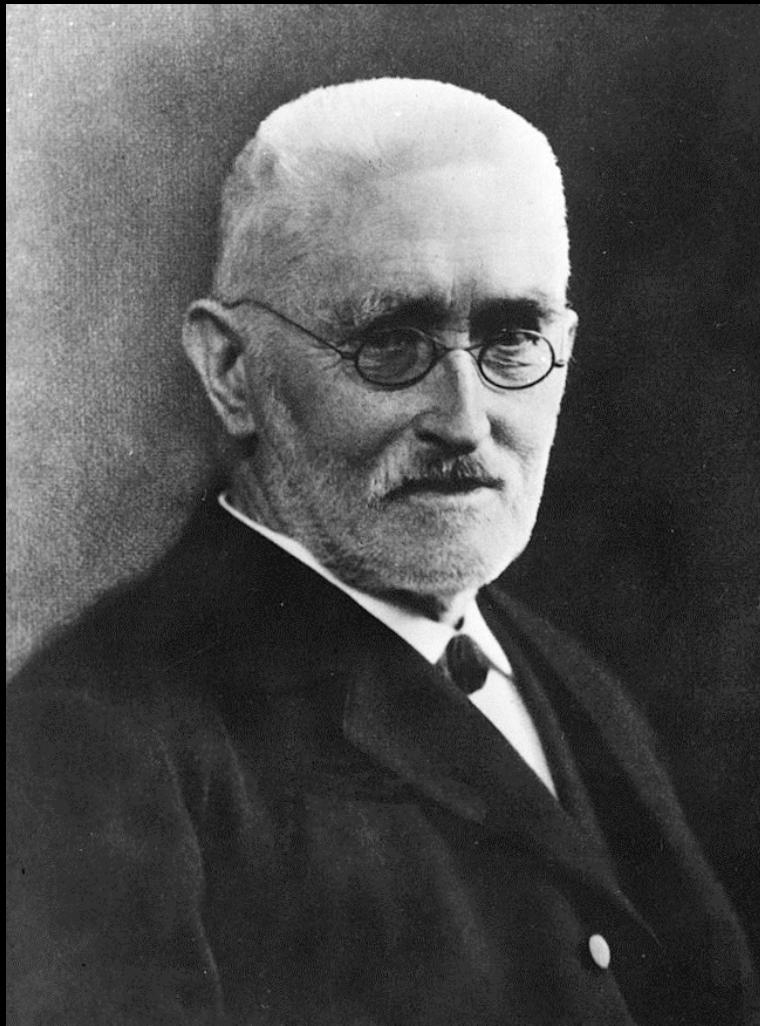


FACT OF THE DAY



Today, October 6, Richard Dedekind
would have turned 191 years old

Slides available at:
tonellicueto.xyz/pdf/8IMM_slides.pdf

Why

does the DESCARTES Solver work?

Alperen A.
ERGÜR

Josué
TONELLI-CUETO

Elias
TSIGARIDAS
Inria





Photo while working on this project

Real Root Isolation I: The Problem

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$$g \in \mathbb{Z}[x]$$

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Intervals J_1, \dots, J_k s.t.

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STURM SOLVER

$$\tilde{\mathcal{O}}_B(d^4 \gamma^2)$$

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PAN'S ALGORITHM

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Q: Can we beat the champion?

Real Root Isolation III:

What do we wish?

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$$\tilde{G}_B(d\gamma)$$

Real Root Isolation III:

What do we wish?

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We wish to find real roots
almost as fast as we read the polynomial!

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Rule of Signs

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Portrait by Frans Hals
Source: Wikimedia Commons

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Moreover,

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\uparrow
 $(0, \infty) \xrightarrow{\text{bijection}} (a, b)$



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DESCARTES SOLVER II:

The Descartes' Oracle

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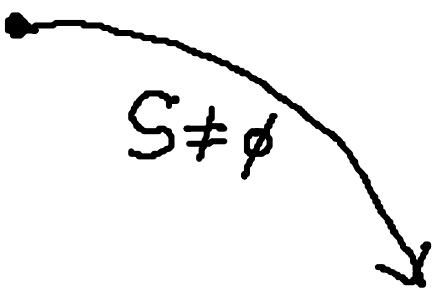
$$\bigcup J_i \subseteq J \Rightarrow \sum V(g, J_i) \leq V(g, J)$$

DESCARTES SOLVER III:

