

P, NP, and Complexity

- Six fundamental facts
- One rule of thumb
- Three fundamental notions
- One fundamental question

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Fundamental Fact #1

Exponential functions are eventually bigger than polynomial functions

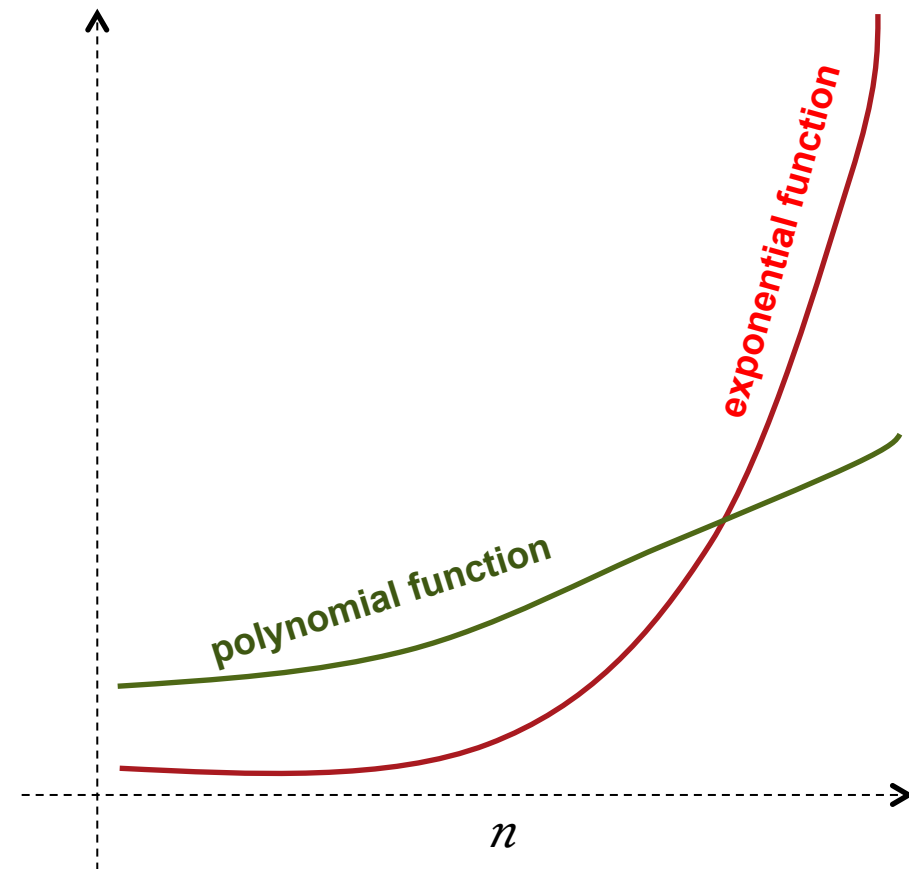
Exponential functions are eventually bigger than polynomial functions

Examples of exponential functions:

- $f(n) = 2^n$
- $f(n) = e^{n-7} + 17$
- $f(n) = n^{2n-1}$

Examples of polynomial functions:

- $f(n) = n^2 + n + 21$
- $f(n) = 26n^7 + 4n^5 + 231n^2 + 21$





Fundamental Fact #2

Some algorithms are efficient, some are not

Some algorithms are efficient, some are not

Sorting problem:

Evelyn
Georgie
Franz
Chen
Bob
Alice
Doug

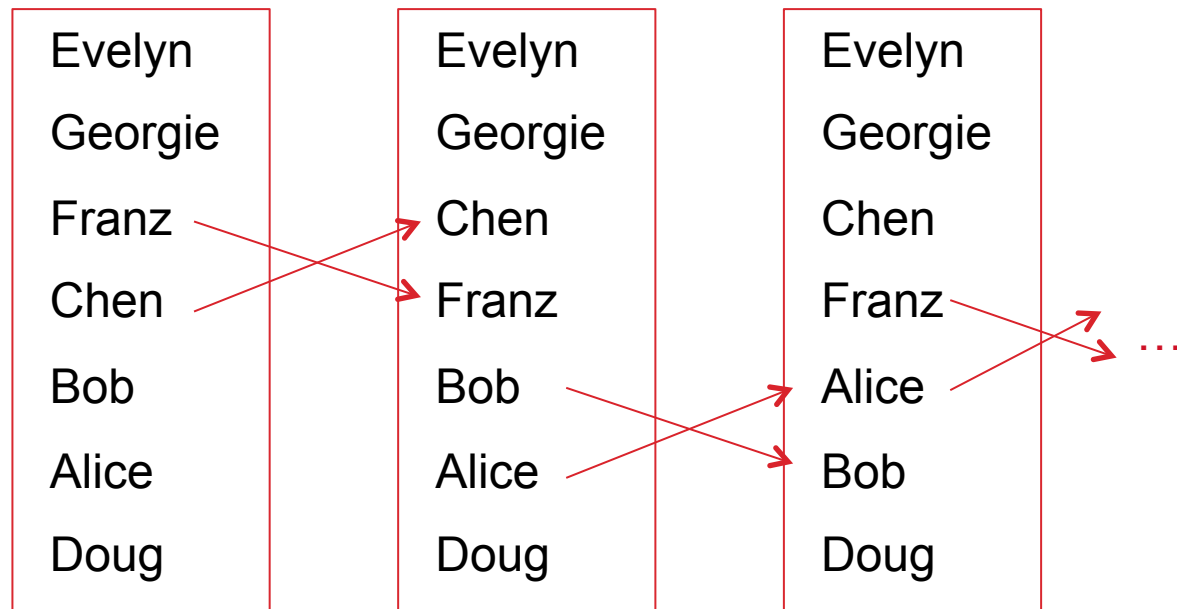
? Algorithm ?

Alice
Bob
Chen
Doug
Evelyn
Franz
Georgie

There are many different algorithms for sorting

Basic step:

- › If two things are out of order, we can swap them to improve the sortedness of the list



exhaustingSort

- Exhaustively swap to get every possible ordering of the list
- At each step, check to see if the list is ordered

Run-time for ***exhaustingSort***:

- You need to consider $n!$ different orderings of the list
- This takes time at least proportional to $n! \cong \sqrt{2\pi n} (n/e)^n$
- This is infeasible

bubbleSort

- Scan the list from top to bottom
- If the i th and $(i+1)$ th elements are out of order, then swap them
- Keep scanning and swapping until the list is sorted

Run-time for ***bubbleSort***:

- you need to scan the list n times
- make $0.5n^2 - 1.5n + 2$ swaps.

This takes time proportional to n^2 .

geeky_bubbleSort

- Same old ***bubbleSort*** algorithm, but use really clever coding in assembler and C languages to make it go faster.
- Run on a fast computer.

Run-time for ***geeky_bubbleSort***:

- you still need to scan the list n times
- you still make $0.5n^2 - 1.5n + 2$ swaps.

It still takes time proportional to n^2 .

But it seems to be much faster

AISASort

- Measure the “sortedness” of the sequence:

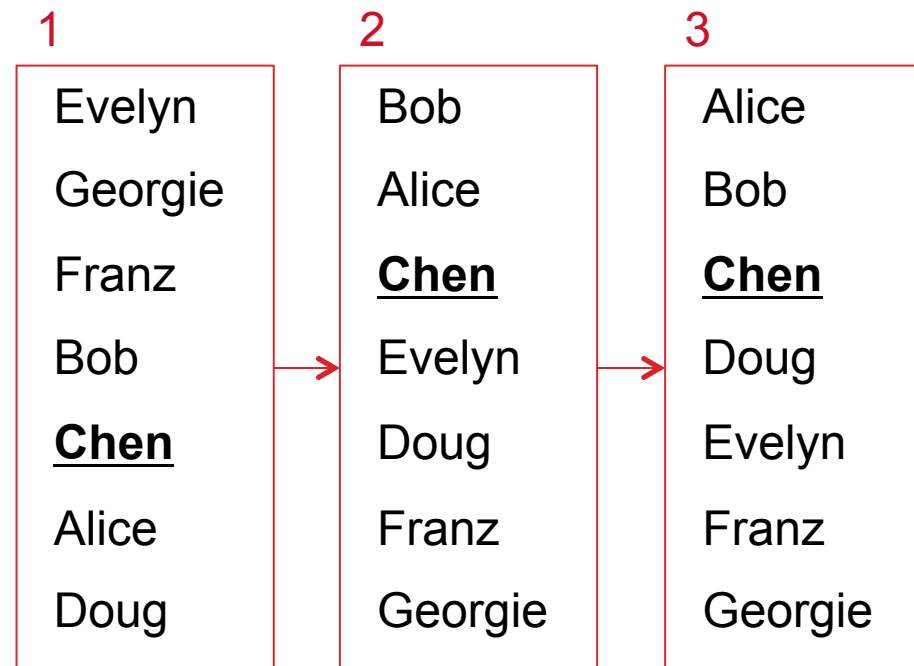
$$\text{sortedness}(x_1, x_2, \dots, x_n) = \sum_{i=1}^{n-1} (x_{i+1} - x_i)$$

- Choose swaps that increase *sortedness* as much as possible
- Use an AI technique called “simulated annealing” to search for a maximally sorted sequence

Run-time for ***AISASort***: exponential

quickSort

1. Choose an element, called a *pivot*, from the list.
2. Swap things around so that smaller things come before the pivot, and larger things come after it
3. Recursively sort the sub-list of lesser elements and the sub-list of greater elements.



