

Soundness is not Sufficient*

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Disclaimer

- This is an idea (= rambling) talk.
- The ideas and the intentions behind them are important, I believe.
- The technical definitions may not be the definite ones (yet).
- I have likely overlooked some results of yours -- tell me.
- Ask, comment, interrupt any time.
- I'll skip slides (which ones I don't know yet)

Goals

- Propose informal criteria for what a static analysis should satisfy to warrant being called a “good” static analysis.
- Propose technical criteria for capturing some aspects of the informal criteria
- Identify questions for further work, both conceptual and technical.


Program property

- A **program property** is a predicate on programs.

- A program property P is **semantic (extensional)** if

$$p \cong q \Rightarrow (P(p) \Leftrightarrow P(q))$$

- A program property P is **trivial** if $P(p)$ for all p , or $\neg P(p)$ for all p .



Behavioral
equivalence

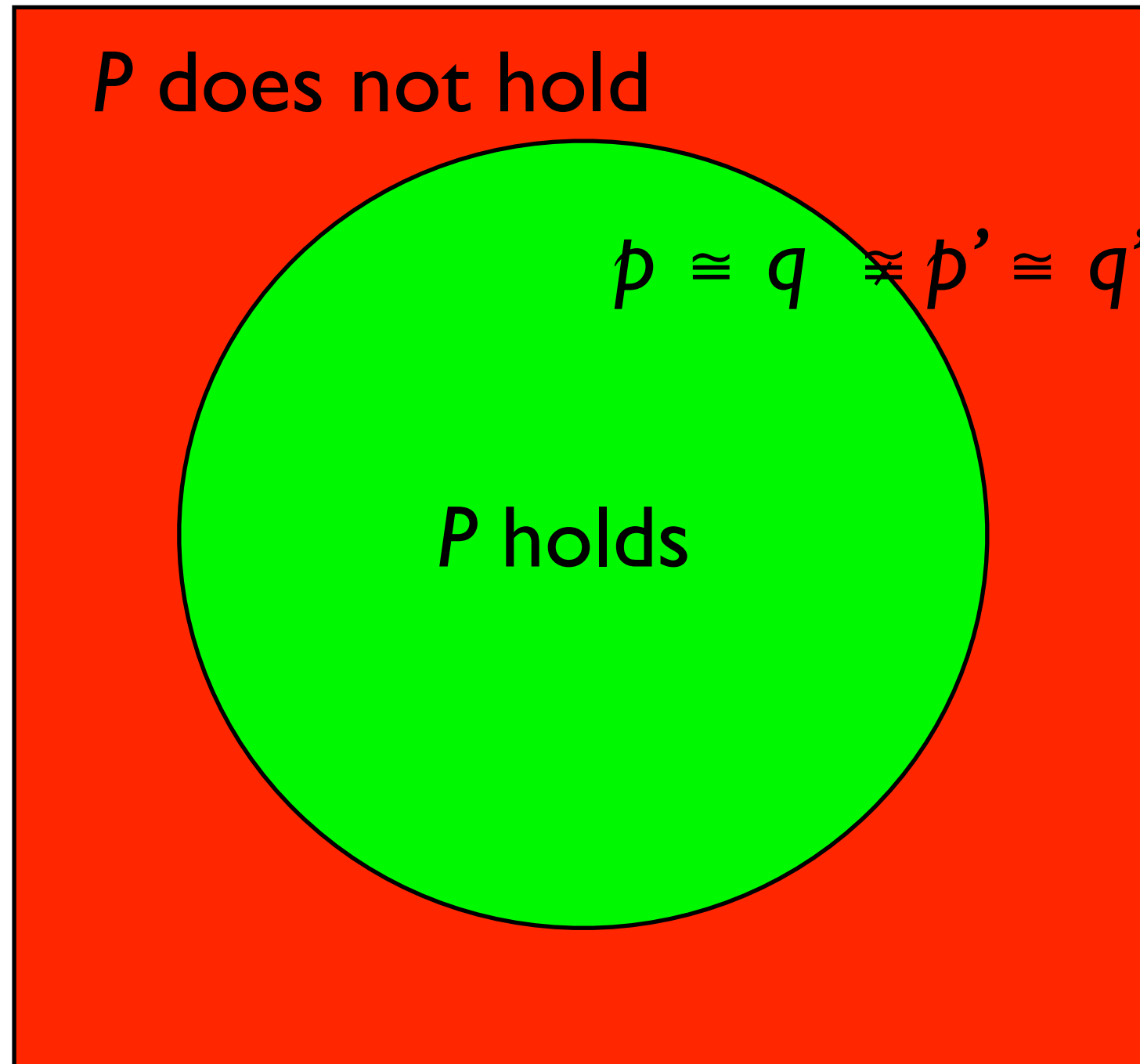
Rice's Curse

Theorem:

Let L be a Turing-complete programming language, P a nontrivial semantic program property.
Then P is undecidable.