The Algorithm

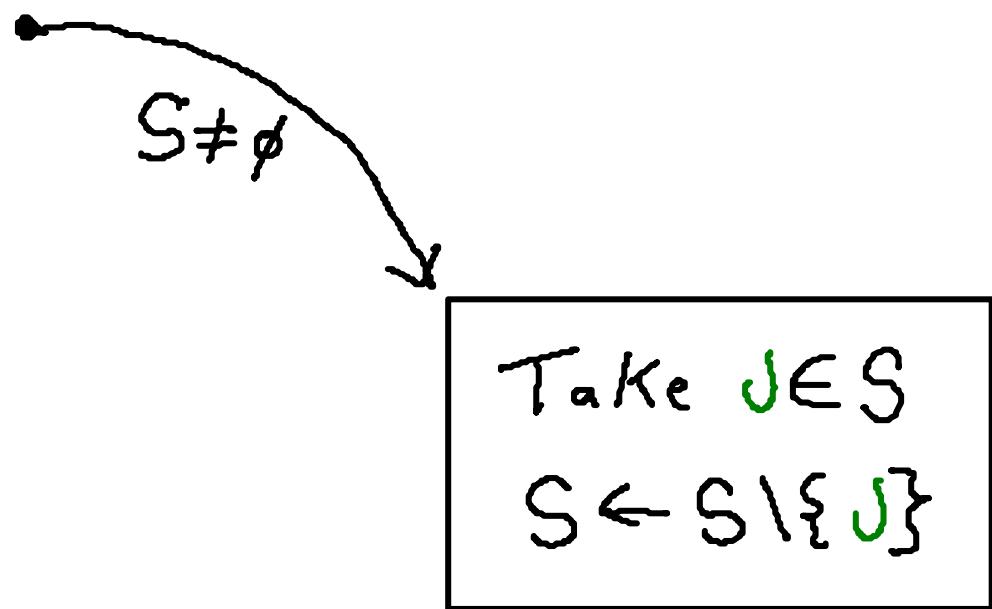
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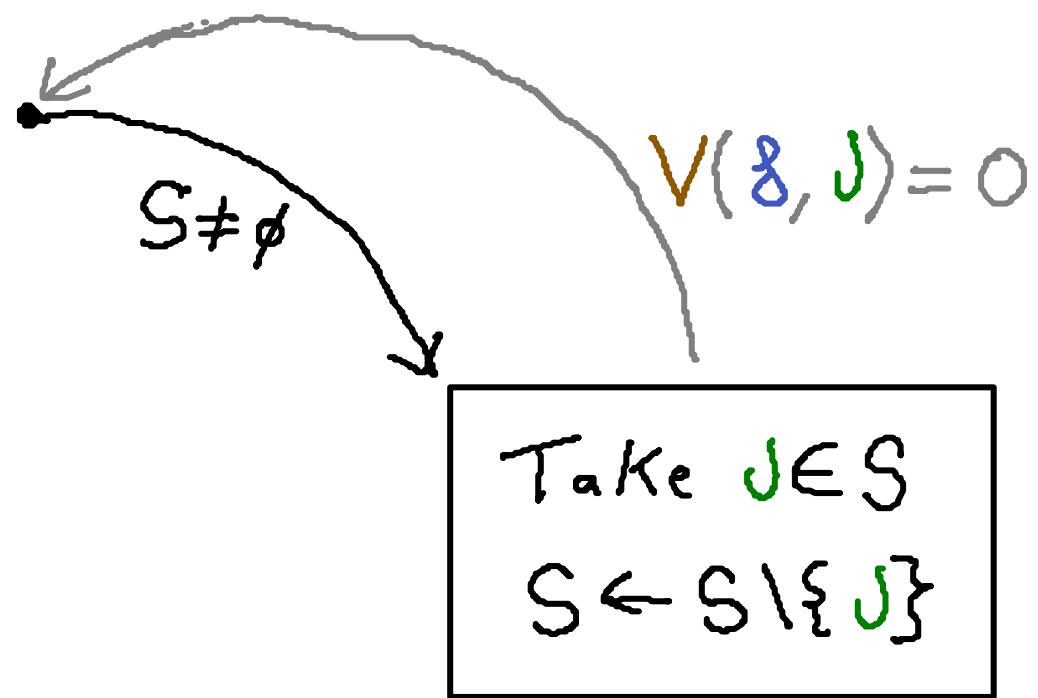
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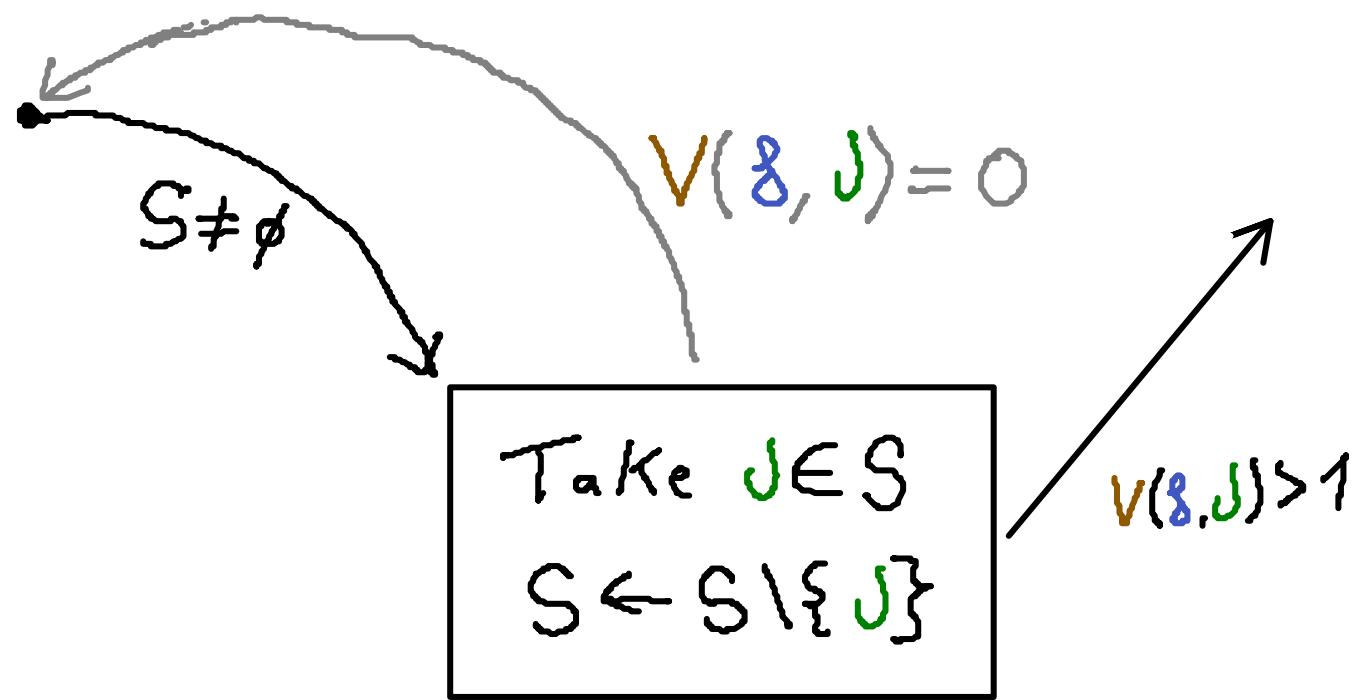
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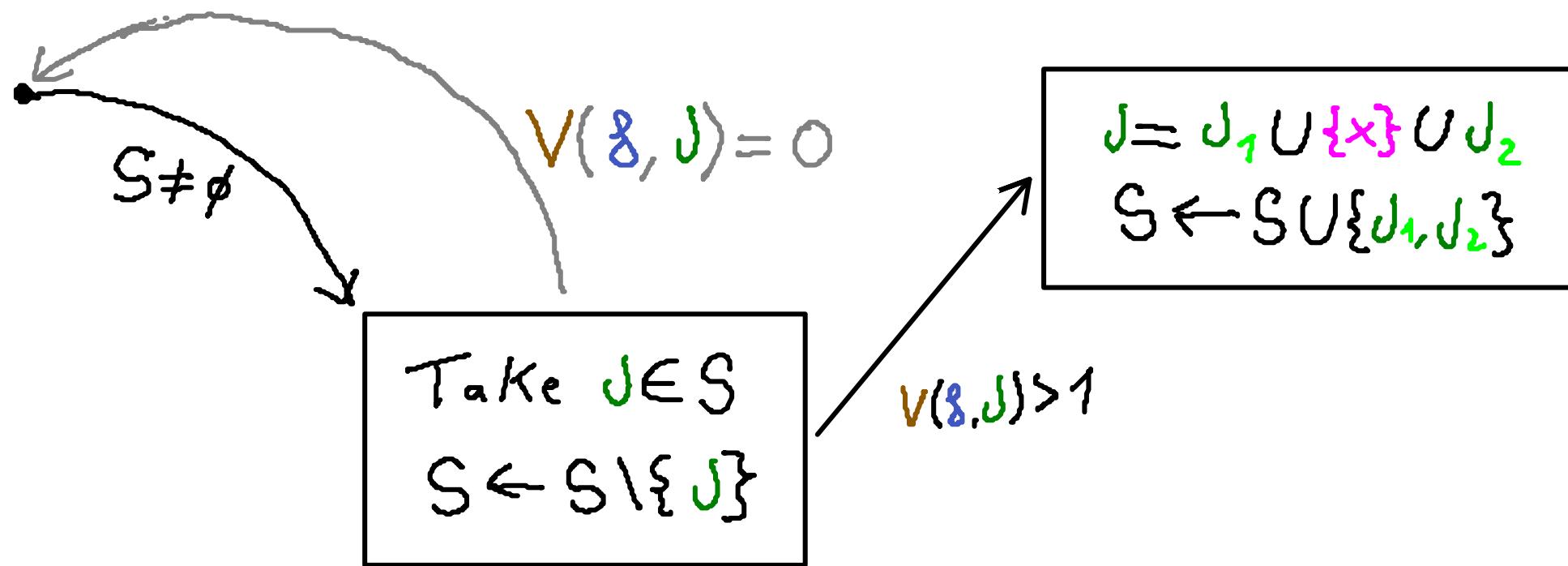
