

# Some Generalised Reductions of Ordered Binary Decision Diagramm (GroBdd)

Joan Thibault

# Boolean Functions

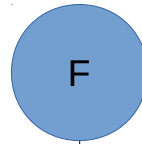
- Why ?
  - Computer Aided Design (e.g. digital circuit synthesis)
  - Knowledge Representation (e.g. Artificial Intelligence)
  - Combinatorial Problems (e.g. N-Queens problem)
- What ?
  - Compact representation
  - Operations (e.g. composing, concatenating, evaluation)
  - Operators (e.g. AND, XOR, ITE, NOT)
  - Reductions (e.g. quantification, partial evaluation, SAT)

# Boolean Functions

- Various representations
  - Truth Table
  - Conjunctive / Disjunctive Normal Form
  - And Inverter Graph
  - Binary Decision Diagramm
    - Reduced Ordered BDD
    - Zero suppressed BDD
    - Xor based BDD

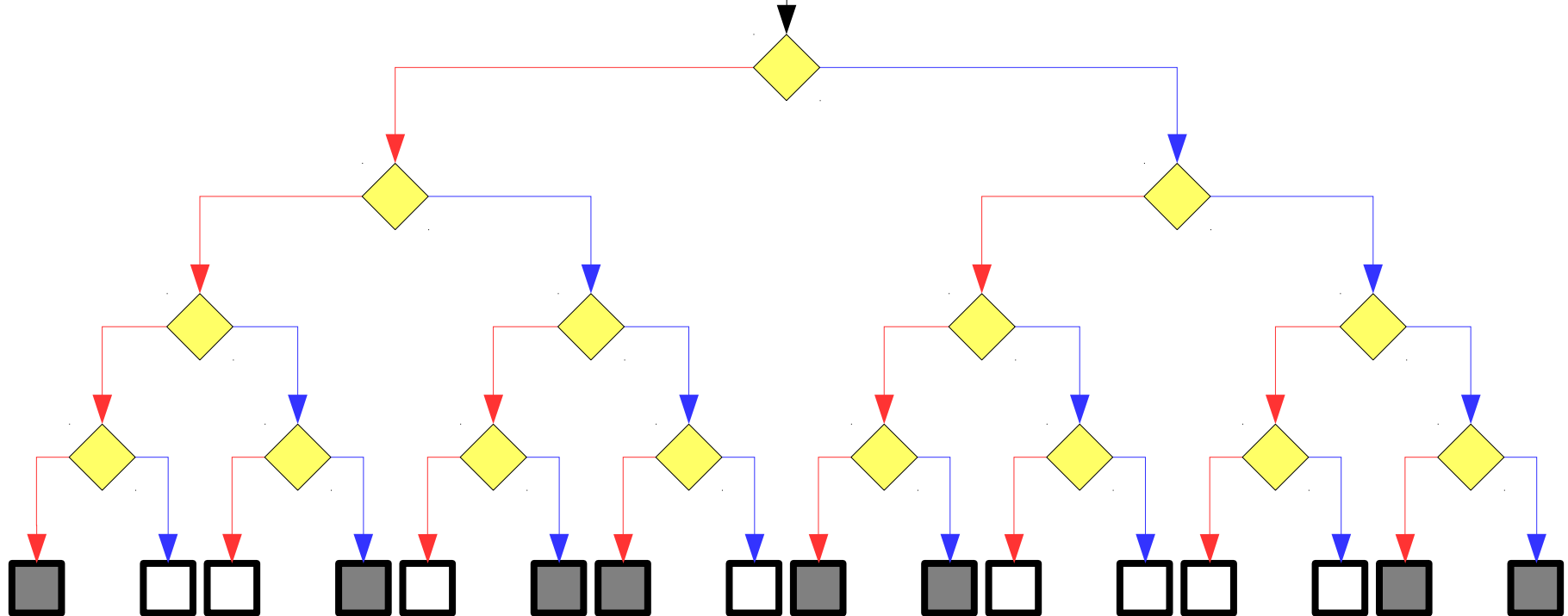
# Section 1

## What is a ROBDD ?



→ = if False

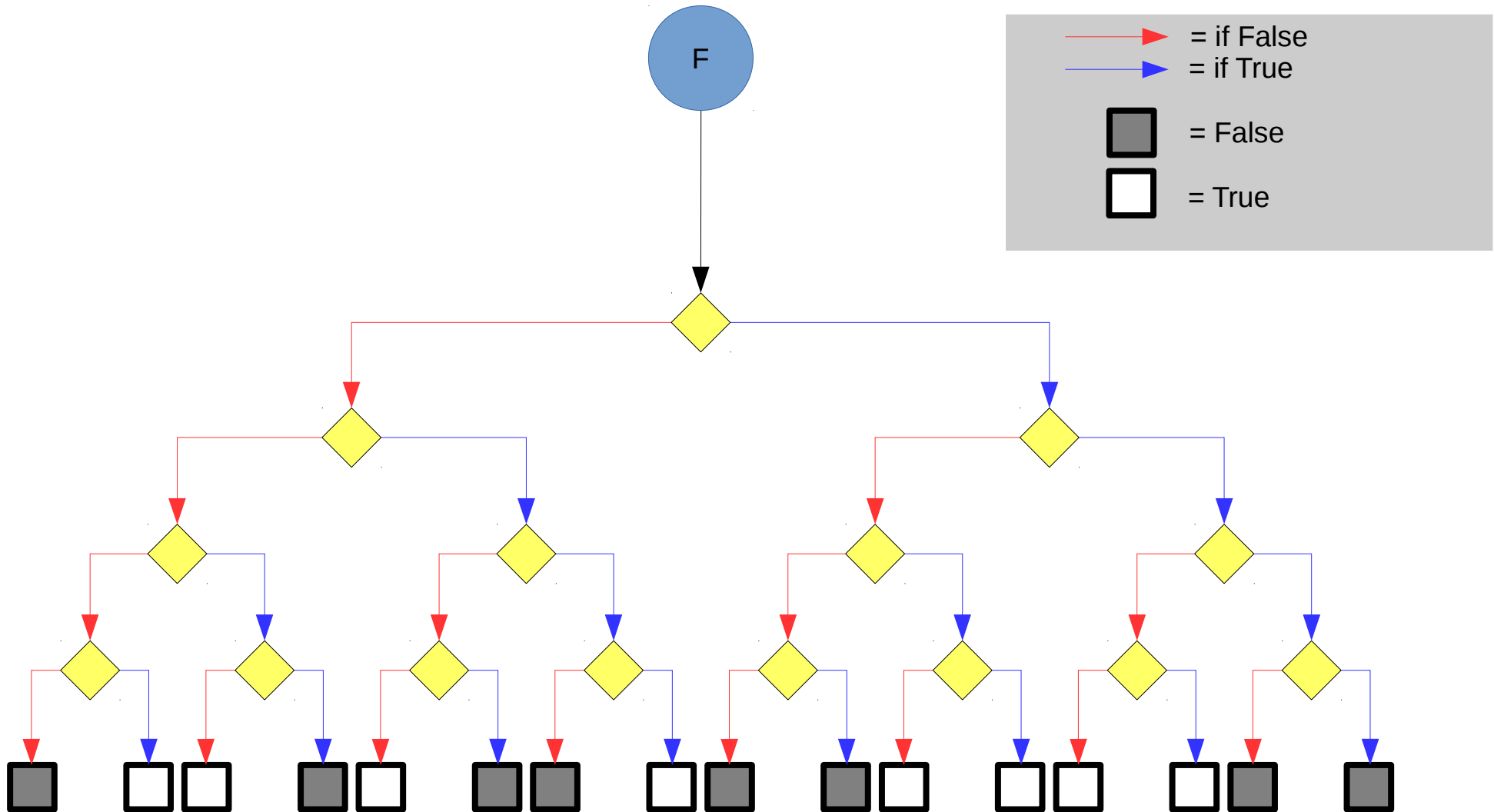
→ = if True



■ = False

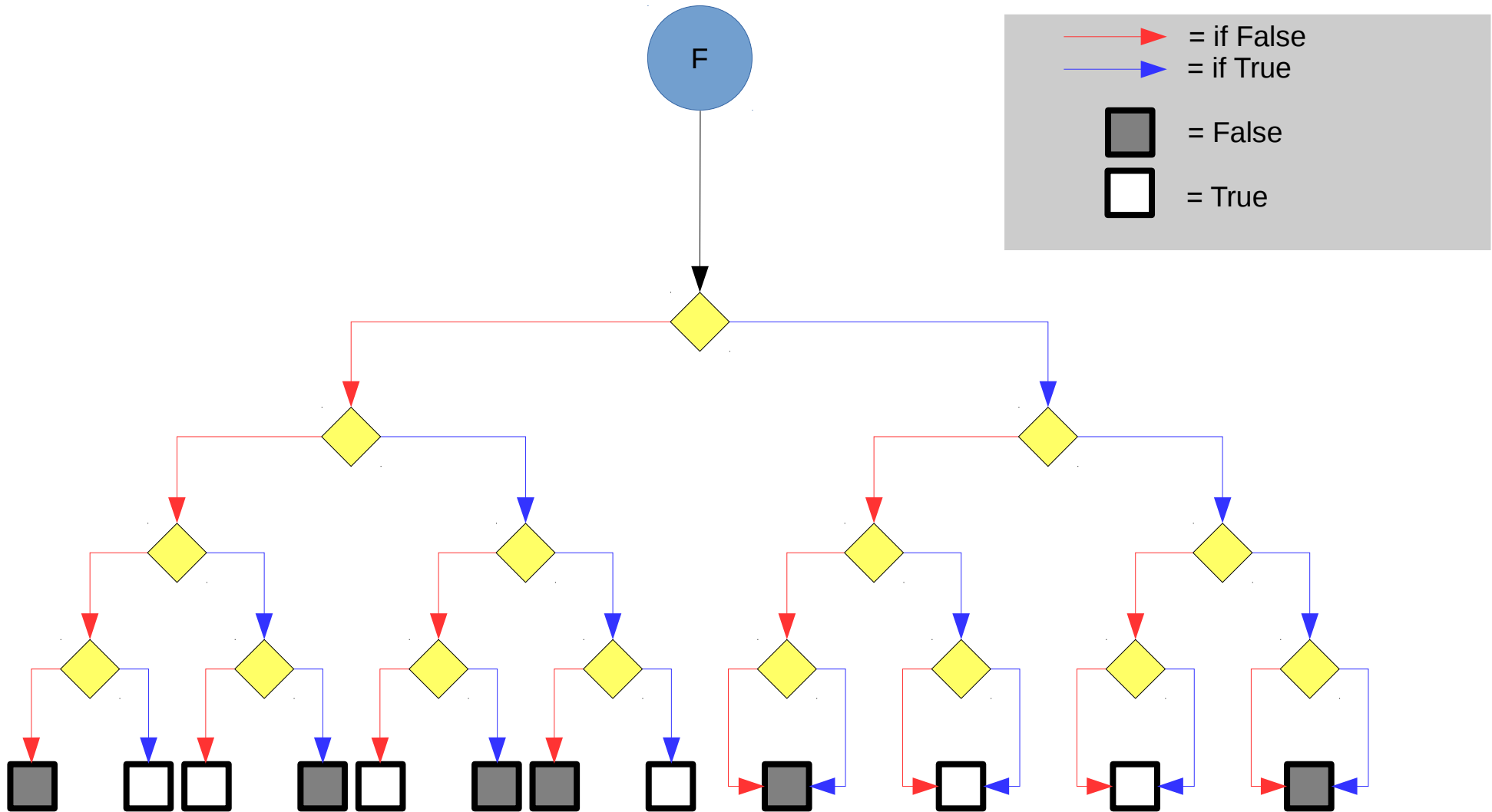
□ = True

(Bryant) Step 1: we merge isomorphic sub-graphs



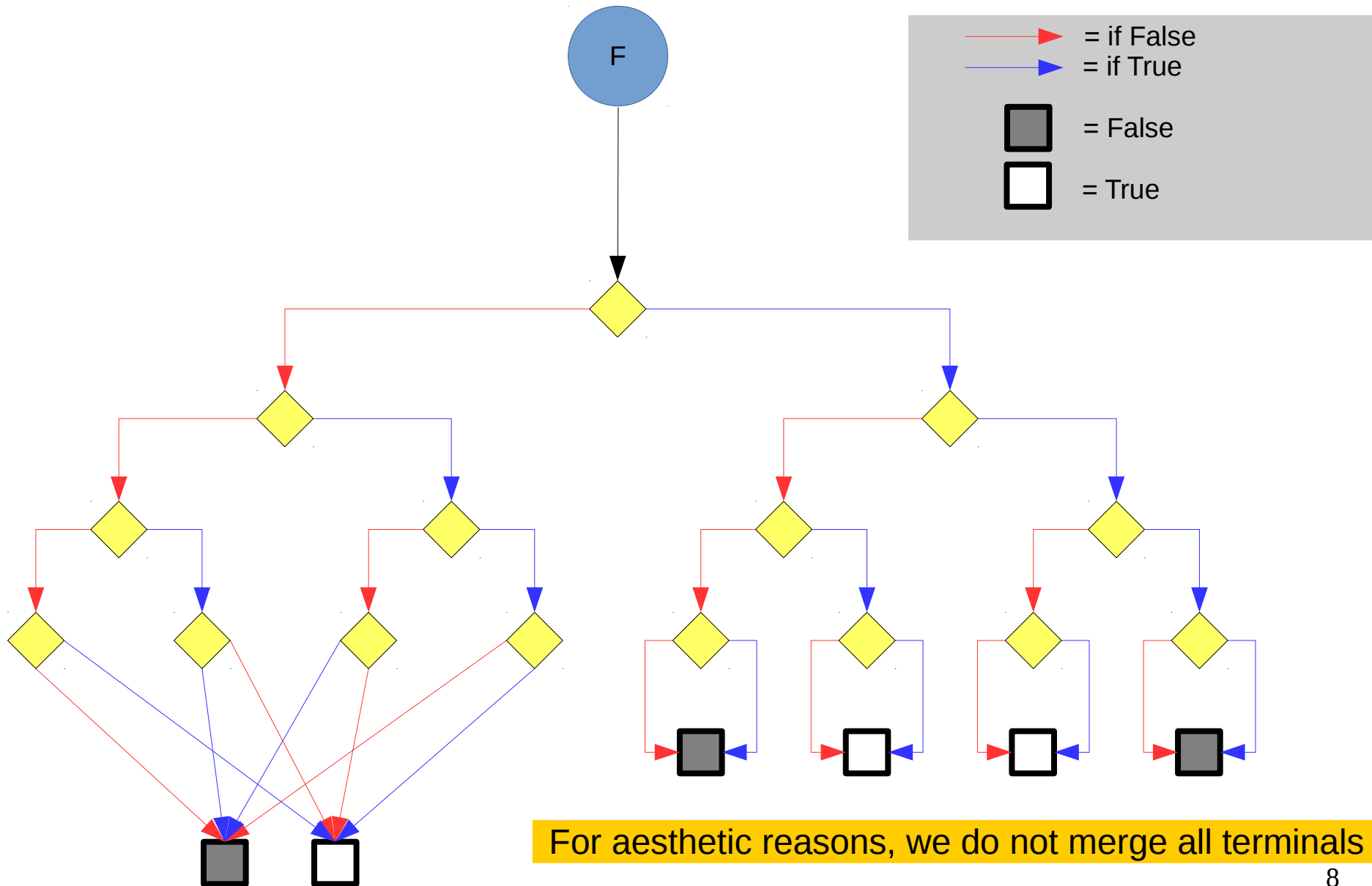
For aesthetic reasons, we do not merge all terminals

