#### CS 3100, Models of Computation, Spring 20, L26

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URL: bit.ly/3100s20Syllabus



# NP-Completeness proofs

### Definition of P-time and NP-time (from book)

P-time: An algo is P-time if its computational tree is bounded in height by a polynomial function
of the length of its input for every input in the language that the DTM decides. For this simple
"101" DTM, here is that DTM's code and here is a computational tree - with paths shown. The
paths are two for rejecting runs and one for an accepting run.

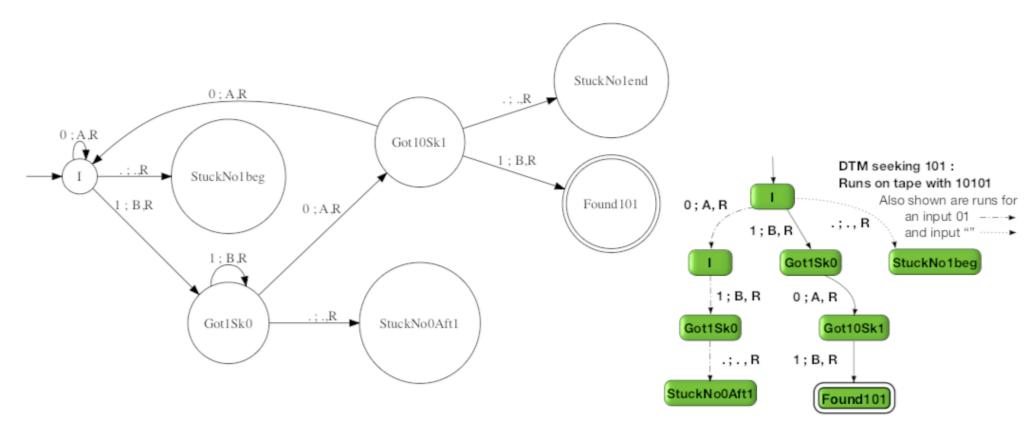
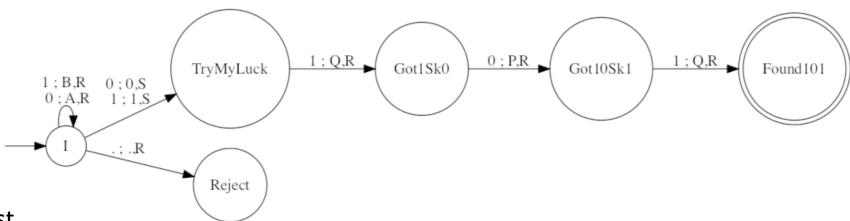


Figure 16.2: Transition diagram and computation tree for a DTM that looks for 101 within given w.

#### Illustration of NP-time (from book)



Look for the deepest path which accepts. That corresponds to The NP-time.

In Jove, the Fuel Models this depth Faithfully even for NDTMs...

(modulo bug-fixes if any...)

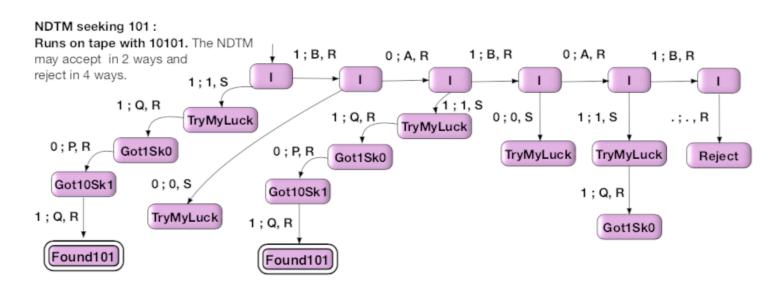
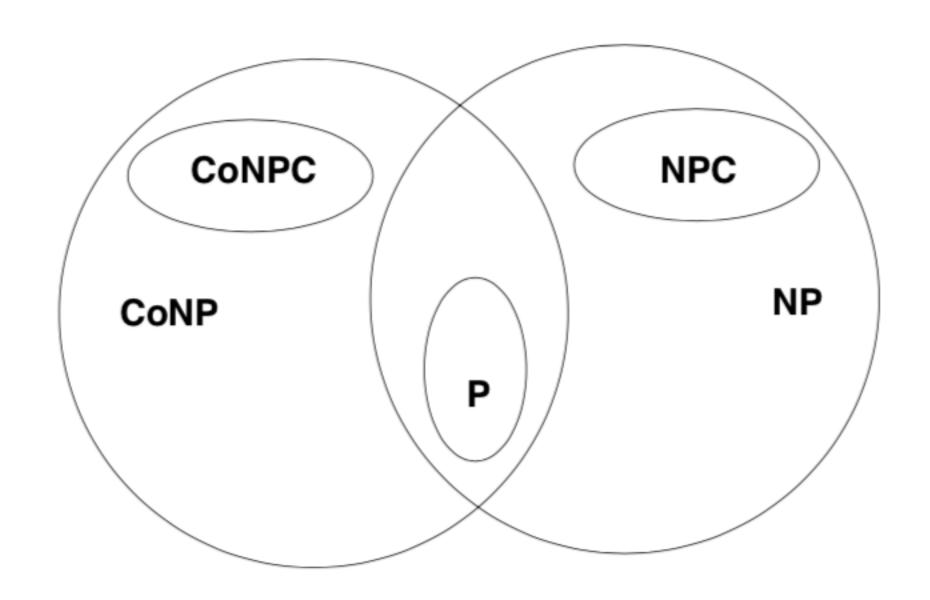