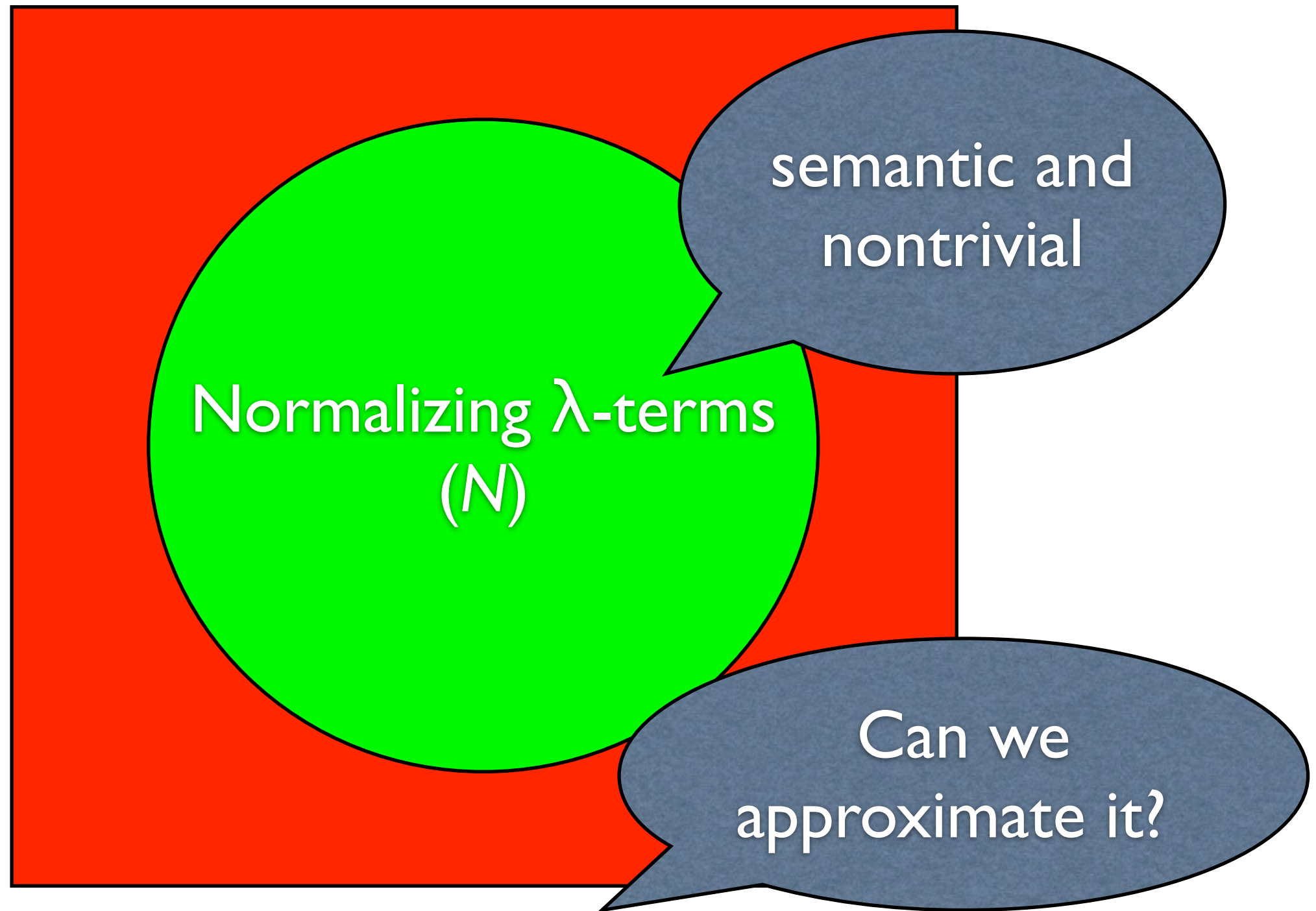
Rice, Classes of recursively enumerable sets and their decision problems, Trans. AMS 1953

Rice's Curse, pictorially



P is not decidable!

Rice's Curse: Example



Corollary: N is not decidable!

Static analysis

- Given:
 - P : Extensional program property
 - (S, S') : Static analysis for P
- We want of (S, S') :
 - **Soundness:** $S \subseteq P, S' \subseteq \neg P$

Is that sufficient? No, we also want...

- **Goodness**

What does “good” mean??


Goodness characteristics

- **Usefulness:**
 - Has some effective use, fitness for a purpose
- **Declarative specification:**
 - Separation of **what** the analysis computes from **how** it computes it (the particular algorithm[s] used)

Goodness characteristics

- **Unimprovability:**
 - Can't get better approximation at lower computational cost
- **Predictability:**
 - Predictability under certain, specified program transformations and changes

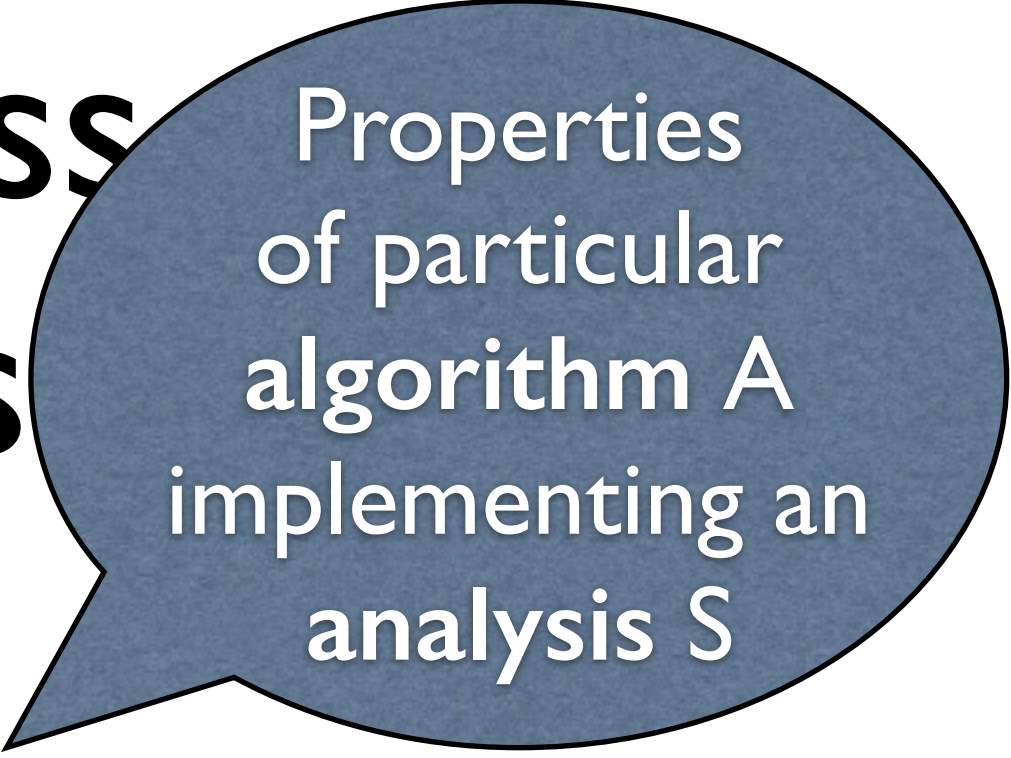
Goodness character



Algorithm need not be
compositional, only its
result

- **Compositional certification**
 - Explicit, modular (syntax-oriented), efficiently checkable logical explanation of analysis results
- **Constructive interpretation**
 - Operational interpretation of certificate, not just of yes/no answer

Goodness characteris



Properties
of particular
algorithm A
implementing an
analysis S

- **Adaptiveness:**
 - Easy instances are handled efficiently
 - Hard instances may take more time.
- **Parameter sensitivity**
 - Scale well with parameter, which captures expectations on input distribution.

