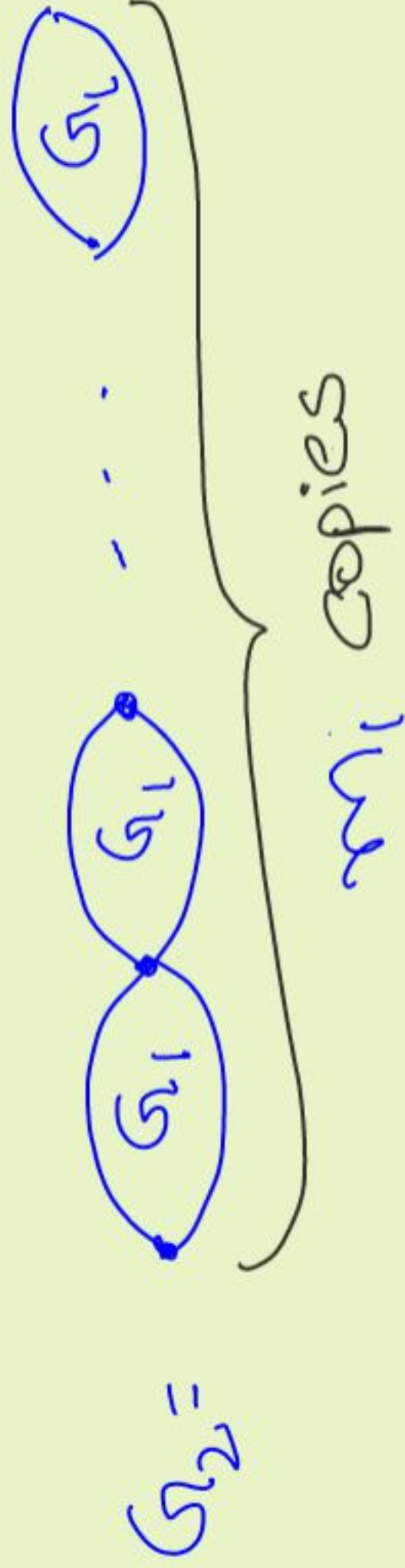
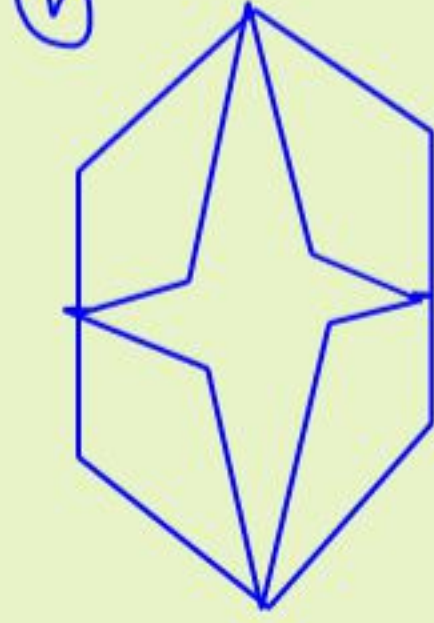


P_1 computed by $\Sigma \Pi$

formula of size $O(n)$.

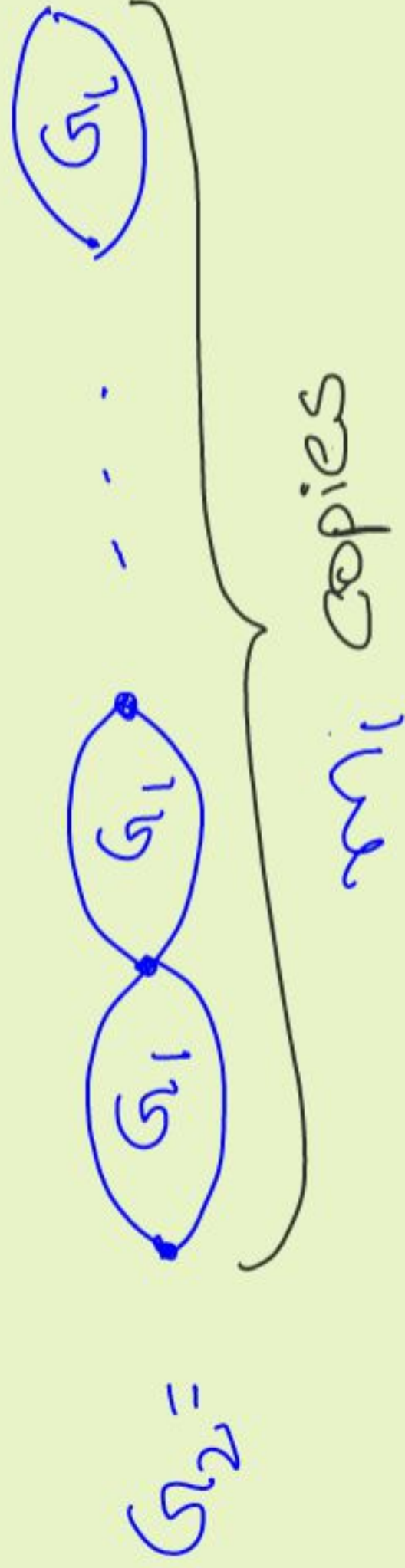
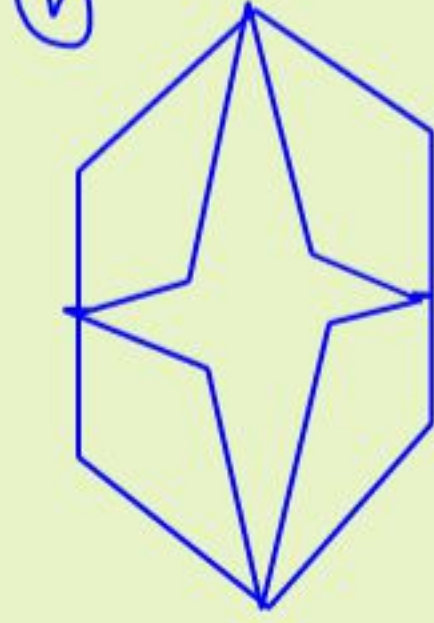
The hard polynomials (Chen et al. 2016)

