

# Topics for this Lecture

- Automatic Test Generation Approaches
  - ~~Random Testing~~ (Guess)
  - Search Based Software Testing ✓✓ (Search)
  - Constraint-Based(symbolic execution) Testing (Deduce)



1 2 3 4 5 6 7 8 9 10



Guess



# Problem Definition

```
float divide(float x, float y){  
    if (y == 0) {  
        // Example handling of this error. Writing a message to stderr, and exiting with failure.  
        fprintf(stderr, "Division by zero! Aborting...\n");  
        return EXIT_FAILURE; /* indicate failure.*/  
    }  
    return x / y;  
}
```



# Problem Definition

```
double divide(float x, float y){  
    if (y == 0) {  
        // Example handling of this error. Writing a message to stderr, and exiting with failure.  
        fprintf(stderr, "Division by zero! Aborting...\n");  
        return EXIT_FAILURE; /* indicate failure.*/  
    }  
    return x / y;  
}
```



## Test Case#1

```
void testcase1(){  
    float r=divide(3.0,2.0);  
    assert(r == 1.5);  
}
```

## Test Case #2

```
void testcase2(){  
    float r=divide(0.0,0.0);  
    assert (r == EXIT_FAILURE);  
    exit(EXIT_FAILURE); /* indicate failure.*/  
}
```



# Problems

1. How to generate the input data?
2. How to generate the assertions and detect if there is a bug?

We can manually generate some inputs. (Think)



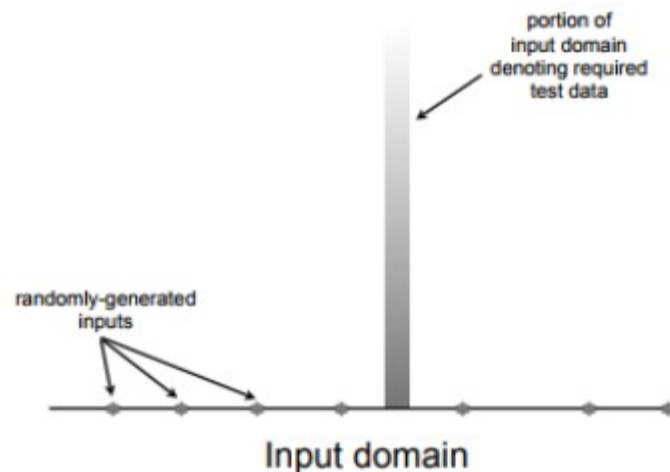
Test Case#1	Test Case #2
<pre>void testcase1(){   float r=<b>divide</b>(3.0,2.0);   assert(r == <b>1.5</b>); }</pre>	<pre>void testcase2(){   float r=<b>divide</b>(0.0,0.0);   assert (r == <b>EXIT_FAILURE</b>);   exit(EXIT_FAILURE); /* indicate failure.*/ }</pre>

# Random Testing (RT)

1. Write code to randomly choose a method
2. Generate random values for each parameter for the chosen method.
3. Execute/invoke the method.