Run-time for *quickSort*: proportional to $n \log n$

How long does it take to sort n things?

n	exhausting Sort	bubbleSort	geeky _bubbleSort	AISASort	quickSort
10	About 100	0	0	0.001	0
100	-	0	0	4.924*	0
10000	-	0.123	0.170	12266.808*	0.006
1000000	-	1296.560	734.422	-	0.111

Remarks:

- **quickSort** is excellent
- **bubbleSort** and **geeky_bubbleSort** are feasible
- **AISASort** is infeasible (as well as **exhaustingSort**)



Rule of Thumb #1

An algorithm that runs in exponential time is not feasible; an algorithm that runs in polynomial time may be feasible.

***An algorithm that runs in exponential time is not feasible;
an algorithm that runs in polynomial time may be feasible.***

› Infeasible

- exhaustingSort
- AISASort

› Feasible

- bubbleSort
- quickSort

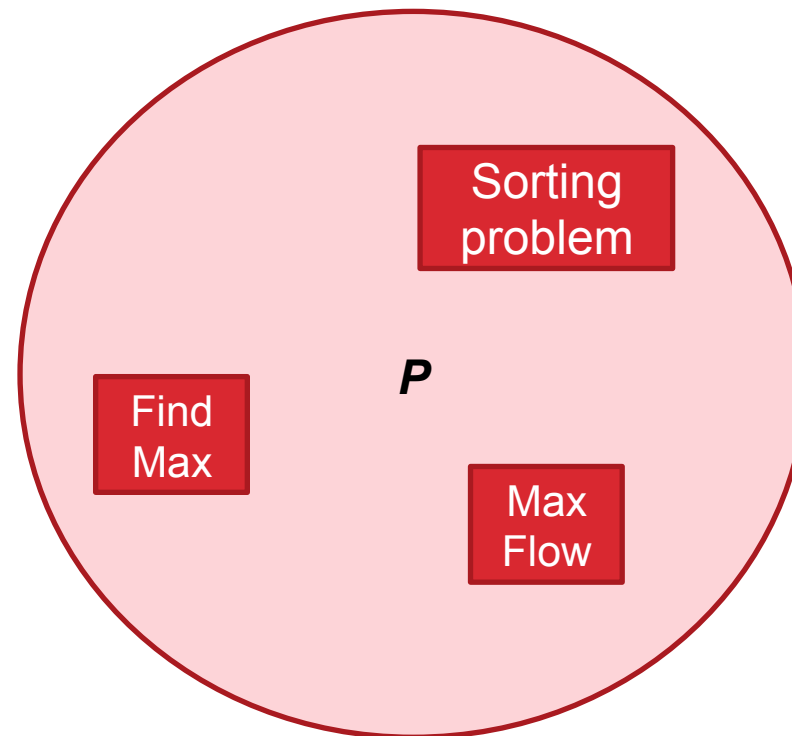


Fundamental Notion #1: P

***P is the set of all problems that can be solved
in polynomial time***



Fundamental Notion #1: P





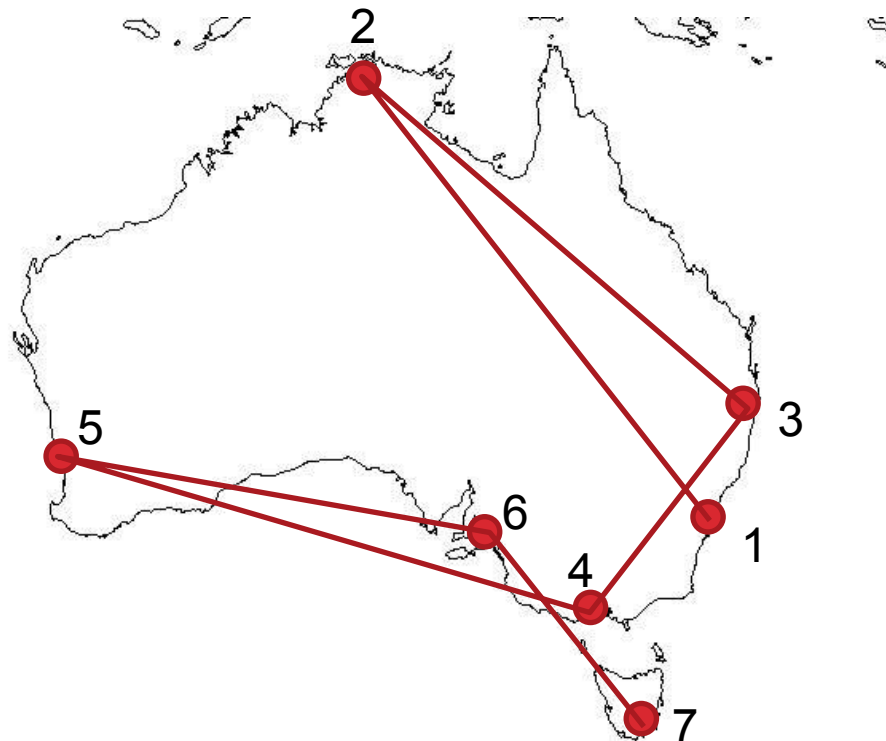
Fundamental Fact #3

Some problems can be solved with efficient algorithms, and some others ... maybe not

Some problems can be solved with efficient algorithms, and some others ... maybe not

Travelling salesman problem:

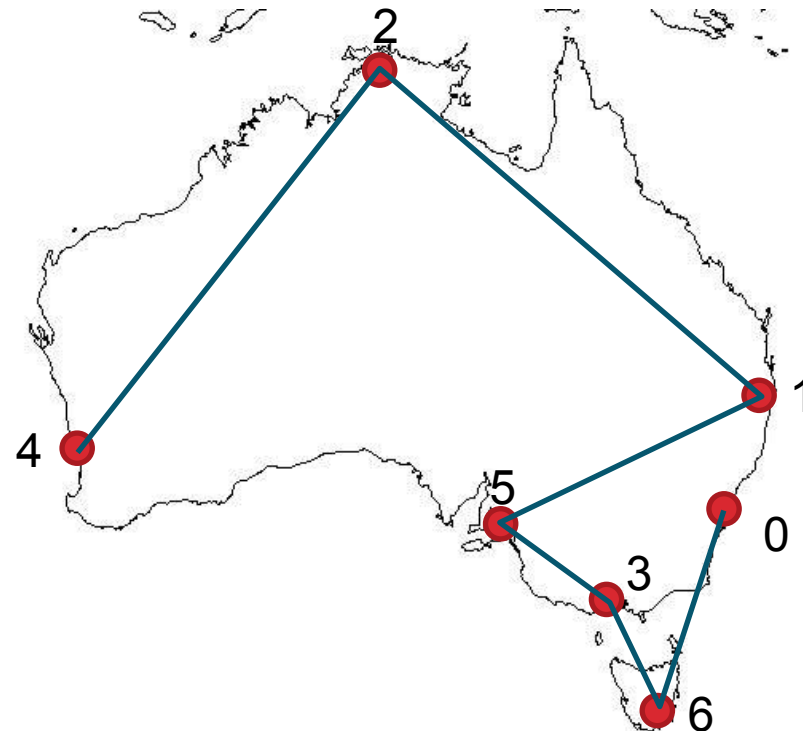
- How can we choose a route around n cities to minimise the total distance travelled?
- We need to order the cities appropriately



Some problems can be solved with efficient algorithms, and some others ... maybe not

Travelling salesman problem:

- How can we choose a route around n cities to minimise the total distance travelled?
- We need to order the cities appropriately



