

# Reductions

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# Reduction

- So far we have shown:
  - There are problems that are not tractable algorithmically
    - Halting Problem
  - We defined a class  $\mathcal{P}$  of problems that are considered computationally tractable, even as the instance size scales up
  - We have defined a class  $\mathcal{NP}$  of problems where we can verify a solution effectively

# Reduction

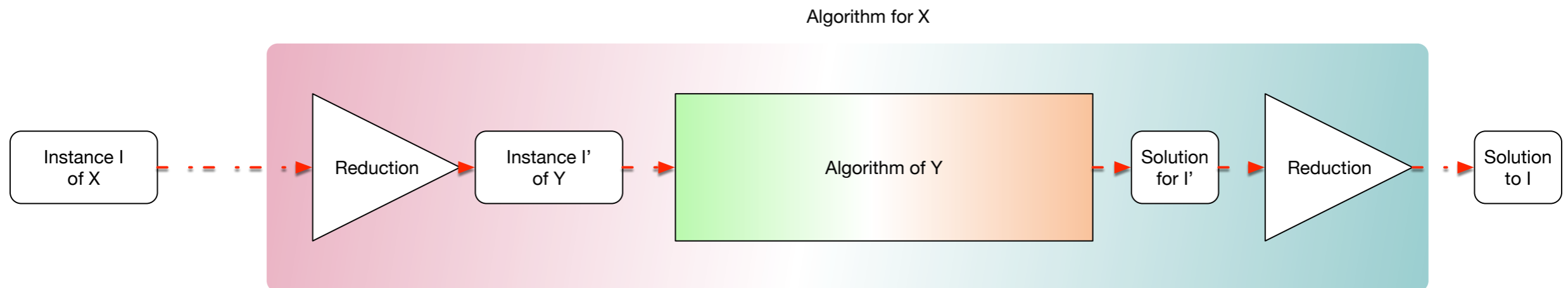
- Fundamental Conjecture in Complexity
  - $\mathcal{P} \neq \mathcal{NP}$
- Reason why people believe in this conjecture
  - There are problems that are  $\mathcal{NP}$ -complete
- A problem  $P \in \mathcal{NP}$  is  $\mathcal{NP}$ -complete if
  - $(P \in \mathcal{P}) \Rightarrow (\mathcal{P} = \mathcal{NP})$

# Reduction

- Can use the solution of one problem to solve another problem
- Example: Matrix multiplication and Matrix Squaring
  - If you can solve matrix multiplication, you can certainly solve matrix squaring
  - However, it also works the other way around:
    - Multiply square matrix  $A$  with square matrix  $B$
    - Calculate left side of  $\begin{pmatrix} 0 & A \\ B & 0 \end{pmatrix}^2 = \begin{pmatrix} AB & 0 \\ 0 & BA \end{pmatrix}$
  - From the algorithmic standpoint: squaring a matrix is exactly as complicated as multiplying two matrices

# Reduction

- Formally, problem  $X$  reduces to problem  $Y$  if there is a reduction that converts instances of  $X$  to instances of  $Y$  and a translation that takes a solution of  $Y$  and makes it into a solution for  $X$  that solves the original instance.



# Reduction

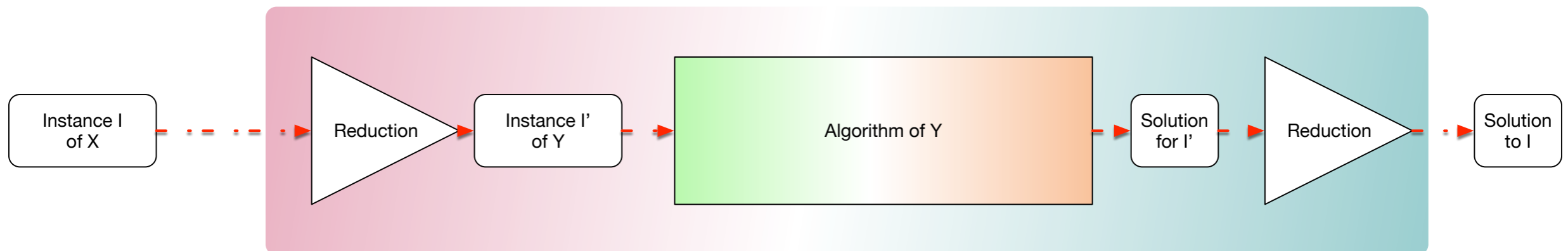
- Matrix multiplication reduces to matrix squaring

- Reduce: Given  $A, B$ , construct  $\begin{pmatrix} 0 & A \\ B & 0 \end{pmatrix}$

- calculate the square

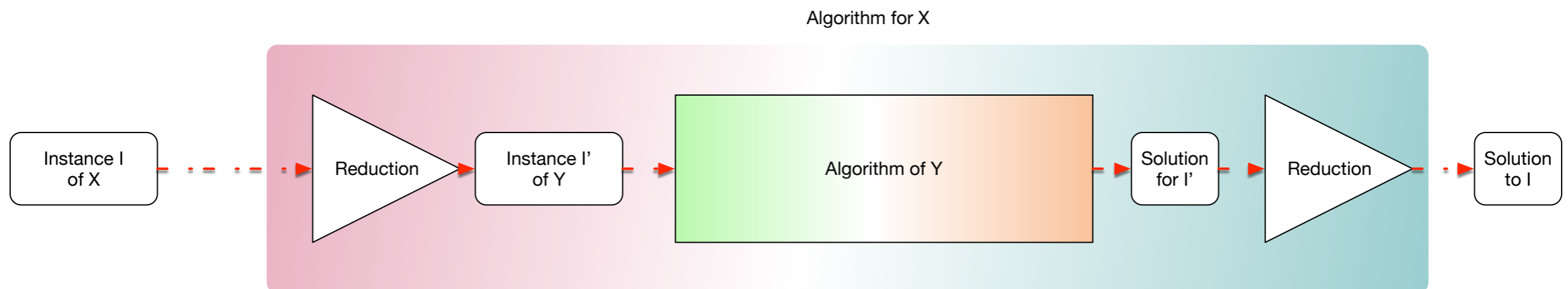
- translate result  $\begin{pmatrix} AB & 0 \\ 0 & BA \end{pmatrix} \longrightarrow AB$

Algorithm for X



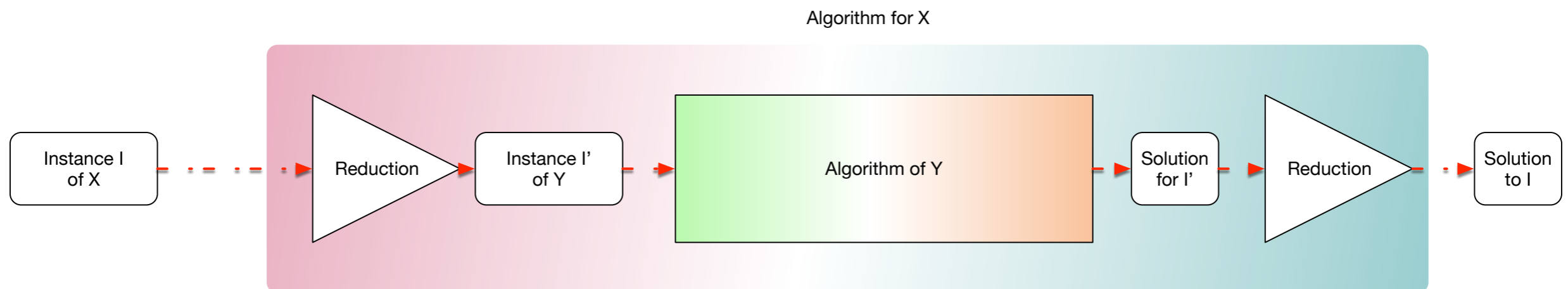
# Reduction

- We usually apply reduction to existence problems
  - Answer is True/False
  - No translation is needed



# Reduction

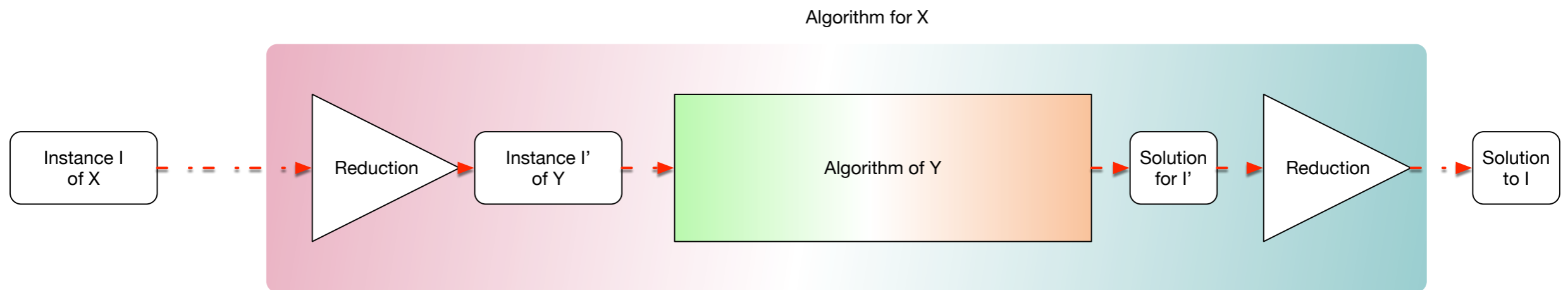
- $X$  reduces *polynomially* to  $Y$ 
  - Reduction works in polynomial time
- Write  $X \leq_p Y$ 
  - Read(  $Y$  is at least as hard as  $X$ )





# Reduction

- Assume  $X \leq_p Y$  and  $Y \in \mathcal{P}$ . Then  $X \in \mathcal{P}$ .
- Proof: if we have a polynomial reduction, then the diagram below explains how we get a polynomial time algorithm for  $X$