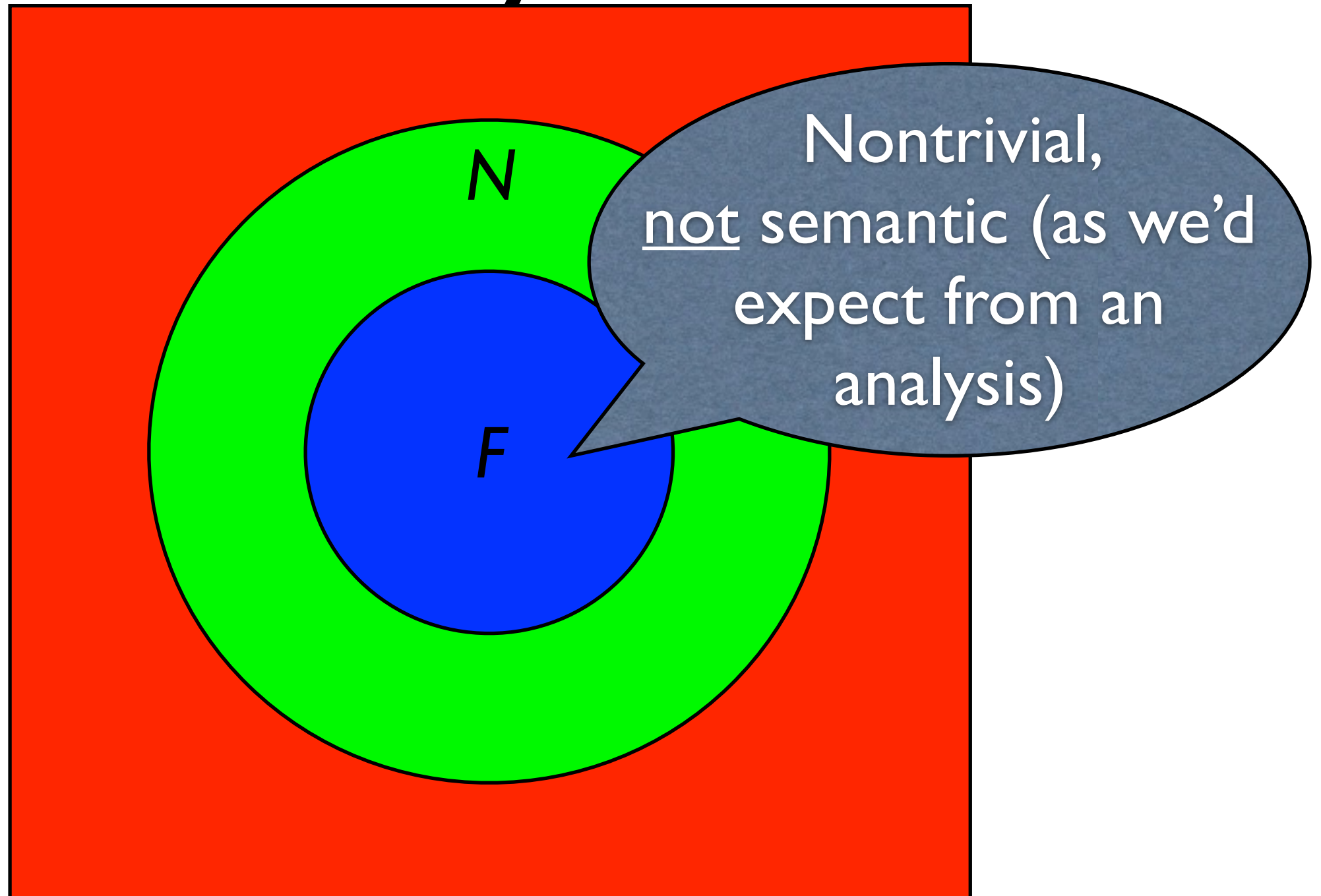
Static Analysis for N

- Consider normalizability of λ -terms.
- Is *System F* typability a “good” static analysis for N ?

System F for N

- Sound? ✓
- Declarative? ✓
- Compositionally certified? ✓
- Useful? ✓
- Predictability properties? (✓)
- Unimprovability? Hmm...

Static Analysis for N



Theorem: F is undecidable

System F for N : Improvability

- Acceptable for *System F* (as a static analysis for N) to be undecidable, as long as there is no **better** approximation of N that is decidable (more efficient).

Unimprovability via separability

Property of analysis, not any particular algorithm

- A static analysis (S, S') for P is **improvable** if there exists (T, T') such that:
 - $S \subseteq T \subseteq P, S' \subseteq T' \subseteq \neg P$, and
 - T and T' have “lower” (structural) complexity than S and S' ; e.g. S is NP-hard, but T is in P (with $S' = T' = \emptyset$).

Recursive inseparability

Classical definition in recursion theory: "... from complement of P if there is no B..."

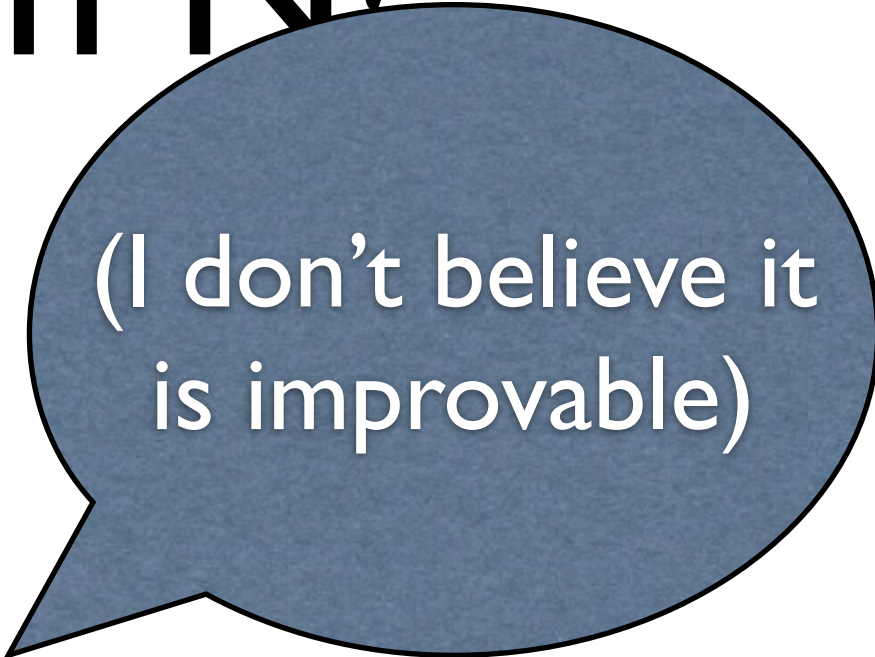
Definition:

Let $A \subseteq P$. A is **recursively inseparable** from P if there is no B such that $A \subseteq B \subseteq P$ and B is decidable ("recursive").

Is F recursively inseparable from N ?

