

Variable Shift SDD: A More Succinct Sentential Decision Diagram

<u>Kengo Nakamura</u> (NTT) Shuhei Denzumi (The University of Tokyo) Masaaki Nishino (NTT)

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Boolean function

 $(A \vee \overline{B}) \land (B \vee \overline{C})$

Input: n Boolean values, Output: a Boolean value

 $f: \{true, false\}^n \rightarrow \{true, false\}$

There are **useful queries** for Boolean function:

- Model count: How many patterns of input make output true?
- Equivalency: Are given two Boolean functions equivalent? etc.

However, these queries are **difficult to solve** Ex.) Using truth table, they take $O(2^n)$ time

			-			
A	B	С	f			
false	false	false	true			
false	false	true	false			
false	true	false	false			
false	true	true	false			
true	false	false	true			
true	false	true	false			
true	true	false	true			
true	true	true	true			
Truth table						

Ordered Binary Decision Diagram (OBDD) [Bryant '86] NTT (

Directed Acyclic Graph (DAG) representation of Boolean function $(A \land B) \lor (B \land C)$ $\lor (C \land D)$

- Size of OBDD is generally exponential, but is significantly smaller than 2ⁿ for many practical cases
- OBDD supports many queries in polytime w.r.t. OBDD size
 - Model count, Equivalency, etc.



OBDD and Apply operation [Bryant '86]

OBDDs support **Apply operation** in $O(|\alpha||\beta|)$ time

– Input: Two OBDDs α , β and <u>binary operator</u> \circ

Size of α

- conjunction(∧), disjunction(∨), etc.
- **Output**: OBDD representing $f \circ g$
 - α (β) represents f (g resp.)

Fundamental in

- constructing OBDD representing any Boolean function
- supporting useful queries



Sentential Decision Diagram (SDD) [Darwiche '11]

- DAG representation of Boolean function
- Generalization of OBDD structure
- Tighter bound on size than OBDD
 - In some cases, SDD can make size exponentially smaller than OBDD [Bova '16]
- Supporting useful queries in polytime w.r.t. SDD size
- Supporting Apply operation in $O(|\alpha||\beta|)$ time

Input: Two **SDDs** α , β and oper. \circ Output: **SDD** representing $f \circ g$ $(A \land B) \lor (B \land C)$ $\lor (C \land D)$



SDD

Substructure sharing



One of the reason why OBDD/SDD is succinct:

- Boolean function is recursively decomposed into subfunctions that are also represented by DAGs (substructures)
- When **same subfunctions** emerge, **only one DAG is needed** instead of having multiple (identical) DAGs

... Share of substructure representing the same function



Proposal: Variable Shift SDD (VS-SDD)

Can we share more substructures to achieve more succinct representation? -> Variable Shift SDD

... SDD-based representation of Boolean function

- VS-SDD can share substructures of subfunctions that are equivalent under specific type of variable substitution
 - Ex.) $C \wedge D$ can be obtained from $A \wedge B$ by substituting A with C and B with D
- Representing Boolean function of substructure is dependent on the path from root to it

Typical example of emerging such sets of subfunctions: modeling time-evolving systems;



Proposal: Variable Shift SDD (VS-SDD)



- VS-SDD is never larger than SDD of the same function
 - There exists a class of function for which VS-SDD is exponentially smaller than SDD
- VS-SDD supports many useful queries as SDD does
 - Supports Apply operation in $O(|\alpha||\beta|)$ time

Input: Two **VS-SDDs** α , β and oper. \circ Output: **VS-SDD** representing $f \circ g$

VS-SDDs incur no additional overhead over SDDs while being potentially smaller than SDDs

 $(A \land B) \lor (B \land C)$ $\vee (C \wedge D)$ **VS-SDD** $A \wedge R$ $C \wedge D$

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Related works

Attempts to share substructures representing "equivalent" Boolean functions up to conversion

For OBDDs:

- Attributed edge [Madre & Billon '88]
- Complement edge [Minato et al. '90] ... up to taking negation
 - Solvability of useful operations is not considered
- ↑ΔBDD [Anuchitanukul et al. '95] ... up to "shifting" of variables
 - Operation like Apply is supported, but its time complexity is not polynomial of DAG sizes

 $f(\mathbf{X}) \leftrightarrow \neg f(\mathbf{X})$

 $\leftrightarrow f(X_4, X_5, X_6)$

 $f(X_1, X_2, X_3)$



Related works

For other DAG structures:



- Sym-DDG/Sym-FBDD [Bart et al. '14] ... up to any variable substitution
 - Based on DDG [Fragier & Marquis '06] and FBDD [Gergov & Meinel '94]
 - Fail to support some important operations such as Apply and Conditioning, as the based structures do