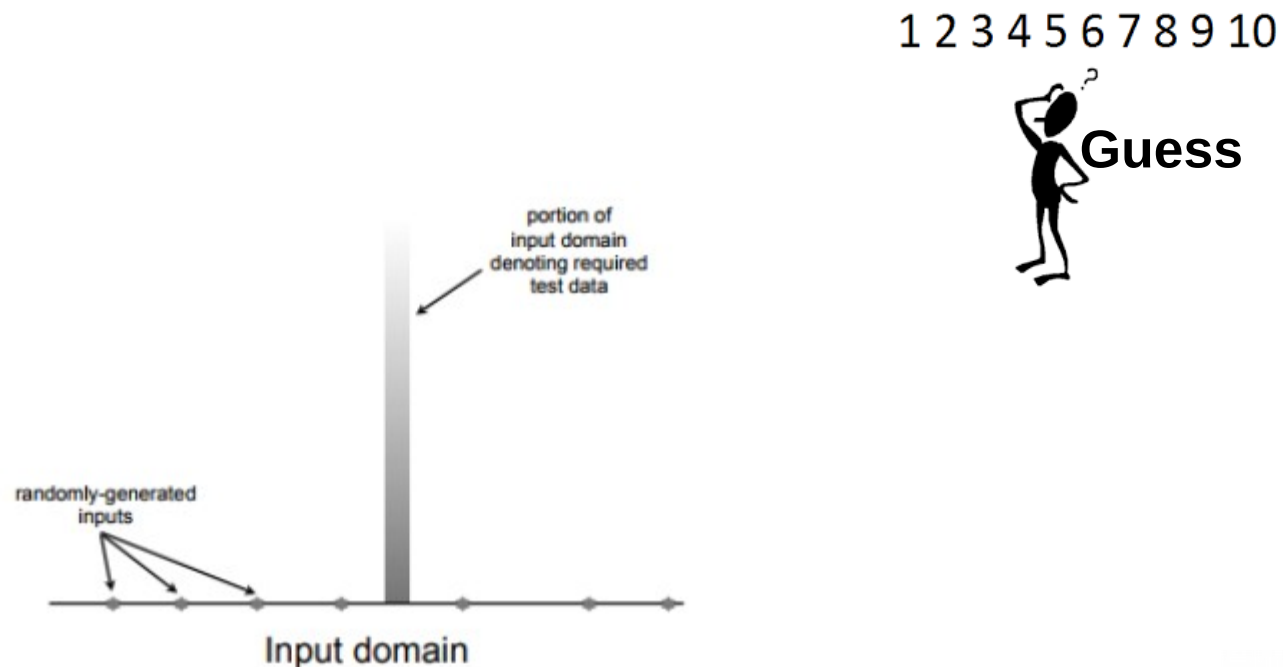
- **Problems:**

1 2 3 4 5 6 7 8 9 10



P. McMinn, "Search-based software testing: Past, present and future," 2011

# Can we be **smarter** than Random Testing (RT)?



P. McMinn, "Search-based software testing: Past, present and future," 2011

# Search Based Software Testing (SBST)

- The problem of test data generation is **converted** to a optimization/search problem, and search algorithms such as Hill Climbing (**HC**) or Genetic Algorithm (**GA**) are used to find the best test data.
- What to optimize? e.g., code coverage (i.e., branch coverage)



# Search Based Software Testing (SBST)

- There are three requirements that need to be fulfilled in order to apply a search optimization technique.



- 1. Search Space:** The set of all possible inputs for a program under test. e.g., divide (float x, float y)
- 2. Search operators.** How to modify the tests to explore the search space. e.g., remove a method call, add a random method call, modify a parameter (e.g., + or -1 for integer values), etc.
- 3. Fitness Function (FF).** Evaluate how good a solution (i.e., test case) is and guides the search toward better solutions.



# Fitness Function (Branch Distance)

- **Fitness Function:** it estimates how close a candidate input (i.e., test case) is to satisfy a coverage goal. A common FF is **Branch Distance**.
- **Branch Distance:** how near the input was to executing the required branch
- Assume you want to solve  

```
if (y==0){/* this our target branch */}
```
- You have test values 1 and 100
- Neither of them solve the  $y==0$ . In other word, both values do not cover the target branch (i.e., the true if statement) because only 0 solves the constraint of the target branch .
- But, 1 is closer than 100 to 0



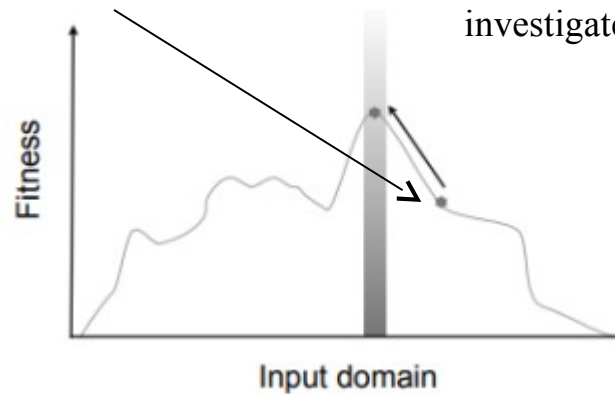
# Local Search: Hill Climbing (HC)

- local search algorithms aim to improve one individual by exploring its neighbors.

