

Sequential ATPG

Introduction

Time-frame expansion methods

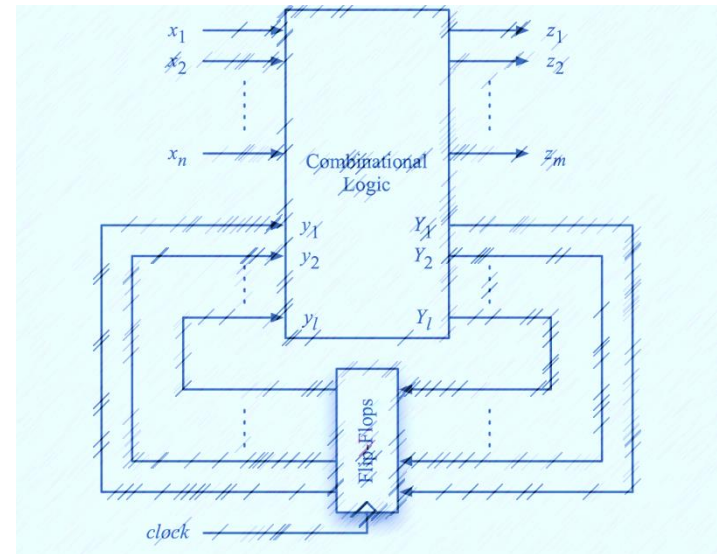
Simulation-based methods (* not in exam)

CONTEST [Agrawal & Cheng 88]

Genetic Algorithm

Issues of Sequential ATPG

Conclusions



Simulation-Based Methods

- Idea: use logic/fault simulators to guide ATPG [Seshu 62]
 - ◆ Simulation is faster than ATPG
- Approach
 - ◆ Generate **candidate** test vectors
 - ◆ ***Fitness**** of candidates evaluated by logic or fault simulation
 - ◆ Select best candidate based on a certain ***cost function***
- Advantage:
 - ◆ No time frame expansion. **Easy memory management**

Simulate Many Vectors and Choose Best

CONTEST– Concurrent Test Generator for Sequential Circuits [Agrawal & Cheng 89]

Based on **event-driven** concurrent fault simulator

- Search for test vectors guided by cost functions
- Three phases
 - ① Initialization
 - ② Concurrent fault detection
 - ③ Single fault detection

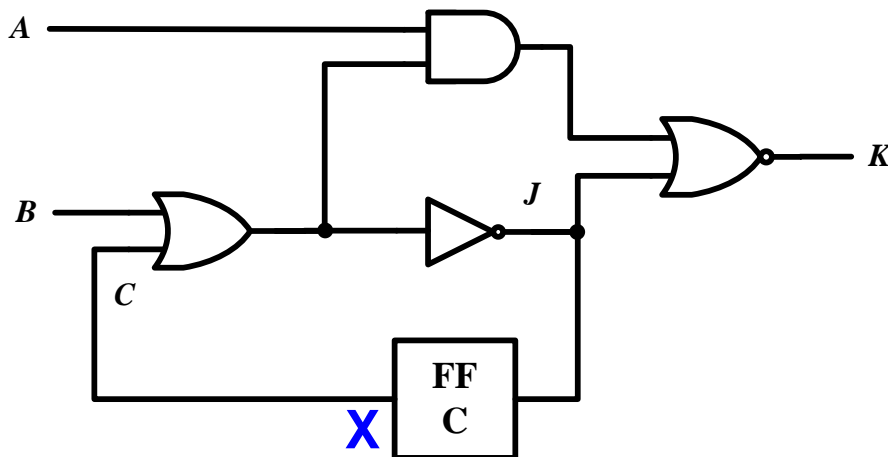
1. Initialization Phase

- Start with arbitrary test vector
 - ◆ Start with FFs in unknown states
- Use logic simulation (not fault simulation)
 - ◆ Cost = number of FFs in unknown state
 - ◆ *Trial vectors* are generated by single-bit change of the current vector. A trial vector is accepted and becomes the current vector if it lowers the cost
- Stop this phase when cost drops below a desired value

QUIZ

Q1: Initially the FF is unknown. Given three trial vectors, please simulate the circuit and decide their costs, where cost = number of unknown FF.

Q2: Which trial vector would you pick?