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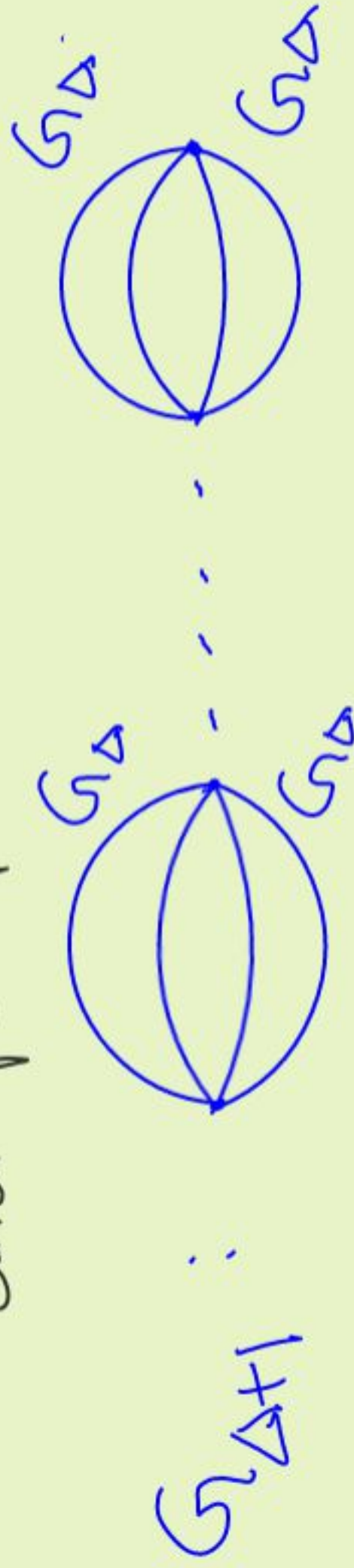
P_2 computed by $\Pi\Sigma\Pi$ formula of size $O(m_1)$.

The hard polynomials (Chen et al. 2016)

$G_\Delta: P_\Delta$ on n_Δ variables that can be
computed by F_Δ of size $O(n_\Delta)$
and p. depth Δ .

