

# On the Complexity of Negative Quantification

Aleksy Schubert, Paweł Urzyczyn, Konrad Zdanowski

Types 2014, Paris, May 14th, 2014

## Prenex normal form

In classical logic:

Every formula is classically equivalent to one of the form:

$$Q_1 x_1 Q_2 x_2 \dots Q_k x_k \cdot \text{Body}(x_1, x_2, \dots, x_k),$$

where *Body* has no quantifiers.

Classification:

Formulas may be classified according to the quantifier prefix, e.g. universal ( $\forall^*$ ) formulas are in  $\Pi_1$ , and  $\Pi_2$  is  $\forall^* \exists^*$ , etc.

In intuitionistic logic:

The prenex fragment is decidable in Pspace.

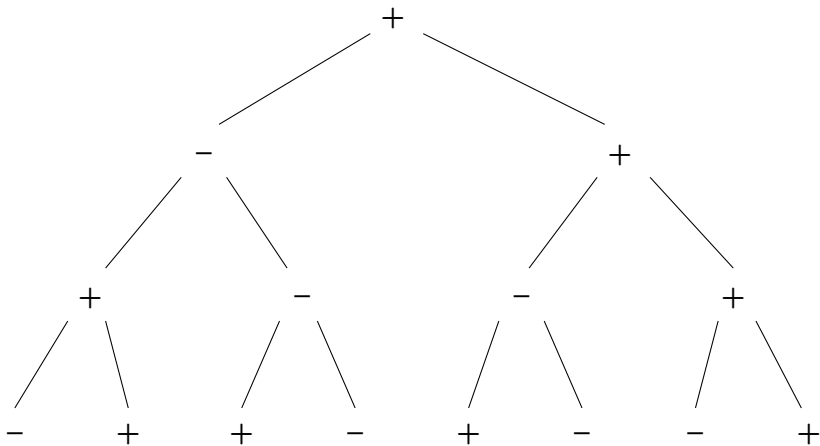
## Mints Hierarchy

Can we restore the prenex classification in intuitionistic logic?

Grigori Mints (1968): Yes, consider the quantifier prefix  
a formula *would get*, if classically normalized.  
(What counts is the alternation of quantifiers.)

For example,  $\forall$  quantifiers occurring at *positive* positions  
will remain  $\forall$  in the prefix.

## Positive and Negative



## Mints Hierarchy

$\Pi_1$  – All quantifiers at positive positions.

$\Sigma_1$  – All quantifiers at negative positions.

$\Pi_2$  – One alternation: some negative quantifiers  
in scope of some positive ones.

$\Sigma_2$  – One alternation: some positive quantifiers  
in scope of some negative ones.

And so on.

## *The language we study*

To make things simpler, we consider first-order formulas

- ▶ with universal quantifiers and implications;
- ▶ without function symbols.

This fragment is known to be undecidable.

# Examples

$(\forall x P(x)) \rightarrow Q$  is a  $\Sigma_1$  (negative) formula;

$((\forall x P(x)) \rightarrow Q) \rightarrow R$  is a  $\Pi_1$  (positive) formula;

$(\forall x (\forall y P(x) \rightarrow \forall z S(y, z))) \rightarrow R$  is a  $\Sigma_2$  formula.

## Complexity of Mints Hierarchy

- ▶  $\Pi_2$  and  $\Sigma_2$  are undecidable.
- ▶  $\Pi_1$  is **co-2-NExptime**-hard  
(conjecture: super-elementary)
- ▶  $\Sigma_1$  (this talk):
  - **Expspace**-complete in general;
  - **co-Nexptime**-complete with monadic predicates.



# $\Sigma_1$ decision problem

Negative formulas are of shape

$$\varphi_1 \rightarrow \cdots \rightarrow \varphi_n \rightarrow \sigma,$$

where  $\sigma$  is an atomic formula and  $\varphi_1, \dots, \varphi_n$  are positive.

A  $\Sigma_1$  decision problem can be seen as

$$\varphi_1, \dots, \varphi_n \vdash \sigma$$

(Derive an atom from positive assumptions.)

## Upper bound for $\Sigma_1$

Proofs in  $\Sigma_1$  may use universal assumptions, which can be instantiated. But introducing new variables is of no use.

If there is a proof then there is one where all eigenvariables are free variables of the original formula  $\varphi$ .