There are many different algorithms for the traveling salesman problem

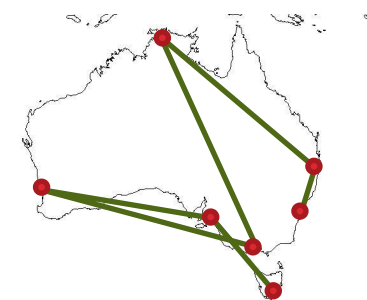
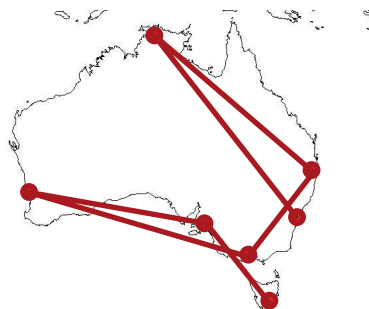
Basic problem:

- › We need to swap around the order of cities to decrease the length travelled

Like the sorting problem

1. Sydney
2. Darwin
3. Brisbane
4. Melbourne
5. Perth
6. Adelaide
7. Hobart

1. Sydney
2. Brisbane
3. Darwin
4. Melbourne
5. Perth
6. Adelaide
7. Hobart



exhaustingTSP

- Exhaustively swap to get every possible ordering of the cities
- At each step, compute the total distance travelled
- Keep the tour with the minimum total distance

Run-time for ***exhaustingTSP***:

- You need to consider $n!$ different orderings of the cities
- This takes time at least proportional to $n! \cong \sqrt{2\pi n} (n/e)^n$
- This is infeasible

bubbleTSP

- Scan the current list of cities
- If swapping the i th and $(i+1)$ th cities would decrease distance, then swap them
- Keep scanning and swapping until no swap decreases distance

Run-time for ***bubbleTSP***:

- Hard theorem: it can take exponentially time, but not as bad as ***exhaustingTSP***

Effectiveness of ***bubbleTSP***:

- Does not give optimal results
- Gives mediocre results

AISA_TSP

- Choose swaps that decrease distance as much as possible
- Use an AI technique called “simulated annealing” to search for an ordering with smallest total distance

Run-time for ***AISA_TSP***:

- Hard theorem: it can take exponentially long
- Faster than ***exhaustingTSP***
- Slower than ***bubbleTSP***

Effectiveness of ***AISA_TSP***:

- Does not give optimal results
- Gives mediocre results, but better than ***bubbleTSP***

How long does it take to find a solution to TSP?

n	AISA_TSP
10	0.052
100	96.013
10000	-
1000000	-

Remarks:

- ***We don't know any algorithm for TSP that runs fast and gives an optimal result***



Fundamental Fact #3

Some problems can be solved with efficient algorithms, and some others ... maybe not

(continued)



Clique problem:

- Given a network \mathcal{N} of size n and an integer k
- Does \mathcal{N} have k nodes that are all connected to each other?

Some people:

- Alice, Andrea, Annie, Amelia, Bob, Brian, Bernard, Boyle

Some connections

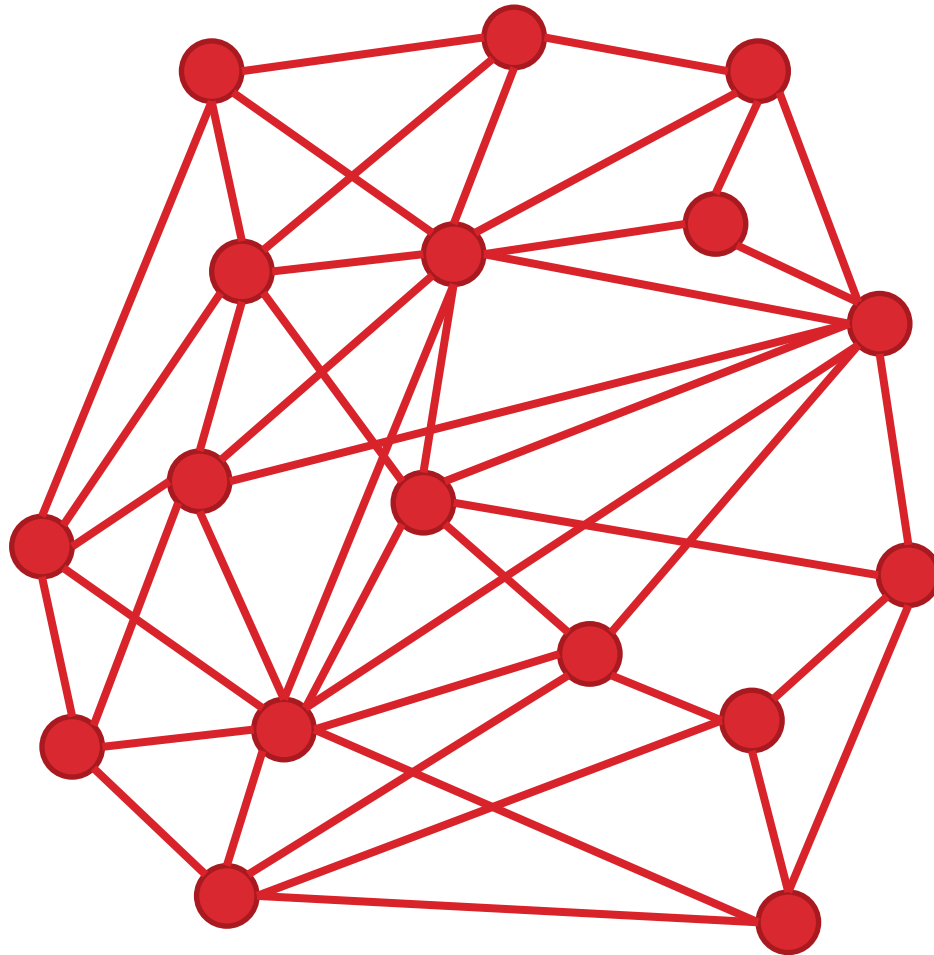
- Bob is connected to Alice
- Bob is connected to Andrea
- Bob is connected to Amelia
- Brian is connected to Alice
- Brian is connected to Andrea
- Brian is connected to Amelia
- Boyle is connected to Alice
- Boyle is connected to Andrea
- Boyle is connected to Annie
- Bernard is connected to Alice
- Bernard is connected to Andrea
- Bernard is connected to Annie

Is there a clique of size 3 among these people?

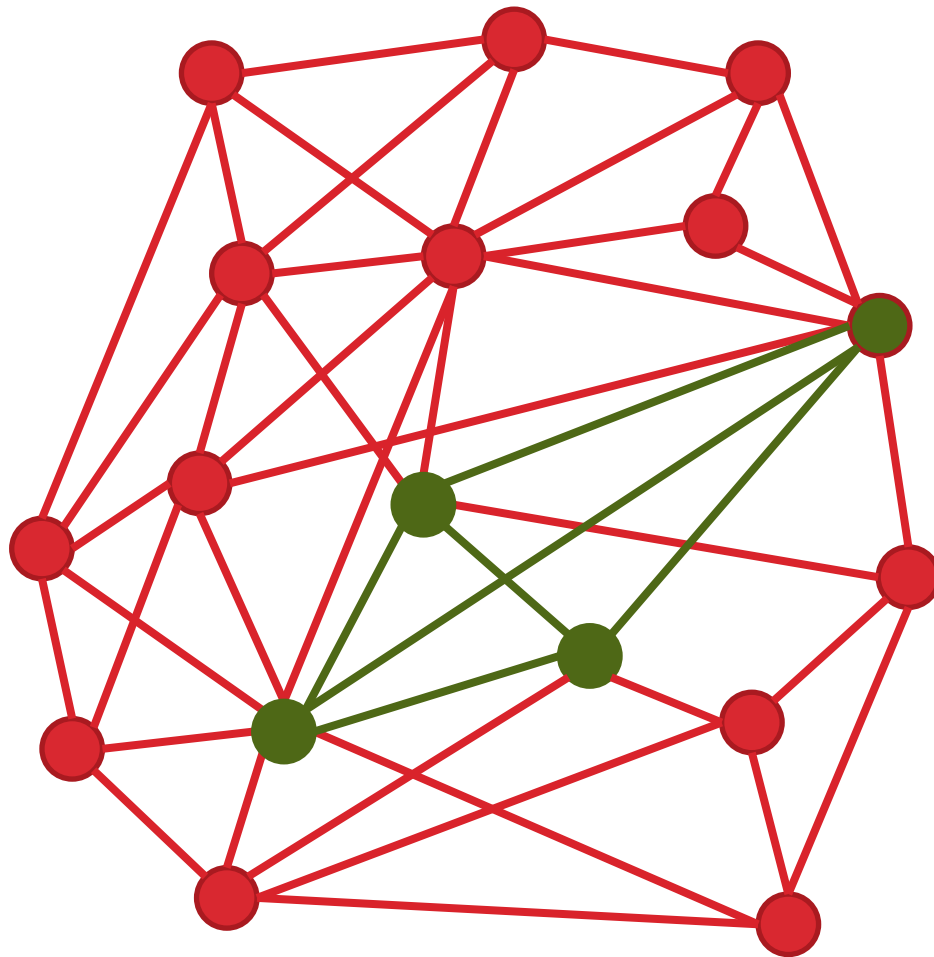
(Clique: a group of people who are all connected to each other)



