

Polynomial Identity Testing for Read-4 Arithmetic Formulas

Nimrod Kaplan

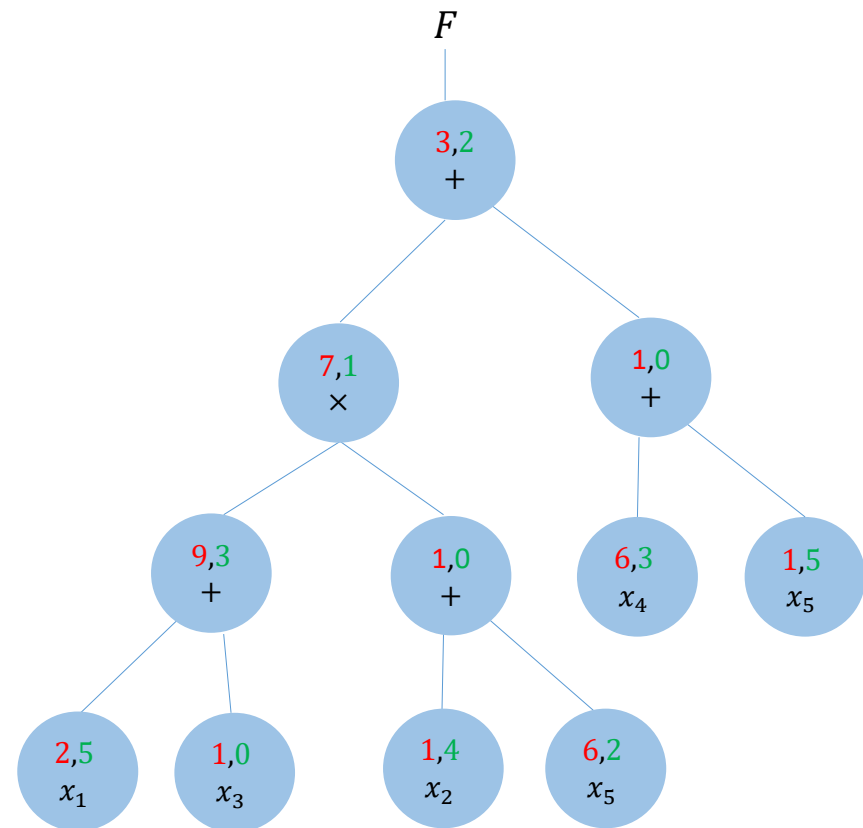
Amir Shpilka

Overview

- Background
- Our results
- Why is Read-4 interesting?
- Some new techniques
- Future directions – higher reads

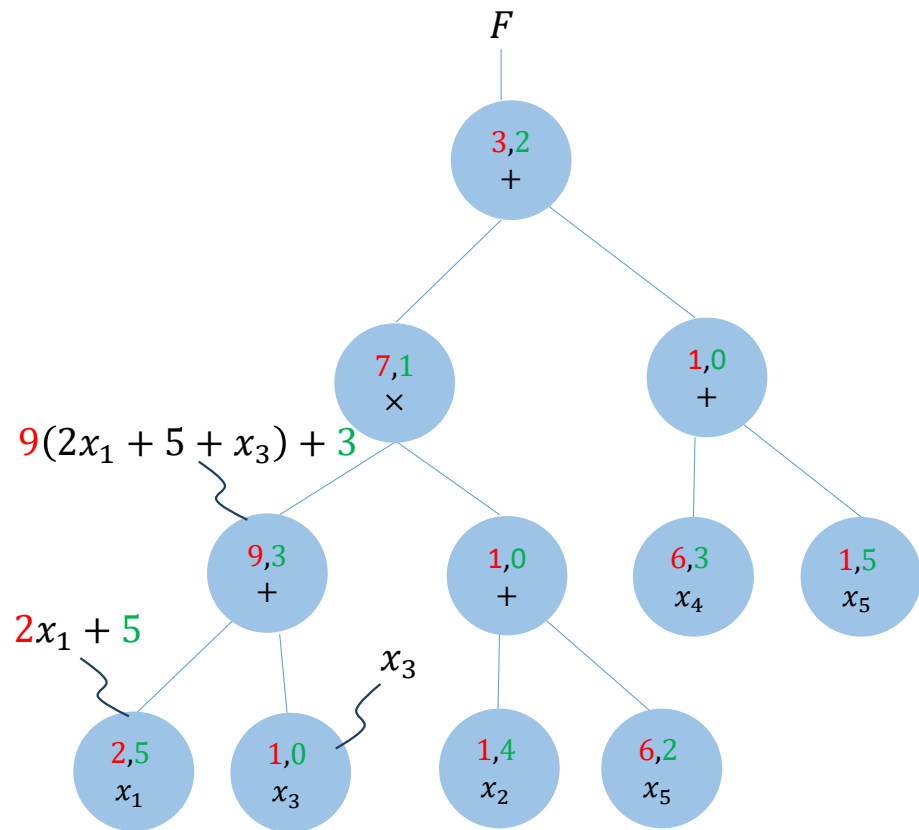
Bounded Read Arithmetic Formulas

- Binary tree
- Inputs are **variables** $x_1, x_2, \dots, x_n = \vec{x}$
- Gates labeled with $+$ and $\times$
- **Multiplicative** and **additive** constants
- Computation in the “natural” way
- **Read- k** formula: Every variable $x_i \in \mathcal{x}$ appears at most k times (RkF, ROF)



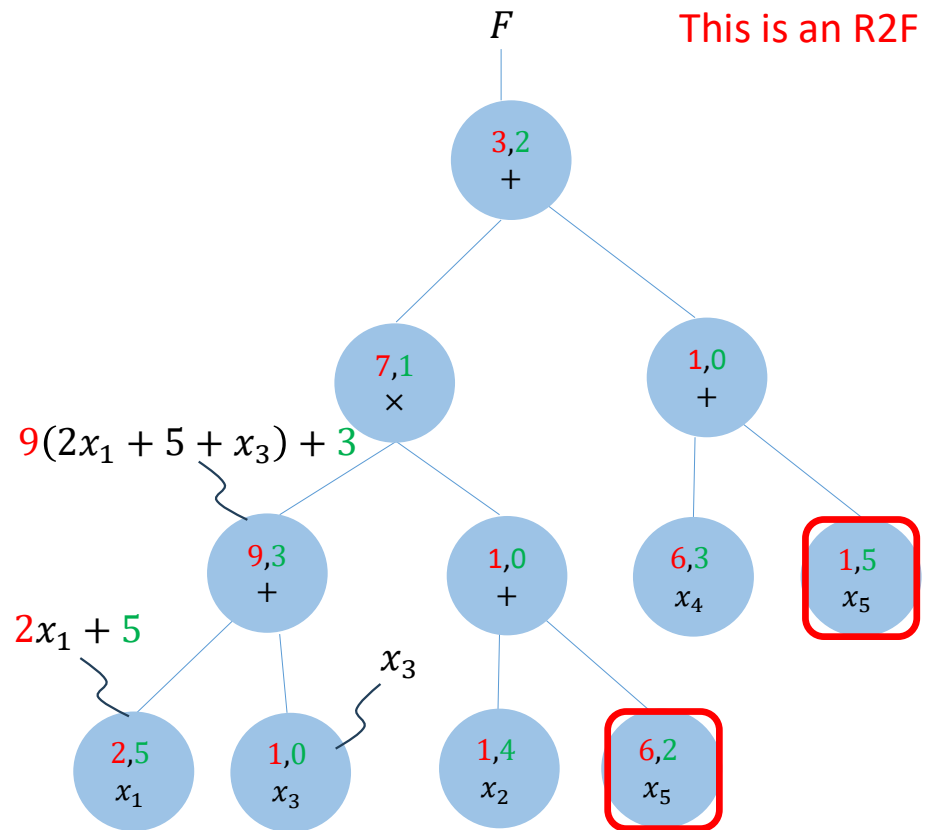
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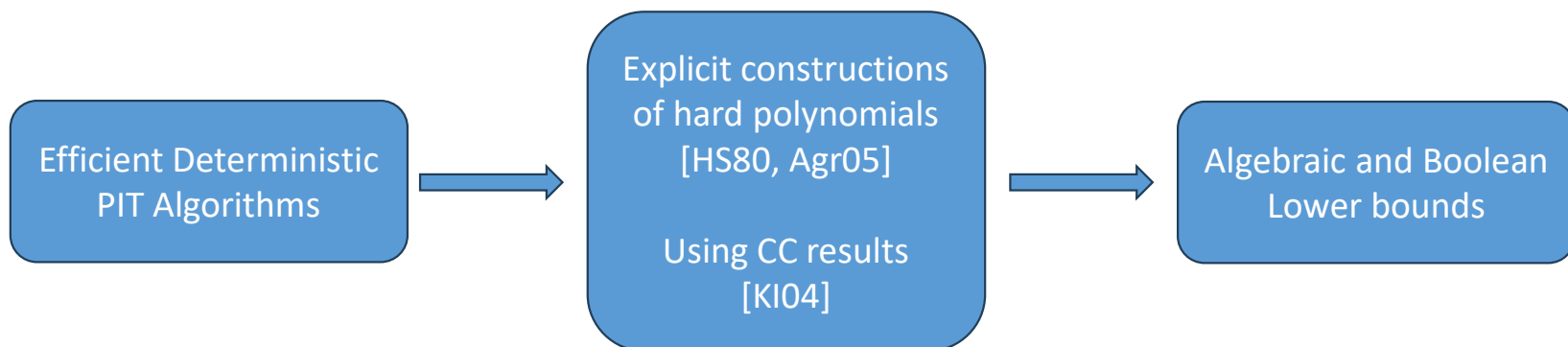
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We want to find a **deterministic** algorithm for PIT.



