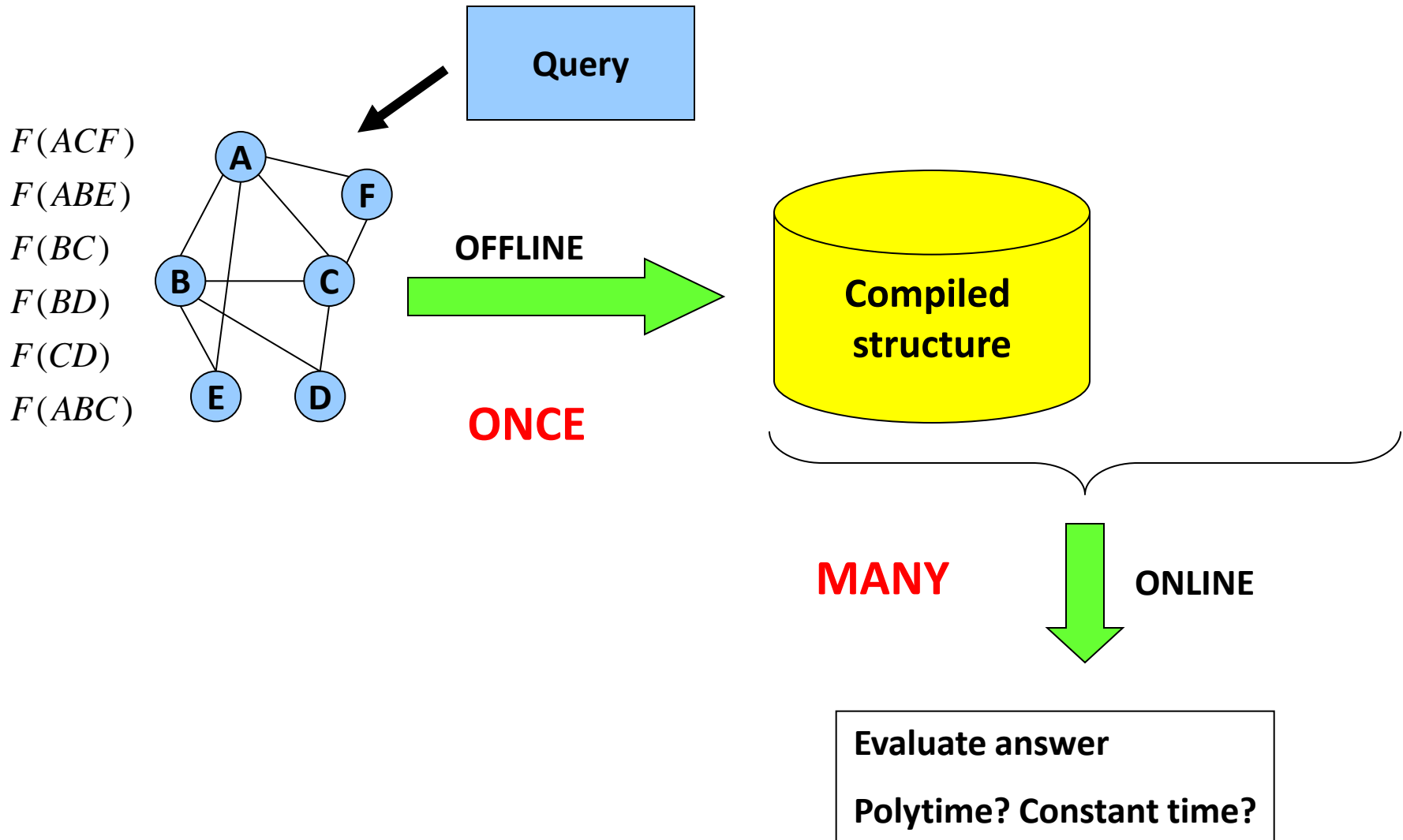




Introduction

- Combining two frameworks
 - AND/OR Search Spaces
 - Multi-valued Decision Diagrams (MDDs)
- Both are more compact ways to represent problems.
- Their combination yields an even more compact representation.
- Decision Diagrams are known to allow online speed queries.

Introduction

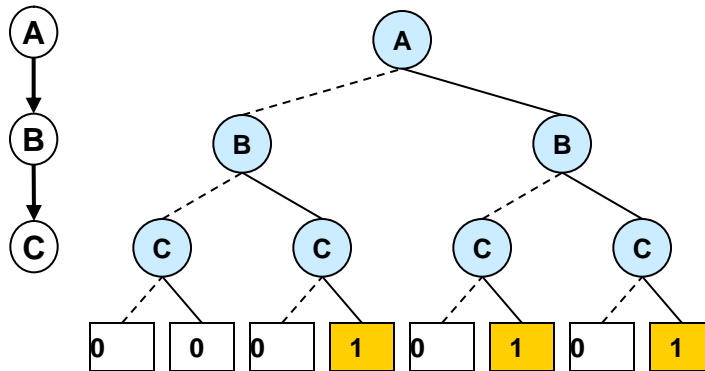


Ordered Binary Decision Diagram

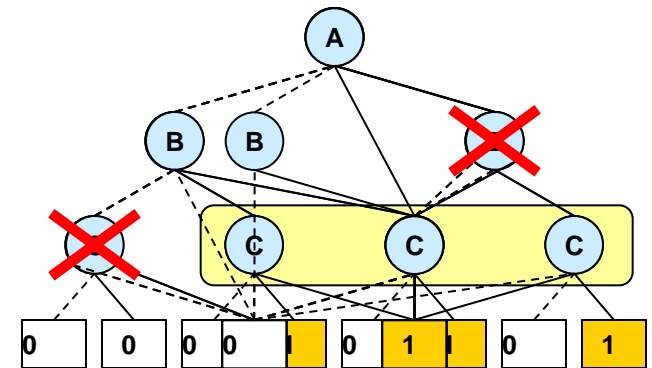
$$B = \{0,1\} \quad f : B^3 \rightarrow B$$

