$P \neq NP$

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Abstract

Not for publication! Still in draft form; probably contains errors.

1 Definitions

If a string s = vw, then v is a prefix of s and w is a suffix of s.

For a language L, define $L_n \subseteq L$ to consist of strings in L of length n: $L_n = \{s \mid s \in L \text{ and } |s| = n\}.$

The notation g_e refers to a graph with e edges. G_e is the set of all graphs with up to e edges. The notation g; e refers to the graph which results from adding edge e to graph g.

A clique is a subset of the vertices of a graph such that every pair in the subset is adjacent. A subclique is a clique smaller than a given desired size. If g is a graph with e edges and k > 1 is a constant, we define P(g,k) to be the Clique decision problem where the answer is "yes" if there is a clique of size $\lceil \frac{e}{k} \rceil$ in g, "no" otherwise. We observe that P(g,k) is an NP-Complete problem. We define Clique(e,k) to be the set of graphs with e edges such that P(g,k) is "yes".

Following Sipser [1], a deterministic finite automaton (DFA) is a tuple $A = (\Sigma, S, F, \delta, s_0)$ where Σ is a finite alphabet, S is a finite set of states, $F \subseteq S$ is the set of final states, $\delta : S \times \Sigma \to S$ is a mapping called the transition function, and $s_0 \in S$ is the initial state. We extend δ to deal with whole strings with the extended transition function $\hat{\delta} : Q \times \Sigma^* \to Q$ which is defined inductively where $a \in \Sigma$ and $\hat{\delta}(q, \epsilon) = q$ and $\hat{\delta}(q, ax) = \hat{\delta}(\delta(q, a), x)$.

If $a \in \Sigma$ and the state $s' = \delta(s, a)$ exists, then we can view the pair (s, s') as a a transition in A. So the automaton defines a directed graph with

vertices S and edges $\{(s, s') \mid s \in S \text{ and } a \in A\}$. We then get the usual graph concepts like a path, shortest path, and depth in their usual meanings. The *depth* of a state $s \in S$ is the length of the shortest path from the start state s_0 to s. A *level* is the set of all states in S at a given depth.

1.1 Creating an Automaton from a Finite Language

We will be creating a finite state automaton from a finite language via the following standard procedure. Since the language is finite, it can be written $\{w_1, w_2, w_3, ..., w_m\}$ where $w_i \in \Sigma^*$. This set can then be turned into the following regular expression: $E = w_1 \cup w_2 \cup w_3 \cup ... \cup w_m$. E is then converted to a non-deterministic finite automaton N. Then N is converted into a deterministic finite automaton D. Finally, D is minimized to yield the minimal automaton we are looking for [1].

1.2 Defining Automata

We will use the adjacency matrix encoding for Clique(e, k). So in particular, there is a finite alphabet $\Sigma_C = \{0, 1\}$ to represent Clique(e, k). By the construction in Section 1.1, we can create the automaton: $C = (\Sigma_C, S_C, F_C, \delta_C, s_{C0})$. So C accepts x iff $x \in \text{Clique}(e, k)$. We observe that this automaton reads one edge at a time.

2 Results

Lemma 2.1. For a fixed c as n grows, the minimal automaton for Clique(e, k) has an exponential number of states $\Omega(2^{\sqrt{e}})$ at level $\lceil \frac{e}{2} \rceil$.

Proof. The Myhill-Nerode theorem will be used to prove this. We will use the usual idea of a distinguishing extension to set this up. In general, given a language L and strings x and y where $x \neq y$, a distinguishing extension is a string z such that only one of the two strings x or y is in L.

In our case, let the language $L_n \subseteq \Sigma_C^*$ be the representation of Clique(e, k) as described above. We set up x and y as two different prefixes of instances of Clique(e, k) which don't themselves contain cliques. In more detail, we first note that in a graph of v vertices, there are potentially $\binom{v}{m}$ cliques of size m. Also, $\binom{2w}{w} = \Omega(2^w)$, and more generally, $\binom{2cw}{w} = \Omega(2^w)$ for $c \geq 1$. So in a graph with 2 * c * w vertices where we can choose cliques of size w,

the number of choices is $\Omega(2^w)$. This means, then, in a graph with v vertices where we can choose cliques of size $\frac{v}{2c}$, the number of choices is still $\Omega(2^v)$. And since the relationship between edges and vertices in a graph is at worst quadratic, the number of choices is $\Omega(2^{\sqrt{e}})$.

So we can choose among these size $\frac{v}{2c}$ subcliques to form the prefix strings x and y.