(Bryant) Step 1: we merge isomorphic sub-graphs



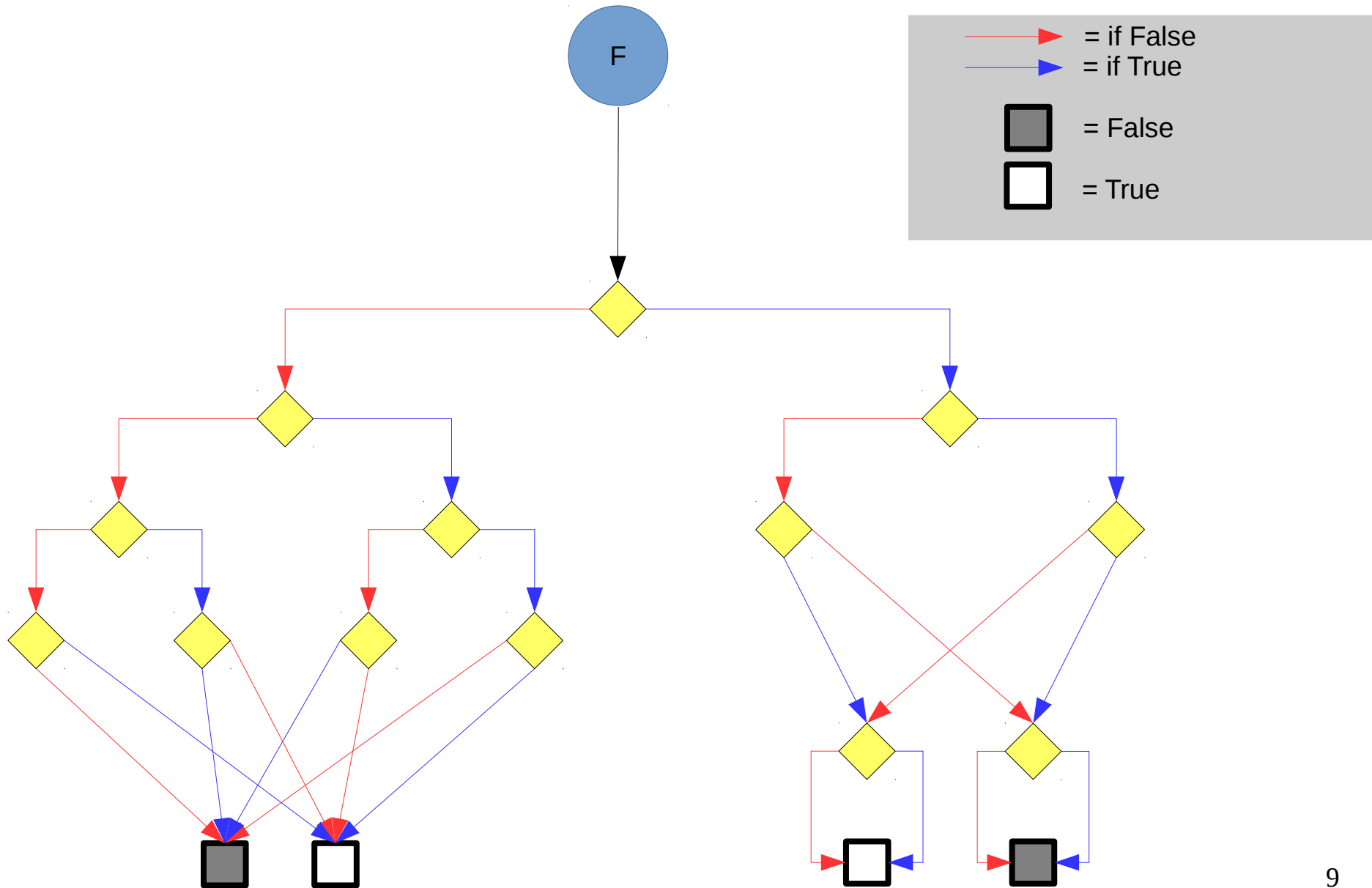
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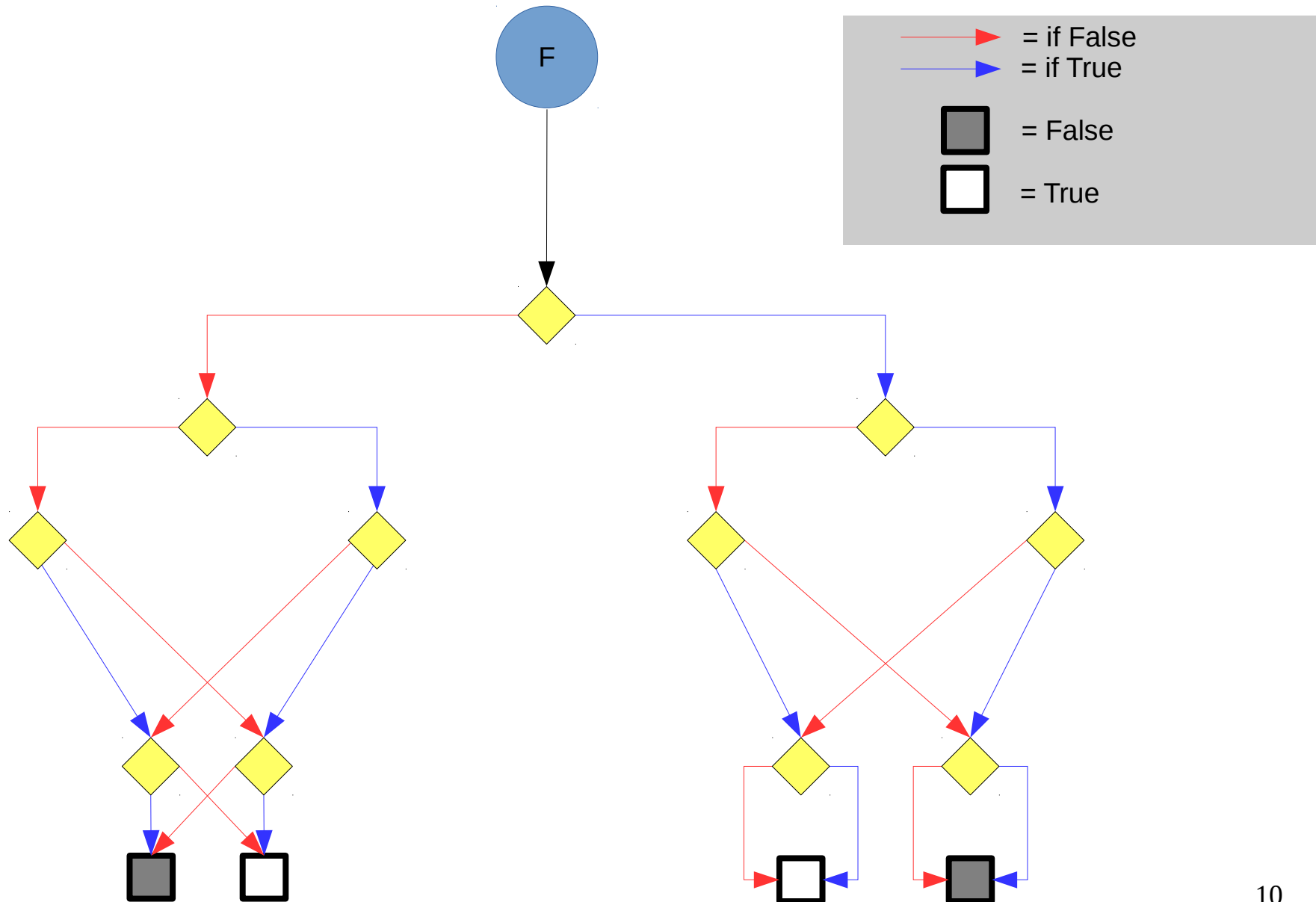




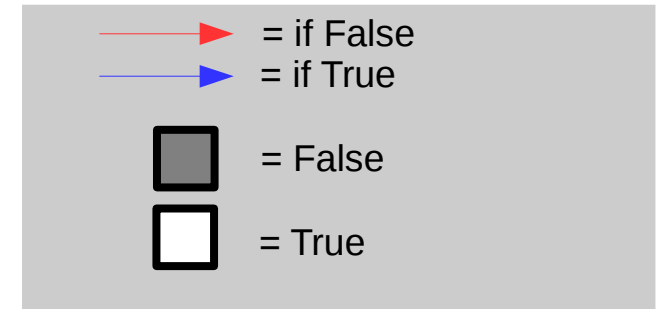
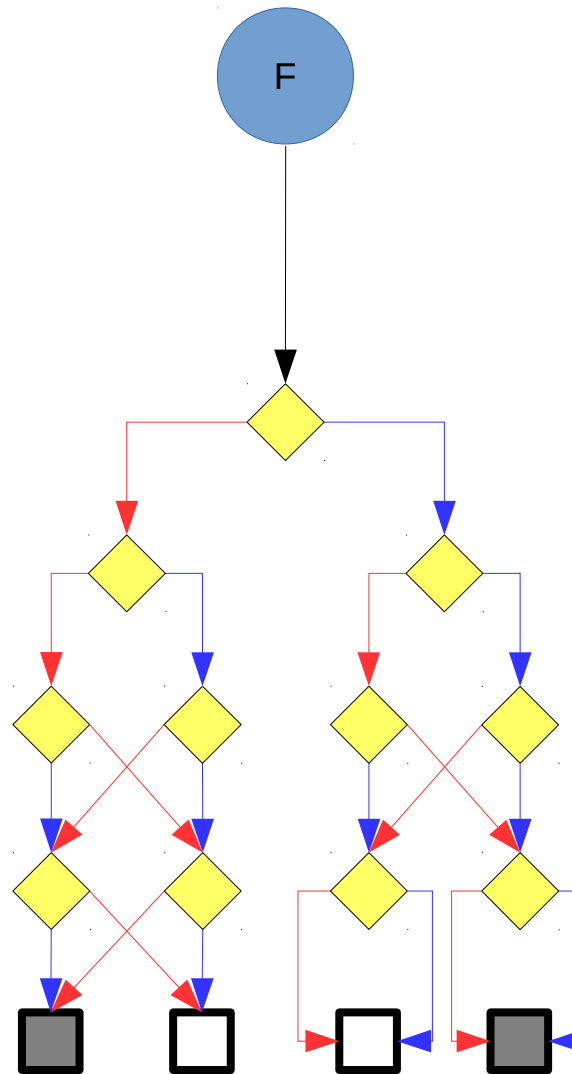
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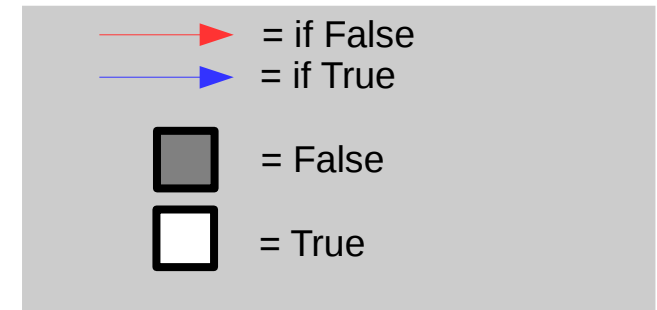
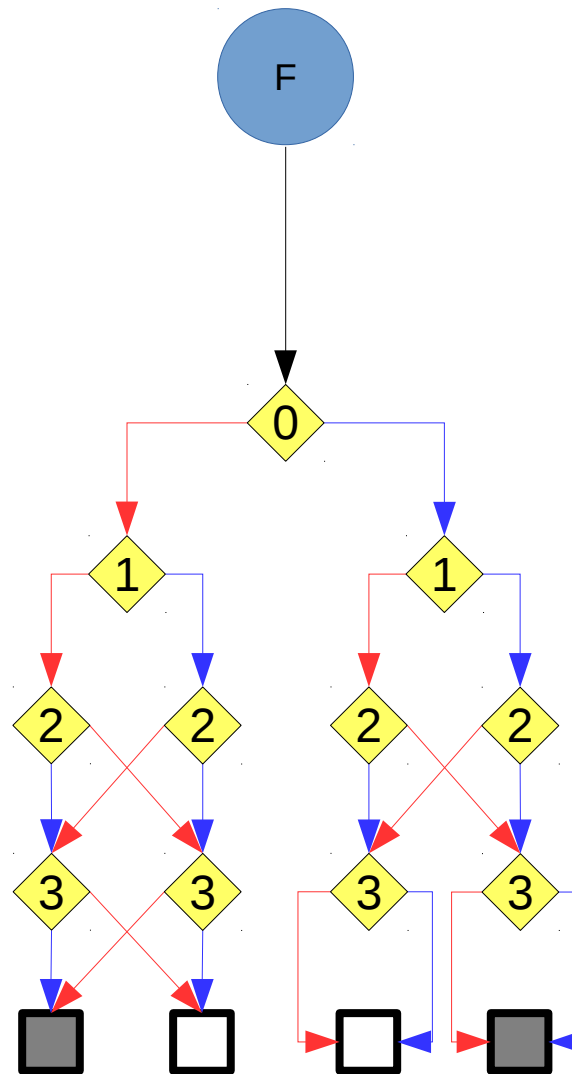
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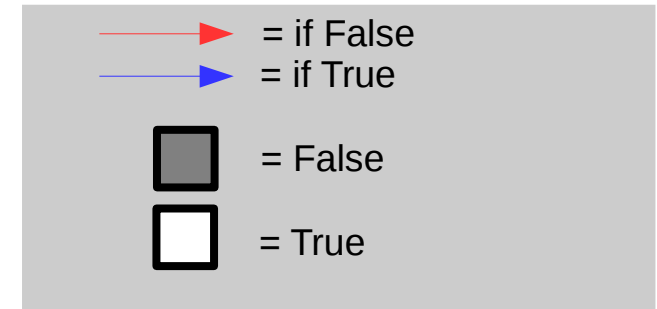
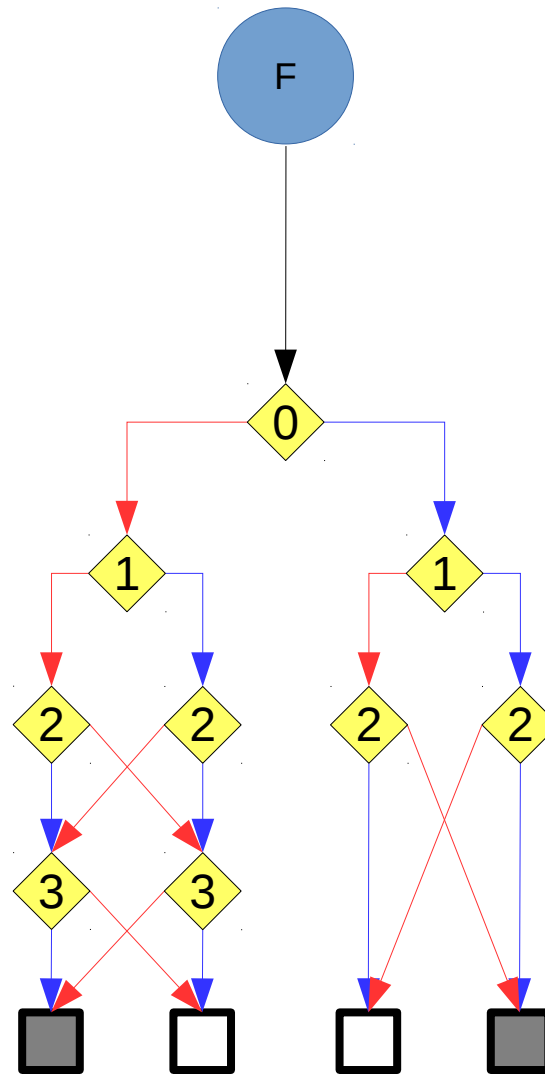
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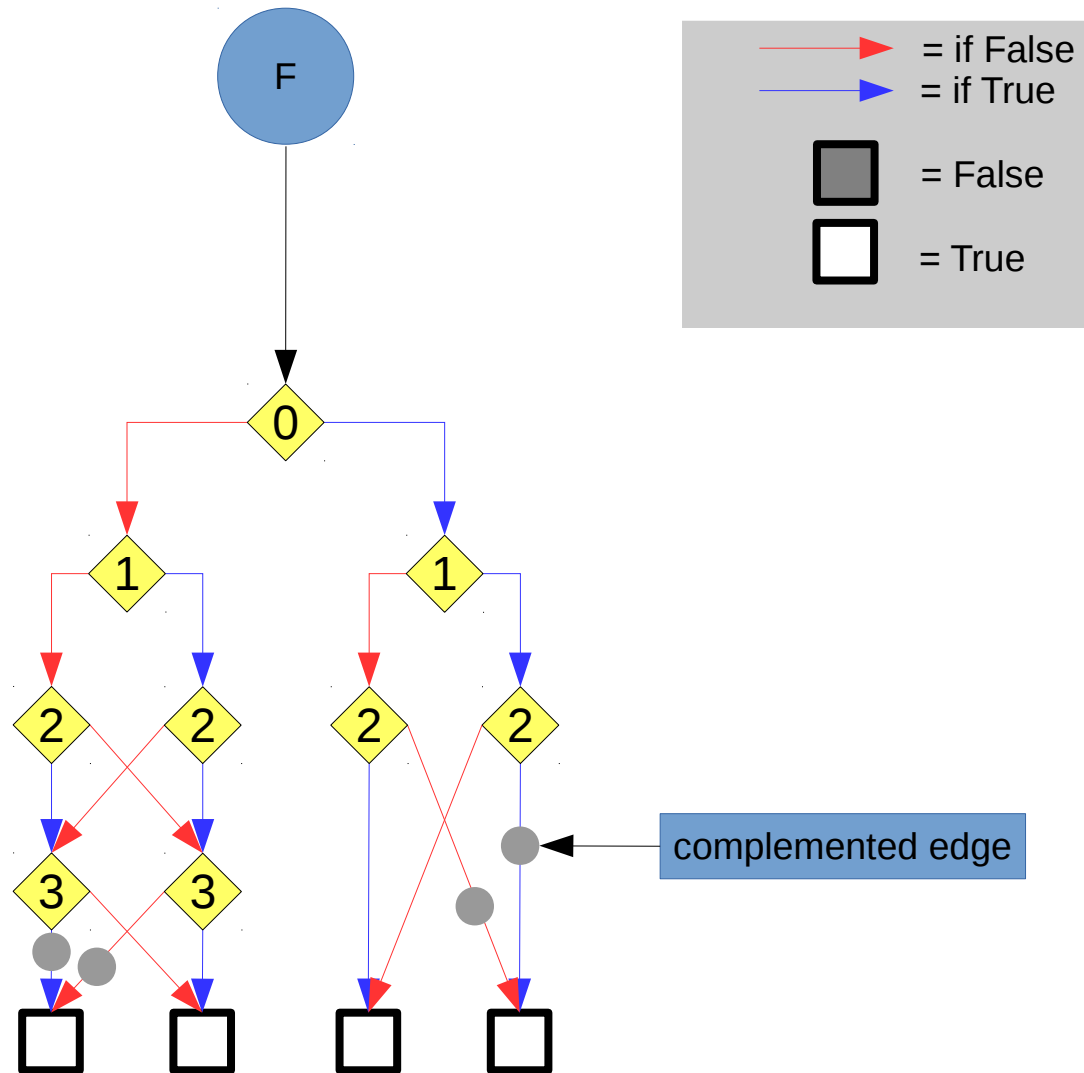
(Bryant) Step 2: we specify for each node:  
on which variable the decision is made