A	B	C	f(ABC)
0	0	0	0
0	0	1	0
0	1	0	0
0	1	1	1
1	0	0	0
1	0	1	1
1	1	0	0
1	1	1	1

Table



Decision tree

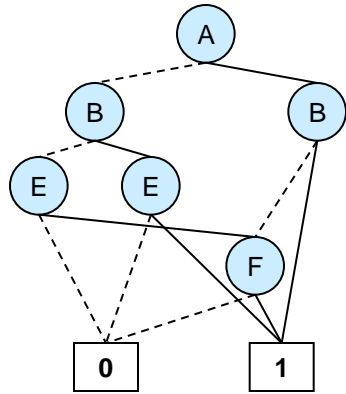


- 1) Merge identical children
[Bryant86]
- 2) Remove redundant nodes

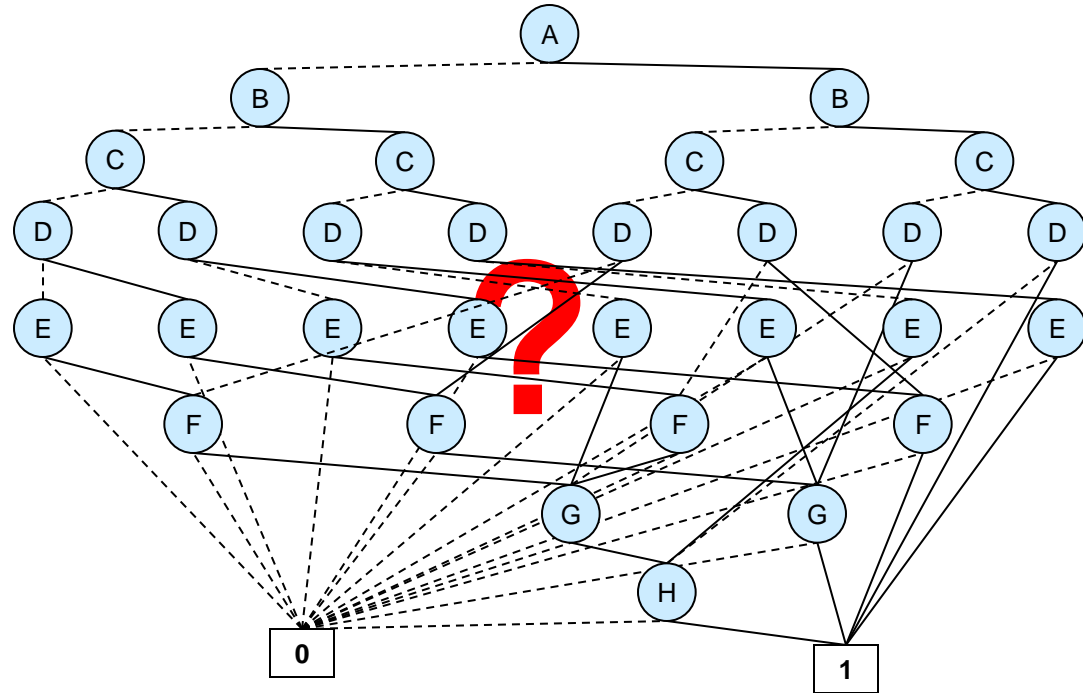
Ordering enables efficient operations

Decision Diagrams

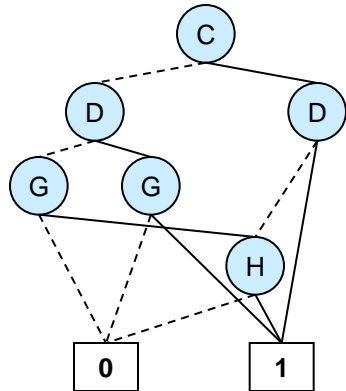
$$f = (A \vee E) \wedge (B \vee F)$$



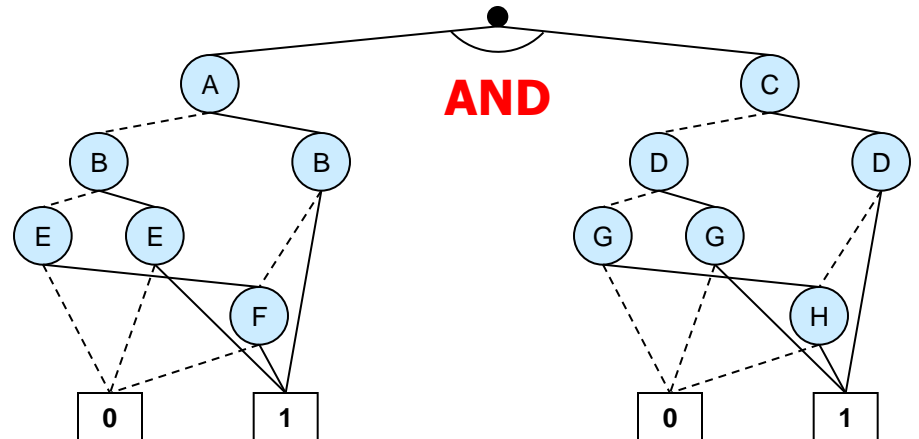
$$f \wedge g =$$



$$g = (C \vee G) \wedge (D \vee H)$$

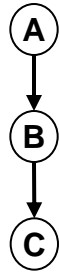


$$f \wedge g =$$



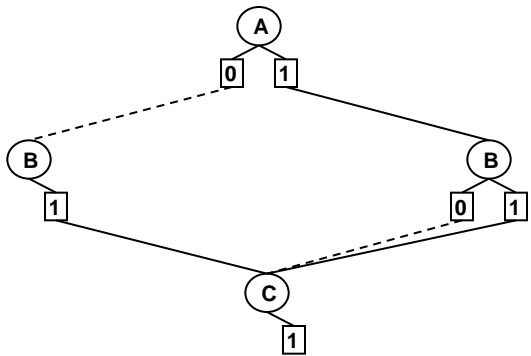
AOMDDs

A	B	C	f(ABC)
0	0	0	0
0	0	1	0
0	1	0	0
0	1	1	1
1	0	0	0
1	0	1	1
1	1	0	0
1	1	1	1

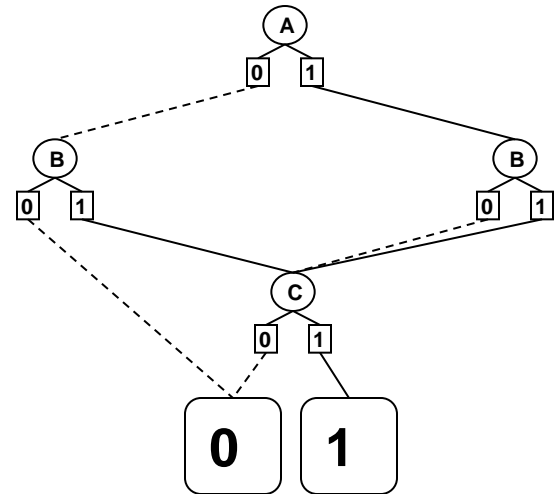
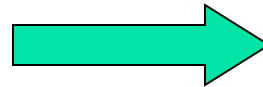


Point dead-ends to terminal node "0"

Point goods to terminal node "1"



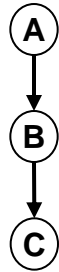
Minimal AND/OR graph



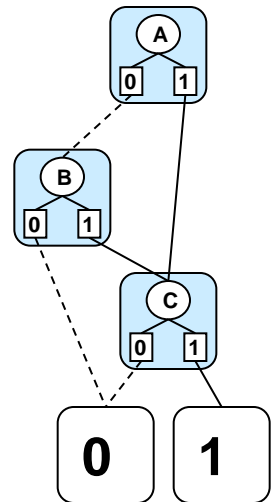
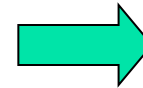
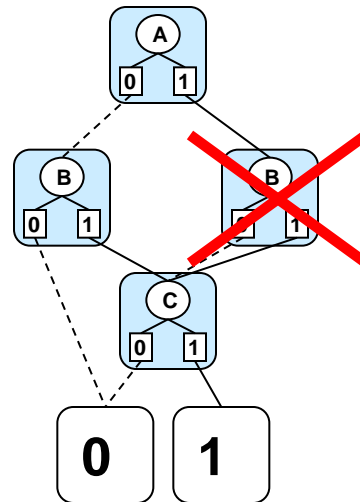
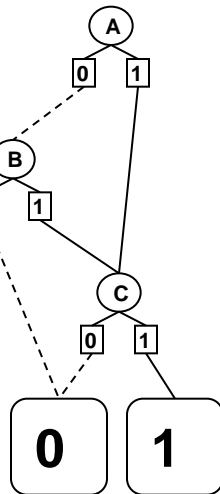
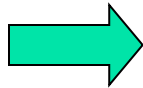
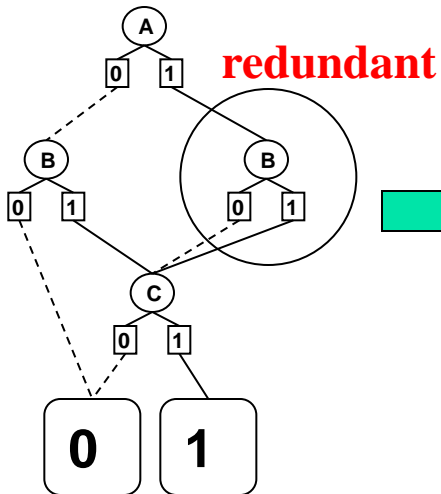
Decision Diagram

Removing Redundancy

A	B	C	f(ABC)
0	0	0	0
0	0	1	0
0	1	0	0
0	1	1	1
1	0	0	0
1	0	1	1
1	1	0	0
1	1	1	1



Group OR node together with its AND children into a meta-node



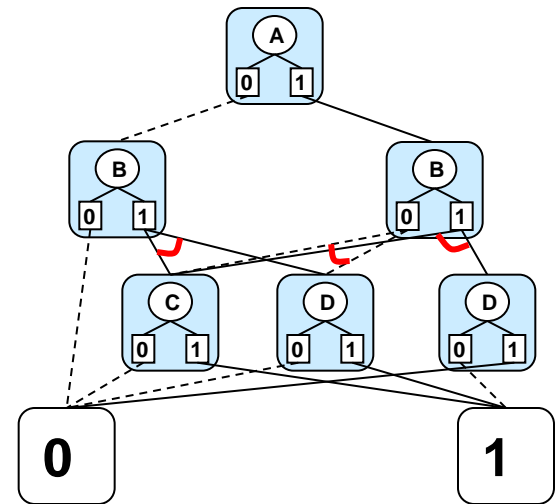
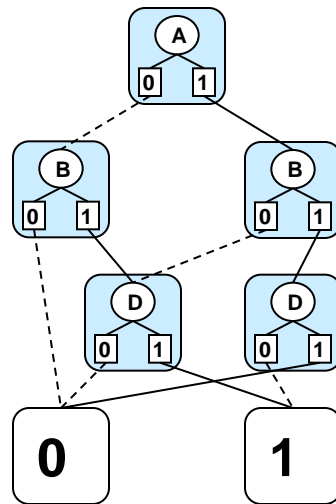
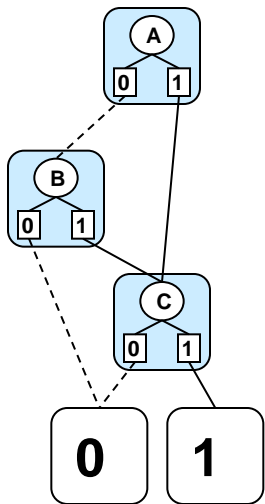
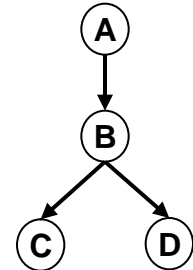
OBDD
(pseudo tree is a **chain**)

AOBDD

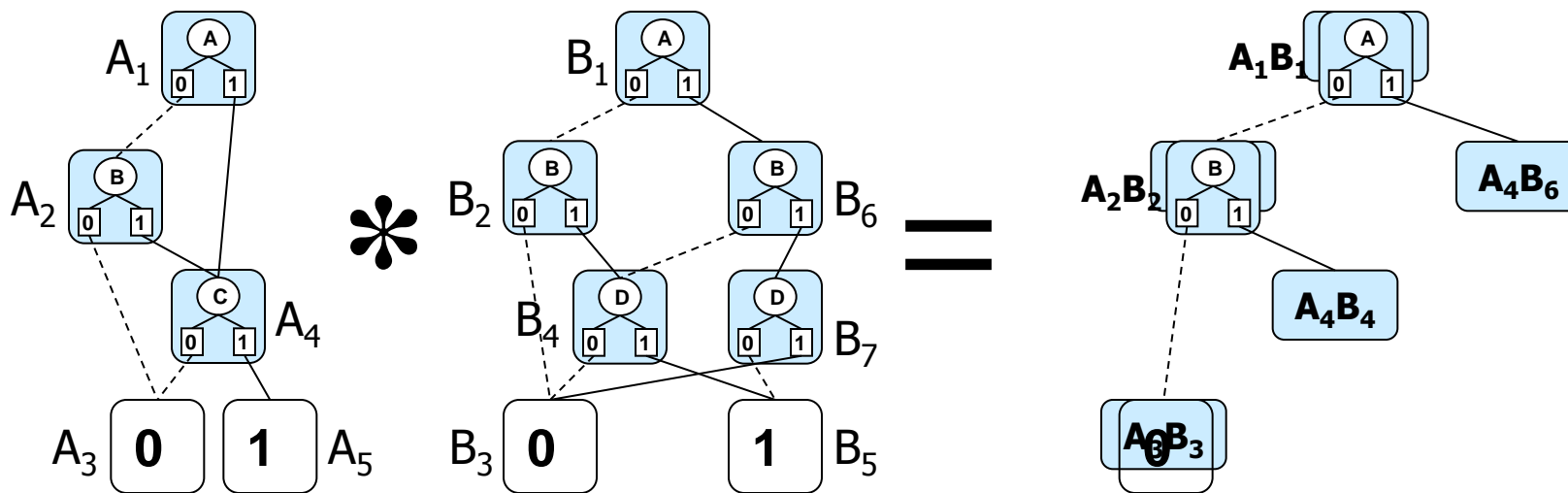
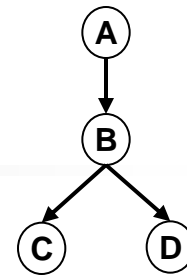
A	B	C	f(ABC)
0	0	0	0
0	0	1	0
0	1	0	0
0	1	1	1
1	0	0	0
1	0	1	1
1	1	0	0
1	1	1	1