Such proof can be constructed in exponential space.  
(Every subformula of  $\varphi$  of size  $n$  has at most  $n^n$  instances.)

# Towards the lower bound

## Assumptions:

$$\forall xyz (P(x, y, z, 1) \rightarrow P(x, y, z, 0)),$$

$$\forall xy (P(x, y, 1, 0) \rightarrow P(x, y, 0, 1)),$$

$$\forall x (P(x, 1, 0, 0) \rightarrow P(x, 0, 1, 1)),$$

$$P(1, 0, 0, 0) \rightarrow P(0, 1, 1, 1).$$

**Goal:**  $P(0, 0, 0, 0)$

# Towards the lower bound

## Assumptions:

$$\forall xyz (P(x, y, z, 1) \rightarrow P(x, y, z, 0)),$$

$$\forall xy (P(x, y, 1, 0) \rightarrow P(x, y, 0, 1)),$$

$$\forall x (P(x, 1, 0, 0) \rightarrow P(x, 0, 1, 1)),$$

$$P(1, 0, 0, 0) \rightarrow P(0, 1, 1, 1).$$

## Goal: $P(0, 0, 0, 1)$

# Towards the lower bound

## Assumptions:

$$\forall xyz (P(x, y, z, 1) \rightarrow P(x, y, z, 0)),$$

$$\forall xy (P(x, y, 1, 0) \rightarrow P(x, y, 0, 1)),$$

$$\forall x (P(x, 1, 0, 0) \rightarrow P(x, 0, 1, 1)),$$

$$P(1, 0, 0, 0) \rightarrow P(0, 1, 1, 1).$$

## Goal: $P(0, 0, 1, 0)$

# Towards the lower bound

## Assumptions:

$$\forall xyz (P(x, y, z, 1) \rightarrow P(x, y, z, 0)),$$

$$\forall xy (P(x, y, 1, 0) \rightarrow P(x, y, 0, 1)),$$

$$\forall x (P(x, 1, 0, 0) \rightarrow P(x, 0, 1, 1)),$$

$$P(1, 0, 0, 0) \rightarrow P(0, 1, 1, 1).$$

## Goal: $P(0, 0, 1, 1)$

# Towards the lower bound

## Assumptions:

$$\forall xyz (P(x, y, z, 1) \rightarrow P(x, y, z, 0)),$$

$$\forall xy (P(x, y, 1, 0) \rightarrow P(x, y, 0, 1)),$$

$$\forall x (P(x, 1, 0, 0) \rightarrow P(x, 0, 1, 1)),$$

$$P(1, 0, 0, 0) \rightarrow P(0, 1, 1, 1).$$

## Goal: $P(0, 1, 0, 0)$

# Towards the lower bound

## Assumptions:

$$\forall xyz (P(x, y, z, 1) \rightarrow P(x, y, z, 0)),$$

$$\forall xy (P(x, y, 1, 0) \rightarrow P(x, y, 0, 1)),$$

$$\forall x (P(x, 1, 0, 0) \rightarrow P(x, 0, 1, 1)),$$

$$P(1, 0, 0, 0) \rightarrow P(0, 1, 1, 1).$$

## Goal: $P(0, 1, 0, 1)$



# Towards the lower bound

## Assumptions:

$$\forall xyz (P(x, y, z, 1) \rightarrow P(x, y, z, 0)),$$

$$\forall xy (P(x, y, 1, 0) \rightarrow P(x, y, 0, 1)),$$

$$\forall x (P(x, 1, 0, 0) \rightarrow P(x, 0, 1, 1)),$$

$$P(1, 0, 0, 0) \rightarrow P(0, 1, 1, 1).$$

**Goal:** a few more steps. . .

# Towards the lower bound

## Assumptions:

$$\forall xyz (P(x, y, z, 1) \rightarrow P(x, y, z, 0)),$$

$$\forall xy (P(x, y, 1, 0) \rightarrow P(x, y, 0, 1)),$$

$$\forall x (P(x, 1, 0, 0) \rightarrow P(x, 0, 1, 1)),$$

$$P(1, 0, 0, 0) \rightarrow P(0, 1, 1, 1).$$

## Goal: $P(1, 1, 1, 1)$

# Towards the lower bound

## Assumptions:

$$\forall xyz (P(x, y, z, 1) \rightarrow P(x, y, z, 0)),$$

$$\forall xy (P(x, y, 1, 0) \rightarrow P(x, y, 0, 1)),$$

$$\forall x (P(x, 1, 0, 0) \rightarrow P(x, 0, 1, 1)),$$

$$P(1, 0, 0, 0) \rightarrow P(0, 1, 1, 1).$$

## Goal:

The proof search is as difficult as rewriting **0000** into **1111**.