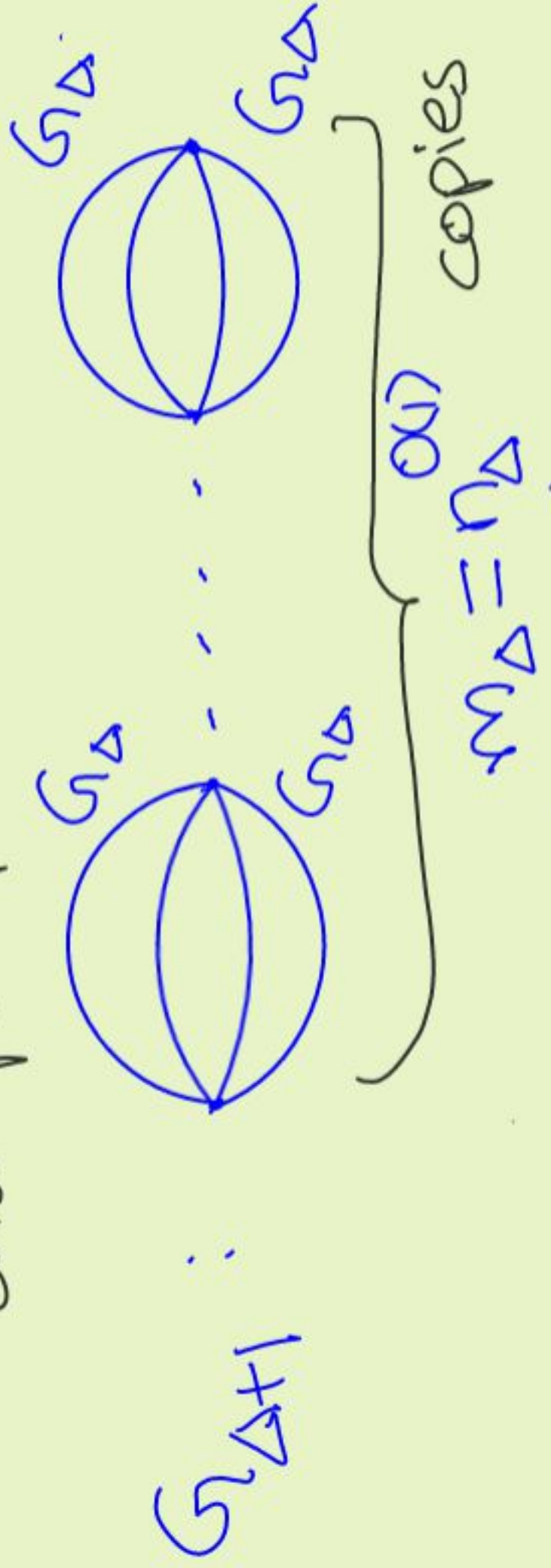
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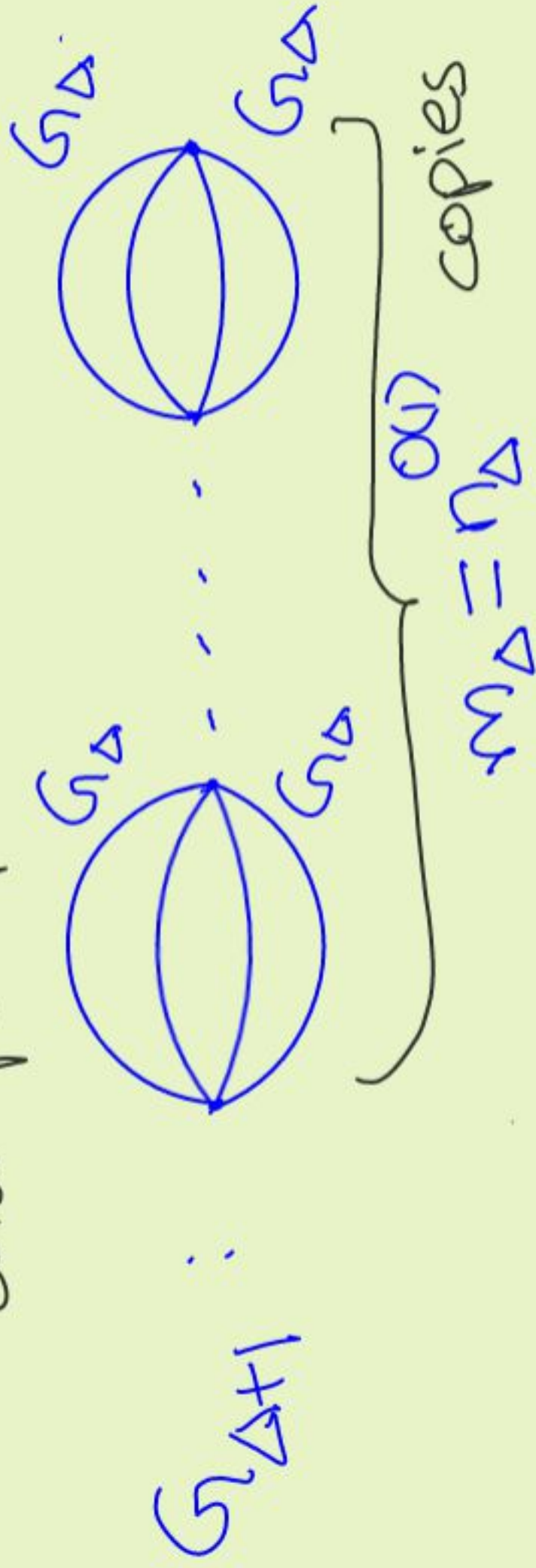
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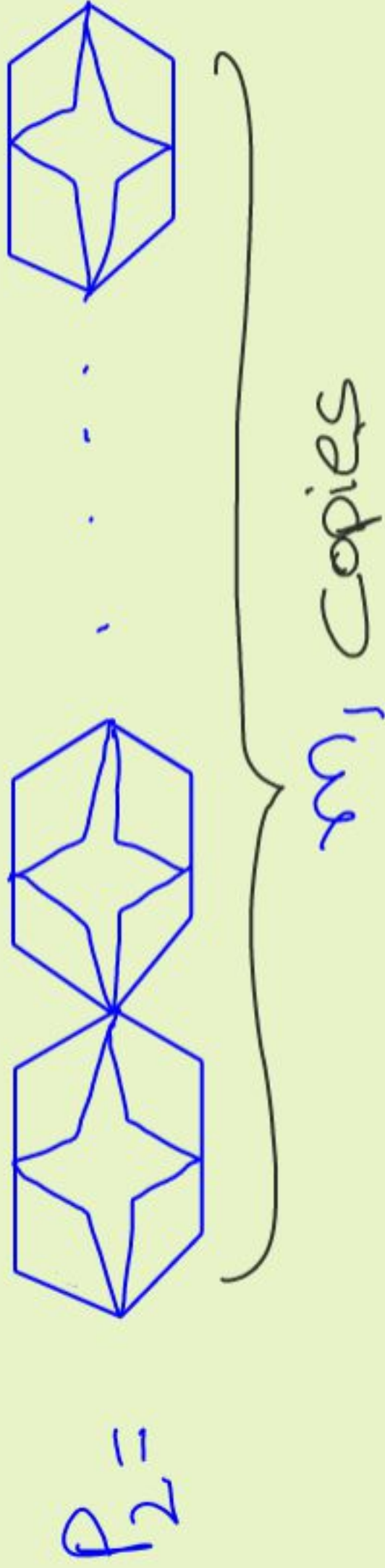
$$P_{\Delta+1} = \prod_{i=1}^{m_\Delta} (P_\Delta^{(i,1)} + P_\Delta^{(i,2)})$$

Main thm (gestated)

Any p-depth Δ formula for $f_{\Delta+1}$
has size $\exp(\Omega(m, \Delta)) = \exp(n^{\Omega(\Delta)})$

Some Proof Details ($\Delta=1$)

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Thm: Any $\sum \Pi \Sigma$ formula for P_2 must
have size $\exp(\Omega(m))$.

Raz's rank measure

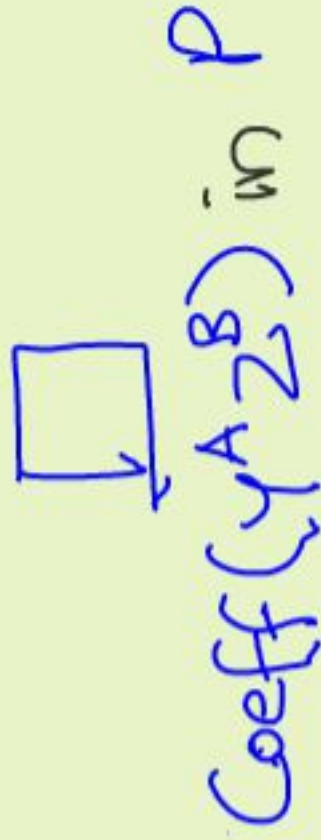
PEFF[yuz]