Another network



*Is there a clique of size 5
among these nodes?*



Clique of size 4

exhaustingClique

- Test every subset of k nodes, check to see if these k nodes form a clique

Run-time for ***exhaustingClique***:

- You need to consider $\binom{n}{k}$ different subsets of nodes
- If $k \cong n/2$ then is $\binom{n}{k}$ exponential
- This is infeasible



Fundamental Fact #3

Some problems can be solved with efficient algorithms, and some others ... maybe not

(still continued)

The Satisfiability (SAT) problem

A statement \mathcal{P}

If Peter is Australian, then Patrick is Australian, and

Either Patrick is Australian, or Patrick is Irish, or Paul is English, and

If Catherine is English then either Paul is not English or Jane is French, and

If Jane is French and Patrick is not Australian, then Peter is English, and

If Peter is English then Peter is not Irish, and

If Paul is not English then either Catherine is Australian or Jane is French, and

....

Question:

- Is there any way that this statement could be true?

SAT

Given an abstract proposition

If $p \downarrow 1$ then $p \downarrow 2$, and

Either $p \downarrow 3$, or $p \downarrow 1$, or $p \downarrow 4$,
and

If $p \downarrow 1$ then either $p \downarrow 2$ or
 $p \downarrow 1$, and

If $p \downarrow 1$ and $p \downarrow 4$, then not
 $p \downarrow 3$, and

If $p \downarrow 2$ then not $p \downarrow 1$, and

If $p \downarrow 3$ then either $p \downarrow 2$ or

$p \downarrow 1$, and

Are there truth values for $p \downarrow 1$, $p \downarrow 2$,

$p \downarrow 3$, $p \downarrow 4$ such that this statement is

true?

exhaustingSAT

- Test every combination of true and false values for p_1, p_2, p_3, \dots , check to see whether the whole statement is true

Run-time for *exhaustingSAT*:

- You need to consider 2^n different combinations of true and false values for p_1, p_2, p_3, \dots
- This is infeasible

Some problems can be solved with efficient algorithms, and some others ... maybe not

The Travelling Salesman Problem, the Clique Problem, and SAT all share some characteristics:

- They can be solved by exhausting algorithms that use exponential time
- No-one knows any algorithms that run fast and always solve these problems

Remarks

- These problems are commercially important, and a huge amount of research has been done on them.
- There are many algorithms for specific versions, and many algorithms that almost work
- There are many other problems with the same characteristics



Fundamental Fact #4

Sometimes we can efficiently check whether an answer is correct, even if we can't efficiently find a correct answer

Sometimes we can efficiently check whether an answer is correct, even if we can't efficiently find a correct answer

› Clique Problem:

- Given a set of k nodes, one can quickly check whether these k nodes are all connected to each other.
- But it seems difficult to find the *right* set of k nodes.

› SAT:

- Given a truth value for each variable p_i it is easy to check whether these truth values make the statement true.
- But it seems hard to find the *right* truth values

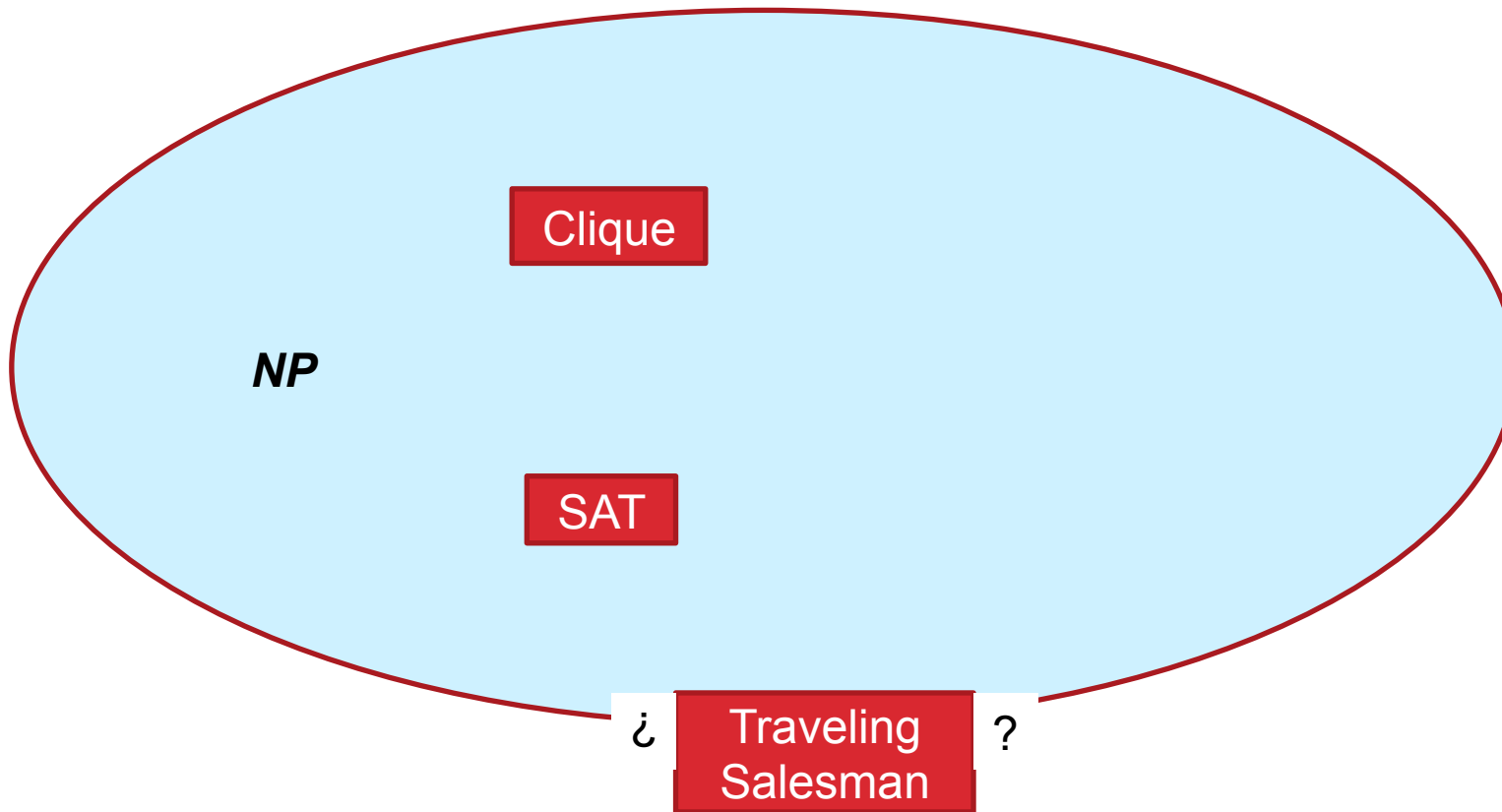
› Traveling Salesman Problem:

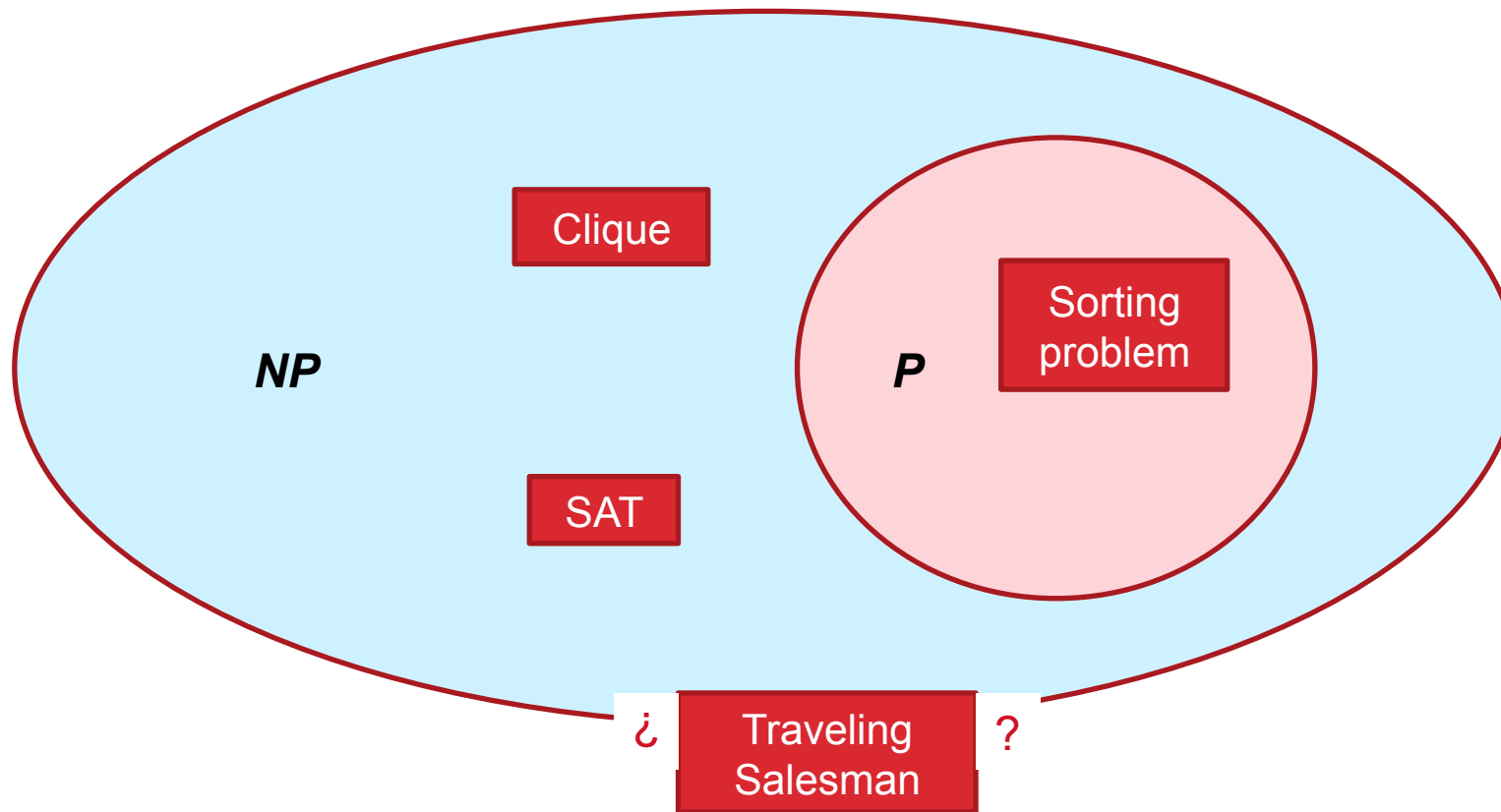
- Hmmmm ... it may be hard even to check whether a given order of cities is optimal?



Fundamental Notion #2: NP

NP is the set of problems for which we can efficiently check to see whether a given answer is correct.







Fundamental Fact #5

Some problems are harder than others

Some problems are harder than others

- › The Clique problem is at least as difficult as SAT.

Theorem

If there were a good algorithm to solve the clique problem, then there would be a good algorithm to solve SAT.

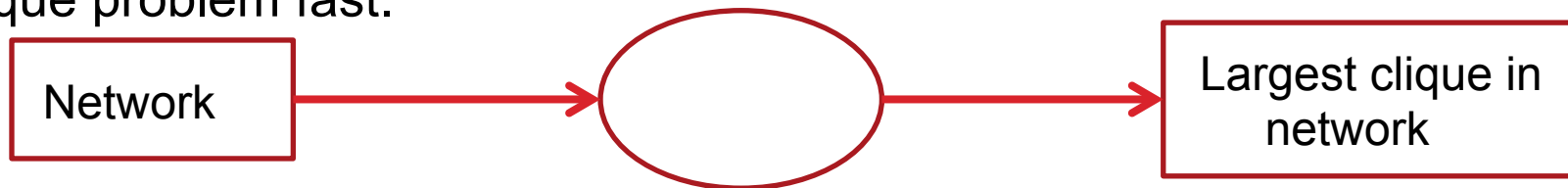
Proof →

Theorem

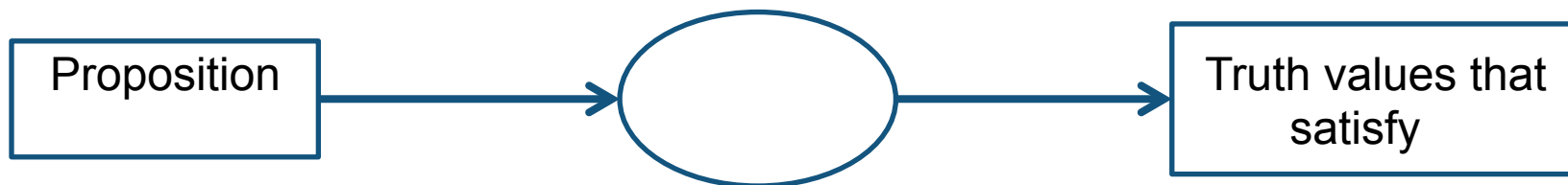
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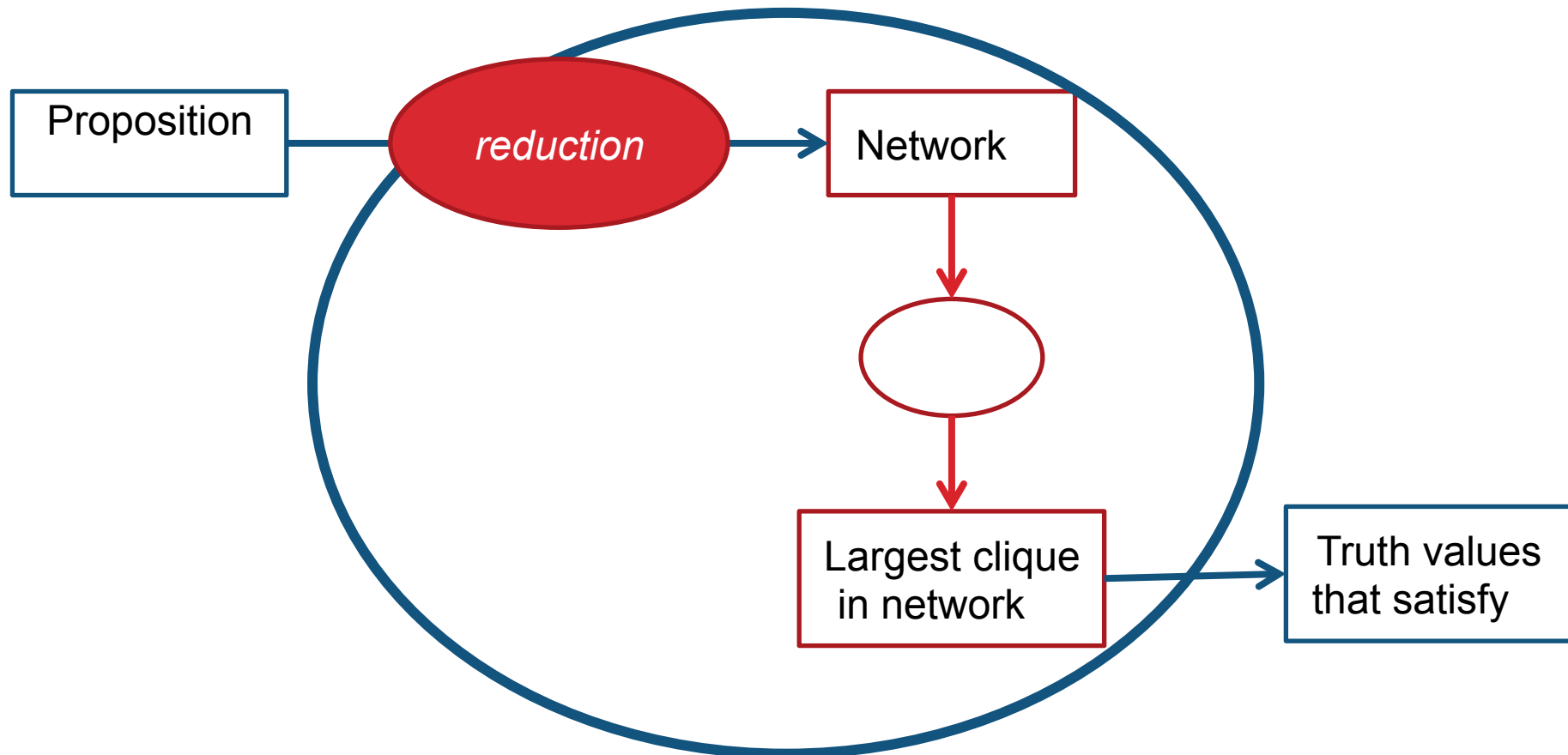
Proof:

Suppose that we have a good fast algorithm *Aclique* that solves the clique problem fast.