## Another example:

$$\forall xyzx'y'z' (D(x, x') \rightarrow D(y, y') \rightarrow D(z, z') \rightarrow \\ ((P(x', y', z', 3) \rightarrow P(x', y', z', 2)) \rightarrow P(x, y, z, 1)) \rightarrow P(x, y, z, 0))$$

$$\forall xyx'y' (D(x, x') \rightarrow D(y, y') \rightarrow \\ ((P(x', y', 3, 2) \rightarrow P(x', y', 2, 3)) \rightarrow P(x, y, 1, 0)) \rightarrow P(x, y, 0, 1))$$

$$\forall xx' (D(x, x') \rightarrow \\ ((P(x', 3, 2, 2) \rightarrow P(x', 2, 3, 3)) \rightarrow P(x, 1, 0, 0)) \rightarrow P(x, 0, 1, 1))$$

$$(P(3, 2, 2, 2) \rightarrow (2, 3, 3, 3)) \rightarrow P(1, 0, 0, 0) \rightarrow P(0, 1, 1, 1),$$

$$P(2, 2, 2, 2) \rightarrow P(1, 1, 1, 1), P(3, 3, 3, 3), D(2, 0), D(3, 1).$$

**Goal:**  $P(0, 0, 0, 0)$

## Another example:

$$\forall xyzx'y'z' (D(x, x') \rightarrow D(y, y') \rightarrow D(z, z') \rightarrow \\ ((P(x', y', z', 3) \rightarrow P(x', y', z', 2)) \rightarrow P(x, y, z, 1)) \rightarrow P(x, y, z, 0))$$

$$\forall xyx'y' (D(x, x') \rightarrow D(y, y') \rightarrow \\ ((P(x', y', 3, 2) \rightarrow P(x', y', 2, 3)) \rightarrow P(x, y, 1, 0)) \rightarrow P(x, y, 0, 1))$$

$$\forall xx' (D(x, x') \rightarrow \\ ((P(x', 3, 2, 2) \rightarrow P(x', 2, 3, 3)) \rightarrow P(x, 1, 0, 0)) \rightarrow P(x, 0, 1, 1))$$

$$(P(3, 2, 2, 2) \rightarrow (2, 3, 3, 3)) \rightarrow P(1, 0, 0, 0) \rightarrow P(0, 1, 1, 1),$$

$$P(2, 2, 2, 2) \rightarrow P(1, 1, 1, 1), P(3, 3, 3, 3), D(2, 0), D(3, 1).$$

**Goal:**  $P(0, 0, 0, 1)$

**New assumptions:**  $P(2, 2, 2, 3) \rightarrow P(2, 2, 2, 2),$

## Another example:

$$\forall xyzx'y'z' (D(x, x') \rightarrow D(y, y') \rightarrow D(z, z') \rightarrow \\ ((P(x', y', z', 3) \rightarrow P(x', y', z', 2)) \rightarrow P(x, y, z, 1)) \rightarrow P(x, y, z, 0))$$

$$\forall xyx'y' (D(x, x') \rightarrow D(y, y') \rightarrow \\ ((P(x', y', 3, 2) \rightarrow P(x', y', 2, 3)) \rightarrow P(x, y, 1, 0)) \rightarrow P(x, y, 0, 1))$$

$$\forall xx' (D(x, x') \rightarrow \\ ((P(x', 3, 2, 2) \rightarrow P(x', 2, 3, 3)) \rightarrow P(x, 1, 0, 0)) \rightarrow P(x, 0, 1, 1))$$

$$(P(3, 2, 2, 2) \rightarrow (2, 3, 3, 3)) \rightarrow P(1, 0, 0, 0) \rightarrow P(0, 1, 1, 1),$$

$$P(2, 2, 2, 2) \rightarrow P(1, 1, 1, 1), P(3, 3, 3, 3), D(2, 0), D(3, 1).$$

**Goal:**  $P(0, 0, 1, 0)$

**New assumptions:**  $P(2, 2, 2, 3) \rightarrow P(2, 2, 2, 2),$

$P(2, 2, 3, 2) \rightarrow P(2, 2, 2, 3),$

## Another example:

$$\forall xyzx'y'z' (D(x, x') \rightarrow D(y, y') \rightarrow D(z, z') \rightarrow \\ ((P(x', y', z', 3) \rightarrow P(x', y', z', 2)) \rightarrow P(x, y, z, 1)) \rightarrow P(x, y, z, 0))$$