Figure 16.3: Transition diagram and computation tree for an NDTM that looks for 101 within given w.

# Definition of NP-time (from book)

• NP-time: An algo is NP-time if an NDTM can be obtained where it can guess a solution nondeterministically but be able to CHECK that solution in P-time. So the depth of the worst-case (deepest) path in that NDTM must be P-bounded for any input. Some problems may not even qualify for the "check phase" being P-bounded... but many useful problems have !! That makes the theory of NPC interesting and relevant in practice!

#### Language hierarchy in NP-land (ignore "Co" for now)



### Mapping reductions are key to "connect-up"

- NPC
  - A language L is NPC
    - If L is in NP
      - It has a P-time NDTM
    - EVERY language in NP has a P-time mapping-reduction to L
    - This is hard to do in practice, so we take the practical approach below.
- In order to show that a NEW language L is NPC in practice
  - We will end up producing a mapping reduction from one of the problems in NPC to L
  - Then we have a mapping reduction from EVERY language in NP to L
- · Study this "funnel diagram" (Ch-16) to be convinced

# The "funnel diagram"

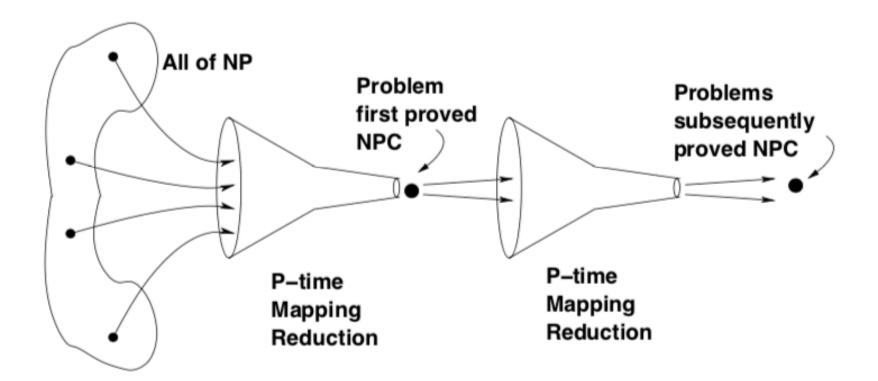


Figure 16.7: Diagram illustrating how NPC proofs are accomplished. The problem first proved NPC is 3-SAT. Definition 16.4(a) is illustrated by the "left funnel" while Definition 16.4(b) is illustrated by the "right funnel." (The funnels serve as a gentle reminder that mapping reductions need not be onto.)