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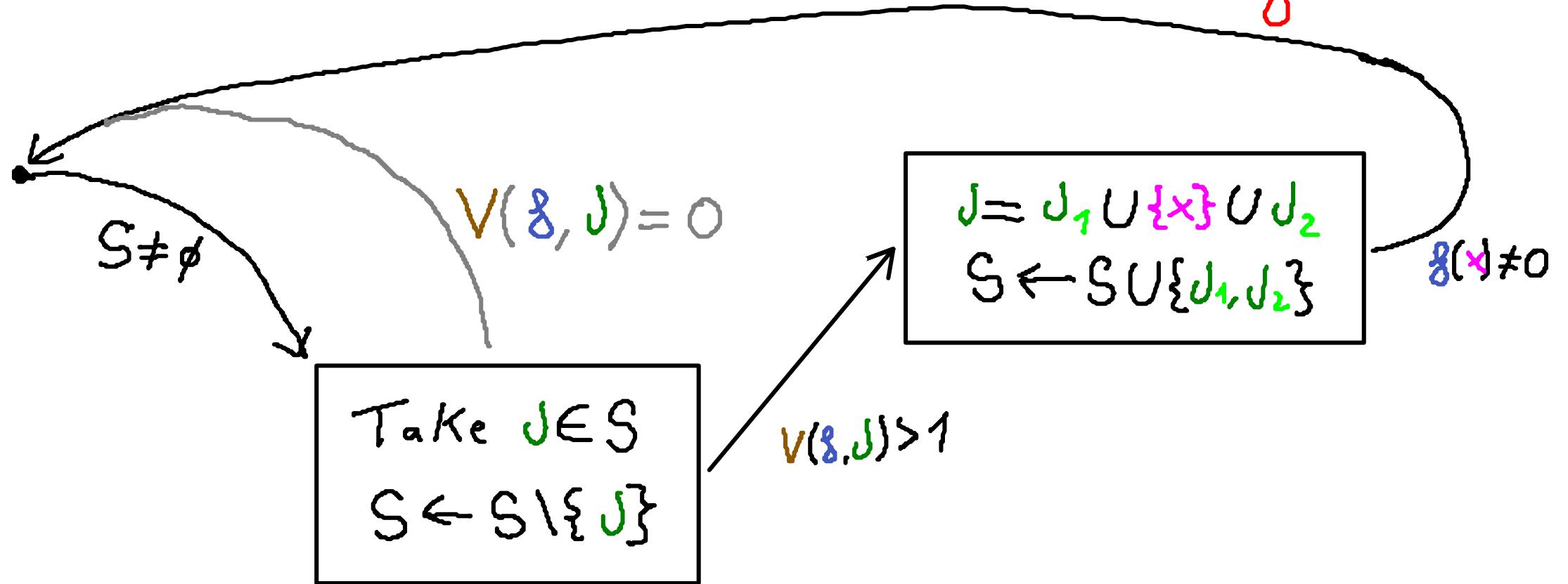
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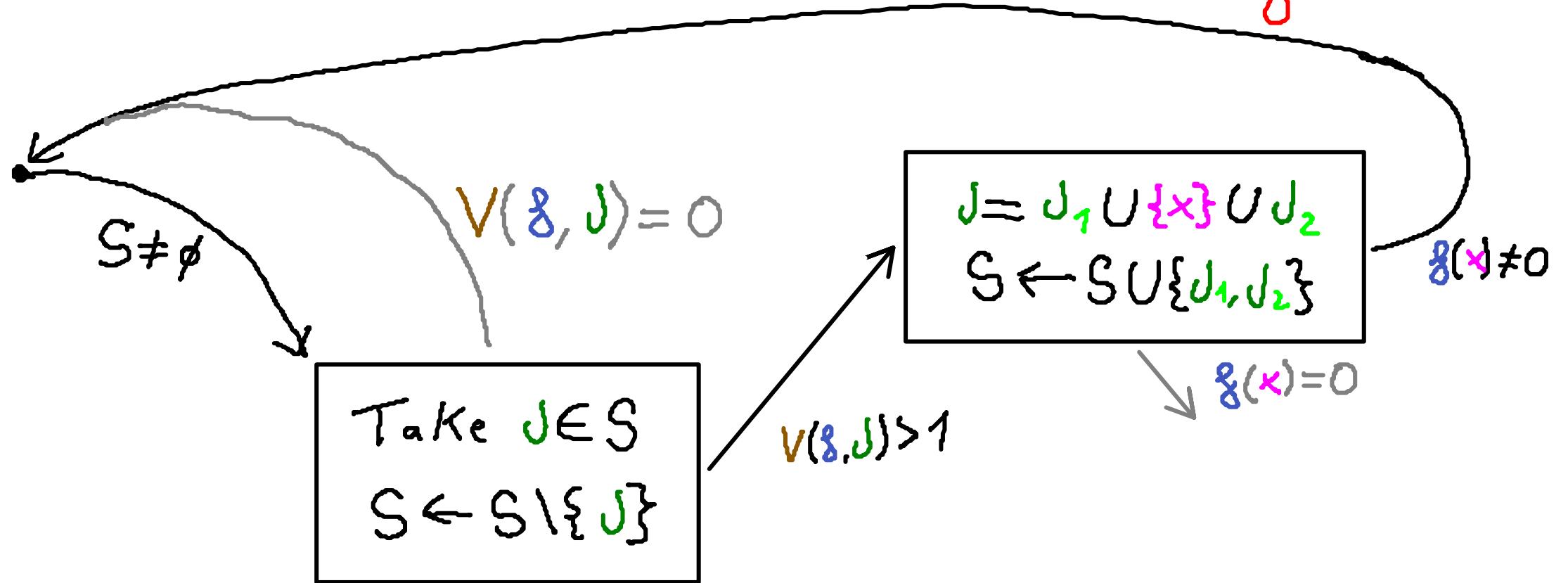
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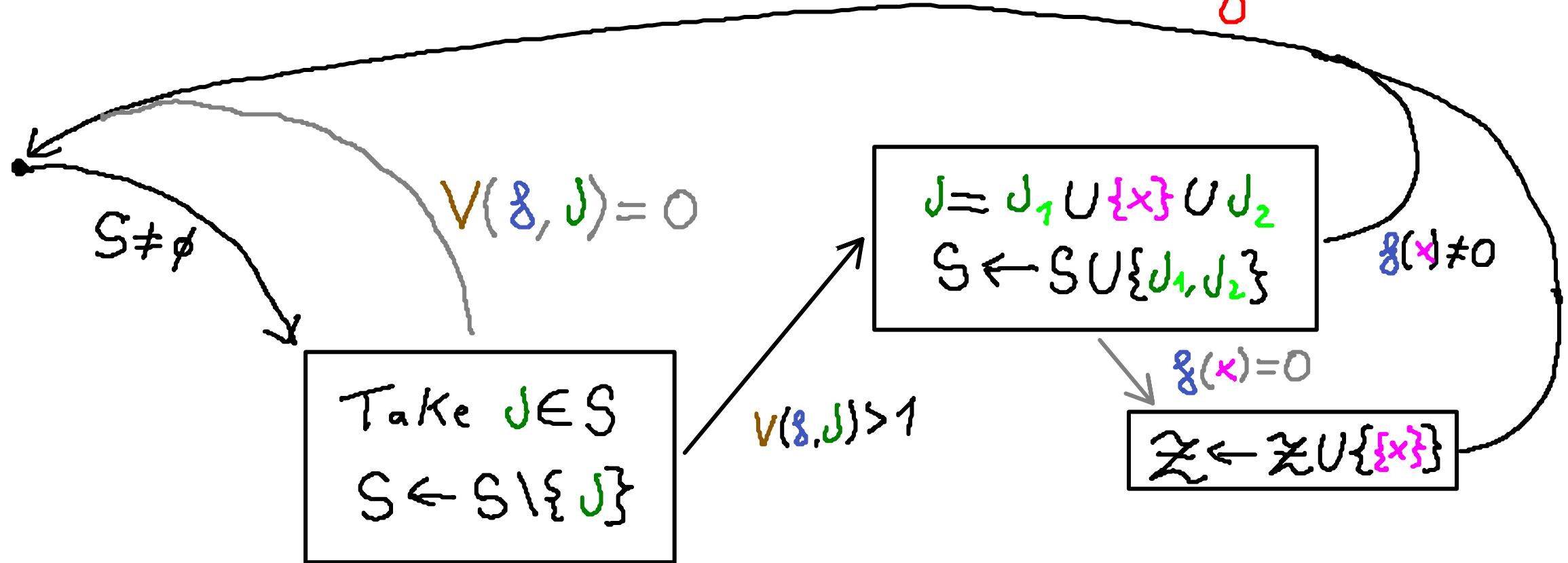