Polynomial Identity Testing (PIT)



Whitebox

$f \in \mathbb{F}[\mathbf{x}]$ is given to us as a
circuit\formula.



Polynomial Identity Testing (PIT)

Blackbox



$f \in \mathbb{F}[\mathbf{x}]$ is given to us as an **oracle**.

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We are **only** allowed to query the polynomial on chosen points.

These sets of points are called Hitting Sets

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We **hit** $\mathcal{C} \setminus f$ – found a HS for $\mathcal{C} \setminus f$.

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Previous Work

Model	Whitebox	Blackbox	BB Orbits
ROF	Poly	Poly [SV15,MV17]	Quasi-poly [MS22]
R2F	Poly [MRS14]	Quasi-Poly [Pra19, Sha23]	?
R3F	Poly [MRS14]	Quasi-Poly [Pra19]	?
R4F	?	?	?
Σ ROF	Poly	Poly [SV15,MV17]	?
Multilinear-R k F	Poly [AvMV15]	Quasi-Poly [AvMV15]	?

[SV15] – Amir Shpilka and Ilya Volkovich (STOC)
 [MV17] – Daniel Minahan and Ilya Volkovich (CCC)
 [AvMV15] - Anderson et al. (CCC)
 [Pra19] - Gautam Prakriya (Doctoral Thesis)

[MS22] – Dori Medini and Amir Shpilka (CCC)
 [Sha23] – Nadav Shamir (Master Thesis)
 [MRS14] – Mahajan et al. (TCS)

Previous Work

$$\text{Orbit}(\mathcal{C}) = \left\{ f(\ell_1, \dots, \ell_n) \mid \begin{array}{l} f \in \mathcal{C}, \\ \ell_i \text{'s linearly independent} \end{array} \right\}$$

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Results

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* Improved algorithm.

Main Result – PIT for R4Fs

Theorem.

There is a deterministic blackbox PIT algorithm for R4Fs in time:

$$O(2n^{c+18 \log n})$$

For constant c .

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What makes read-4 so **special**?

Known Techniques – Partial Derivatives

Previous results:

PIT for New Model $\xrightarrow[\text{via partial derivatives}]{\text{reduce}}$ Known PIT

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 g_i 's nonzero & simple

Known Techniques – Partial Derivatives

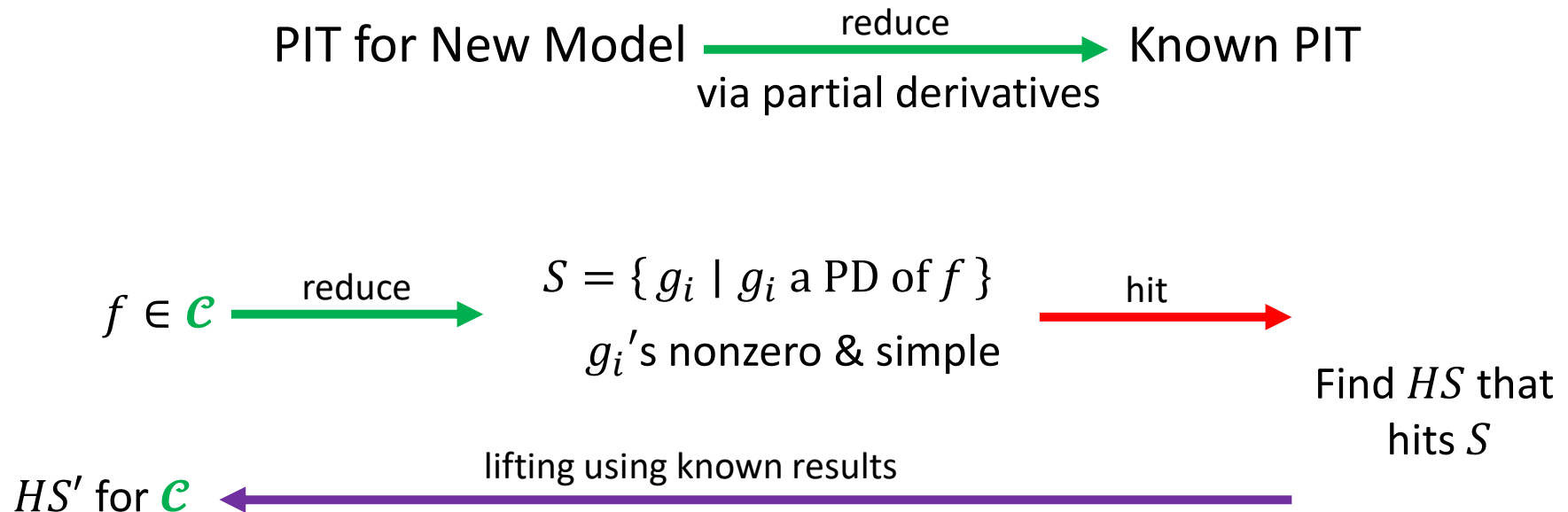
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PIT for New Model $\xrightarrow[\text{via partial derivatives}]{\text{reduce}}$ Known PIT

$f \in \mathcal{C}$ $\xrightarrow{\text{reduce}}$ $S = \{g_i \mid g_i \text{ a PD of } f\}$
 g_i 's nonzero & simple $\xrightarrow{\text{hit}}$ Find HS that hits S

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PIT for New Model $\xrightarrow[\text{via partial derivatives}]{\text{reduce}}$ Known PIT

- ROF: Homogenization
- $\sum$ ROF: PDs + Hardness of Representation $\rightarrow$ ROF PIT
- R2F: Fragmenting (repeated PDs) $\rightarrow \sum^2$ ROF PIT
- R3F: Fragmenting (repeated PDs) $\rightarrow$ R2F PIT
- ml-RkF: Fragmenting/Shattering (repeated PDs) $\rightarrow$ ml-R($k - 1$)F
- Orbits for ROFs: Directional derivatives

Known Techniques – Partial Derivatives

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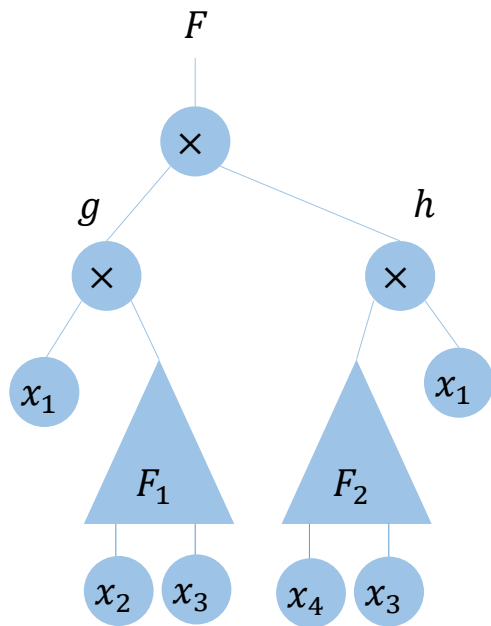
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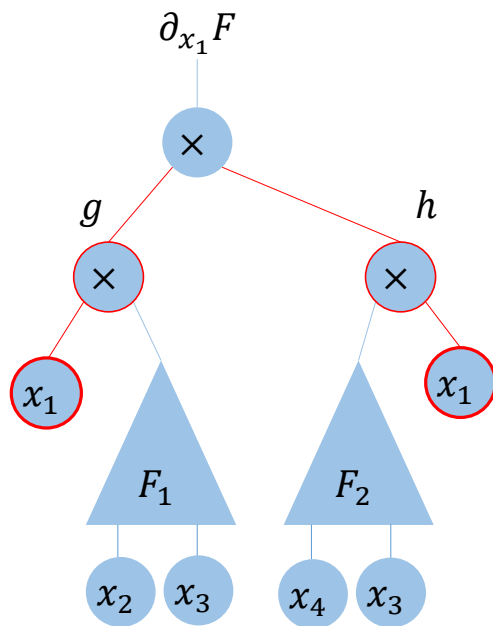
PDs **simplify** the problem