# The "funnel diagram"

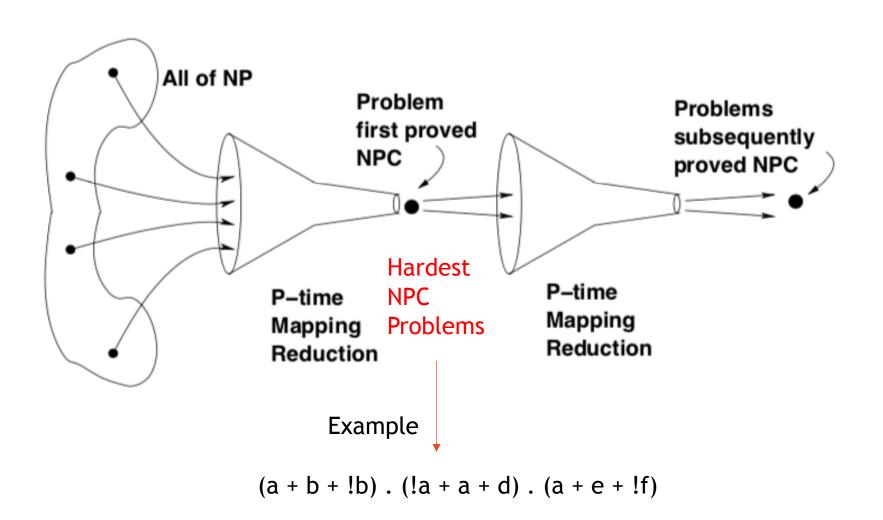
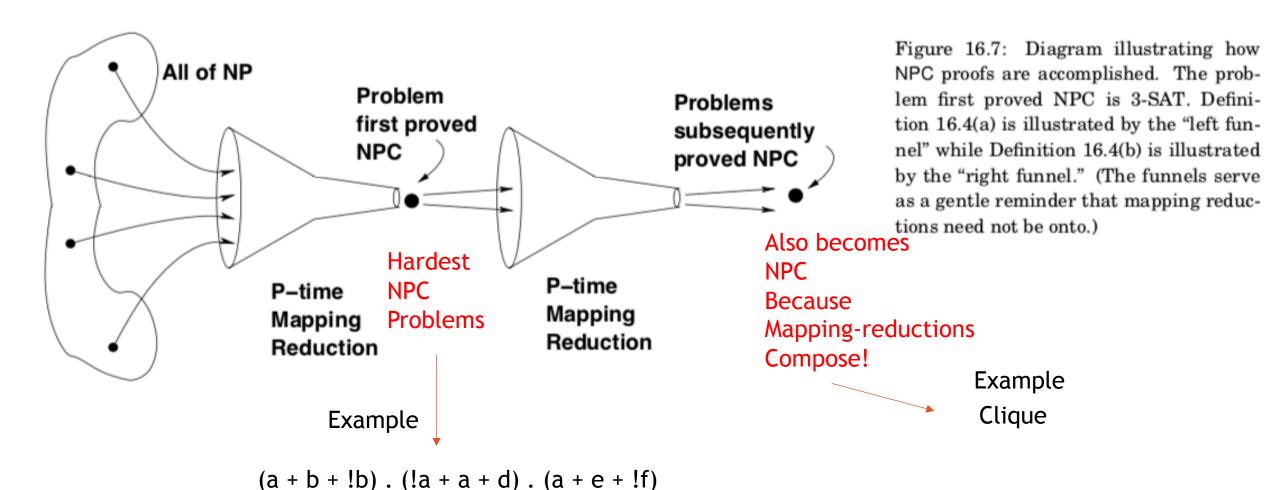


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# The "funnel diagram"

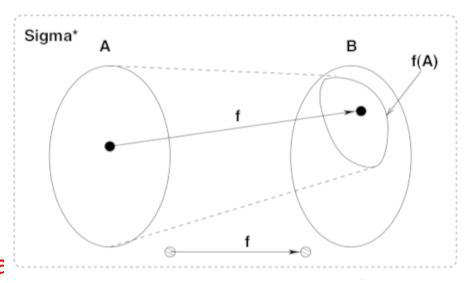


# The basic idea of Mapping Reduction NPC proofs

- A way to take advantage of an already proven hard-fought theorem ("A" below) in establishing the truth of new conjectures ("B" below)
  - We do a proof via diagonalization that 3-SAT is NP-Complete (the "A" mentioned)
  - We can then handle 1000s of other (often more relevant) questions easily (the "B"s)
    - Prove that "A" <= "B" i</li>
    - i.e. A mapping-reduces-to B via a Polynomial-time computable function f (i.e. you can code-up f in any of your favorite prog. languages in a way that f NEVER loops on any A-input and runs in P-time. Easy to do you map Boolean formulae to Graphs, etc as we will see!)
    - i.e. Given a P-time solver for B, we can take ANY A-instance, map it via "f" into a B-instance and feed it to the claimed P-time B-solver (P-time decider)
- Given that an MR means
- "x in A iff f(x) in B", i.e.
  - $x \text{ in } A \Rightarrow f(x) \text{ in } B$
  - x not in A => f(x) not in B