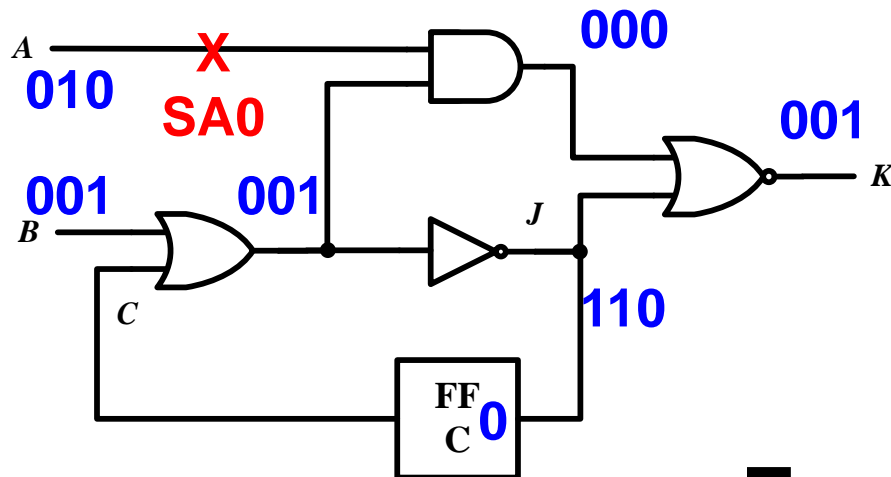


AB	cost = number of unknown FF
00	1
01	
10	

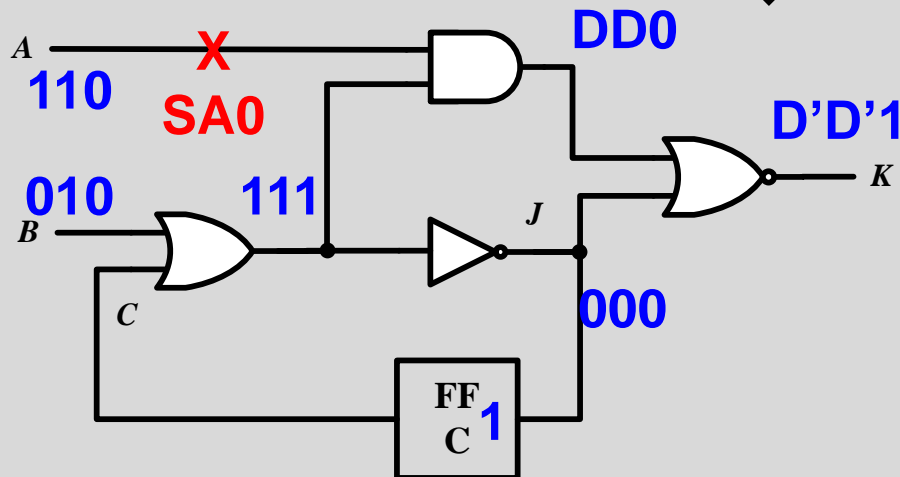
2. Concurrent Fault Detection Phase

- Start with fault simulation of generated initialization sequence
 - ◆ Detected faults are dropped from the fault list
- Compute the cost of the last vector
 - ◆ Cost of an undetected fault f
 - **COST(f)** = minimum distance of its fault effect to a PO
 - distance = level of logic gates
 - ◆ Cost of a vector
 - Sum of costs of all undetected faults
- Trial vectors are generated by **single-bit change**
 - ◆ Only accept the vectors that reduce the cost