Raz's rank measure

$$P \in \mathbb{F}[y, u, z]$$
$$M_p \in \mathbb{F}^{2^{|y|} \times 2^{|z|}}$$

B
 Z

A



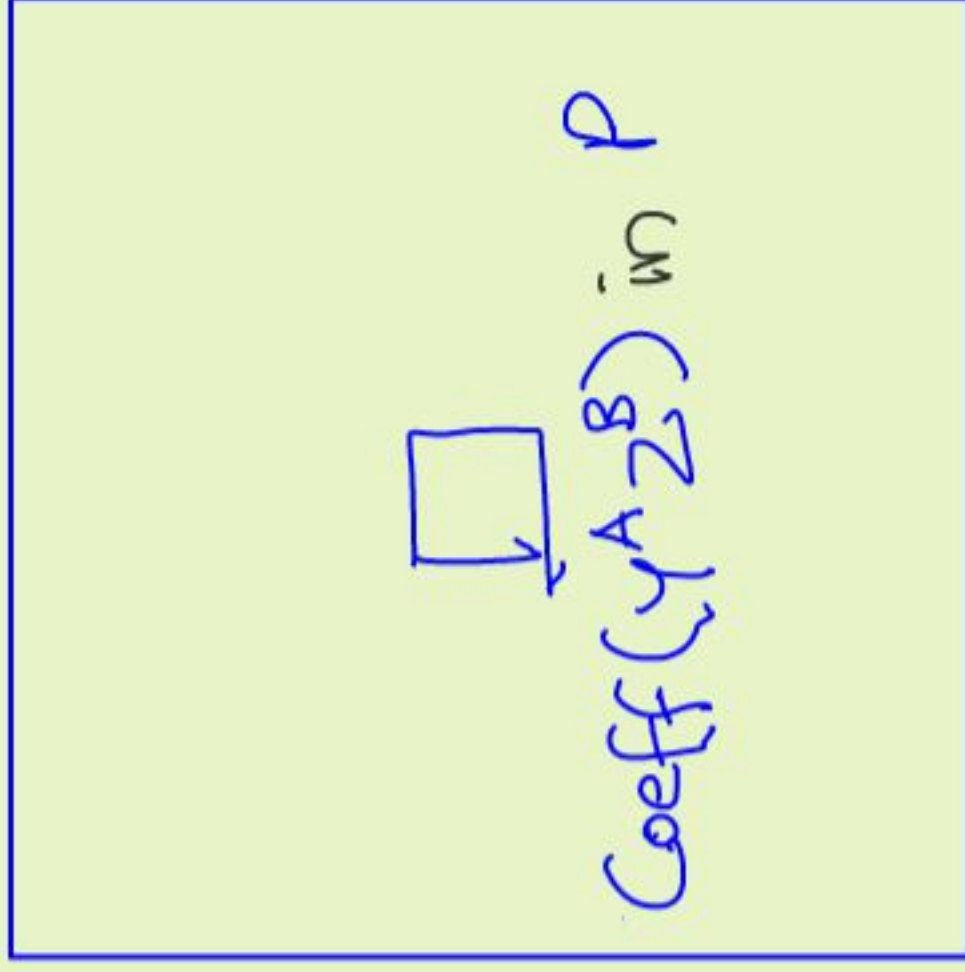
Coeff($y^A z^B$) in P

Raz's rank measure

$$P \in \mathbb{F}^{2^{|Y|} \times 2^{|Z|}}$$
$$M_P \in \mathbb{F}^{2^{|Y|} \times 2^{|Z|}}$$

$$\mu(P) = \text{rank}(M_P)$$
$$\leq 2^{\min\{|Y|, |Z|\}}$$

\downarrow



Raz's rank measure

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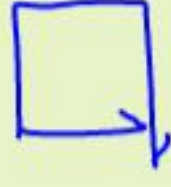
$$M_p =$$

$$\mu(p) = \text{rank}(M_p)$$

$$\leq 2^{\min\{|Y|, |Z|\}}$$

Idea: $\mu(\cdot)$ small for small formulas but large for hard polynomial.

B
Z



Coeff $(Y^A Z^B)$ in P

A simple example

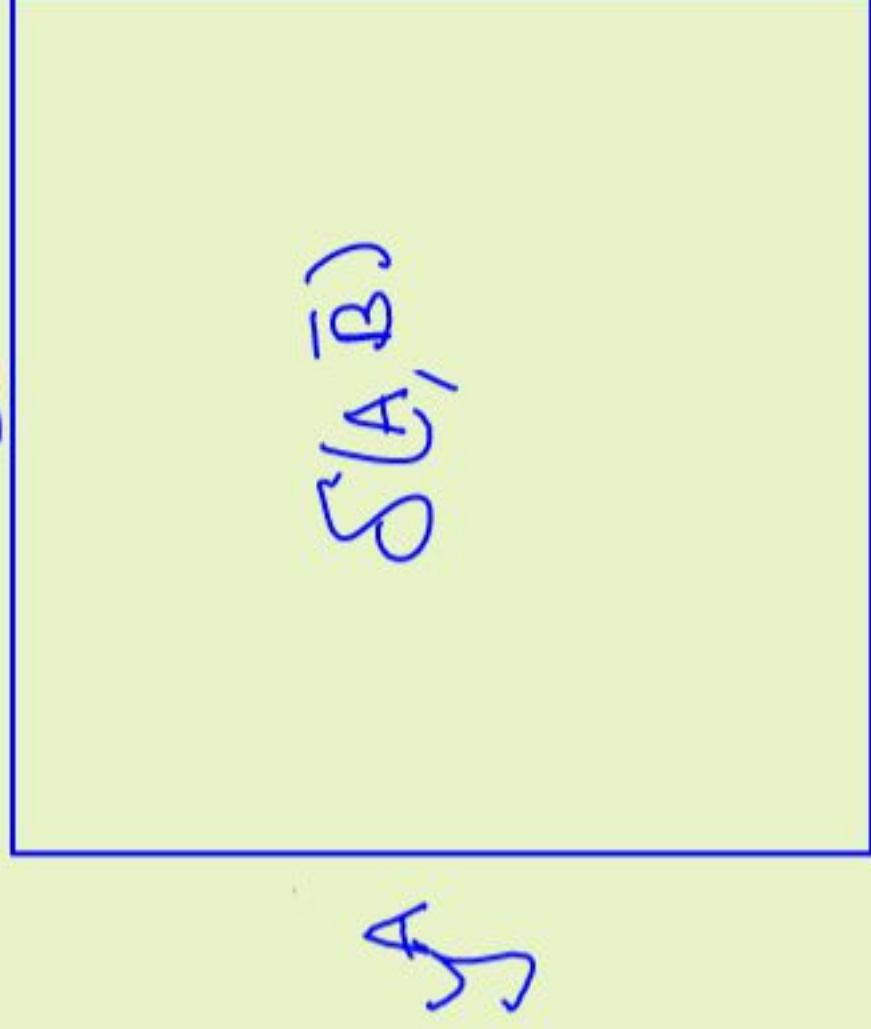
$$\begin{aligned} P(y_1, \dots, y_m, z_1, \dots, z_m) &= \prod_{i=1}^m (y_i + z_i) \\ &= \sum_{A \subseteq [m]} y^A z^{\bar{A}} \end{aligned}$$