We can conclude that

"solving B in P-time means Solving A in P-time --> A Turing Awa



### The language K-Clique

- Given an undirected graph, is there a set of K nodes such that they are all pairwise connected?
  - They form a K-Clique?

K-Clique - { < G > : G is a graph that has a K-Clique in it }

### We are about to show 3-SAT <=p K-Clique

 <=p is a mapping reduction but with a polynomial bound on runtime

That is, we can translate a 3-SAT instance to a K-Clique instance in polynomial time

• This means that if K-Clique has a Polynomial Algorithm, then 3-SAT will also have a Polynomial Algorithm

# 3SAT <=p K-Clique (the "Translate" fn.)

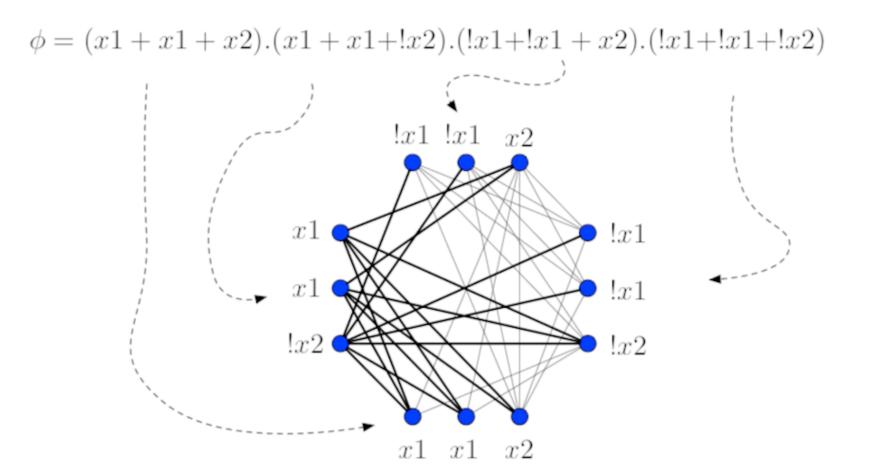


Figure 16.9: The Proof that Clique is NPH using an example formula  $\varphi = (x1+x1+x2).(x1+x1+!x2).(!x1+!x1+x2).(!x1+!x1+x2).(!x1+!x1+1x2)$ . We never connect the nodes within each clause "island" (there are four such islands, each with three nodes). Across each clause island, we draw edges in all possible ways *provided* we never connect a literal and its complement. For visual clarity, we show through dark edges all the edges emanating from the clause island for (x1+x1+!x2) going to all other clause islands. We also show the remaining edges, but using fainter lines.

#### Your Asg-7's problems

- They help you practice these notions using the Binary Decision Diagram tool (BDD tool)
  - Part of Jove
- Binary Decision Diagrams will be explained now
  - They also are minimal DFA "in disguise"
- You'll also gain exposure to Boolean Satisfiability (SAT tools)
- BDD and SAT are industrial tools (used for verifying chips and SW to be bug-free wrt deep properties you can state)
  - I gave you a SAT demo on 4/13/20. You can solve Asg-7 qns using that.
- SAT is fun too (for solving Sudoku etc)
- <a href="https://medium.com/@rvprasad/sat-encoding-solving-simpler-sudoku-d92671206d1e">https://medium.com/@rvprasad/sat-encoding-solving-simpler-sudoku-d92671206d1e</a>