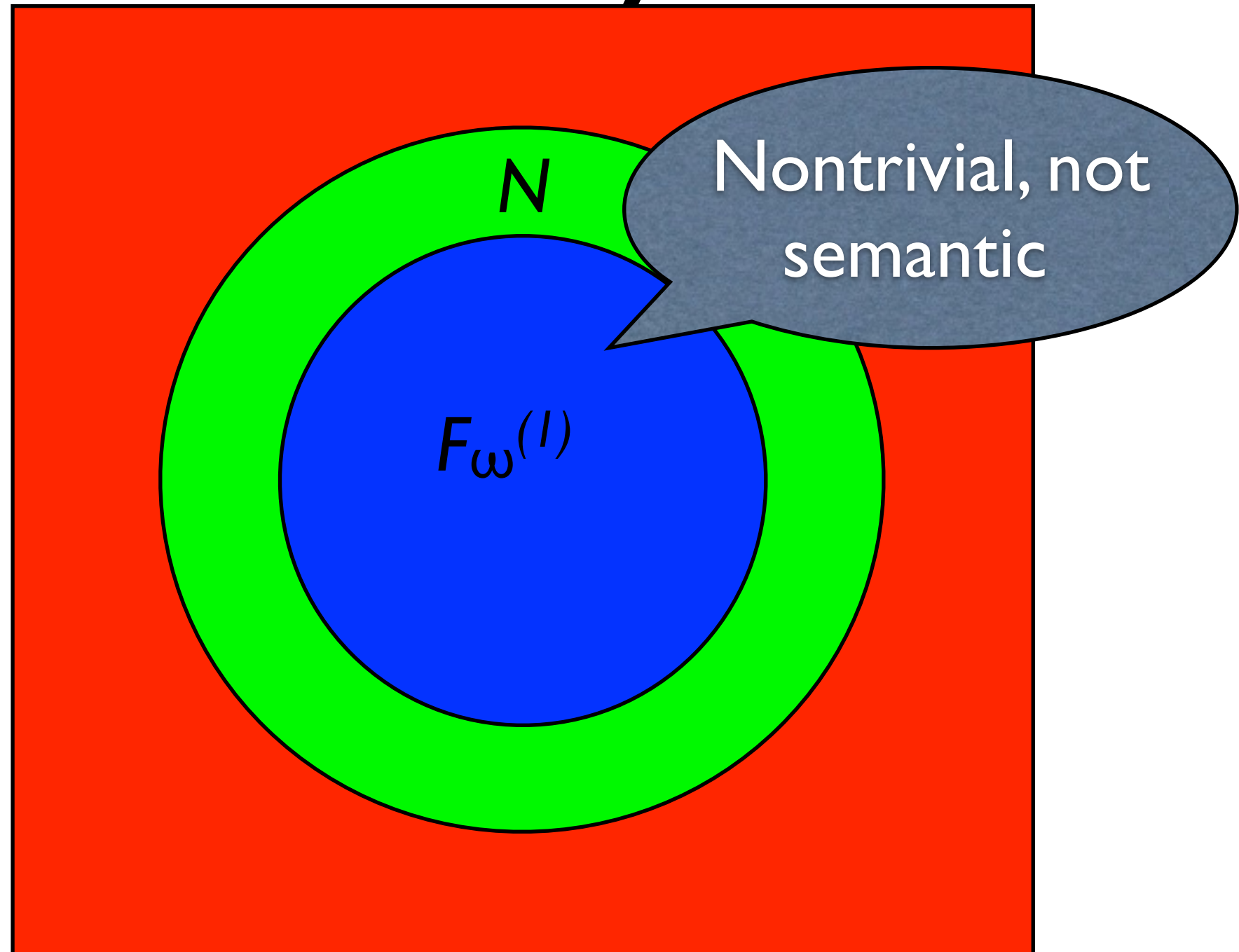
Is F recursively inseparable from N ?

- The answer is...
- **We don't know!**
 - Does not follow from Well's proof
- We don't know whether F is improvable:
 - There may be a (type) system out there that extends System F , guarantees N and is decidable.



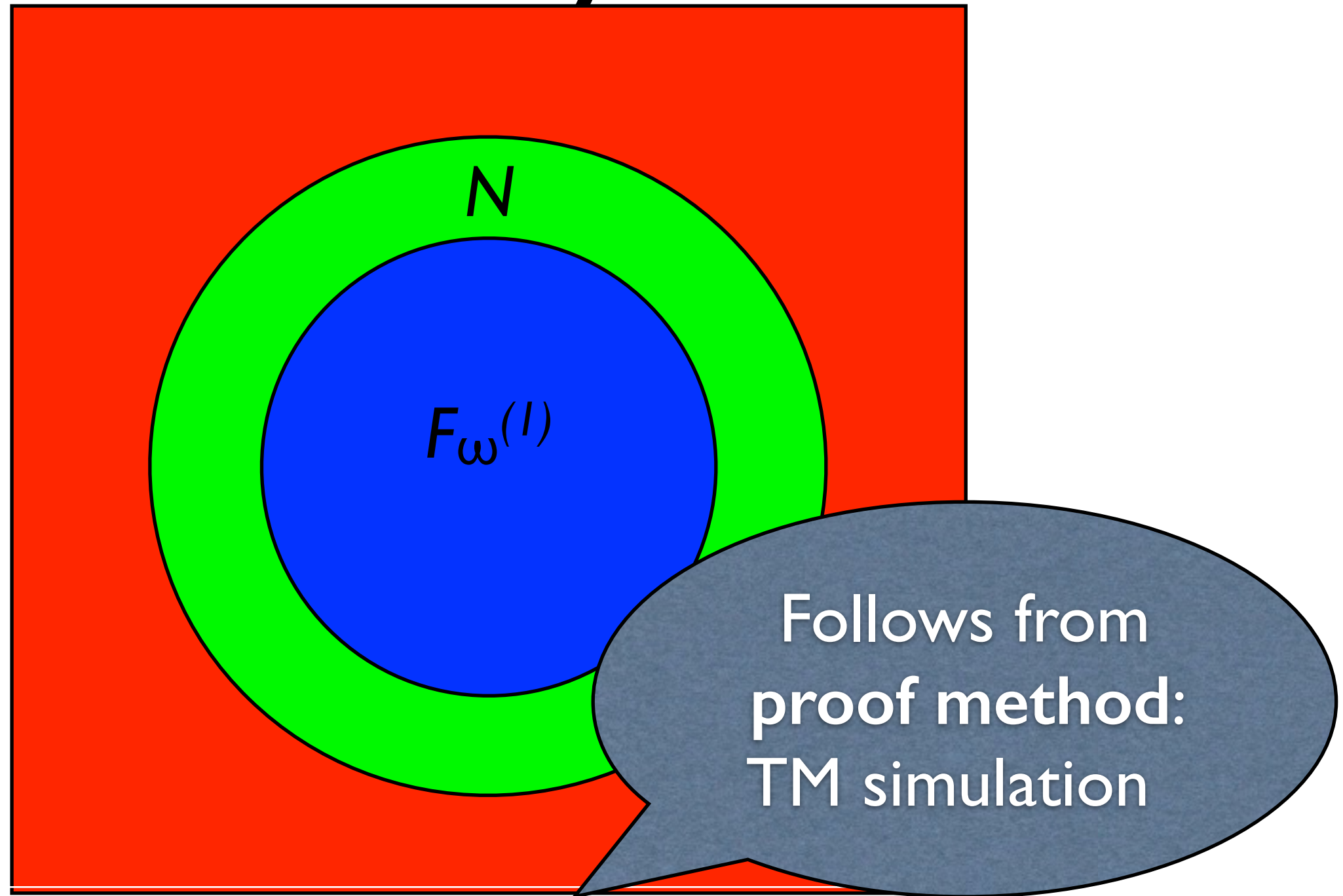
(I don't believe it is improvable)

