

SWVE3002-42: Introduction to Software Engineering

Lecture 11 – Grey-Box Testing

Sooyoung Cha

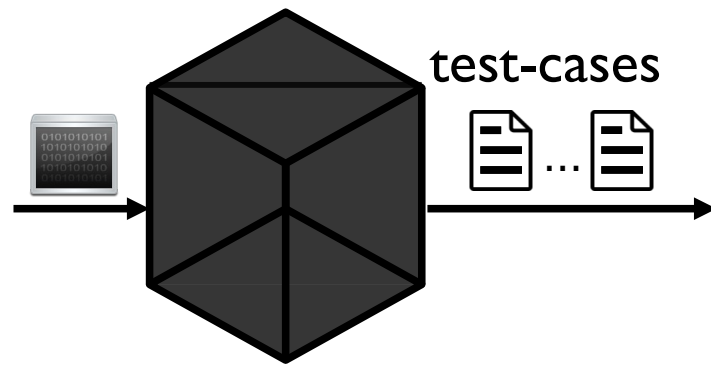
Department of Computer Science and Engineering

Today's Lecture

- Grey-Box Testing (Fuzzing)
 - Mutation-based Fuzzing
 - AFL: A representative mutation-based fuzzer
 - Recent trend in mutation-based fuzzing
 - Program-adaptive mutation-based fuzzing
 - Generation-based Fuzzing
 - Grammar generation
 - Test-case generation from grammar

Black-Box vs White-Box Testing

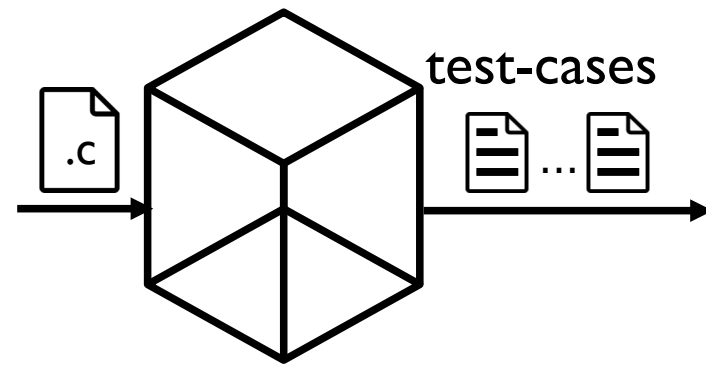
- Difference: **with/without** using of source code



“**without** source code”

(+) cheap

(-) naive



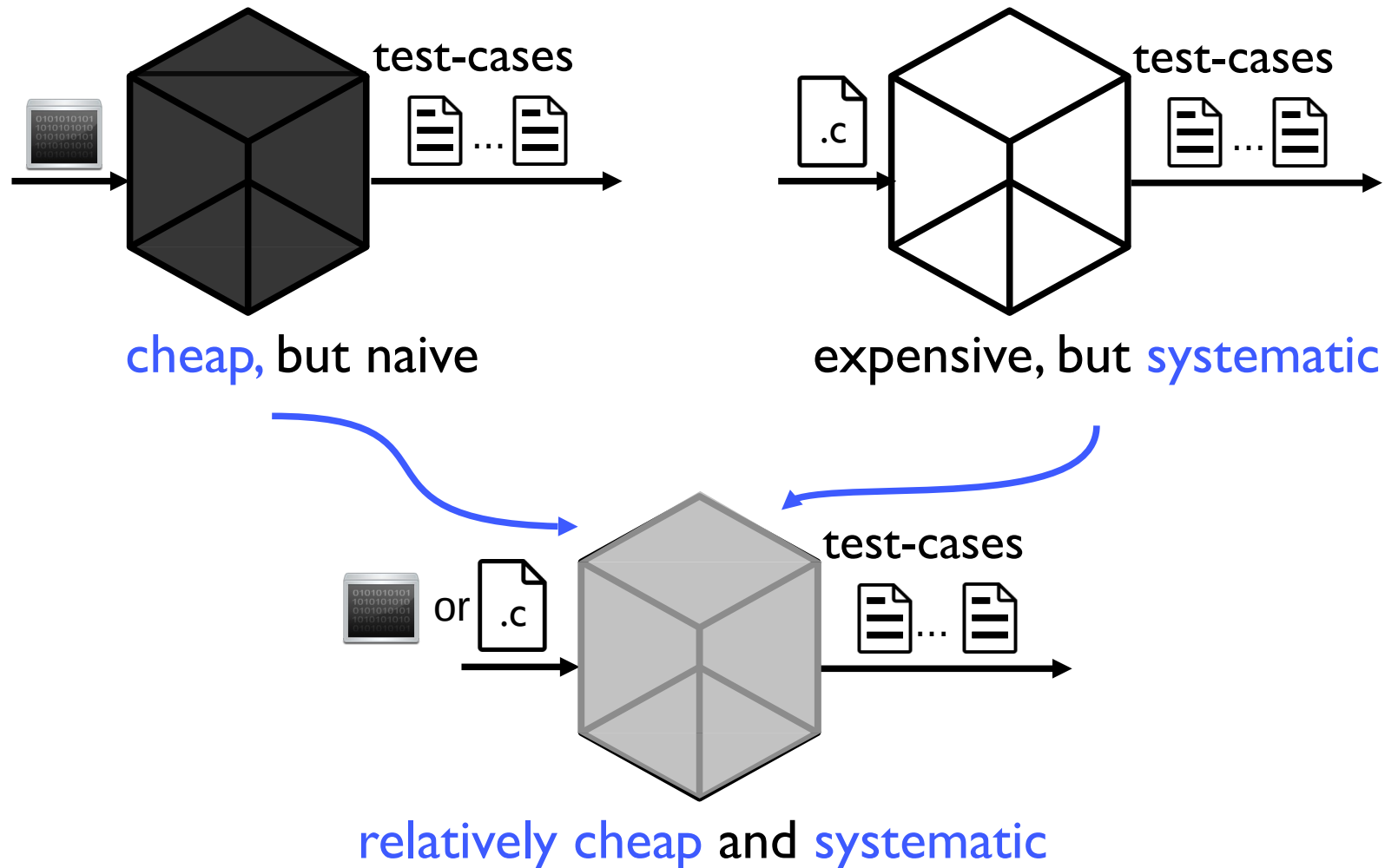
“**with** source code”

(-) expensive

(+) systematic

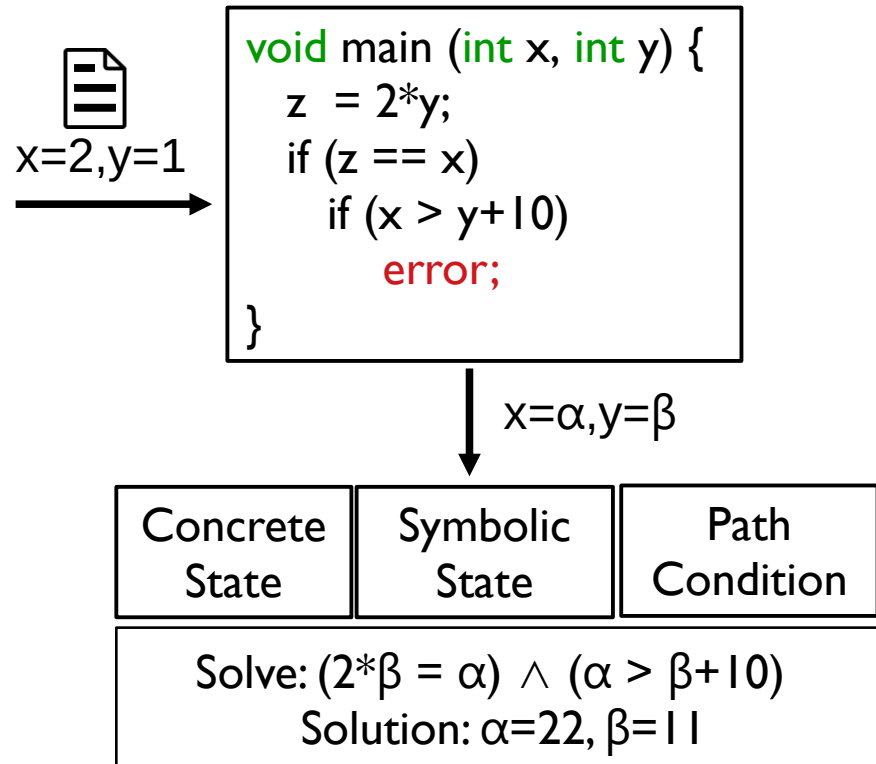
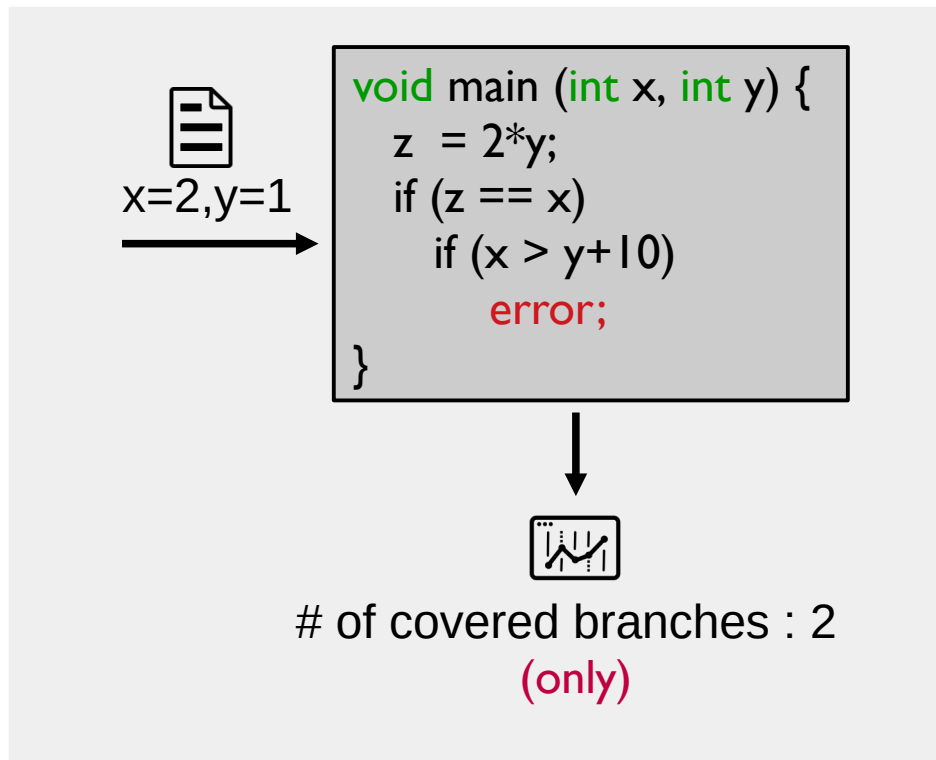
Grey-Box Testing (Fuzzing)