For SDDs:

- No such previous study is known
- VS-SDD cannot be obtained by straightforward application of techniques for OBDDs (like [↑]ΔBDD)

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SDD: Boolean circuit view



SDD = Boolean circuit with structured decomposability and strong determinism

- Represents Boolean function by recursive application of **OR** and **AND**
 - OR gate corresponds to circle node
 - AND gate corresponds to box pair



Decomposability

- NTT 🕐
- **Decomposable**: for each **AND** gate, <u>used variables are non-overlapping</u> for any two inputs of it





Vtree and structured decomposability



Decomposable circuit whose decomposition of variables is structured with vtree

- Vtree: <u>full binary tree</u> whose leaves are labeled with variables
- Structured decomposable (given vtree v): for each AND gate, it has two inputs α, β and there is an internal node in v s.t. used variables of α (β) are all left (right resp.) descendants



Strong determinism

For structured decomposable Boolean circuits:

• Strongly deterministic: for each OR gate, let p_i be a Boolean function of the left input of *i*-th child AND gate. Then, $p_i \wedge p_j = false$ ($i \neq j$) holds.

Due to structured decomposability and strong determinism, **SDD supports polytime Apply** and has preferable properties such as canonicity (uniqueness)





SDD: structure and respecting vtree node



Seeing SDD as data structure representing **Boolean function (given vtree)**

- **OR** gate of circuit is represented as **decomposition node** (circle)
 - It has corresponding vtree node called respecting node
- Bottom literal (constant) of circuit is represented as literal (constant resp.) node
- Size of SDD is defined as sum of #(inputs) of all OR gates



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Observation



Equivalent substructures (except for labels) emerges if:

- their respecting vtrees are identical (same shape) except for labels
- their representing functions are equivalent under variable substitution induced from vtree isomorphism
- Ex.) SDD representing $f = (A \land B) \lor (B \land C) \lor (C \land D)$
- Vtrees rooted at 2 and 5 are identical
 - Exchanged by following substitution: $B \leftrightarrow D$, $A \leftrightarrow C$
- $A \wedge B$ and $C \wedge D$ are equivalent under above substitution
- -> Bottom-middle and bottom-right nodes have same shape



VS-SDD: key idea

- SDD retains **respecting vtree node (ID)** of each decomposition and literal **explicitly**
- -> Substructures representing Boolean functions with different variables cannot be shared
- VS-SDD retains it differentially in edges
 ... It is represented as sum of numbers from root to decomposition/literal node
- To facilitate this idea, vtree nodes are numbered following preorder traversal of vtree



VS-SDD: example



Given vtree, VS-SDD is also data structure to represent Boolean function

- Based on SDD structure
- Respecting vtree node ID of node is retained by the sum of offset and the numbers from root to it
- Ex.) VS-SDD representing $f = (A \land B) \lor (B \land C) \lor (C \land D)$

Marked substructure represents $A \wedge B$ when traversing yellow path, and $C \wedge D$ when traversing purple path



Size of VS-SDD



- The size of VS-SDD is **not larger than its SDD counterpart**
- There is a class of function for which VS-SDD is exponentially smaller than SDD
- Ex.) Boolean function with high symmetry
- Attach a variable for each edge of complete binary tree of depth *j*
- Consider a Boolean function that evaluates *true* iff the edges corresponding to *true* variables constitute a matching
 - SDD size is $\Omega(2^j)$ given any vtree
 - VS-SDD size is O(j) given vtree with recursive structure
 - For j = 14, SDD size is 278377 while VS-SDD size is only 259



Other properties and operations of VS-SDD

Properties:

- VS-SDDs have canonicity (under some assumption), as SDDs do
 - Given vtree, there is **unique VS-SDD** for a Boolean function

Operations:

- VS-SDDs support **Apply operation** in $O(|\alpha||\beta|)$ time
- Useful queries listed in [Darwiche & Marquis '02] that are supported in polytime by SDDs are also supported in polytime by VS-SDDs
 - Including model count, equivalency, ...

