

LORIA - INRIA, NANCY (FRANCE)
ENS PARIS-SACLAY (FRANCE)

Descriptive complexity on non-Polish spaces

Antonin CALLARD & Mathieu HOYRUP

STACS, March 2020



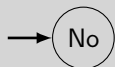
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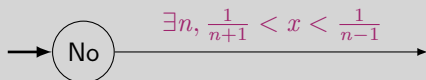
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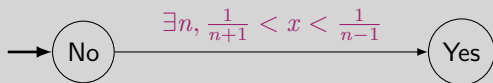
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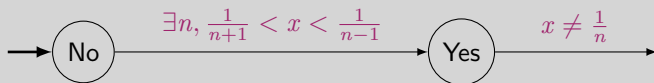
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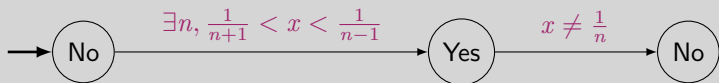
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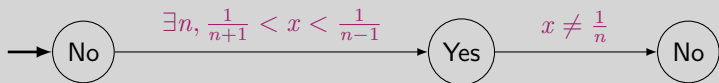
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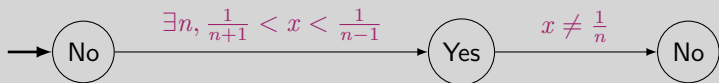
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QUESTION:

Are those two approaches always equivalent?

OUTLINE

Formalization of the problem

Countably-based spaces

Non countably-based spaces

Conclusion

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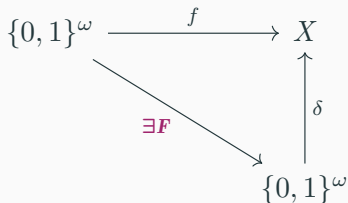
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Key idea: semi-decidable problems are similar to (effective) open sets!

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- $[\Pi_1^0]$: problems P verifying $P^c \in [\Sigma_1^0]$, ie. :
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$\Delta_2^0 = \Sigma_2^0 \cap \Pi_2^0$,

where $\Sigma_2^0 = \{U_{n \in \mathbb{N}} U_n \setminus U'_n\}$.

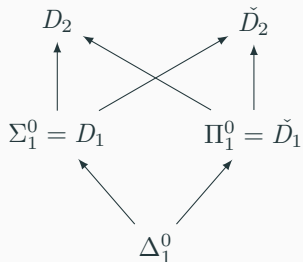
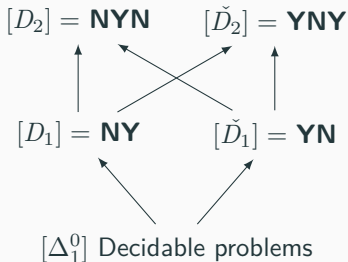
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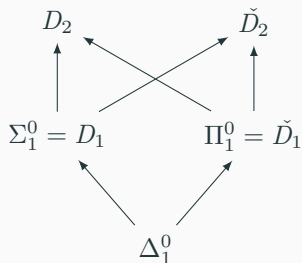
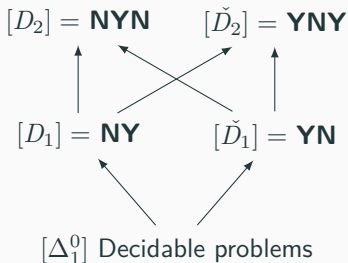
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For $\Gamma \in \{D_\eta : \eta < \omega_1\}$, when does one have $\Gamma = [\Gamma]$?

Countably-based spaces

Countably-based spaces: matching of the two hierarchies

Countably-based: a countable number of open sets describes the topology.

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Theorem 2.1: M. De Brecht, 2012

For (X, δ) a countably-based represented space,

$$S \in \Gamma \iff S \in [\Gamma]$$

Countably-based spaces: matching of the two hierarchies

Countably-based: a countable number of open sets describes the topology.

[New] - Theorem 2.2: Hierarchies coincide in an effective way

For (X, δ) a countably-based represented space,

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in an effective and uniform way.

Countably-based: about hardness

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[New] - Theorem 2.6: Hardness-criterion

$$S \notin D_\eta \iff S \text{ is } \check{D}_\eta\text{-hard.}$$

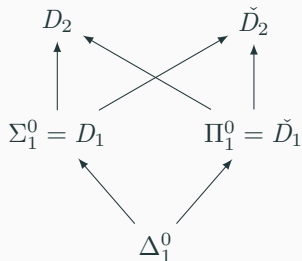
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Theorem 2.7: Hardness-criterion

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Non countably-based spaces

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- The coefficients p_0, \dots, p_n such that

$$P = p_0 + p_1X + \dots + p_nX^n.$$

Property 3.1: Complexity of S_1 (Part 1)