A simple example

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$$= \sum_{A \subseteq [m]} y^A z^{\bar{A}}$$

$$\mu(P) = Z^m.$$



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\mathcal{F}

$$\mathcal{S}(A, \bar{B})$$

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Fix: but not in a
"robust" way.

$$\mathcal{S}(A, \bar{B})$$

Restriction

Variable set X

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$f: X \rightarrow YUZEUF$

Restriction

Variable set X

$f: X \rightarrow YUZYUF$

$f \in \# [X]$

substitution \rightarrow

$f|_f \in \# [YUZY]$

Restriction

Variable set X

$f: X \rightarrow YUZYUF$

$P \in \# [X]$ substitution \rightarrow

$P|_P \in \# [YUZY]$

Design a restriction P such that

$\rightarrow \mu(P|_P)$ large,

Restriction

Variable set X

$P: X \rightarrow Y \cup Z \cup F$

$P \in \# [X]$ substitution $P|_P \in \# [Y \cup Z]$

Design a restriction P such that

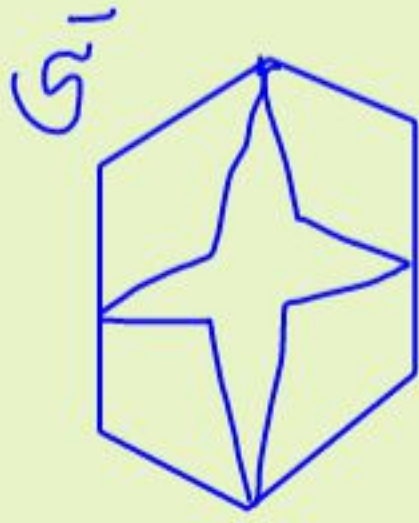
$\rightarrow \mu(P|_P)$ large,

$\rightarrow \mu(F|_P)$ small for any small $\Sigma \Pi \Sigma$

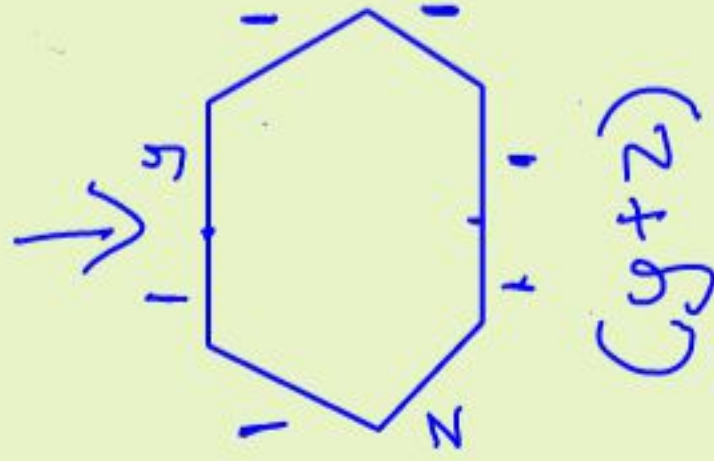
formula

Random Restriction

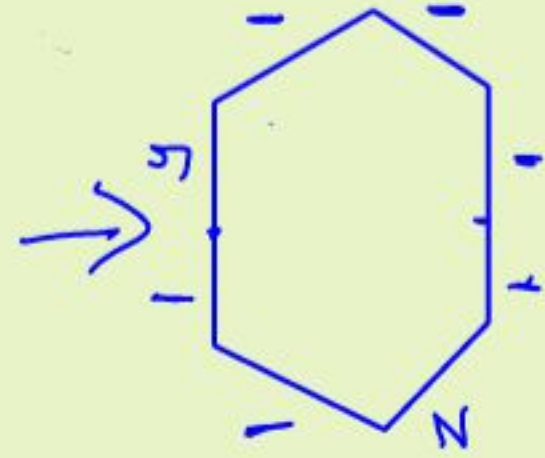
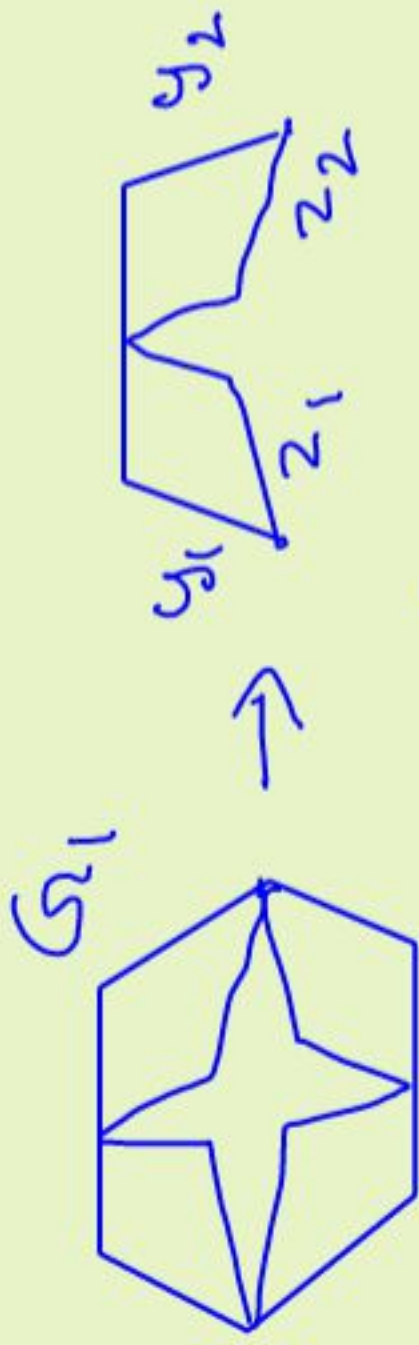
$$f: X \rightarrow Y \cup Z \cup \{0, 1\}$$



