

CS 6810 – Theory of Computing

Lecture 1: Introduction & Review

David Steurer

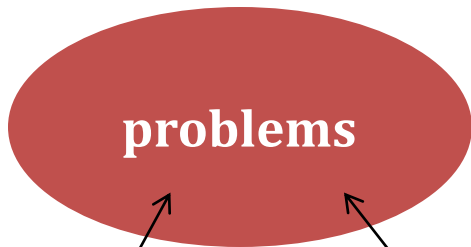
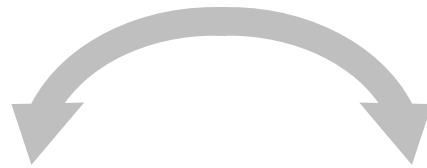
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What is complexity theory?

How do resource limitations impact our ability to solve problems?

lower bounds
(impossibility results)

upper bounds
(algorithms)



3SAT
={satisfiable 3CNF formulas}

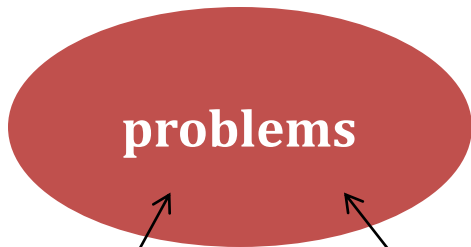
LP
={feasible linear programs}

P
={poly-time Turing machines}

L
={log-space Turing machines}

Which problems are easier than others?