# A New Formal Definition of

## $\mathcal{NP}$ - complete

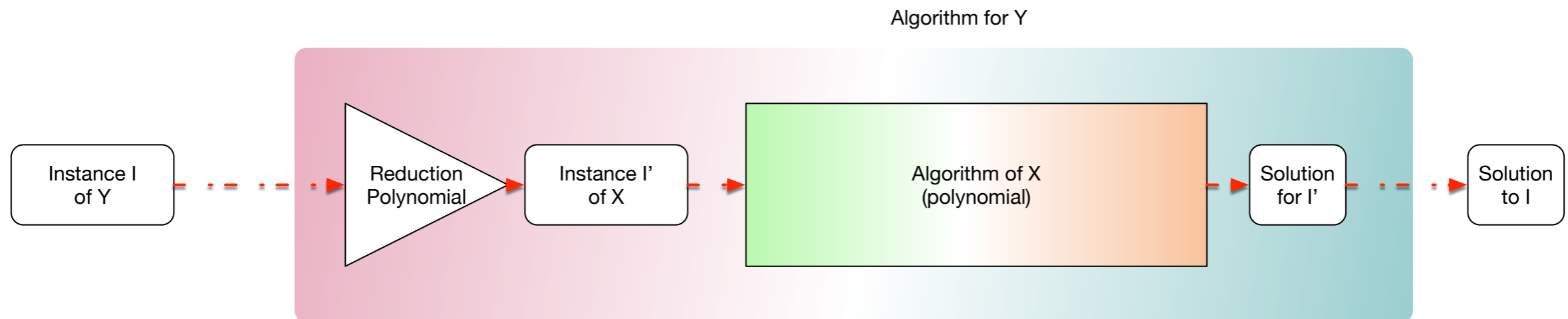
- $X \in \mathcal{NP}$  is  $\mathcal{NP}$ -complete
  - IF AND ONLY IF
- $\forall Y \in \mathcal{NP} : Y \leq_p X$

# A New Formal Definition of $\mathcal{NP}$ - complete

- Theorem:
  - $X \in \mathcal{NP}$ -complete AND  $X \in \mathcal{P}$ 
    - IMPLIES
  - $\mathcal{P} = \mathcal{NP}$

# A New Formal Definition of $\mathcal{NP}$ - complete

- Proof:
  - If  $Y \in \mathcal{NP}$ , then by completeness
  - $Y \leq_p X$
  - Thus:



- There is a polynomial time algorithm for Y
- Thus:  $Y \in \mathcal{P}$

# A New Formal Definition of

# $\mathcal{NP}$ - complete

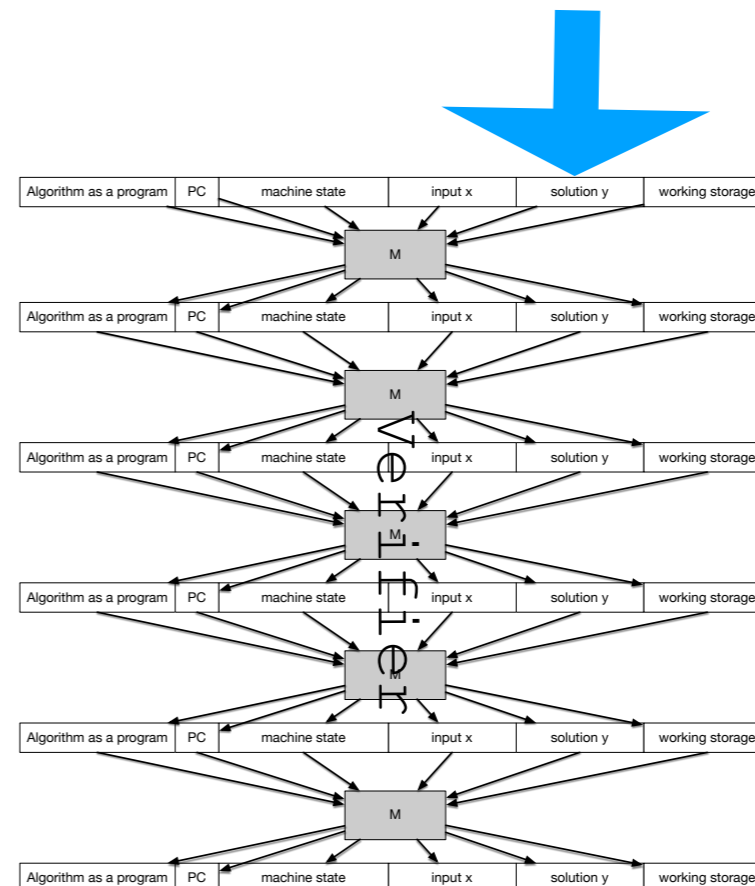
- Circuit Satisfiability is  $\mathcal{NP}$  - complete
  - We have argued that we can express any problem in  $\mathcal{NP}$  using a circuit

- This reduces the problem to Circuit Satisfiability

- Therefore:

- Circuit Satisfiability remains  $\mathcal{NP}$  - complete

Gussed Solution



# Family of $\mathcal{NP}$ -complete problems

- Boolean Formula Satisfiability:
  - Given a boolean formula, determine whether there is an assignment to the variables such that the formula becomes true

$$(x \Rightarrow \bar{y}) \vee (x \wedge y \vee z) \vee (x \wedge \bar{y} \wedge \bar{z} \wedge (x \oplus y))$$

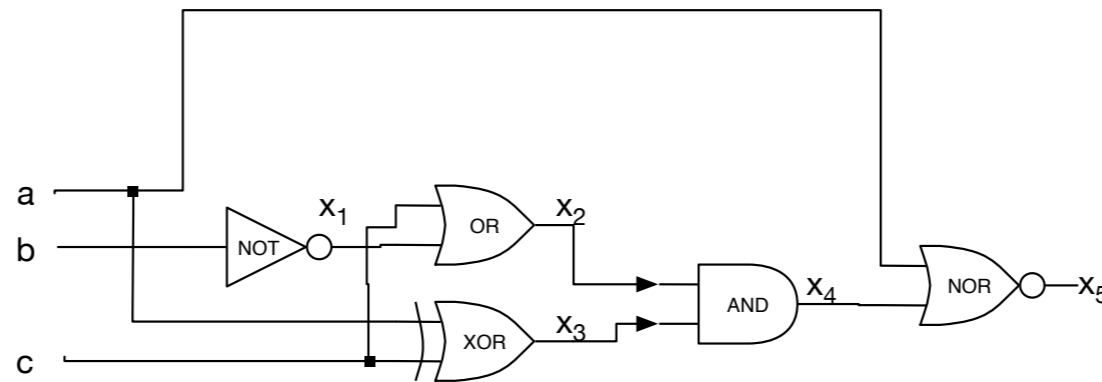
- Given an assignment, can check the truth value of the formula in polynomial time
- Thus, Boolean Formula Satisfiability is in  $\mathcal{NP}$

# Family of $\mathcal{NP}$ -complete problems

- We can reduce boolean formula satisfiability to boolean circuit satisfiability
  - Need to show that boolean circuit satisfiability is at least as hard as boolean formula satisfiability
  - Given an instance of boolean circuit satisfiability, show that it can be reduced to boolean formula satisfiability

# Family of $\mathcal{NP}$ -complete problems

- Need to "translate" a boolean circuit into a formula



- Each circuit element becomes part of a Boolean formula
- $x_1 = \bar{b}$ ,  $x_2 = c \vee x_1$ ,  $x_3 = a \oplus c$ ,  $x_4 = x_2 \wedge x_3$   
 $x_5 = \neg(x_4 \vee a)$
- Make this into a single formula
- $x_1 = (\neg b) \wedge (x_2 = c \vee x_1) \wedge \dots$



# Family of $\mathcal{NP}$ -complete problems

- Final formula is satisfiable exactly if the circuit is satisfiable
- Translation is in polynomial time

# Family of $\mathcal{NP}$ -complete problems

- 3-SAT : Satisfiability of a boolean formula in conjunctive normal form with clauses with three literals

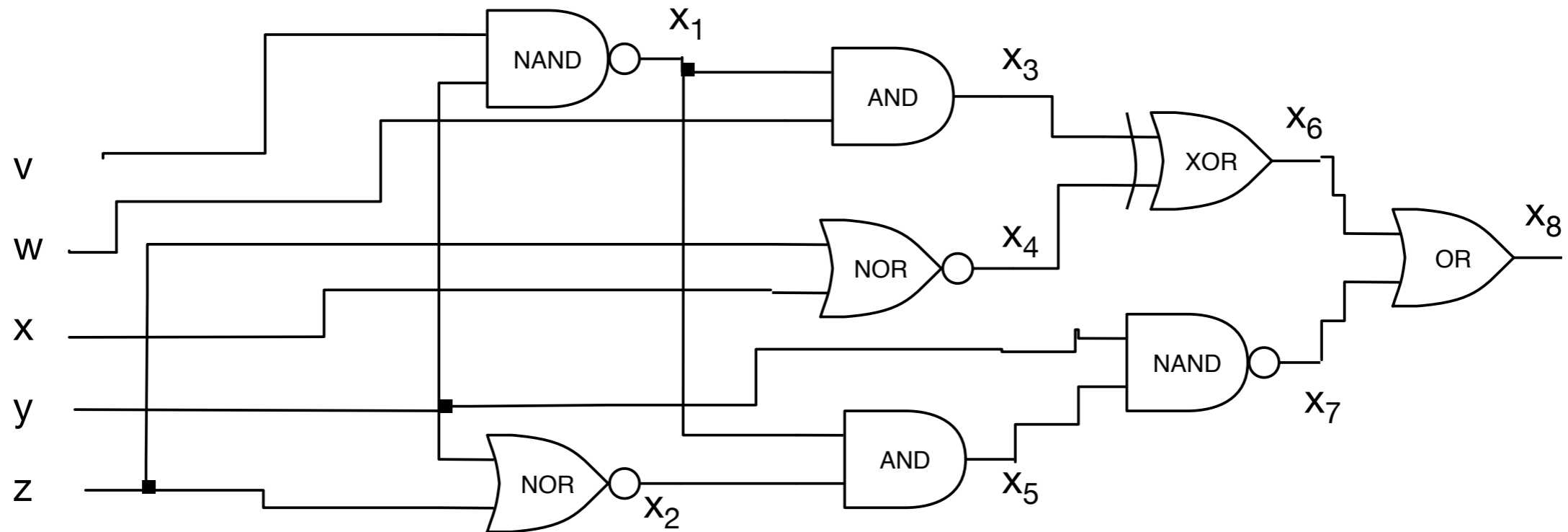
$$(a \vee \bar{b} \vee c) \wedge (a \vee a \vee c) \wedge (b \vee \bar{d} \vee d) \wedge (\bar{a} \vee \bar{d} \vee \bar{f}) \dots$$

# Family of $\mathcal{NP}$ -complete problems

- Reduction to Boolean Formula Satisfiability
  - Need to transform the boolean circuit satisfiability problem to 3-SAT
  - Reduction:  $3\text{-SAT} \leq_P \text{Boolean Circuit Sat}$
  - As before, describe the circuit in a boolean formula
  - However, now each gate can be expressed with only three-clause conditions:

# Family of $\mathcal{NP}$ -complete problems

- Translate the circuit into one with two-entry boolean gates



$$\begin{aligned}
 & (x_1 = \neg(v \wedge y)) \wedge ((x_2 = \neg(z \vee y))) \wedge (x_3 = x_1 \wedge w) \wedge (x_4 = \neg(x \vee z)) \wedge \\
 & (x_5 = x_1 \wedge x_2) \wedge (x_6 = x_3 \oplus x_4) \wedge (x_7 = \neg(y \wedge x_5)) \wedge (x_8 = x_6 \vee x_7) \wedge x_8
 \end{aligned}$$

# Family of $\mathcal{NP}$ -complete problems

- Clauses directly represent gates
- Transform the clauses into conjunctive normal form with three literals
  - Trick
    - $x = (x \vee p \vee q) \wedge (x \vee \neg p \vee q) \wedge (x \vee p \vee \neg q) \wedge (x \vee \neg p \vee \neg q)$
    - $x \vee y = (x \vee y \vee p) \wedge (x \vee y \vee \neg p)$
- This blows up the representation, but only by a constant amount

# Family of $\mathcal{NP}$ -complete problems

- As before, the circuit is satisfiable if and only if the formula is satisfiable
- Ergo: We can solve 3-SAT implies we can solve Boolean Circuit Satisfiability

# Family of $\mathcal{NP}$ -complete problems

- What about 2-SAT
  - Clauses are of the form  $(a \vee b)$ , which is equivalent to  $\neg a \Rightarrow b$
  - Create a graph:
    - Vertices are variables and their negation

$a$

$\bar{b}$

$\bar{c}$

$\bar{d}$

$\bar{a}$

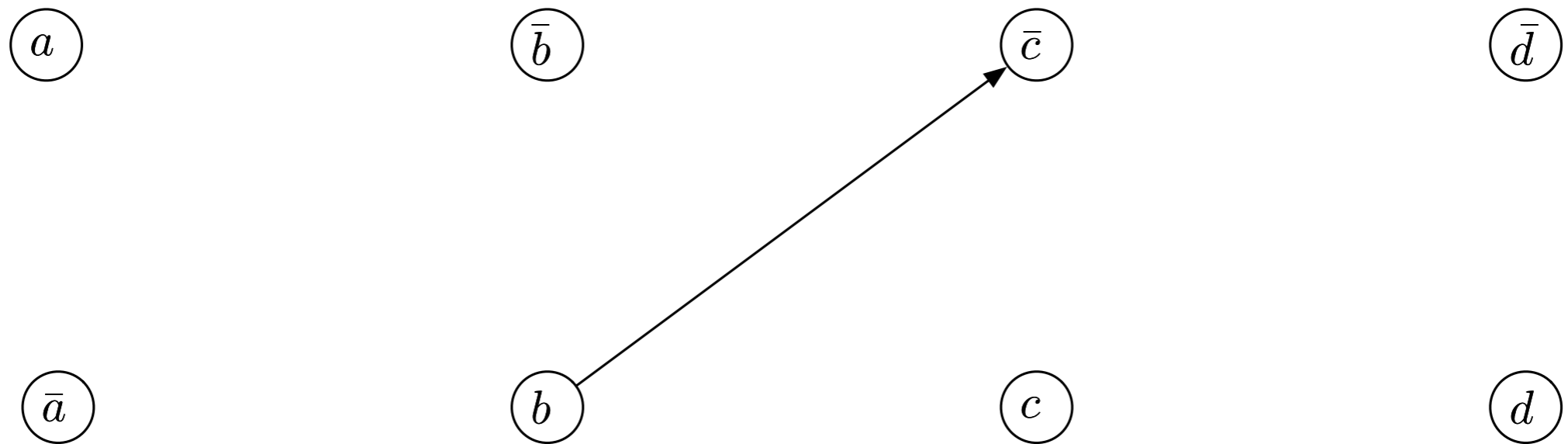
$b$

$c$

$d$

# Family of $\mathcal{NP}$ -complete problems

- Draw an edge if there is a clause equivalent to  $x \Rightarrow y$  in the formula

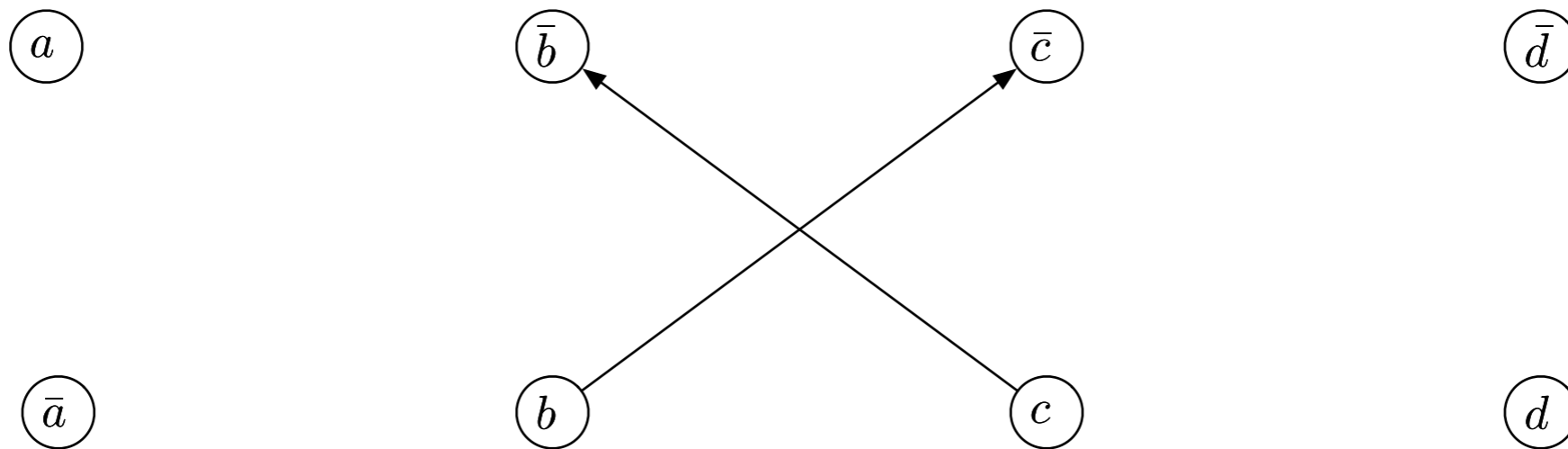


$b \Rightarrow \neg c$  is equivalent to  $(\neg b \vee \neg c)$



# Family of $\mathcal{NP}$ -complete problems

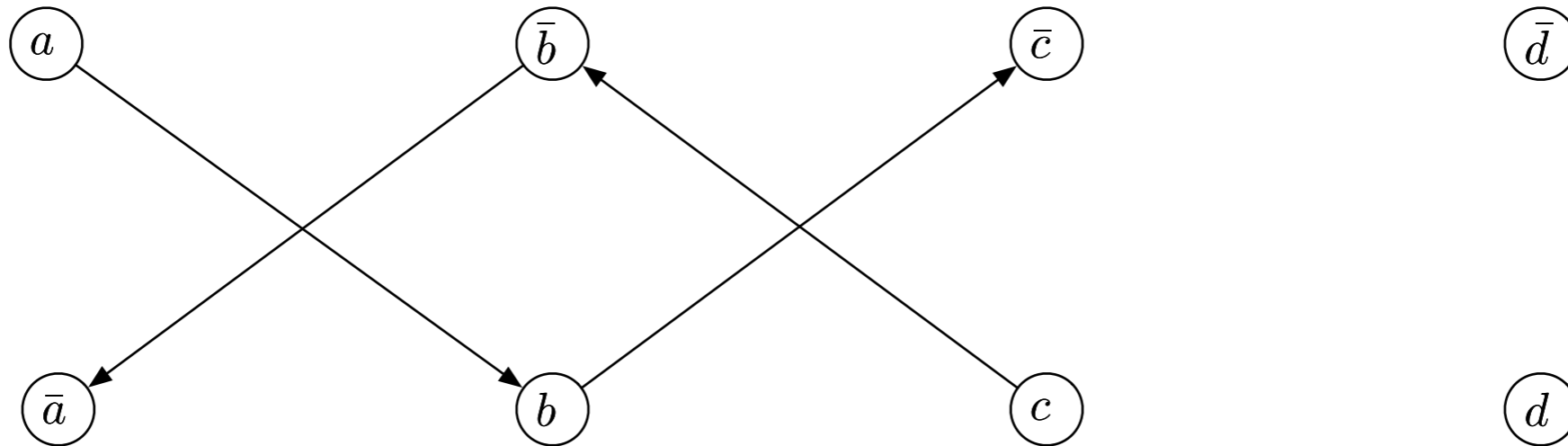
- Draw an edge if there is a clause equivalent to  $x \Rightarrow y$  in the formula



$b \Rightarrow \neg c$  is equivalent to  $(\neg b \vee \neg c)$   
which is equivalent also to  $c \Rightarrow \neg b$

# Family of $\mathcal{NP}$ -complete problems

- Draw an edge if there is a clause equivalent to  $x \Rightarrow y$  in the formula