$$\forall xyx'y' (D(x, x') \rightarrow D(y, y') \rightarrow \\ ((P(x', y', 3, 2) \rightarrow P(x', y', 2, 3)) \rightarrow P(x, y, 1, 0)) \rightarrow P(x, y, 0, 1))$$

$$\forall xx' (D(x, x') \rightarrow \\ ((P(x', 3, 2, 2) \rightarrow P(x', 2, 3, 3)) \rightarrow P(x, 1, 0, 0)) \rightarrow P(x, 0, 1, 1))$$

$$(P(3, 2, 2, 2) \rightarrow (2, 3, 3, 3)) \rightarrow P(1, 0, 0, 0) \rightarrow P(0, 1, 1, 1),$$

$$P(2, 2, 2, 2) \rightarrow P(1, 1, 1, 1), P(3, 3, 3, 3), D(2, 0), D(3, 1).$$

**Goal:**  $P(0, 0, 1, 1)$

**New assumptions:**  $P(2, 2, 2, 3) \rightarrow P(2, 2, 2, 2),$   
 $P(2, 2, 3, 2) \rightarrow P(2, 2, 2, 3), P(2, 3, 2, 2) \rightarrow P(3, 2, 2, 2),$

## Another example:

$$\forall xyzx'y'z' (D(x, x') \rightarrow D(y, y') \rightarrow D(z, z') \rightarrow \\ ((P(x', y', z', 3) \rightarrow P(x', y', z', 2)) \rightarrow P(x, y, z, 1)) \rightarrow P(x, y, z, 0))$$

$$\forall xyx'y' (D(x, x') \rightarrow D(y, y') \rightarrow \\ ((P(x', y', 3, 2) \rightarrow P(x', y', 2, 3)) \rightarrow P(x, y, 1, 0)) \rightarrow P(x, y, 0, 1))$$

$$\forall xx' (D(x, x') \rightarrow \\ ((P(x', 3, 2, 2) \rightarrow P(x', 2, 3, 3)) \rightarrow P(x, 1, 0, 0)) \rightarrow P(x, 0, 1, 1))$$

$$(P(3, 2, 2, 2) \rightarrow (2, 3, 3, 3)) \rightarrow P(1, 0, 0, 0) \rightarrow P(0, 1, 1, 1),$$

$$P(2, 2, 2, 2) \rightarrow P(1, 1, 1, 1), P(3, 3, 3, 3), D(2, 0), D(3, 1).$$

**Goal:**  $P(0, 1, 0, 0)$

**New assumptions:**  $P(2, 2, 2, 3) \rightarrow P(2, 2, 2, 2),$   
 $P(2, 2, 3, 2) \rightarrow P(2, 2, 2, 3), P(2, 3, 2, 2) \rightarrow P(3, 2, 2, 2),$   
 $P(3, 2, 2, 2) \rightarrow P(2, 3, 3, 3),$



## Another example:

$$\forall xyzx'y'z' (D(x, x') \rightarrow D(y, y') \rightarrow D(z, z') \rightarrow \\ ((P(x', y', z', 3) \rightarrow P(x', y', z', 2)) \rightarrow P(x, y, z, 1)) \rightarrow P(x, y, z, 0))$$

$$\forall xyx'y' (D(x, x') \rightarrow D(y, y') \rightarrow \\ ((P(x', y', 3, 2) \rightarrow P(x', y', 2, 3)) \rightarrow P(x, y, 1, 0)) \rightarrow P(x, y, 0, 1))$$

$$\forall xx' (D(x, x') \rightarrow \\ ((P(x', 3, 2, 2) \rightarrow P(x', 2, 3, 3)) \rightarrow P(x, 1, 0, 0)) \rightarrow P(x, 0, 1, 1))$$

$$(P(3, 2, 2, 2) \rightarrow (2, 3, 3, 3)) \rightarrow P(1, 0, 0, 0) \rightarrow P(0, 1, 1, 1),$$

$$P(2, 2, 2, 2) \rightarrow P(1, 1, 1, 1), P(3, 3, 3, 3), D(2, 0), D(3, 1).$$

**Goal:**  $P(0, 1, 0, 1)$

**New assumptions:**  $P(2, 2, 2, 3) \rightarrow P(2, 2, 2, 2),$   
 $P(2, 2, 3, 2) \rightarrow P(2, 2, 2, 3), P(2, 3, 2, 2) \rightarrow P(3, 2, 2, 2),$   
 $P(3, 2, 2, 2) \rightarrow P(2, 3, 3, 3), P(3, 2, 2, 3) \rightarrow P(3, 2, 2, 2),$

