

NP-Completeness

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Polynomial Time $\mathcal{O}(n^k)$ is Good.



- Most (if not all) problems we have studied so far can be solved efficiently
 - Searching
 - Sorting
 - Single-Source Shortest Path
 - All-Pair Shortest Path
 - Fractional Knapsack
 - etc.
- ► All of these problems can be solved in polynomial time wrt. the input length.
- Polynomial Time \Rightarrow Easy (or Tractable)

Exponential Time $\Omega(c^n)$ is Bad.



- Some problems are hard(er) to solve.
 - Tower of Hanoi
 - N-Chess
- Lowers bound on their running time have been shown to be exponential wrt. the input length.
- ► Exponential Time ⇒ Hard (or Intractable)



Figure: Tower of Hanoi¹

¹Image Courtesy of Wikipedia

Don't Worry about the Extremes !!!



What if k is large for $\mathcal{O}(n^k)$?

What about a problem with a polynomial runtime $\mathcal{O}(n^{1000000})$?

Is it still considered efficient?

Don't be concerned too much with such extremes:

- Such extreme cases are extremely rare (if ever existing) in practice unless deliberately designed to be slow.
- Many problems could be solved with less efficient polynomial algorithms when they were first discovered than the currently best ones.

Bridging the Algorithmic Gap



Before merge sort was invented,

• we knew only algorithms like selection sort, insertion sort, all of which can sort in $\mathcal{O}(n^2)$ time

Later, it was proven that

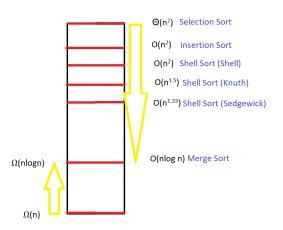
- ▶ any comparison-based sorting algorithm needs at least $\Omega(n \log n)$ time.
- ▶ people then started to search for a sorting algorithm that needs at most $\mathcal{O}(n \log n)$.

When merge sort was discovered, which needs $\mathcal{O}(n \log n)$,

- the algorithmic gap for the sorting problem was closed
- we are certain that we cannot find any better algorithm than $\mathcal{O}(n \log n)$ because of the lower bound $\Omega(n \log n)$

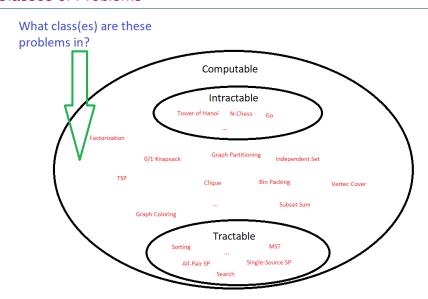
Bridging the Algorithmic Gap





Classes of Problems





What class(es) do those problems in the middle belong to?



- They are problems whose polynomial-time algorithms have not been yet discovered.
- However, we still cannot prove that there exist lower bounds that are exponential.
- That is, their algorithmic gaps are still large.

So are they considered tractable or intractable?

- ▶ No one knows the answer yet.
- But, most theoretical computer scientists believe that they are intractable and cannot be solved in polynomial time.
- These problems are the topic of our discussion today.

Decision Problems



Decision problem

- X is a set of strings
- Instance: string s
- ▶ Algorithm A solves X: $A(s) = yes \iff s \in X$

Every optimization problem can be turned into a corresponding decision problem.

- ► Shortest-Path-Opt: Find a shortest path between *u* and *v* in *G*.
- ► Shortest-Path-Dec: Is there a shortest path between *u* and *v* of length at most *k* in *G*?

Optimization Problems ⇒ Decision Problems



By introducing an integer k, an optimization problem can be cast into a corresponding decision problem.

- ▶ Minimization Problem ⇒ at most k
- ► Maximization Problem ⇒ at least k

Optimization Problems ⇒ Decision Problems : Minimization Problem



Travelling Salesman (Minimization Problem):

- ▶ TSP-Opt: Given a directed graph G = (V, E), find a shortest tour that visits each vertex in V exactly once except for the first one.
- ▶ TSP-Dec: Given a directed graph G = (V, E), is there a tour with length at most k that visits each vertex in V exactly once except for the first vertex?

Optimization Problems ⇒ Decision Problems : Maximization Problem



Independent Set (Maximization Problem):

- ▶ IS-Opt: Given an undirected graph G = (V, E), find a largest possible subset $W \subseteq V$ s.t. no pair of vertices in W is connected by an edge.
- IS-Dec:Given an undirected graph G = (V, E), is there a subset W ⊆ V of size at least k s.t. no pair of vertices in W is connected by an edge?

Hardness: Decision Problems VS. Optimization Counterparts



Let Q_{DEC} be the decision problem of an optimization problem Q_{OPT} .