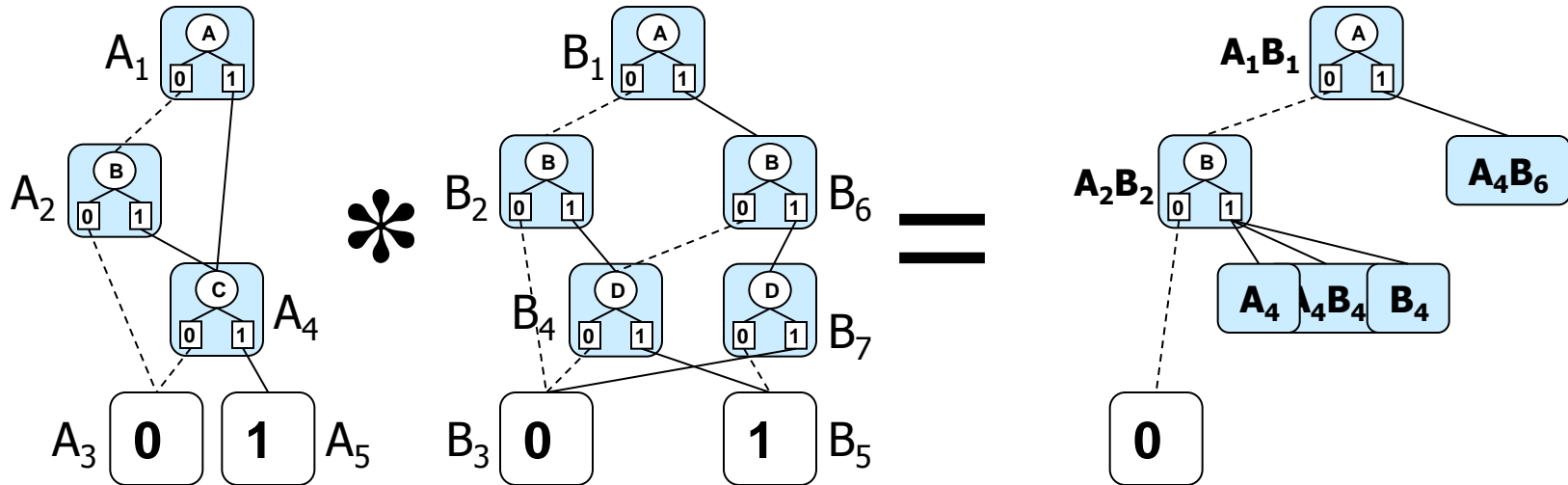
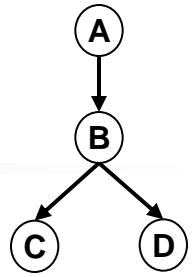
A	B	D	g(ABD)
0	0	0	0
0	0	1	0
0	1	0	0
0	1	1	1
1	0	0	0
1	0	1	1
1	1	0	1
1	1	1	0



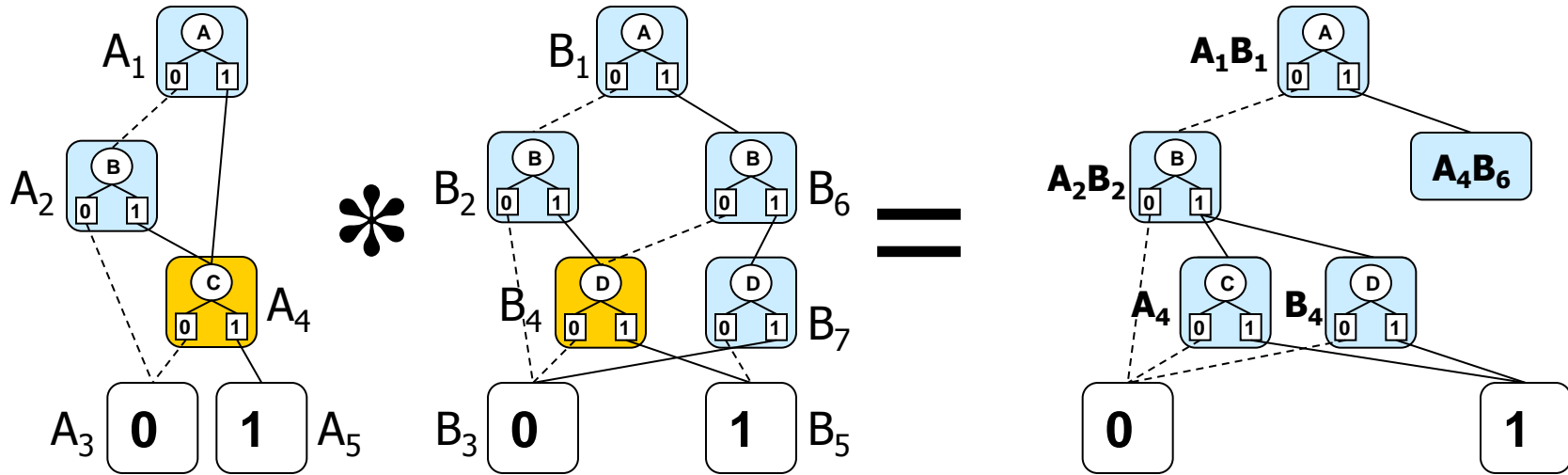
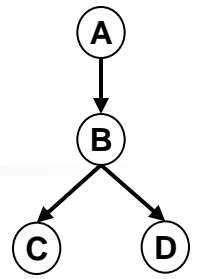
Apply Operator



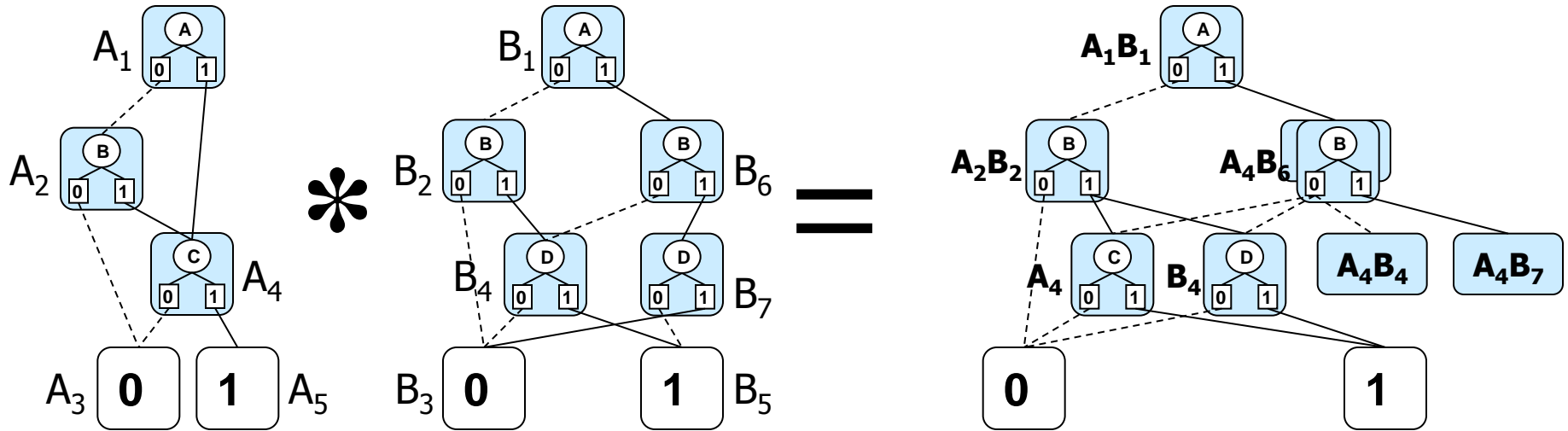
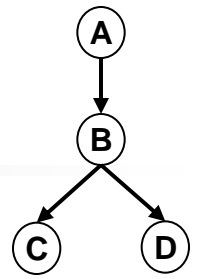
Apply Operator



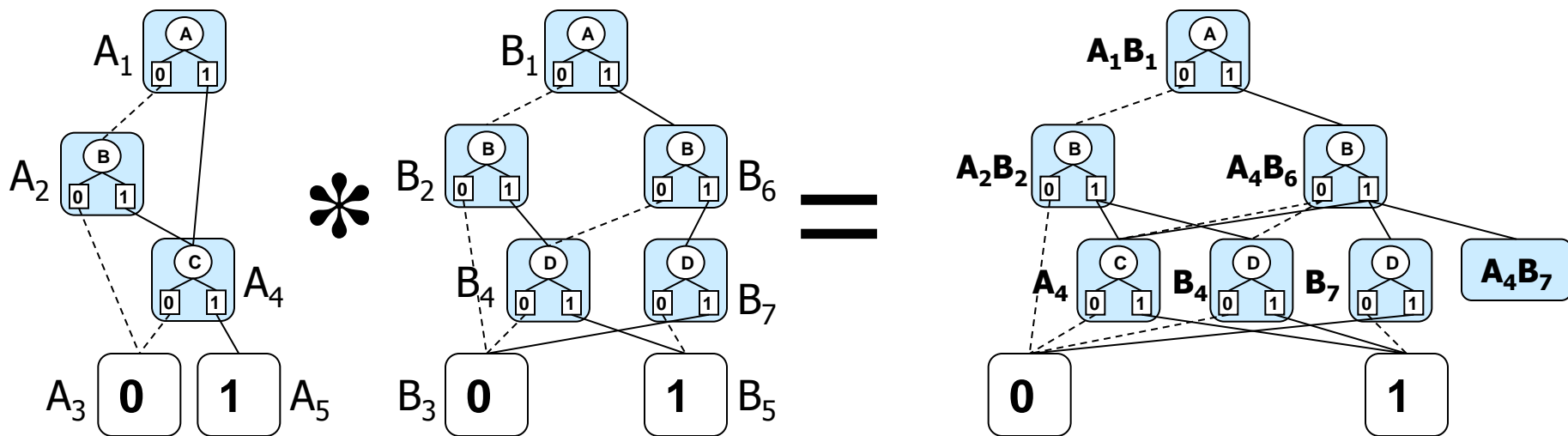
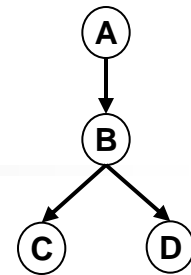
Apply Operator