- Include the characteristics of black- and white-box testing.



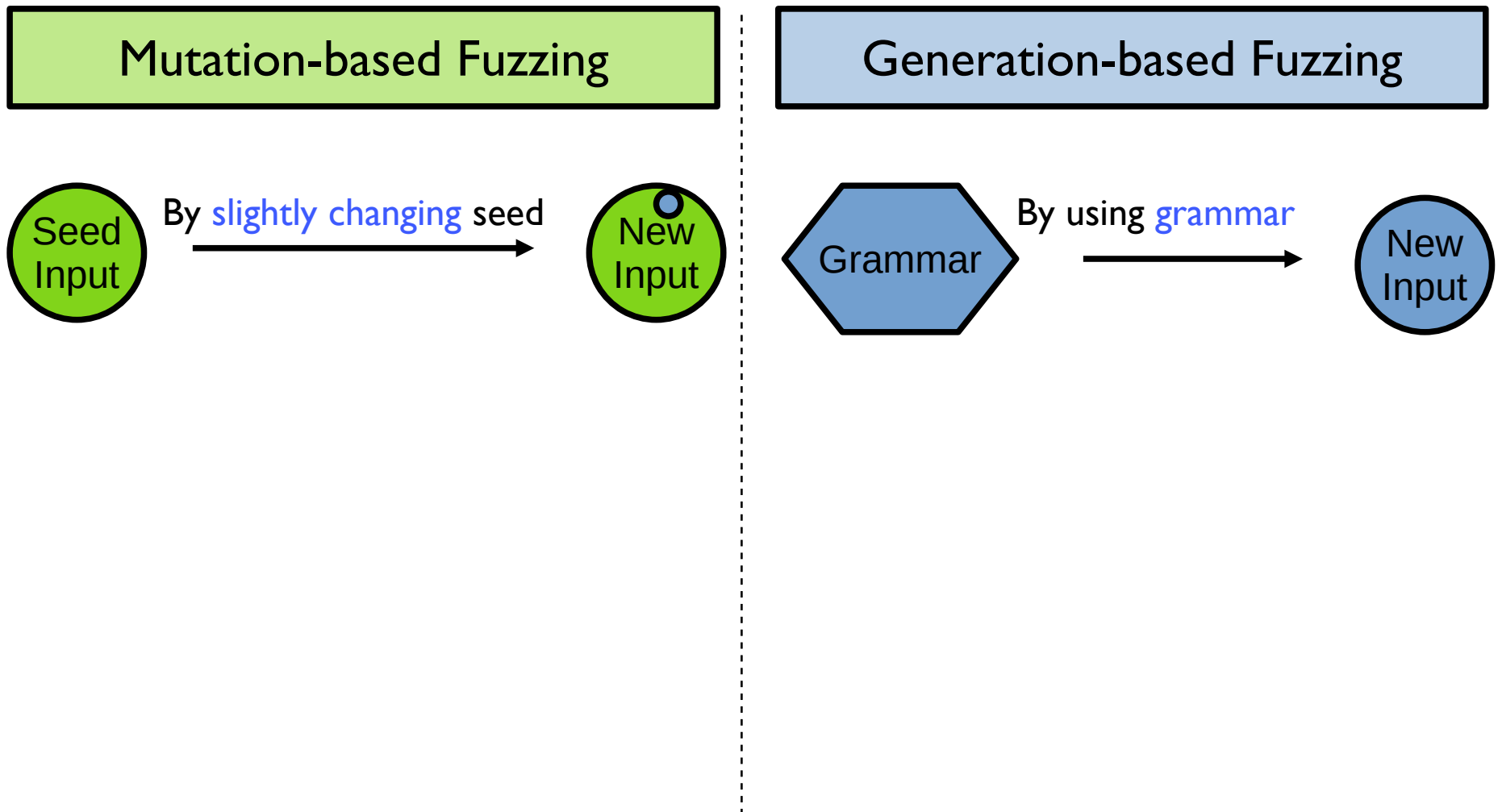
Grey-Box vs White-Box Testing

- Key Difference: using source code **partially/fully**.
- Grey-box Testing
 - **Obtaining only partial information from** each program execution.



Grey-Box Testing (Fuzzing)

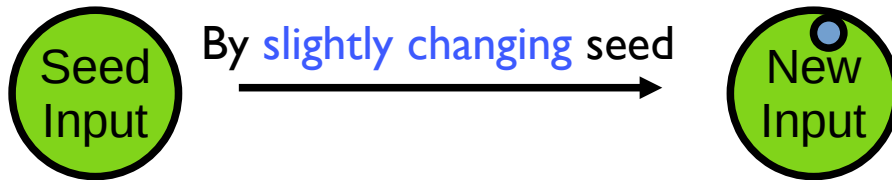
- Categorize into two fuzzing techniques



Grey-Box Testing (Fuzzing)

- Categorize into two fuzzing techniques

Mutation-based Fuzzing



Seed Input = "0 1 0 0 0 0 0 1"

↓ *"flip the first bit of seed input"*

New Input = "1 1 0 0 0 0 0 1"

Generation-based Fuzzing



Grey-Box Testing (Fuzzing)

- Categorize into two fuzzing techniques

Mutation-based Fuzzing

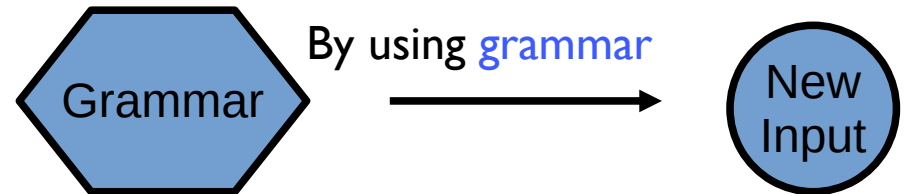


Seed Input = "0 1 0 0 0 0 0 1"

"flip the first bit of seed input"

New Input = "1 1 0 0 0 0 0 1"

Generation-based Fuzzing



<Grammar>

$S \rightarrow A C$

$A \rightarrow B C$

$C \rightarrow D | E$

New Input = "BDD"