This means, these models are **closed** under PDs

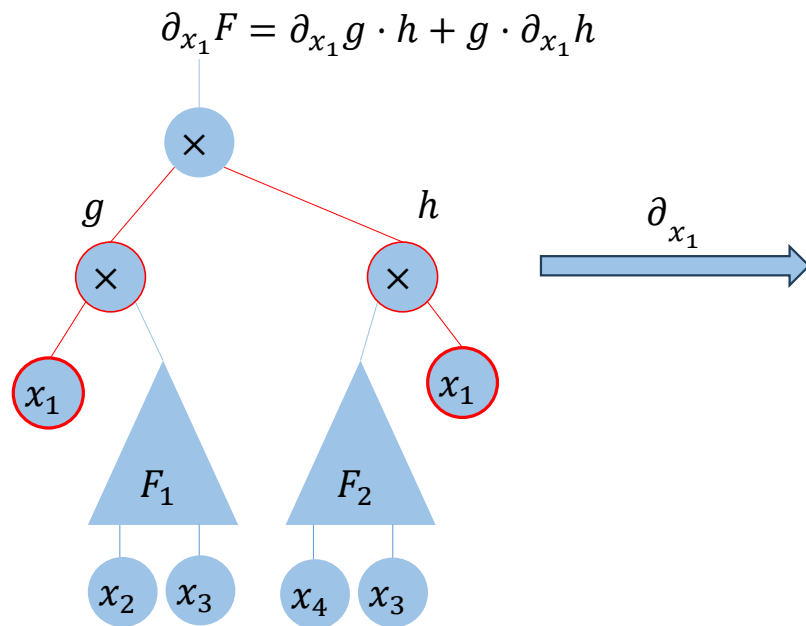
Differentiation Problems



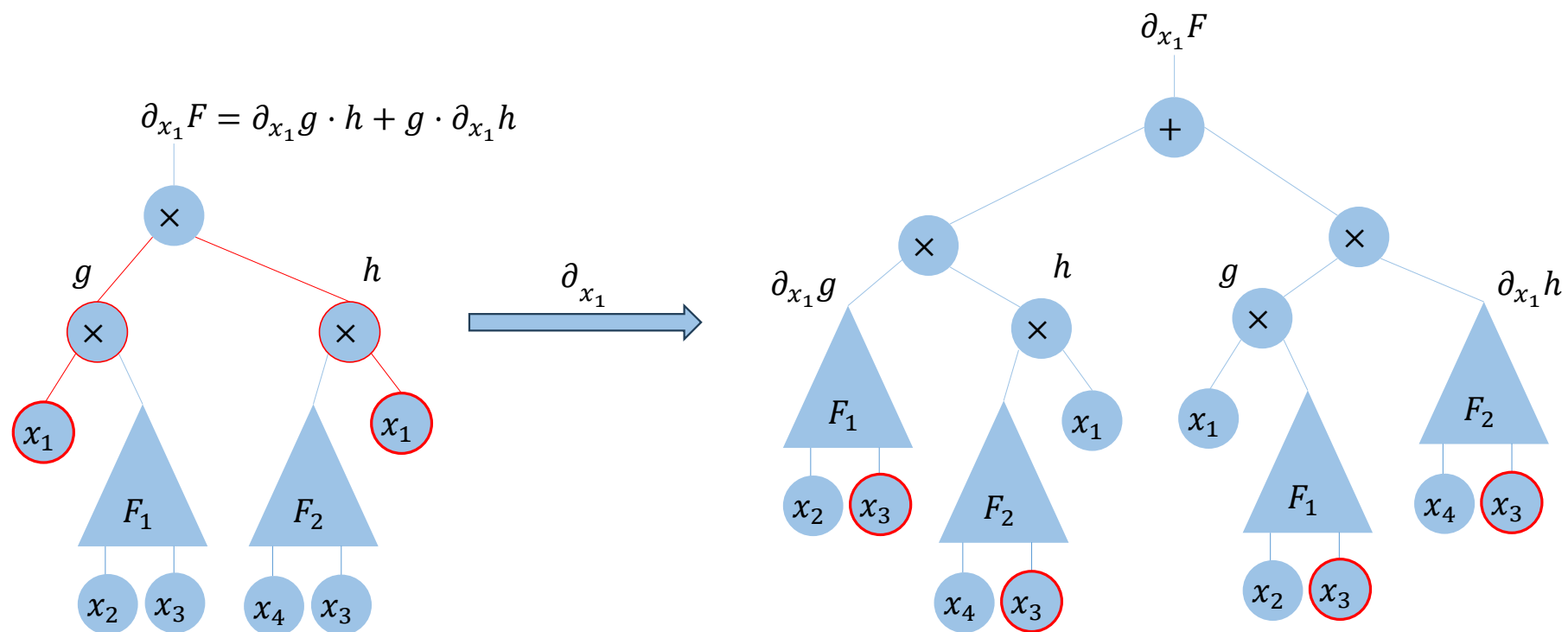
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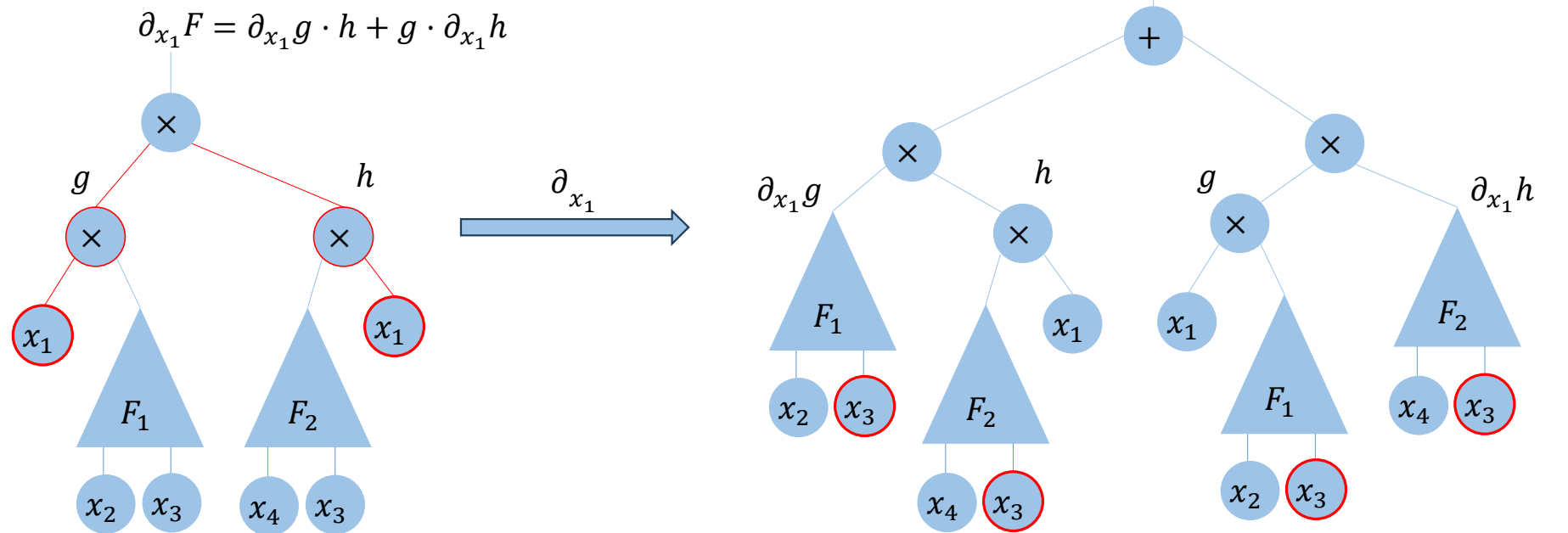
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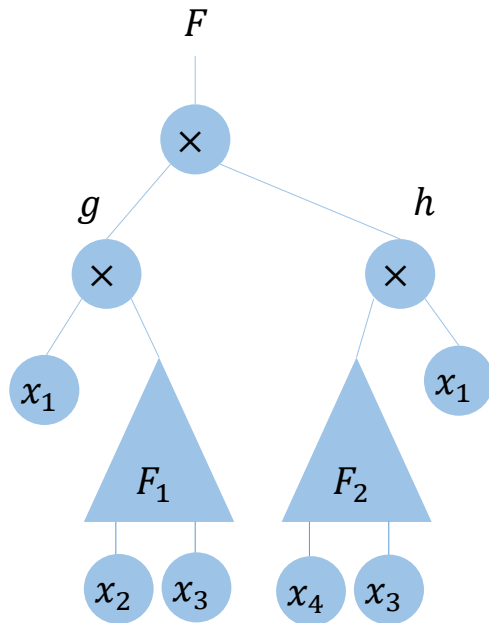