Using *Aclique*, let's make an algorithm *ASAT* to solve SAT.







Either $p \downarrow 3$, or $not\ p \downarrow 1$, or $p \downarrow 4$,
 and
 Either $p \downarrow 5$, or $not\ p \downarrow 2$, or $not\ p \downarrow 1$, and
 Either $p \downarrow 6$, or $p \downarrow 2$, or $not\ p \downarrow 3$,
 and
 Either $not\ p \downarrow 3$, or $p \downarrow 4$, or $p \downarrow 6$,
 and
 Either $p \downarrow 5$, or $p \downarrow 2$, or $not\ p \downarrow 3$,
 and



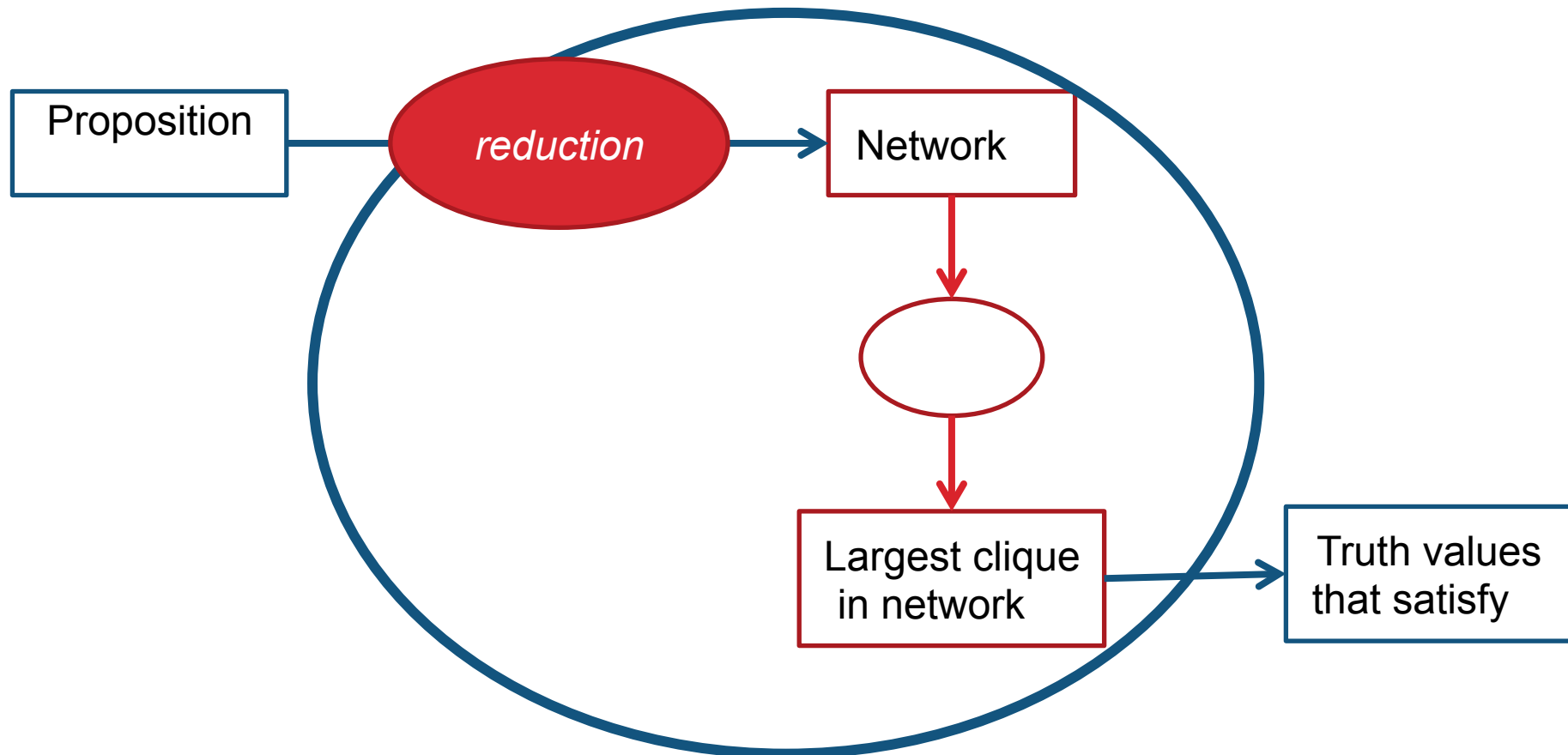
Nodes:

- > For each i , there is a node for variable p_i and a node for the negation $not\ p_i$ of p_i .

Connections:

- > Connect two nodes if they do not occur in the same clause, and one is not the negation of the other.

Either $not\ p \downarrow 1$, or $not\ p \downarrow 4$, or



Theorem

If there were a good algorithm to solve the clique problem, then there would be a good algorithm to solve SAT.

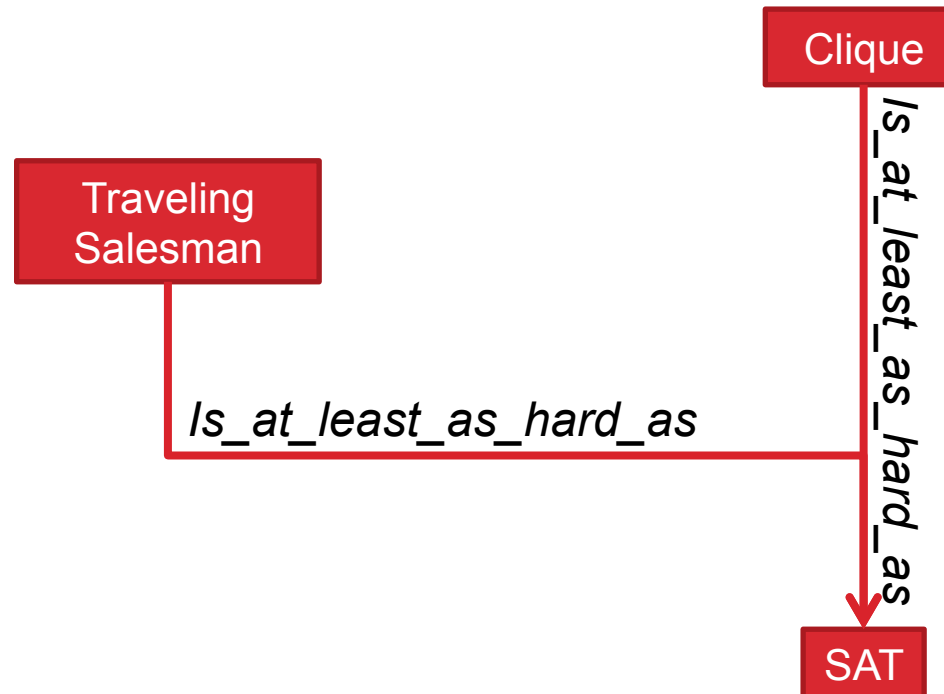
Proof:

If we knew a good fast algorithm *Aclique* for the clique problem, then we could use it to make an algorithm *ASAT* to solve SAT.