## Another example:

$$\forall xyzx'y'z' (D(x, x') \rightarrow D(y, y') \rightarrow D(z, z') \rightarrow \\ ((P(x', y', z', 3) \rightarrow P(x', y', z', 2)) \rightarrow P(x, y, z, 1)) \rightarrow P(x, y, z, 0))$$

$$\forall xyx'y' (D(x, x') \rightarrow D(y, y') \rightarrow \\ ((P(x', y', 3, 2) \rightarrow P(x', y', 2, 3)) \rightarrow P(x, y, 1, 0)) \rightarrow P(x, y, 0, 1))$$

$$\forall xx' (D(x, x') \rightarrow \\ ((P(x', 3, 2, 2) \rightarrow P(x', 2, 3, 3)) \rightarrow P(x, 1, 0, 0)) \rightarrow P(x, 0, 1, 1))$$

$$(P(3, 2, 2, 2) \rightarrow (2, 3, 3, 3)) \rightarrow P(1, 0, 0, 0) \rightarrow P(0, 1, 1, 1),$$

$$P(2, 2, 2, 2) \rightarrow P(1, 1, 1, 1), P(3, 3, 3, 3), D(2, 0), D(3, 1).$$

**Goal:** a few more steps...

**New assumptions:**  $P(2, 2, 2, 3) \rightarrow P(2, 2, 2, 2),$

$P(2, 2, 3, 2) \rightarrow P(2, 2, 2, 3), P(2, 3, 2, 2) \rightarrow P(3, 2, 2, 2),$

$P(3, 2, 2, 2) \rightarrow P(2, 3, 3, 3), P(3, 2, 2, 3) \rightarrow P(3, 2, 2, 2),$

.....,

## Another example:

$$\forall xyzx'y'z' (D(x, x') \rightarrow D(y, y') \rightarrow D(z, z') \rightarrow \\ ((P(x', y', z', 3) \rightarrow P(x', y', z', 2)) \rightarrow P(x, y, z, 1)) \rightarrow P(x, y, z, 0))$$

$$\forall xyx'y' (D(x, x') \rightarrow D(y, y') \rightarrow \\ ((P(x', y', 3, 2) \rightarrow P(x', y', 2, 3)) \rightarrow P(x, y, 1, 0)) \rightarrow P(x, y, 0, 1))$$

$$\forall xx' (D(x, x') \rightarrow \\ ((P(x', 3, 2, 2) \rightarrow P(x', 2, 3, 3)) \rightarrow P(x, 1, 0, 0)) \rightarrow P(x, 0, 1, 1))$$

$$(P(3, 2, 2, 2) \rightarrow (2, 3, 3, 3)) \rightarrow P(1, 0, 0, 0) \rightarrow P(0, 1, 1, 1),$$

$$P(2, 2, 2, 2) \rightarrow P(1, 1, 1, 1), P(3, 3, 3, 3), D(2, 0), D(3, 1).$$

**Goal:**  $P(1, 1, 1, 1)$

**New assumptions:**  $P(2, 2, 2, 3) \rightarrow P(2, 2, 2, 2),$   
 $P(2, 2, 3, 2) \rightarrow P(2, 2, 2, 3), P(2, 3, 2, 2) \rightarrow P(3, 2, 2, 2),$   
 $P(3, 2, 2, 2) \rightarrow P(2, 3, 3, 3), P(3, 2, 2, 3) \rightarrow P(3, 2, 2, 2),$   
.....,  $P(3, 3, 3, 3) \rightarrow P(3, 3, 3, 2)$

# $\Sigma_1$ and **Expspace**

## Theorem

*The problem to decide if a given  $\Sigma_1$  formula is derivable in intuitionistic predicate calculus is **Expspace**-complete.*

## Unary and multi-ary predicates (a diggression)

Can we translate this example to a monadic signature?