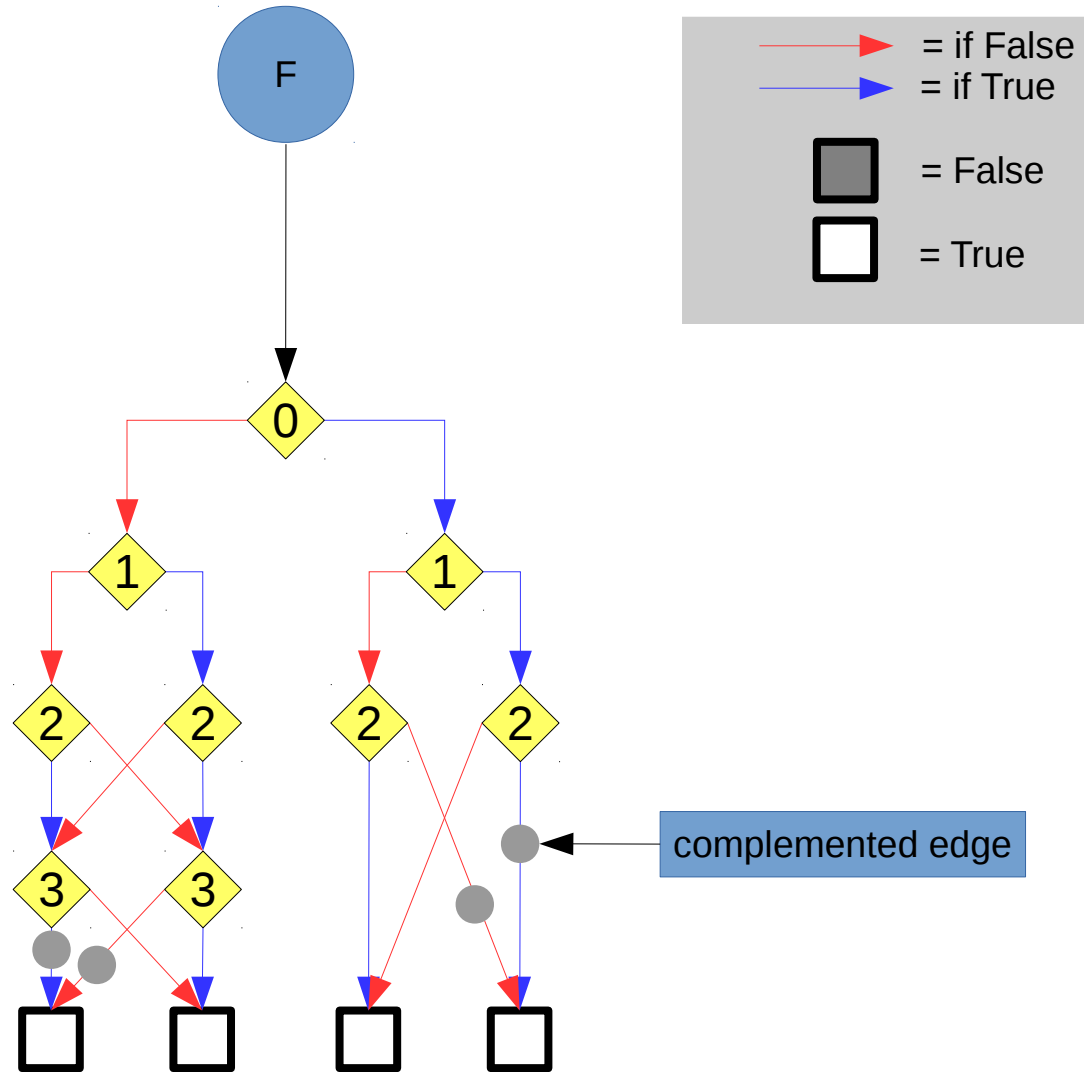
(Bryant) Step 3: we remove useless decisions



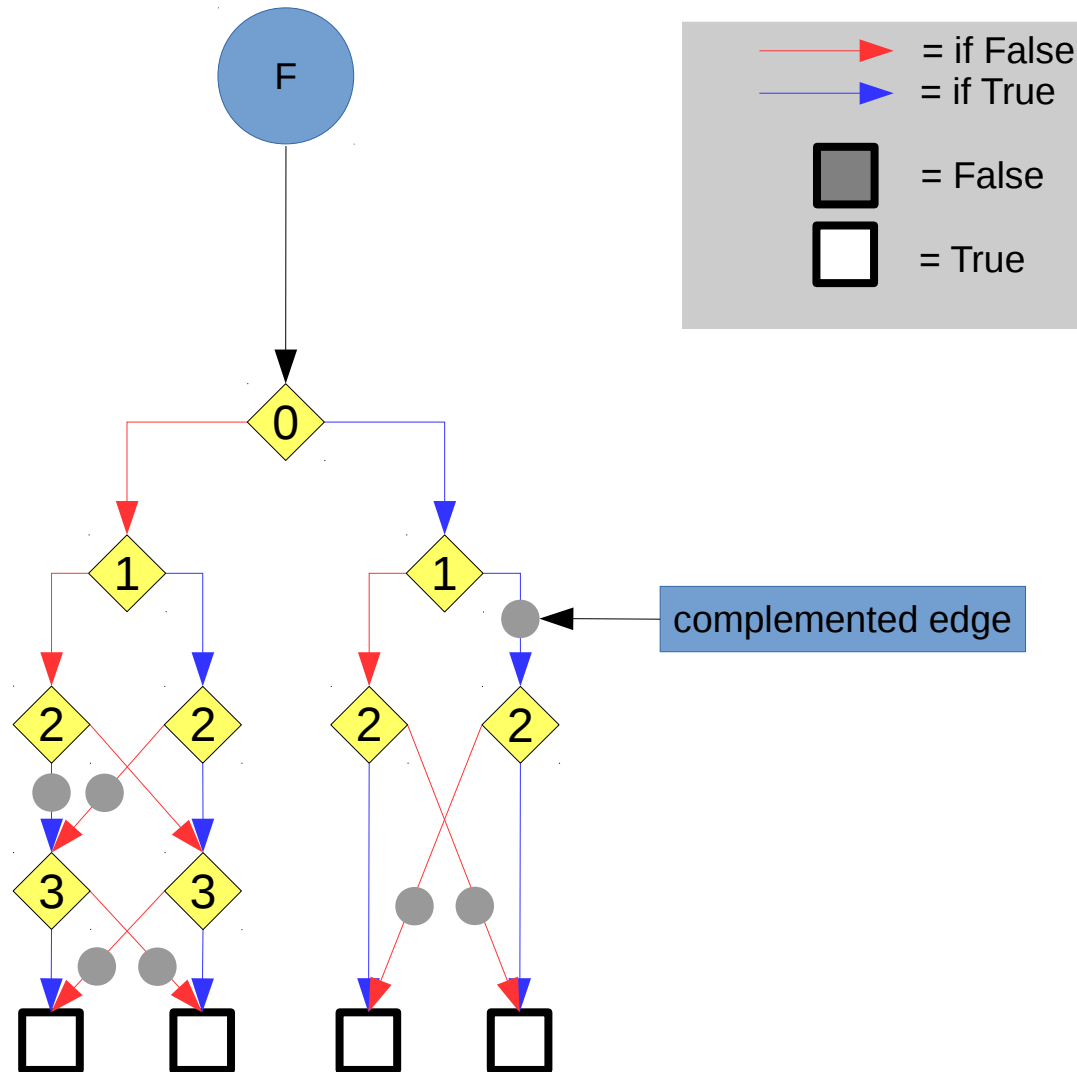
(Complemented Edges) Step 1: we replace the False node by a complemented edge to True



(Complemented Edges) Step 2 : we propagate inverted edges upward, ensuring that no “if True” edge is complemented

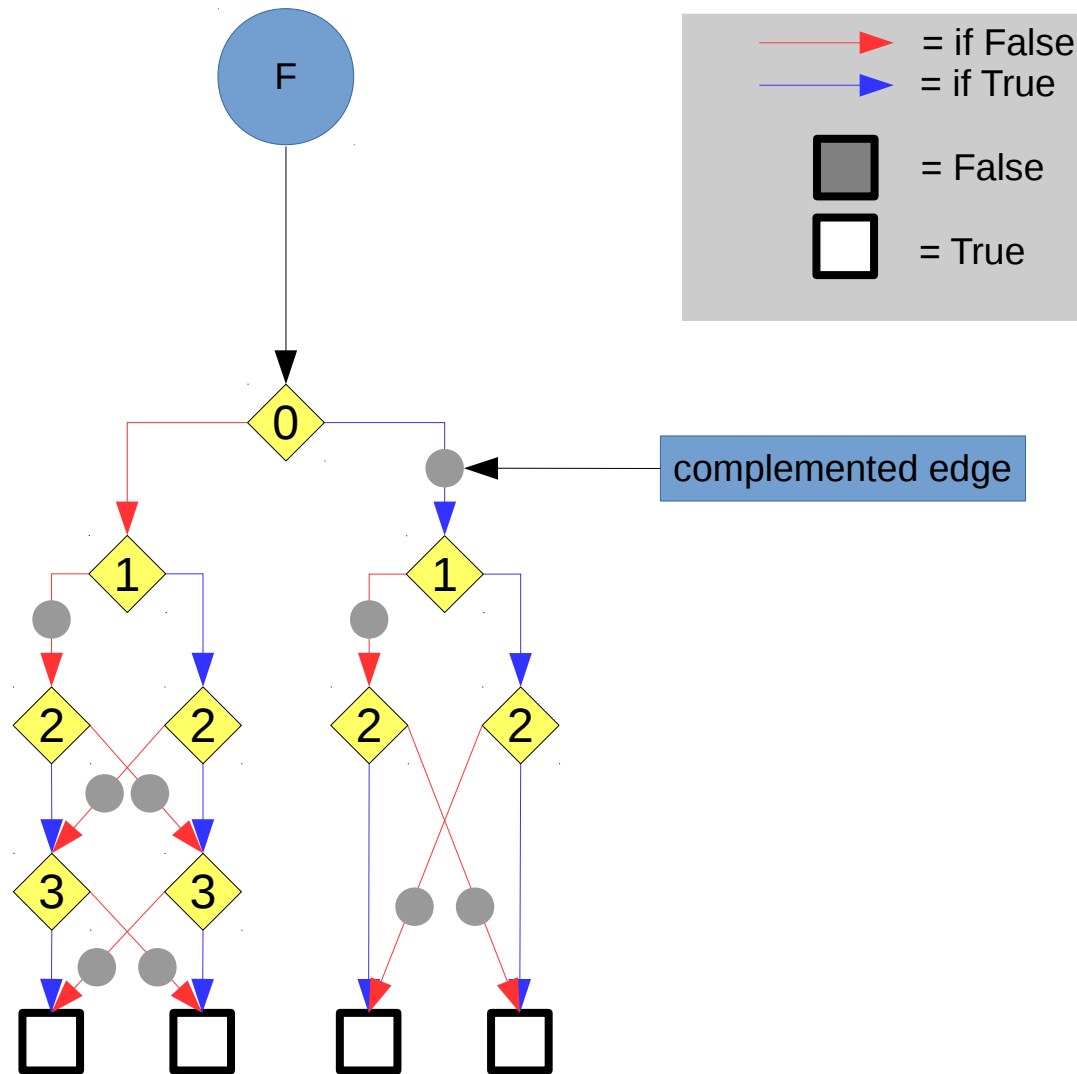


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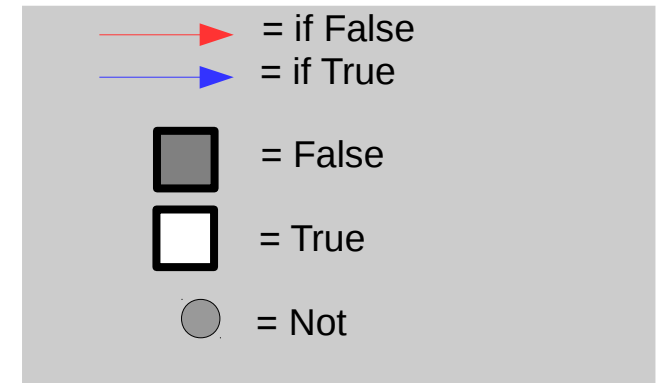
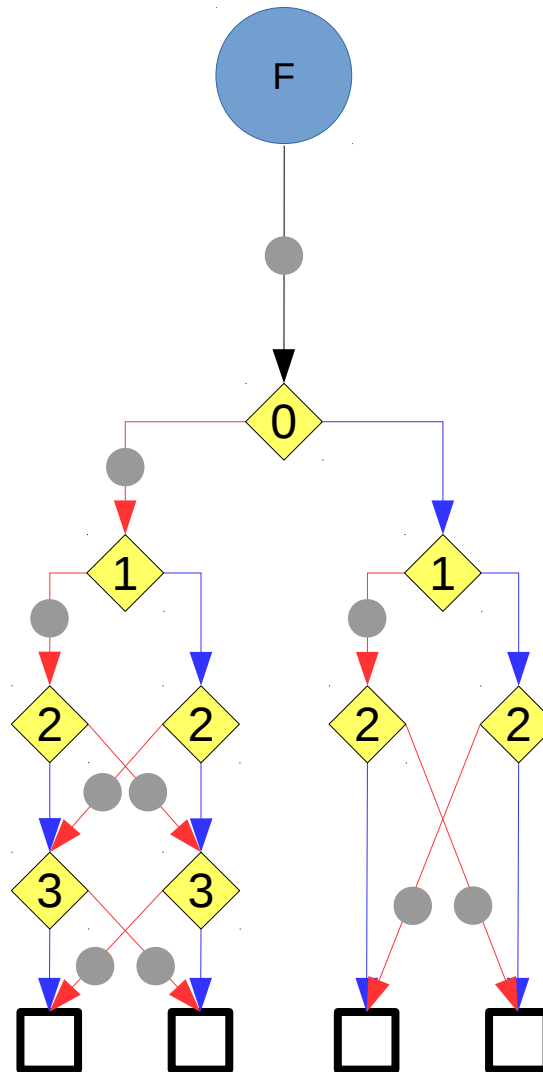