Apply Operator



Apply Operator



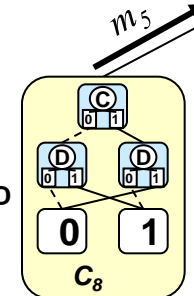
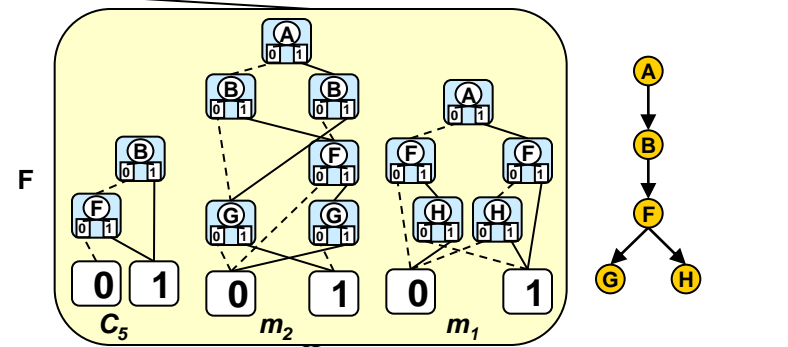
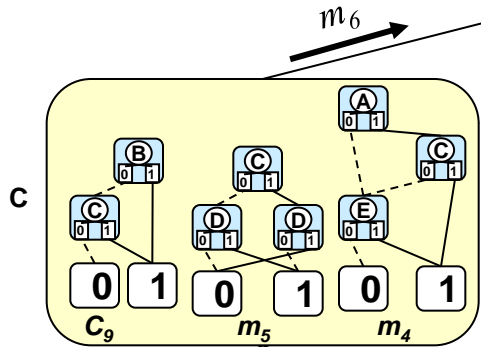
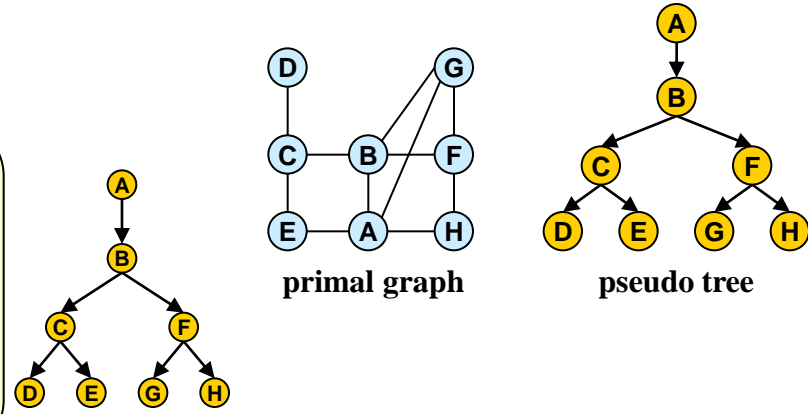
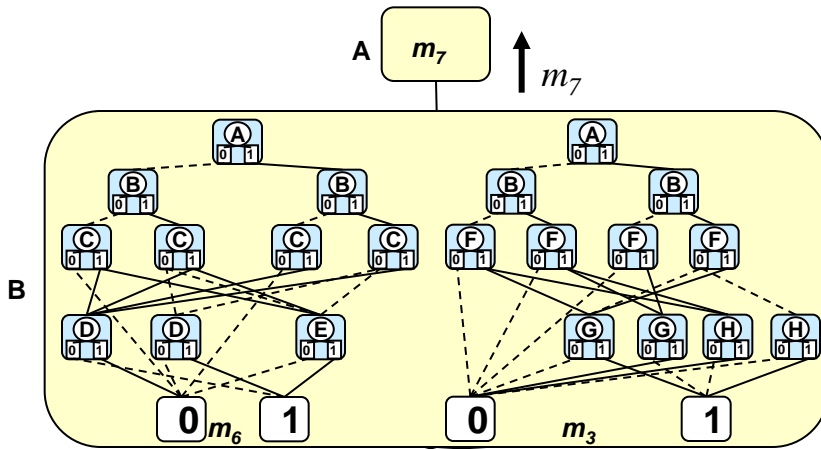


And/Or Multi-Valued Decision Diagrams

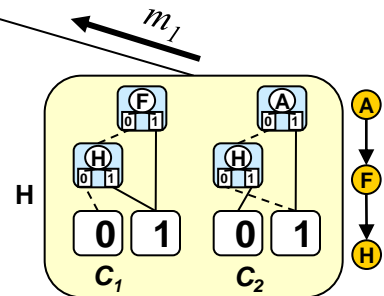
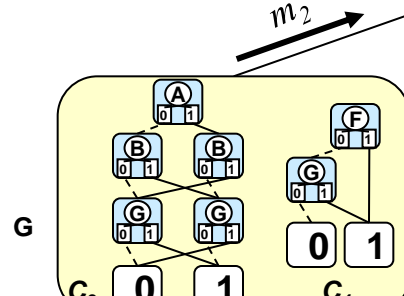
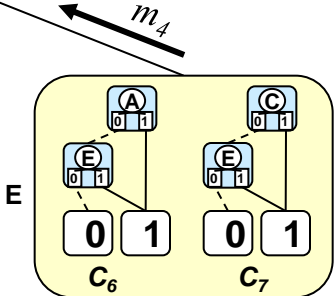
- AOMDDs are:
 - AND/OR search graphs
 - **canonical representations**, given a pseudo tree
 - Defined by two rules:
 - All isomorphic subgraphs are merged
 - There are no redundant (meta) nodes

Example:

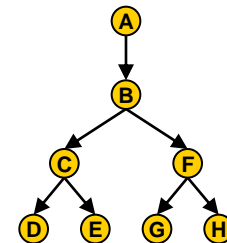
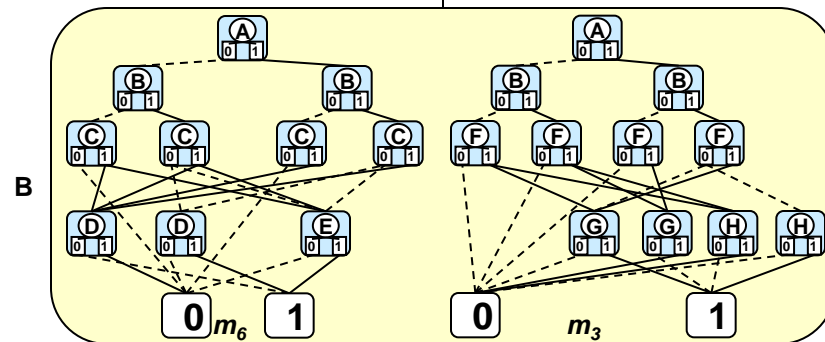
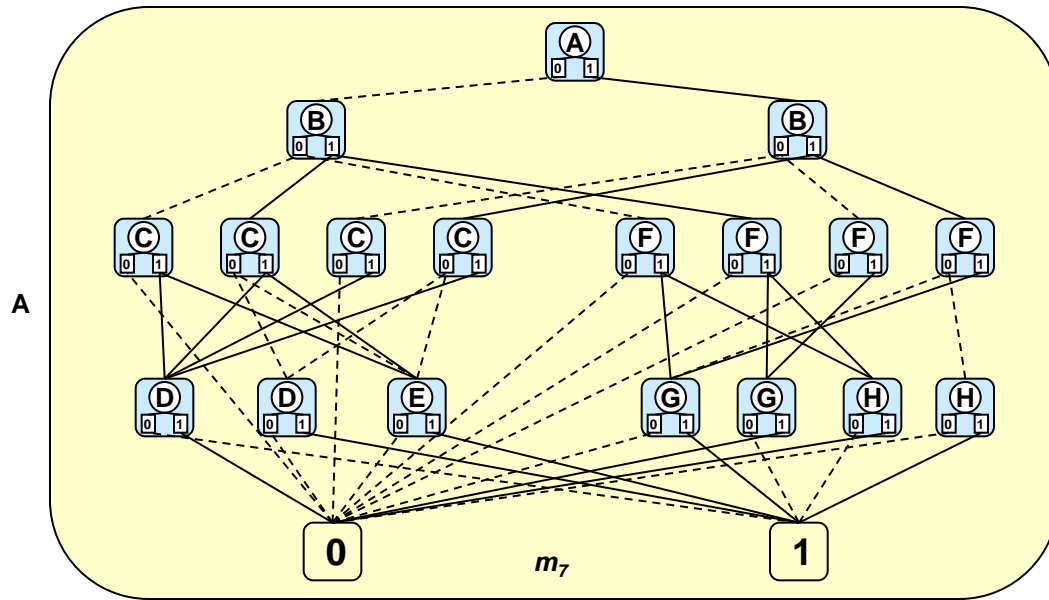
$$(f \vee h) \wedge (a \vee !h) \wedge (a\#b\#g) \wedge (f \vee g) \wedge (b \vee f) \wedge (a \vee e) \wedge (c \vee e) \wedge (c\#d) \wedge (b \vee c)$$