# Binary Decision Diagrams

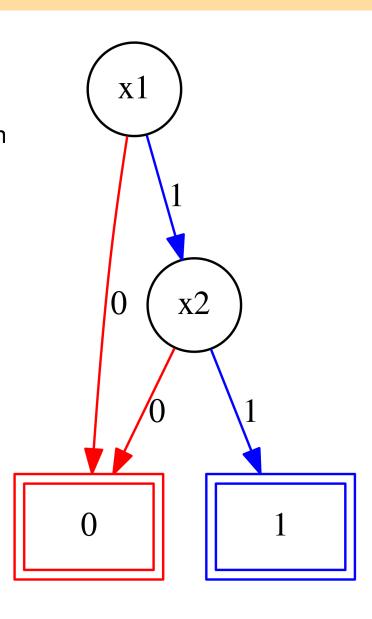
#### **BDDs**

- They are a data structure for representing Boolean Functions
- Included in Jove (see First\_Jove\_Tutorial/CH17/CH17.ipynb)

- All the details of BDDs is not that important
- But practice with our BDD tool will fix ideas in your mind better (seeing Boolean formulae as graphs is often edifying)

#### Use of BDD tool

```
# http://formal.cs.utah.edu:8080/pbl/BDD.php
# Very simple example to show-off syntax
#First declare the variables and specify variable orderin
Var_Order: x1 x2 x3
#Then define formula
fmla = (x1 | x2) & (x1 | !x2) & (!x1 | x2)
Main_Exp: fmla
# Type "build BDD"
# Right-click and save PNG
# SAT, UNSAT, Valid - from shape of BDD
# SAT: paths exist to "1"
# UNSAT: [0] BDD
# Valid: [1] BDD
```



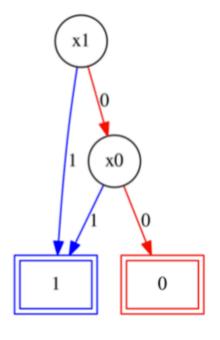
#### Things to observe about BDDs

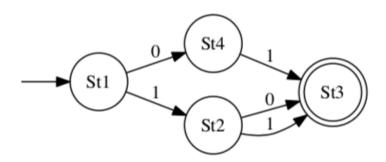
- They are DAGs with one "layer" per variable
- They "decode" the formula in a top-to-bottom order

- The order of decoding determines the BDD size
- There are often good heuristics to select this order

# BDDs are minimal DFA in disguise

 $L_0R = "(01+10+11)" \# Regexp for the on-set of the OR function 
 dot0bj_dfa(min_dfa(nfa2dfa(re2nfa(L_0R))), STATENAME_MAXSIZE=4)$ 

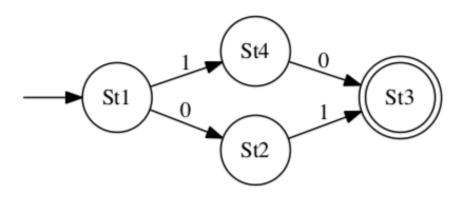




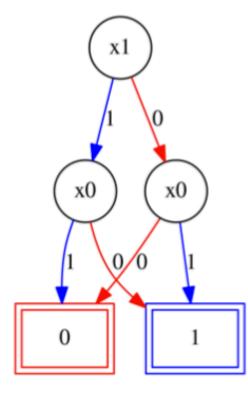
By entering these commands and clicking "build BDD," one obtains the Or BDD of Figure 17.5.

#### BDDs are minimal DFA in disguise

 $L_XOR = "(01+10)" \# The \ regexp \ for \ the \ on-set \ of \ the \ XOR \ function \\ dot0bj_dfa(min_dfa(nfa2dfa(re2nfa(L_XOR))), STATENAME_MAXSIZE=4)$ 



We can see that the minimal DFA<sup>3</sup> for Xor accepts 01 and 10. The BDD for Xor can be obtained using an online tool called PBDD<sup>4</sup> that can be invoked as follows (it will open the BDD tool in a new browser tab):



### Good vs Bad BDD var order (poly vs exp size)

study how to build a BDD for x2x1x0 < y2y1y0

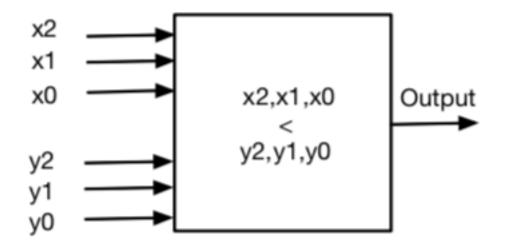


Figure 17.6: The < comparator.

# Good vs Bad BDD var order (poly vs exp size)

study how to build a BDD for x2x1x0 < y2y1y0: All satisfying Boolean combinations as a regexp  $\rightarrow$  give it to Jove!!