Example



AB	cost = distance of D or D' to PO
00	∞
10	2 ←
01	∞

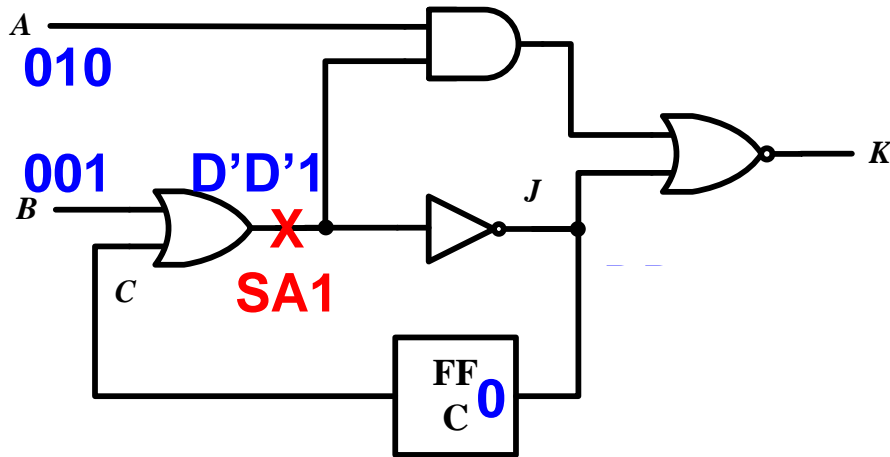


AB	cost = distance of D or D' to PO
10	0 ←
11	0 ←
00	∞

QUIZ

Q1: Given FF initial state is zero, please evaluate the cost of three trial vectors: 00, 10, 01

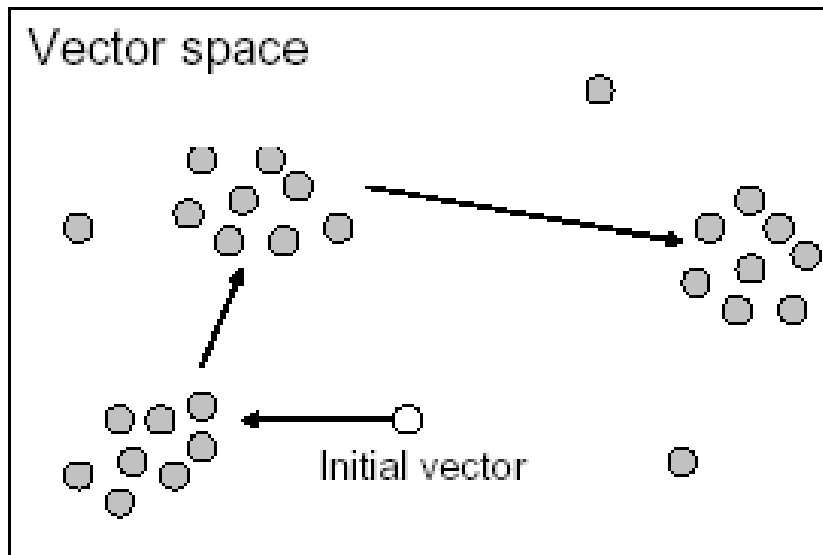
Q2: Which test vector would you pick?



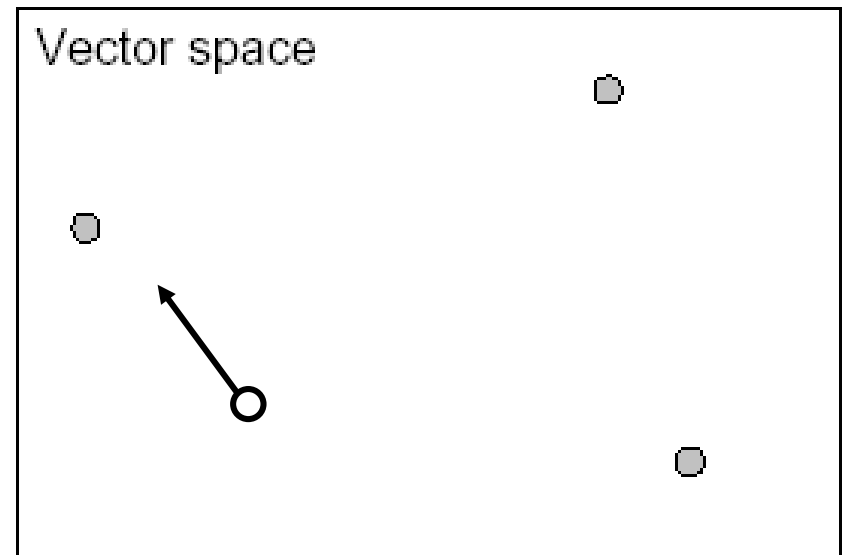
AB	cost = distance of D or D' to PO
00	
10	
01	

Need Phase 3

- Experience shows test patterns for all stuck-at faults are usually **clustered** instead of being evenly distributed
- When only a few faults are left, their tests will be isolated vectors and we need a different test generation strategy