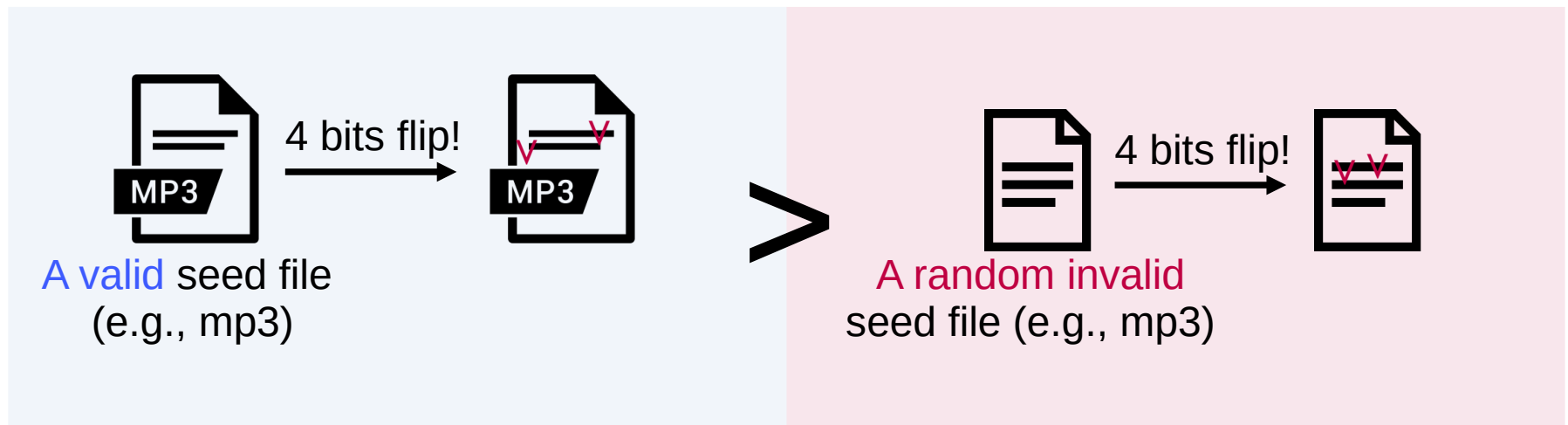
A parse tree for the string "BDD" using the grammar rules. The root node is S, which expands to A and C. A expands to B and C. The first C expands to D. The second C expands to D. The leaf nodes are B, D, and D, which form the string "BDD".

```
graph TD
    S --> A1[A]
    S --> C1[C]
    A1 --> B[B]
    A1 --> C2[C]
    C1 --> D1[D]
    C2 --> D2[D]
```


Mutation-based Fuzzing

- Intuition

- Mutating **only a fraction of a “valid” seed** may generate good inputs!



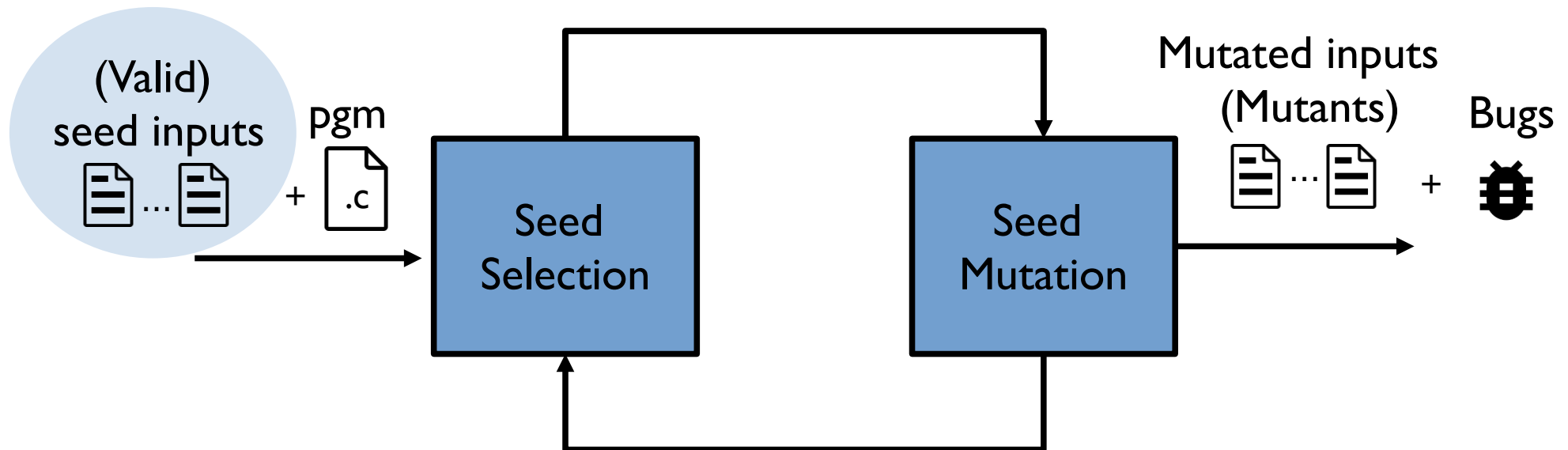
Detail of an MP3 Header

Bits	1	2	3	4	5	6	7	8	9	10	11	12	13	14	15	16	17	18	19	20	21	22	23	24	25	26	27	28	29	30	31	32				
Binary	1	1	1	1	1	1	1	1	1	1	1	1	1	0	1	1	1	0	1	0	0	0	0	0	0	1	0	0	0	0	0	0				
Hex	F			F			F			B							A					0					4					0				
Meaning	MP3 Sync Word												Version	Layer	Error Protection	Bit Rate					Frequency		Pad. Bit	Priv. Bit	Mode		Mode Extension (Used With Joint Stereo)		Copy	Original	Emphasis					
Value	Sync Word												1 = MPEG	01 = Layer 3	1 = No	1010 = 160					00 = 44100 Hz		0 = Frame is not padded		Unknown	01 = Joint Stereo		0 = Intensity Stereo Off	0 = MS Stereo Off	0 = Not Copyrighted	0 = Copy Of Original Media	00 = None				