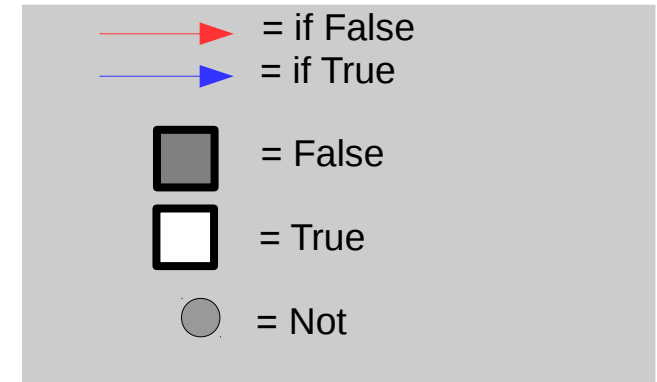
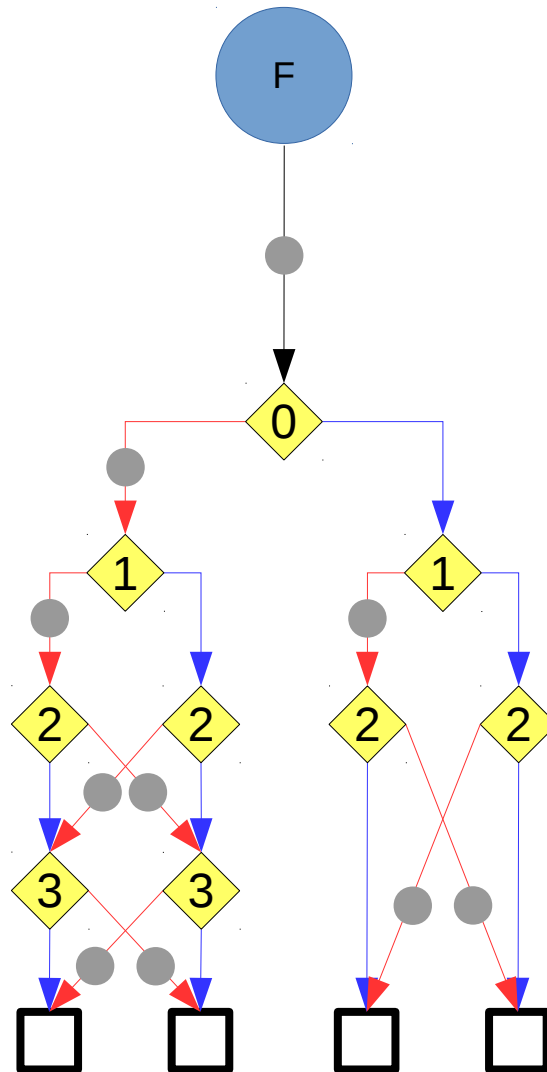
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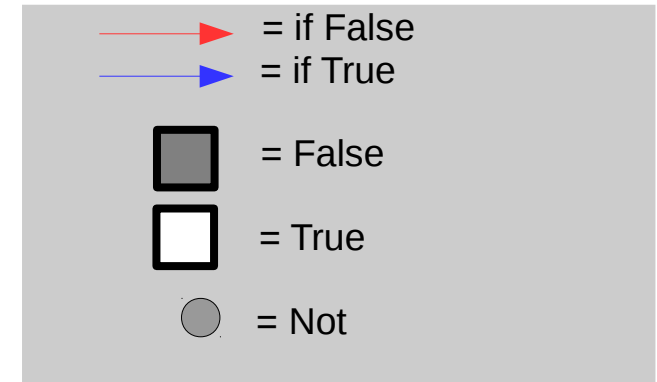
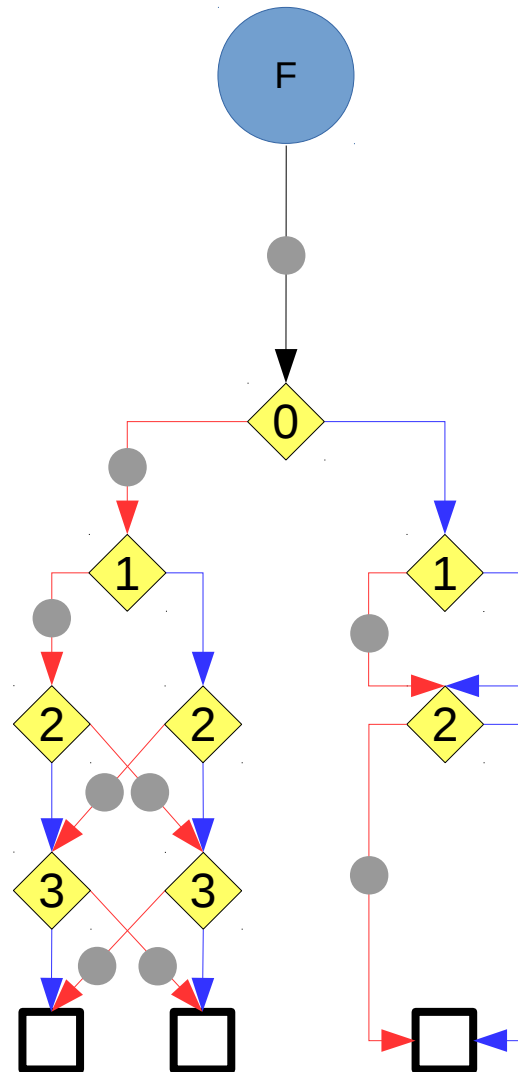
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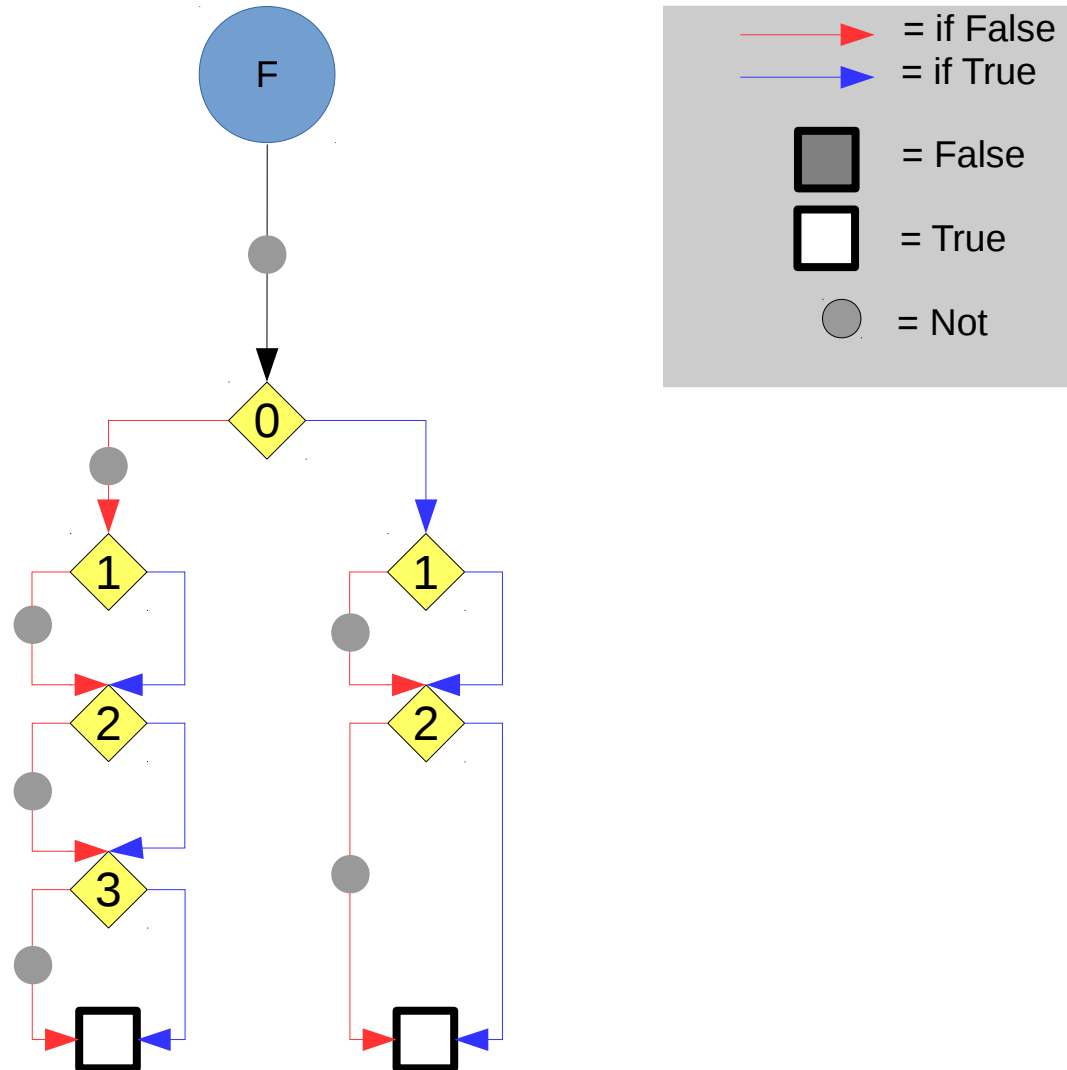
we merge isomorphic sub-graphs



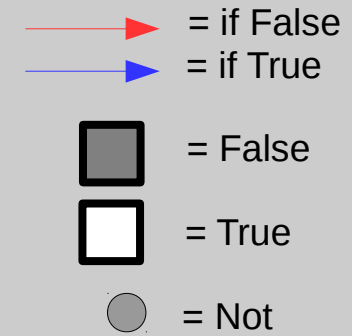
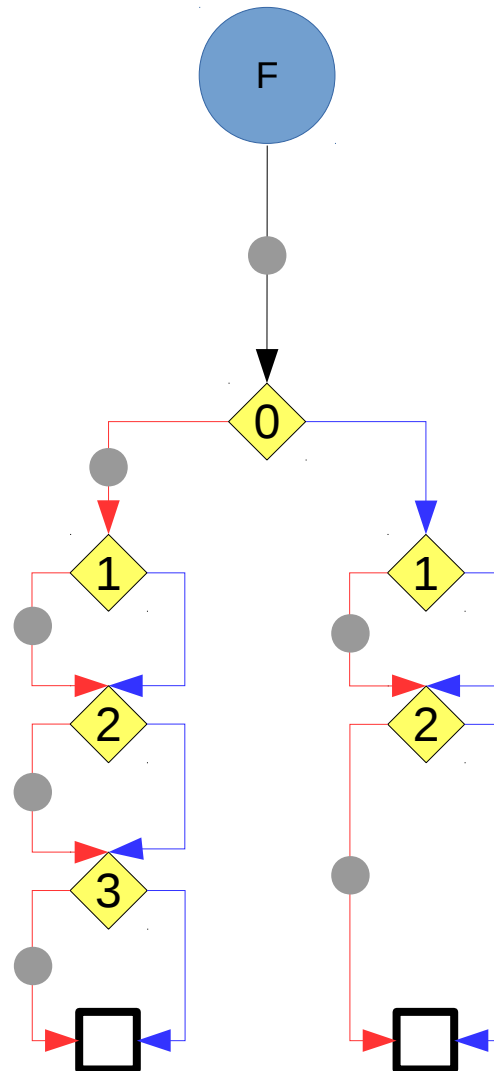
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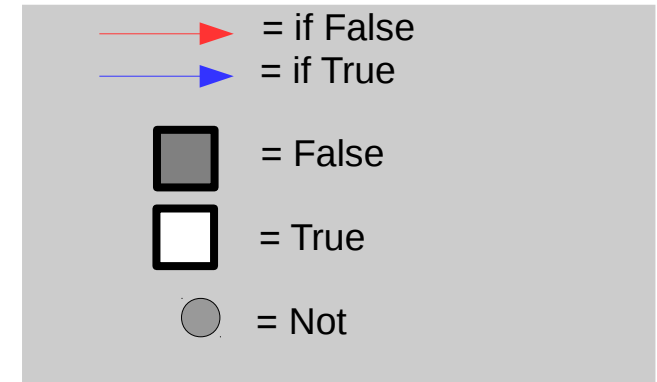
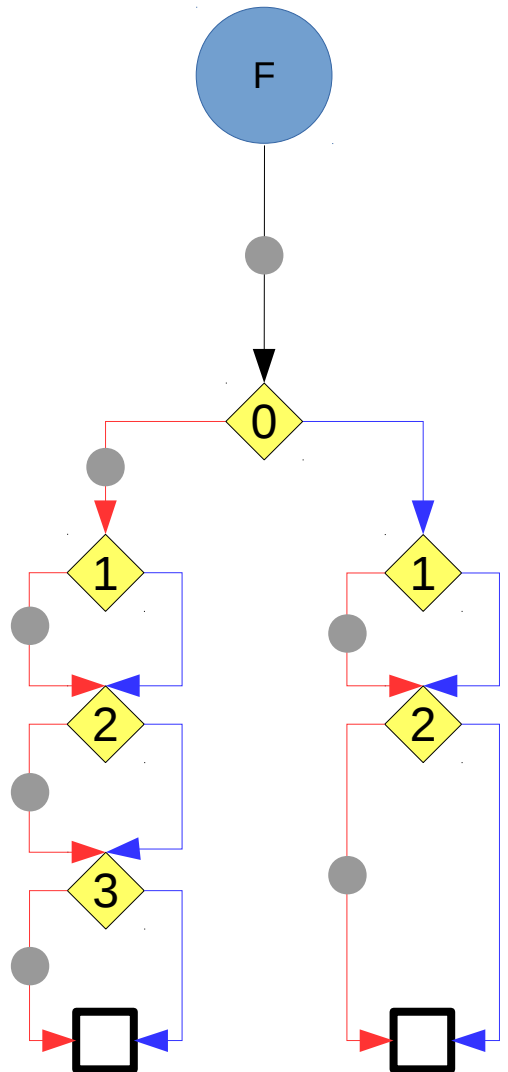


we merge isomorphic sub-graphs



Augmenting edges with  
negation can be  
performed in linear time in #node

## State Of The Art since 2000s



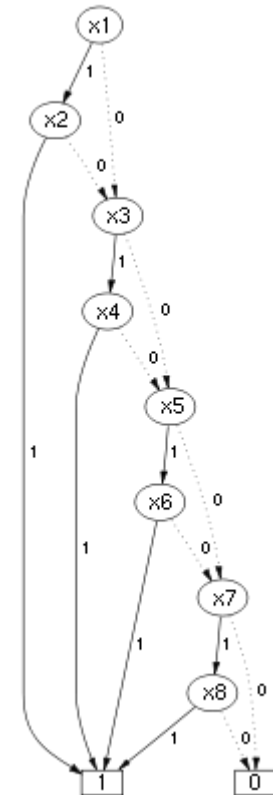
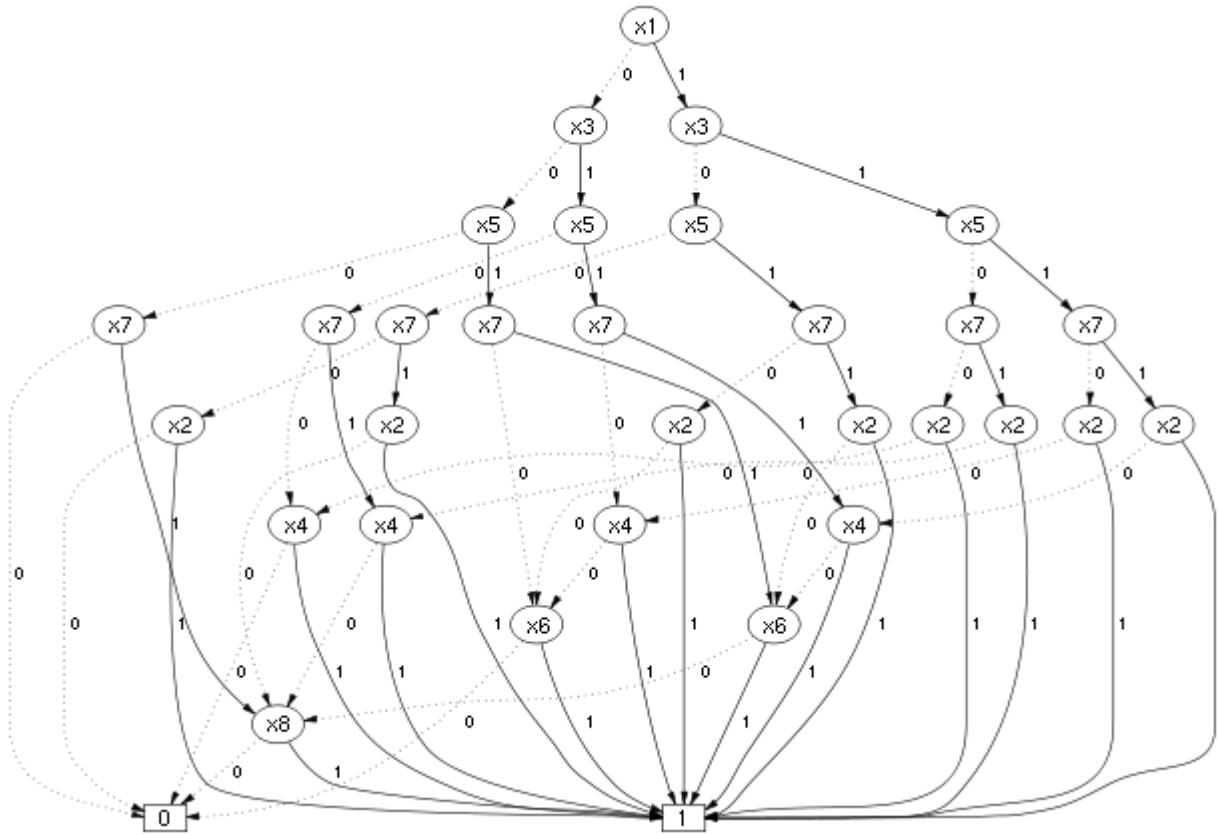
# Reduced Ordered BDD

- $=)$  :
  - SAT : constant time
  - Any/Max/Min SAT : linear time (#variable)
  - #SAT : linear time (#node)
  - NOT : constant time
- $=()$  :
  - AND, XOR : quadratic time/space (#node)
  - #node is order dependent



# #node is order dependent

$$(x_1 \wedge x_2) \vee (x_3 \wedge x_4) \vee (x_5 \wedge x_6) \vee (x_7 \wedge x_8)$$