reductions 



LP

= {feasible linear programs}

3SAT

= {satisfiable 3CNF formulas}



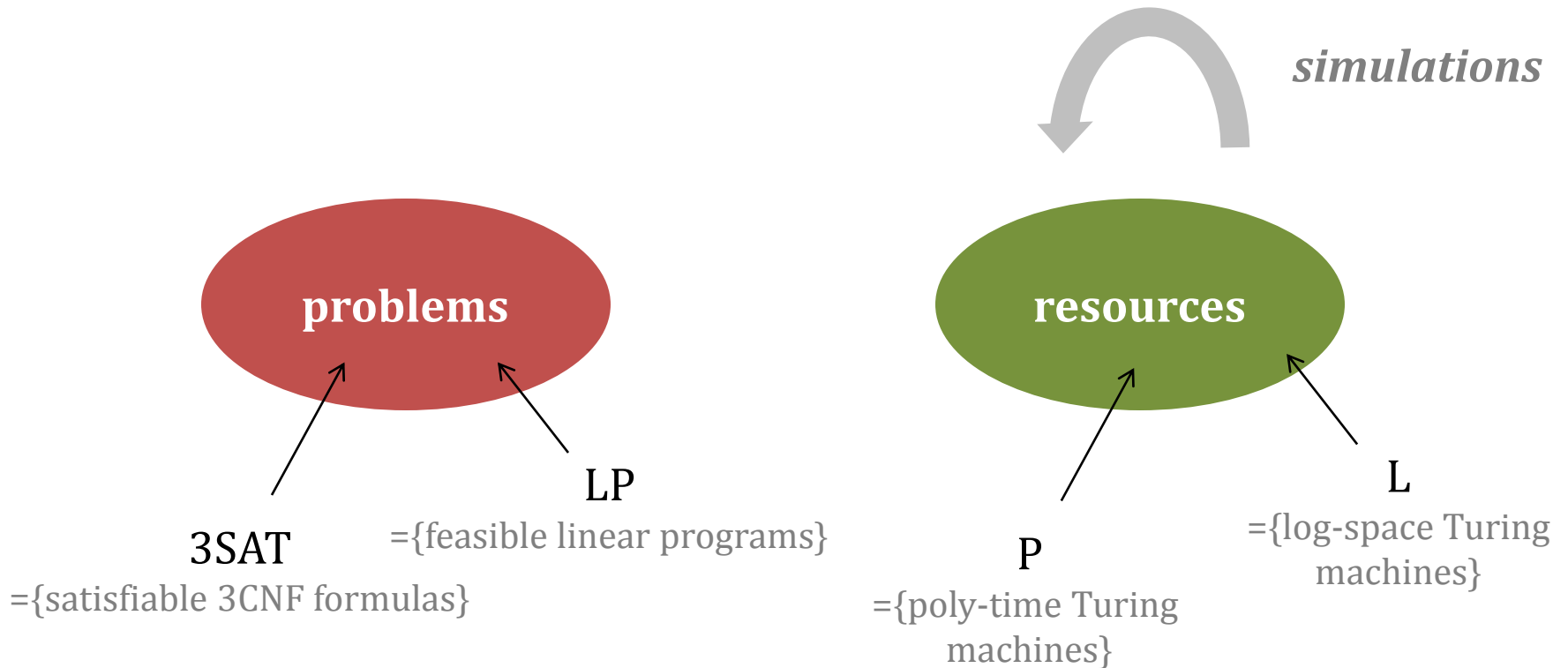
L

= {log-space Turing machines}

P

= {poly-time Turing machines}

What resources are more powerful than others?





problems

decision problems

(simplest, default)

search problems

(most general)

promise problems

(very useful case between decision and search)

distributional problems

(average-case complexity, later in course)

problems

decision problems

(simplest, default)

$$f: \{0,1\}^* \rightarrow \{0,1\}$$

or

$$L \subseteq \{0,1\}^*$$

(YES/NO answer for each input)

(set of YES inputs)

Example

$$3\text{SAT} = \{\text{satisfiable 3CNF formulas } \phi\}$$

problems

search problems *(most general)*

for every input x , there is a set of acceptable outputs y
relation $R = \{(x, y) \mid \text{output } y \text{ is acceptable on input } x\}$

Examples

$3\text{SAT}^s = \{(\phi, z) \mid \text{assignment } z \text{ satisfies 3CNF formula } \phi\}$

$\text{Max3SAT}^s = \{(\phi, z) \mid \text{assignment } z \text{ satisfies as many clauses as possible in 3CNF formula } \phi\}$

problems

promise problems (*very useful case between decision and search*)

for every input x , there is a set of acceptable outputs $y \in \{0,1\}$

partition into YES inputs, NO inputs, and DON'T-CARE inputs

Example

Max3SAT(α)

YES: satisfiable 3CNF formula

NO: at most α fraction of clauses satisfiable



resources

model

Turing machine

circuits

protocols

resolution proofs

linear program

measure

time and space

non-determinism, alternation

randomness

advice

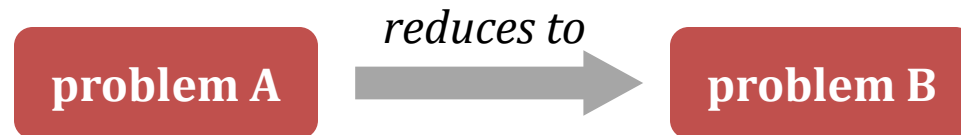
size and depth

communication cost

length

number of constraints (facets)

Reductions



Turing/Cook

solve A using an algorithm for B as a subroutine
(use B as an oracle)

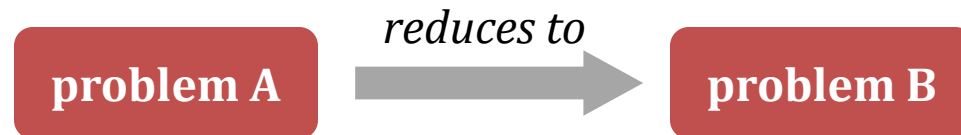
Karp

map instances of A to instances of B such that
answer is preserved (*for decision & promise problems*)

Levin

map A-instances to B-instances such that any answer
to B-instance can be pulled back (*for search problems*)

Reductions



Shows:

B is harder than A

meaningful if A is hard (*plausibly*)