The string z, then, is a suffix such that $x \in L_n$ and $y \notin L_n$. This is accomplished by choosing z so that it completes the subclique encoded in x but not in y. Since there are $\Omega(2^{\sqrt{e}})$ subcliques encoded in strings x, y, and z, we have $\Omega(2^{\sqrt{e}})$ states in the minimal automaton at level $\lceil \frac{e}{2} \rceil$ by the Myhill-Nerode theorem.

Lemma 2.2. The minimal automaton for a sparse language has only a polynomial number of states at each depth. That is, for all sparse languages L and for all n, there exists a constant c such that the subset $L_n \subseteq L$ produces a minimal DFA such that each level has only $O(n^c)$ states.

Proof. By the definition of a sparse language,

$$|L_n| = O(n^c)$$

Now if we take the set of all prefixes at a certain depth d,

$$S_n = \{ w \mid wv \in L_n, |w| = d \}$$

 S_n is also polynomially bounded:

$$|S_n| = O(n^c)$$

So by the Myhill-Nerode theorem, there are only a polynomial number of states at each depth in the minimal automaton, even if all the prefixes are distinguishable from one another by different suffixes. \Box

3 Conclusions

Proof Idea: Using the assumption that there exists an NP-Complete sparse language, we will use that to build a "smaller" automaton for Clique, smaller than the minimal finite automaton based on the language of Clique(e, k). This is a contradiction, so there can be no such NP-Complete sparse language.

Theorem 3.1. $P \neq NP$.

Proof. The proof technique here is Reductio ad Absurdum. Suppose there exists an NP-Complete sparse language L. Since L is NP-Complete, there is a polynomial-time reduction which takes a graph g and maps it to an element of $x \in L$. Let's call this function r(g).

Let's restrict L to inputs of size p where p is big enough to accommodate all the input graphs with up to e edges:

$$p = \max\{r(g) \mid |g| \le e\}$$

For notational convenience later, we'll call this restriction L(p). Using the standard construction described in section 1.1, we create a minimal finite-state automaton for L(p); we'll call it $M(L(p)) = (\Sigma_L, S_L, F_L, \delta_L, s_{L0})$. So M(L(p)) accepts x iff $x \in L(p)$.

Let's define the function $s(x) = \hat{\delta}(s_{L0}, x)$; this takes us from the start state of M(L(p)) to the state in M(L(p)) corresponding to input x. So s(x) takes $x \in L(p)$ and maps it to a state s in M(L(p)).

Our job now is to build a machine for Clique based on the machine for L: $X = (\Sigma_X, S_X, F_X, \delta_X, s_{X0})$.

Alphabet, equation 1: The alphabet for the X machine is the same as the machine for clique because they are reading in the same inputs: graphs.

States, equation 2: the states for the X machine are the same as the sparse language machine L, because we are translating graphs into the sparse language.

Transition function, equation 3: the transition function for the X machine is more complicated. What we are going to do is read in a graph, translate to the sparse language, then find the resulting state in the sparse automaton. This gives us the starting state in the transition we are about to create. We then take that same input graph and add a new edge. We then go through this same translation to the sparse language and then find the resulting state in the sparse automaton. This gives us the ending state of our new transition. Finally we just add the edge to complete the whole transition.

The start state of the X machine is the same as the start state in the sparse automaton (equation 5).

Similarly, the final states of the X machine (equation 4) are the same as the final states of the sparse automaton.

$$\Sigma_X = \Sigma_C \tag{1}$$

$$S_X = \{ s(r(g)) \mid g \in G_e \} \tag{2}$$

$$\delta_X = \{ (s(r(g)), e) \to s(r(z)) \mid g \in G_{e-1}, \ z = (g; e) \}$$
 (3)

$$F_X = F_L \tag{4}$$

$$s_{X0} = s_{L0} \tag{5}$$

So we've built a machine X that accepts iff the original Clique machine C accepts. By Lemma 2.2, S_X has only $O(n^c)$ states at level $\lceil \frac{n}{2} \rceil$ for some fixed c. But by Lemma 2.1, there are $\Omega(2^{\sqrt{n}})$ states at level $\lceil \frac{n}{2} \rceil$ in the minimal machine. This is a contradiction. So by Reductio ad Absurdum, our original assumption that there exists a sparse NP-Complete language is wrong.

The biconditional form of Mahaney's Theorem states that there is a sparse NP-Complete language iff P = NP. Since we've just shown there is no such NP-Complete language, $P \neq NP$.

References

[1] Michael Sipser. Introduction to the Theory of Computation. Cengage, Boston, 3 edition, 2013.