Another analysis for N



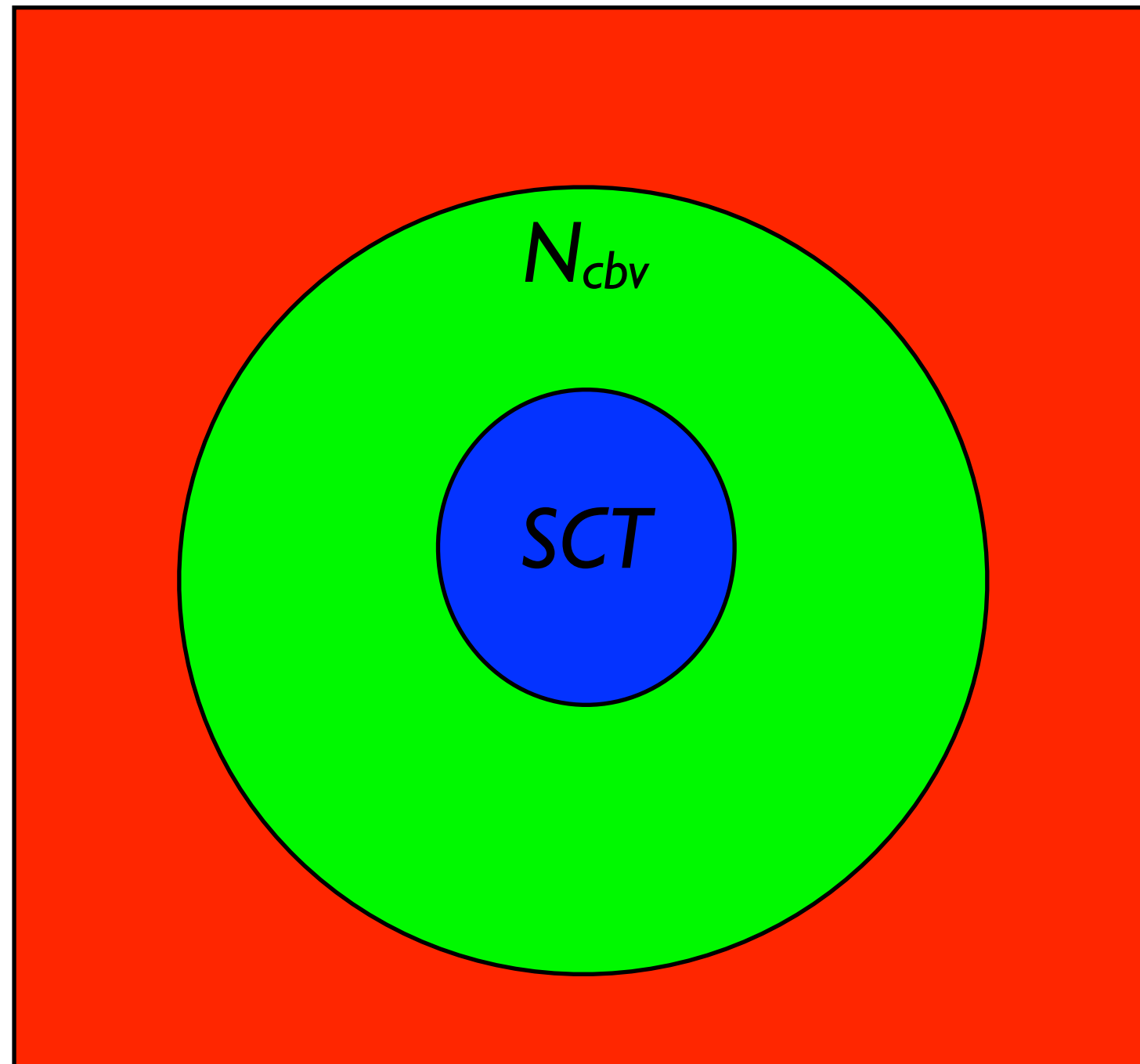
Theorem: $F_{\omega}^{(I)}$ is undecidable

Another analysis for N



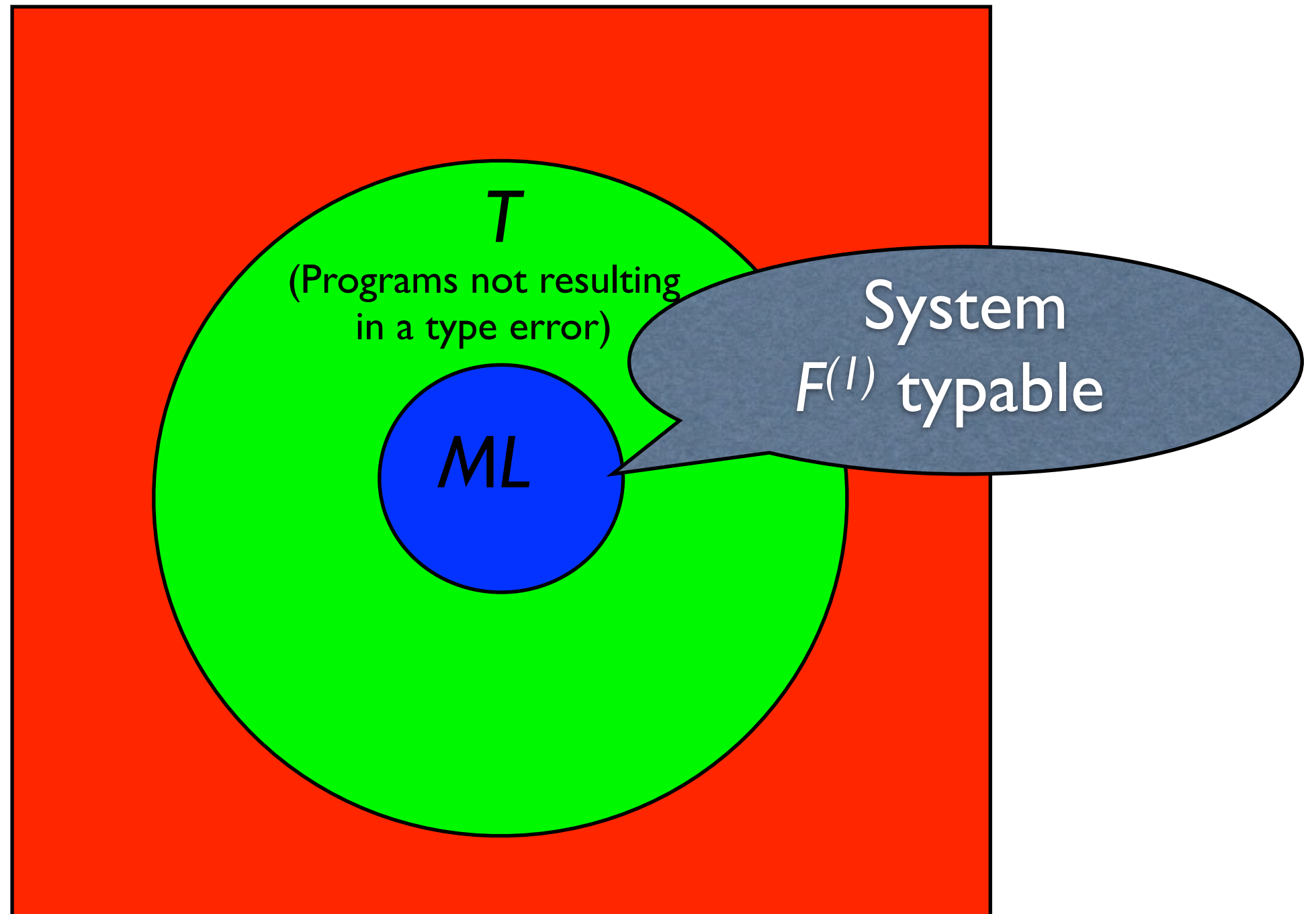
Theorem: $F_{\omega}^{(l)}$ is recursively inseparable from N