Phase 2: Concurrent Fault Detection



Phase 3: Single Fault Detection

3. Single Fault Detection Phase

- Start with any vector
- Generate new vectors by single-bit change to reduce cost of the selected fault until it is detected
 - ◆ The lowest cost fault is picked first
- Cost of a fault f at signal line g is
 - ◆ If not activated yet:
 $K C_A(f) + C_P(f)$
 $K = \text{constant}$; $C_A = \text{activation}$; $C_P = \text{propagation cost}$
 - ◆ If activated:
 $\text{Min}(C_P(i))$, $i \in \text{the set of inputs to signal } g$



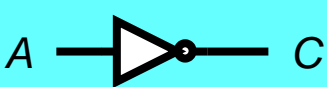
C_A and DC

- **Activation Cost, C_A**

- ◆ $C_A(g \text{ stuck-at-}v) = DC_v(g) = \text{dynamic controllability of line } g \text{ at } v'$

- **Dynamic Controllability, DC**

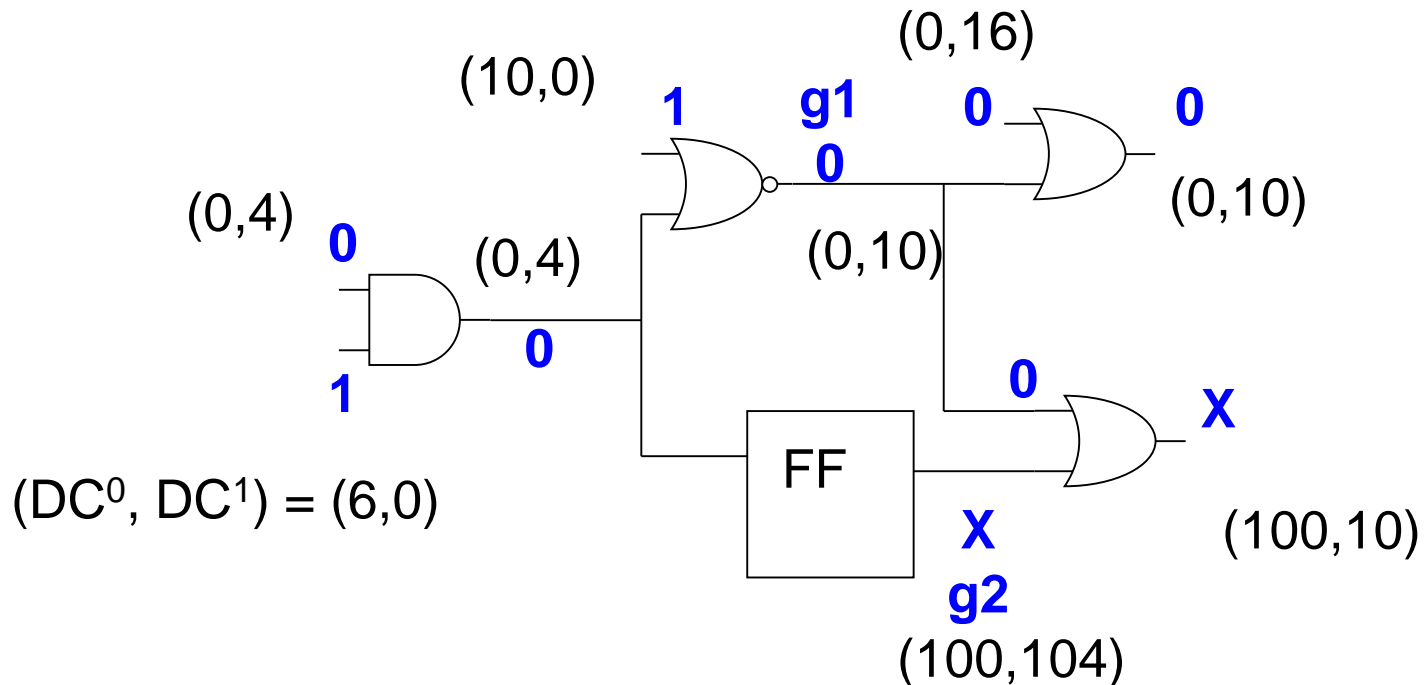
- ◆ Similar to **sequential controllability** in SCOAP except logic values known