It is obvious that

▶ Q_{DEC} is hard $\implies Q_{OPT}$ is also hard.

By the contrapositive,

▶ Q_{OPT} is easy $\Longrightarrow Q_{DEC}$ is also easy.

This result suggests that

- ▶ we can efficiently solve Q_{DEC} by
 - efficiently solving Q_{OPT}
 - 2. comparing the output of Q_{OPT} and the value k of Q_{DEC} .

But But, some Q_{DEC} can be used to solve their corresponding Q_{OPT} .

▶ Caveat: not always the case for any pair of (Q_{DEC}, Q_{OPT}) .

Graph Colouring



Graph Colouring:

- ▶ GC-OPT:Given an undirected graph G = (V, E), find the minimum number of colours that can be assigned to the vertices s.t. no two adjacent vertices are of the same colour.
- ▶ GC-DEC:Given an undirected graph G = (V, E), is there a colour assignment to the vertices using at most k colours s.t. no two adjacent vertices are of the same colour?

Graph Colouring: Using GC_{OPT} to solve GC_{DEC}



Use GC_{OPT} to solve GC_{DEC} :

```
GC_{DEC}(G, k){
c = GC_{OPT}(G)
return c <= k
}
```

Graph Colouring: Using GC_{DEC} to solve GC_{OPT}



Use GC_{DEC} to solve GC_{OPT} :

```
GC_{OPT}(G=(V,E))\{\\ \text{for } (k=1:k<=|V|:k++) \text{ do}\\ \text{if } (GC_{DEC}(G,k)) \text{ then}\\ \text{return } k\\ \text{end if}\\ \text{end for} \}
```

 GC_{DEC} can solve GC_{OPT} by repeatedly asking $GC_{DEC}(G, k)$ outputs a yes for different values of k.

- Q1: Is there a better algorithm Q_{OPT} than this one?
- Hint: Divide and Conquer

The P Class



The problems in the P Class are decision problems that can be solved (decided) in polynomial time.

► In other words, given a problem instance, they output yes or no in polynomial time.

Examples of the problems in P includes:

- ► Minimum-Spanning-Tree: Given an undirected graph *G*, is there a spanning tree having a total length of at most *k*?
- ► Longest-Common-Subsequent: Given two strings *X* and *Y*, is there a common substring of both *X* and *Y* having a length of at least *k*?
- ▶ Majority: Given an array A of length N, if there is a majority s.t. the majority appears in A at least $\frac{N}{2} + 1$ times?
- Primality-Test***: Given an integer N, determine if N is a prime number.

The NP Class



The problems in the NP Class are decision problems for which any yes can be verified in polynomial time.

- ► In other words, given a yes instance together with a certificate, they can verify it in polynomial time.
- ► The encoding length of a certificate must be at most polynomial in length wrt. the encoding length of the given instance.

Majority ∈ NP?



We know that Majority $\in P$

 \blacktriangleright there exists a polynomial-time algorithm that can decide in $\mathcal{O}(\textit{N}^2)$ time.

```
\begin{array}{l} \textit{MAJ}_{\textit{DEC}}(A[1...N]) \{\\ & \textbf{for} \; (i=1:i<=N:i++) \; \textbf{do} \\ & c = count(A,A[i]) \\ & \text{if} \; (c>N/2) \; \textbf{then} \\ & \text{return} \; \; \textit{true} \\ & \textbf{end} \; \textbf{if} \\ & \textbf{end} \; \textbf{for} \\ & \text{return} \; \; \textit{false} \\ \} \end{array}
```

How can we show Majority \in NP?

MAJ^{YES}_{CER}: Majority Yes-Certifier



- ► Instance: A[1...N]
- Yes-Certificate: maj
- Runtime: Θ(N)

```
egin{align*} & \emph{MAJ}_{CER}^{YES}(A[1...N], \emph{maj}) \{ & \emph{c} = 0 & \emph{for} \ (\emph{i} = 1 : \emph{i} <= \emph{N} : \emph{i} + +) \ \emph{do} & \emph{if} \ (A[\emph{i}] == \emph{maj}) \ \emph{then} & \emph{c} = \emph{c} + 1 & \emph{end if} & \emph{end for} & \emph{return} \ \emph{c} > (\emph{N}/2) & \emph{} \} \end{split}
```

There exists a polynomial-time algorithm that can verify in $\Theta(N)$ time that the array A of length N has a majority by giving it the majority maj as a certificate.

$MAJ_{DFC} \in NP$



As we have found MAJ_{CER}^{YES} , which is a polynomial-time certifier, we can now conclude that

► MAJ_{DEC} ∈ NP

Actually, since we know $MAJ_{DEC} \in P$, we need not have looked for such a polynomial-time certifier MAJ_{DEC} to prove that it is in NP.