Random Restriction $f: X \rightarrow Y \cup Z \cup \{0, 1\}$



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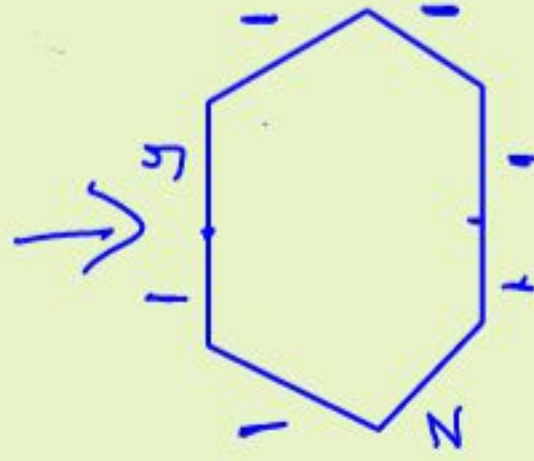
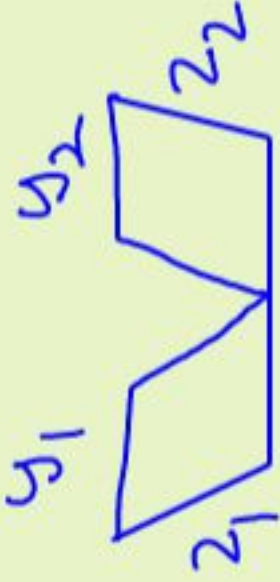
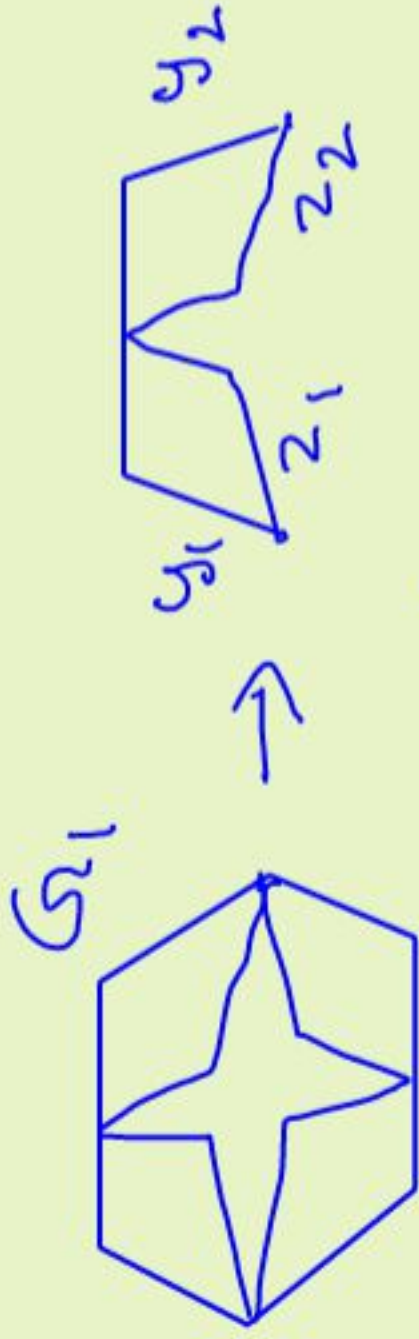


$$(y_1 + z_1) (y_2 + z_2)$$

$$(y + z)$$

Random Restriction

$$f: X \rightarrow Y \cup Z \cup \{0, 1\}$$



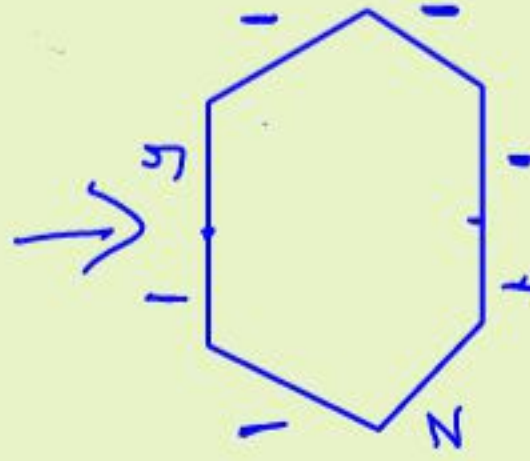
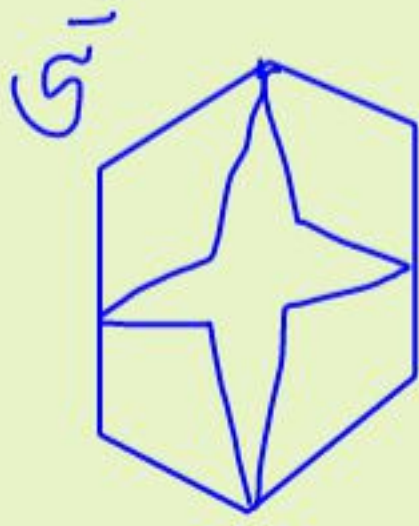
$$(y_1 + z_1)(y_2 + z_2)$$

$$(y + z)$$

Choose a uniform
random option
for each G_1 .

Random Restriction

$$P: X \rightarrow Y \cup Z \cup \{0, 1\}$$



Choose a uniform random option for each G_1 .



$$(y_1+z_1)(y_2+z_2)$$

$$P_{2|P} = \prod_{i=1}^m (y_i+z_i)$$

full rank w.p. 1.