	$DC^0(C)$	$DC^1(C)$
	$\min[DC^0(A), DC^0(B)]$, if $C=1$ or x 0 , if $C=0$	$DC^1(A) + DC^1(B)$, if $C=0$ or x 0 , if $C=1$
	$DC^0(A) + DC^0(B)$, if $C=1$ or x 0 , if $C=0$	$\min[DC^1(A), DC^1(B)]$, if $C=0$ or x 0 , if $A=1$
	$DC^1(A)$, if $C=1$ or x 0 , if $C=0$	$DC^0(A)$, if $C=0$ or x 0 , if $C=1$
Primary inputs	1 , if $C=1$ or x 0 , if $C=0$	1 , if $C=0$ or x 0 , if $C=1$
$C = FF(A)$	$DC^0(A)+K$, if $C=1$ or x 0 , if $C=0$	$DC^1(A)+K^*$, if $C=0$ or x 0 , if $C=1$

* K is a chosen constant

C_A and DC Example

- $C_A(g_1 \text{ stuck-at } 0) = DC_1(g_1) = 10 \leftarrow \text{easier}$
- $C_A(g_2 \text{ stuck-at } 1) = DC_0(g_2) = 100$



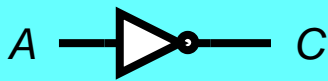
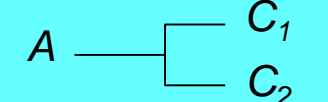


Propagation Cost, C_p

- $C_p(g)$ = Dynamic Observability of node g
- Dynamic observability (DO)
 - ◆ Similar to combinational observability in SCOAP
 - ◆ Measure the effort to observe the fault on a given node
 - the number of gates between N and PO's, and
 - the minimum number of PI assignments required to propagate the logical value on node N to a primary output.

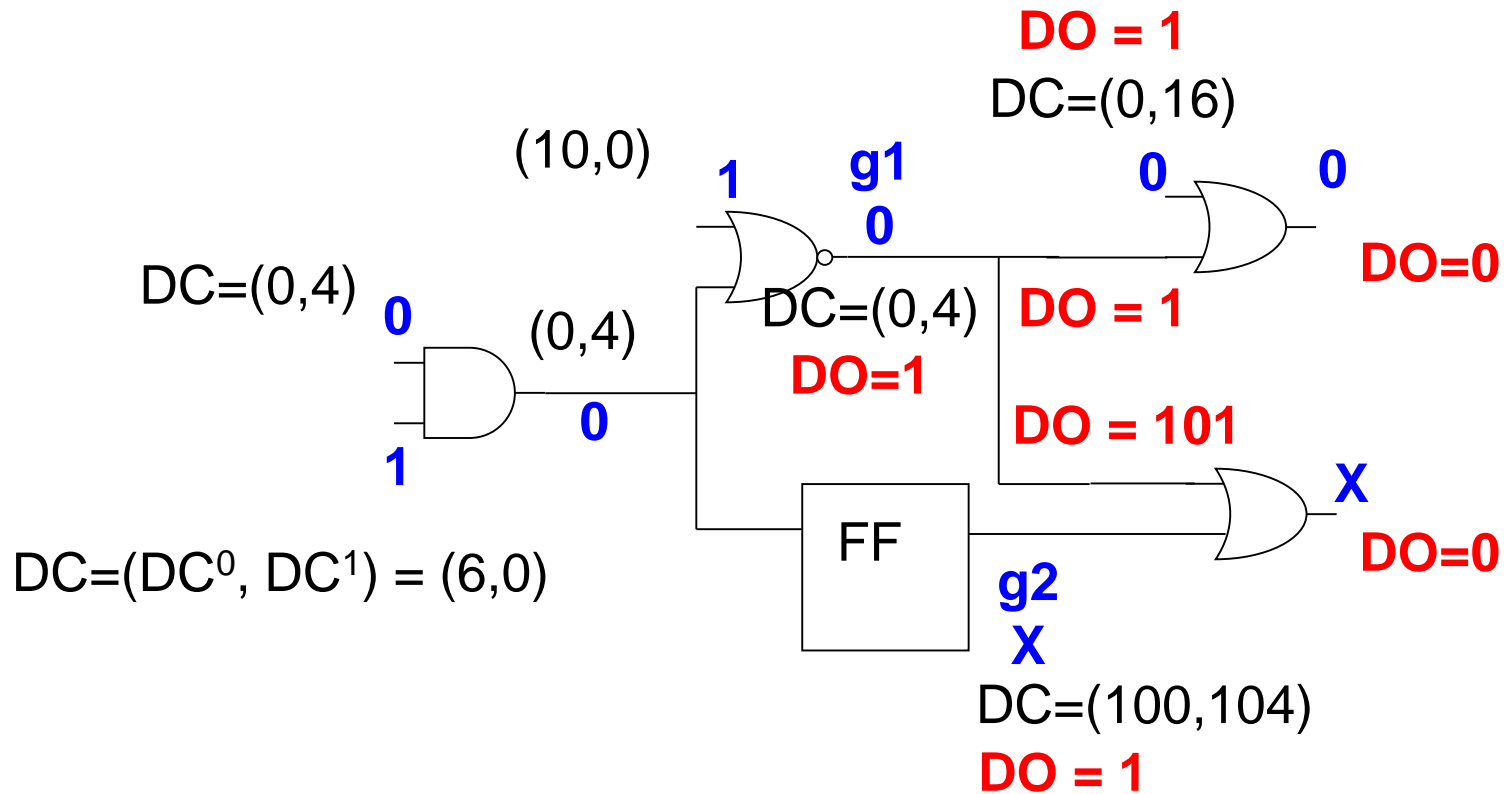
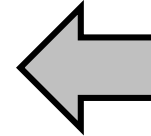
Dynamic Observability (DO)