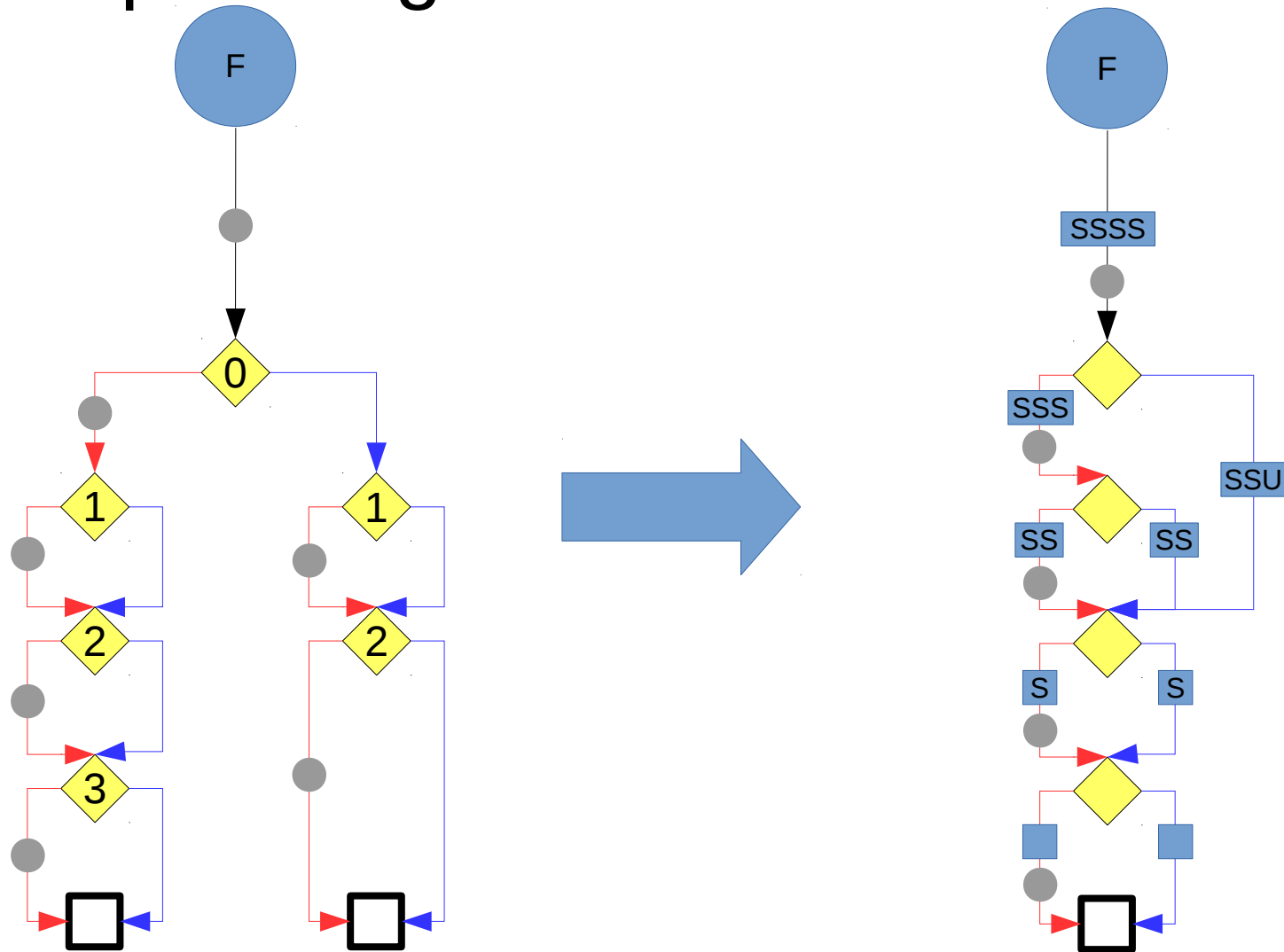


# Objective

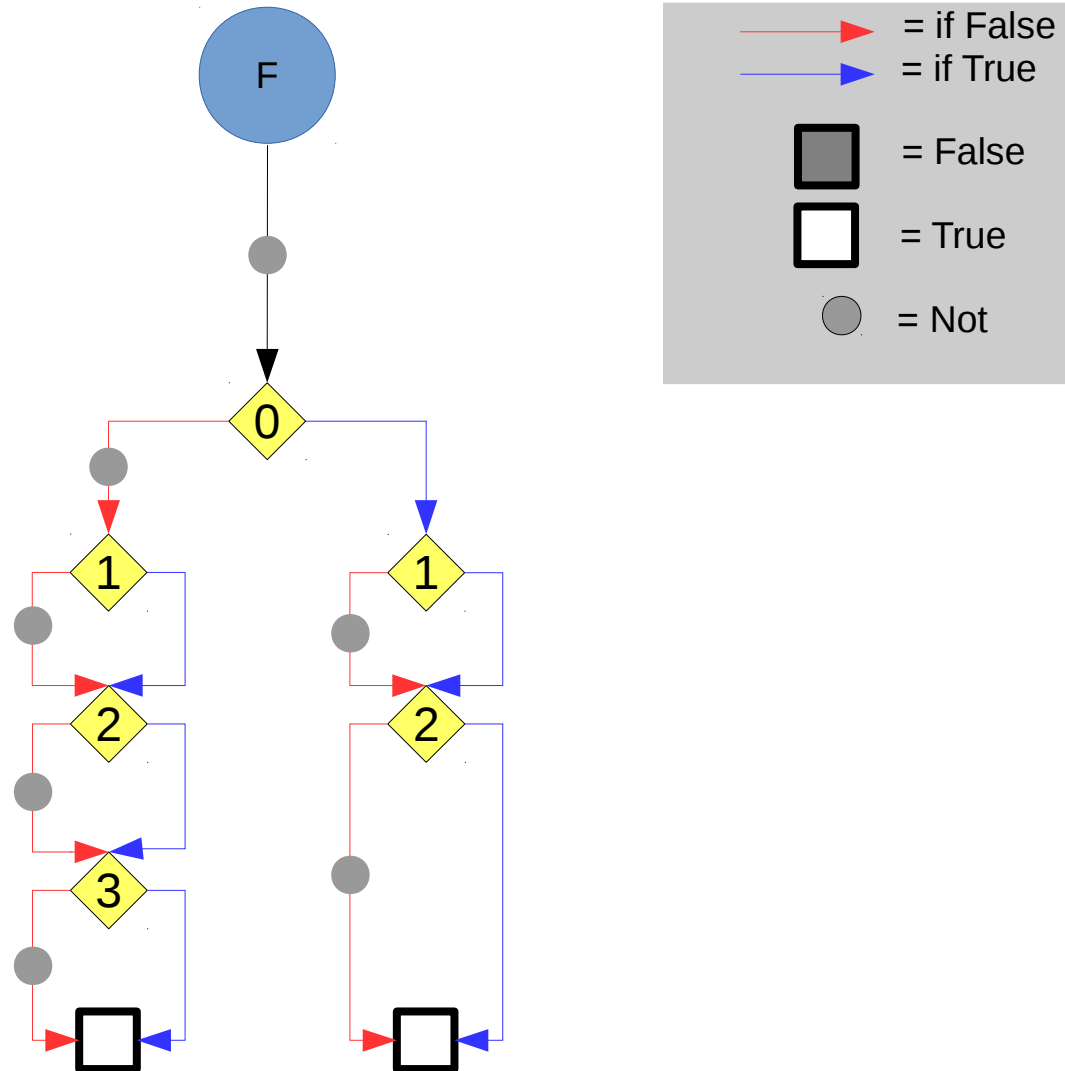
## Reduce #node

# Section 2

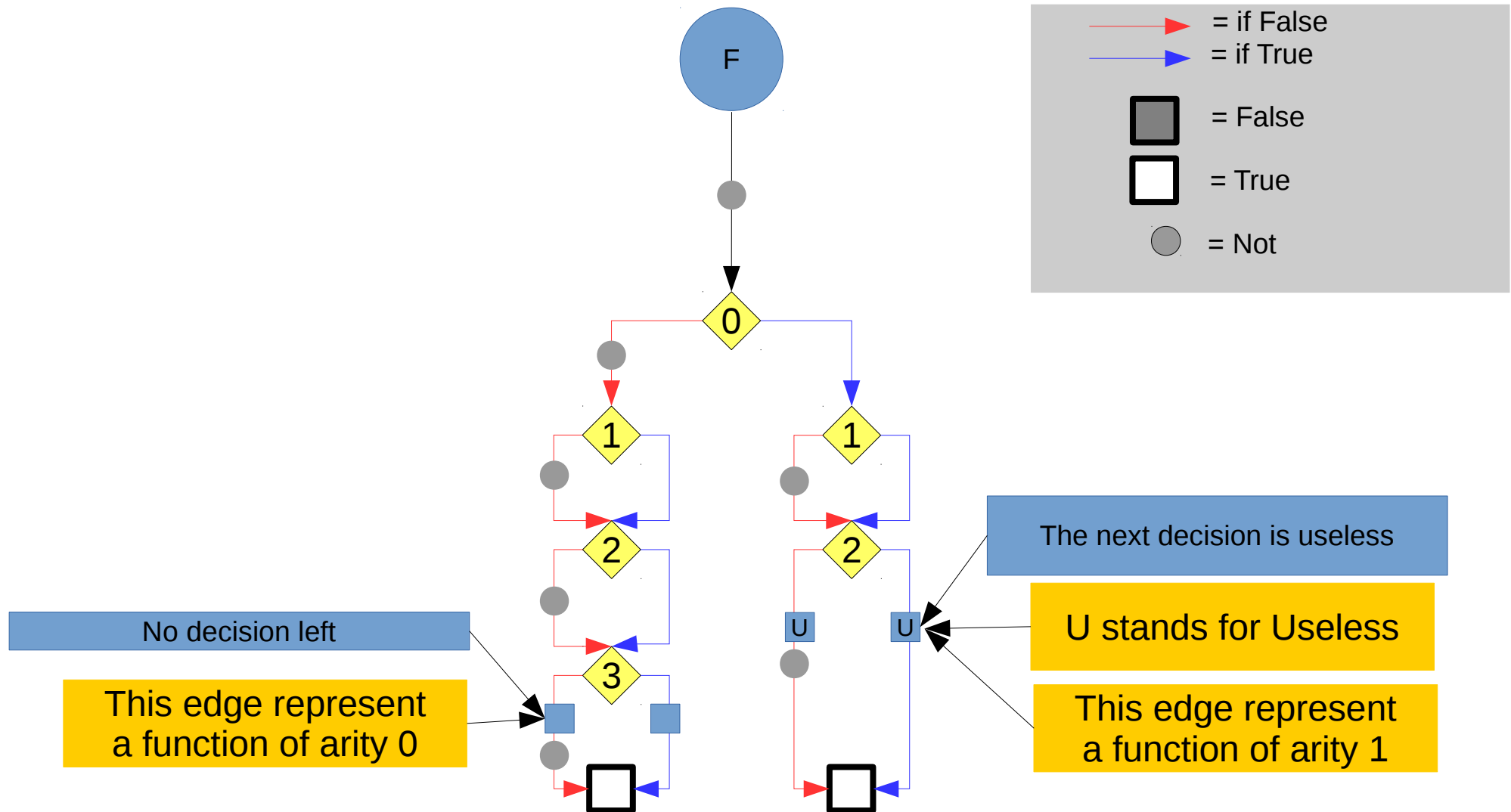
## Compressing a ROBDD into a GroBdd



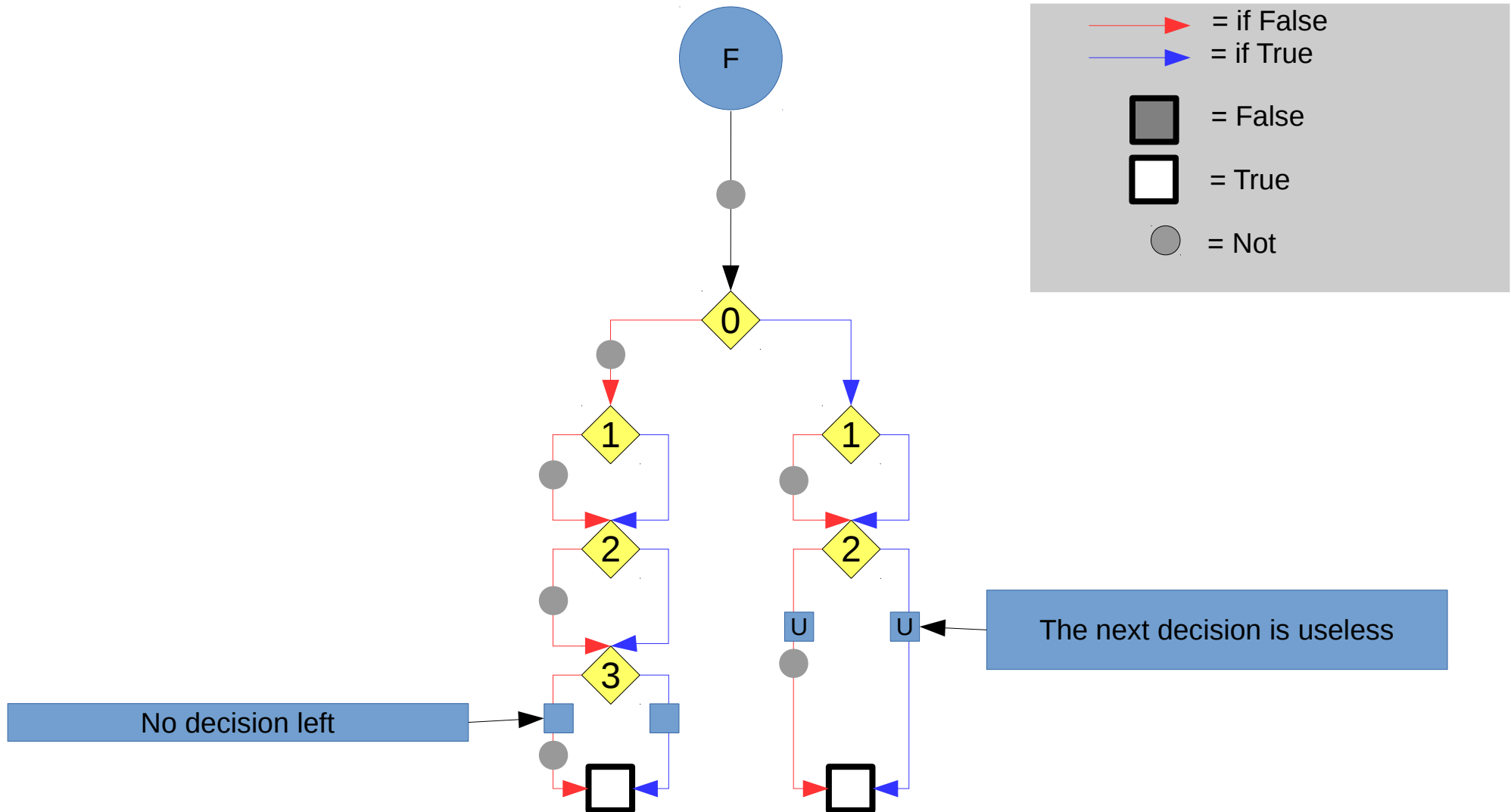
(Model 1) Step 1: for terminal leading edges, we unary represent the number of useless decisions



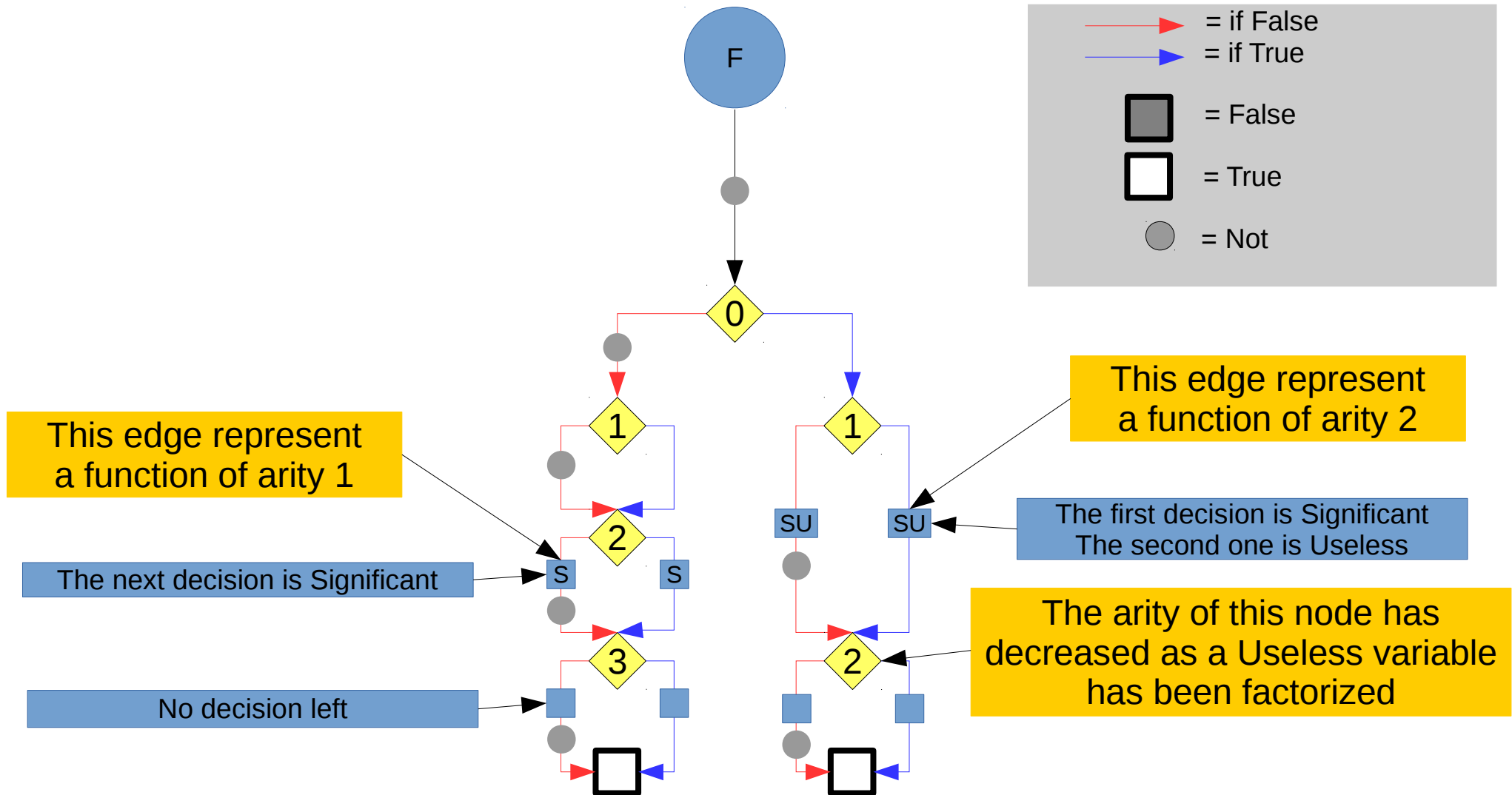
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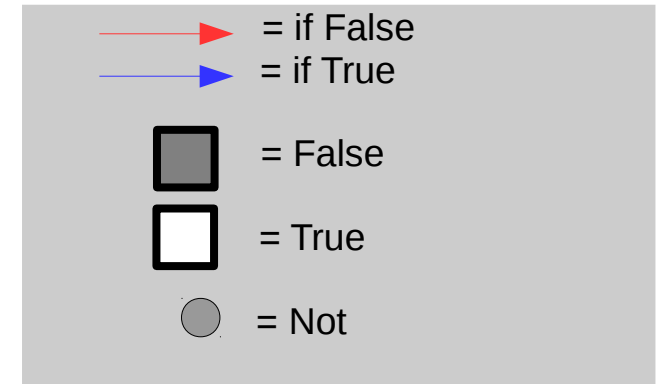
(Model 1) Step 2: we factorize useless variables