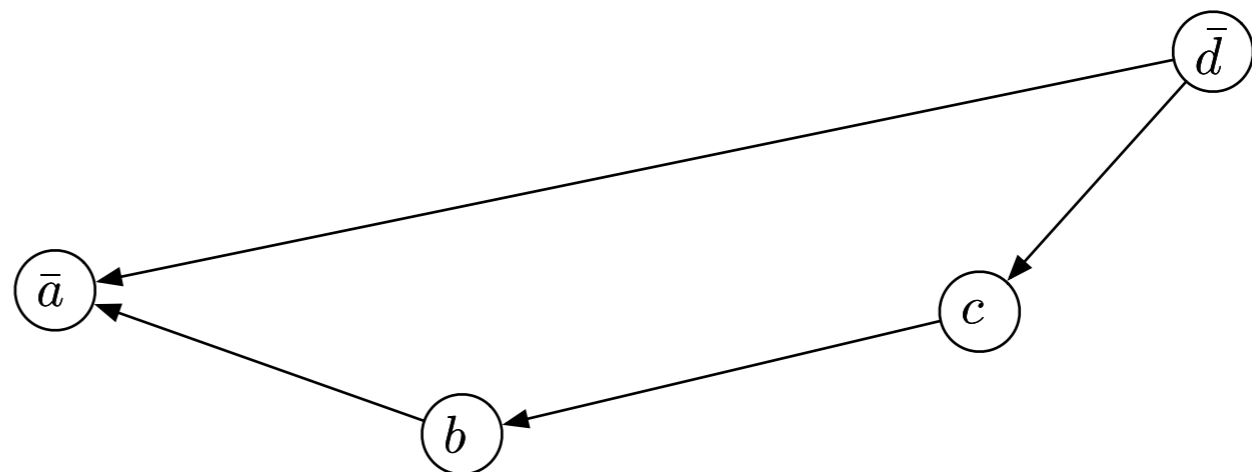
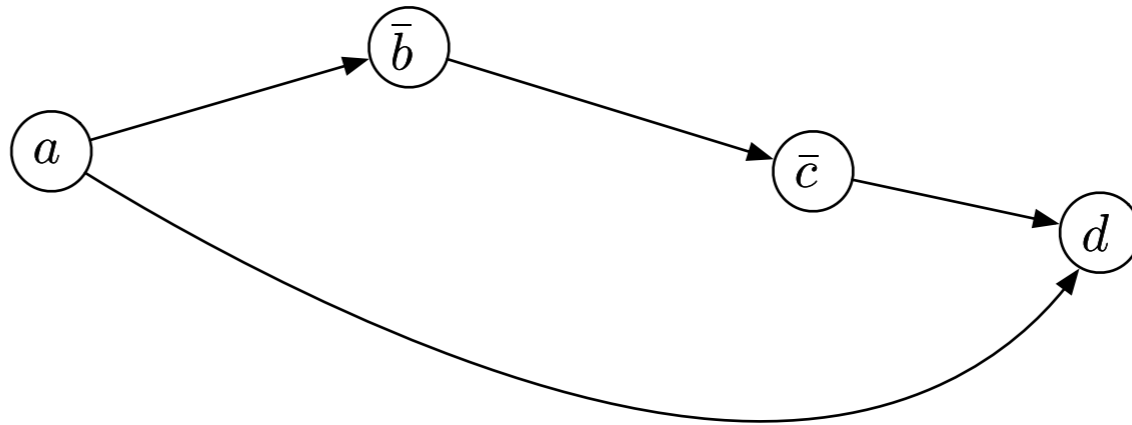
$a \Rightarrow b$  means  $(\neg a \vee b)$   
which also means  $\neg b \Rightarrow \neg a$

# Family of $\mathcal{NP}$ -complete problems

- Assume that  $x$  and  $\neg x$  are in a cycle. Then we can conclude that  $x$  implies  $\neg x$  that implies  $x$ .
- Thus, there is no possibility to find a truth assignment that works
- If there is never such a cycle, then we just start with one variable and assign it true, then we follow the implications et cet.

# Family of $\mathcal{NP}$ -complete problems

- Example:
  - Assign  $b = \text{True}$
  - Then  $a = \text{False}$
  - Then  $c = \text{True}$
  - Then  $d = \text{False}$

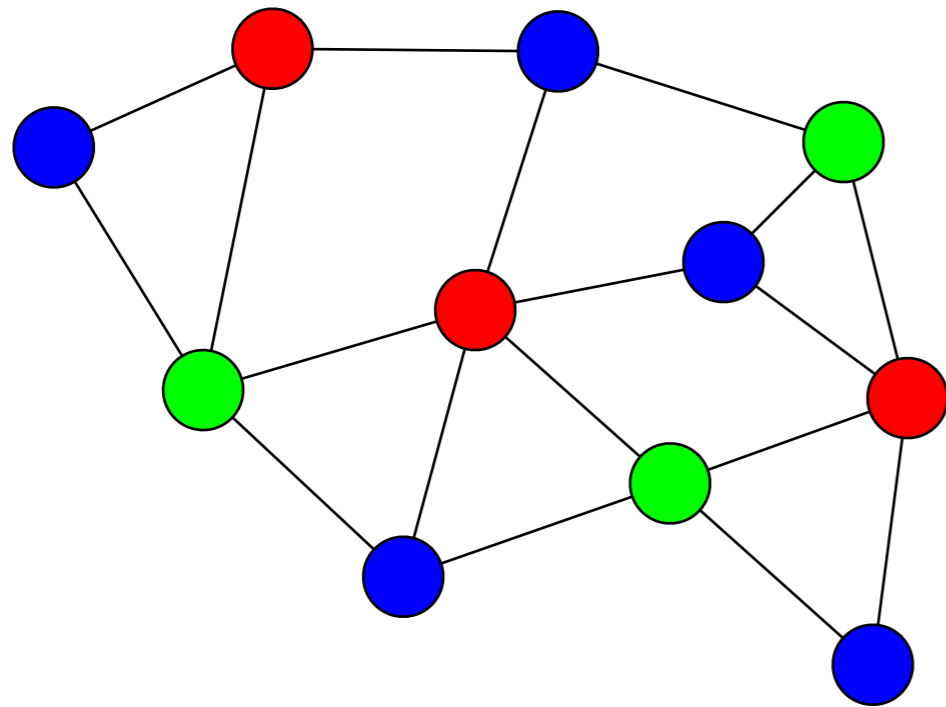


# Family of $\mathcal{NP}$ -complete problems

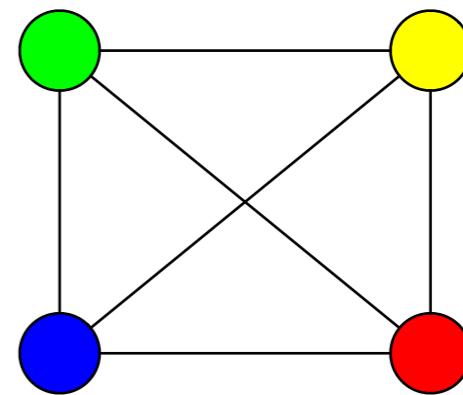
- To calculate this, we calculate the strongly connected components
- If there is a component that contains at most one of a variable and its negation, then assign those literals True
  - By working through all the components, we obtain a satisfiable assignment
- If there is a component that contains both a variable and its negative, then the formula is not satisfiable
- ERGO: we can solve 2-SAT with DFS in polynomial time

# Graph 3-Colorability

- Graph 3-colorability
  - Given a graph, can we color its vertices with three colors such that no edge is between vertices with the same color



3 colorable



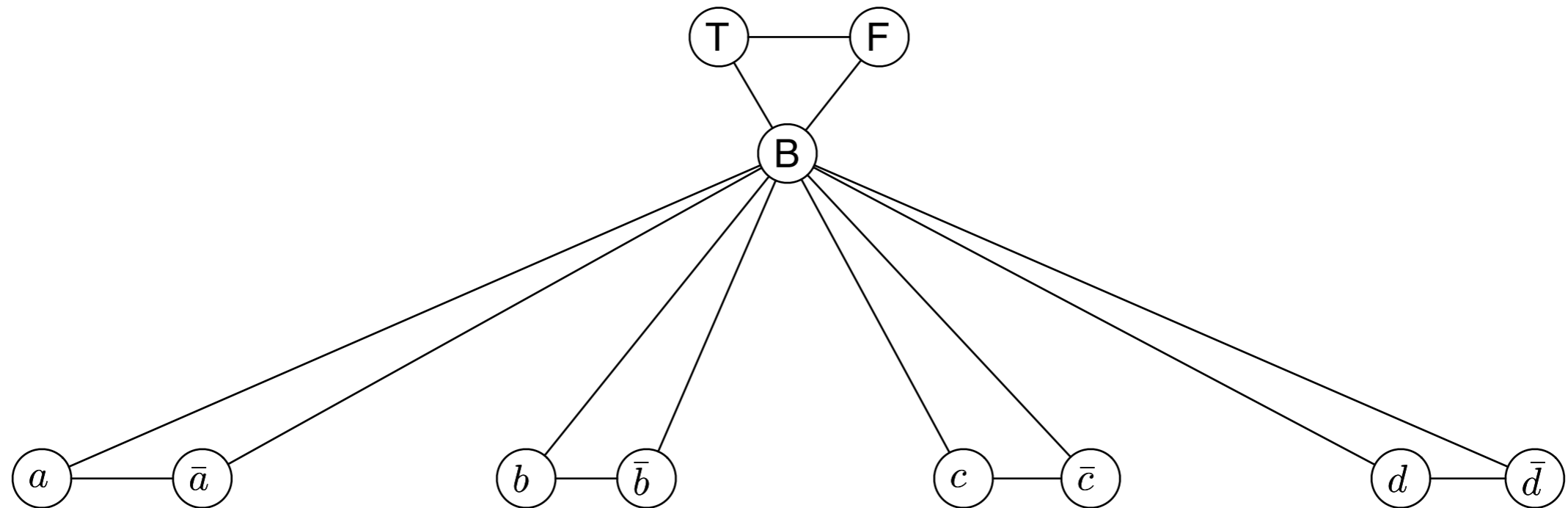
not 3 colorable

# Graph 3-Colorability

- Claim:  $3\text{-SAT} \leq_p 3\text{-color}$
- Given a  $3\text{-SAT}$  instance, we construct (in polynomial time) an instance of  $3\text{-color}$  such that the graph is colorable if and only if the boolean formula is satisfiable