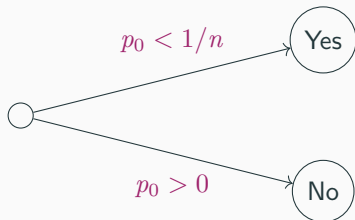
$$S_1 = \left\{ P \in \mathbb{R}[X] : p_0 = 0 \text{ or } p_0 > \frac{1}{\deg(P)} \right\} \in [D_2]$$

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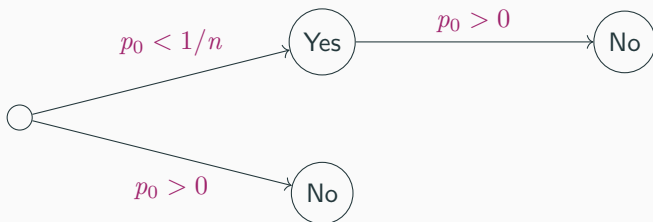


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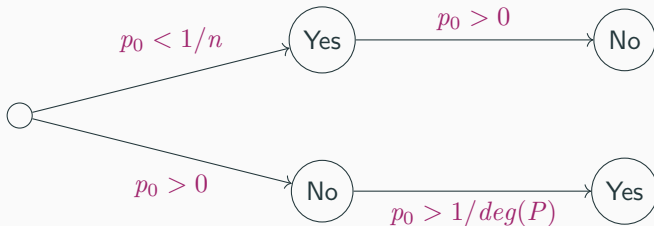


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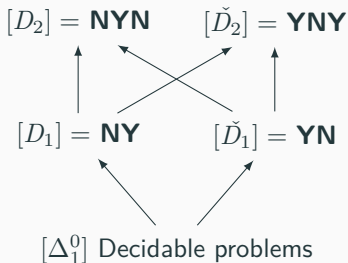
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Conclusion : $S_1 \in [D_2]$ but $S_1 \notin D_2$

Complexity of S_1

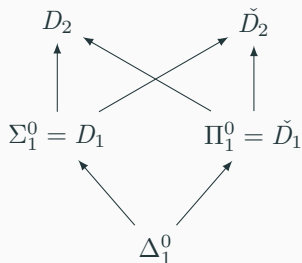
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To go even further... Introduce S_2 .

$$S_2 = \left\{ \frac{1}{k_1} + \frac{X^{k_1}}{k_2} + \dots + \frac{X^{k_n}}{k_{n+1}} : \exists n, k_1 < \dots < k_{n+1}, k_n \text{ even} \right\}$$

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[New] - Theorem 3.3: Complexity of S_2

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2. But $\forall \eta < \omega_1, S_2 \notin D_\eta$.

To go even further... Introduce S_2 .

$$S_2 = \left\{ \frac{1}{k_1} + \frac{X^{k_1}}{k_2} + \dots + \frac{X^{k_n}}{k_{n+1}} : \exists n, k_1 < \dots < k_{n+1}, k_n \text{ even} \right\}$$

[New] - Theorem 3.3: Complexity of S_2

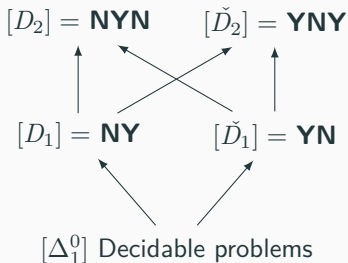
1. $S_2 \in [D_2]$.
2. But $\forall \eta < \omega_1, S_2 \notin D_\eta$.

Conclusion : $S_2 \in [D_2]$ but $S_2 \notin \bigcup_{\eta < \omega_1} D_\eta$

Complexity of S_2

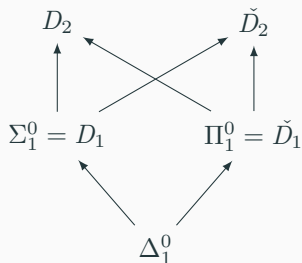
$[\Delta_2^0]$ = converging sequences

...



$\Delta_2^0 = \Sigma_2^0 \cap \Pi_2^0$

...



What happened?

Why is this happening?

Guess: The difference is related to the mismatch between **topological** and **sequential** aspects.

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Why do we think that?

[New] - Theorem 3.4: $[\Gamma]$ depends on sequentiality

For a coPolish space X , 1. and 2. are equivalent.

1. Closure and sequential closure coincide on X .
2. For every $n < \omega$, $A \in [D_n]$ iff $A \in D_n$.

Consequences - About hardness

Actually, the definition of Γ -hardness involves $\{0, 1\}^\omega$.

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For example:

[New] Theorem 3.5: Countably-based spaces

$$A \notin D_\eta \iff A \text{ is } \check{D}_\eta\text{-hard.}$$

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[New] Theorem 3.6: co-Polish spaces

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For example:

[New] Theorem 3.6: co-Polish spaces

$$A \notin [D_\eta] \iff A \text{ is } \check{D}_\eta\text{-hard.}$$

DST works better with algorithmic complexity (and
represented spaces) rather than with topology.

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Let us recap:

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Let us recap:

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Questions?