1- Hill Climbing (HC) usually starts with a random input value

2- Then it considers the set of near neighbors to this input value

3- If a fitter neighbor is found, HC moves to it and again it investigates its neighbors



(space of all possible solutions)

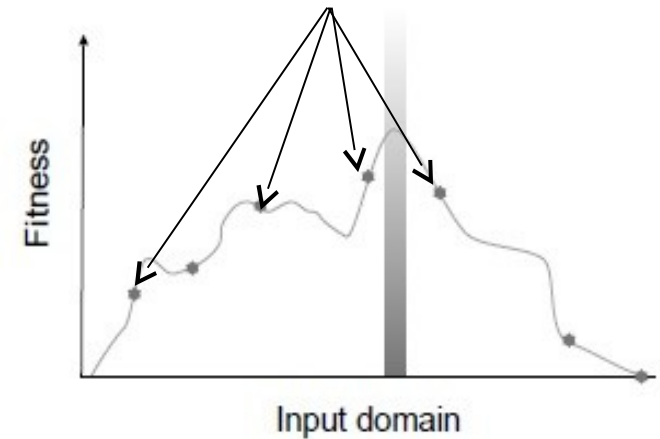
# HC example

- Assume that you want to solve
  - `if (y==0) { /* this our target branch */ }`
- Starting with random values  $y=10$
- The **branch distance (BD)** is  $|y-0| = |10-0|=10$
- Neighborhood  $\pm 1$  :  $y=9$  and  $y=11$ 
  - The **BD** is  $|y-0| = |9-0|$  and  $|11-0|$  which is 9 and 11
  - 9 is closer to 0  $\rightarrow$  moving to  $y=9$  (new value), and discard 11
- Repeat for  $y=9$ , i.e.,  $\pm 1$ :  $y=8$  and  $y=10$ 
  - The **BD** is  $|y-0| = |8-0|$  and  $|10-0|$  which is 8 and 10
  - 8 is closer to 0  $\rightarrow$  moving to  $y=8$  (new value) and discard 10
- Repeat until the solution  $y=0$