<Mp3 header>

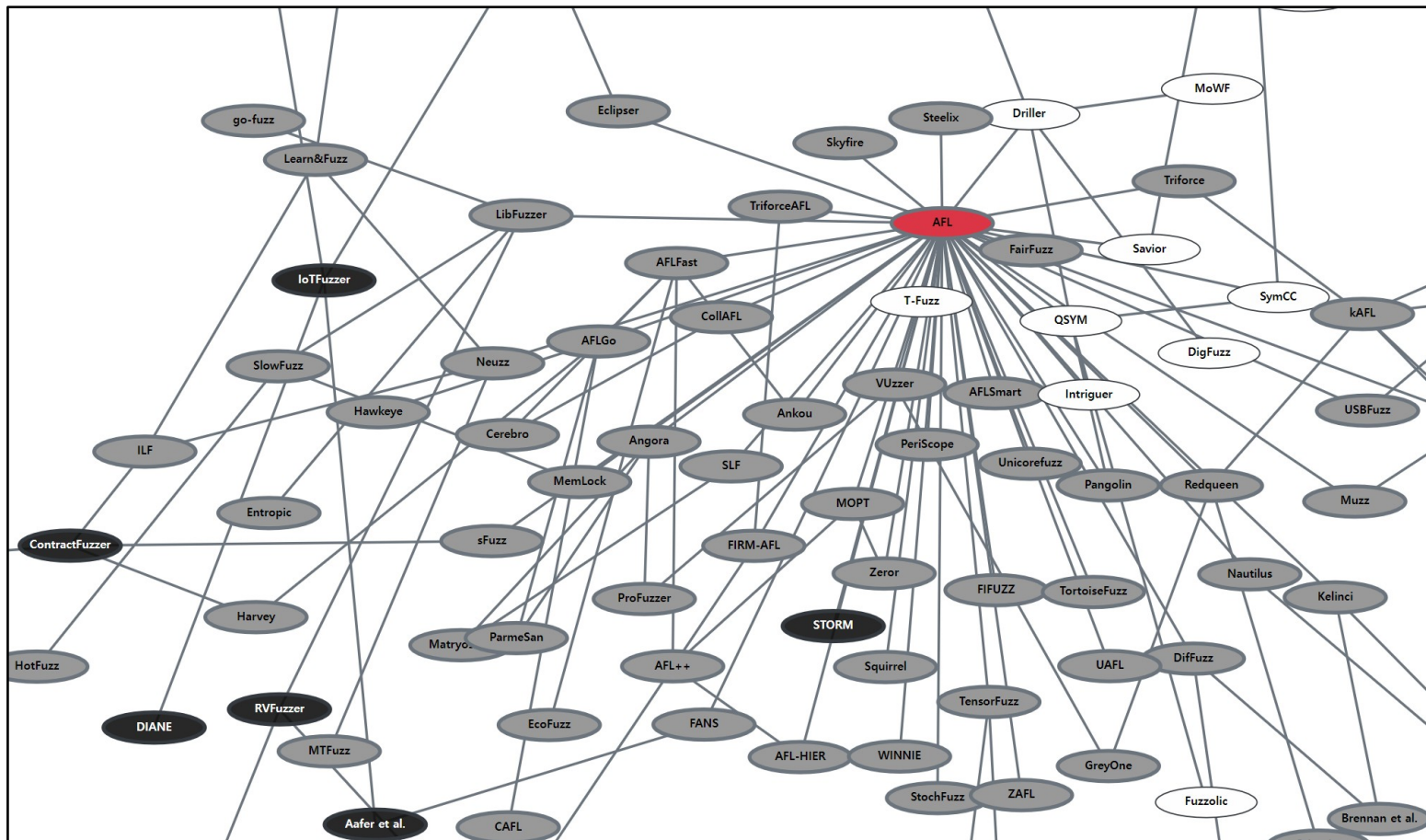
Mutation-based Fuzzing

- Key component
 - Seed Selection & Seed Mutation



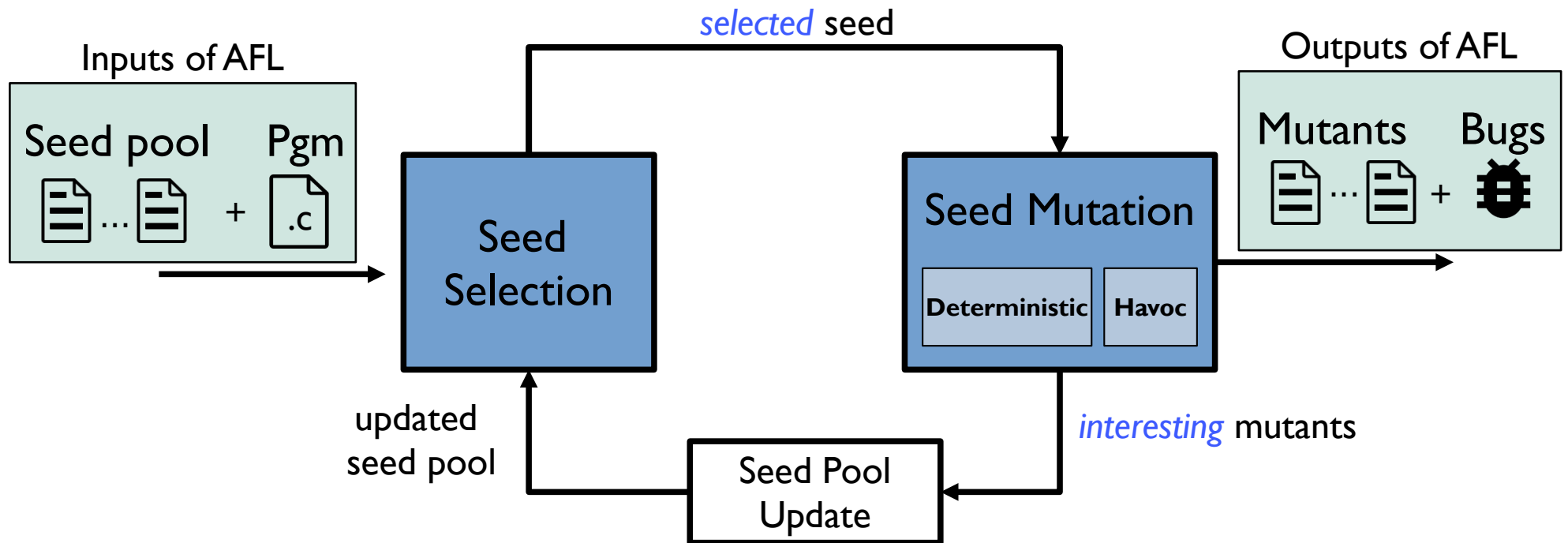
AFL (American Fuzzy Lop)

- A representative mutation-based fuzzer
 - Numerous techniques have been implemented on top of AFL!



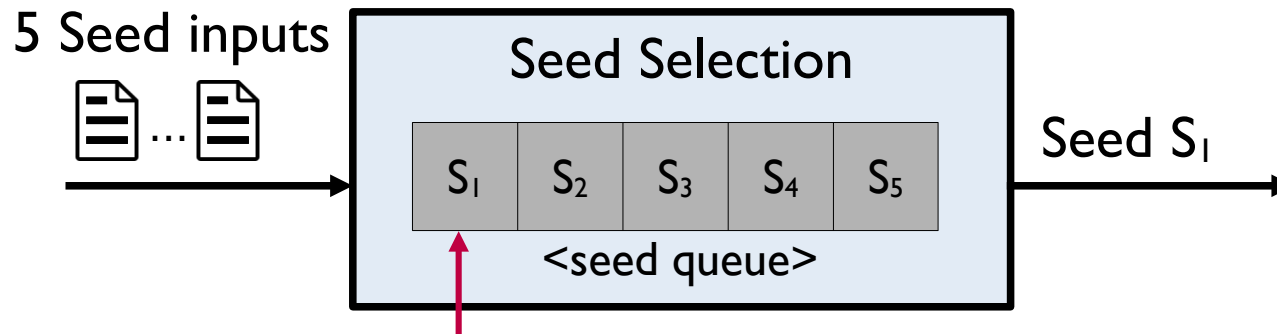
AFL (American Fuzzy Lop)

- Workflow of AFL



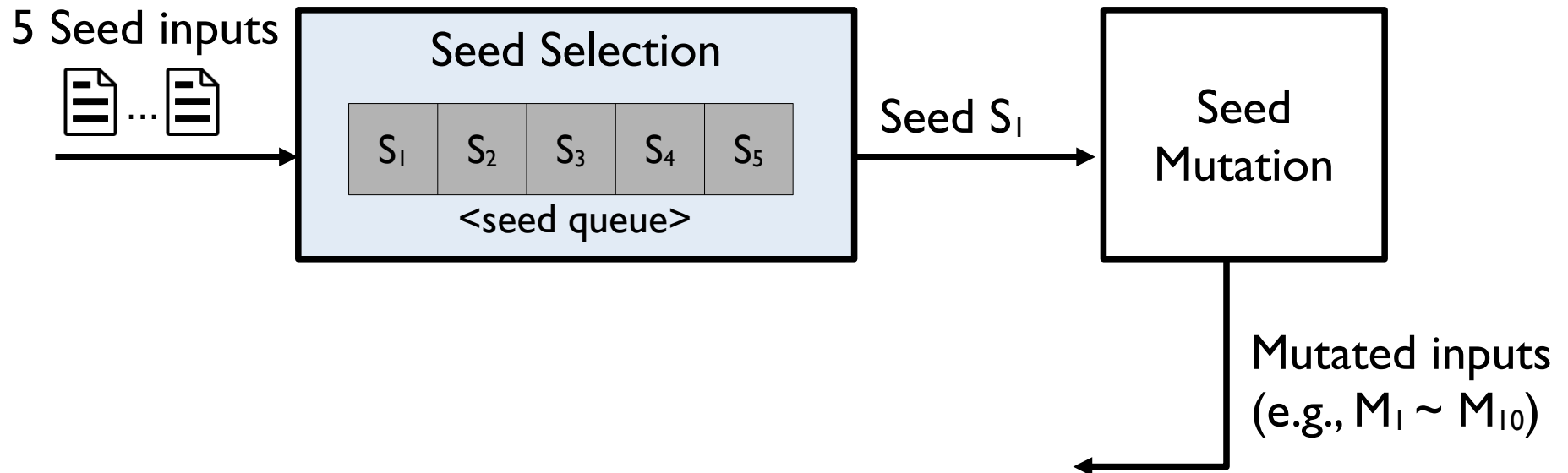
Seed Selection

- Load a set of seed inputs into a queue.
- Sequentially select a seed from the queue.



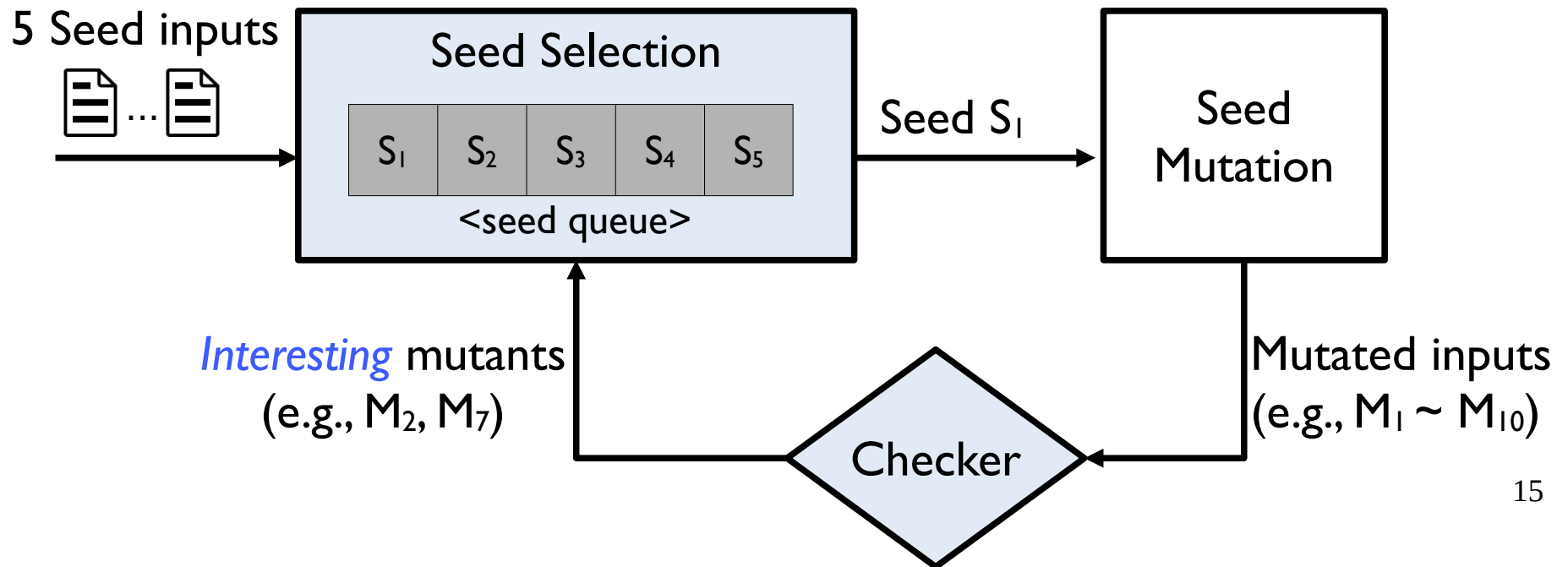
Seed Selection

- Load a set of seed inputs into a queue.
- Sequentially select a seed from the queue.
- Generate the mutated inputs by using the seed.