SCT for N

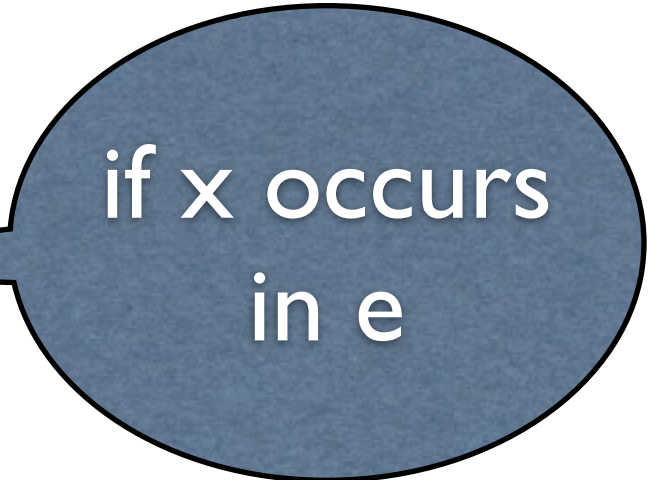


Theorem: SCT is decidable.
(Complexity: $PSCACE$ -complete)

An analysis for type error freeness



ML goodness

- **Invariant under let-reduction:**
 $ML(\text{let } x = e \text{ in } e') \Leftrightarrow ML(e'[e/x])$
- **Preservation under β -reduction:**
 $ML((\lambda x.e)e') \Rightarrow ML(e[e'/x])$ 
- **Preservation under eta-reduction:**
 $ML(\lambda x.ex) \Rightarrow ML(e)$
- ML is invariant under arbitrary unfolding (inlining) and folding (refactoring) of (nonrecursive) definitions

ML typability as static analysis for type error freeness

- Is ML typability improvable?

ML typability as static analysis for type error freedom

No, ML is not
improvable for type
error detection

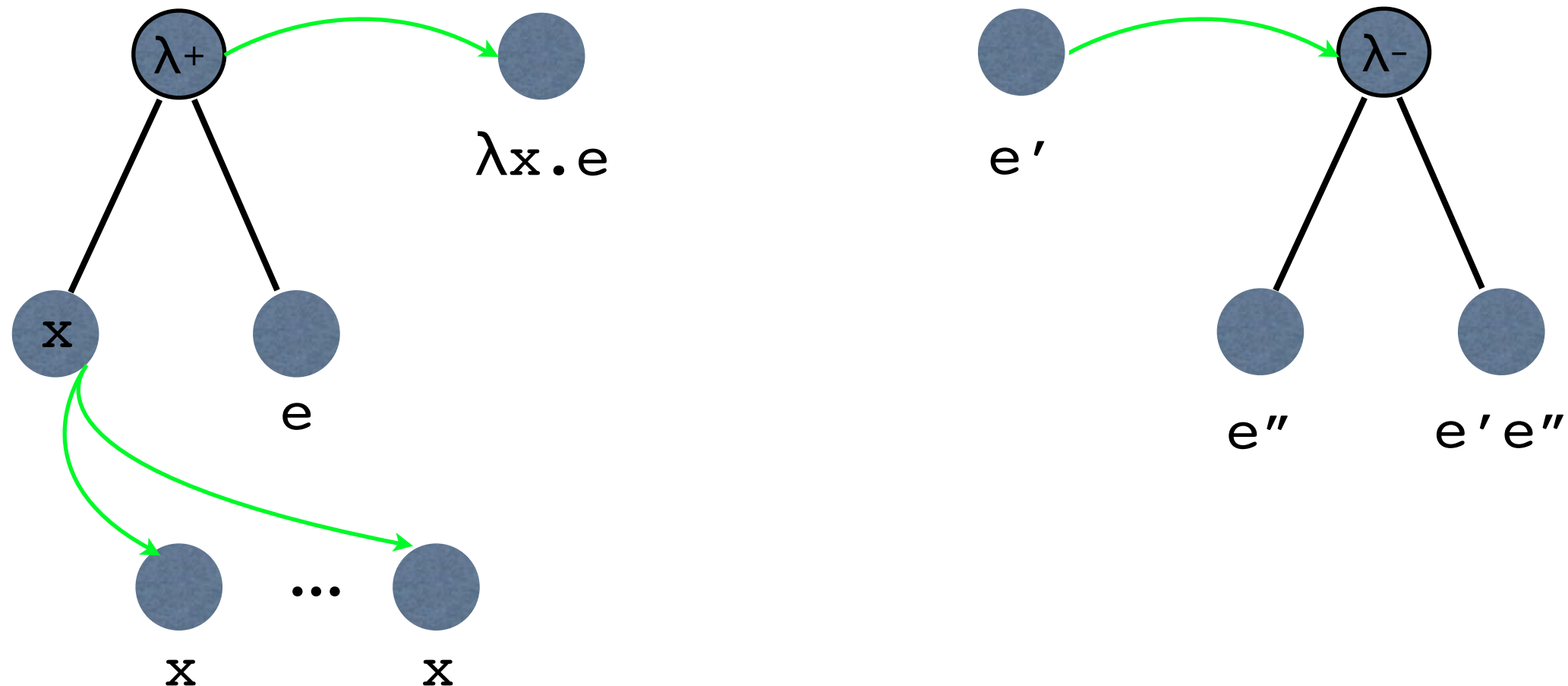
Theorem: Let $ML \subseteq B \subseteq T$.
Then B is *DEXPTIME*-hard.

mVFA

(OCFA in direct style)

Build graph with **flow** and **tree** edges. One node per subexpression, plus some extra ones (λ^-).