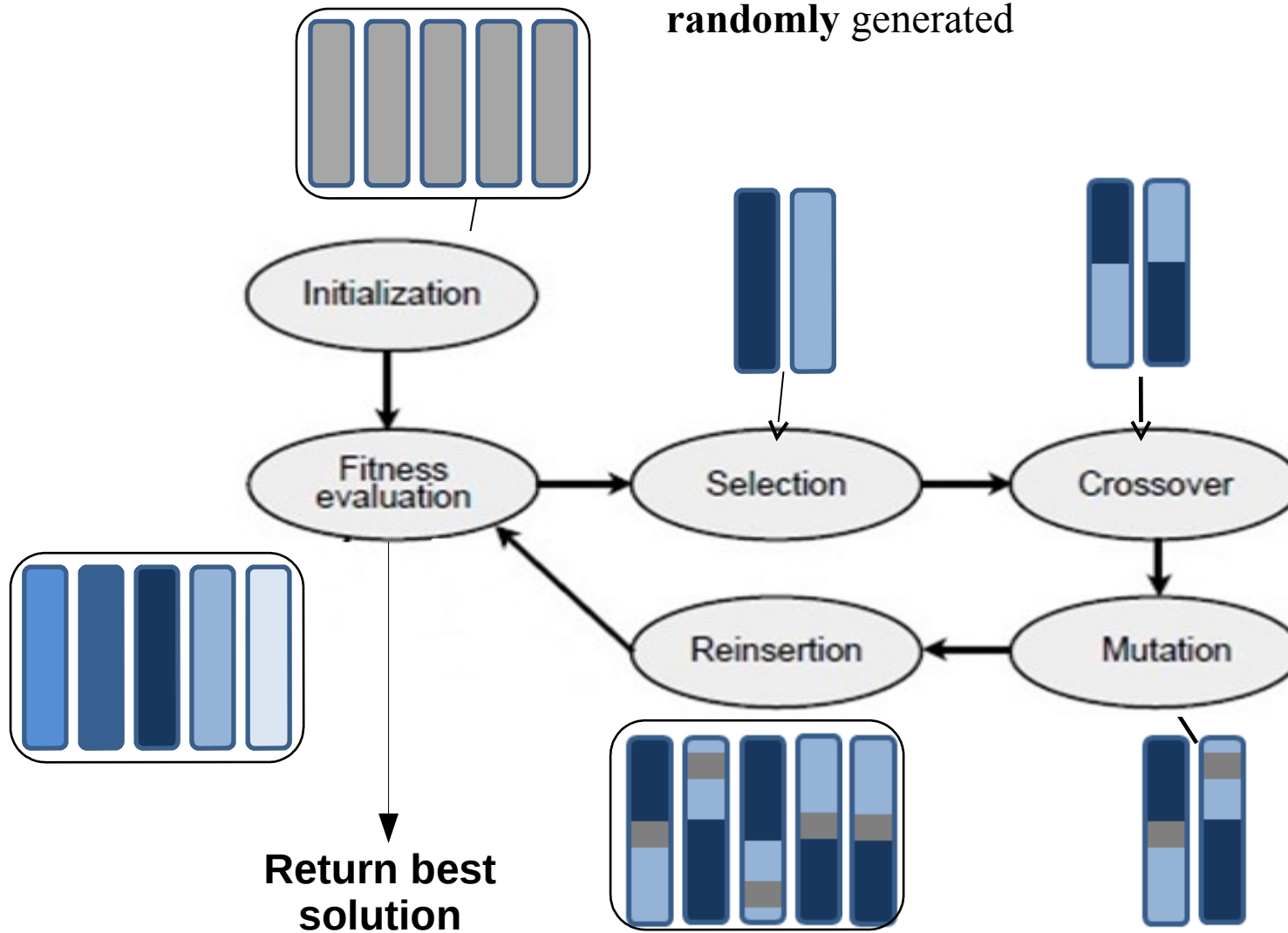
# Genetic Algorithm (Global Search)

- **Global** search, sampling **many points** in the search space at once
- **Global** search is more effective than **local** search, but less efficient, as it is more costly.
- The most commonly applied global search algorithms is a ***Genetic Algorithm*** (GA).
- A GA tries to imitate the natural processes of evolution
- Genetic Algorithms are inspired by Darwinian evolution and the concept of survival of the fittest.
- **Evolution** is change in the heritable characteristics of biological populations over successive generations