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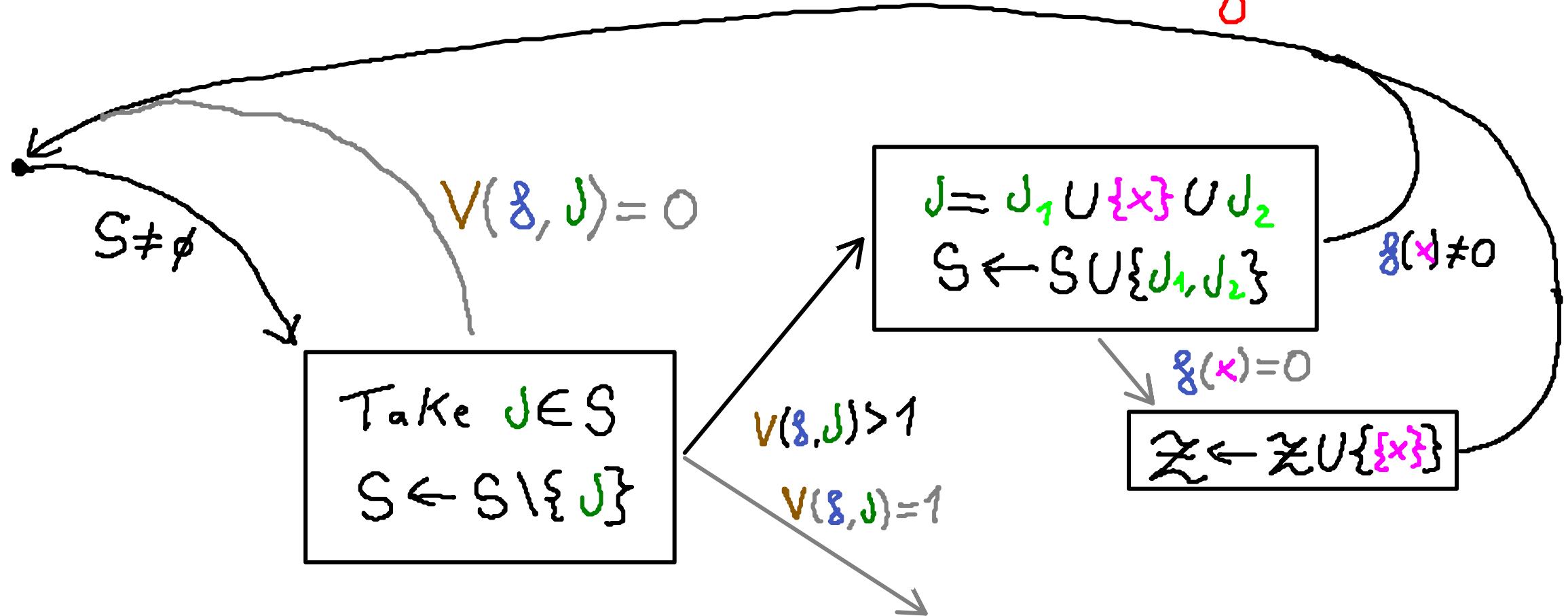
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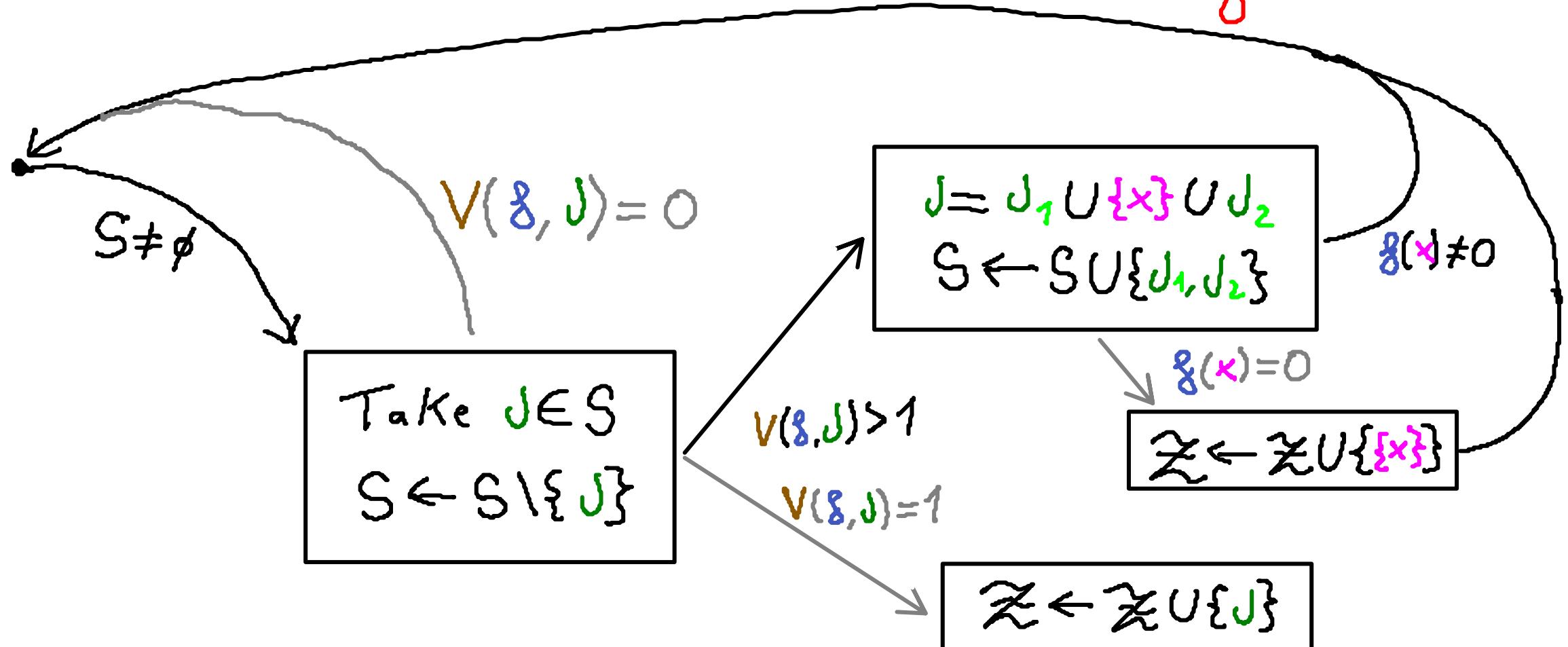
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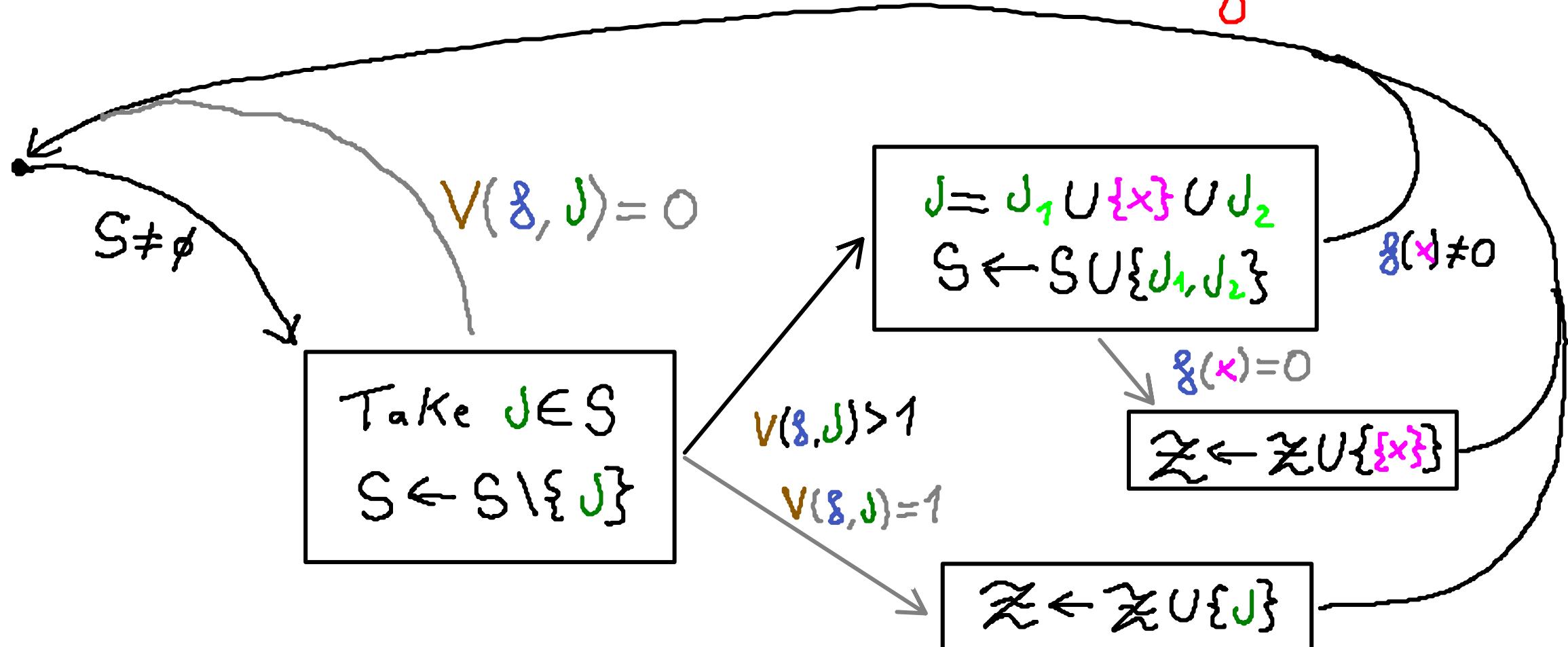