- Similar to combinational observability in SCOAP

	DO(A)
	$DO(C) + DC^1(B) + 1$
	$DO(C) + DC^0(B) + 1$
	$DO(C) + 1$
	$\min[DO(C_1), DO(C_2)]$
Primary outputs	0

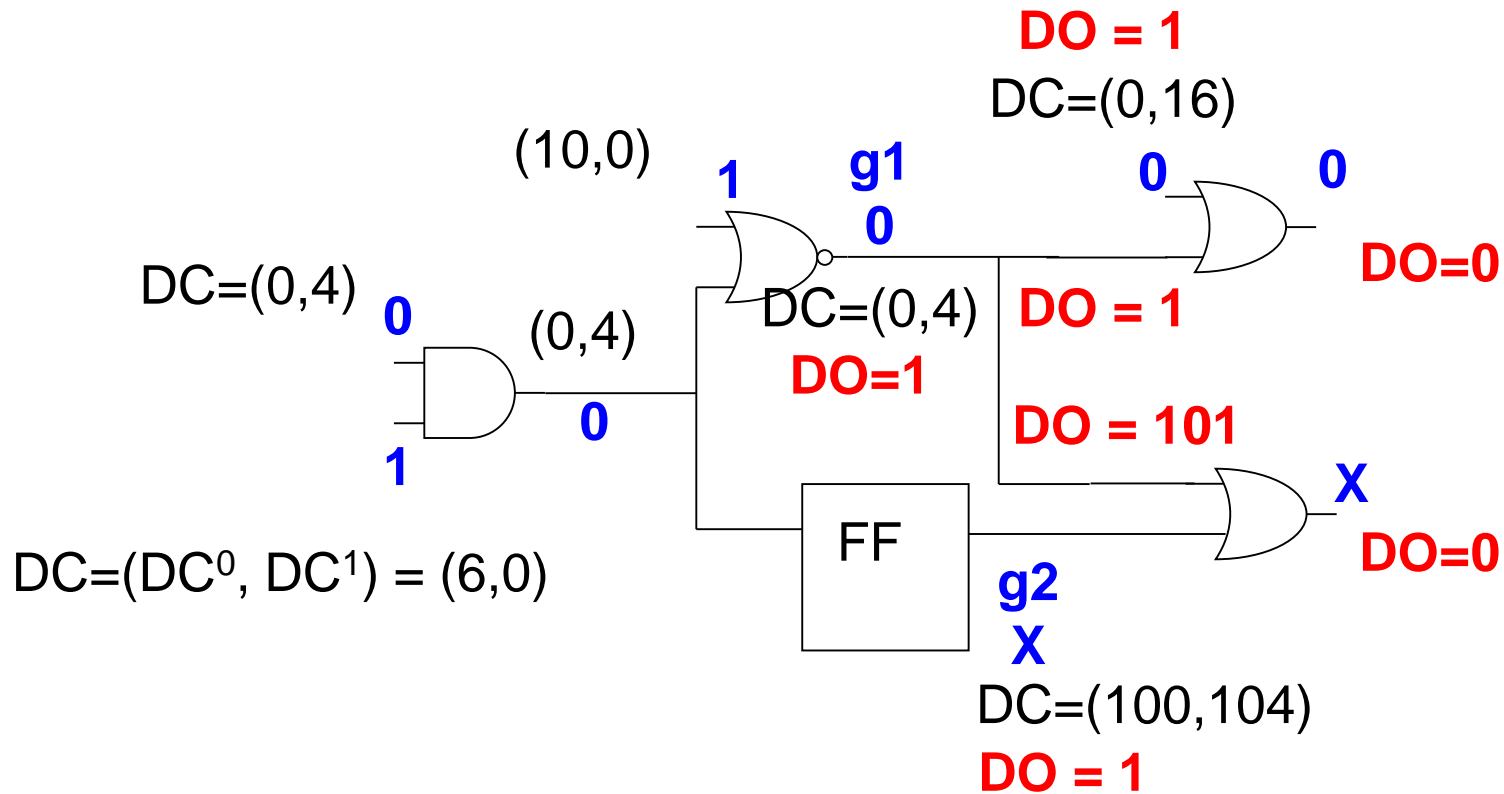
C_p and DO Example

- $C_p(g_1) = DO(g_1) = 1$
- $C_p(g_2) = DO(g_2) = 1$



Total Cost

- Fault g_1 : $C_A = 10$, $C_p = 1$
- Fault g_2 : $C_A = 100$, $C_p = 1$