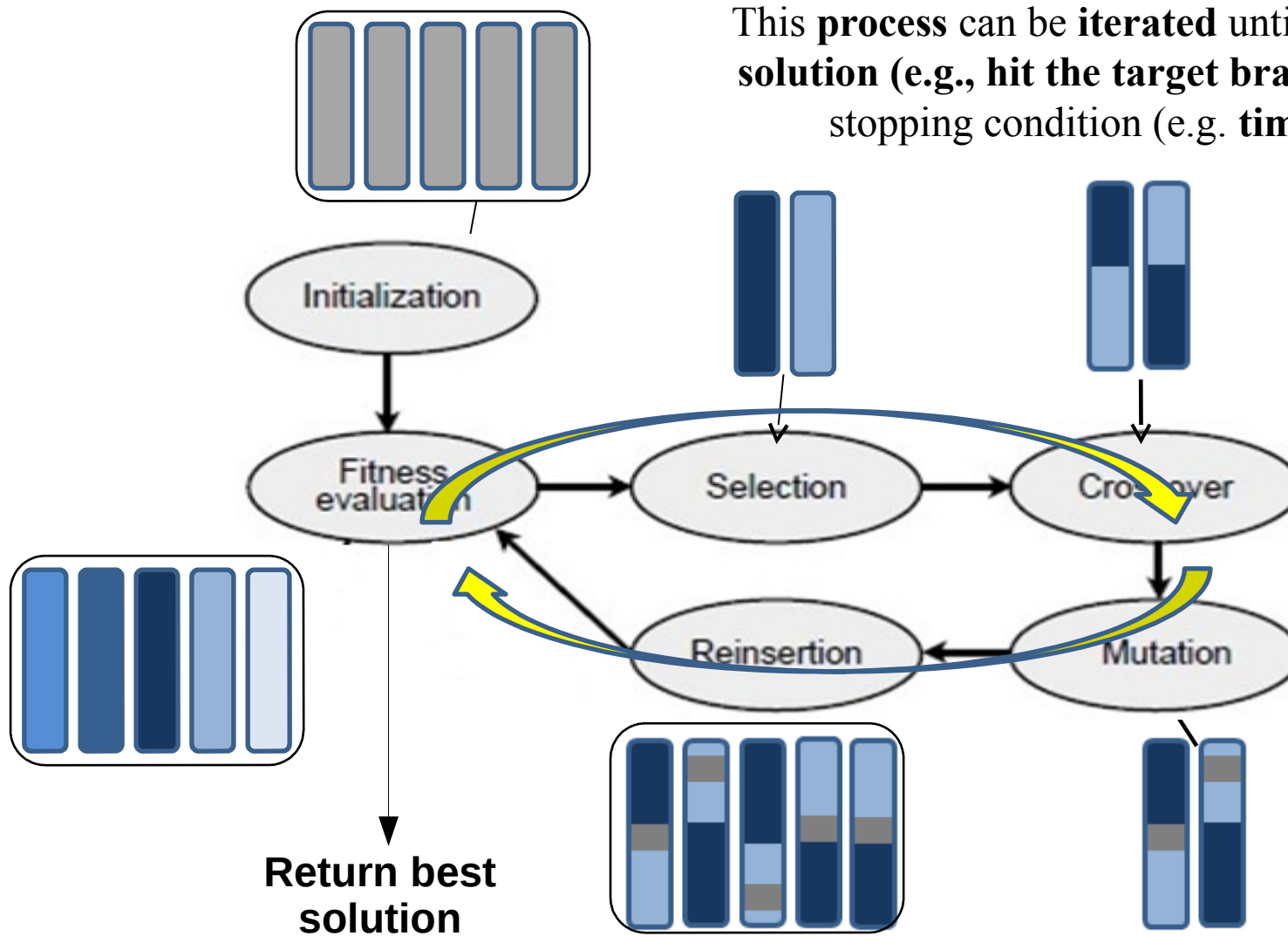


# Global Search (Genetic Algorithm (GA))

- Initial population (i.e., values/test cases) is randomly generated



# Global Search (Genetic Algorithm (GA))



# GA vs HC

- Assume that you want to solve  
if  $(x==3 \ \&\& \ x*y=72)\{/* \text{ this our target branch */}\}$
- We have two constraints to solve which are  $x=3$  and  $x*y=72$
- Lets say that you start point is  $x=9$  and  $y=8$
- $+1$  or  $-1$  on  $x$  or  $y$  would not improve your solution and HC get stuck.
- Having a population (more than one solution), GA can overcome these local problems.

# Some automatic SBST input generation tools for Java

- **Java**

- **EvoSuite** tool: (<http://www.evosuite.org/>)

- **AVMf** tool: (<http://avmframework.org/>)

- **C Language**

- **AUSTIN** tool (<https://github.com/kiranlak/austin-sbst>)

- They are freely available for use





## References:

Noraini, Mohd Razali, and John Geraghty. "Genetic algorithm performance with different selection strategies in solving TSP." (2011).

[https://en.wikipedia.org/wiki/Mutation\\_\(genetic\\_algorithm\)](https://en.wikipedia.org/wiki/Mutation_(genetic_algorithm))