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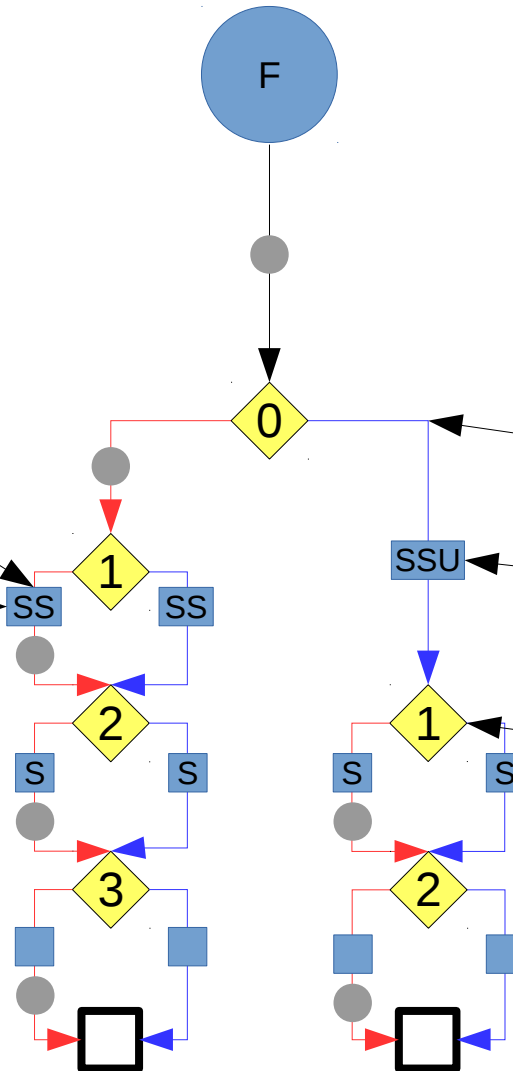
This edge represent a function of arity 2

This edge represent a function of arity 3

The two next decisions are Significant

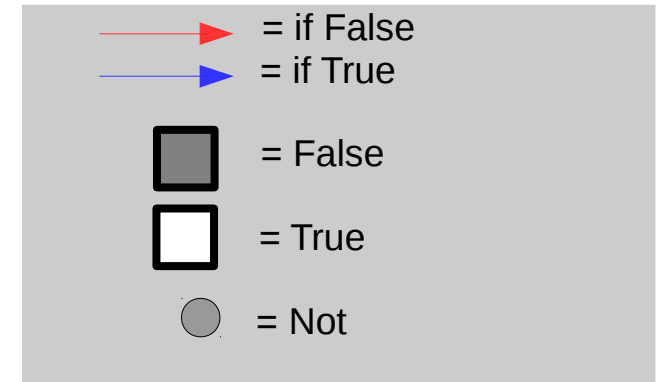
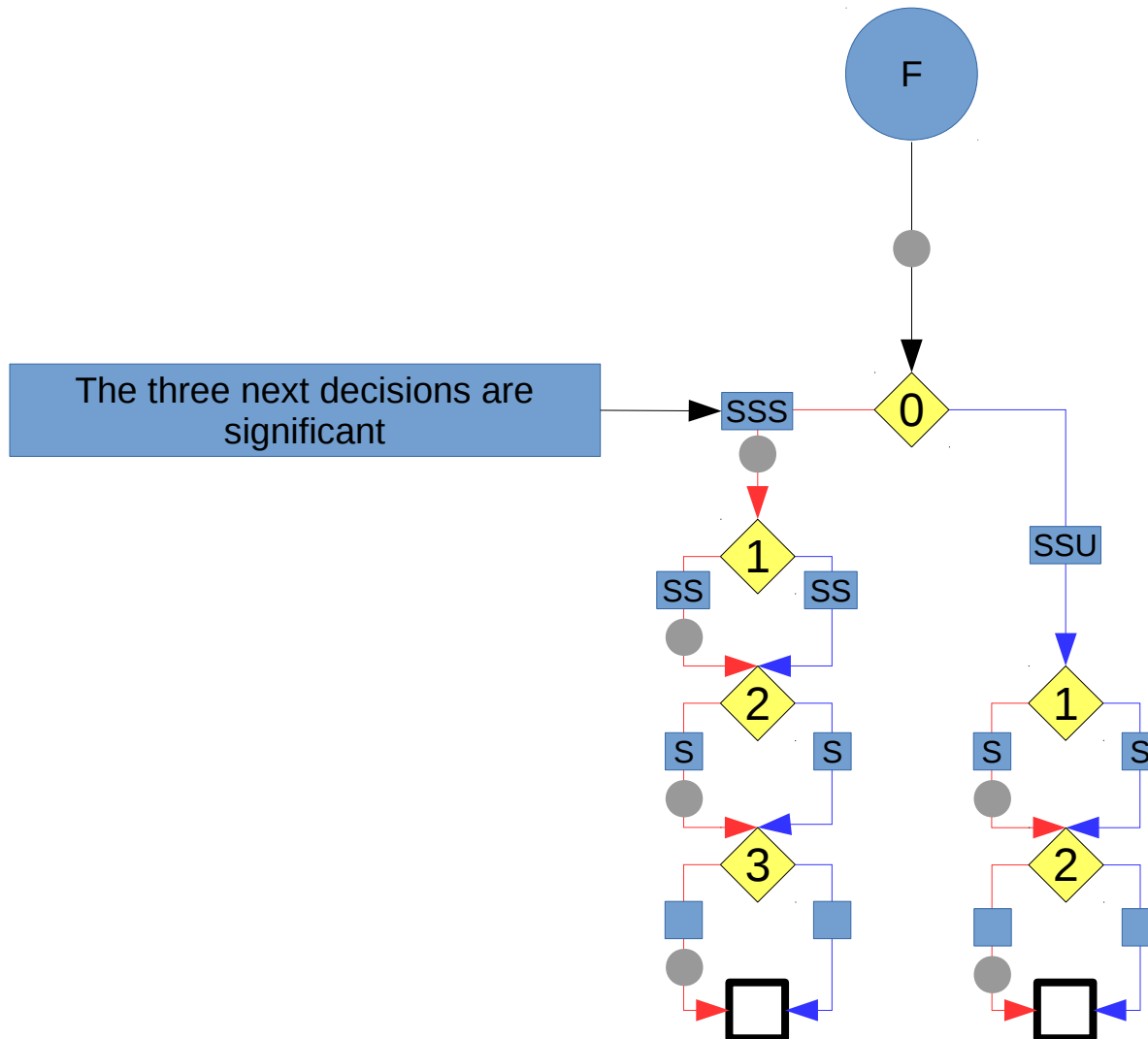
The two next decisions are significant, the third one is useless

The arity of this node has decreased as a Useless variable has been factorized

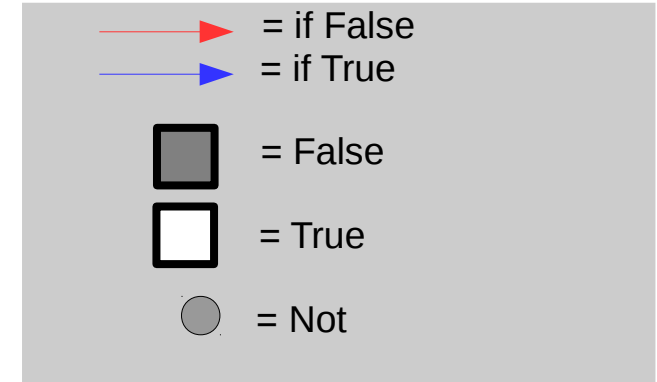
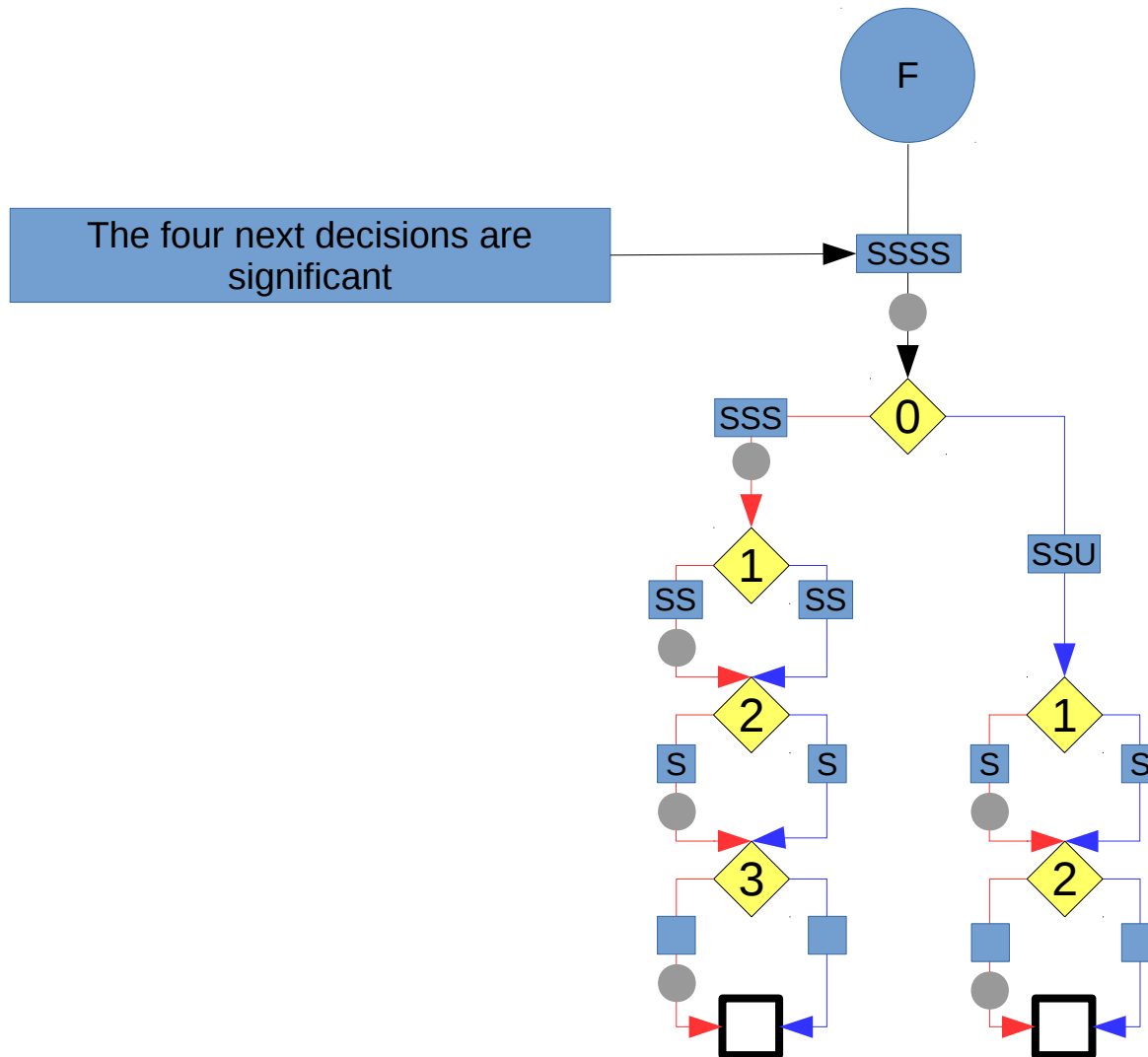




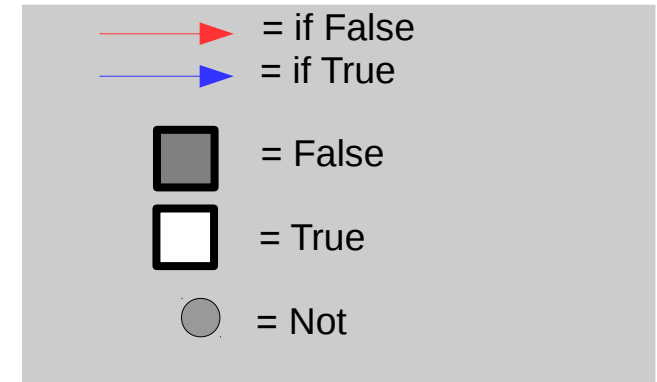
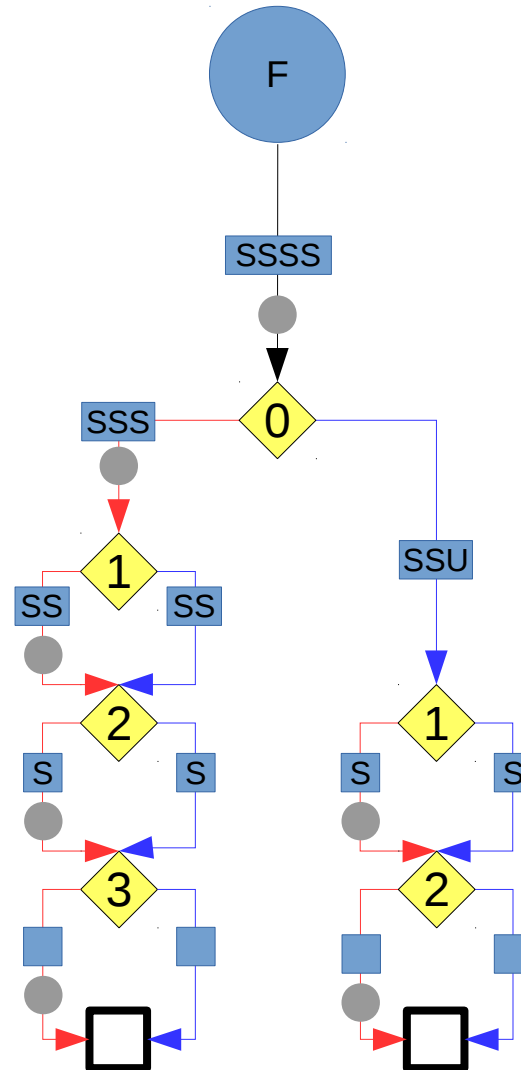
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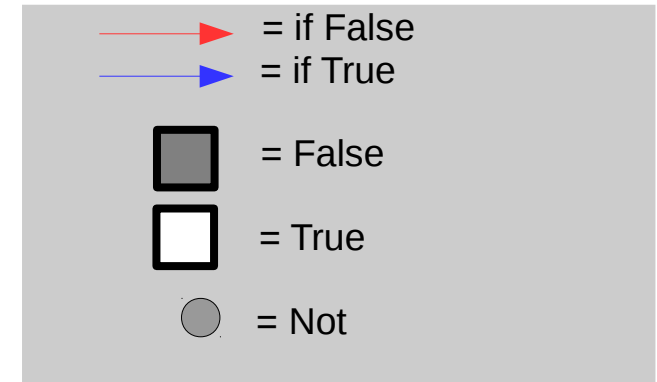
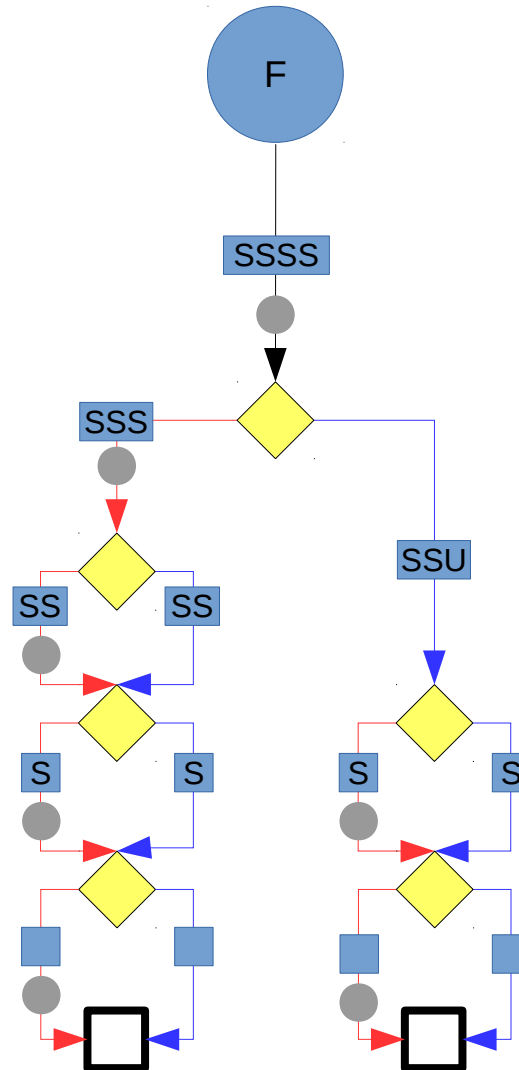
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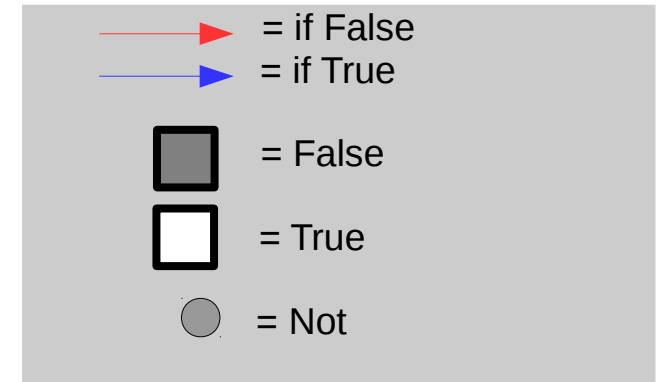
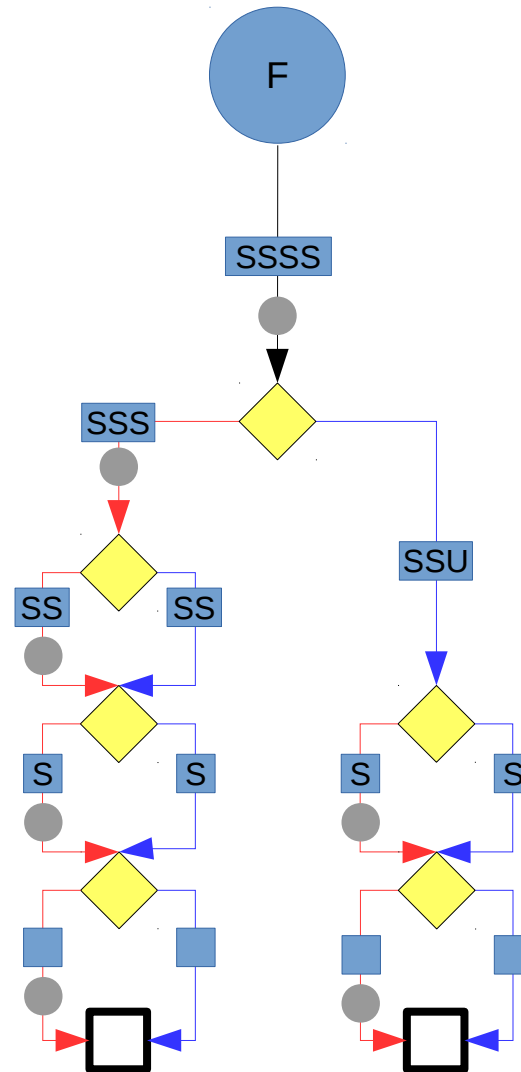
(Model 1) Step 3: we forget every node's depth