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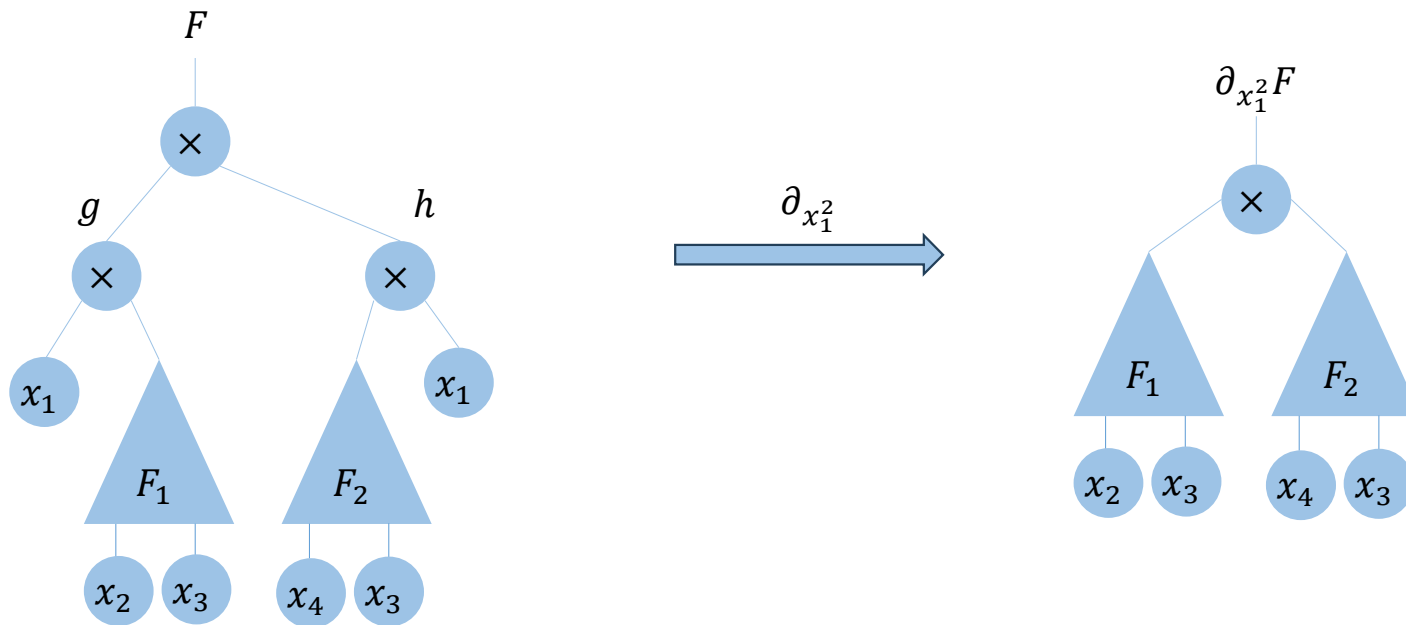
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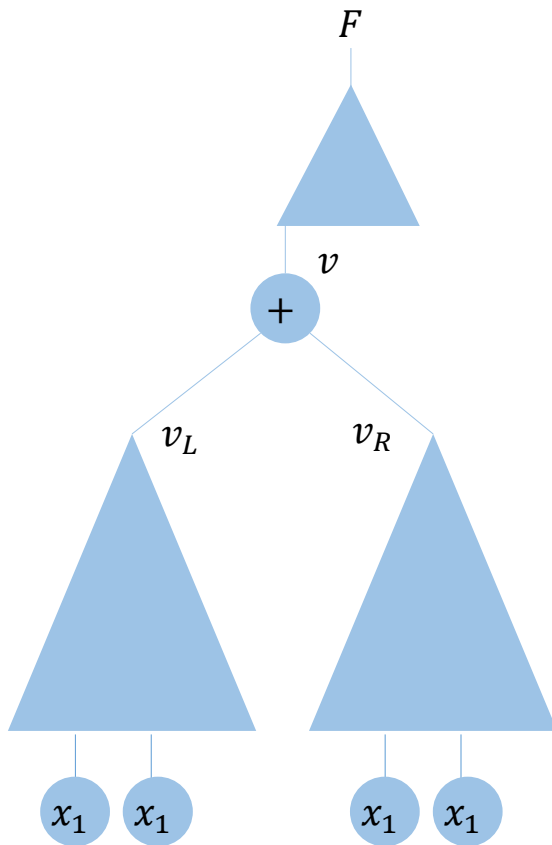


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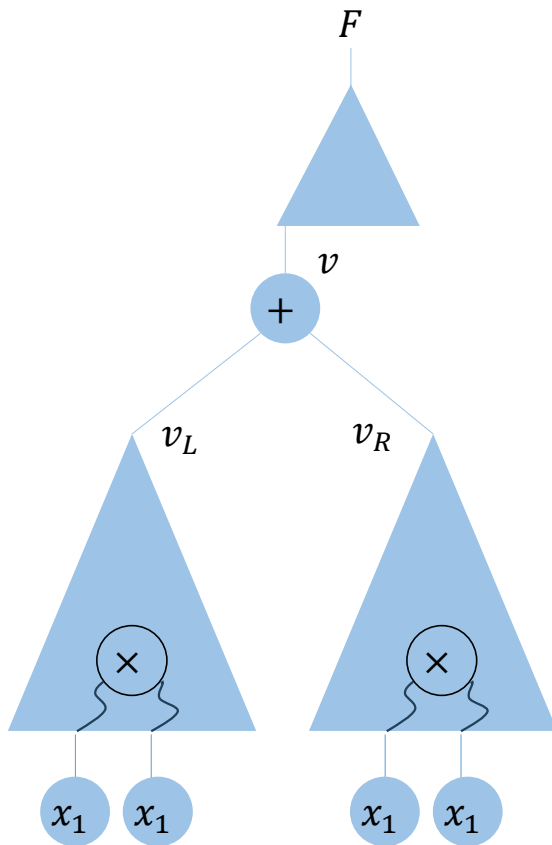
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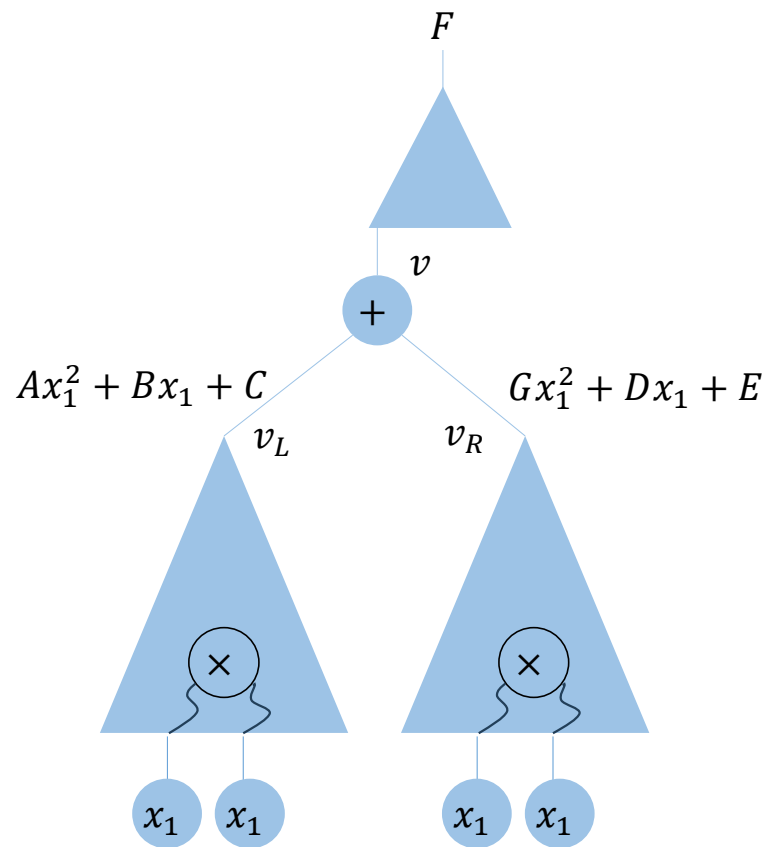
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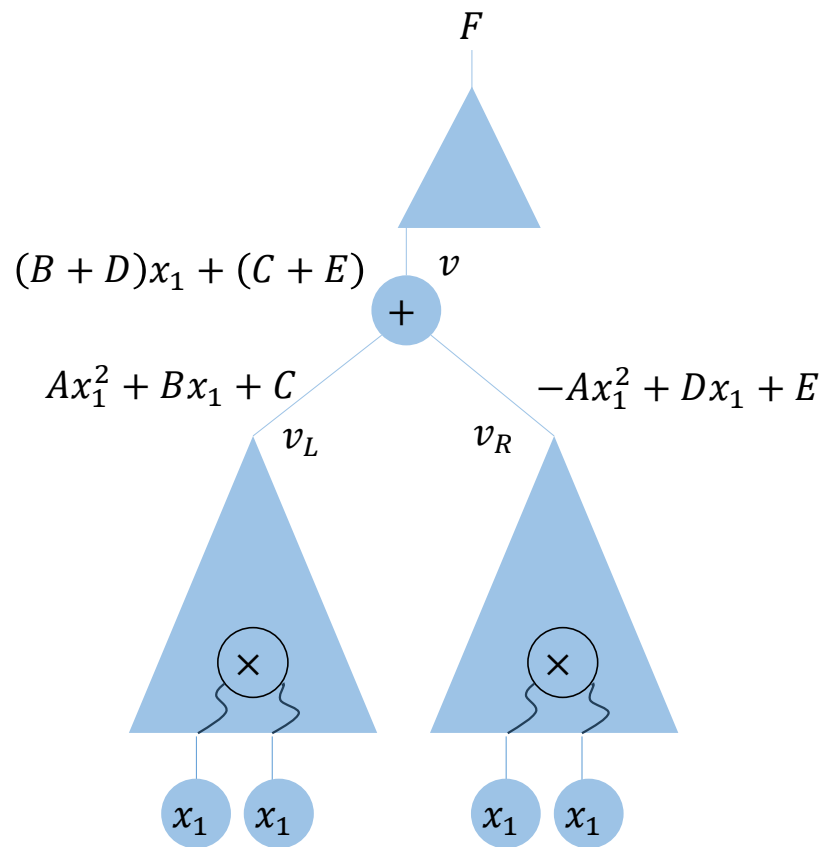
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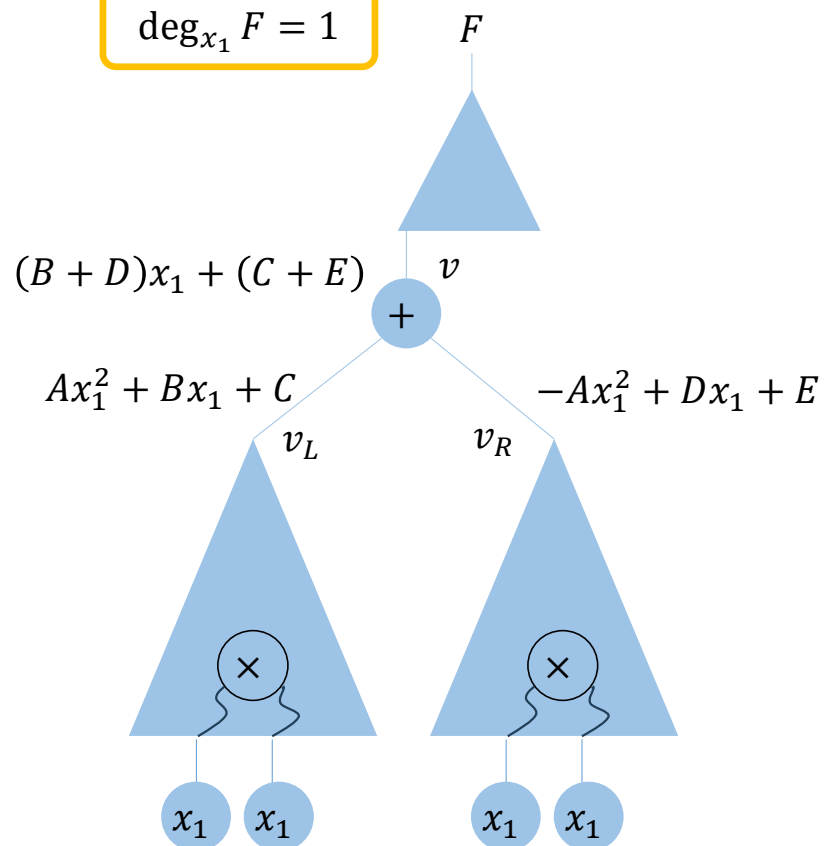


