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Bound number of calls to `arc` (iterations of search)

# Feasibility and non-determinism: P vs NP

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$P = NP$  says non-determinism makes no difference to feasibility.

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## Boolean satisfiability (SAT)

**SAT.** Given a Boolean expression  $\varphi$  with variables  $x_1, \dots, x_n$ , can we make  $\varphi$  true by assigning true/false to  $x_1, \dots, x_n$ ?

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*Horn*-SAT: every clause has at most one positive literal — linear

# Prolog and SAT

Prolog KB (definite clauses)

$x1 \text{ :- } x2, x4.$

$x2 \text{ :- } x3. \quad \rightsquigarrow \quad [[x1, x2, x4], [x2, x3], [x4]]$

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From proofs to **unsatisfiability**:

$\underbrace{KB \text{ proves } \varphi}$	iff	$\underbrace{KB, \overline{\varphi} \text{ is not satisfiable}}$
Prolog		Horn (linear SAT)