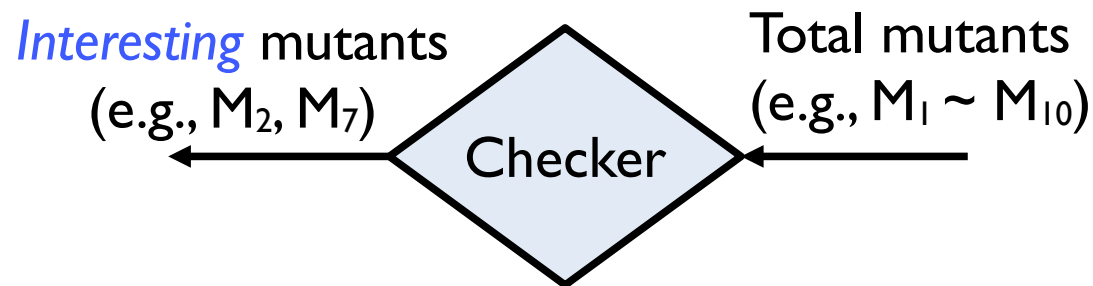
Seed Selection

- Load a set of seed inputs into a queue.
- Sequentially select a seed from the queue.
- Generate the mutated inputs by using the seed.
- Check whether the mutated inputs are interesting



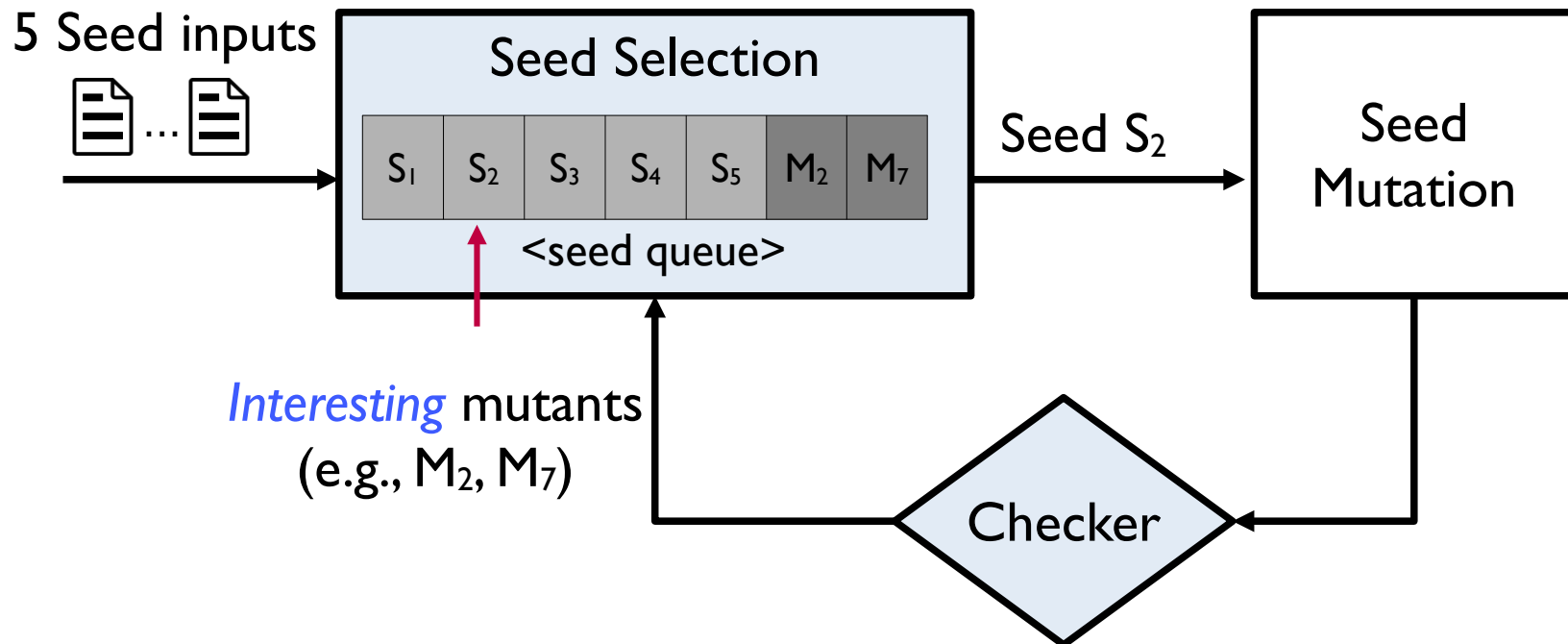
Checker

- Two criteria for **interesting** inputs.
 - The input **causes a crash** in the target program.
 - The input **covers a new path** which has not been covered.



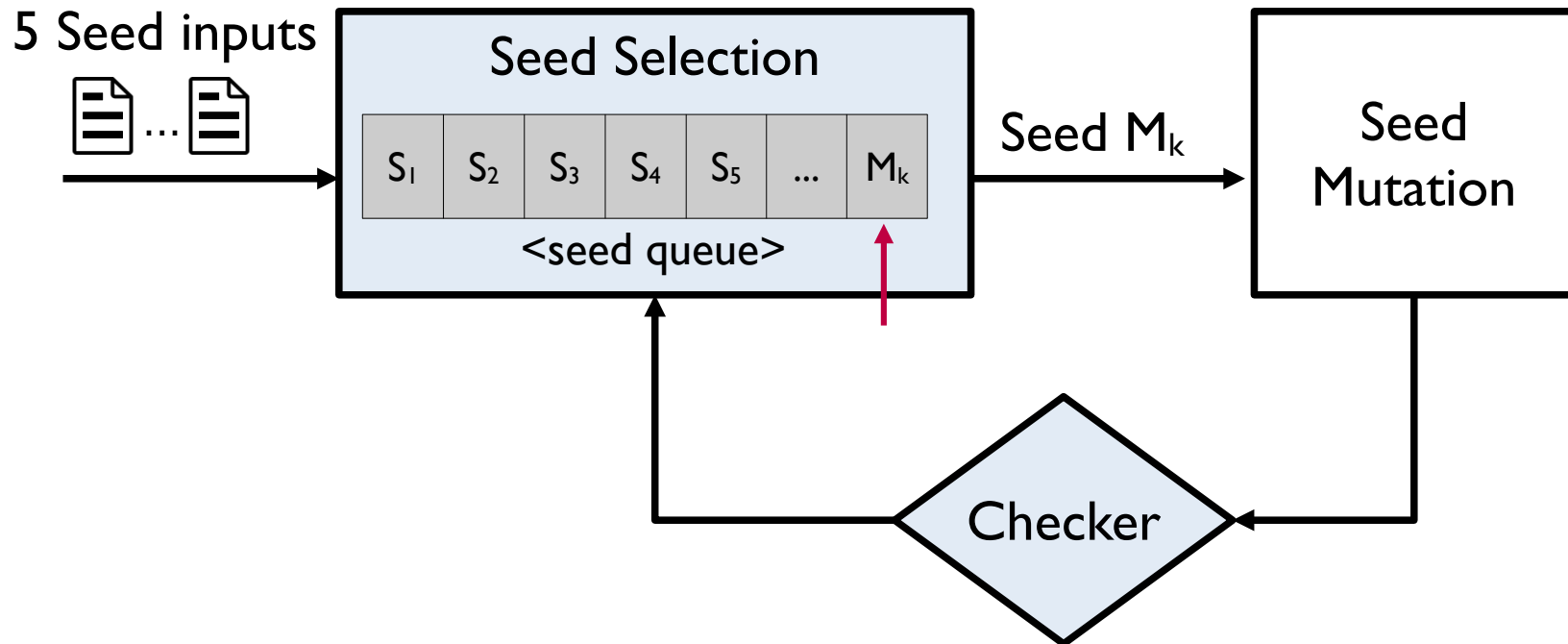
Seed Selection

- Add the interesting inputs into the queue.
- Sequentially select the next seed from the queue.



