

A Measure of Space for Computing over the Reals

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Plan of the Talk

- The BSS Model of Computation over the Reals
- Michaux's Result - Computing in Constant Space
- Koiran's Weak Model
- A Weak Measure of Space

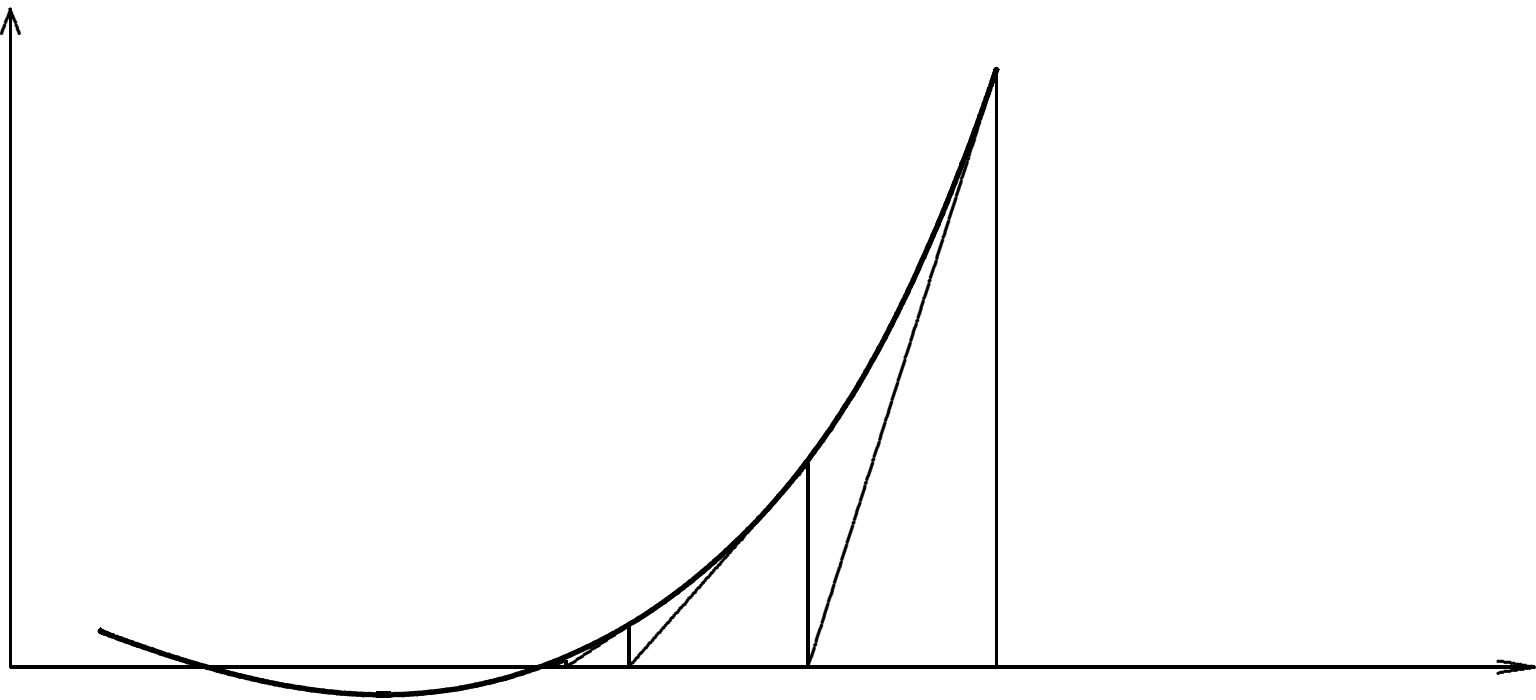
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Example:



Newton's Method for finding a zero of a function.

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- some **shift nodes** move the scanning head on the tape,
- some **copy nodes** duplicate the content of some cells.

Input - Output

Convention: One Input Tape, one Work Tape, one Output Tape.

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- Decision Problems - or Languages - are subsets of \mathbb{R}^* .

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- The set of points that converge under Newton's algorithm is undecidable.

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- $EXP_{\mathbb{R}}$: subsets of \mathbb{R}^* decided in exponential time.