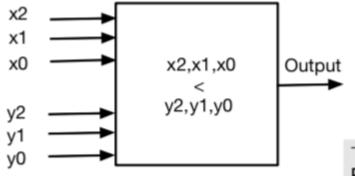


Figure 17.6: The < comparator.

```
R = "(000001+000011+000111+001011+001111+010011+010111+011111+\
100101+100111+101111+110111+000010+000101+000110+001010+001101+\
001110+010101+010110+011101+011110+100110+101110+000100+001100+\
010100+011100)"
```

R express a "bad order" (not bad odor)  $\rightarrow$  we fed in order of x2 x1 x0 THEN y2 y1 y0

A quick decision is facilitated by x2 y2 x1 y1 x0 y0 (coming later)

#### Bad BDD var --> blowup

study how to build a BDD for x2x1x0 < y2y1y0: All satisfying Boolean combinations as a regexp  $\rightarrow$  give it to Jove!!

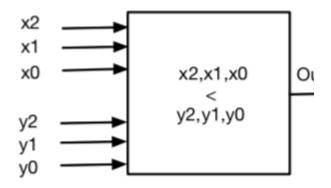


Figure 17.6: The < comparator.

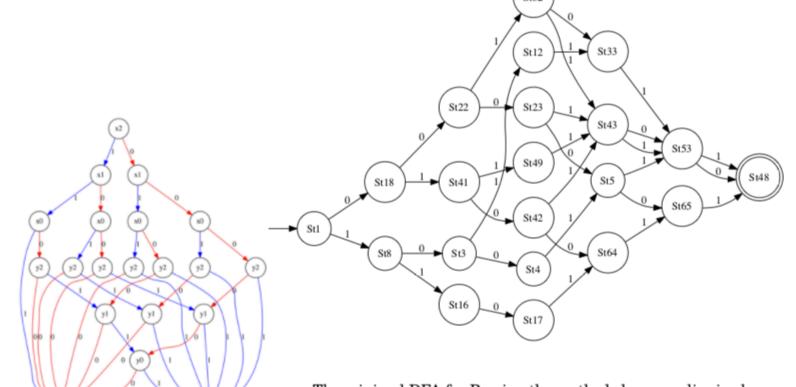


Figure 17.7: BDD for the magnitude comparator: bad input-variable order

The minimal DFA for R using the method shown earlier is above, and the BDD for it is in Figure 17.7 (for Var\_Order being x2 x1 x0 y2 y1 y0). This DFA is in fact exponential in the x2,x1,x0 bits (those are the first three bits to arrive at this machine, and the machine grows exponentially with respect to those inputs). It must represent every x2 x1 x0

### Good BDD var --> no blowup

study how to build a BDD for x2x1x0 < y2y1y0: All satisfying Boolean combinations as a regexp  $\rightarrow$  give it to Jove!!

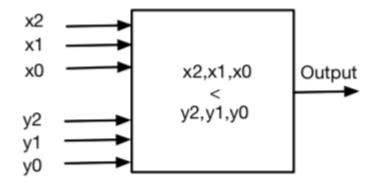
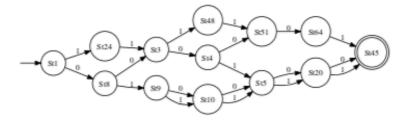


Figure 17.6: The < comparator.

Let us call the regular expression obtained by interleaving the input bits "Rmix":

The minimal DFA for Rmix is below, and the BDD for it is in Figure 17.8 (for Var\_Order being x2 y2 x1 y1 x0 y0). Again, it is easy to see that the BDD does not "trudge through" the redundant decodings. For instance, in the DFA, state St8 is reached when  $x_2 = 0$ , and then when  $y_2 = 1$  is seen, a pathway of redundant decodings leading to the accept state is entered. Correspondingly, in the BDD, after seeing  $x_2 = 0$ , we reach a node which decodes  $y_2$ , and if  $y_2 = 1$ , the BDD jumps to the "1" leaf node.