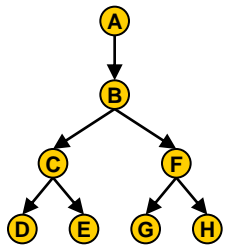
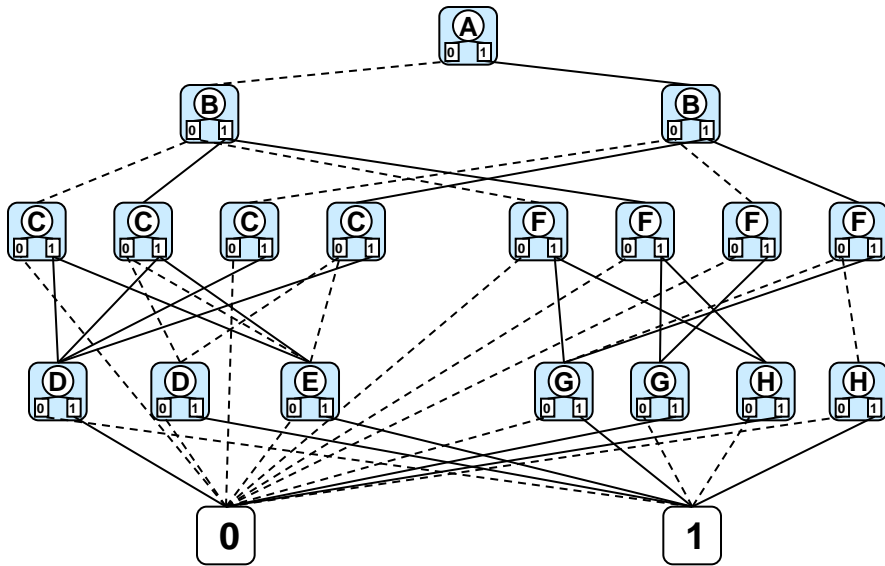
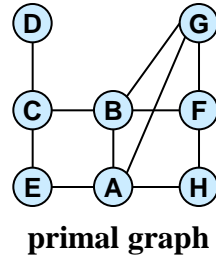
14



Example (continued)



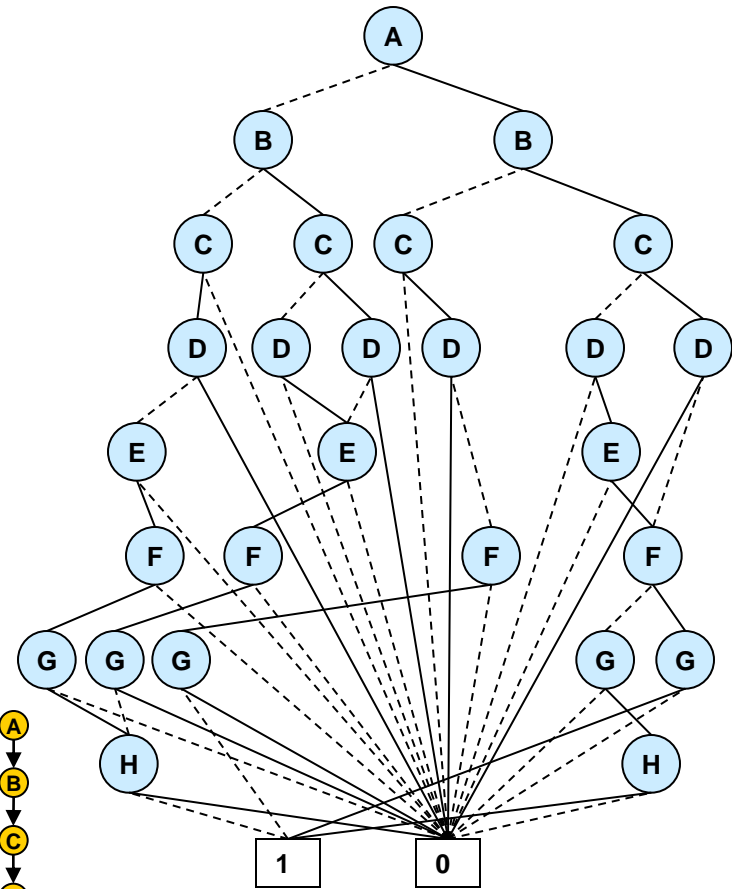
AOBDD vs. OBDD



AOBDD

18 nonterminals

47 arcs



OBDD

27 nonterminals

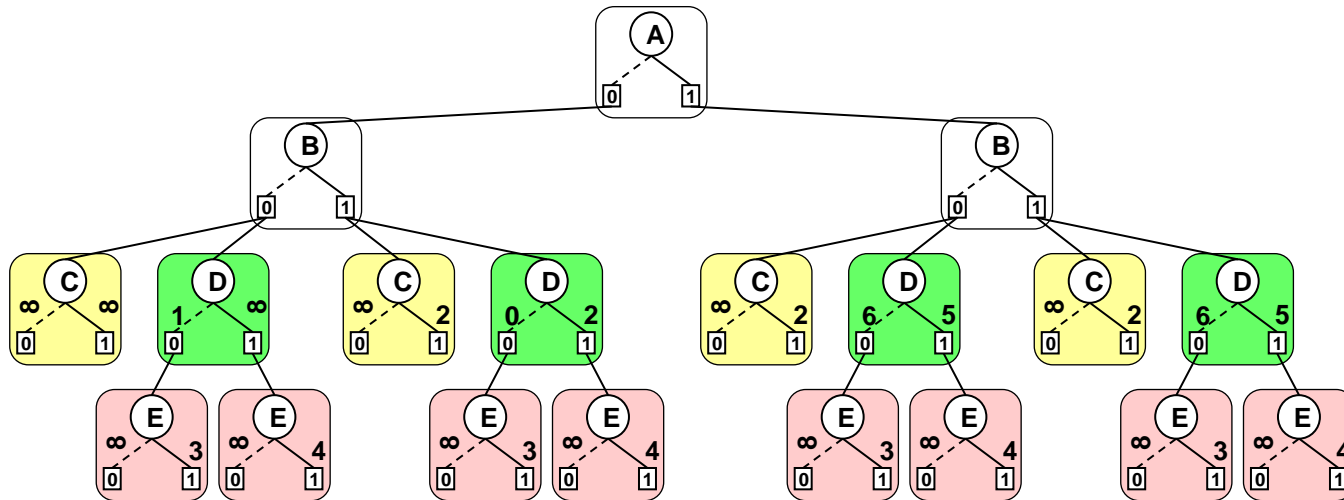
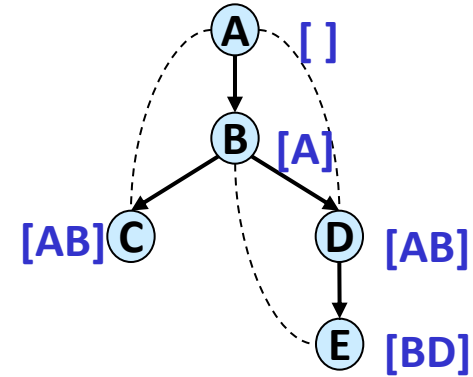
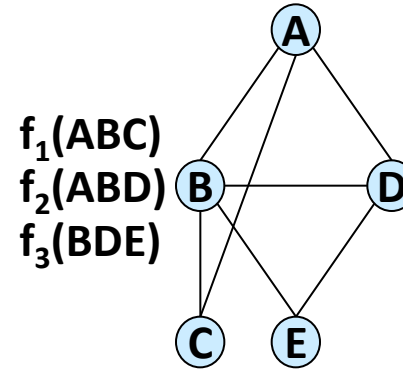
54 arcs