Not really: it is essential that the goals to prove represent strings being rewritten.

What counts are the targets of non-atomic formulas.

More precisely: if  $\Gamma \vdash \sigma$  is a decision problem where

– all targets of non-atomic formulas in  $\Gamma$  are unary or nullary,  
then one can find an equivalent monadic problem  $\Gamma' \vdash \sigma'$ .

# Alternation

## Existential choice

because there may be more than one usable assumption.

## Universal choice

because an assumption may have more than one premise.

(To derive  $P$  from  $\varphi \rightarrow \psi \rightarrow P$

one has to prove both  $\varphi$  and  $\psi$ .)

# The monadic case

**Theorem:** *The decision problem for  $\Sigma_1$  with at most unary predicates is co-Nexptime-complete.*

**Proof:**

1. A  $\Sigma_1$  formula can express the halting problem for a *Branching Turing Machine* (alternating Turing Machine without existential states) working in exponential time.
2. A non-provable  $\Sigma_1$  formula has an exponential size refutation.

# Branching Turing Machine (BTM)

A branching Turing Machine only has deterministic and universal states. Its computation is a tree which may consist of many branches, but there is only one computation for any given input.

This means that a BTM is like a deterministic tree automaton.

Yet differently: the problem of finding an accepting computation of a BTM has at most one solution.



# Branching Turing Machine (BTM)

A computation of BTM is a tree of machine configurations, each configuration consisting of a number of tape cells.

It can be seen as a finite graph of tape cells.

For a given input this graph is unique.

Such a graph can be reconstructed using a nondeterministic loop (after constructing an initial node):

1. Add a new node  $n$  to the graph;
2. Nondeterministically set the label of  $n$ ;
3. Verify the correctness of the label;
4. If  $n$  is the last node then stop else go to 1.

# Simulating a BTM

Reconstruction of a computation of BTM:

1. Add a new node  $n$  to the graph;
2. Nondeterministically set the label of  $n$ ;
3. Verify the correctness of the guess;
4. If  $n$  is the last node then stop else go to 1.

This can be simulated by a proof-construction process for a  $\Sigma_1$  decision problem, where all non-atomic formulas have the same nullary target *loop*.

With no multi-ary targets, the decision problem can be designed as a unary decision problem.

## Monadic lower bound

The decision problem for monadic  $\Sigma_1$  is hard for the class of languages recognized by exponential time bounded branching Turing machines. In other words, it is co-Nexptime-hard.

NB: the positive assumption formulas use in the coding are actually universal formulas (all quantifiers occur at the beginning).

## Towards monadic upper bound

The process of deciding a judgment  $\Gamma \vdash \phi$  can be seen as a game between an existential player E, trying to prove the judgment, and a universal player A, trying to refute it.

Player E attempts to construct a long normal proof. At every step he chooses an assumption  $\psi \in \Gamma$  to be used in the next step. Player A then determines the next goal, choosing one of the premises of  $\psi$ . Player E wins when A has no choice.

## Example

For example, suppose in position  $\Gamma \vdash P(x)$  player E chooses the formula  $\forall y((R(z) \rightarrow Q(y)) \rightarrow R(y) \rightarrow P(y))$  from  $\Gamma$ .

Then player A chooses the next position.

It is either  $\Gamma \vdash R(x)$ , or  $\Gamma, R(z) \vdash Q(x)$ .

# Refutations

The game is determined. One of the players must have a strategy: there is either a proof or a refutation.

What is a refutation?

It is an infinite tree of judgments representing a strategy of player A. Every node has as many children as there are possible choices of E. Each child represents a move of A in response to some choice of E.

# The size of a refutation

- ▶ If a judgment  $\Gamma \vdash \sigma$  occurs in a refutation of an initial formula of size  $n$  then the size of  $\Gamma$  is exponential in  $n$ .
- ▶ There is at most  $n^2$  possible choices of  $\sigma$ .
- ▶ Every node has at most exponentially many children.
- ▶ A refutation is infinite but can be trimmed to a finite tree, of at most exponential depth...
- ▶ ...with only polynomial number of branchings.

# Conclusions

- ▶ The decision problem for  $\Sigma_1$  is Expspace-complete.
- ▶ Important: only free variables in the original formula are used.
- ▶ Important: the number of targets is exponential.
- ▶ The decision problem for  $\Sigma_1$  with at most unary predicates is co-Nexptime-complete.
- ▶ Important: small number of targets.