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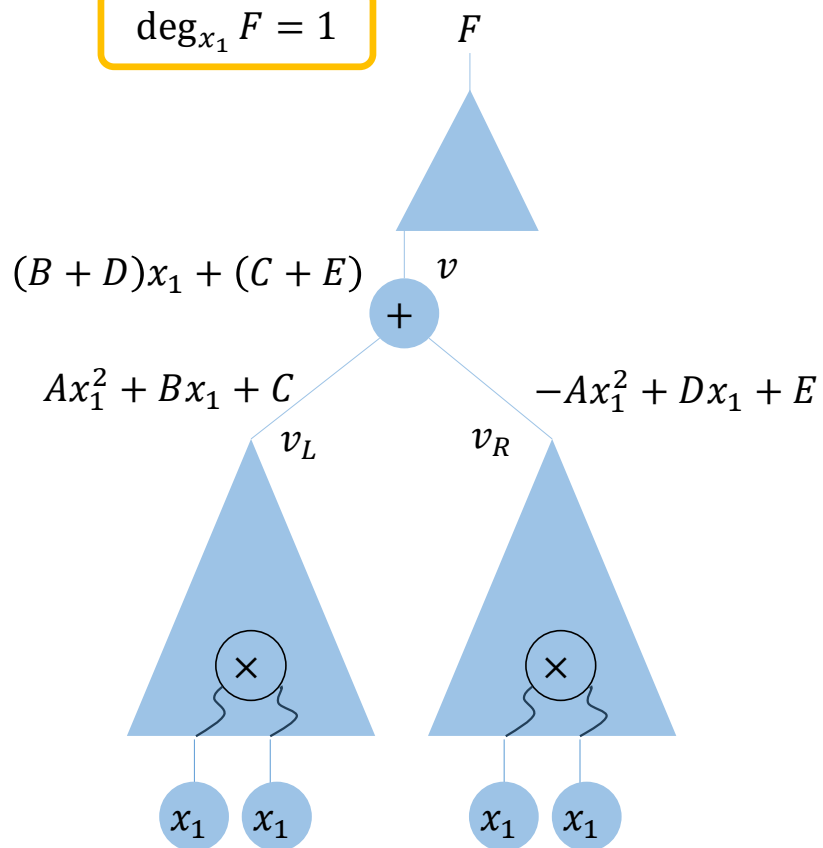
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$$\deg_{x_1} F = 1$$



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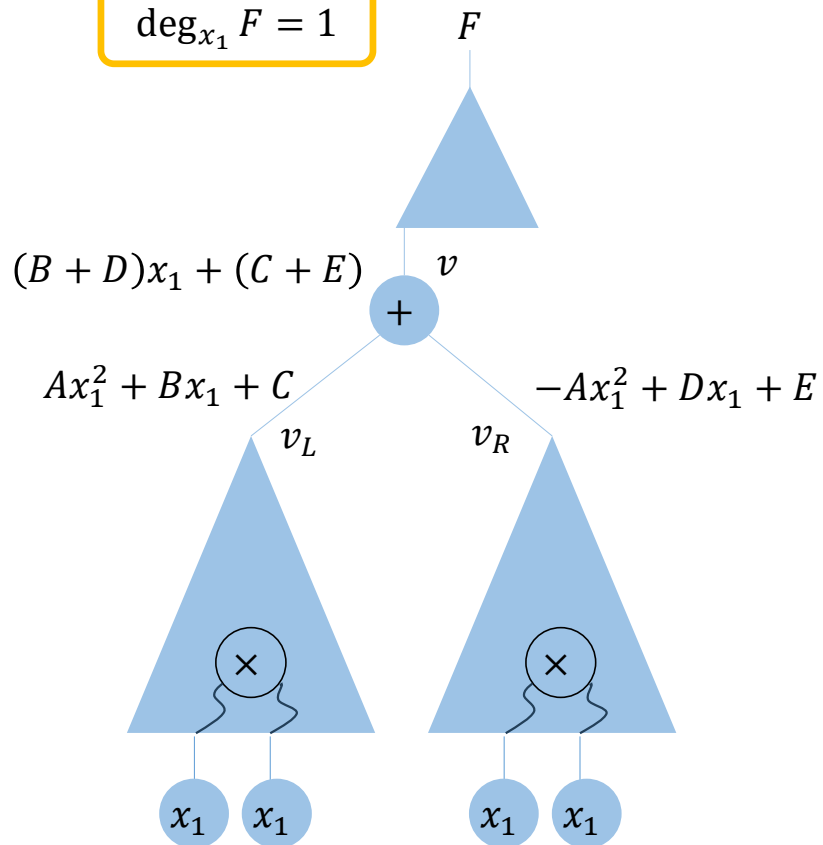
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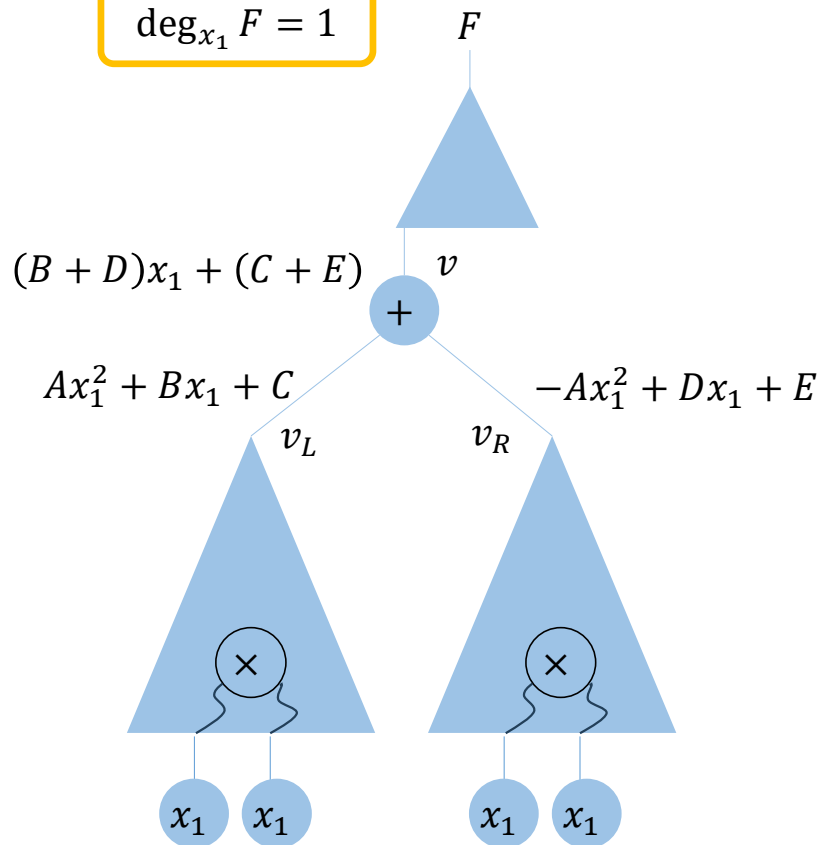