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Experiment



Evaluating how much our approach reduces the DAG size

- For benchmark Boolean functions (in CNF form), we compare the sizes of SDD and VS-SDD, given the same vtree
 - Vtree is generated by SDD package version 2.0 [Choi & Darwiche '18] with balanced initial vtree
 - ... Meaning that vtree is searched to suit for SDDs
- Used benchmarks:
 - Planning dataset (used as benchmark for Sym-DDG [Bart et al. '14])
 - **N-queens problem** (used as benchmark for ZDD [Minato '93])
 - Grid matching enumeration
- Problems in which SDD compilation took more than 10 minutes are omitted
- Environment: 64-bit macOS, Intel Core i7@2.5GHz, 16GB RAM, Language: C++

Planning dataset



- Given: state variables (Boolean) $f_1, ..., f_n$, set of actions $\{a_1, ..., a_k\}$
- Action changes assignment of state variables, depending on current assignment
- Given initial and goal assignments, we want to find out a sequence of actions that maps initial assignment to goal assignment

Modeling this as a Boolean function of following variables, it exhibits high symmetry

- $f_{t,i}: f_i$ of timestamp t (= 0, ..., T)
- $a_{t,j}$: whether action a_j is performed at timestamp t (= 1, ..., T)

Ex.) Sokoban puzzle

- Player wants to move boxes to storage by pushing box
- Player can only push box when the square next to box is empty
- * In experiment, more simple planning problems are used



Other datasets

• N-queens puzzle

... Place N queens on $N \times N$ chessboard s.t. no two queens attack each other

- Associate variable for each square, and consider function that evaluates *true* iff constituting solution of *N*-queens puzzle
- Matching enumeration on graph [Kawahara et al. '17]
 - Associate variable for each edge, and consider function that evaluates *true* iff constituting matching
 - ... For grid graphs, such function exhibits symmetry





Results

<u>Planning dataset</u> (Suffix " $_tn$ " stands for T = n):

- VS-SDD reduces the sizes to around 60-80% of original SDD
 - This compression ratio is competitive to (or for some cases better than) that of Sym-DDG compared to DDG

Many nodes representing substitution-equivalent functions are found among bottom nodes of SDD, yielding substantial size decrease of VS-SDD

	-SDD	size	
SDD size		$\overline{}$	'NTT (°)
Problem	#vars	S	V ratio
$blocks-2_t3$	248	8811	$7057\ 80.1\%$
$blocks-2_t5$	406	31861	28858 90.6%
$bomb-5-1_t3$	348	3798	$2278\ 60.0\%$
$bomb-5-1_t5$	564	6327	$3960\ 62.6\%$
bomb-5-1t7	780	11212	$7287\ 65.0\%$
$bomb-5-1_t10$	1104	16514	$10426\ 63.1\%$
$\operatorname{comm-5-2}$ t3	488	20584	$18033\ 87.6\%$
emptyroom-4 t3	116	1822	$1146\ 62.9\%$
emptyroom-4 t5	188	3090	$1885\;61.0\%$
emptyroom-4 ^{t7}	260	5073	$3001 \ 59.2\%$
emptyroom-4 t10	368	106737	$103417\ 96.8\%$
emptyroom-8_t3	244	10511	$8549\ 81.3\%$
safe-5_t3	54	567	$441\ 77.8\%$
${ m safe-5}{ m t5}$	86	898	$640\ 71.2\%$
safe-5t7	118	1710	$1314\ 76.8\%$
$safe-5_t10$	166	2506	$1756\ 70.1\%$
safe- 30 t 3	304	5476	$4067\ 74.3\%$
safe-30t5	486	8710	$6328\ 72.7\%$
$safe-30_t7$	668	14449	$10371\ 71.8\%$
$safe-30_t10$	941	23469	$17421\ 74.2\%$
8-Queens	64	2222	$1624\ 73.1\%$
9-Queens	81	5559	$4767\ 85.8\%$
10-Queens	100	10351	$9159\ 88.5\%$
11-Queens	121	30611	$28876 \ 94.3\%$
Matching-6x6	60	13091	$12671\ 96.8\%$
Matching-8x8	112	98200	$97103\ 98.8\%$
Matching-6x18	192	36228	$34241 \ 94.5\%$

Results



Other datasets:

- N-queens data exhibits better compression ratios
- Matching data does not fit to VS-SDD

Due to the asymmetry in strong determinism ...?

Problem	#vars	S	V	ratio
blocks-2 $t3$	248	8811	7057	80.1%
blocks-2 t5	406	31861	28858	90.6%
bomb-5-1 $t3$	348	3798	2278	60.0%
$bomb-5-1_t5$	564	6327	3960	62.6%
bomb-5-1 $t7$	780	11212	7287	65.0%
$bomb-5-1_t10$	1104	16514	10426	63.1%
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Conclusion



- VS-SDD represents Boolean function succinctly as OBDD and SDD do
- VS-SDDs incur no additional overhead over SDDs while being potentially smaller than SDDs
- Experiments show that VS-SDD effectively captures the symmetries of functions especially for planning datasets

Future work:

- Suitable vtree search for VS-SDD
- Top-down construction of VS-SDD