Effect on $\sum \Pi \Sigma$ formula

$$F = \sum_{i=1}^8 \prod_{j=1}^4 L_{i,j} (X_{i,j}) , \quad X = \prod_j X_{i,j}$$

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$$F = \sum_{i=1}^8 \prod_{j=1}^4 L_{i,j} (X_{i,j}), \quad X = \bigcup_j X_{i,j}$$

→ Each linear poly. is "rank-deficient"
with constant probability.

Effect on $\sum \Pi \Sigma$ formula

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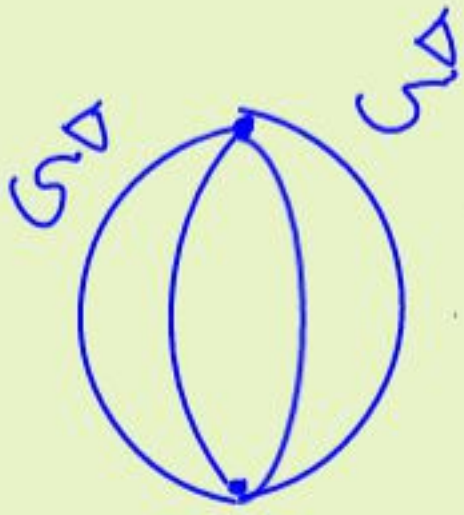
→ Each product term is highly "rank-deficient" with high probability

Effect on $\sum \Pi \Sigma$ formula

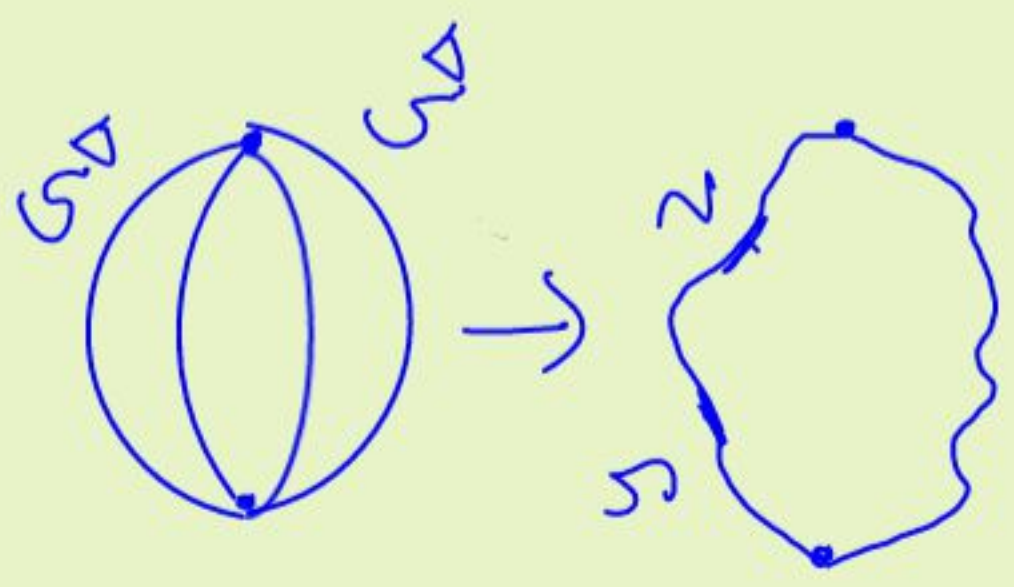
$$F = \sum_{i=1}^8 \prod_{j=1}^4 L_{i,j} (X_{i,j}), \quad X = \bigcup_j X_{i,j}$$

- Each linear poly. is "rank-deficient" with constant probability. "rank-deficient"
- Each product term is highly "rank-deficient" with high probability
- Subadditivity of rank \Rightarrow # of terms must be large.

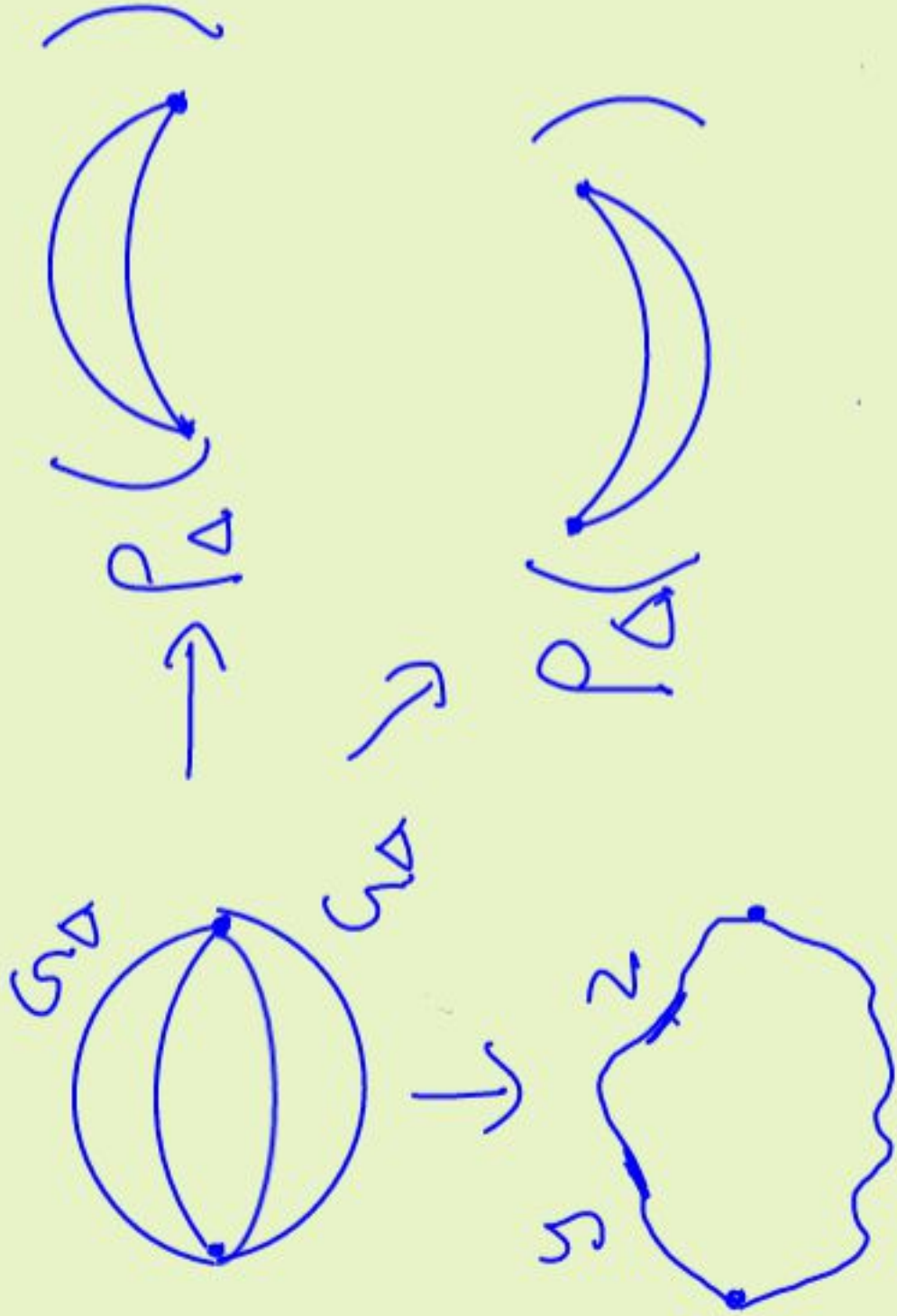
Larger depths $P_{\Delta+1}: X \rightarrow Y \cup Z \cup \{0, 1\}$



