# **Metaheuristic Binary Decision Diagram**

Finding best BDD ordering on efficient way by metaheuristic solution

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# **Binary Decision Diagram**

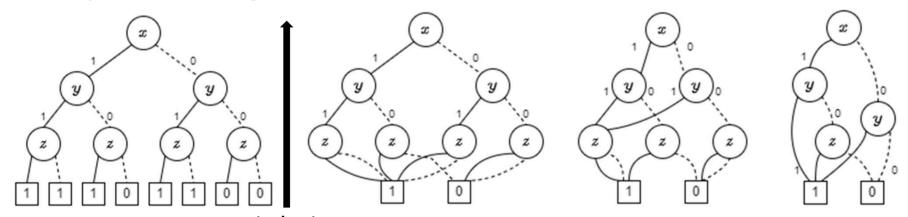


Fig 1. binary decision tree (reduce)

Fig 2. reduced ordered binary decision diagram

#### non-terminal nodes + terminal nodes

boolean variables 0 or 1

dashed line ---- 0 solid line — 1

# Variable ordering problem (NP-hard)

f(a,b,c,d,e,f) = (a&b)|(c&d)|(e&f)

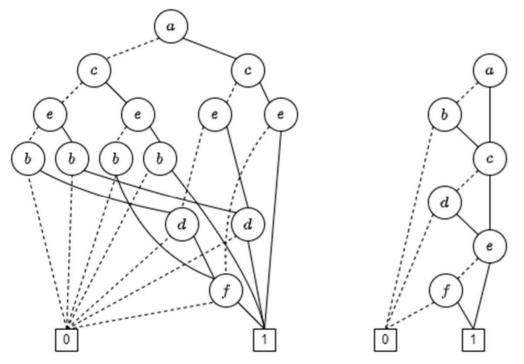
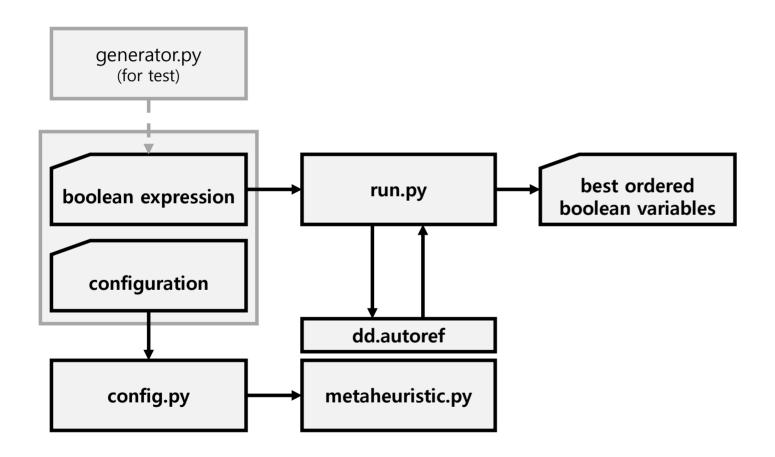


Fig 3. ROBDD ordering [a, c, e, b, d, f]

Fig 4. ROBDD ordering [a, b, c, d, e, f]

# **Overall Structure**



## **Parameters (configuration)**

| • | POP_SIZE    | creating population | size of population  |
|---|-------------|---------------------|---|
| • | POOL_SIZE   | crossover, mutation | size of parent pool                                       |
| • | N_LIMIT     | stopping criteria   | iteration limitation                                      |
| • | LIMIT_RATIO | ıı                  | constant ratio for CRITERIA                               |
| • | CRITERIA    | II                  | minimum number of new offsprings which means no changing. |

**Individual** ordering sequence of boolean variables. **all numbers are unique!** These individuals make population with POP\_SIZE.

```
for [x_0, x_1, x_2, x_3, x_4, x_5] ... [0, 1, 2, 3, 4, 5], [0, 2, 4, 1, 3, 5], [3, 4, 0, 5, 2, 1] ...
```

# **Stopping Criteria**

```
if len(newPop) < CRITERIA:
    nIter += 1
    if nIter == N_LIMIT:
        break</pre>
```

If there is not enough change, increase iteration limitation. When that chance reaches to the limitation, the process is over. Fitness building reordered BDD from individual, return number of new ordered BDD nodes.

For boolean function  $f(x_1, ..., x_n)$ 

$$fitness = N_{BDD\_nodes}$$
  $2n + 2 \le N_{BDD\_nodes} \le 2^{n+1}$ 

**Selection** FPS (Fitness Probability Selection)

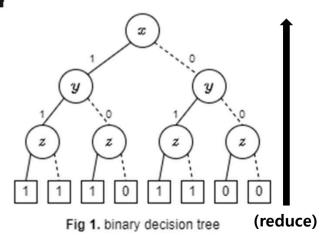
```
def fps(pop):
    total_fitness = sum(fitness)
    weight = [f/total_fitness for f in fitness]
    return weight
```

Crossover keeping pivot and its backward, shuffle others.

$$[0, 1, 2, 3, 4, ..., 7, 8, 9, 10] \rightarrow [3, 7, 4, 1, 2, ..., 6, 8, 9, 10]$$