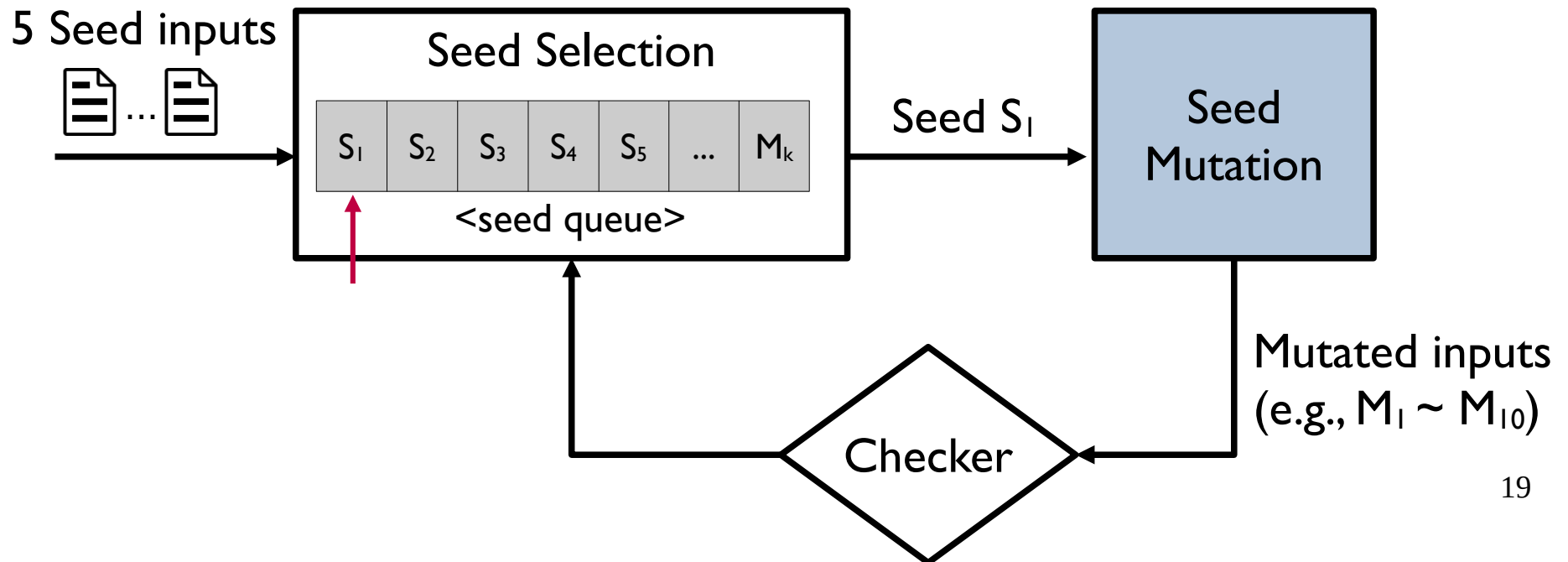
Seed Selection

- Add the interesting inputs into the queue.
- Sequentially select the first seed in the queue again.



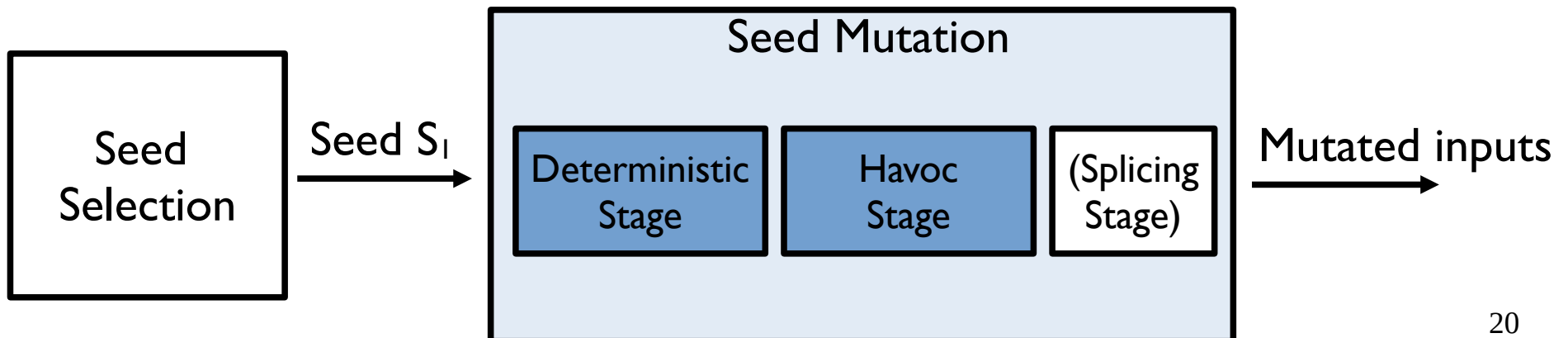
Seed Selection

- Add the interesting inputs into the queue.
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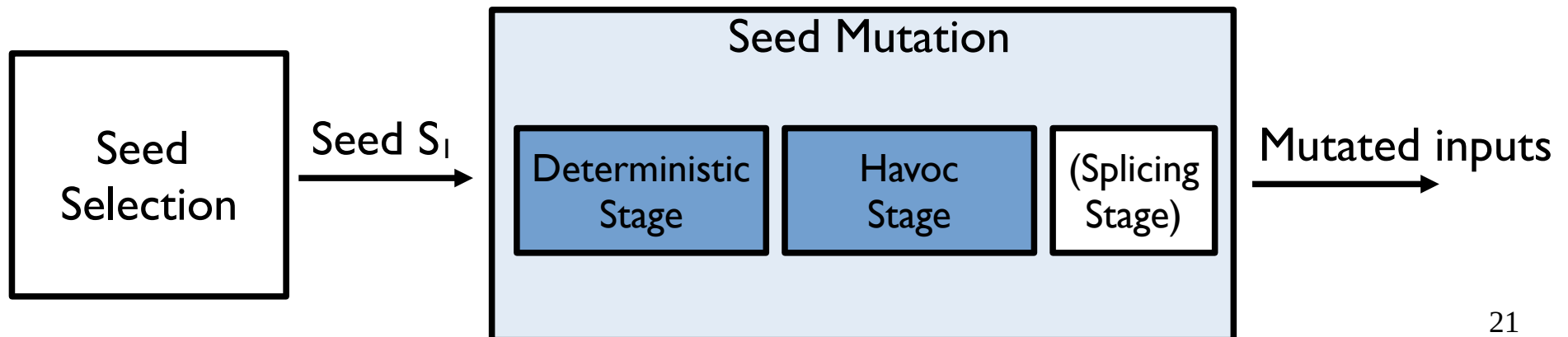
Seed Mutation

- **Deterministic** stage (**with no** randomness)
- **Havoc** stage (**with** randomness)



Seed Mutation

- **Deterministic** stage (**with no** randomness)
 - Generating mutants **by using a mutation operator**
(**Small change** to seed input)
- **Havoc** stage (**with** randomness)
 - Generating mutants **by using multiple operators**
(**Big change** to seed input)



Seed Mutation

- Deterministic stage
 - Apply each mutation operator to the input from the start to the end
 - Operators: `bitflip(1,2,4 bit)`, `byteflip(1,2,4 byte)`, arithmetic inc/dec, ...

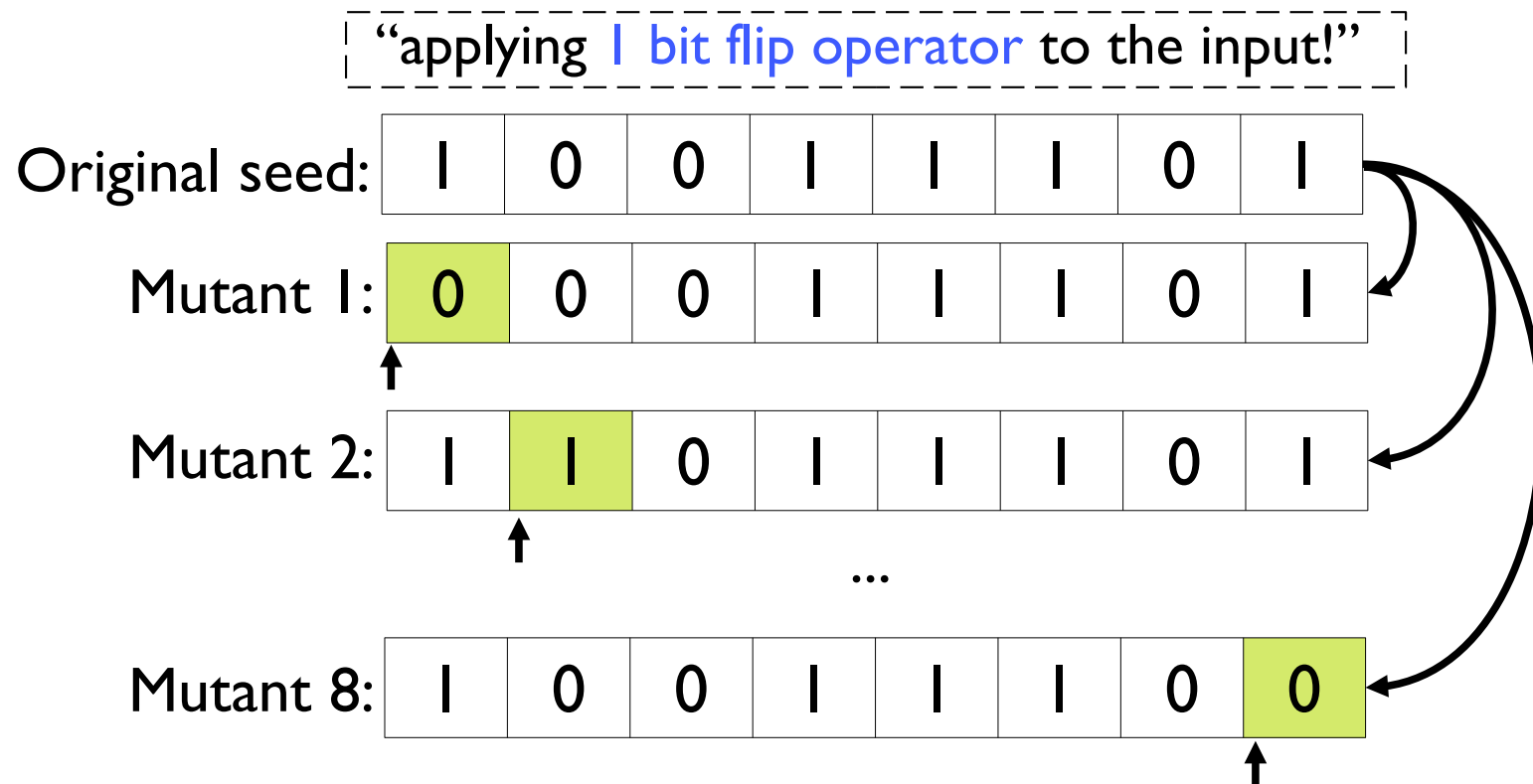
“applying 1 bit flip operator to the input!”