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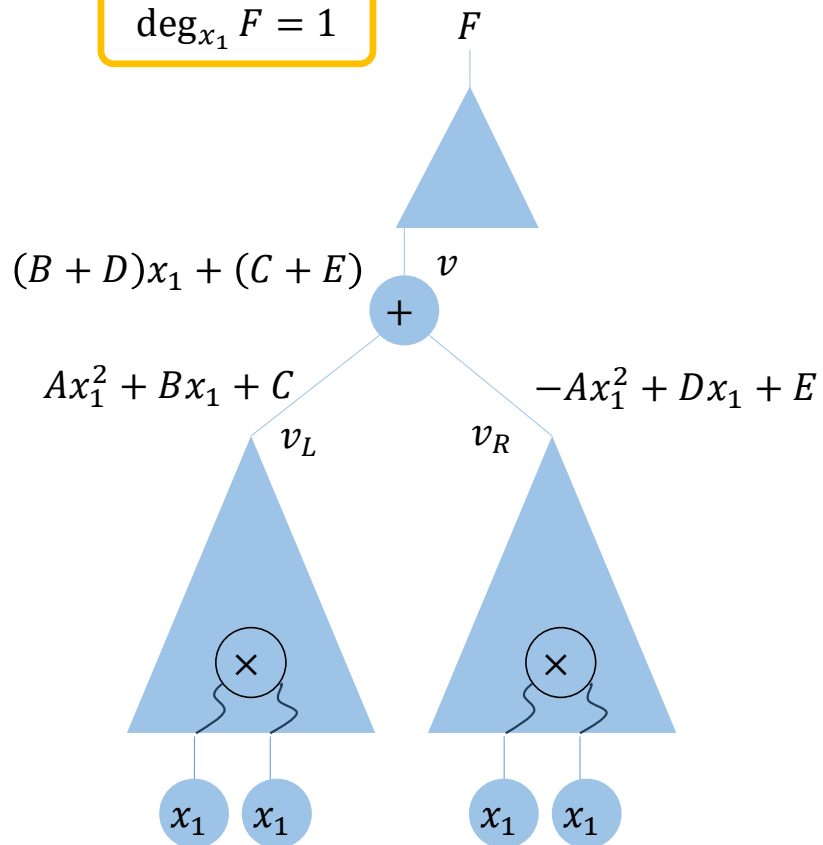


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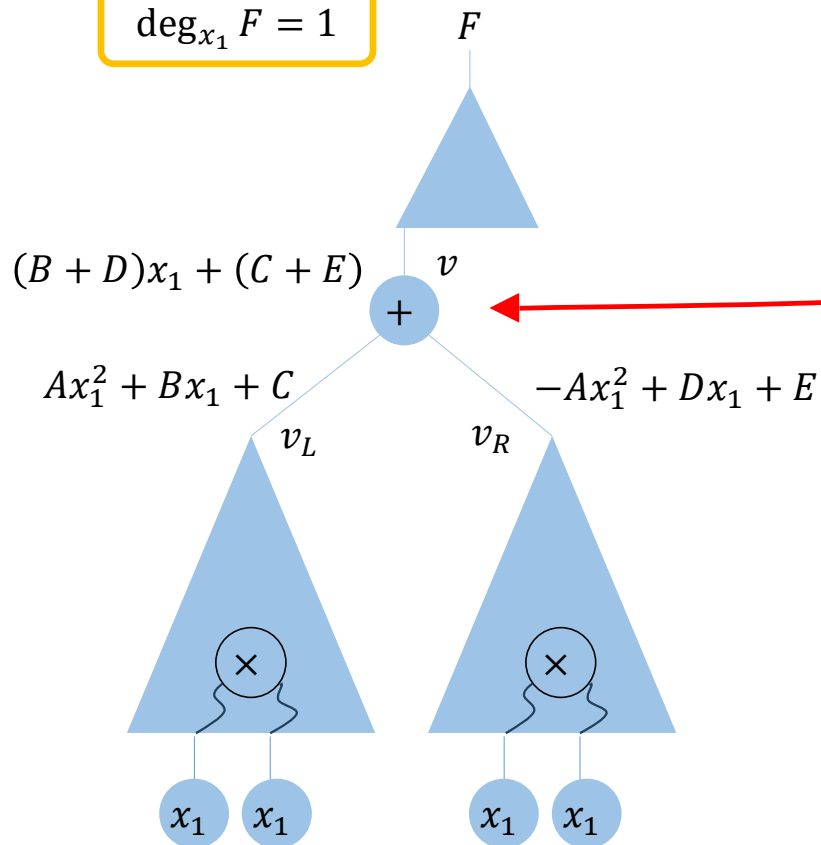


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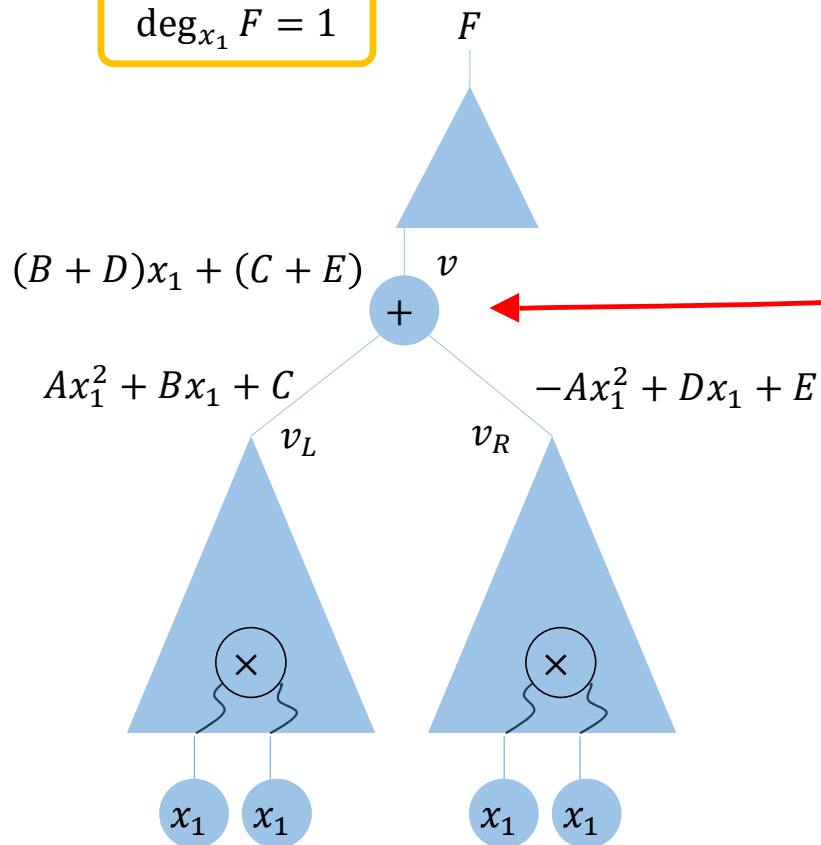
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The problem is here

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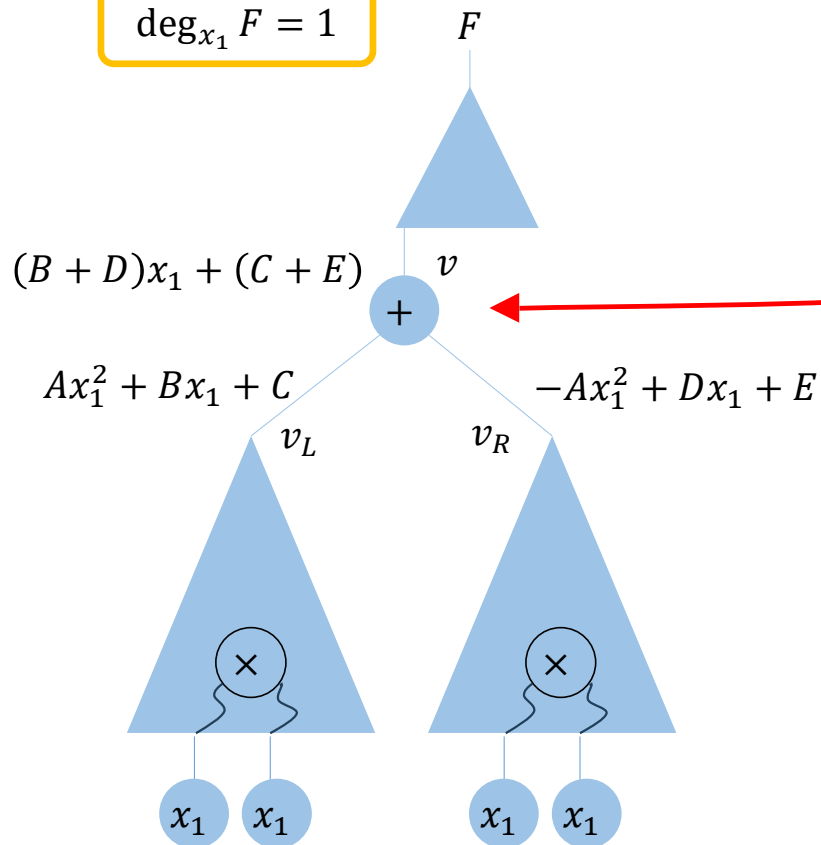
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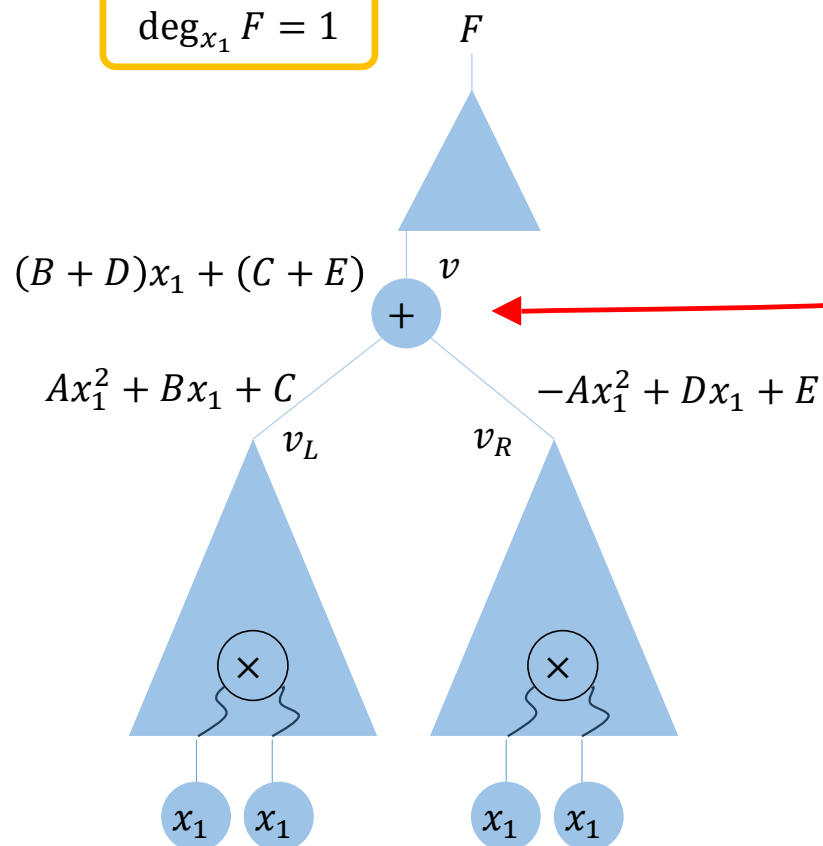
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This is the barrier that breaks
the previous methods