Constraint Optimization - AND/OR Tree

A	B	C	$f_1(ABC)$
0	0	0	∞
0	0	1	∞
0	1	0	∞
0	1	1	2
1	0	0	∞
1	0	1	2
1	1	0	∞
1	1	1	2

A	B	D	$f_2(ABD)$
0	0	0	1
0	0	1	∞
0	1	0	0
0	1	1	2
1	0	0	6
1	0	1	5
1	1	0	6
1	1	1	5

B	D	E	$f_3(BDE)$
0	0	0	∞
0	0	1	3
0	1	0	∞
0	1	1	4
1	0	0	∞
1	0	1	3
1	1	0	∞
1	1	1	4



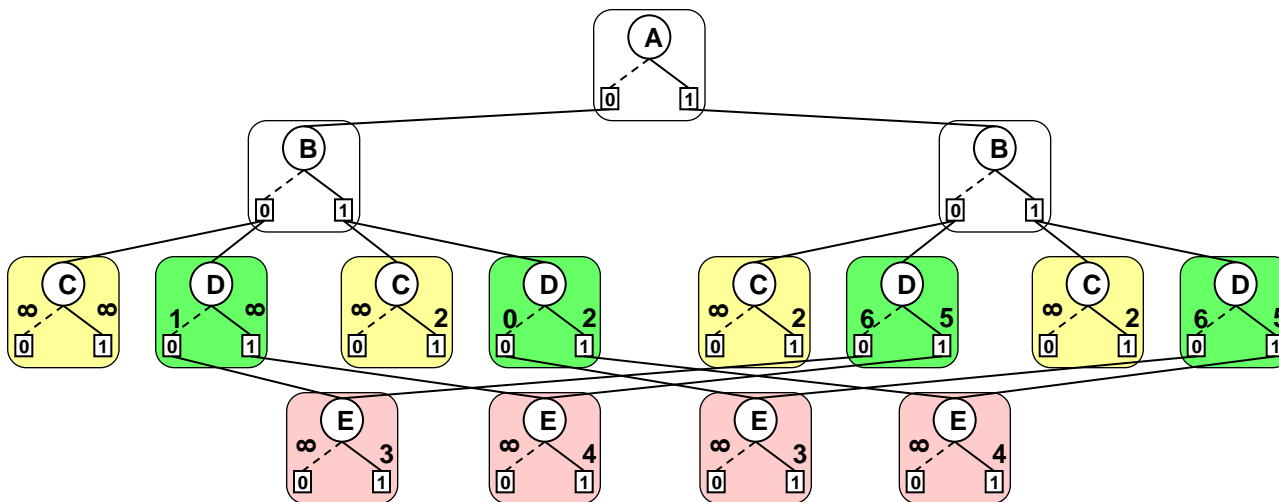
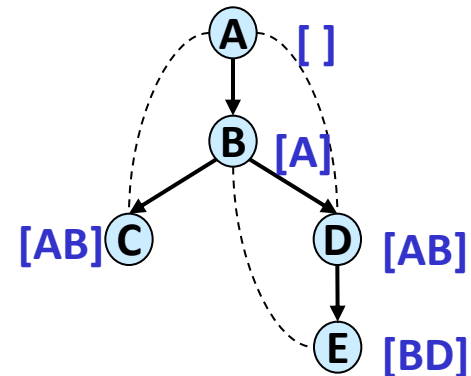
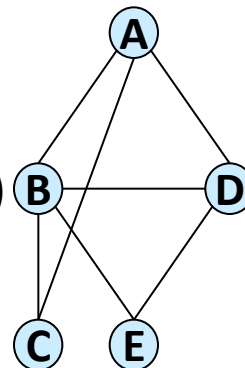
AND/OR Context Minimal Graph

A	B	C	$f_1(ABC)$
0	0	0	∞
0	0	1	∞
0	1	0	∞
0	1	1	2
1	0	0	∞
1	0	1	2
1	1	0	∞
1	1	1	2

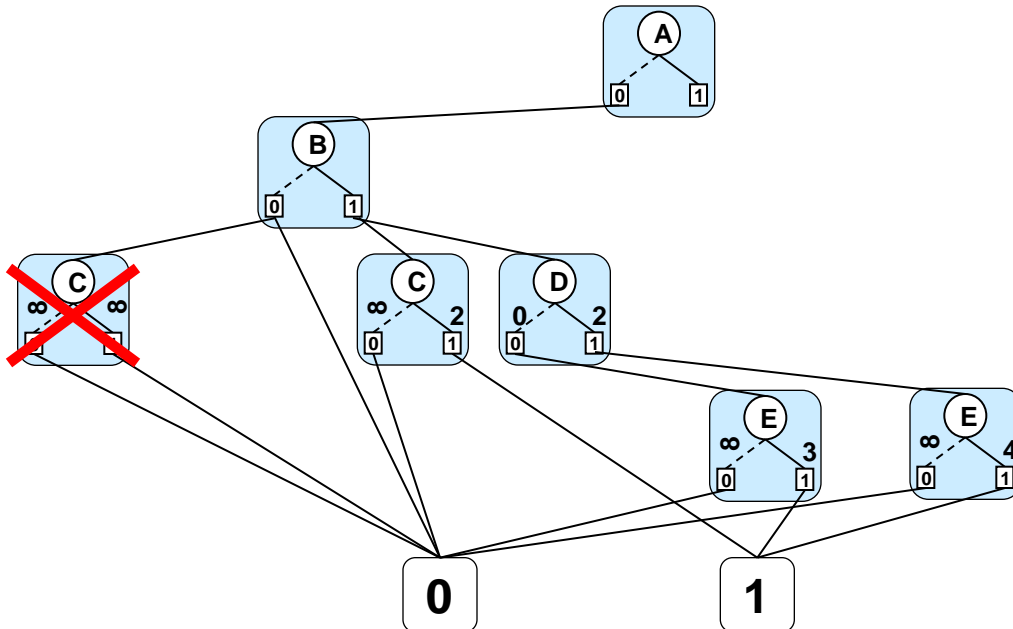
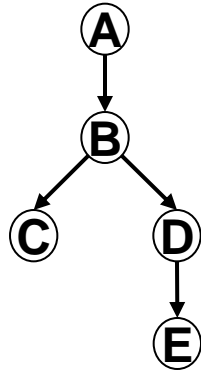
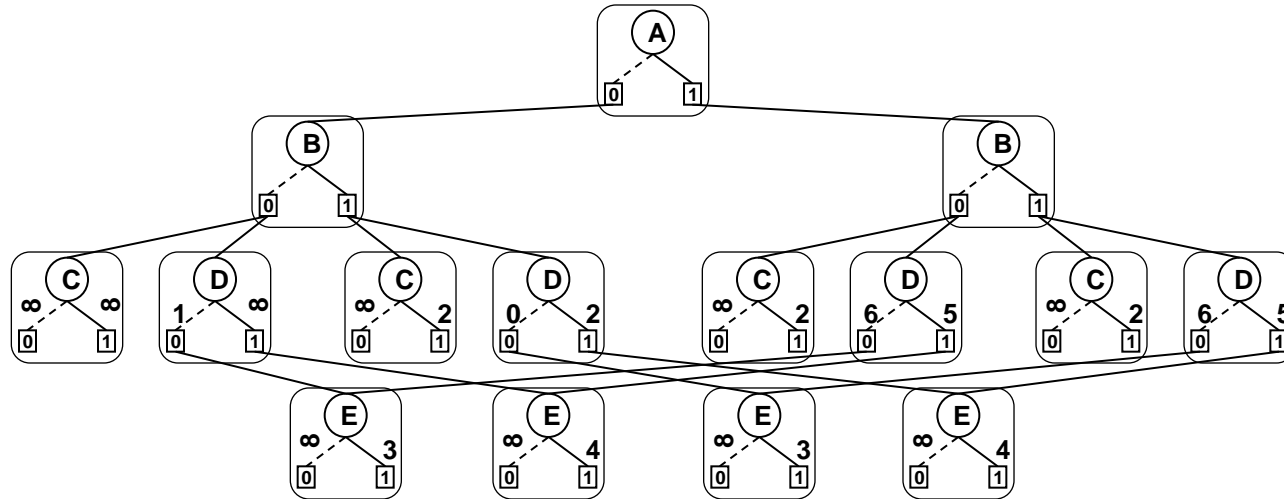
A	B	D	$f_2(ABD)$
0	0	0	1
0	0	1	∞
0	1	0	0
0	1	1	2
1	0	0	6
1	0	1	5
1	1	0	6
1	1	1	5

B	D	E	$f_3(BDE)$
0	0	0	∞
0	0	1	3
0	1	0	∞
0	1	1	4
1	0	0	∞
1	0	1	3
1	1	0	∞
1	1	1	4

$f_1(ABC)$
 $f_2(ABD)$
 $f_3(BDE)$

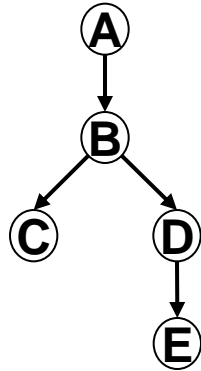
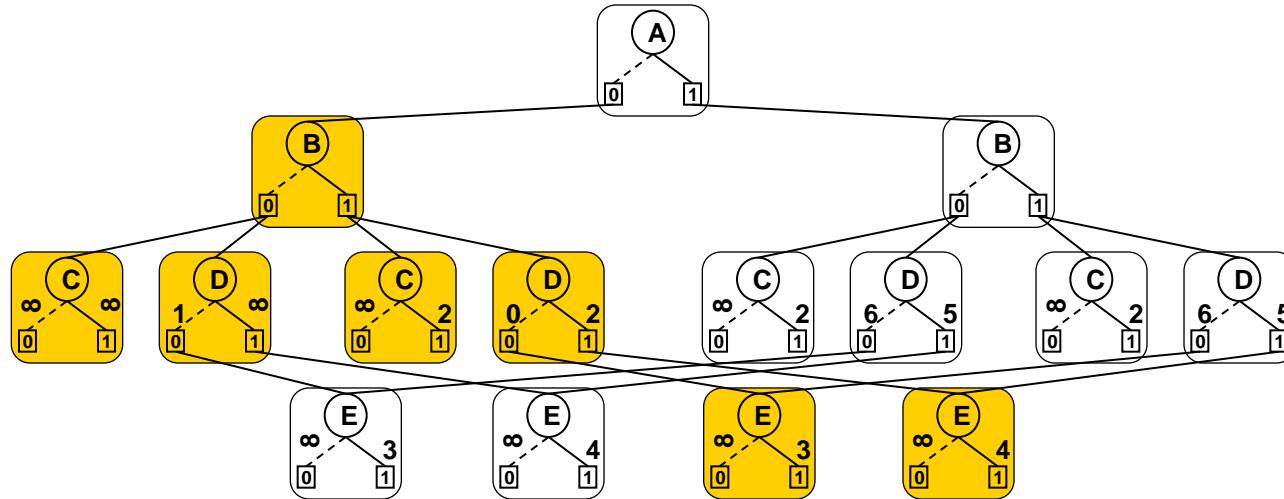