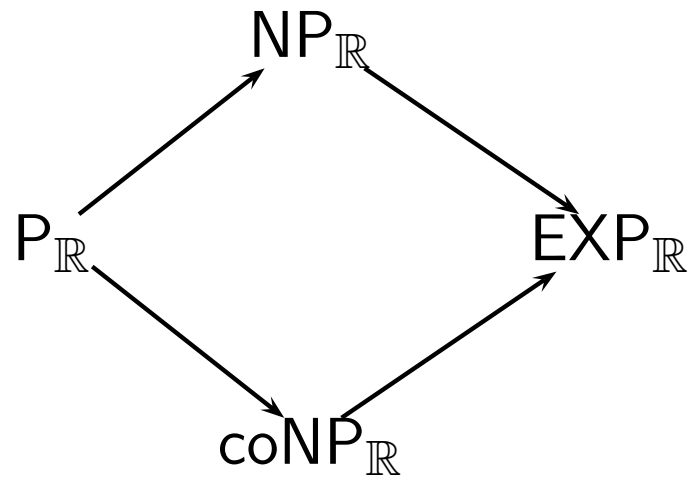
Complexity (2)

There exist natural $\text{NP}_{\mathbb{R}}$ and $\text{coNP}_{\mathbb{R}}$ -complete problems.
Ex: **4FEAS _{\mathbb{R}}** (existence of a zero for a real polynomial of degree 4) is $\text{NP}_{\mathbb{R}}$ -complete (reductions in $\text{P}_{\mathbb{R}}$).

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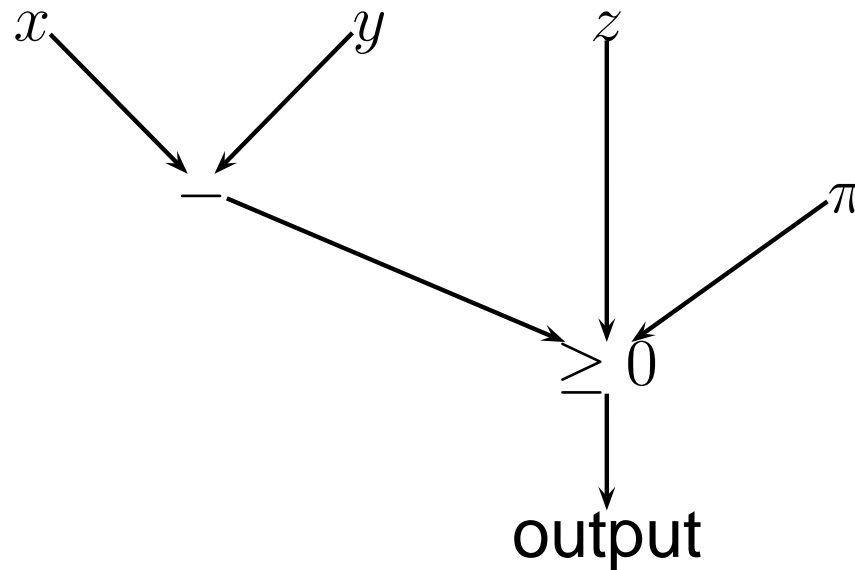
Inclusions:



Question: $\text{P}_{\mathbb{R}} = \text{NP}_{\mathbb{R}}$?

Algebraic Circuits

An algebraic circuit \mathcal{C} computes $F_{\mathcal{C}} : \mathbb{R}^n \rightarrow \mathbb{R}^m$.



if $x - y \geq 0$ then z else π , $n = 3, m = 1$

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there exist natural $P_{\mathbb{R}}$ -complete problems (reductions in $NC_{\mathbb{R}}^2$).