- Choose **g1 SA0** as target fault to generate test vector



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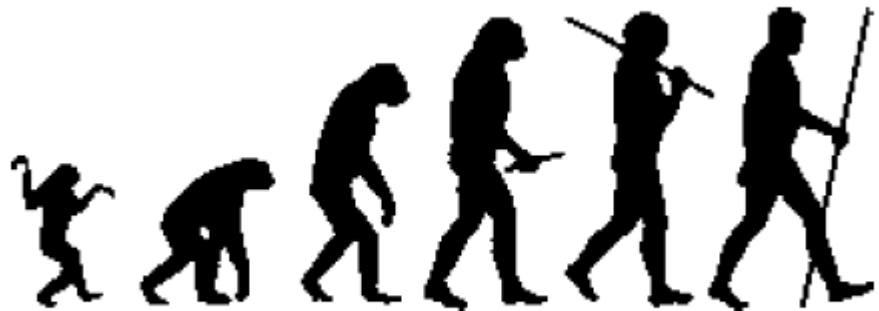
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CONTEST

Genetic Algorithm

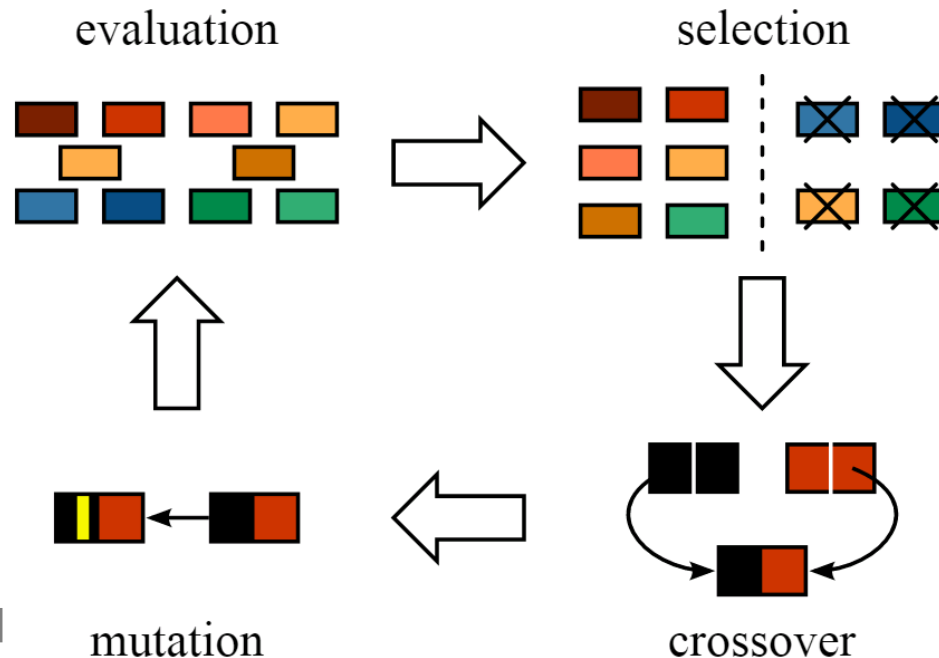
Issues of Sequential ATPG*

Conclusions



Genetic Algorithms (GA) [Holland 1975]