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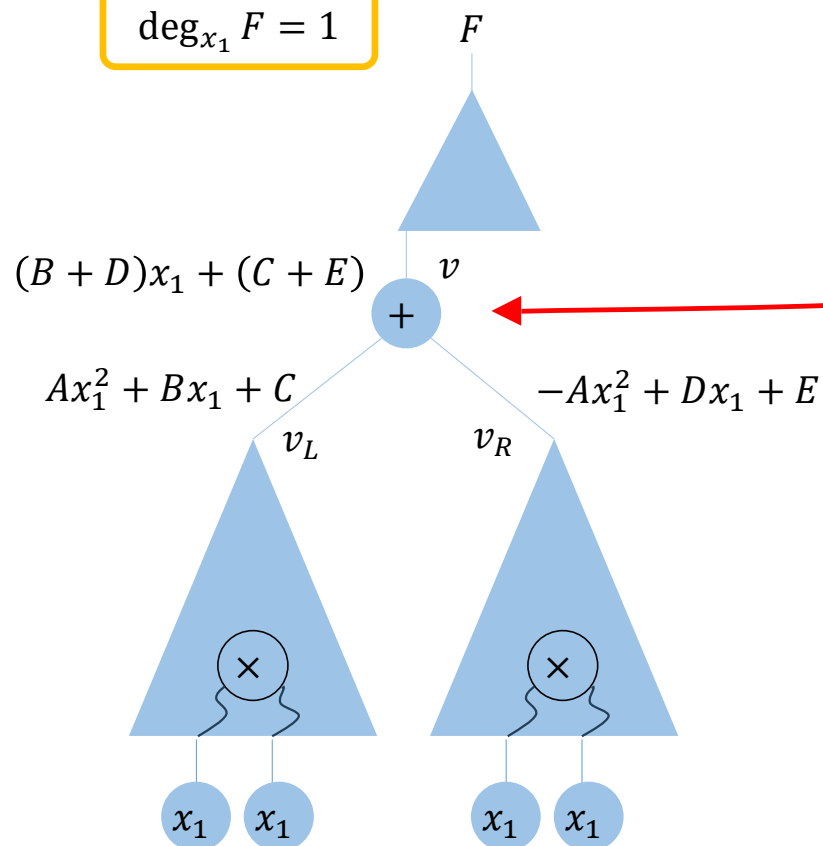
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Cancellations are exactly what makes PIT hard in general

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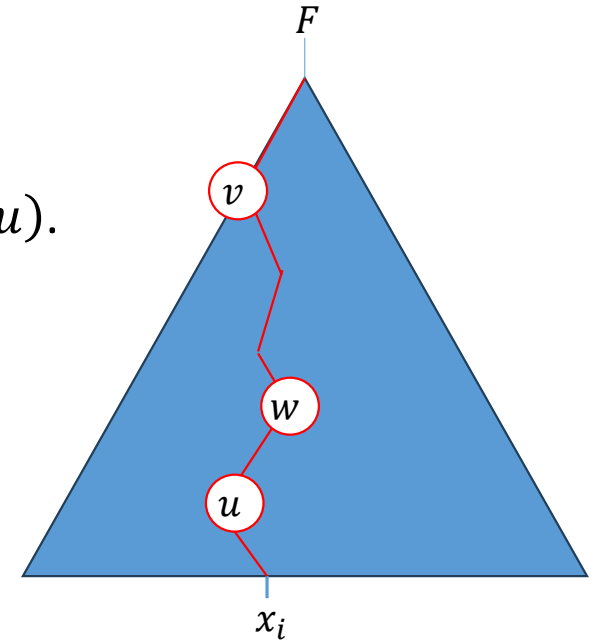
Requires read ≥ 4 !

Can also happen in Σ^2 R2F

The Structural Property

x_i is **structural** in F if:

$\forall$ gates v, u : v is an **ancestor** of $u \Rightarrow \deg_{x_i}(v) \geq \deg_{x_i}(u)$.



$$\deg_{x_i} u \leq \deg_{x_i} w \leq \deg_{x_i} v$$

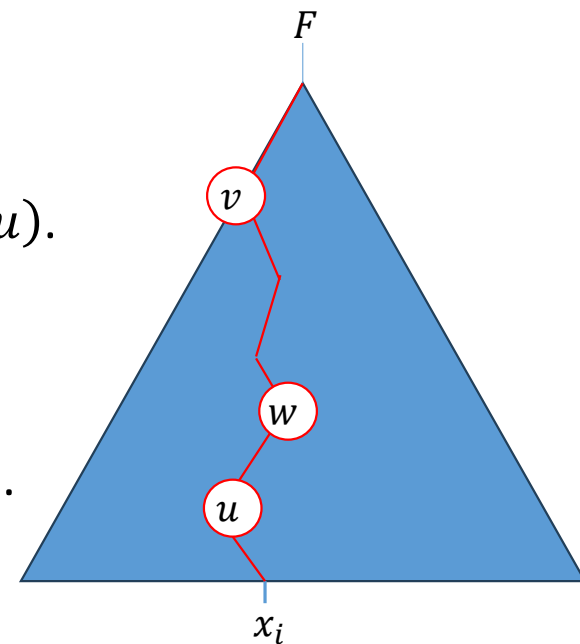
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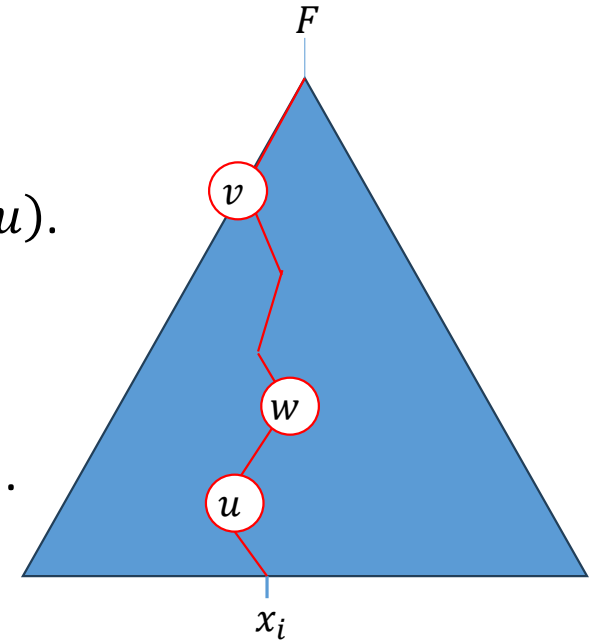
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F is **structural** $\Rightarrow$ previous methods **work** (mostly)



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New Techniques – Two Examples

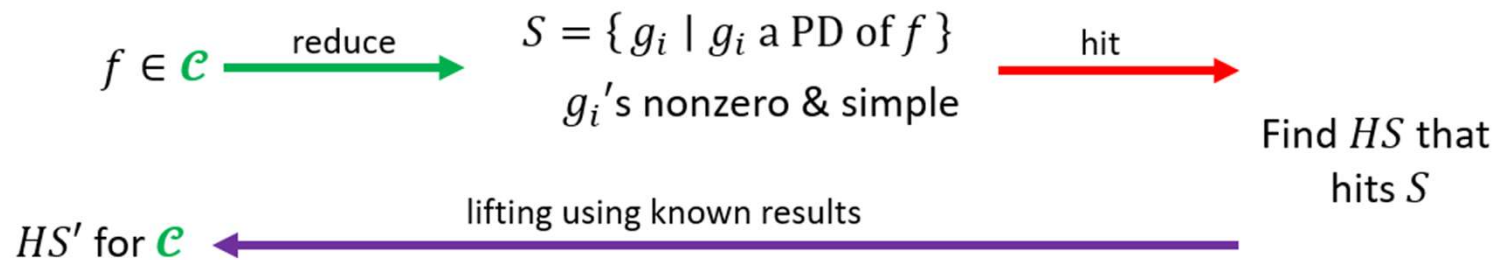
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Divide and conquer

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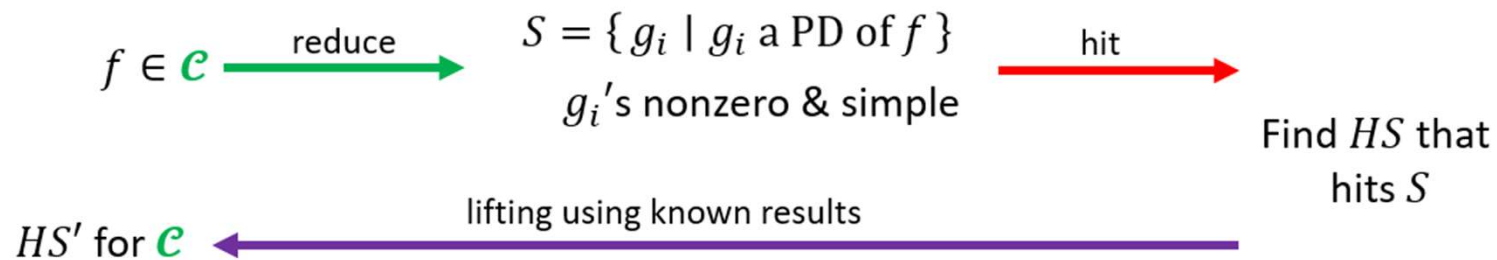
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Divide and conquer, replaces old framework.