- › That is, if the clique problem were easy, then SAT would be easy
- › That is, the clique problem is at least as difficult as SAT

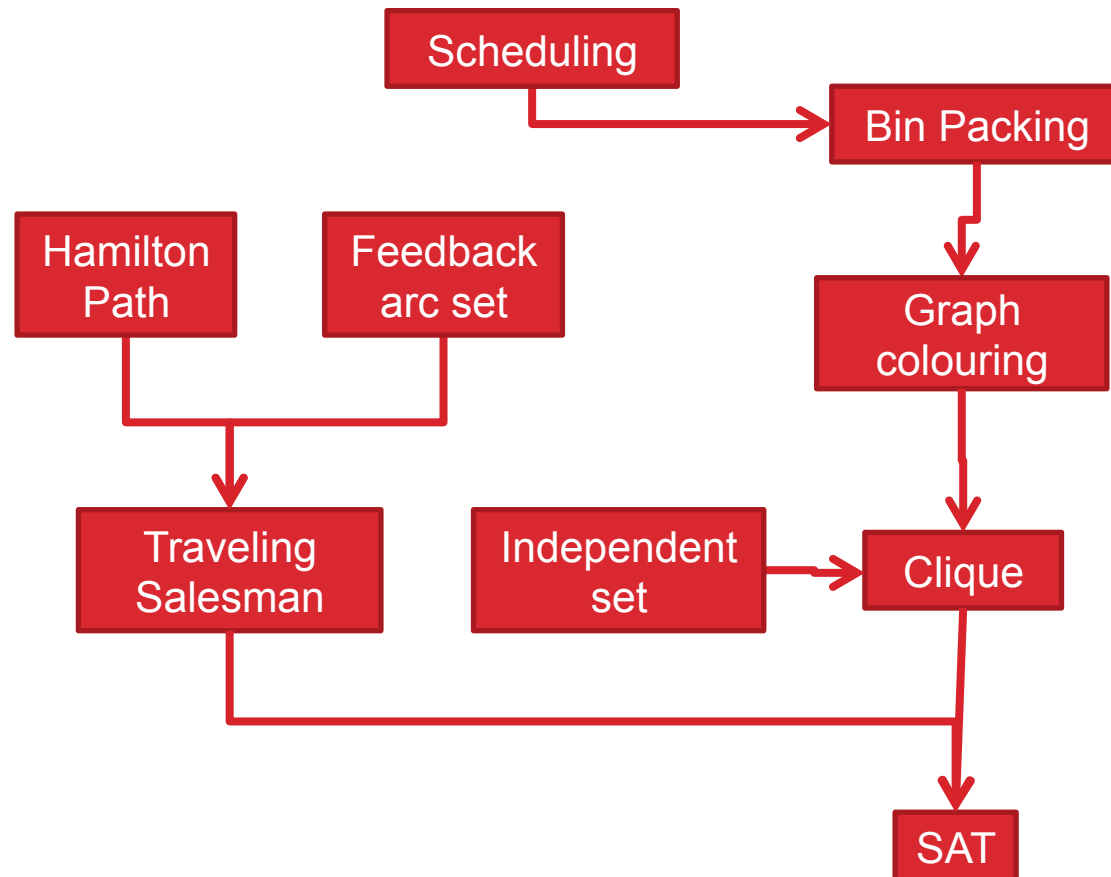
Some problems are harder than others

- › Also, the travelling salesman problem is at least as hard as SAT.



In fact, the relationship “*at_least_as_hard_as*” between problems has been very well investigated.

For many problems, there is a proof that this problem is at least as hard as SAT.



In fact, the relationship “*at_least_as_hard_as*” between problems has been very well investigated.

For many problems, there is a proof that this problem is harder than SAT.

Many problems are *at_least_as_hard_as* SAT.

- *Feedback Arc Set Problem*: solutions needed to detect deadlocks in communications systems
- *Scheduling problems*: solutions need in logistics industry, and in computer systems
- *Bin Packing problems*: solutions needed in manufacturing: steel industry, clothing industry
- *Many layout problems*: solutions needed for network visualization, newspaper layout, integrated circuit layout
- *Number theory problems*: solutions needed for cryptography

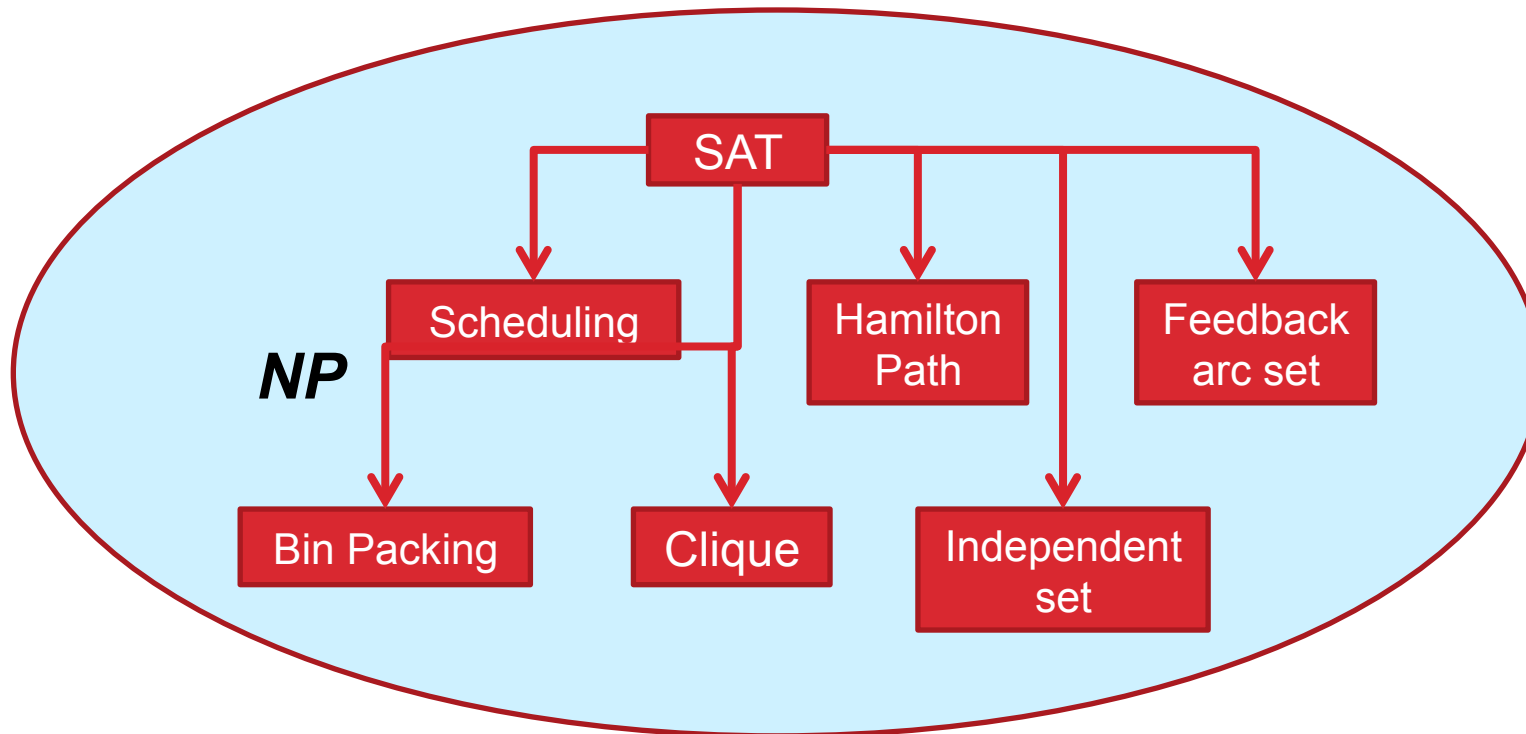


Fundamental Fact #6

SAT is at least as hard as every other problem in NP

(Cook's Theorem)

SAT is *at_least_as_hard_as* every problem in NP





Cook's Theorem

SAT is "NP-complete"



