Polynomial-Time Reductions

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Formal Languages and Decision Problems



Languages and Decision Problems

Language: A set of strings over some alphabet

Decision problem: A decision problem can be viewed as the formal language consisting of exactly those strings that encode YES instances of the problem.

Yes instance:



No instance:



The Language Prime

Let us encode positive integers in binary representation. The decision problem "Is x a prime?" has the following representation as a formal language: $L_{Primes} = \{10, 11, 101, 111, ...\}$ where 10 encodes 2, 11 encodes 3, 101 encodes 5, and so on.





Let L_1 be a language over an alphabet V_{1} . Let L_2 be a language over an alphabet V_{2} . A polynomial-time reduction from L_1 to L_2 is a function f: $V_1^* \rightarrow V_2^*$ such that (1) f is computable in polynomial time (2) for all x in V_1^* , x is in L_1 if and only if f(x) is in L_2















- YES instances map to YES instances
- NO instances map to NO instances
- computable in polynomial time
- Notation: $L_1 \leq_p L_2$
- [Think: L_2 is at least as hard as L_1]

Theorem If $L_1 \leq_p L_2$ and L_2 is in P, then L_1 is in P.

Proof. Let A_2 be a polynomial time algorithm for L_2 . Here is a polynomial time algorithm A_1 for L_1 .

- •input: x
- compute f(x)
- -run A_2 on input f(x)
- return whatever A₂ returns

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 $|\mathbf{x}| = \mathbf{n}$ takes p(n) time takes q(p(n)) time takes O(1) time

Implications

- Suppose that $L_1 \leq_p L_2$
- If there is a polynomial time algorithm for L_2 , then there is a polynomial time algorithm for L_1 .
- If there is no polynomial time algorithm for L_1 , then there is no polynomial time algorithm for L_2 .

Traveling Salesman Problem

Suppose that we are given a set of cities, distances between all pairs of cities, and a distance bound B.

Traveling Salesman Problem: Does there exist a route that visits each city exactly once and returns to the origin city with a total travel distance <= B?

TSP is in NP: Given a candidate solution (a tour), add up all the distances and check if total is at most B.

Example of a Reduction

Theorem HC \leq_p TSP.

Proof. Given a graph G, the Hamiltonian circuit decision problem tries to decide whether or not G has a Hamiltonian circuit.

A polynomial reduction from HC to TSP has to transform G into an input for the TSP decision problem. More precisely, the graph G needs to be transformed in polynomial time into a configuration of (cities, distances, and bound B) such that

G has a Hamiltonian circuit iff the resulting TSP input has a tour of cities that has a total distance <= B.

The Reduction

Given undirected graph G = (V, E) with m nodes, construct a TSP input like this:

set of m cities, labeled with names of nodes in V

• distance between u and v is 1 if (u,v) is in E, and is 2 otherwise

• bound B = m

This TSP input be constructed in time polynomial in the size of G.

Figure for Reduction

dist(1,2) = 1dist(1,3) = 1dist(1,4) = 1dist(2,3) = 1dist(2,4) = 2dist(3,4) = 1bound = 4

tour w/ distance 4: 1,2,3,4,1

Hamiltonian cycle: 1,2,3,4,1

1

4

HC input

2

3

Figure for Reduction

(1)-2 HC input 4 3

dist(1,2) = 1TSP input dist(1,3) = 1dist(2,4) = 2dist(2,3) = 2dist(1,4) = 1dist(3,4) = 1bound = 4

no tour w/ distance at most 4

no Hamiltonian cycle

Correctness of the Reduction

- Check that input G is in HC (has a Hamiltonian cycle) if and only if the input constructed is in TSP (has a tour of length at most m).
- => Suppose G has a Hamiltonian cycle $v_1, v_2, ..., v_m, v_1$.
 - Then in the TSP input, v_1 , v_2 , ..., v_m , v_1 is a tour (visits every city once and returns to the start) and its distance is $1 \cdot m = B$.

Correctness of the Reduction

- <=: Suppose the TSP input constructed has a tour of total length at</p> most m.
 - Since all distances are either 1 or 2, and there are m of them in the tour, all distances in the tour must be 1.
 - Thus each consecutive pair of cities in the tour correspond to an edge in G.
 - Thus the tour corresponds to a Hamiltonian cycle in G.

Implications

- If there is a polynomial time algorithm for TSP, then there is a polynomial time algorithm for HC.
- If there is no polynomial time algorithm for HC, then there is no polynomial time algorithm TSP.

Transitivity of Reductions

Theorem: If $L_1 \leq_p L_2$ and $L_2 \leq_p L_3$,

then $L_1 \leq_p L_3$.

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