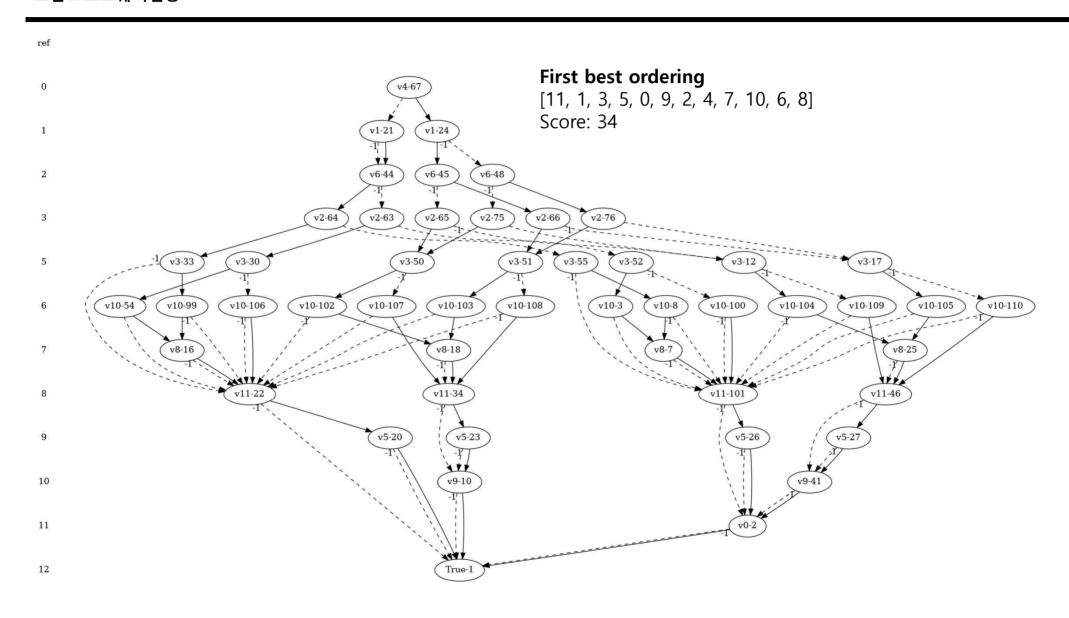
Mutation just swap 2 variables to avoid local optima.

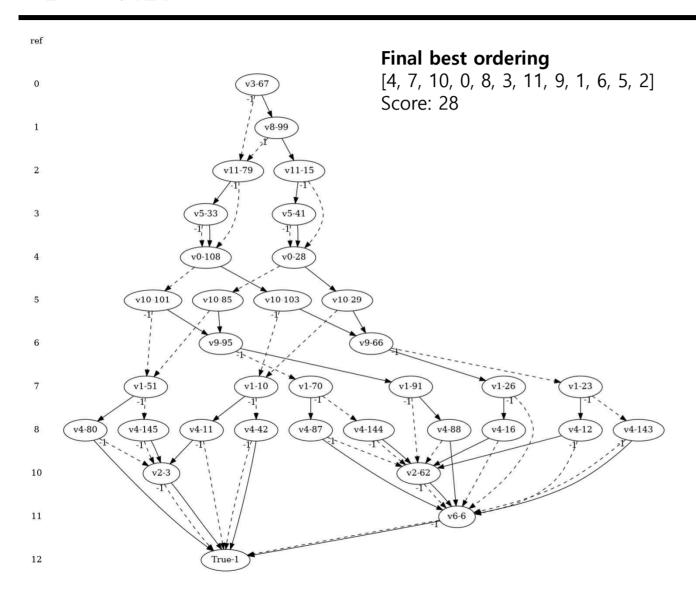
**Design** elitism.



```
Target expr: v0 | v2 ^ v10 & ~ v6 ^ ~ v9 | ~ v10 | ~ v4 & v9 ^ ~ v5 | ~ v11 ^ ~ v4 ^ ~ v3 | ~ v8 | ~ v10 ^ ~ v1 | v4 & ~ v2 ^ v1 & ~ v9 | ~ v7 | v7
Best Ind : [11, 1, 3, 5, 0, 9, 2, 4, 7, 10, 6, 8]
Best Score: 34
Limitation: 0
Best Ind : [5, 9, 10, 2, 8, 1, 11, 7, 6, 4, 0, 3]
Best Score: 32
Limitation: 0
Best Ind : [6, 7, 0, 8, 5, 4, 1, 3, 10, 11, 9, 2]
Best Score: 31
Limitation: 0
_______
Best Ind : [3, 2, 0, 11, 4, 6, 1, 8, 5, 10, 9, 7]
Best Score: 30
Limitation: 0
Best Ind : [4, 7, 10, 0, 8, 3, 11, 9, 1, 6, 5, 2]
Best Score: 28
Limitation: 0
Best Ind : [4, 7, 10, 0, 8, 3, 11, 9, 1, 6, 5, 2]
Best Score: 28
```

**Best Order** [4, 7, 10, 0, 8, 3, 11, 9, 1, 6, 5, 2] v10, v4, v3, v8, v1, v9, v6, v0, v2, v7, v11, v5





## Limitation

## Time Consuming

- building BDD takes a lot of time.
- only work in small size (no more than 30)

### Mutation

- too weak to avoid local optima.

## **Further work**

#### Mutation

- do more than swapping only two elements.

# Applicability

- design for applying BDD-based model checker.