2. Dominating Degree Patterns.

New **closure** rules (under taking PDs).

Restrictions and HoR

Idea: Divide and conquer

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Such an assignment maintains “any property” of f .

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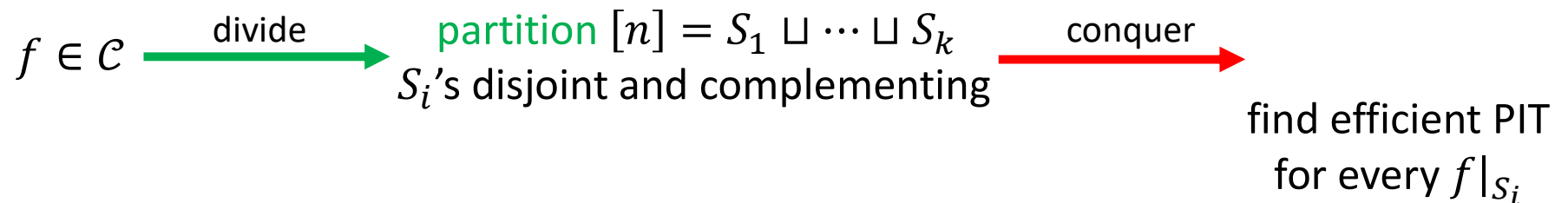
$f|_S$: Substituting a generic/random assignment to variables outside of S .
Such an assignment maintains “any property” of f .

$f \in \mathcal{C} \xrightarrow{\text{divide}} \text{partition } [n] = S_1 \sqcup \cdots \sqcup S_k$
 S_i 's disjoint and complementing

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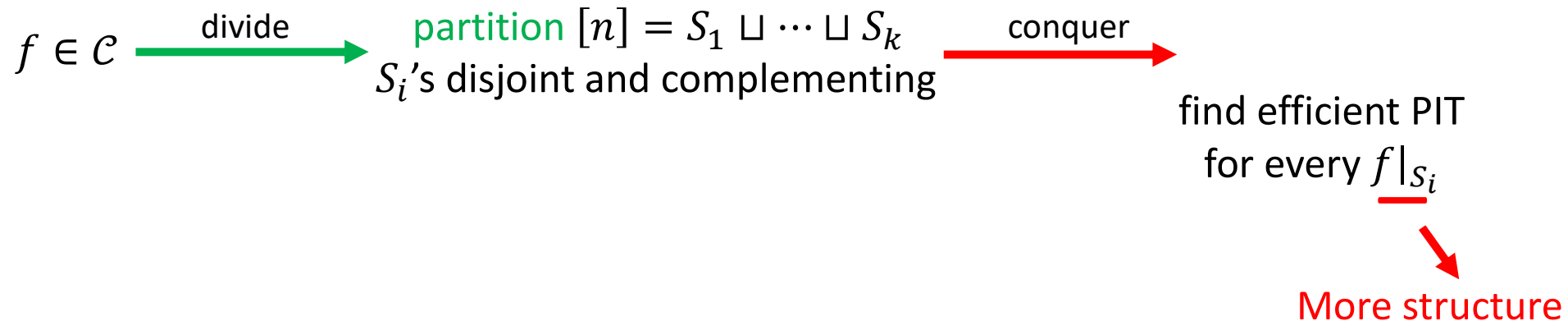
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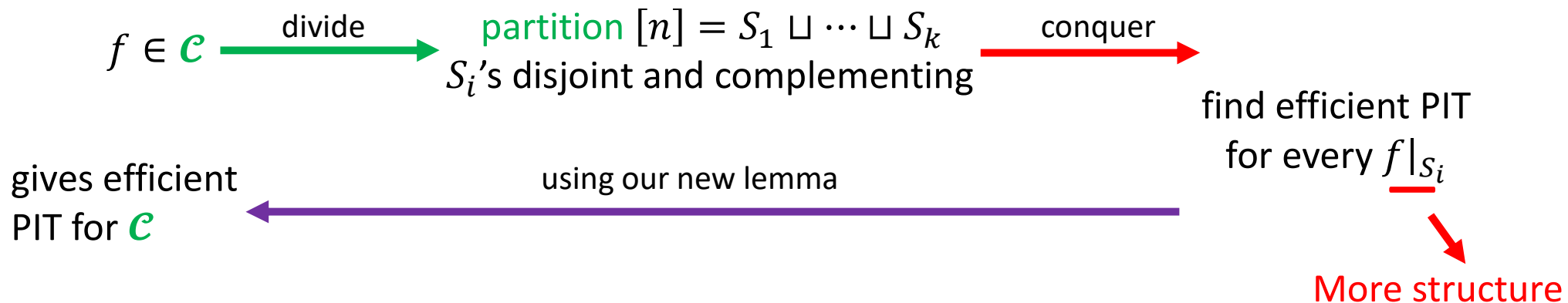
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1. $f|_{S_1} \in \text{“Structural” } \Sigma^2 \text{R2F}$

2. $f|_{S_2} \in \text{“Totally non-structural” } \Sigma^2 \text{R2F}$

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2. $f|_{S_2} \in \text{“Totally non-structural” } \Sigma^2 \text{R2F}$ - strong structural constraints.

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Gives more **closure rules** under PDs.

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Old closure lemmas:

$$f \in RkF \quad \Rightarrow \quad \partial_{x^d}(f) \in RkF$$

This is **local**: **one** variable at a time.

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This is **local**: **one** variable at a time.

New idea:

What about **global** conditions involving **several** variables?

Choose $J \subseteq [n]$ and differentiate by a **maximal** exponent vector on J .

Dominating Degree Patterns - Example

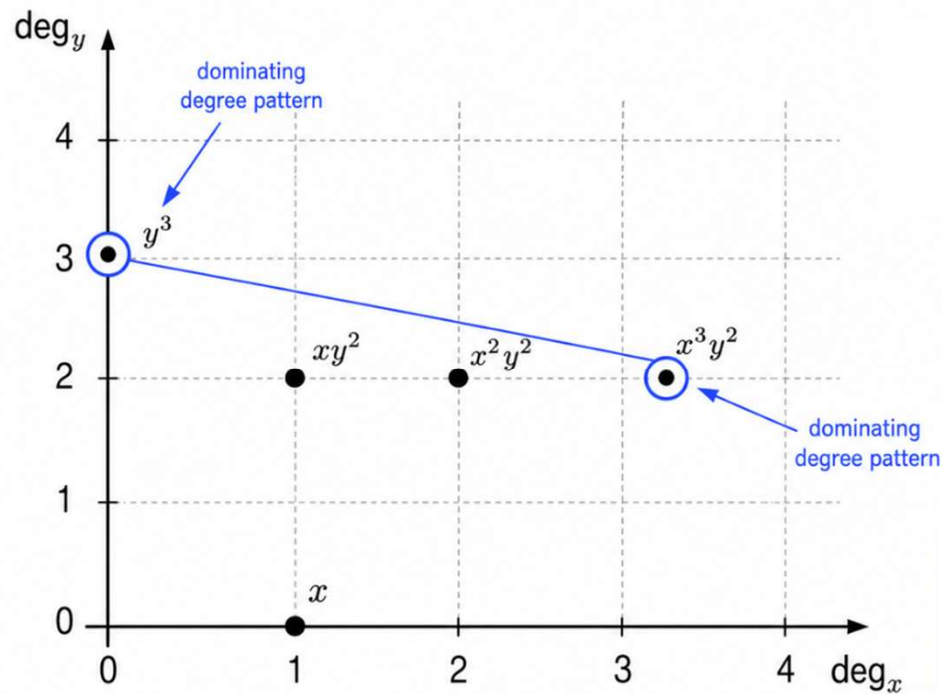
$$f(x, y, z, w) = (5zw^3)xy^2 + (2z^2w)x^3y^2 + (5)x^2y^2 + (zw)x + (7w)y^3$$

Take $J = \{x, y\}$;

Dominating Degree Patterns - Example

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Take $J = \{x, y\}$;



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Let $\vec{e} \in \mathbb{N}^J$.

Current limitations:

- Variables in J are read-2
- Size of J is at most 3

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More concretely:

1. Can a dominating degree pattern lemma be shown for general RkFs and general set size?
2. Can a better Hardness of Representation result be shown for Resultants of ROFs?

Thanks!