I. Base flow rules, resulting in graph G :

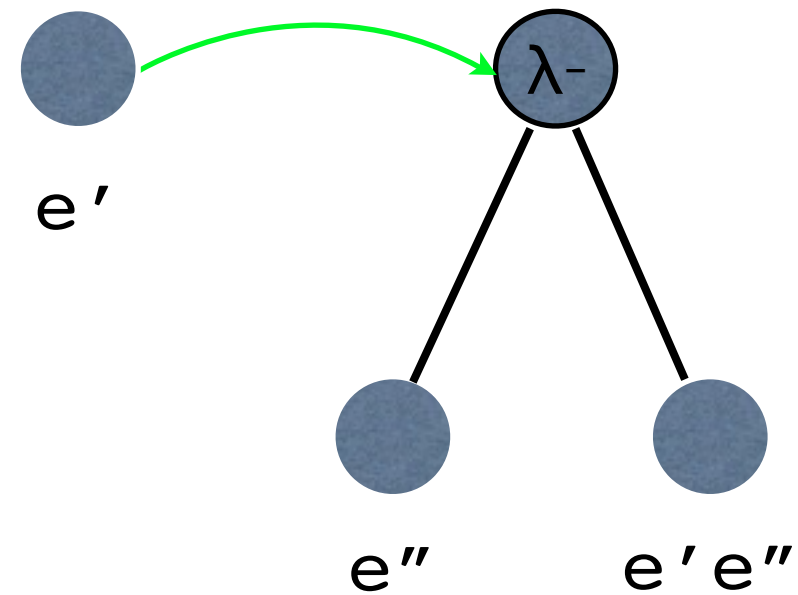
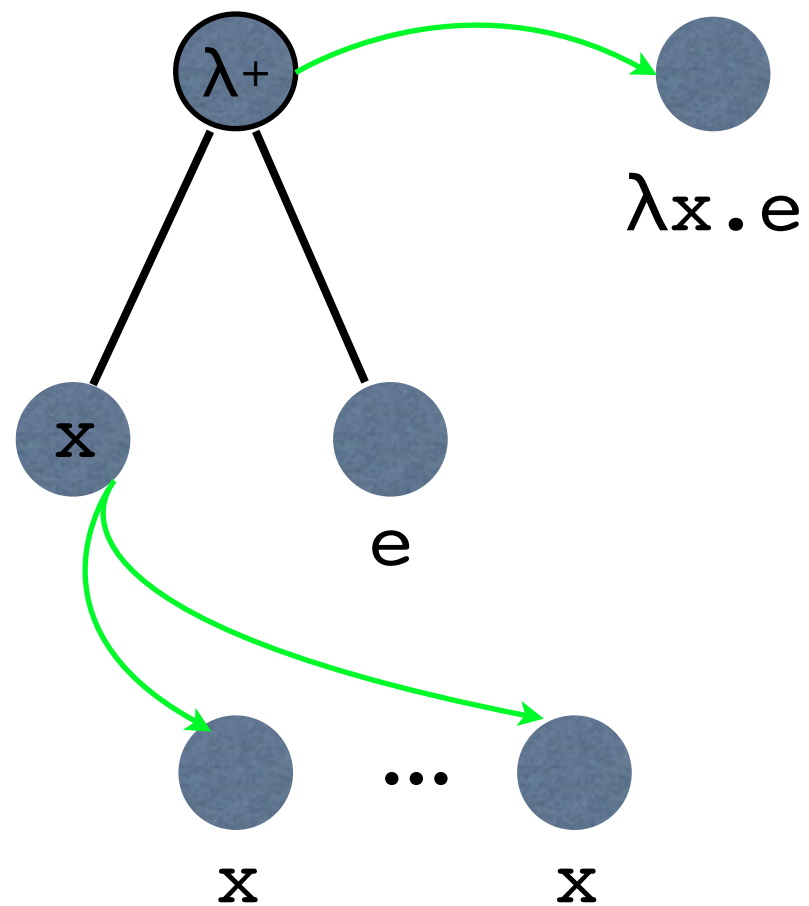


mVFA

OCFA in direct style

$O(n)$ nodes

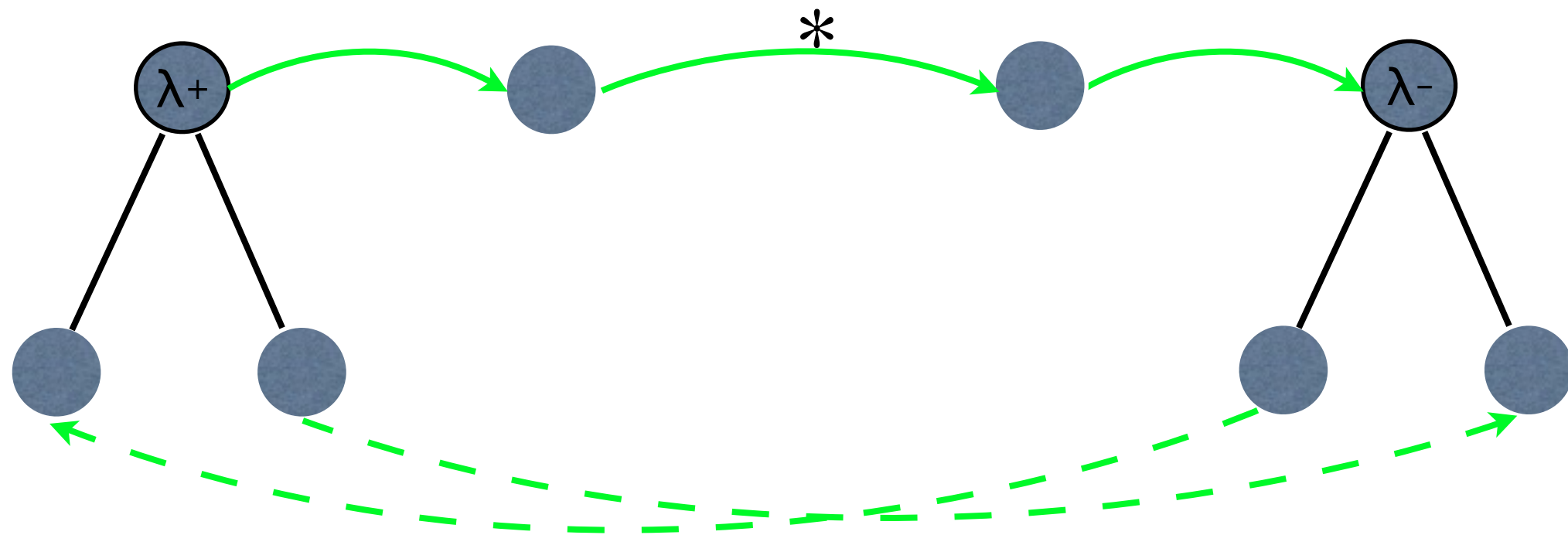
$O(n)$ edges



mVFA

OCFA in direct style

2. Closure rule:



mVFA

OCFA in direct style

Algorithm:

Close base graph under closure rule, resulting in graph G.

mVFA

OCFA in direct style

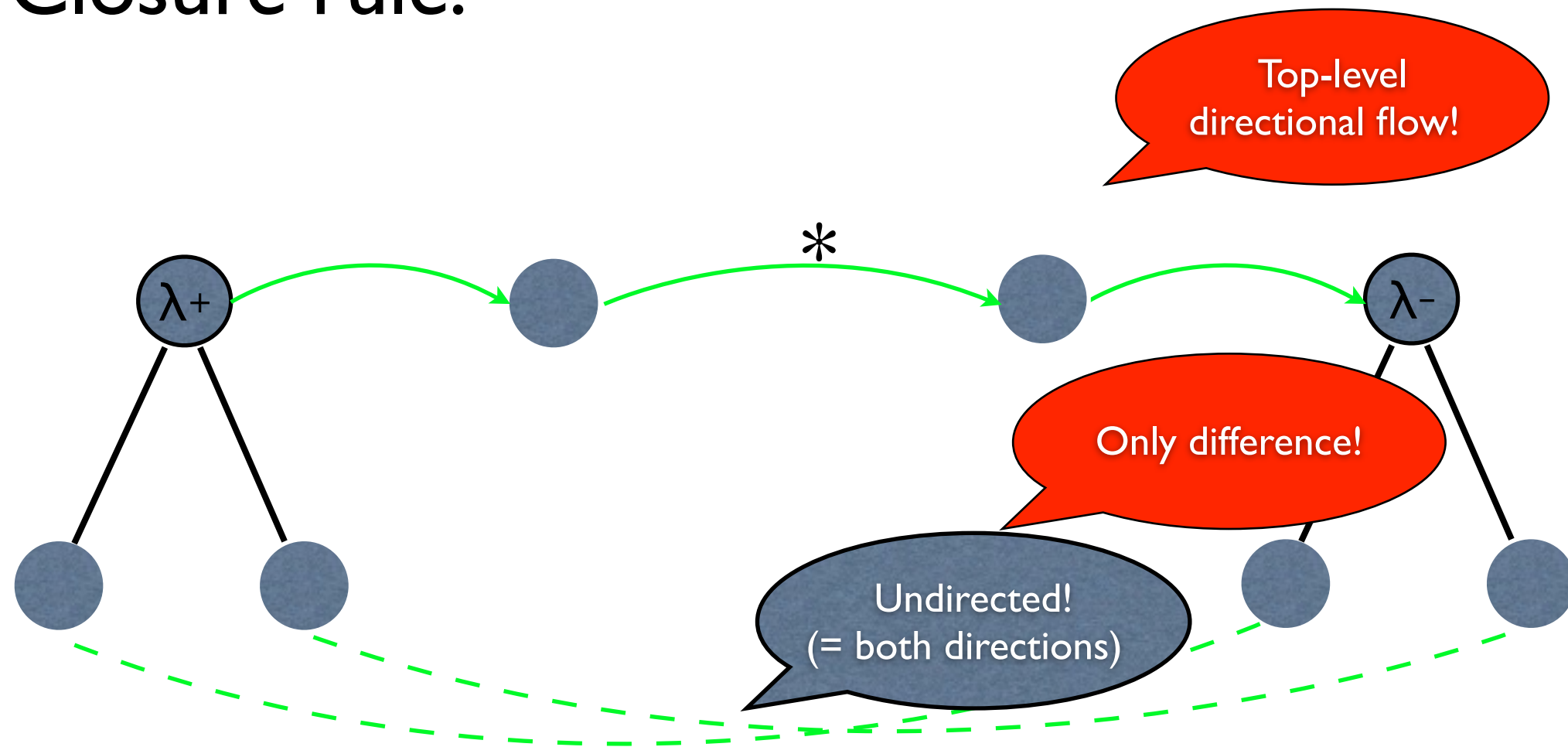
Theorem: mVFA can be implemented in time $O(d(m^* + pn + q))$, where

- n : number of nodes
- d : maximum outdegree of nodes in G ,
- m^* : number of flow edges in G^*
(flow-transitive closure of G),
- p : number of closure rule applications.
- q : number of reachability queries

sVFA

Simple closure analysis

1. Base rules: As for mVFA
2. Closure rule:



sVFA

Simple closure analysis

Algorithm:

Close base graph under closure rule by unification closure, using union/find data structure.

sVFA

Simple closure analysis

Theorem: sVFA can be implemented in time

$O(n \alpha(n,n) + q n)$, where

- $\alpha(m,n)$: inverse Ackerman function
- q : number of reach set queries

Henglein, Simple Closure Analysis, TOPPS TR D-193, 1992

sVFA

Simple closure analysis

- Very fast in practice
- Applications:
 - Binding-time analysis
Henglein, Efficient Type Inference for Higher-Order Binding-Time Analysis, FPCA 1991
 - Dynamic type inference for Scheme
Henglein, Global tagging optimization by type inference, LFP 1992
 - Closure analysis in Similix
Bondorf, Jørgensen, Efficient Analysis for Realistic Off-Line Partial Evaluation, JFP 1993
- No significant reduction in precision vis a vis mVFA observed
- Flows are *not* unidirectional (“equational”)

sVFA predictability

- sVFA is invariant under
 - linear beta-reduction
 - eta-reduction (for pure λ -terms)

sVFA predictability

Theorem:

sVFA-reachability is P -complete

Van Horn, Mairson, Flow Analysis, Linearity, and PTIME, SAS 2008

Not a corollary. Follows from proof method used: invariance under linear λ -term reduction

Theorem:

Let B be such that $sVFA \subseteq B \subseteq R$, where R is semantic (un)reachability. Then B is P -hard.

Adaptiveness

- Assume $S_0 \subseteq S_1 \subseteq P$, with algorithms A_0, A_1 for S_0, S_1 , respectively. ($S_0' = S_1' = \emptyset$)

Not a proper definition formal

- A_1 is naturally **adaptive** over A_0 if its (time) complexity is as good as the complexity of A_0 on instances from S_0 , without explicitly invoking A_0
- A_1 is allowed to take substantially more time than A_0 on instances outside S_0 . (where A_0 and A_1 give different results).

Adaptiveness

- **Intuition:** A static analysis algorithm should **not be slower** on instances where a **less precise** analysis algorithm manages to compute the **semantically correct** result (on “easy instances”).

Questions

- Are the various k CFA-algorithms adaptive over sVFA?
- Is (functional) k CFA improvable for $k \geq 1$?
- Is SCT improvable? How predictable is it?
- ...

End of talk