AOMDD – Compilation by Search

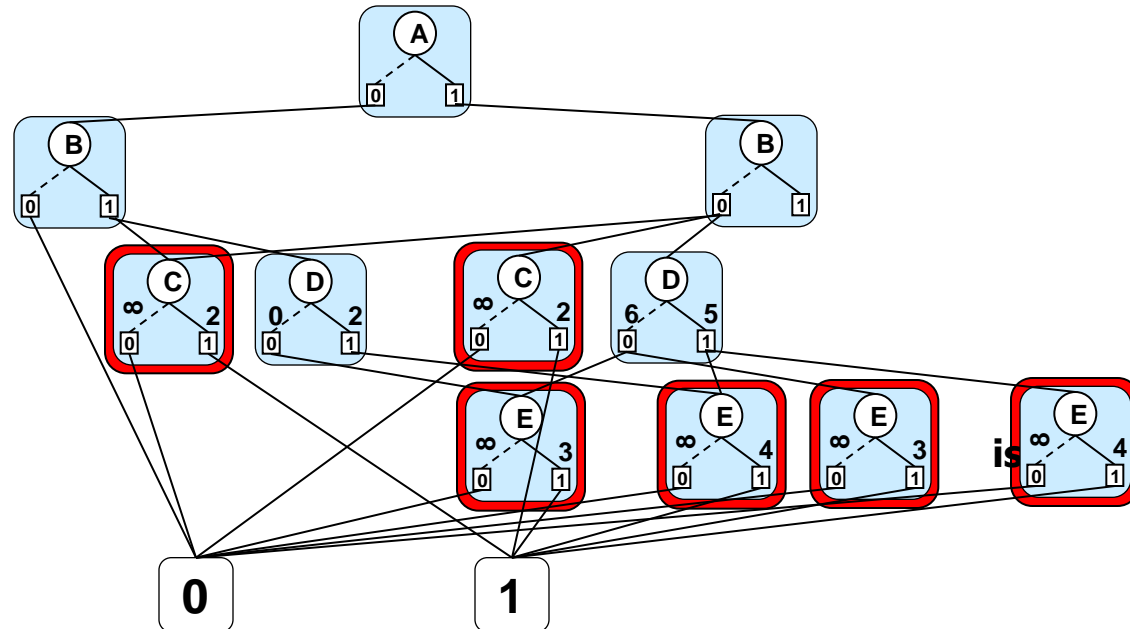


redundant

AOMDD – Compilation by Search

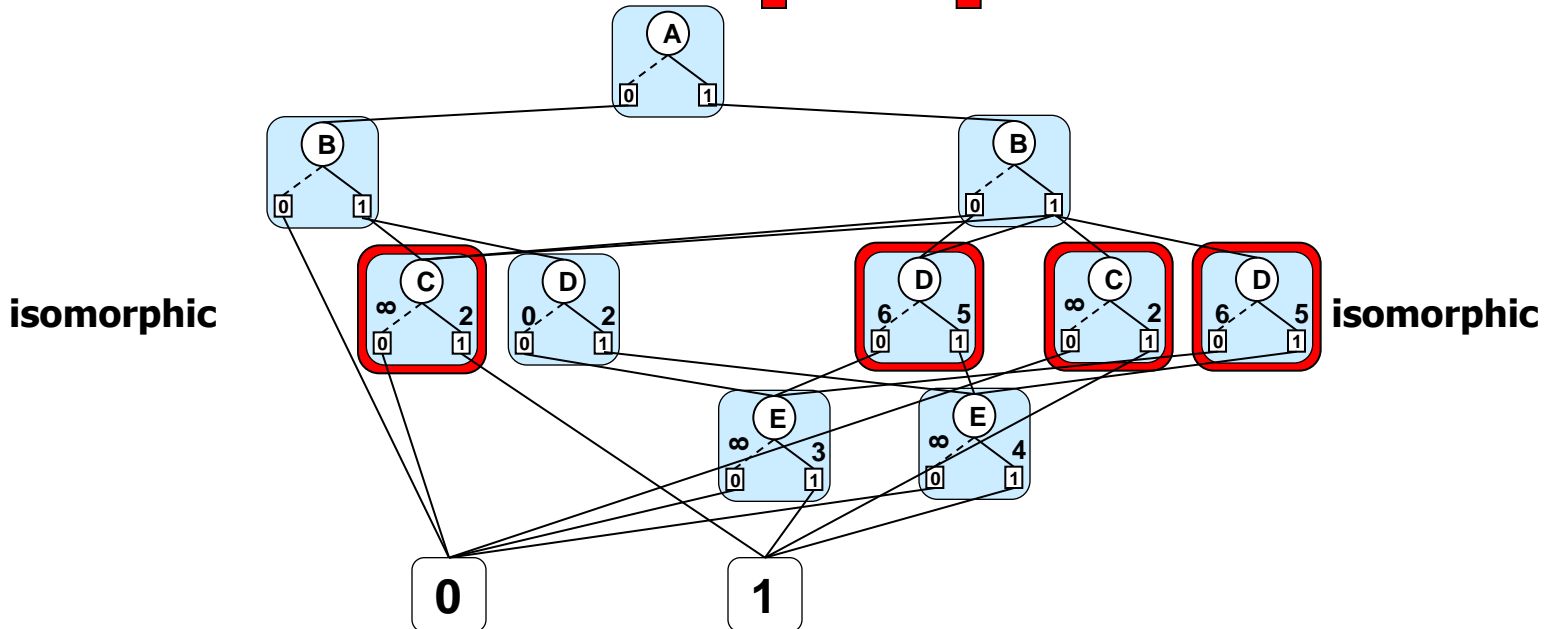
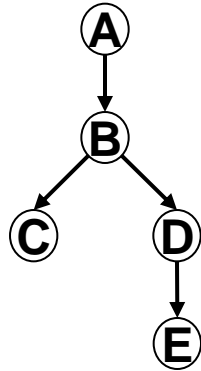
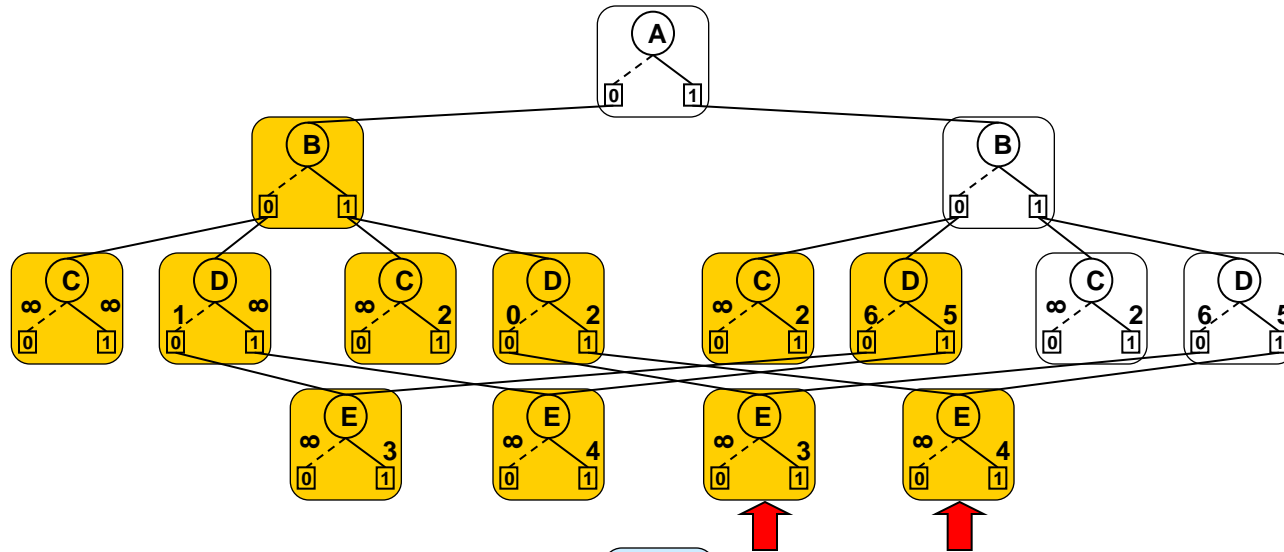


isomorphic

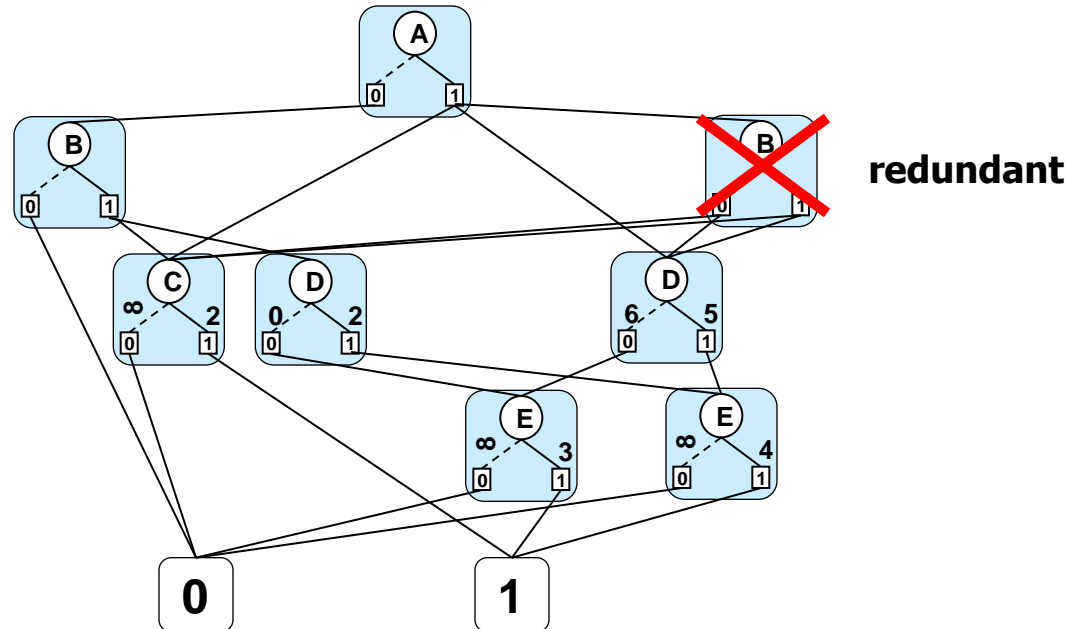
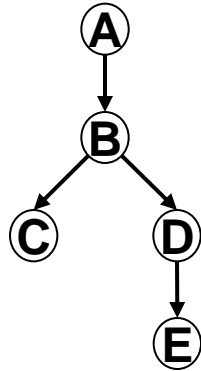
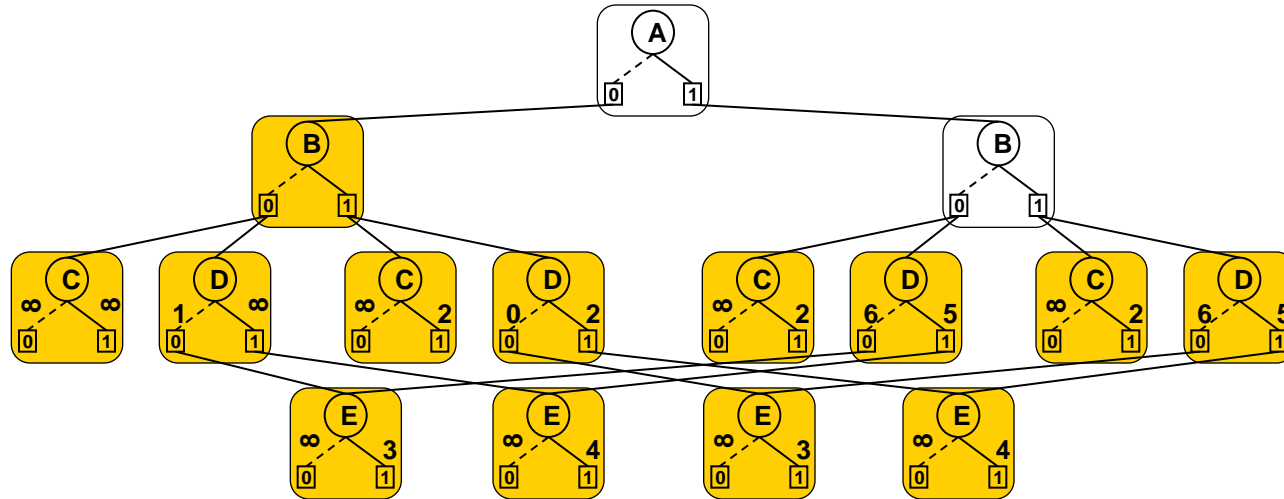