Example: NP-hardness reductions

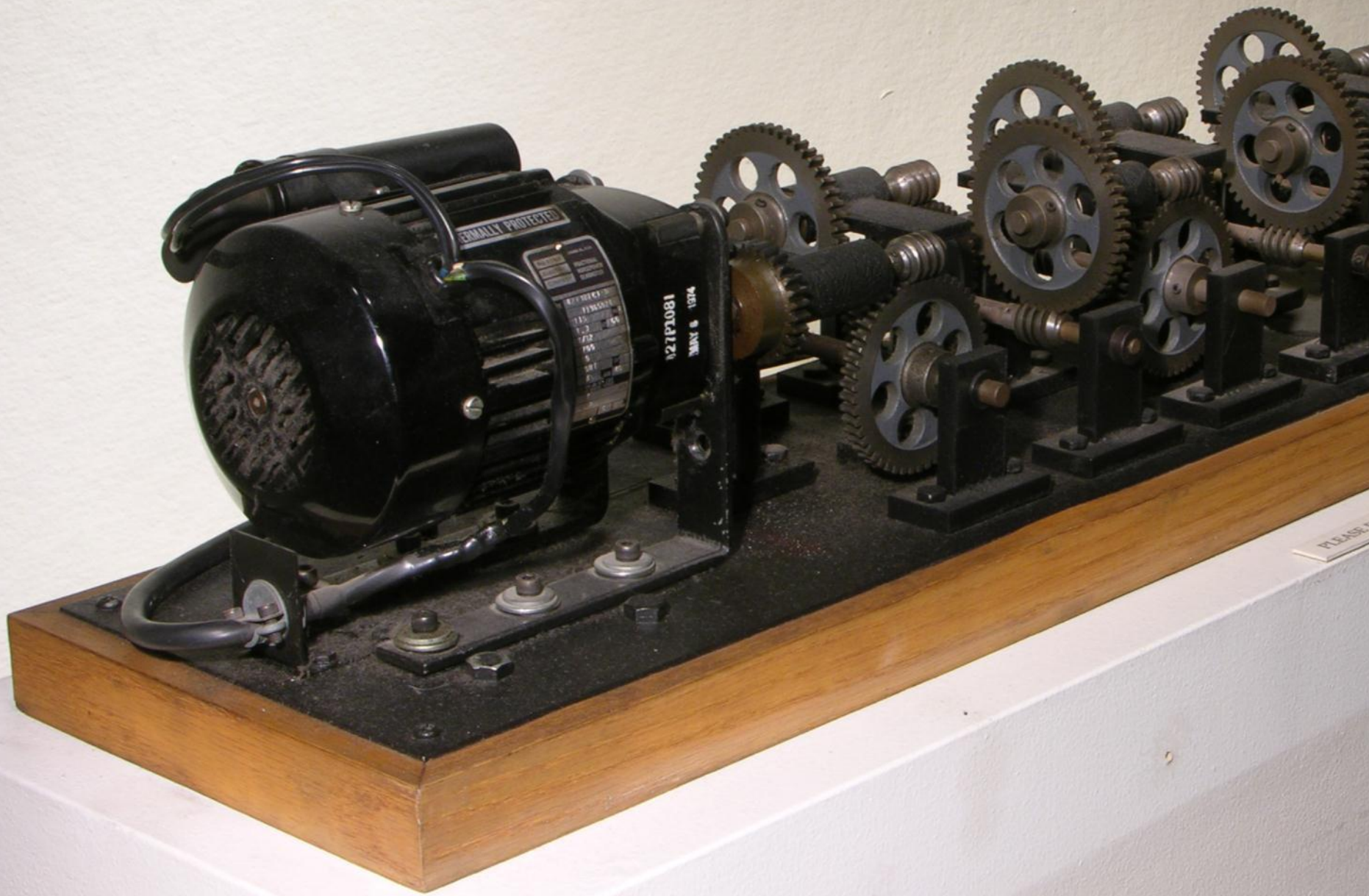
Warning:
3UNSAT is NP-hard w.r.t.
Cook reductions but *not* w.r.t.
Karp reductions (unless NP=coNP)

A is easier than B

meaningful if B is easy (*plausibly*)

Example: LP relaxations
(*reduction to LP*)

rounding algorithm is pull-back
from LP to original problem
(*constructive Levin reduction*)



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MAY 8 1974

PLEASE

Machine with Concrete

Arthur Ganson