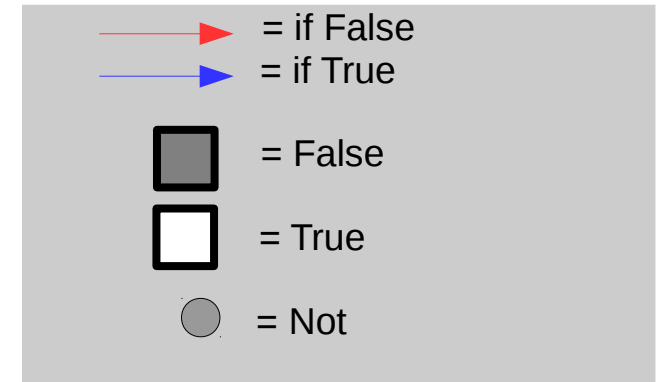
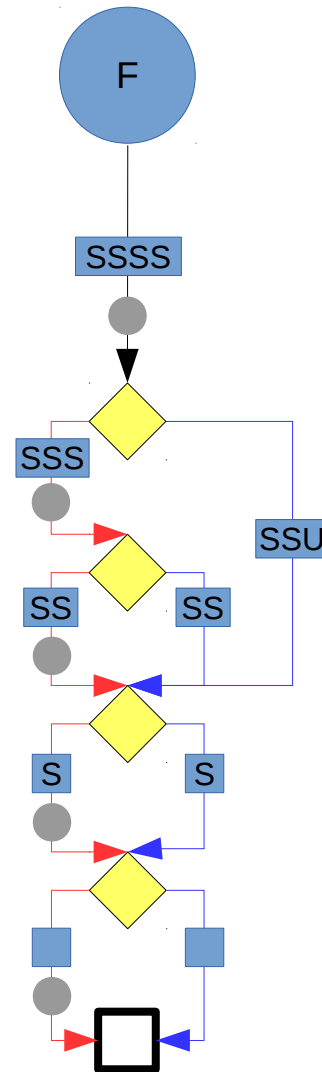
(Model 1) Step 3: we forget every node's decision variable



we merge isomorphic sub-graphs



we merge isomorphic sub-graphs

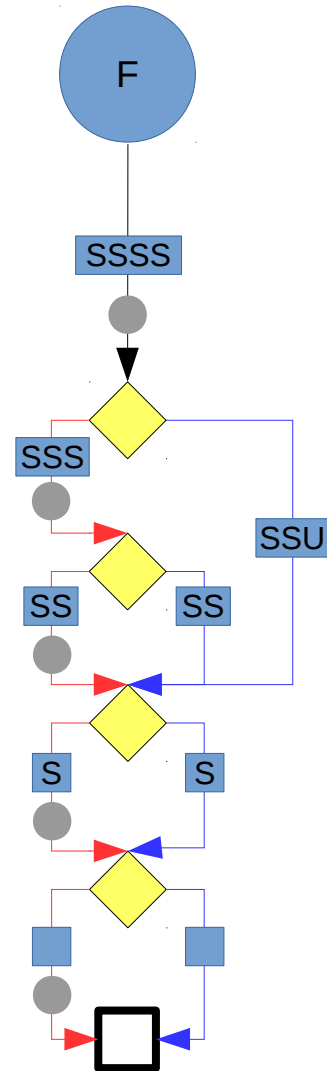


# Compiling a formula into a GroBdd

$$f(x_0^3) = x_1 \oplus x_2 \oplus (\neg x_0 \wedge x_3)$$

Represents a vector of four elements:

$$X_0 = (x_0, x_1, x_2, x_3)$$



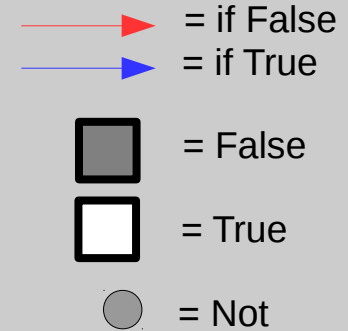
How to compile a formula into a GroBdd

$$f(x_0^3) = x_1 \oplus x_2 \oplus (\neg x_0 \wedge x_3)$$

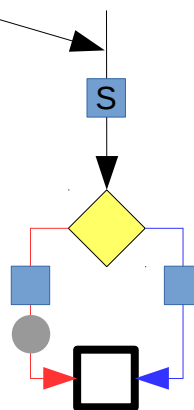


## How to compile a formula into a GroBdd

$$f(x_0^3) = x_1 \oplus x_2 \oplus (\neg x_0 \wedge x_3)$$

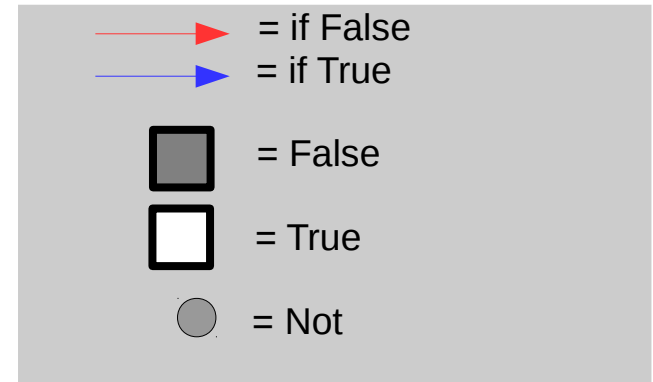


Step 1: we build  
the identity function



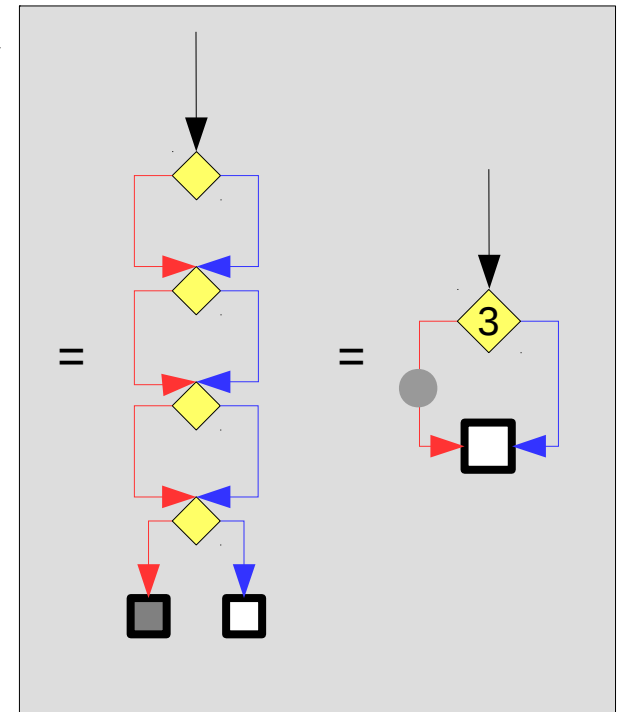
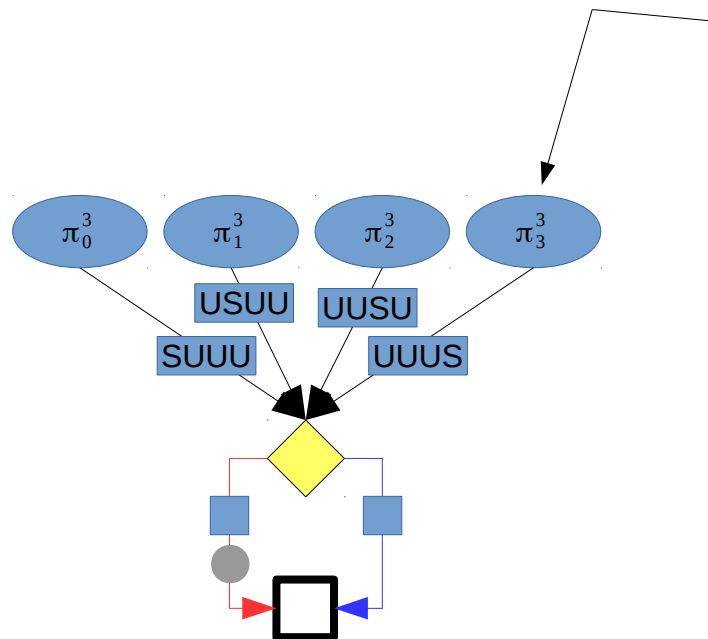
## How to compile a formula into a GroBdd

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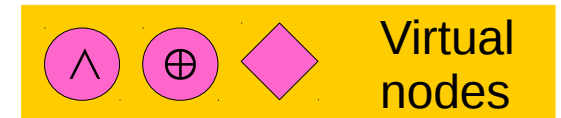
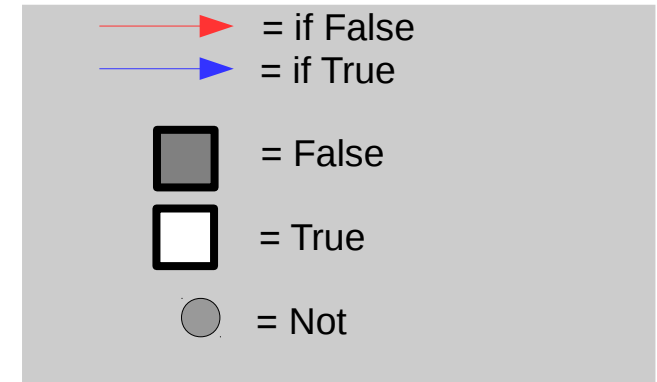
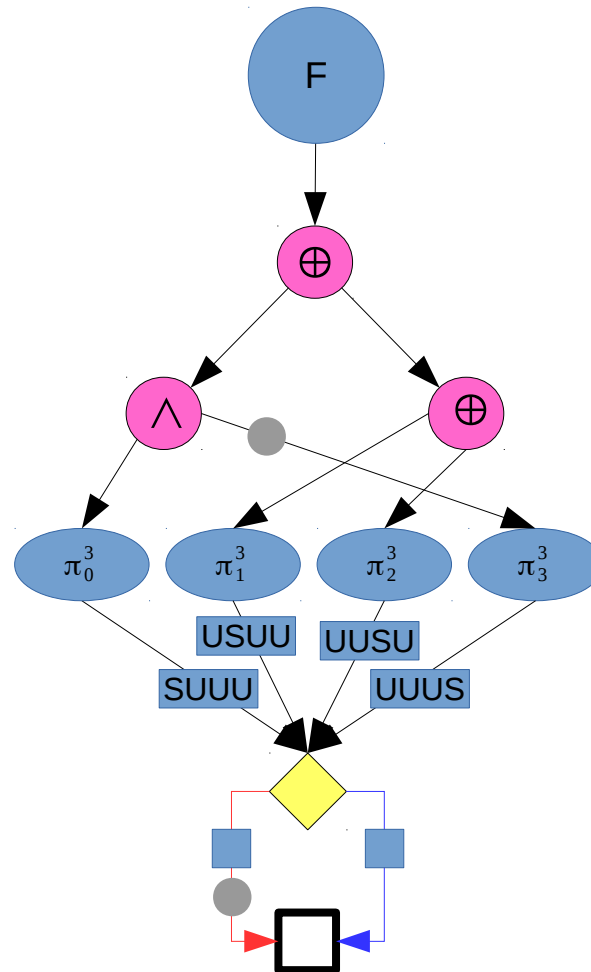
Step 2: we build one projection per variable

$$\pi_k^n(x_0^n) = x_k$$



## How to compile a formula into a GroBdd

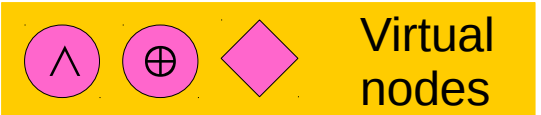
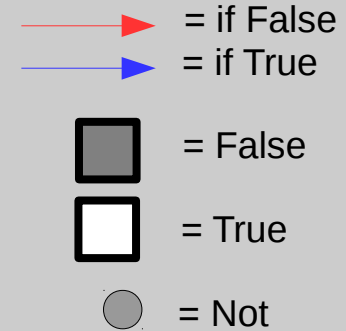
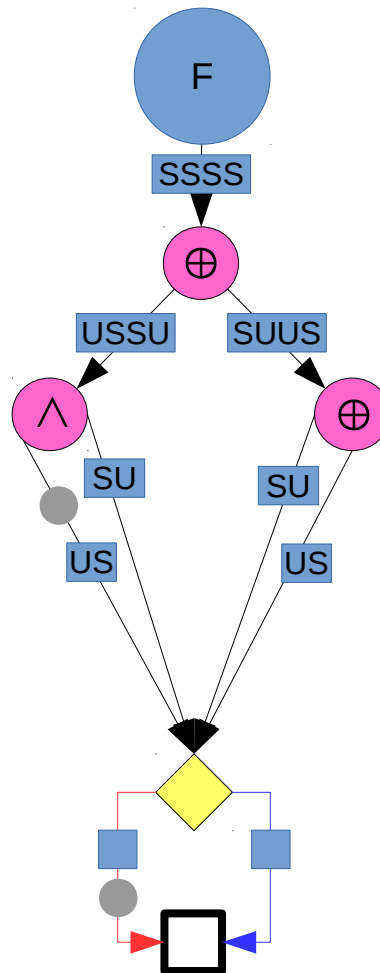
$$f(x_0^3) = x_1 \oplus x_2 \oplus (\neg x_0 \wedge x_3)$$



Step 3: we build the formula

## How to compile a formula into a GroBdd

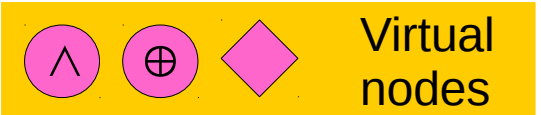
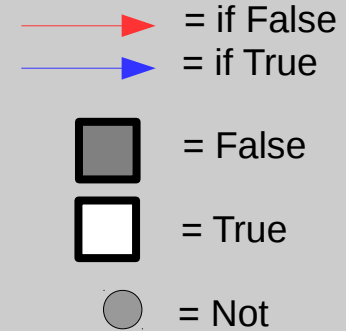
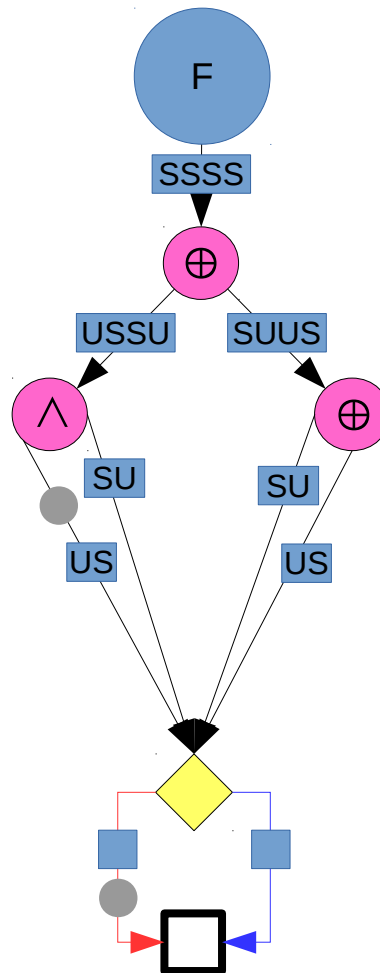
$$f(x_0^3) = x_1 \oplus x_2 \oplus (\neg x_0 \wedge x_3)$$