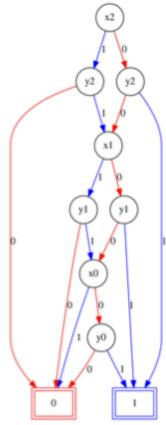


Figure 17.8: BDD for the magnitude comparator: good input-variable order

# Good BDD var --> no blowup for multi-input xor

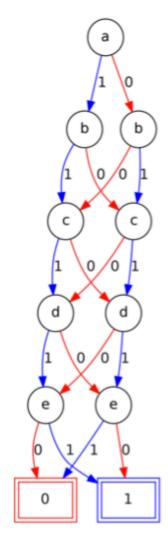


Figure 17.9: BDD for a 5-input Xor gate

#### Practice Problems: Sat? BDD for var order x1,x2,x3?

$$(x_1 + x_2 + x_3) \cdot (x_1 + !x_2 + !x_3)$$

#### Sat? Sat instance with BDD var order x1,x2,x3?

$$(x_1 + x_2 + x_3) \cdot (!x_1 + !x_2 + x_3) \cdot (!x_1 + !x_2 + !x_3) \cdot (!x_1 + x_2 + !x_3)$$

#### Sat? Sat instance with BDD var order x1,x2,x3?

$$(x_1 + x_2 + x_3) \cdot (!x_1 + !x_2 + x_3) \cdot (!x_1 + !x_2 + !x_3) \cdot (!x_1 + x_2 + !x_3)$$

This is the online interface for a BDD package written using PBL.

About this BDD implementation | Language Spec

#### **ENTER YOUR FORMULA HERE**

Var\_Order: x1 x2 x3

Main\_Exp: (x1 | x2 | x3) & (|x1 | |x2 | |x3) & (|x1 | |x2 | |x3) & (|x1 | |x2 | |x3)

#### PROGRAM OUTPUT

Number of satisfying assignments: 5

Number of Variables: 3

Number of Nodes : 6

Variable Ordering

['x1', 'x2', 'x3']

All satisfying assignts:

[0, 0, 1]

[0, 1, 1]

[1, 1, 0]

[0, 1, 0]

[1, 0, 0]

#### Sat? Sat instance with BDD var order x1,x2,x3?

$$(x_1 + x_2 + x_3) \cdot (!x_1 + !x_2 + x_3) \cdot (!x_1 + !x_2 + !x_3) \cdot (!x_1 + x_2 + !x_3)$$

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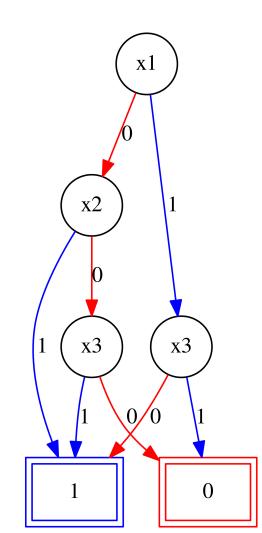
[0, 0, 1]

[0, 1, 1]

[1, 1, 0]

[0, 1, 0]

[1, 0, 0]





Build BDD

**Build minimum BDD** 

#### Show relative hardness of problems via <=p

People observed that many real-world problems are expensive to solve in the worst case

E.g. Is there a clique in a large graph

The internet is a huge graph

Update of internet node software, reliable direct links between cities, ... are all hard problems

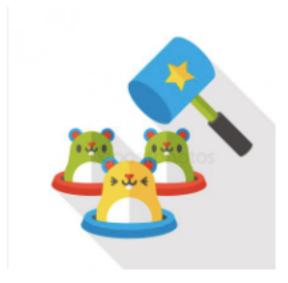
They are all formally connected via <=p (poly-time mapping reductions)



#### NPC problems are easy to check; difficult to solve

Both for 3SAT  $(a + b + !b) \cdot (!a + a + d) \cdot (a + e + !f)$ 

And clique



### Definition of P-time and NP-time (from book)

- P-time: Illustrated wrt 3SAT
  - Obtain computation tree of DTM for the language 3SAT (all members in it)
  - Can we claim anything about the depth of the computational tree for all inputs?
    - As far as we know, any DTM working on 3SAT appears to incur an exp depth for at least some of the instances
- NP-time: there is an NDTM that can guess the solution for a 3SAT problem (in P-time) and also check this guess in P-time
  - Question: will we ever get a DTM that does this in P-time??
  - This is what the question of P =?= NP really means