# Graph 3-Colorability

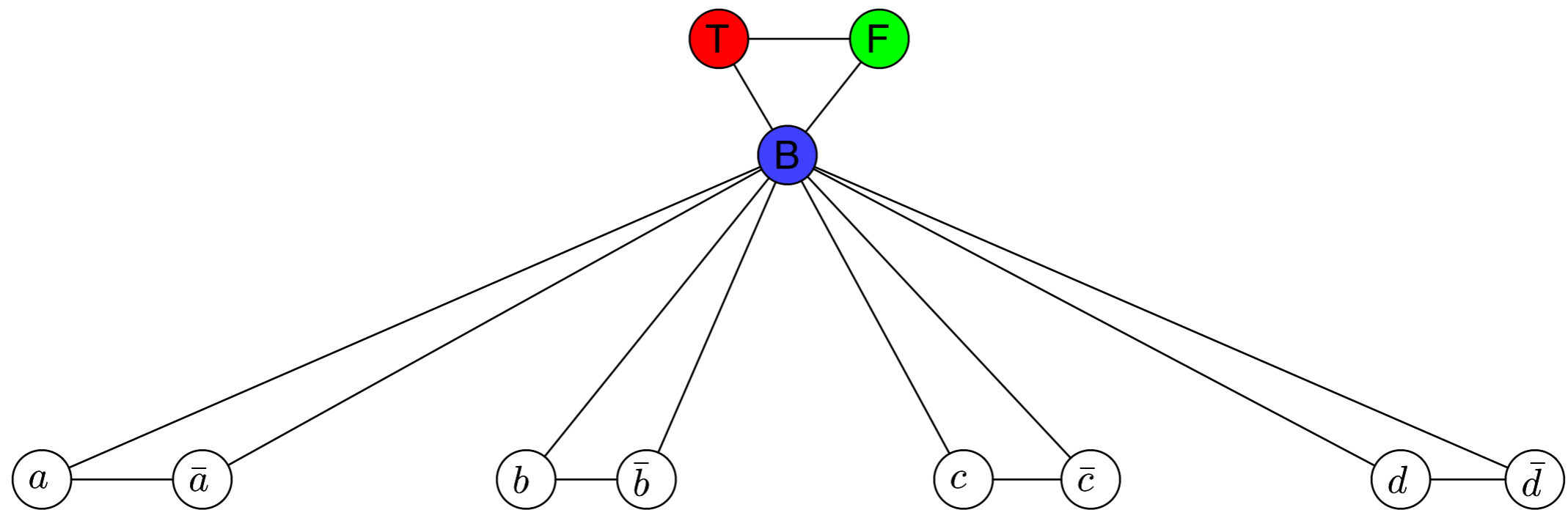
- We create a node for each variable and its negation
- We also create a triangle with labels B, False, and True
- Connect as shown below





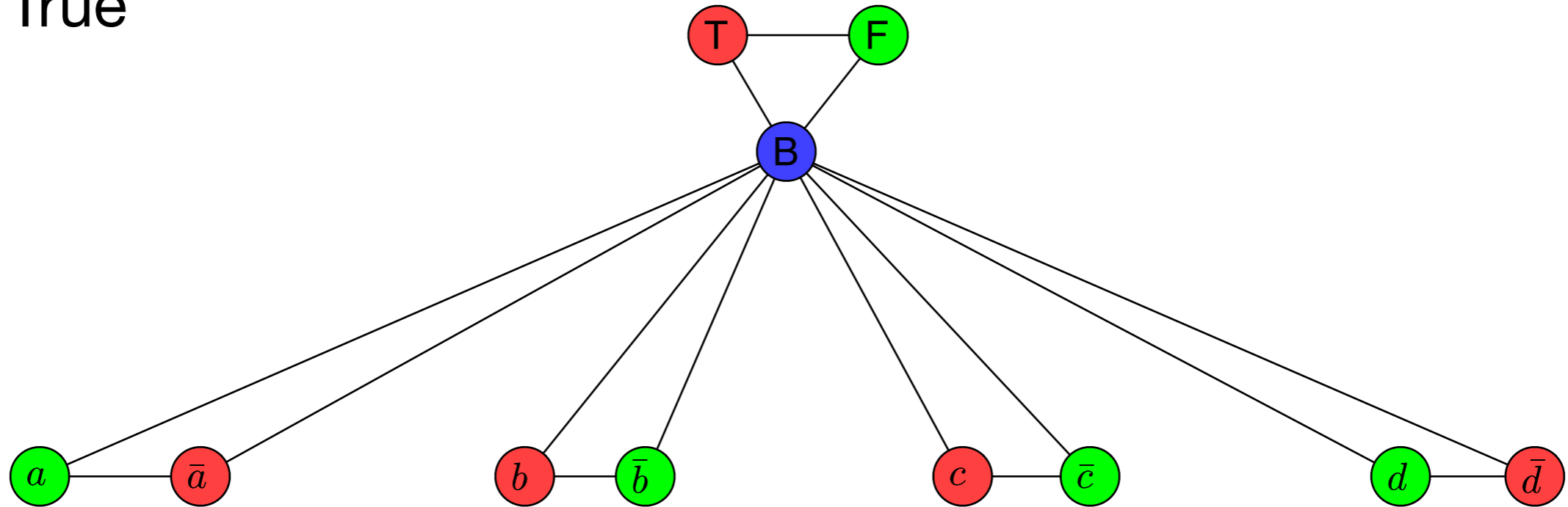
# Graph 3-Colorability

- This graph is clearly colorable
  - Start with the upper triangle



# Graph 3-Colorability

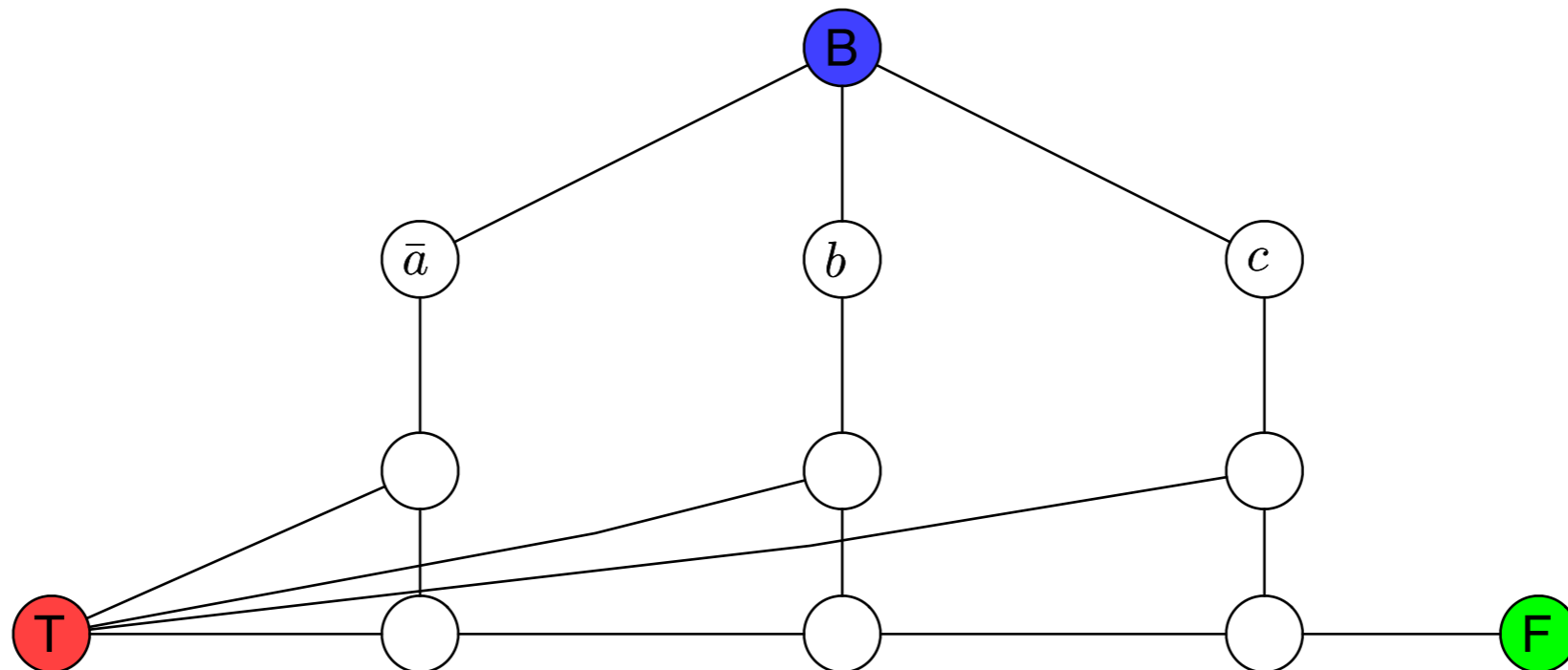
- Then for each variable, we get to decide whether we want to color the variable or its negation with the same color as True



- This "encodes" a truth assignment to the variables

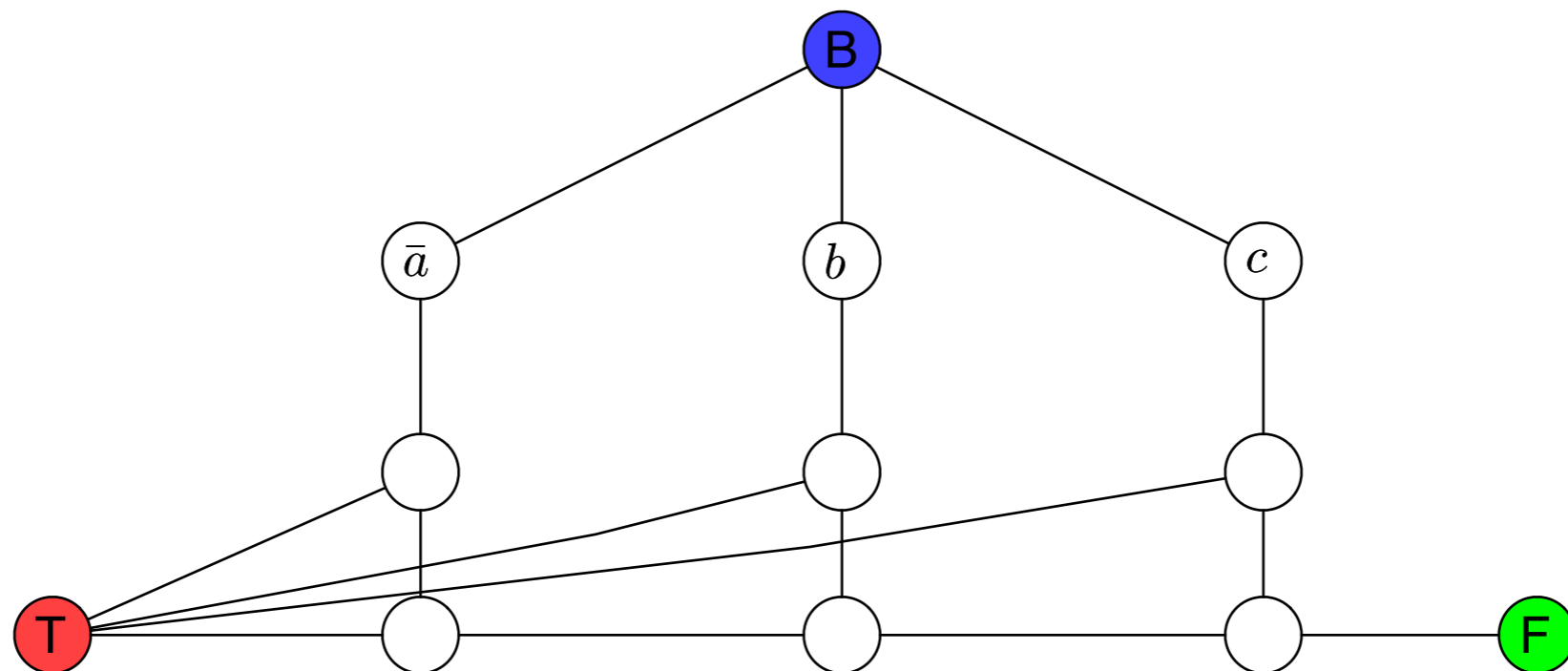
# Graph 3-Colorability

- Add for each clause in the formula a 6-node 13-edge gadget
  - Add the six nodes, then add edges as shown below
- Example:  $\neg a \vee b \vee c$