Step 4: we factorize  
useless variables

## How to compile a formula into a GroBdd

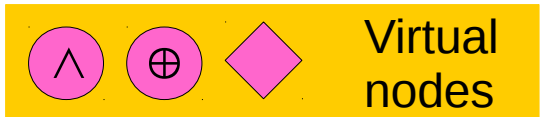
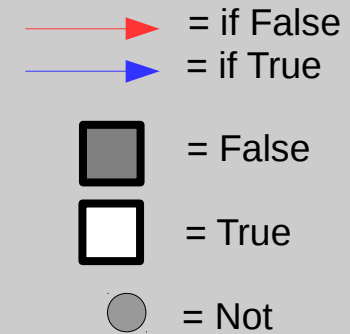
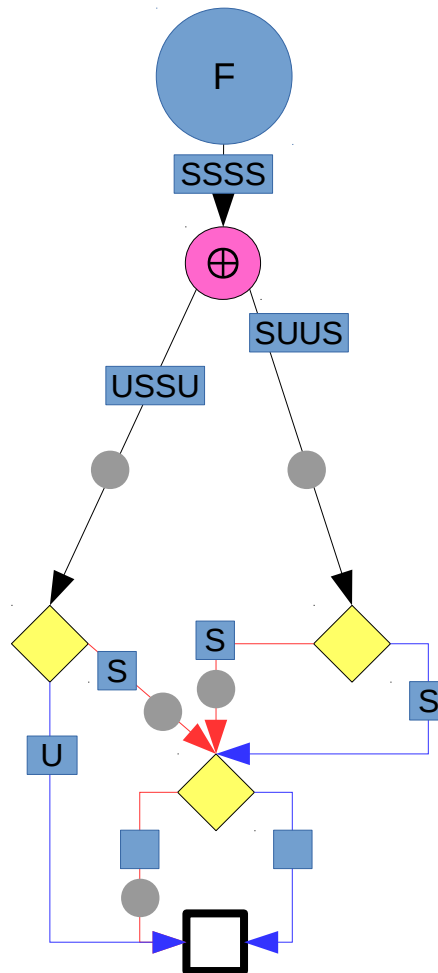
$$f(x_0^3) = x_1 \oplus x_2 \oplus (\neg x_0 \wedge x_3)$$



Step 5: we compute operator nodes

## How to compile a formula into a GroBdd

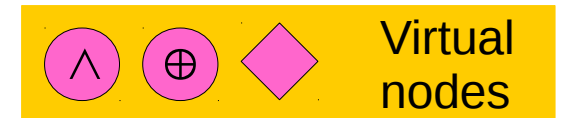
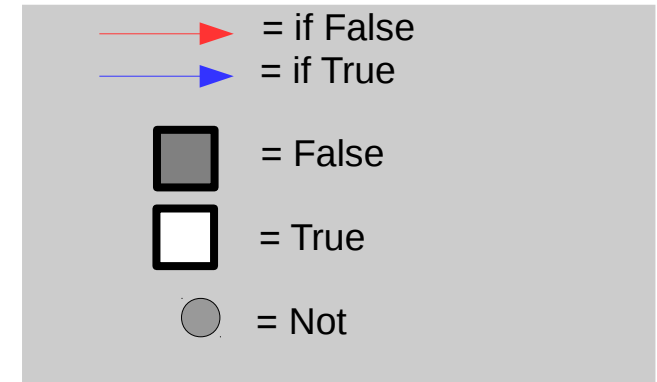
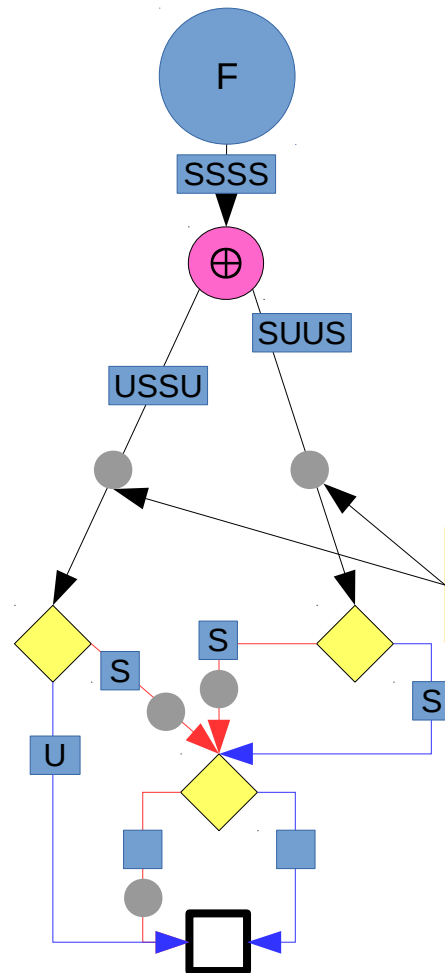
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Step 5: we compute operator nodes

## How to compile a formula into a GroBdd

$$f(x_0^3) = x_1 \oplus x_2 \oplus (\neg x_0 \wedge x_3)$$

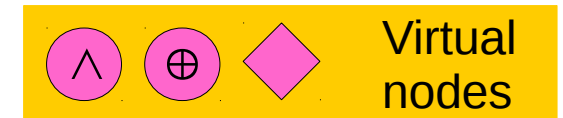
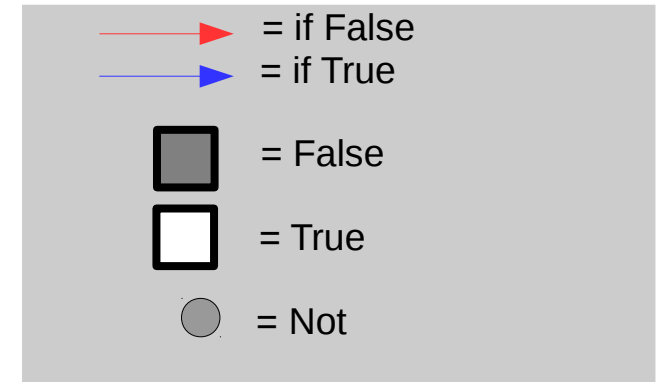
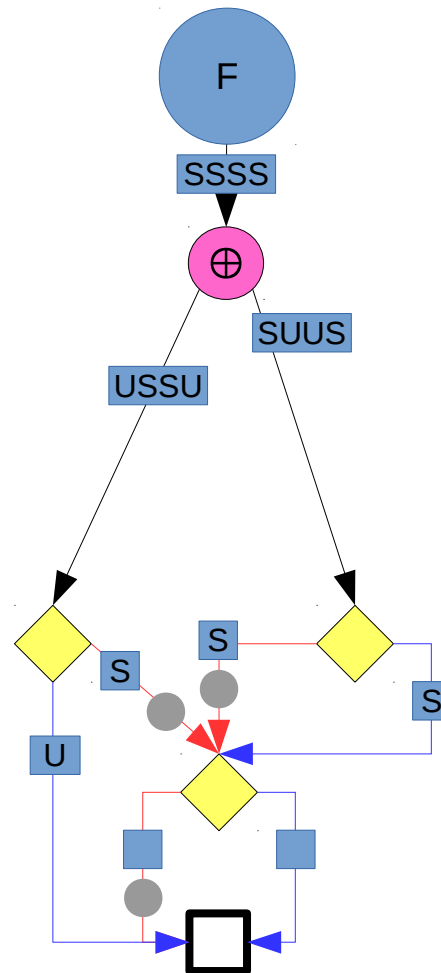


Step 5: we compute operator nodes

$$\neg f \oplus \neg g = f \oplus g$$

## How to compile a formula into a GroBdd

$$f(x_0^3) = x_1 \oplus x_2 \oplus (\neg x_0 \wedge x_3)$$

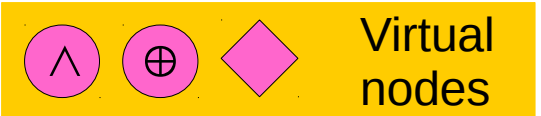
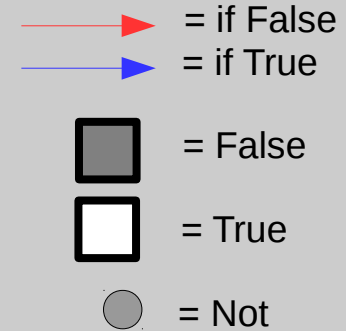
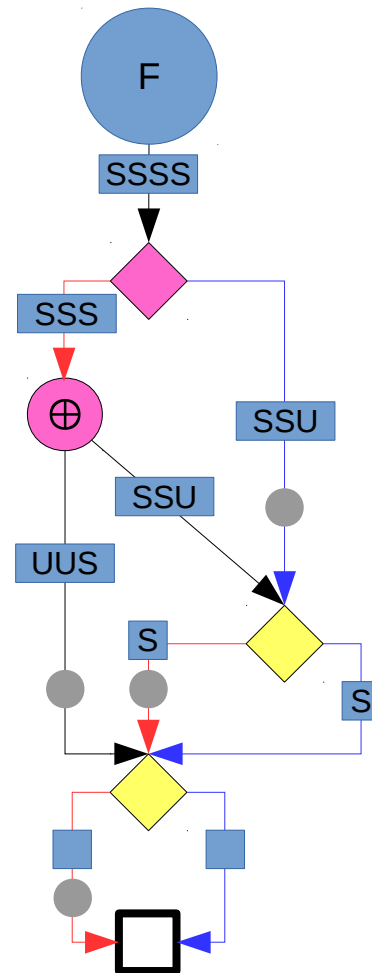


Step 5: we compute operator nodes



## How to compile a formula into a GroBdd

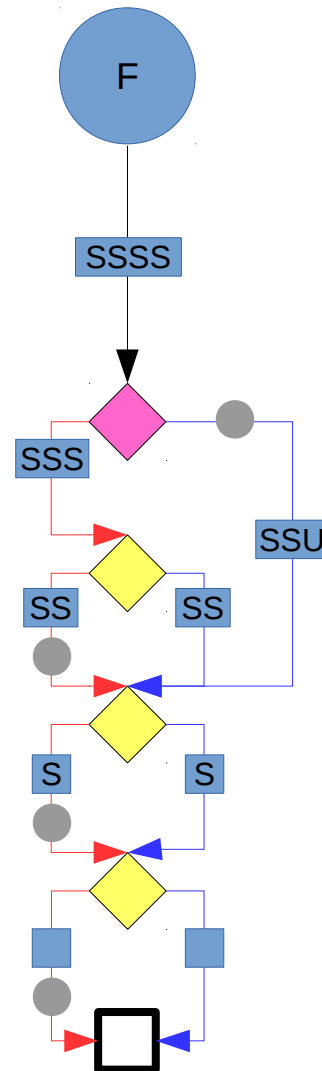
$$f(x_0^3) = x_1 \oplus x_2 \oplus (\neg x_0 \wedge x_3)$$



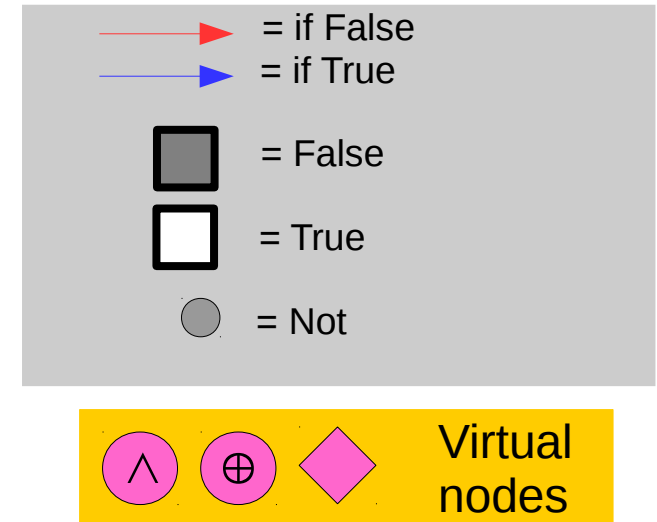
Step 5: we compute operator nodes

## How to compile a formula into a GroBdd

$$f(x_0^3) = x_1 \oplus x_2 \oplus (\neg x_0 \wedge x_3)$$

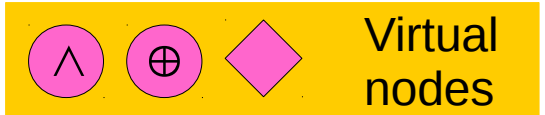
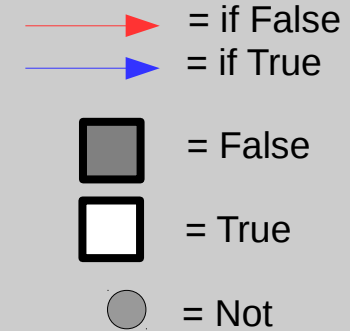
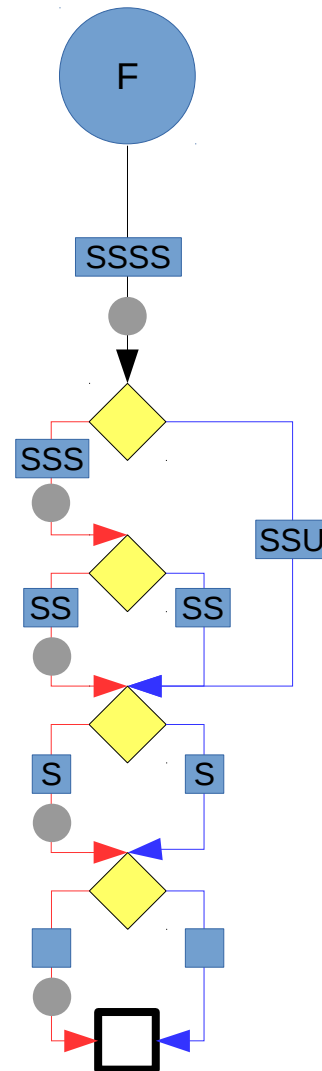


Step 5: we compute operator nodes



## How to compile a formula into a GroBdd

$$f(x_0^3) = x_1 \oplus x_2 \oplus (\neg x_0 \wedge x_3)$$



Step 5: we compute operator nodes

# Section 4

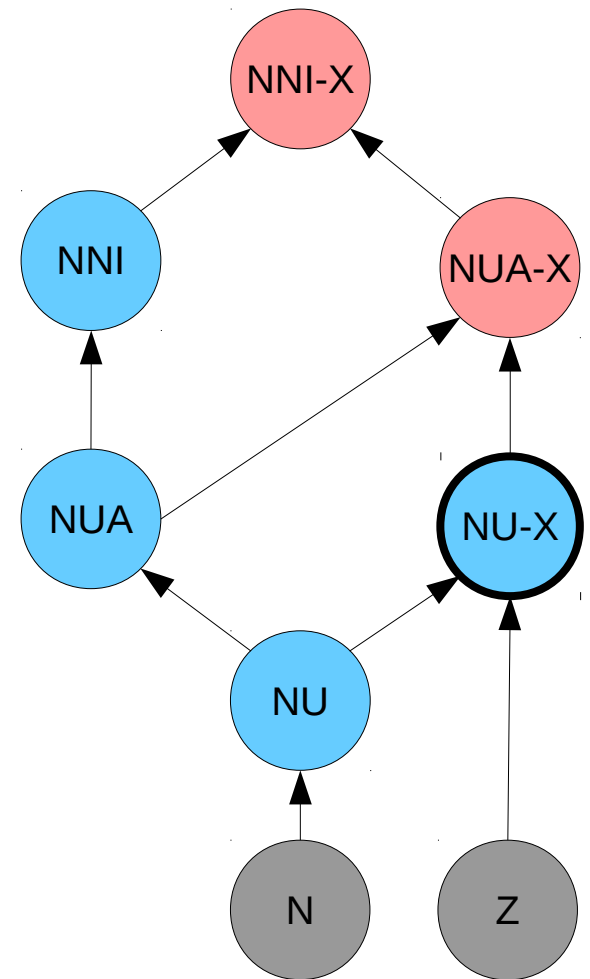
## Results

Average reduction of the number of nodes in four benchmarks

	GroBdd vs RoBdd
arithmetic	-40,43%
mcnc	-14,74%
iscas99	-25,47%
satlib/uf20-91	-2,86%

# Conclusion

- Software implemented in OCaml:
  - <https://github.com/JoanThibault/DAGaml/tree/grobdd-dev>
  - ~ 10 000 lines of OCaml
- Fewer nodes
  - NU : -0.35 d (-55%)
  - NNI : -0.51 d (-69%)
  - NU-X : -0.13 d (-26%)
- Future Work
  - Quantify the dependency between variables' order and #node
  - Solve & Implement NUA-X and NNI-X versions
- TO DO
  - Parallelism & hardware acceleration
  - Quantification Operators
  - Variable Reordering



# Extended Results

	Z	NU	NNI	NU-X
arithmetic	44,10%	-40,43%	-74,17%	-55,69%
mcnc	129,66%	-14,74%	-46,86%	-51,79%
iscas99	162,20%	-25,47%	-56,47%	-55,01%
satlib/uf20-91	-41,02%	-2,86%	-29,50%	-93,00%
(log10)	Z	NU	NNI	NU-X
arithmetic	0,15	-0,44	-1,00	-0,60
mcnc	0,09	-0,07	-0,42	-0,35
iscas99	0,36	-0,15	-0,49	-0,40
satlib/uf20-91	-0,24	-0,01	-0,17	-1,19