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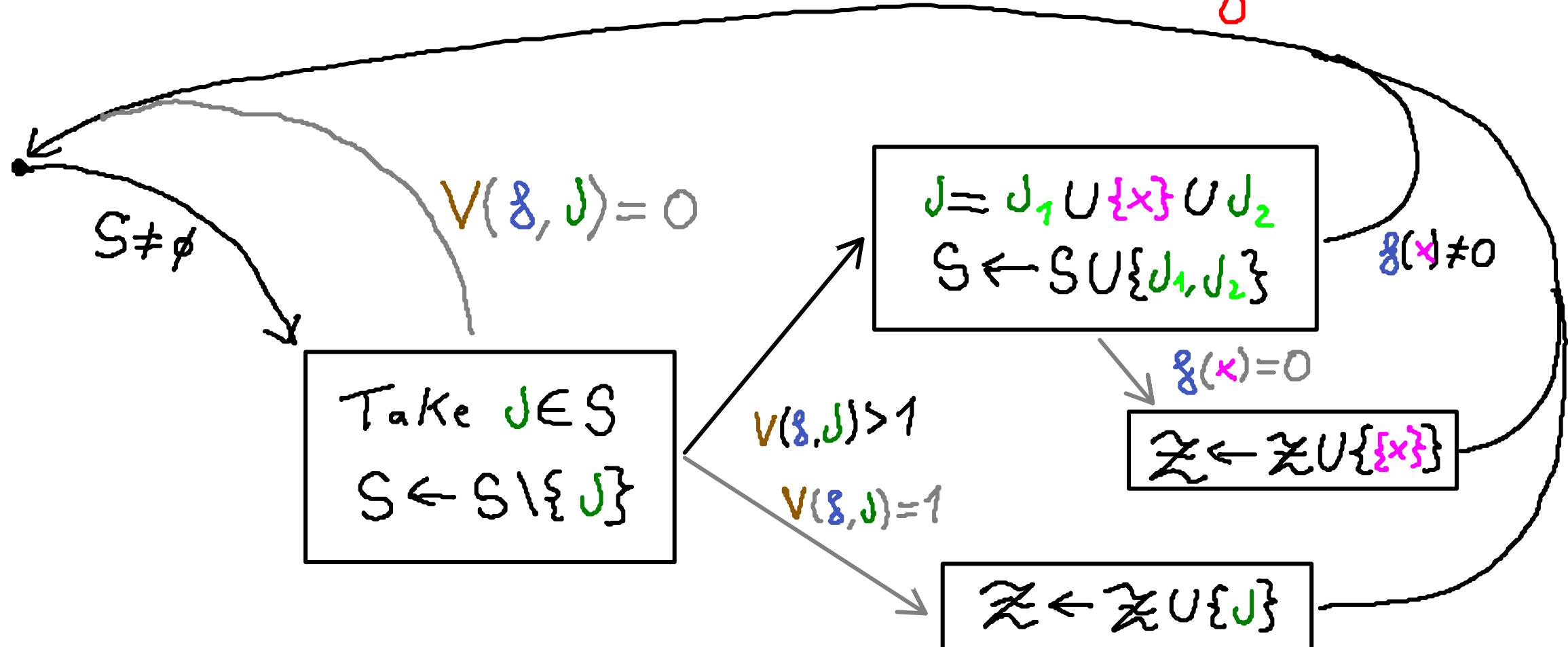
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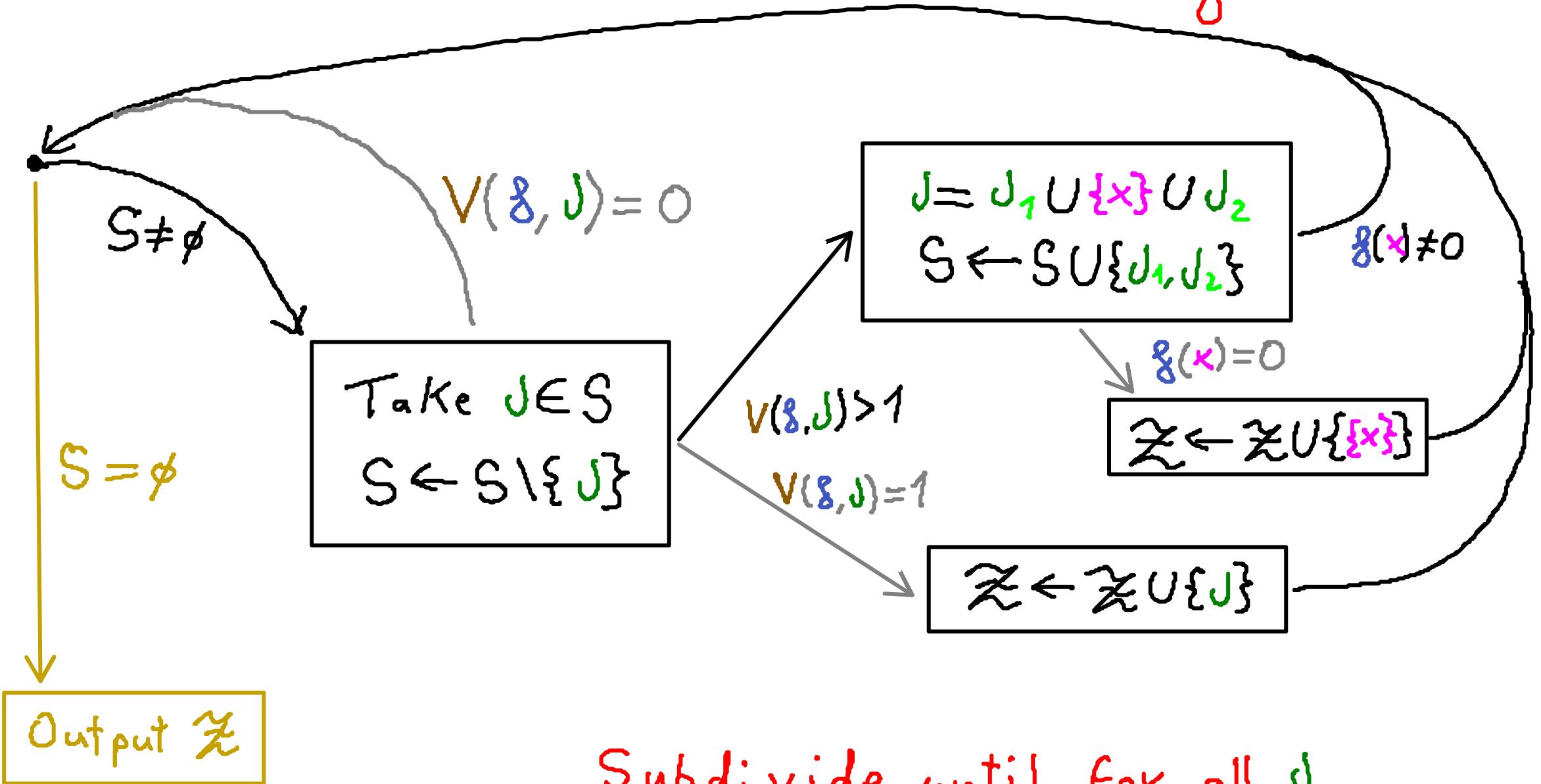
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Subdivide until for all J ,
 $V(g, J) \leq 1!$