- ▶ Q2: Why?
- ▶ Q3: Show that $P \subseteq NP$.

GC^{YES}_{CER}: Graph-Coloring Yes-Certifier



- ► Instance : G[1...N]G[1...N] and k
- Yes-Certificate : color[1...N]
- ► Runtime : O(N²)

```
GC_{CEP}^{YES}(G[1...N][1...N], k, color[1...N])
    nColors = max(color)
    if (nColors > k) then
       return false
    end if
    for (i = 1 : i \le N : i + +) do
       for (j = i + 1 : j \le N : j + +) do
              if (G[i][j] == true  and color[i] == color[j])  then
                     return false
              end if
       end for
    end for
    return true
```

$GC_{DFC} \in NP$



Since there exists a polynomial-time algorithm GC_{CER}^{YES} that can verify a yes instance,

▶ $GC_{DEC} \in NP$

Excercise:

► Q4: Prove the Subset-Sum Problem ∈ NP.

The co-NP Class



The problems in the co-NP Class are decision problems for which any no can be verified in polynomial time.

- ► In other words, given a no instance together with a certificate, they can verify it in polynomial time.
- ► The encoding length of a certificate must be at most polynomial in length wrt. the encoding length of the given instance.

Note that the co- prefix stands for complementary in the sense that it is the complementary problem class to the NP class.

Primality Testing (Prime)



Provided a certificate c, $Prime_{CER}^{NO}$ verifies a given no instance N in polynomial time.

- Instance: N
- No-Certificate: c
- ▶ Runtime: $\Theta((\log N)^2)$

```
Prime_{CFR}^{NO}(N, c){
     return (N \mod c) == 0
```

Therefore, Prime \in co-NP.

Actually,

- Prime is also in NP.
- But, a yes-certificate (called Pratt certificate) is quite tricky to find.

Reduction



That Q_i is polynomially reducible to Q_i is denoted by

- $ightharpoonup Q_i \leq_p Q_j$.
- ▶ all instances of Q_i are transformed to corresponding instances of Q_i.

 $Q_i \leq_p Q_j$ is equivalent to saying:

- $ightharpoonup Q_i$ is no harder than Q_i .
- $ightharpoonup Q_i$ is at least as hard as Q_i .
- ▶ If Q_i is efficiently solvable, so is Q_i .

Reduction : SQR \leq_p MULT



- Instances of SQR: x
- ► Instances of MULT: a, b
- ▶ Instance Transformation: a = x, b = x

```
SQR(x){
a = x, b = x
return MULT(a, b)
}
```

We have just shown that SQR \leq_{ρ} MULT.

SQR is no harder than MULT.

Reduction : $MULT \leq_{D} SQR$



- Instances of MULT: a, b
- ▶ Instances of SQR: x , y
- ▶ Instance Transformation: x = a + b, y = a b

```
MULT(a,b)\{ \\ x = a+b, y = a-b \\ \textbf{return} \ (SQR(x) - SQR(y))/4 \\ \}
```

We have just shown that MULT \leq_p SQR.

► MULT is no harder than SQR.

$MULT \leq_{p} SQR$ and $SQR \leq_{p} MULT$



Since MULT \leq_{ρ} SQR and SQR \leq_{ρ} MULT,

- MULT and SQR are equally hard (also equally easy).
- Their hardness levels are the same.

What can reduction tell?



Suppose $A \leq_{p} B$.

- ▶ If *B* has a polynomial time algorithm, so does *A*.
- ► If *B* is easy, so is *A*.
- ▶ If A is hard, so is B.
- A is no harder than B.

If we can polynomially reduce any pair of problems A and B to each other,

A and B are equally hard.

the NPC Class



Cook-Levin Theorem shows that SAT is NP-Complete:

That is, all problems in NP can be reduced to SAT.

▶ $\forall q \in NP : q \leq_{p} SAT$

What does it mean for a problem A to be NP-Complete?

- ▶ *A* is among the hardest problems in *NP*.
- ▶ If we know how to solve A in polynomial time, we also know how to solve all problems in NP in polynomial time.
- ➤ That is, we would be able to solve hard problems such as Vertex Cover, Independent Set etc in polynomial time as well.
- ▶ One immediate result is that P = NP.

But, until now, we have not discovered any polynomial-time algorithm for any problems in NP.

So it still remains a mystery whether P = NP.

Graph Colouring: Solution to Q1



Use GC_{DEC} to solve GC_{OPT} :

```
GC_{OPT}(G = (V, E)){
    low = 1, high = |V|, k = high
    while low <= high do
      mid = (low + high)/2
      if GC_{DFC}(G, mid) then
             if mid < k then
                    k = mid
             end if
             high = mid - 1
      else
             low = mid + 1
      end if
    end while
    return k
```