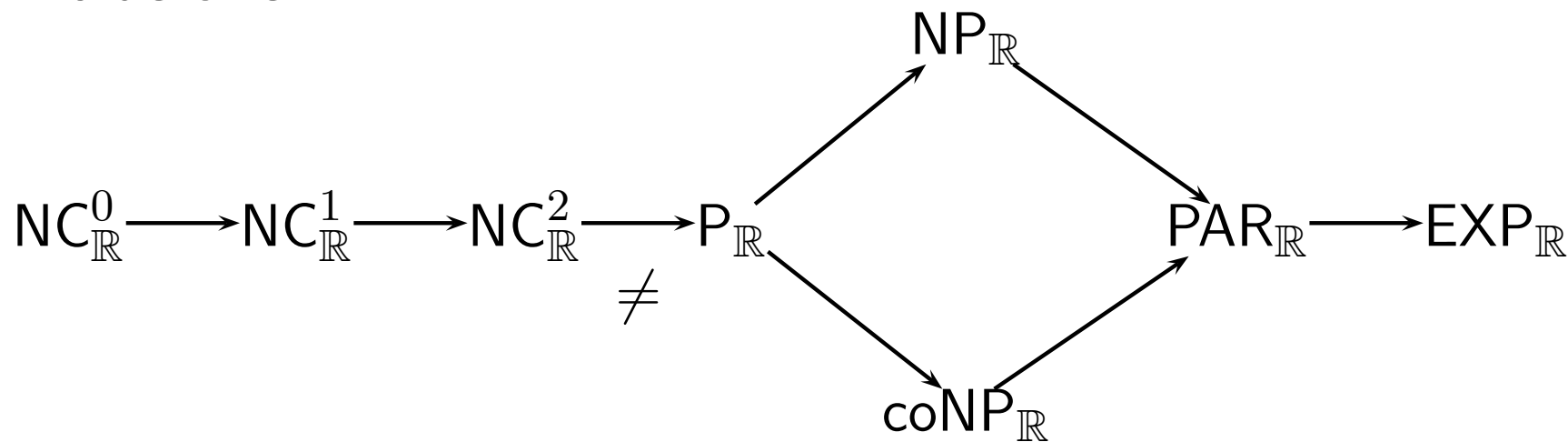
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Inclusions:



Questions of Space

Unit measure of space: # of tape cells used.

Michaux's Result

Let $L \subseteq \mathbb{R}^*$ be a language decided in bounded time by a machine M . There exists $k \in \mathbb{N}$ and a machine M' deciding L in bounded time and working space less than k .

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$M' \in \text{NC}_{\mathbb{R}}^2$. Existence of a more natural class?

Michaux's Result: Computed Values

Let M be a machine, with

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Lemma:

On any input $x_1, \dots, x_n \in \mathbb{R}^n$, at any computation step k , any non-empty cell e_l on the work tape holds the evaluation of a rational fraction $f_{l,k} \in \mathbb{Z}(X_1, \dots, X_{n+m})$ on $(x_1, \dots, x_n, A_1, \dots, A_m)$.

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Simulation of one **arithmetical** computation step: symbolic binary computation.

Simulation of one **branch** step: numerical evaluation of the arguments, and comparison.

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Weak cost of $y * z$: max of the **degrees** and of the **coefficient heights** of f_y and f_z .

Complexity (4)

Lemma: $L \in P_W$ if and only if:

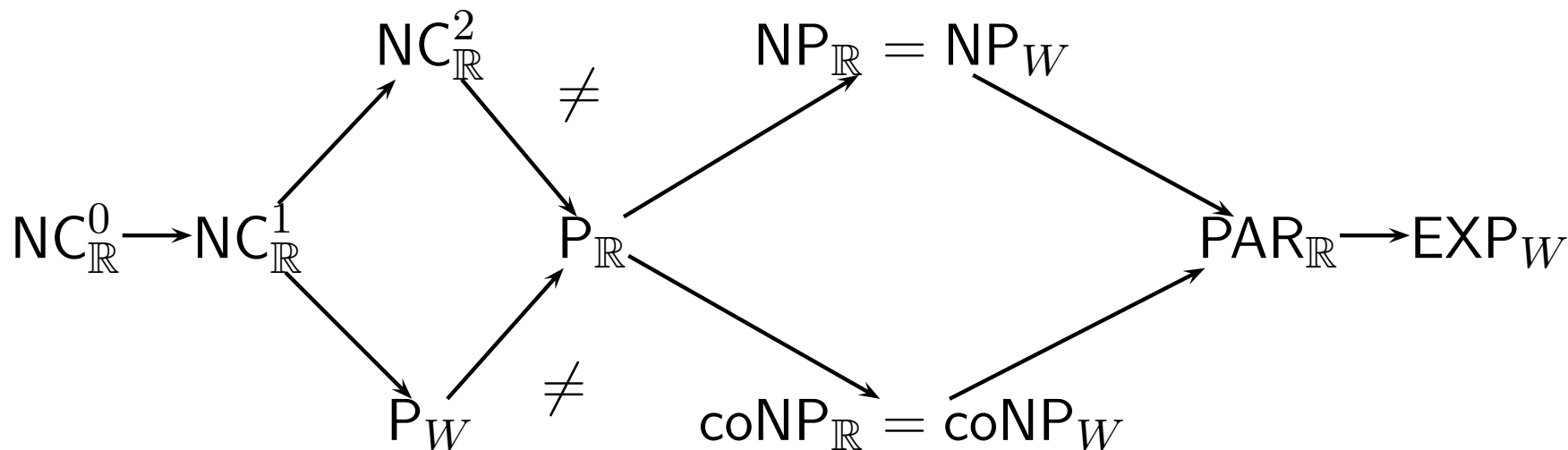
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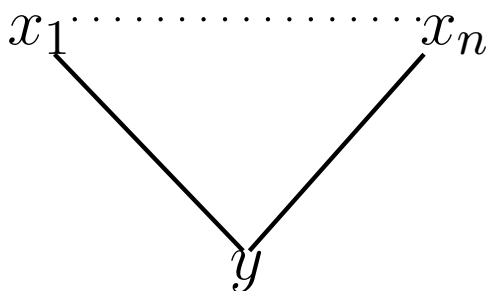
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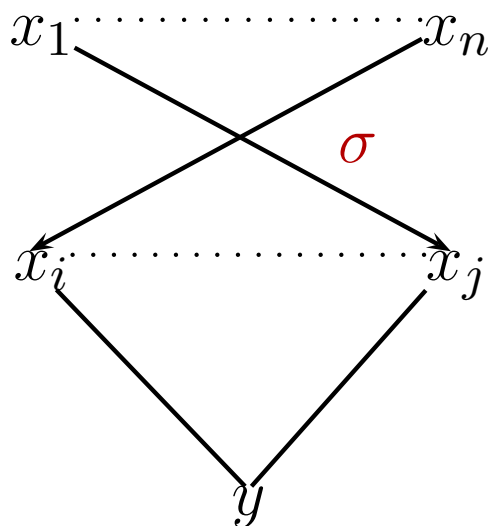