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DESCARTES SOLVER

seems to behave faster in practice!

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Can we explain this?

Real Root Isolation V:

Beyond pessimism

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Beyond pessimism

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(Goldstine & von Neumann, 1951)
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↓
(Roughgarden, 2021)

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↑
Many choices of randomness 😱

Beyond pessimism I:
Uniform Random Bit Polynomials
& A SIMPLE MAIN THEOREM

Beyond pessimism I: Uniform Random Bit Polynomials & A SIMPLE MAIN THEOREM

$$F = \sum_{k=0}^d F_k X^k$$

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On average, DESCARTES is almost optimal!

Beyond pessimism II: Random Bit Polynomials

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uniformity of F : $u(F) := \ln(w(F)(1 + 2^{\gamma(F)+1}))$

Beyond pessimism III: MAIN THEOREM

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IF $\gamma = \Omega(d)$, almost like reading!

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MAIN THEOREM

MAIN THM (Ergür, T-C, Tsigaridas)

$$\mathbb{E} \text{cost}(\text{DESCARTES}, F) = \tilde{O}_B(d^2 + d\gamma)(1 + u(F))^4$$

Note: F uniform $\Rightarrow u(F) = 0$

Claim: For many cases, $u(F) = O(1)$

IF $\gamma = \Omega(d)$, almost like reading!

On average, DESCARTES is almost optimal!

Beyond pessimism IV:

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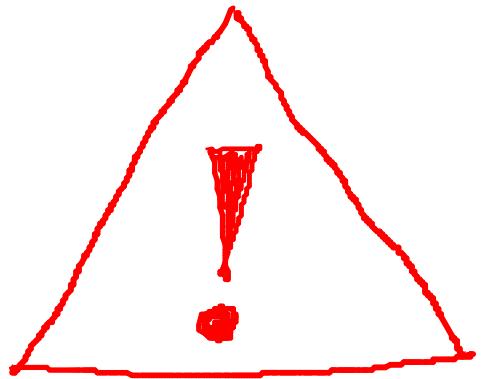
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Our random model is flexible!



LOTS OF DETAILS
WILL BE OMITTED

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the run-time of DESCARTES?

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$\gamma(g) \leftarrow$ DESCARTES' Computation Tree

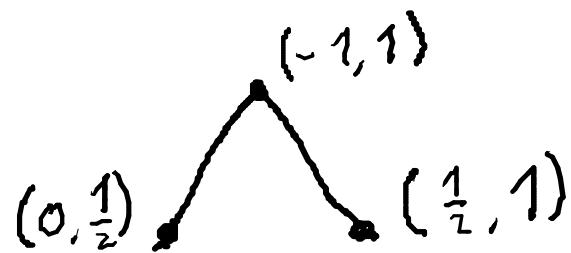
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$\langle -1, 1 \rangle$