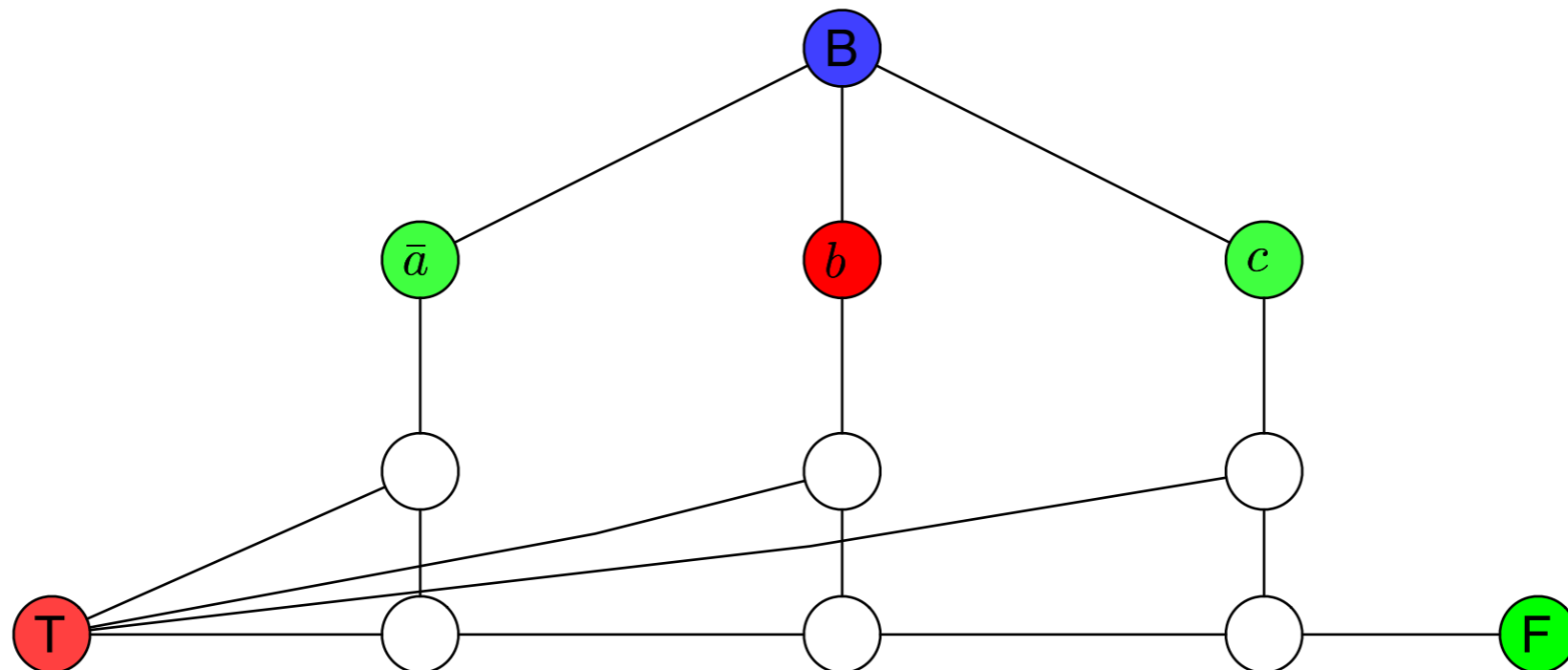
# Graph 3-Colorability

- If the graph is three colorable, then:
  - The gadget ensures that at least one element in the clause is True
- Example:  $\neg a \vee b \vee c$



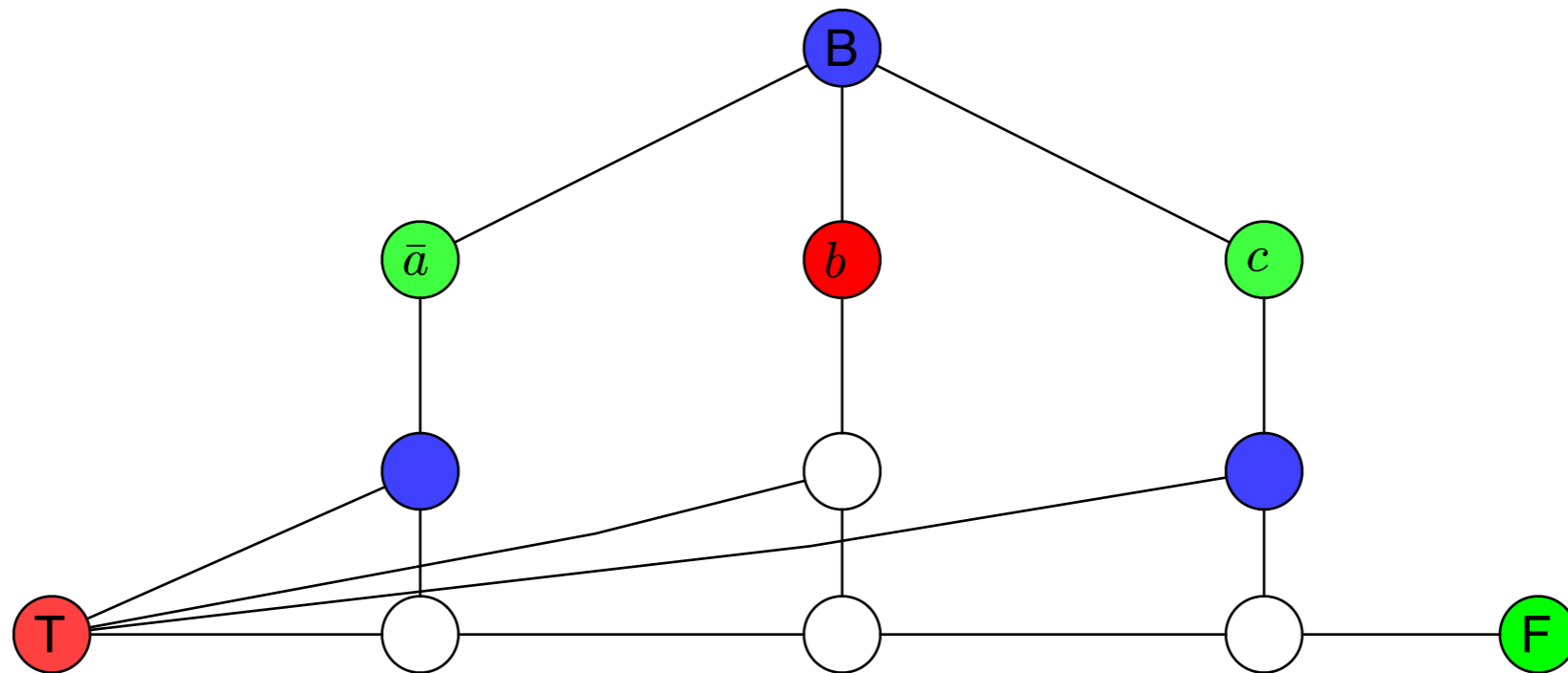
# Graph 3-Colorability

- Assume that we have an assignment to the variables that satisfies the formula and all of the clauses
  - Clause is  $\neg a \vee b \vee c$  and  $a$  and  $b$  are true and  $c$  is false
  - Color the variable nodes accordingly:



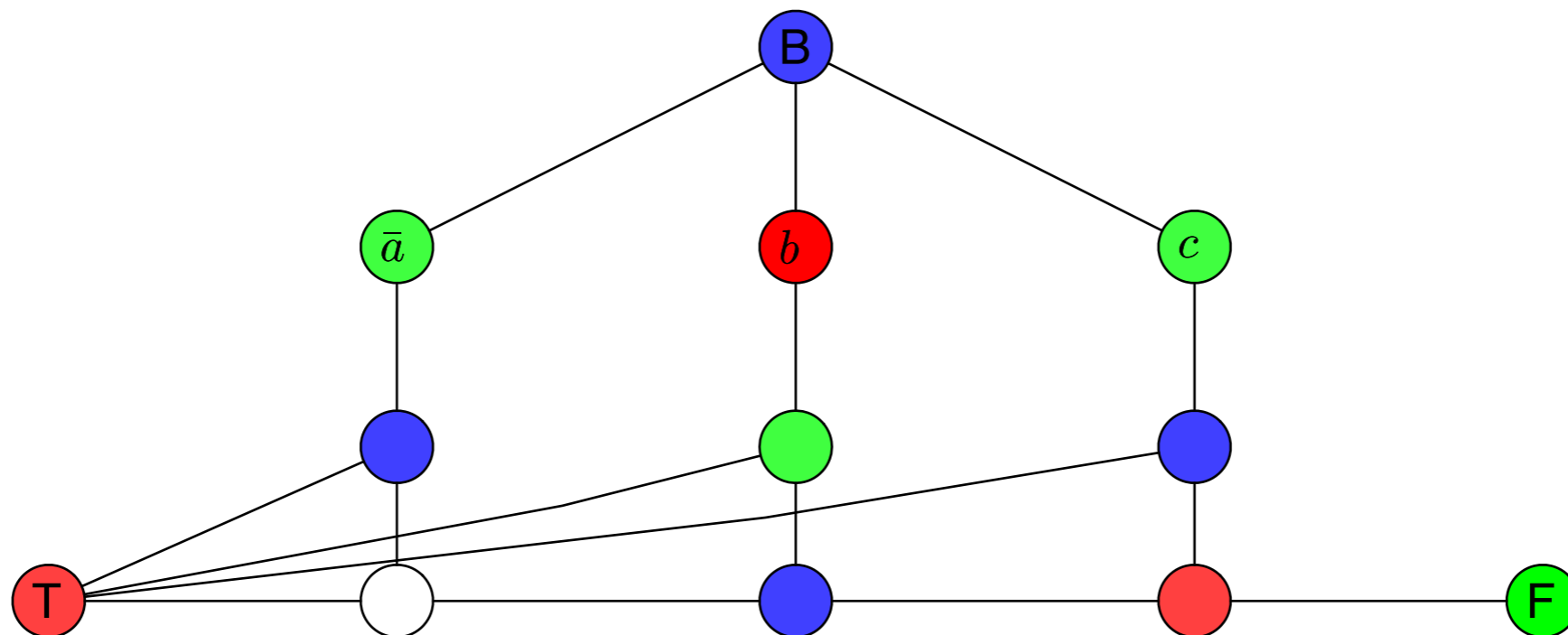
# Graph 3-Colorability

- Some node colors now follow by necessity



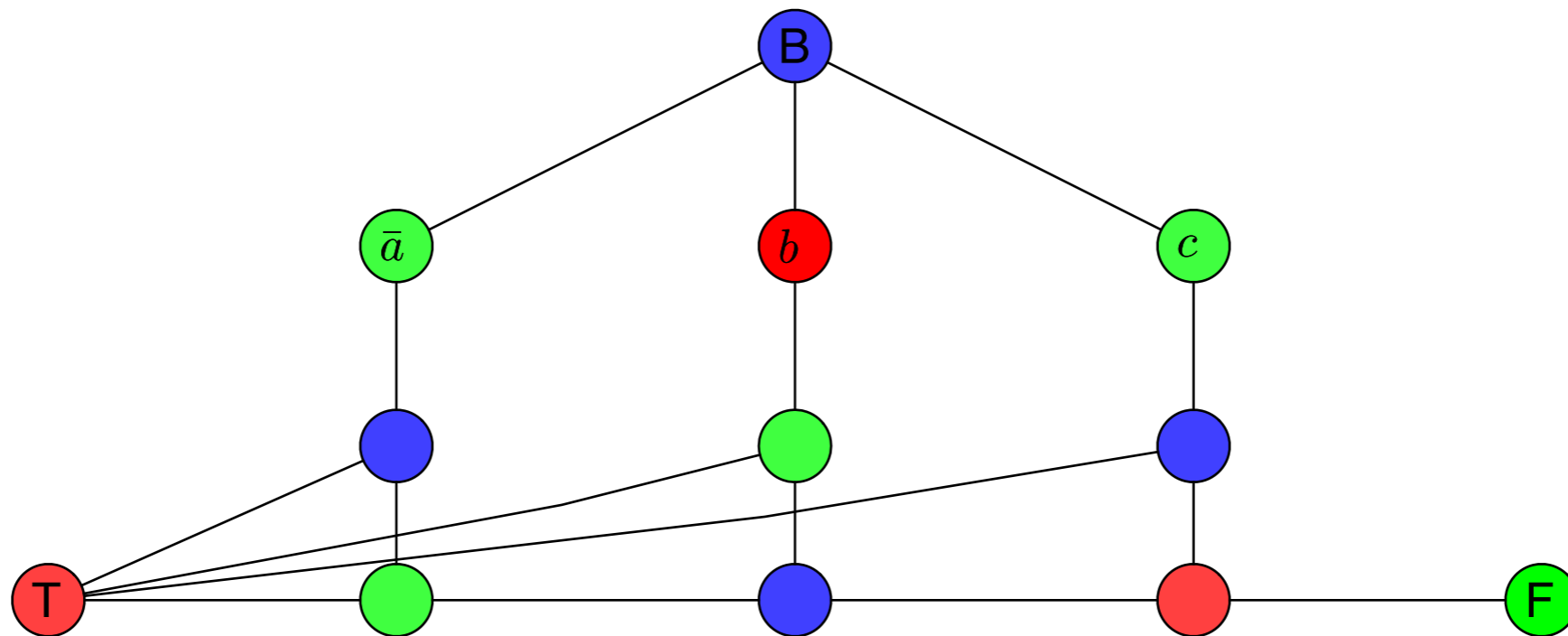
# Graph 3-Colorability

- Color the node under a True-colored vertex with the False color and the one underneath with Blue



# Graph 3-Colorability

- The last vertex color follows



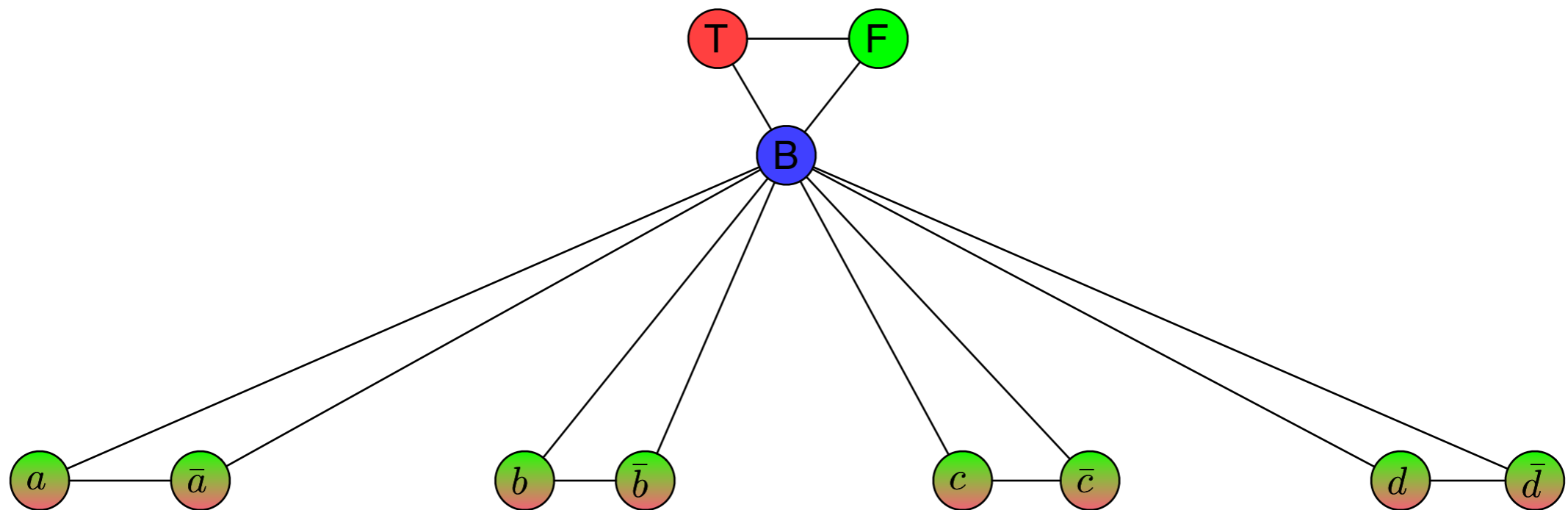


# Graph 3-Colorability

- This gadget is now three-colored
  - Do the same thing for all gadgets
  - Get a three coloring of the graph

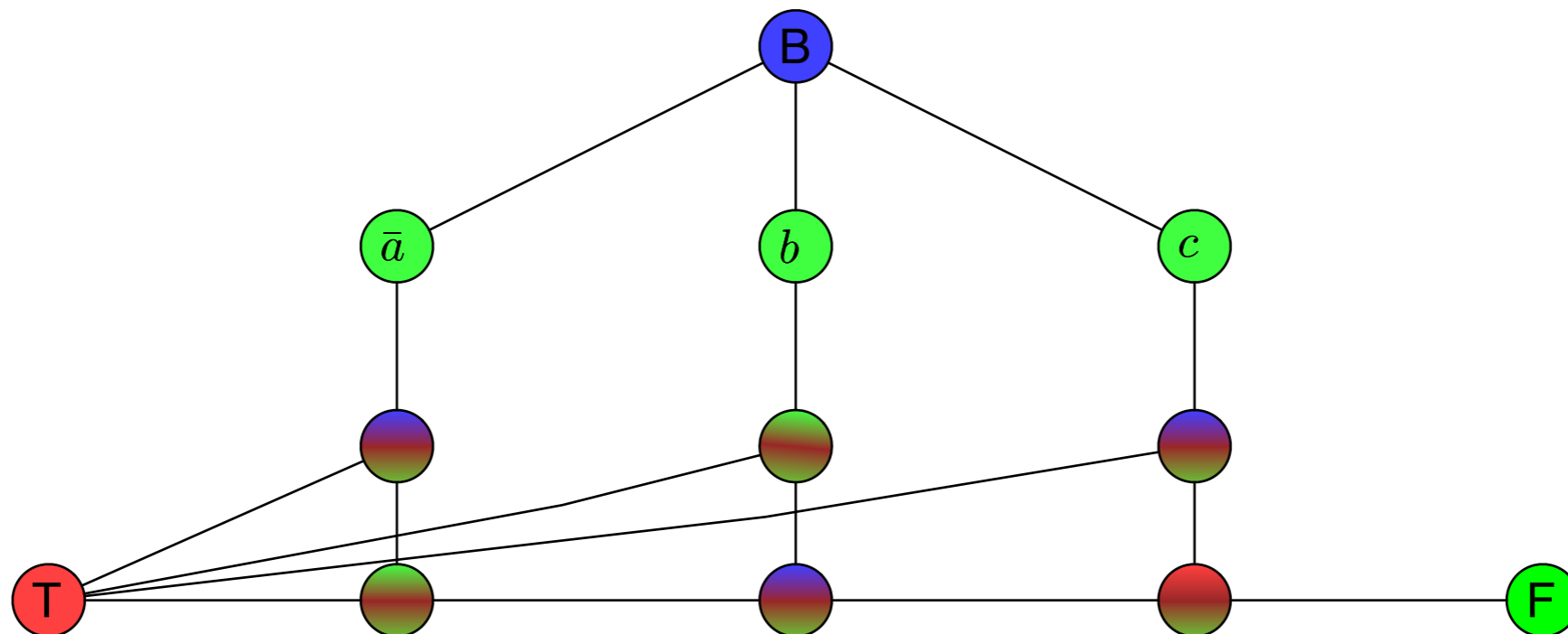
# Graph 3-Colorability

- Vice versa, assume that you have a three coloring
- The base part of the graph assures that each literal is colored either with Red (for True) or with Green (for False)