isomorphic

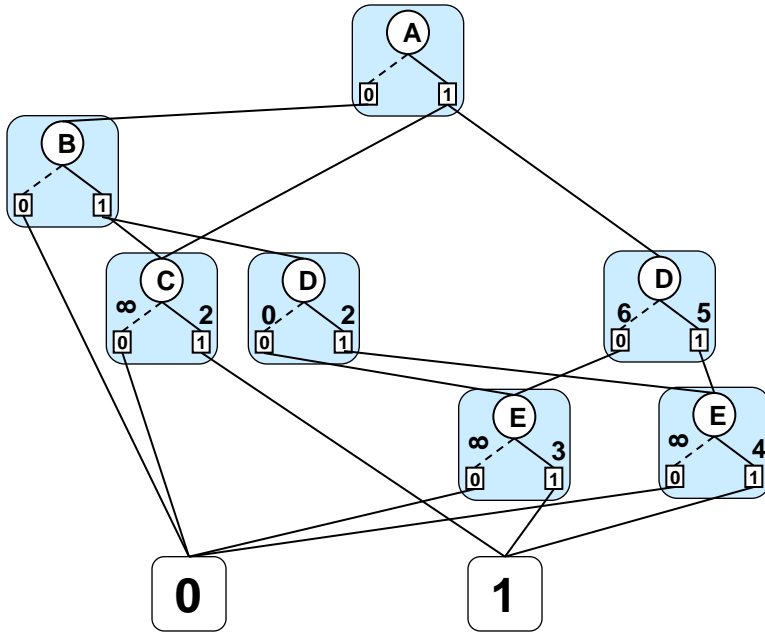
AOMDD – Compilation by Search



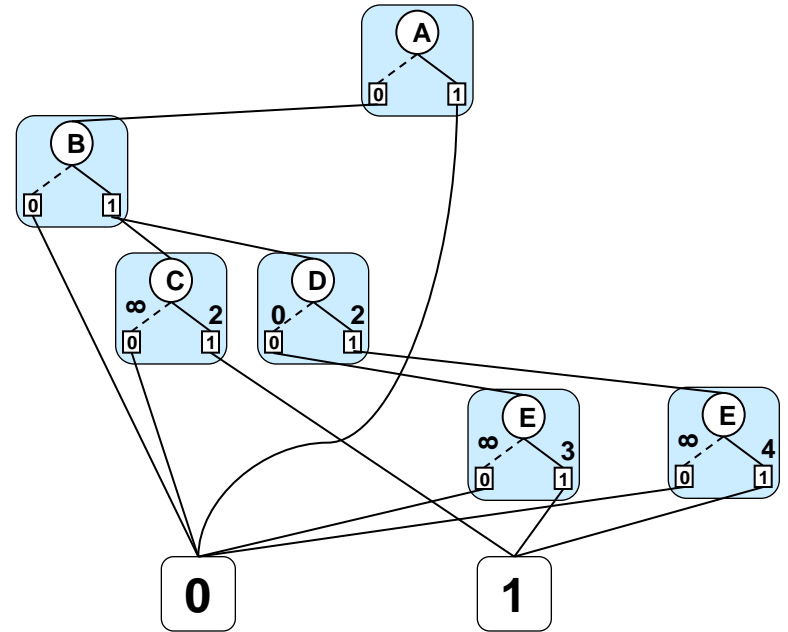
AOMDD – Compilation by Search



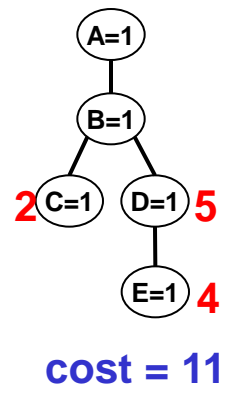
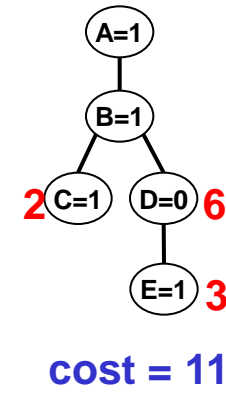
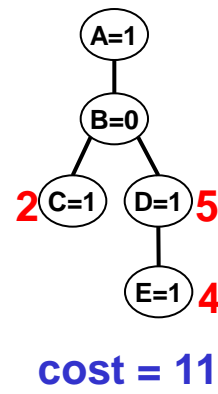
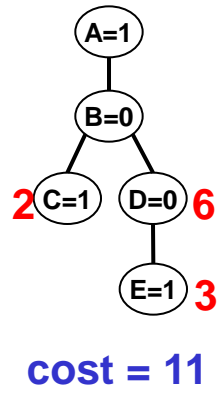
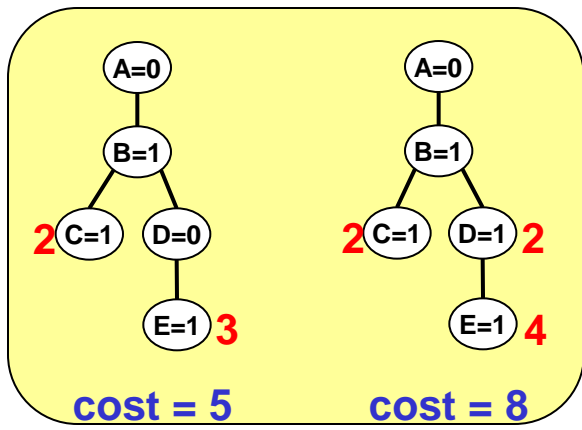
AOMDD for Constraint Optimization



AOMDD for all solutions



AOMDD for two best solutions





Complexity of Compilation

- The size of the AOMDD is $O(n k^{w^*})$
- The compilation time is also bounded by $O(n k^{w^*})$

k = domain size

n = number of variables

w^* = treewidth



Semantic Treewidth

- Given a network, there may exist a sparser equivalent network.
- Challenges the idea of using induced width to measure the difficulty of the problem
- AOMDD sizes are much smaller than the bound



Semantic Treewidth

- With respect to a pseudo tree, this is the smallest treewidth over all equivalent networks that can have that pseudo tree
- With respect to the network, this is the smallest semantic treewidth over all pseudo trees that can express the set of solutions
- Instead of the induced width bounding AOMDD size, we can use semantic treewidth.

Semantic Treewidth

eq(n,10)	graph	c	w*	h	time	#aomdd
n=10	chain	9	1	5	0.0240	91
	complete	45	9	9	0.0660	91
n=50	chain	49	1	25	0.1420	491
	complete	1225	49	49	1.1130	491
n=100	chain	99	1	50	0.3120	991
	complete	4950	99	99	5.5900	991

Equality constraint network
results



Constraint Propagation

- We can also prune the search space during compilation without removing possible solutions.
- In Bayesian networks, prune a subtree if the weight of the assignment is 0.



Experiments

- What about the pseudo-tree height parameter?
- Problems: WCSP instances
 - ISCAS 89 Circuits
 - SPOT5 Satellites
 - Mastermind
 - CELAR6 Radio Frequencies
- Time bound for compilation: 3 hours



Experiments

- Compilation was for finding the optimal solution
- Used AOBB with static mini-bucket heuristics (i-bound = 10)
- Tried different implementations of MinFill
 - Existing implementation in the compiler
 - daoopt (gets lower h because it considers it too)
 - CVO



Experiments

- BnB pruning makes the size unpredictable as a function of the parameters
- Need to modify the routine for solution counting so the entire AOMDD is actually compiled



Experiments (BN)

- Reproduce and extend BN results in JAIR 2008 paper
- UAI 2006 Bayesian network benchmarks
 - Domain sizes of 2
 - Evidence on 30 random variables (to simplify the networks slightly)
 - Many elements with “0” support
- Compile with constraint propagation



To do

- Perform experiments to compare optimization vs. full compilation
 - Need to extend code for WCSPs
- Another way to deal with unpredictability of w/h vs. size in optimization?
 - Compute many orderings with equal w/h and average the search space/AOMDD sizes.



Future Work?

- Try regressing curves that depend on w , h , or both
- Evaluation of bottom-up version of compilation (Robert's algorithm in CP 2006)