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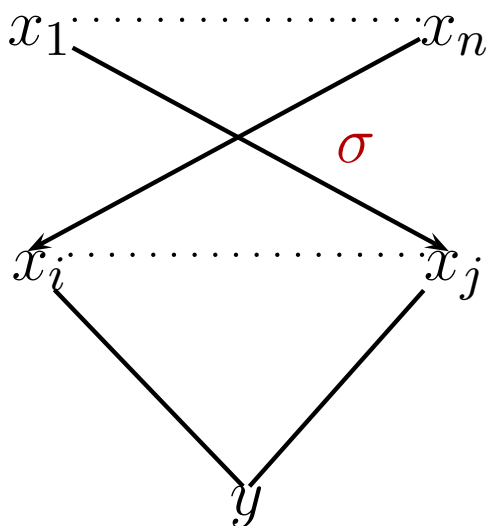
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PB 2: computation of a minimal tree \rightarrow factorization problem.

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$|y|_W$ is the size of a **explicit** (sequence of monomials) boolean description of f_y , modulo the permutation.

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- The algorithm M' is in **LOGSPACE_W**.
- Michaux's simulation of M' is in **LOGSPACE_W**.
- There exists some problems decidable in bounded time, not decidable in constant weak space.

$P_{\mathbb{R}}$ -completeness

Theorem: $\text{RCDP}_{\mathbb{R}}$ (Real Circuit Decision Procedure) is $P_{\mathbb{R}}$ -complete under LOGSPACE_W reductions.

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Proof: P-completeness of the Boolean Circuit Decision Problem under LOGSPACE-reductions.

Structural Complexity

Theorem:

- $\text{LOGSPACE}_W \subset P_W \cap \text{NC}_{\mathbb{R}}^2$.
- $\text{PSPACE}_W \subset \text{PAR}_{\mathbb{R}}$.