Original seed:

1	0	0	1	1	1	0	1
---	---	---	---	---	---	---	---

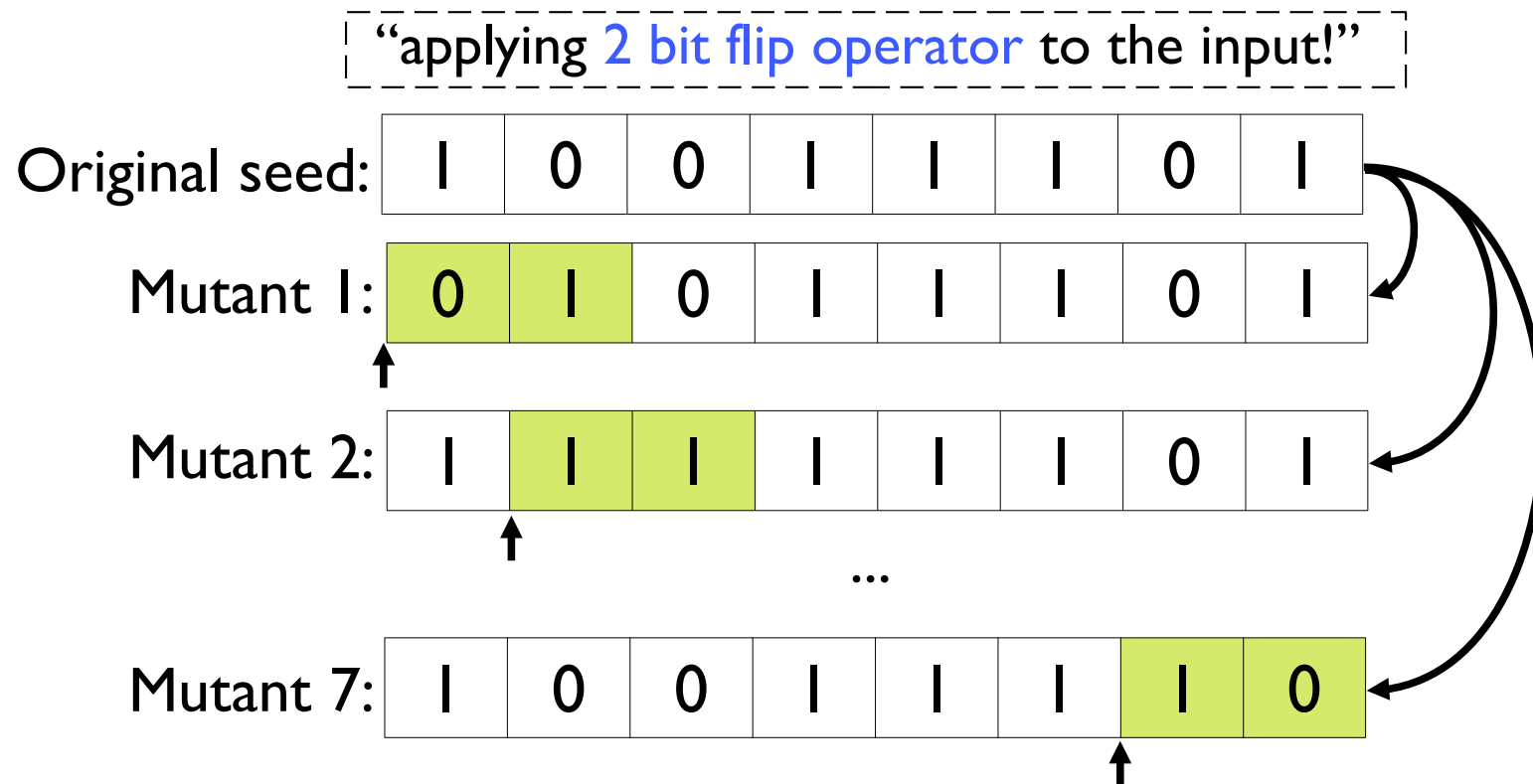
Seed Mutation

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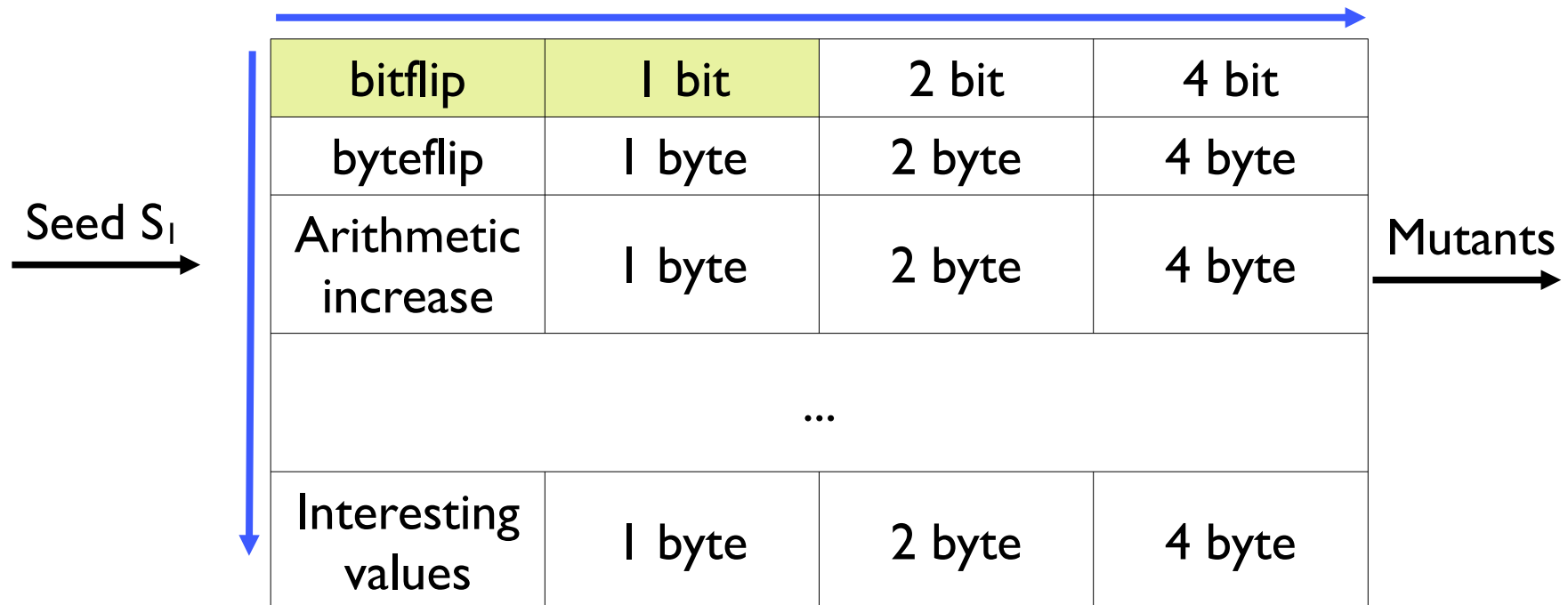
Seed Mutation