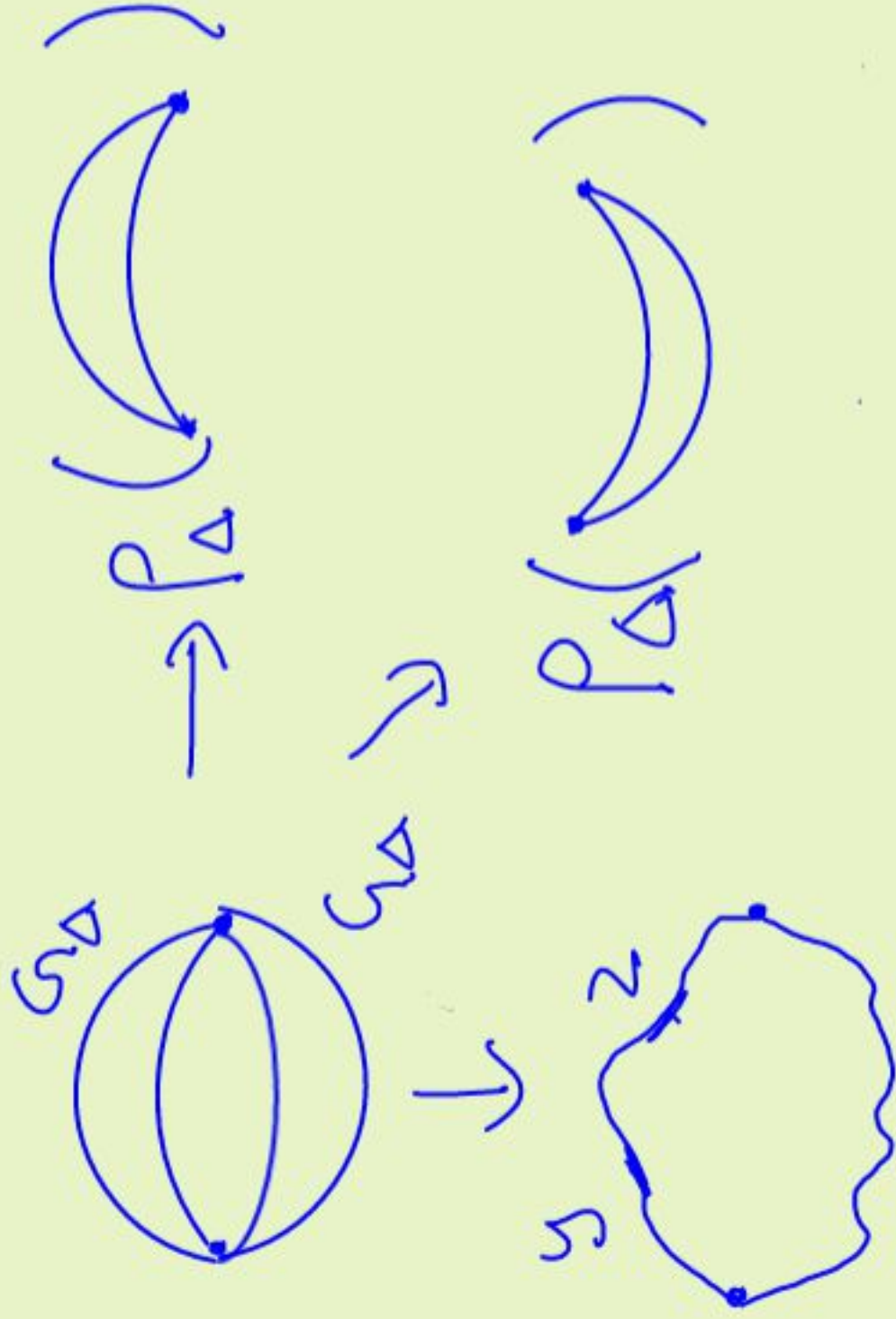
Larger depths $P_{\Delta+1}: X \rightarrow Y \cup Z \cup \Sigma_0, 1, 3$



Larger depths $f_{\Delta+1}: X \rightarrow Y \cup Z \cup \{0,1\}$



Larger depths $P_{\Delta+1}$: $X \rightarrow Y \cup Z \cup \{0, 1\}$



Choose uniformly
at random
among these
options

Summary

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→ Relative succinctness of P-depth Δ
 $\Delta+1$ in comparison to P-depth Δ .

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Summary

- Relative succinctness of P-depth Δ .
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- YES in multilinear setting.

Thank you!