#### Smart idea

- Define the idea of the hardest problems in NP
- Call it NPC
- I.e. A language L is NPC if
  - L is in NP --- has a NP-time algo (guess check on NDTP is P-time)
  - For every problem L' in NP, we have L' <=p L</li>
    - That is, L is harder than anything there is in NP
  - Any problem such as L is "NP-hard" i.e. harder than anything in NP
- So, NPC = NP-hard + in NP
- Finding such an NPC language was the open question that Cook and Levin solved

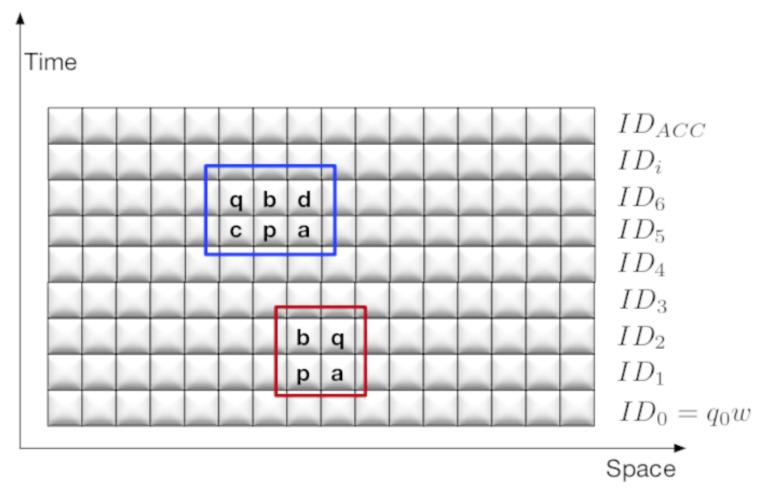
### First NPC problem

• 3SAT was shown NPC as follows

- First show that 3SAT has an NP algorithm (easy; build an NDTM)
- Then show that EVERY NP problem has a P-time M.R. to 3-SAT
- This was achieved by imagining the solution of any NP problem as a collection of "tape histories"
- Then encoding these histories using 3CNF formulae!

#### How 3SAT was shown NP-hard (from book)

This is the history of how a TM chugs along. Fortunately, each move from ID\_k (full tape) to ID\_k+1 (full tape) can be modeled as changes that occur within a 2x2 or 2x3 window. These changes can be captured using a 3CNF formula. .... The rest is history! (see the book for more)



#### Boolean Satisfiability: First NPC problem

- From any NDTM, we can compile a 3-SAT formula
- If the NDTM is NP-time, then we can decide the truth of the generated 3-SAT formula in NP-Time

• If 3-SAT is deterministic P-time, then we can decide any NDTM in P-time (not in NP-time)

Details given in the book

#### Aside: DNF does not capture the complexity of NPC properly!

$$(x_1 + x_2 + x_3) \cdot (x_1 + !x_2 + !x_3)$$

Given this or any other CNF with N variables, an NDTM can be built such that

- \* its first N moves are to write out a variable assignment on the tape
- \* then check that under that assignment, the formula is true

But hey, DNF is linear-time SAT-checkable. Multiply the above out to get a DNF

$$x1 + x1.!x2 + x1.!x3 + x2.x1 + x2.!x2 + x2.!x3 + x3.x1 + x3.!x2 + x3.!x3$$
 $\Rightarrow$  then simplify

SAT if ANY product-term is ..... ?(what)

This is a linear check
But expansion to DNF turns the formula EXP LONG !!! So no real advantage.

In a sense, DNF is spelling out every possible solution and we have to check one by one !!!

### How does it relate to primality testing?

- General result
  - Researchers believe that it is highly unlikely that a language is NPC and the complement of that language is in NP
  - I.e. NPC and Co-NP IFF NP = Co-NP (highly likely)
  - So if a problem is in NP and the complement of the problem is in NP also
    - i.e. L in NP and N in Co-NP
  - Then it is unlikely that L is NPC
- This allows people to predict that some algos are "easy"
- This happened in 1977 and 2002
  - Primes are in NP (1977)
  - Primes are in P (2002) because Primes were in NP and Co-Primes were also in NP; so unlikely that Primes were NPC. This was established in 2002
- Note this is only primality testing not Prime Factorization which is still of unknown hardness!