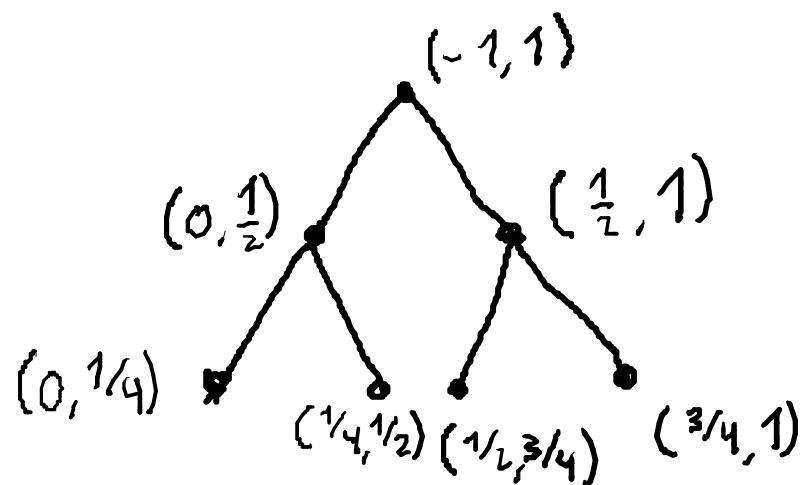
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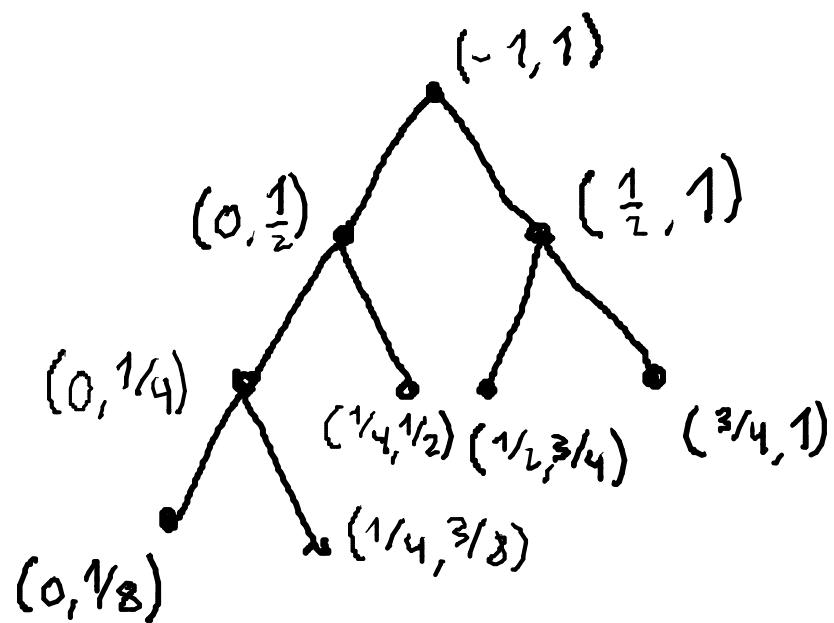
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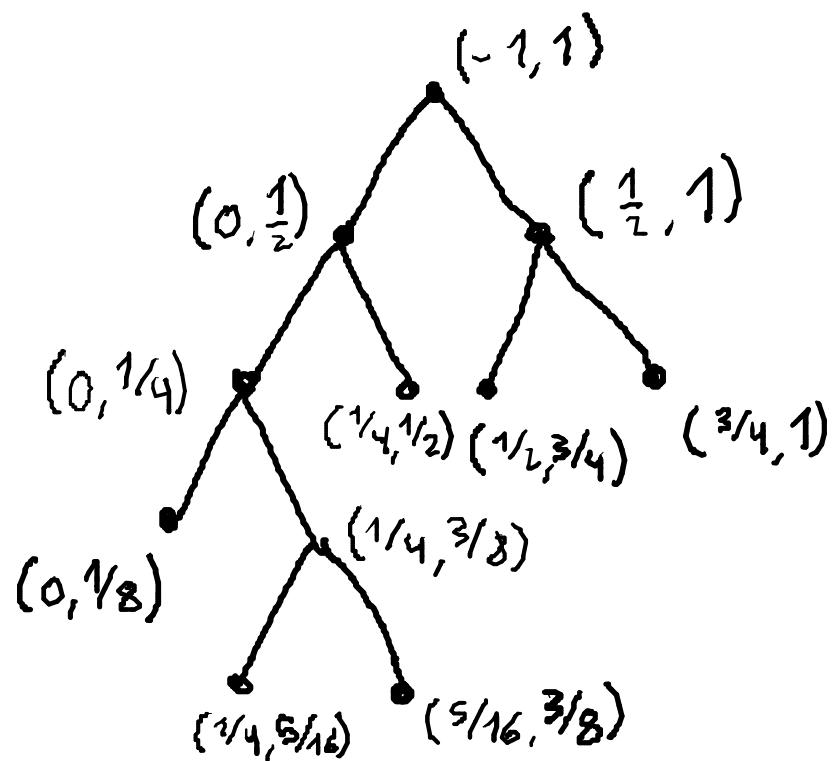
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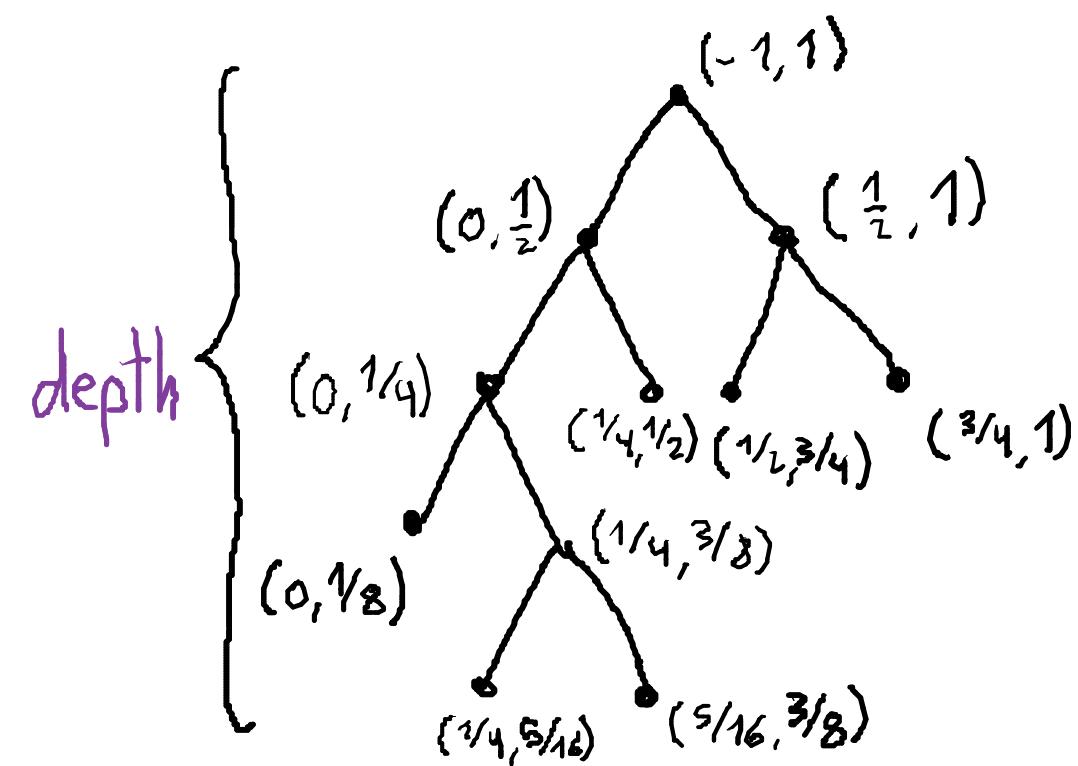
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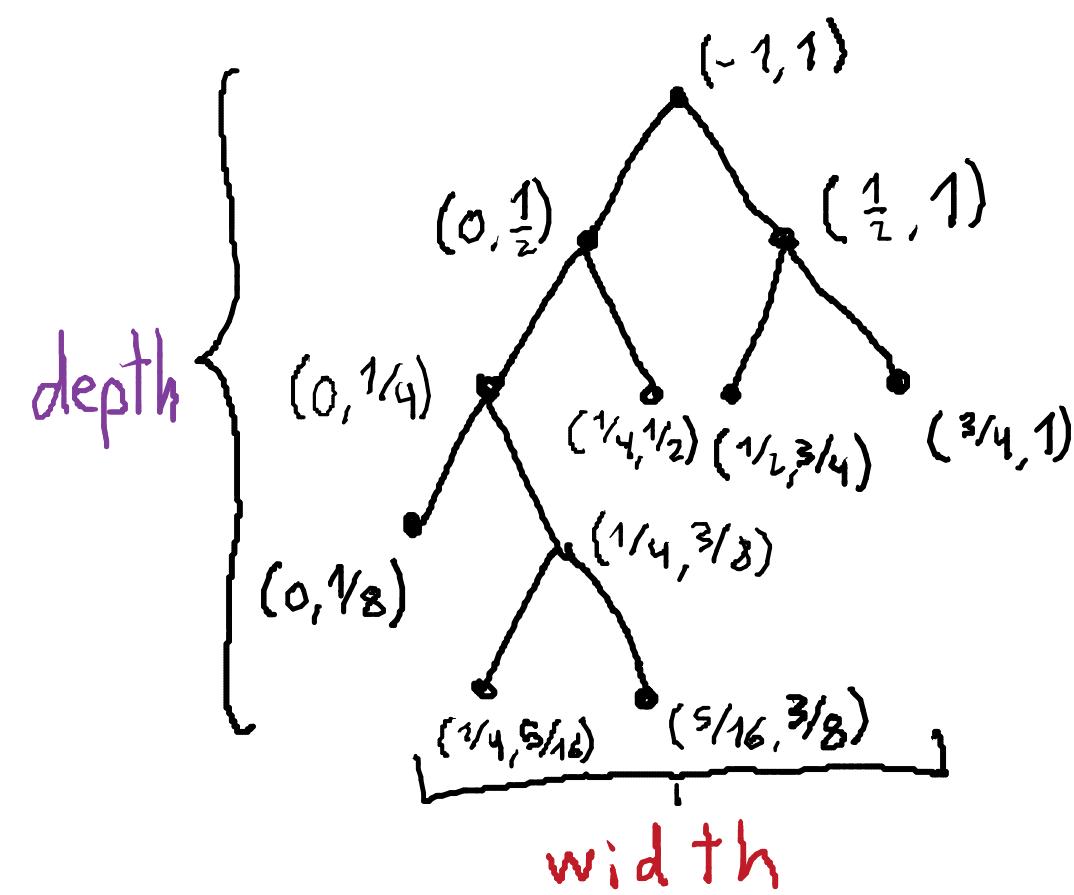
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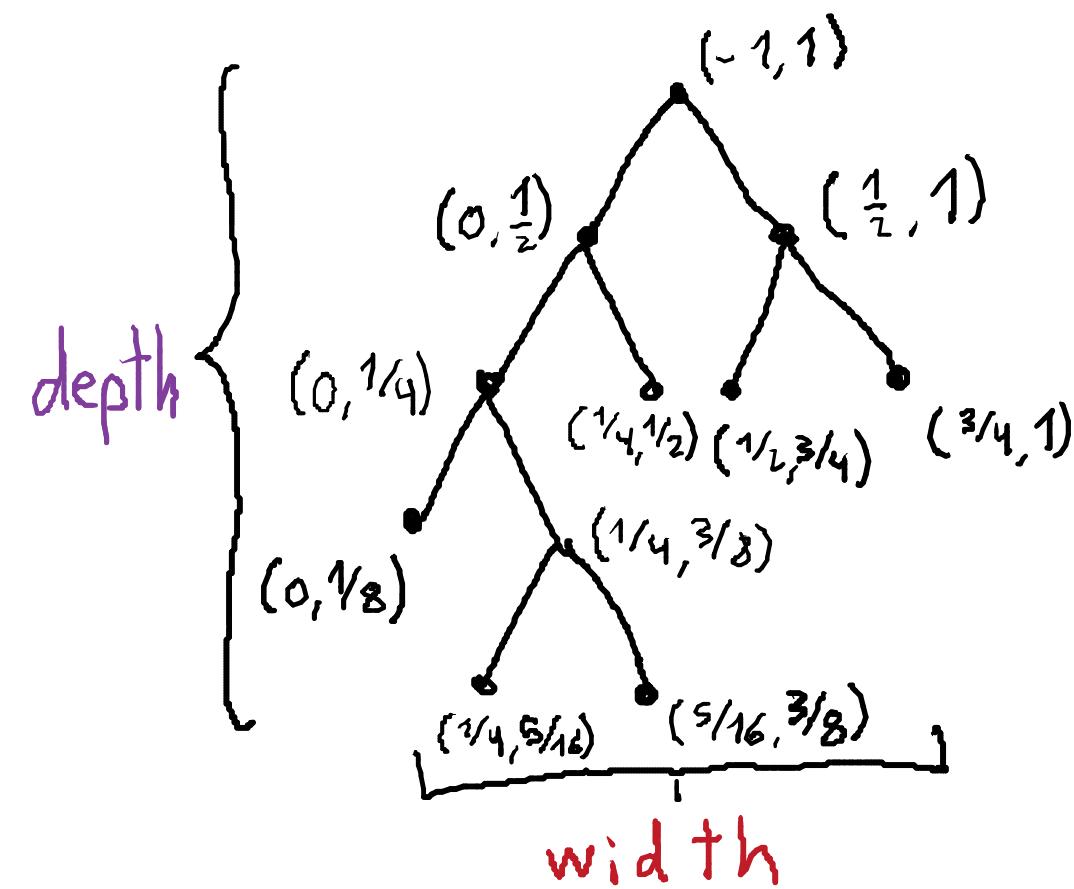