Fundamental Notion #3: NP-complete

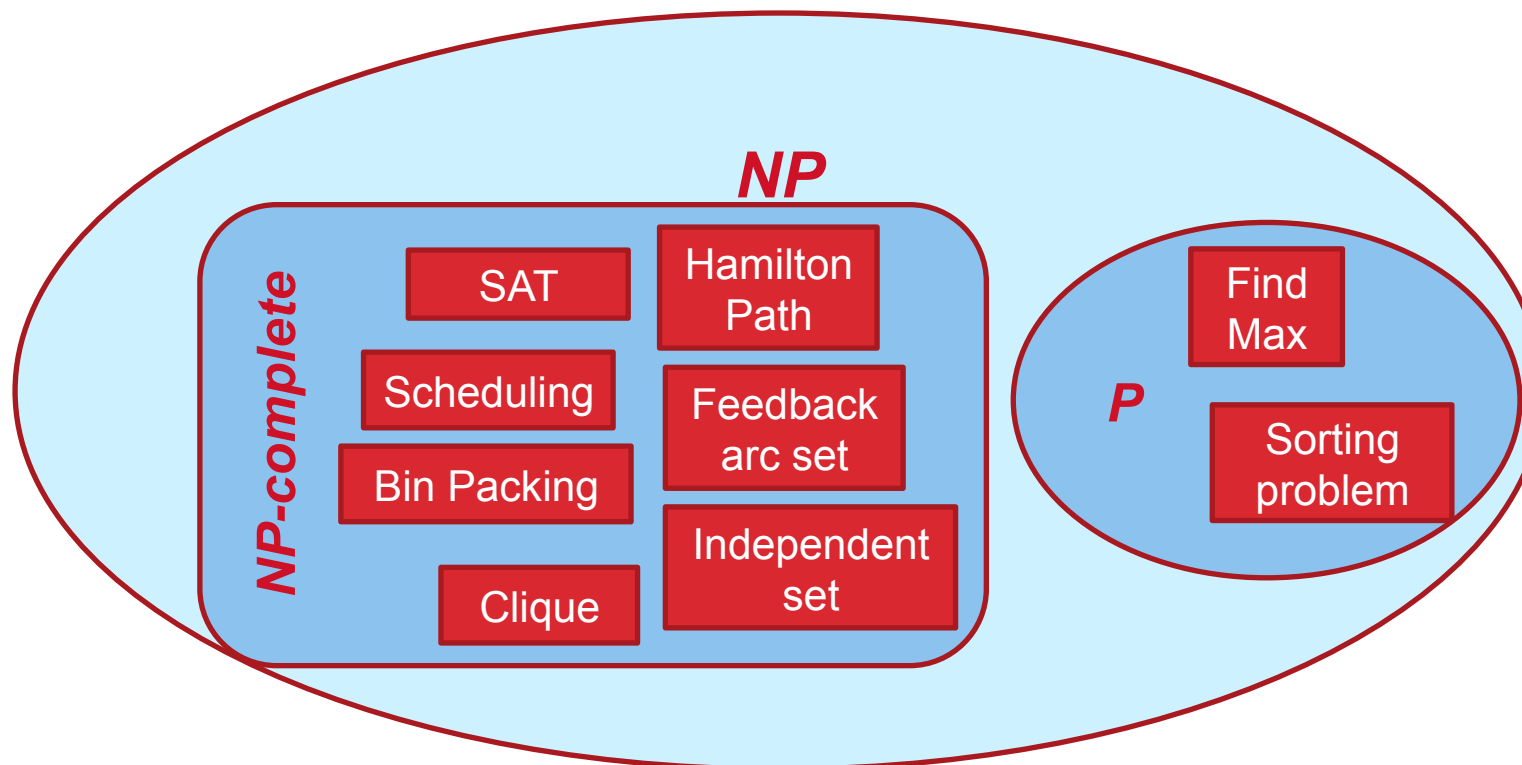
Fundamental Notion #3

A problem is ***NP-complete*** if it is

- In NP, and
- At least as difficult as every other problem in NP

A problem is *NP-complete* if it is

- In NP, and
- At least as difficult as every other problem in NP





Fundamental Fact #7

Many real-world problems are NP-complete.

Many real-world problems are ***NP-complete***.

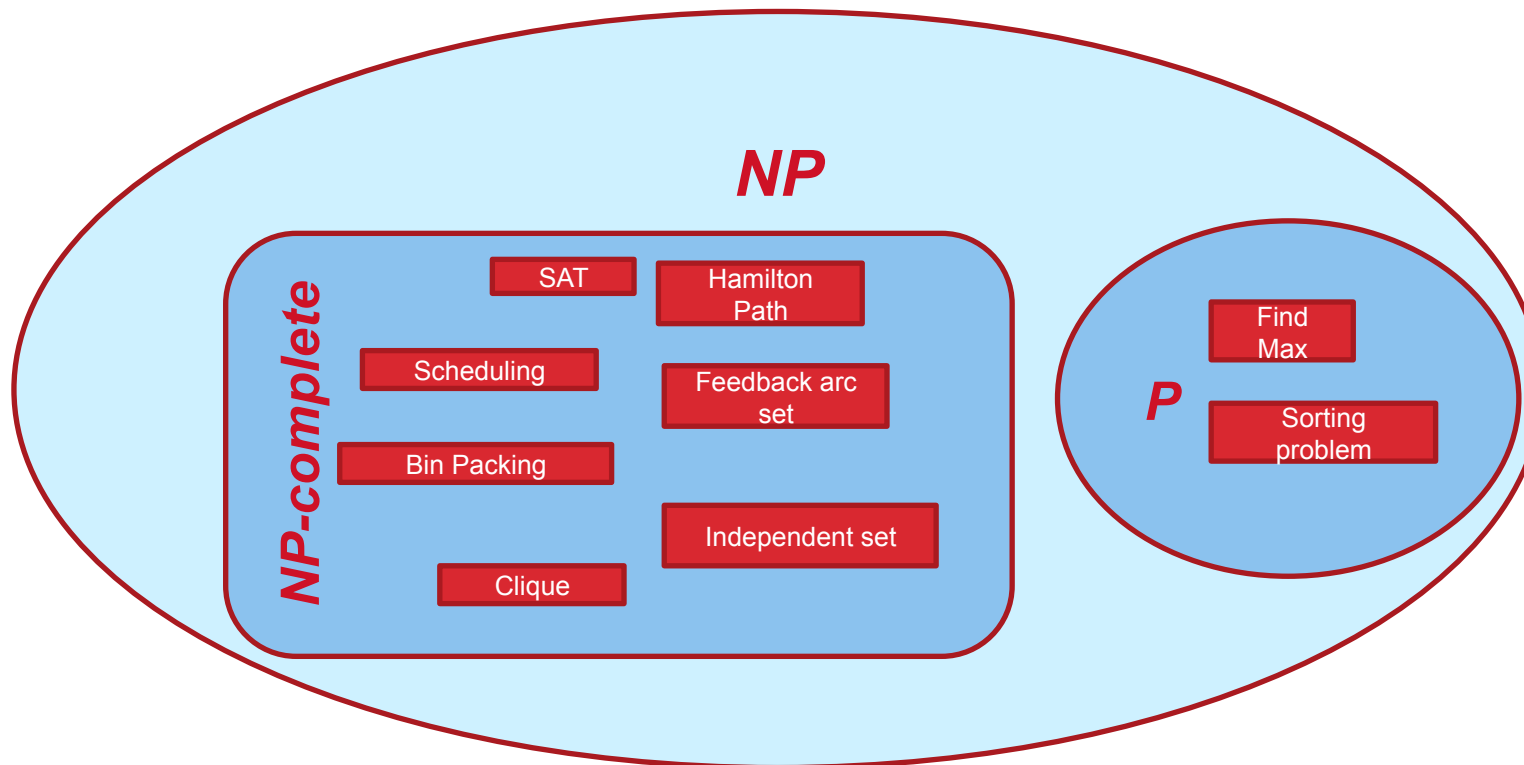
- *Feedback Arc Set Problem*: solutions needed to detect deadlocks in communications systems
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- *Number theory problems*: solutions needed for cryptography



Fundamental Question #1

Does P equal NP ?

$P=NP$???



$P=NP$???

› If $P=NP$ then

- All the real-world problems in NP have polynomial-time algorithms, and can be feasibly solved.

› If $P \neq NP$ then

- We must be satisfied with algorithms that do not work entirely.

$P=NP$???

- › To prove $P=NP$
 - You need to show that one NP-complete problem has a polynomial-time solution.
- › To prove $P \neq NP$
 - You must show that one NP-complete problem does not have a polynomial-time solution



Fundamental Question #1

$P=NP$???

- › This is the most fundamental issue in Computer Science
- › The problem is still unsolved



Fundamental Question #1

$P=NP$???

- › The investigation of this question and others like it is called “complexity theory”

The Fundamentals of P, NP, and Complexity

Fundamental Fact #1: ***Exponential functions are eventually bigger than polynomial functions***

Fundamental Fact #2: ***Some algorithms are efficient, some are not***

Rule of Thumb #1: ***An algorithm that runs in exponential time is not feasible; an algorithm that runs in polynomial time may be feasible.***

Fundamental Notion #1: ***P is the set of all problems that can be solved in polynomial time***

Fundamental Fact #3: ***Some problems can be solved with efficient algorithms, and some others ... maybe not***

Fundamental Fact #4: ***Sometimes we can efficiently check whether an answer is correct, even if we can't efficiently find a correct answer***

Fundamental Notion #2: ***NP is the set of problems for which we can efficiently check to see whether a given answer is correct.***

Fundamental Fact #5: ***Some problems are harder than others***

Fundamental Fact #6: ***SAT is at least as hard as every other problem in NP***

Fundamental Notion #3: ***A problem is NP-complete if it is in NP, and at least as difficult as every other problem in NP***

Fundamental Fact #7: ***Many real-world problems are NP-complete***

Fundamental Question #1: ***Does P equal NP?***