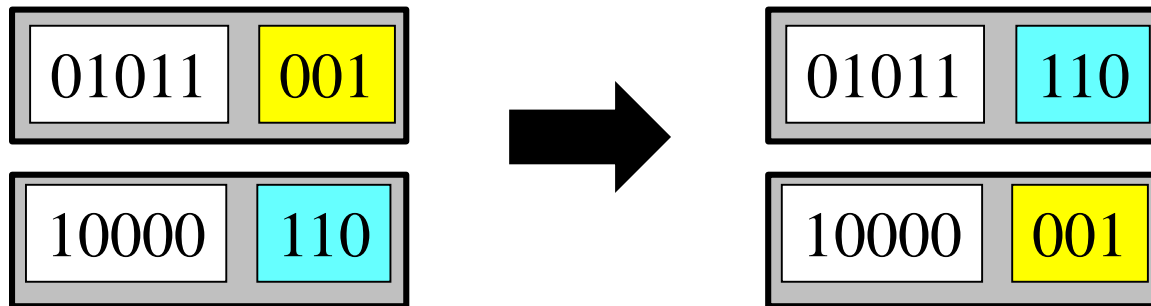
- General Principle: **Survival of fittest(s)**
 - ◆ Keep a **population** of feasible solutions, not just one
 - ◆ Parent population generates child population
 - by gene **crossover**, **mutation** etc
 - ◆ Select only best children, remove weak children
 - ◆ Repeat the above for many generations



[www.jade-cheng.com]

Crossover and Mutation

- Test vectors are represented by **bit-stream “gene”**
- **Crossover**: Two feasible solutions generate child by switching gene



- **Mutation**: some gene can change by a random probability

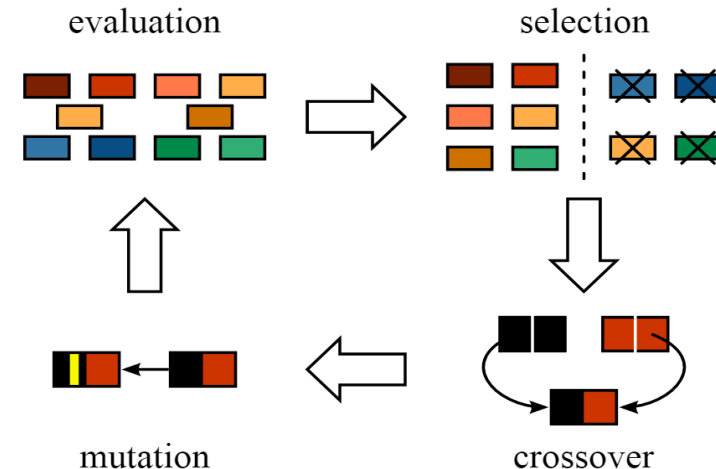


Pseudo Code of GA

GENETICALGORITHM

```
1  pop = set of initial solutions
2  do
3    childpop =  $\emptyset$ 
4    for ( $i = 1$  to ( $n \times \text{pop.size}$ )) //  $n$  times size
5      crossover = random 0 or 1
6      if (crossover)
7        parent1 = random_choose(pop)
8        parent2 = random_choose(pop)
9        child = crossover(parent1, parent2)
10     else // mutate
11       parent = random_choose(pop)
12       child = mutate(parent)
13     childpop = childpop  $\cup$  {child}
14   pop = evaluate&select(childpop)
15 while (!stop)
16 return (best solution)
```

- **Need to decide**
- **1. initial solution**
- **2. corssover/ mutation**
- **3. evaluate & select**
- **4. stop criterion**



Summary

- **Simulation-based methods**
 - ◆ Randomly generate many trial test vectors
 - ◆ Evaluate test vectors by simulation and pick the best
 - ◆ Need many testability measure to help smart decision
- **Advantages**
 - ◆ Better **memory management** than time frame expansion
 - ◆ **Timing** can be considered
 - ◆ Use **genetic algorithm** to optimize
- **Disadvantages**
 - ◆ Cannot identify **untestable faults**
 - ◆ Test length can be **longer** than time frame expansion