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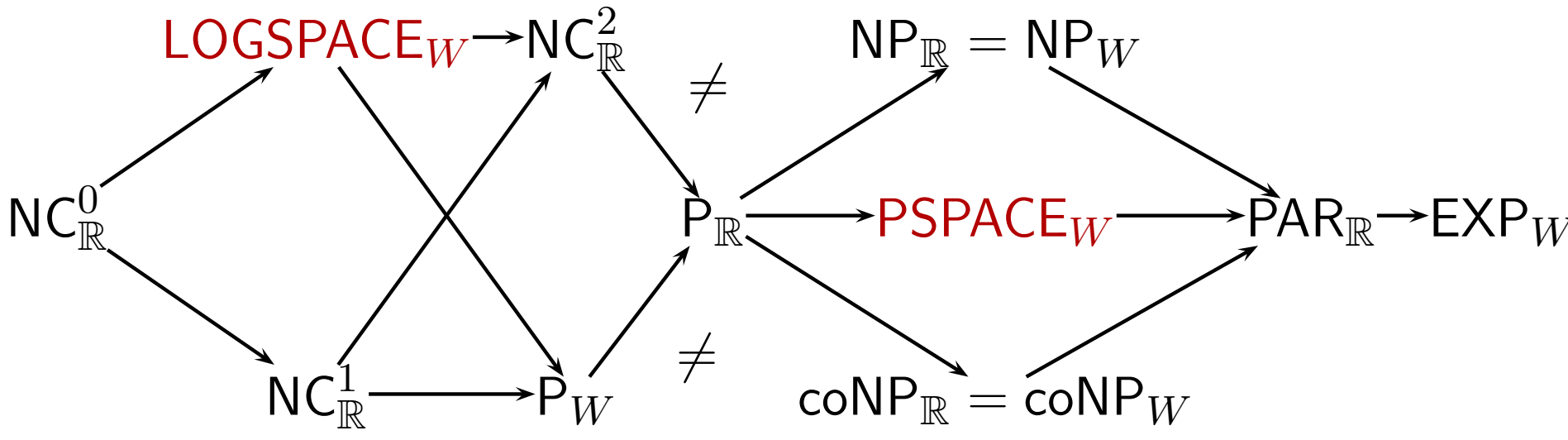
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- $\text{PSPACE}_W \subset \text{PAR}_{\mathbb{R}}$.

Proof:

- $\text{LOGSPACE}_W \subset P_{\mathbb{R}}$: enumeration of all configurations.
- $\text{LOGSPACE}_W \subset P_W$: Koiran's Lemma.
- $\text{LOGSPACE}_W \subset \text{NC}_{\mathbb{R}}^2$: $P_{\mathbb{R}}$ -uniform construction of the configuration graph of a LOGSPACE_W machine, and graph reachability in NC^2 .
- $\text{PSPACE}_W \subset \text{PAR}_{\mathbb{R}}$: Corollary.

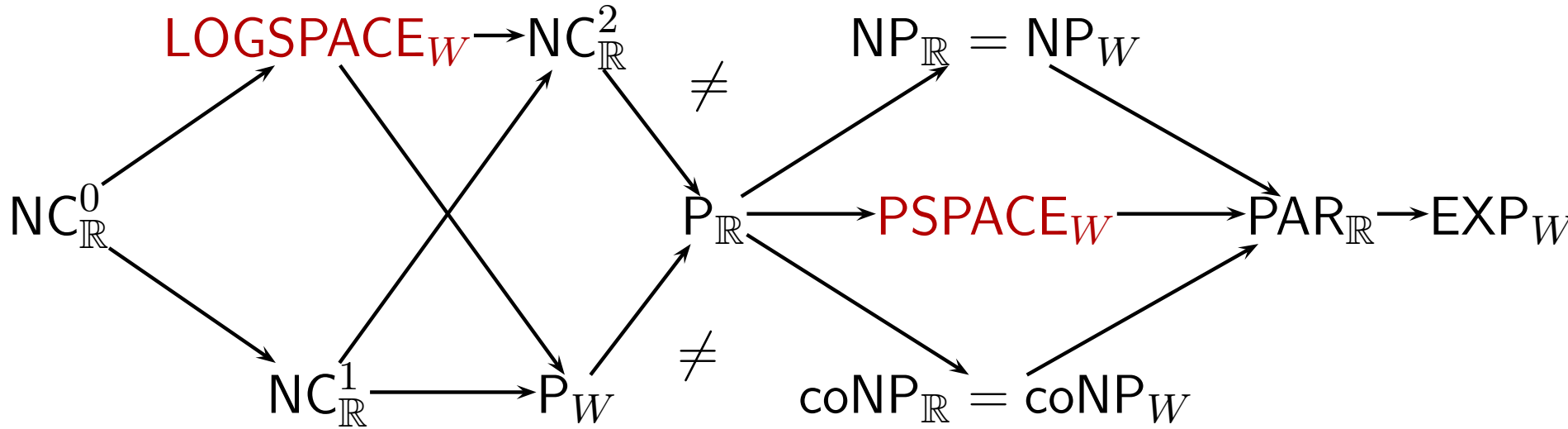
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Conjectures:

- $NC_{\mathbb{R}}^1 \not\subseteq LOGSPACE_W$, $LOGSPACE_W \not\subseteq NC_{\mathbb{R}}^1$.
- $PSPACE_W = PAR_{\mathbb{R}}$