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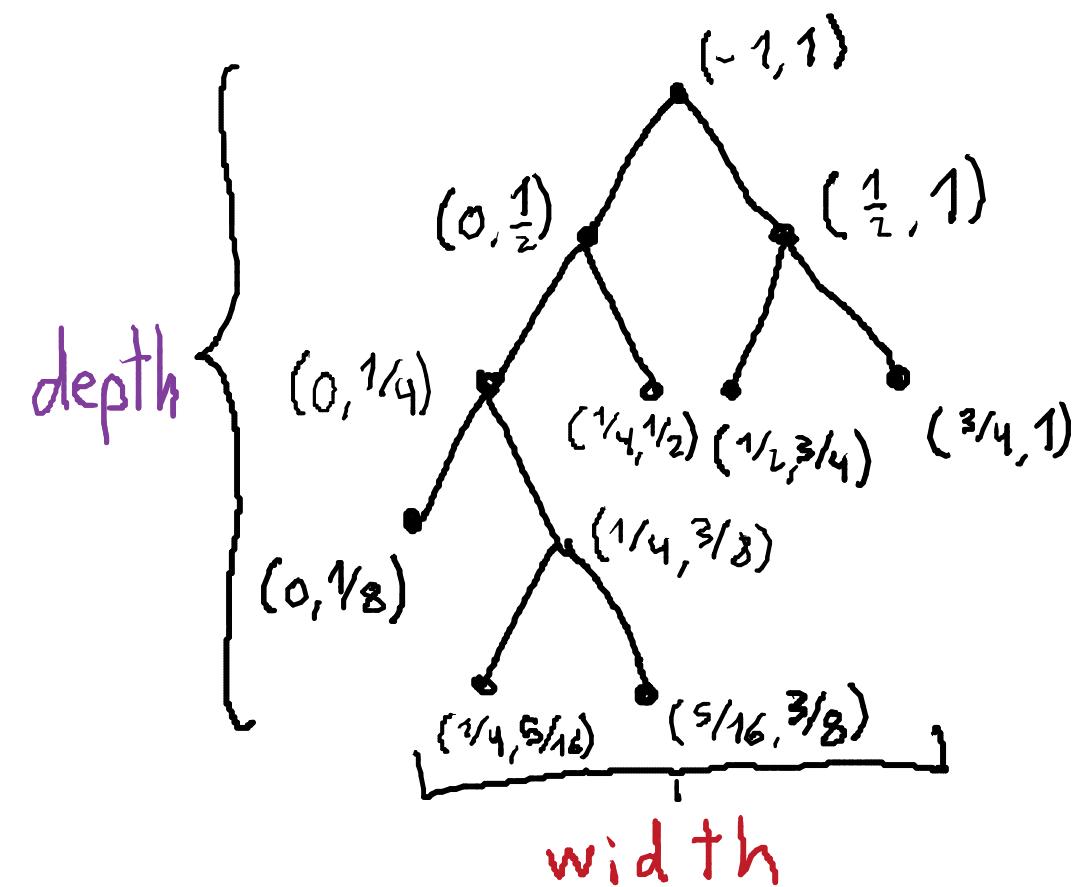
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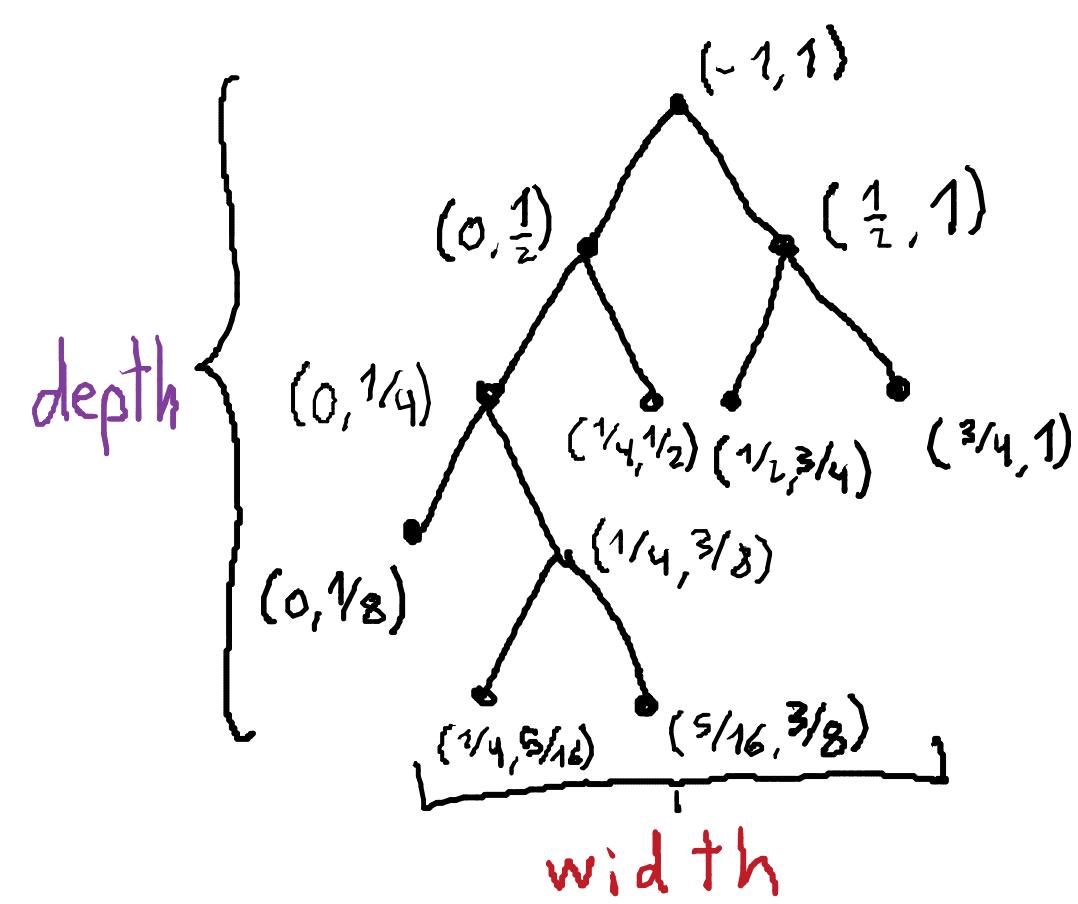
\wedge

$$\mathcal{O}(\text{d} \cdot \text{width} \gamma(g) \cdot \text{depth} \gamma(g) + d^2 \cdot \text{width} \gamma(g) \cdot \text{depth}^2 \gamma(g))$$



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 $\text{cost}(\text{DESCARTES}, g)$
 \propto
 $O(d \gamma \text{width} \gamma(g) \text{depth} \gamma(g))$
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I.e. size of $\gamma(g)$ bounds
run-time of DESCARTES!

The Ingredients of the Analysis I: Condition Number

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$\frac{1}{C(g)} \sim \left\{ \begin{array}{l} \text{How much I have to perturb} \\ \text{the coefficients of } g \text{ so that} \\ \tilde{g} \text{ has a singular root in } [-1,1] \end{array} \right.$

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$$\text{depth } \gamma(g) \leq 5 + \log d + \log C(g)$$

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Here $\rightarrow 1/\lambda$

no

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... but we don't have a continuous dist.

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We can use our old cont. toolbox!

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... AND THAT'S WHY DESCARTES WORKS SO WELL

Muito Obrigado

pela Atenção!

¡Muchas Gracias

por su Atención!