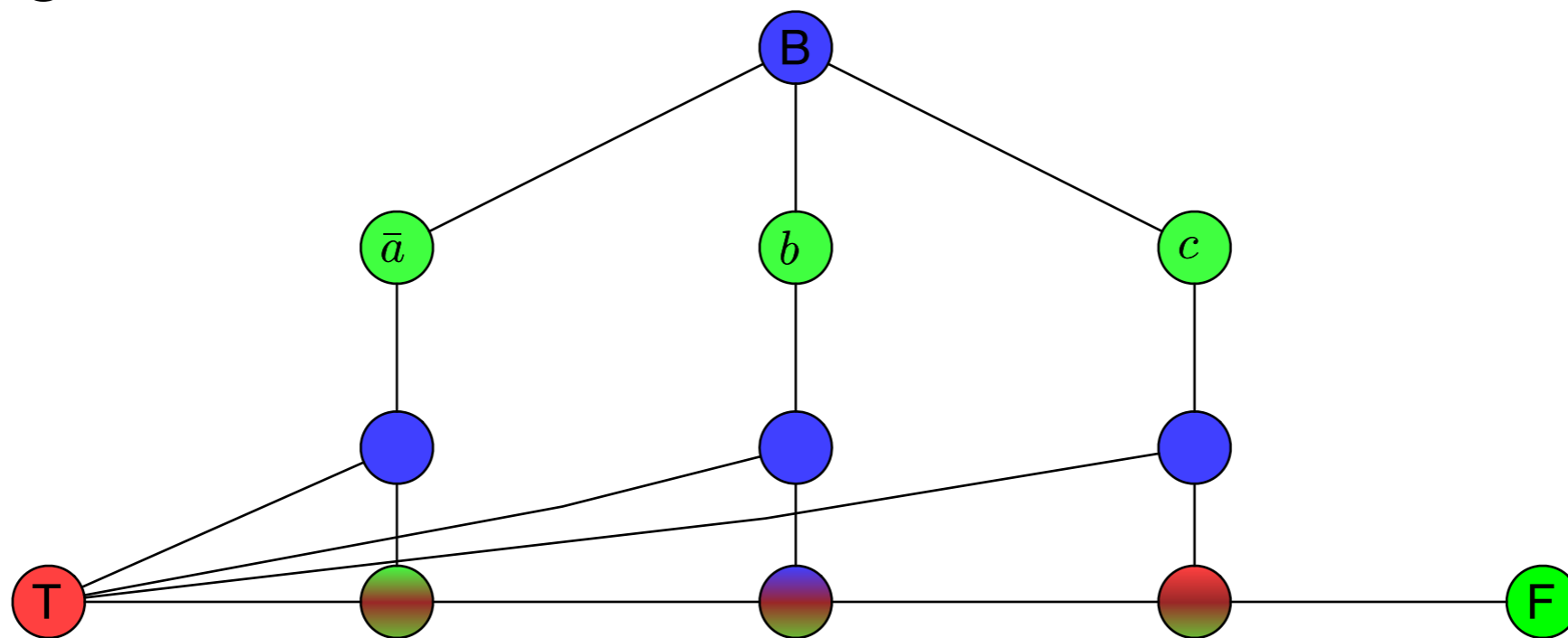
# Graph 3-Colorability

- If the three coloring where to assign Green (the color of Falsehood) to the literal nodes, then what would happen



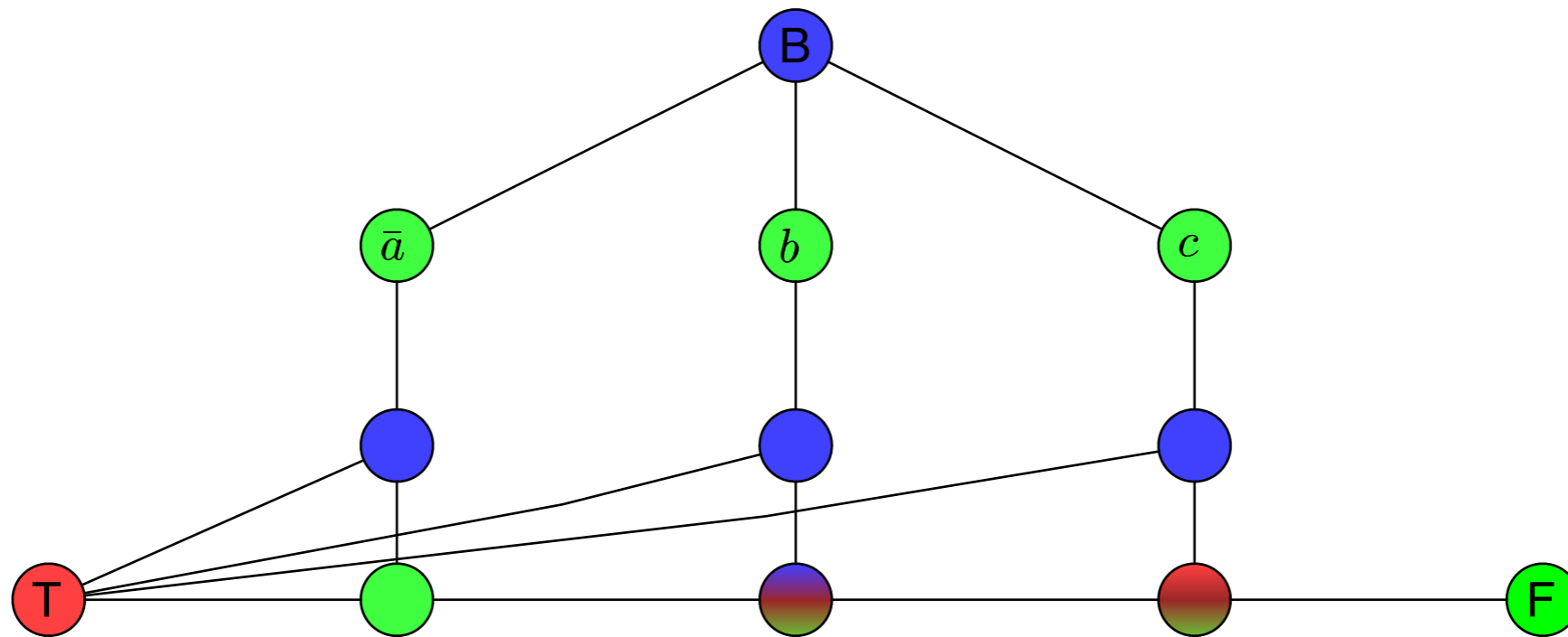
# Graph 3-Colorability

- We can deduce some colors
  - The middle layer of nodes has to be blue, because they have a Red neighbor (the true node) and a Green neighbor



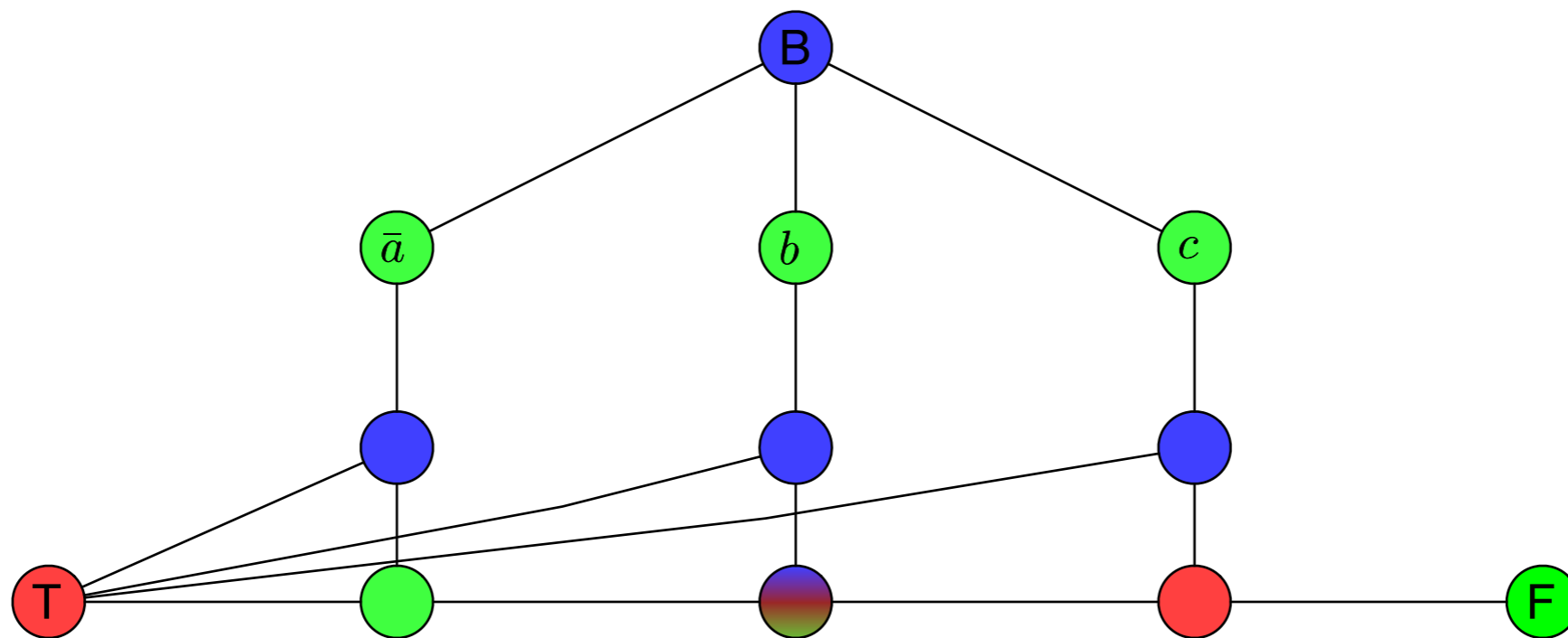
# Graph 3-Colorability

- The vertex on the lower row and to the left needs to be Green because it has a Blue and a Red neighbor



# Graph 3-Colorability

- The vertex on the lower row on the right has to be Red



# Graph 3-Colorability

- But now the middle node has a Red, a Green, and a Blue neighbor
- Thus the graph is not 3-colorable
- Therefore: The graph is 3-colorable implies that in the gadget, one vertex is colored Red (the color of Truth)
  - Not to be confused with <https://www.youtube.com/watch?v=8epG4AezdYg> which is just propaganda but the color of the Holy Spirit

# Graph 3-Colorability

- So, the complete graph is 3-colorable if in all gadgets at least one of the literal nodes is colored Red
- Then all clauses are satisfied
- Then the formula is satisfied



# Consequences

- There is a large set of known NP-complete problems
  - Many problems that appear in practice can be shown to be NP-complete
    - You look through the catalogue of NP-complete problems and reduce your problem to one of them
    - (If I can solve my problem polynomially, then I can solve that problem polynomially, but that would mean that  $\mathcal{P} = \mathcal{NP}$ , which we believe to be false)
- If this is the case:
  - You cannot expect a solution that scales well
  - But it might be perfectly solvable for the instance sizes that you need to solve