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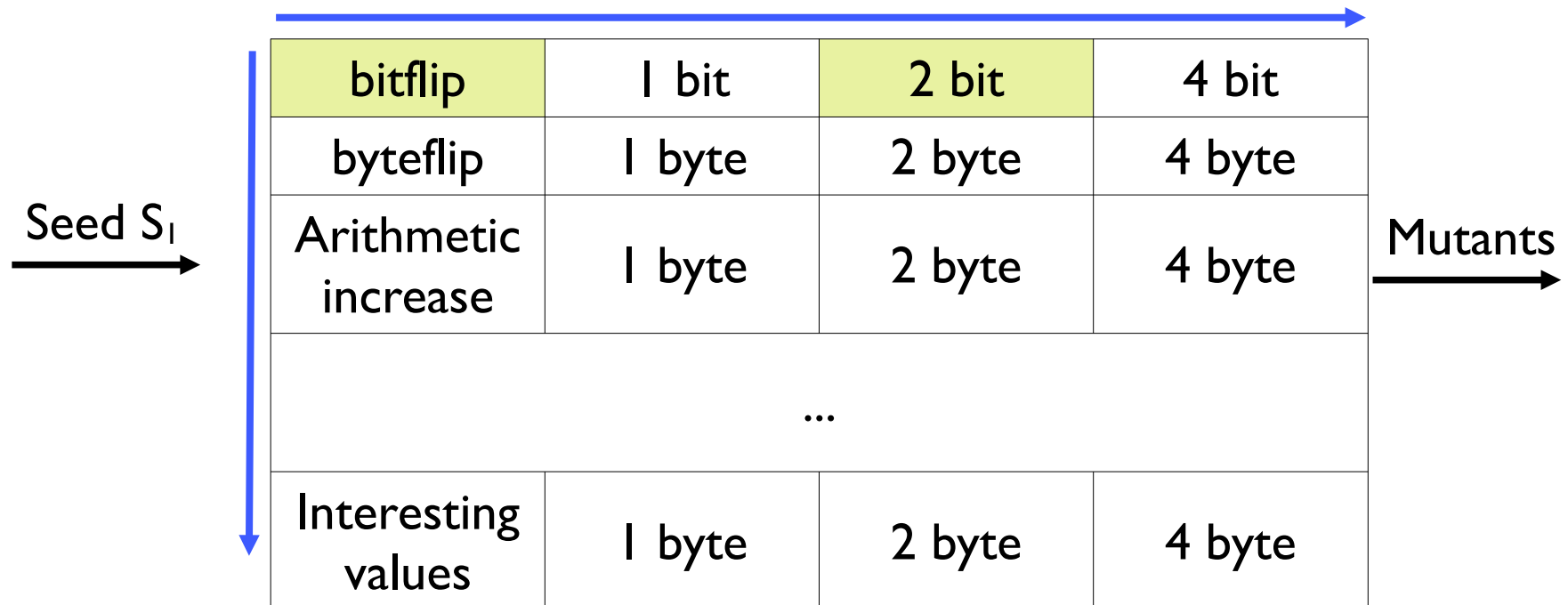
Seed Mutation

- Deterministic stage
 - Apply each mutation operator to the input from the start to the end
 - Perform all mutation operators in order



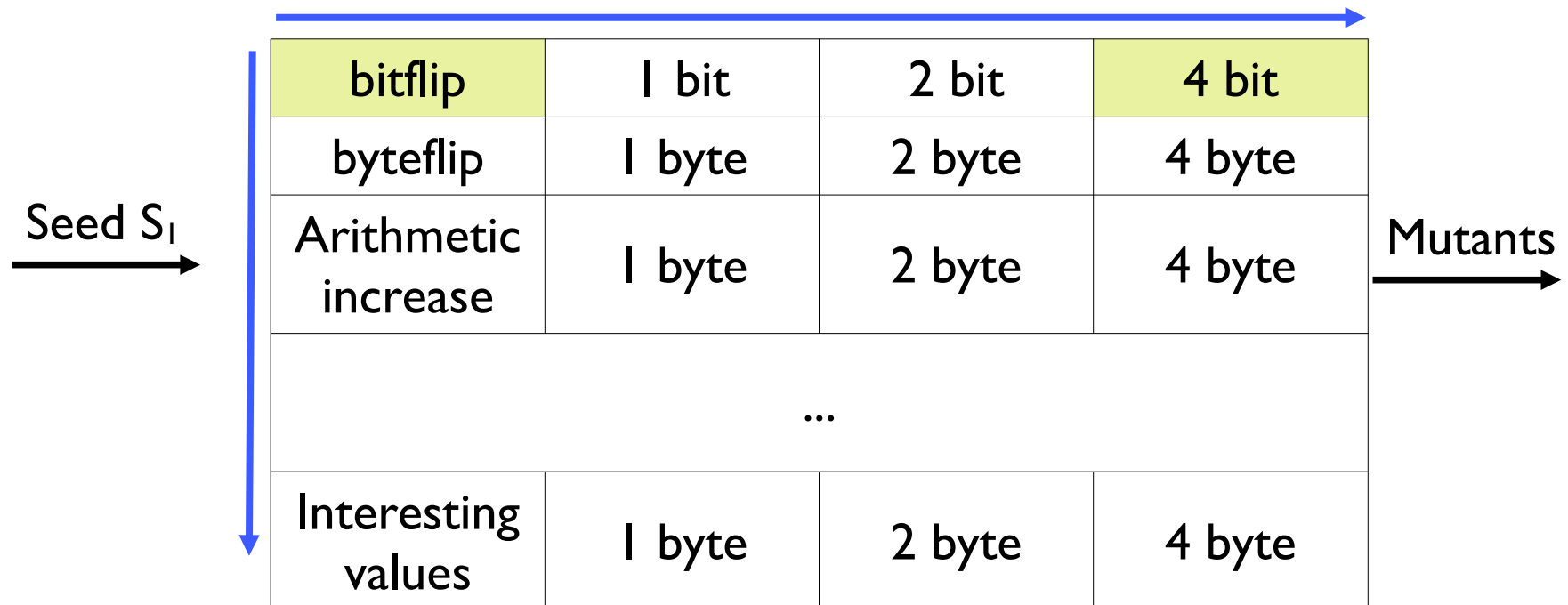
Seed Mutation

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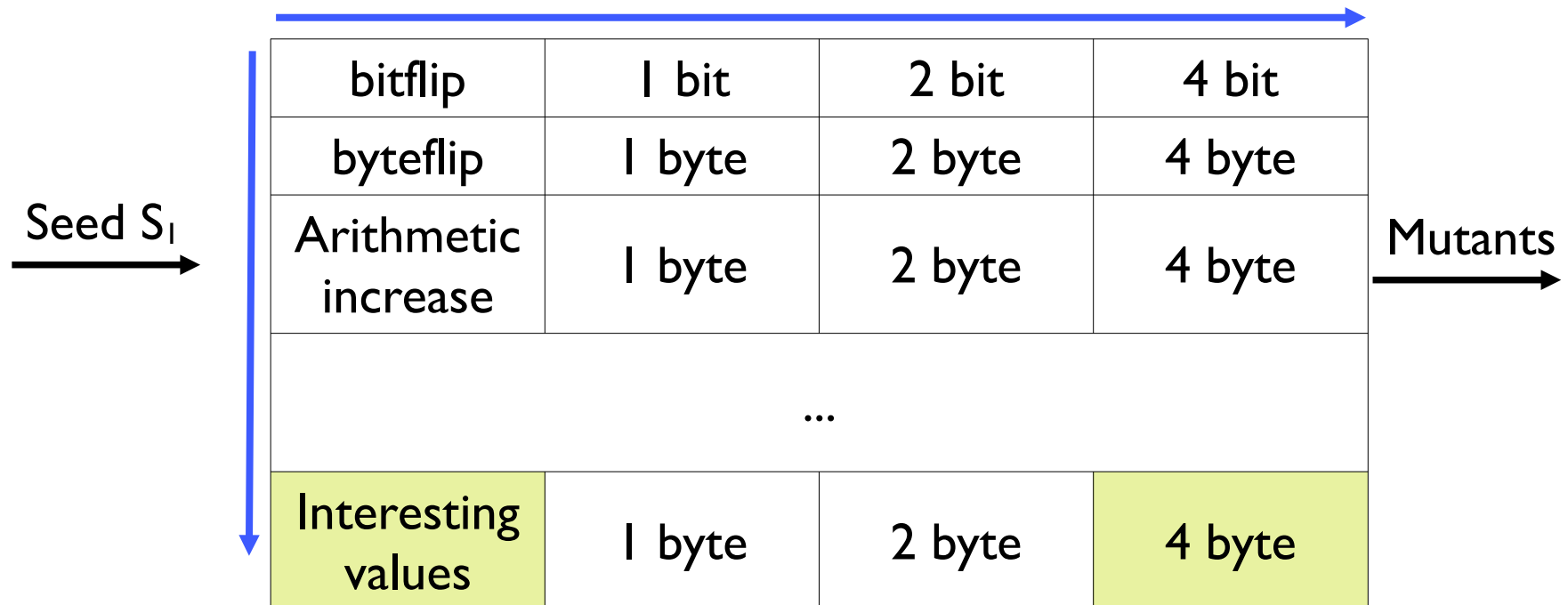
Seed Mutation

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 - Apply each mutation operator to the input from the start to the end
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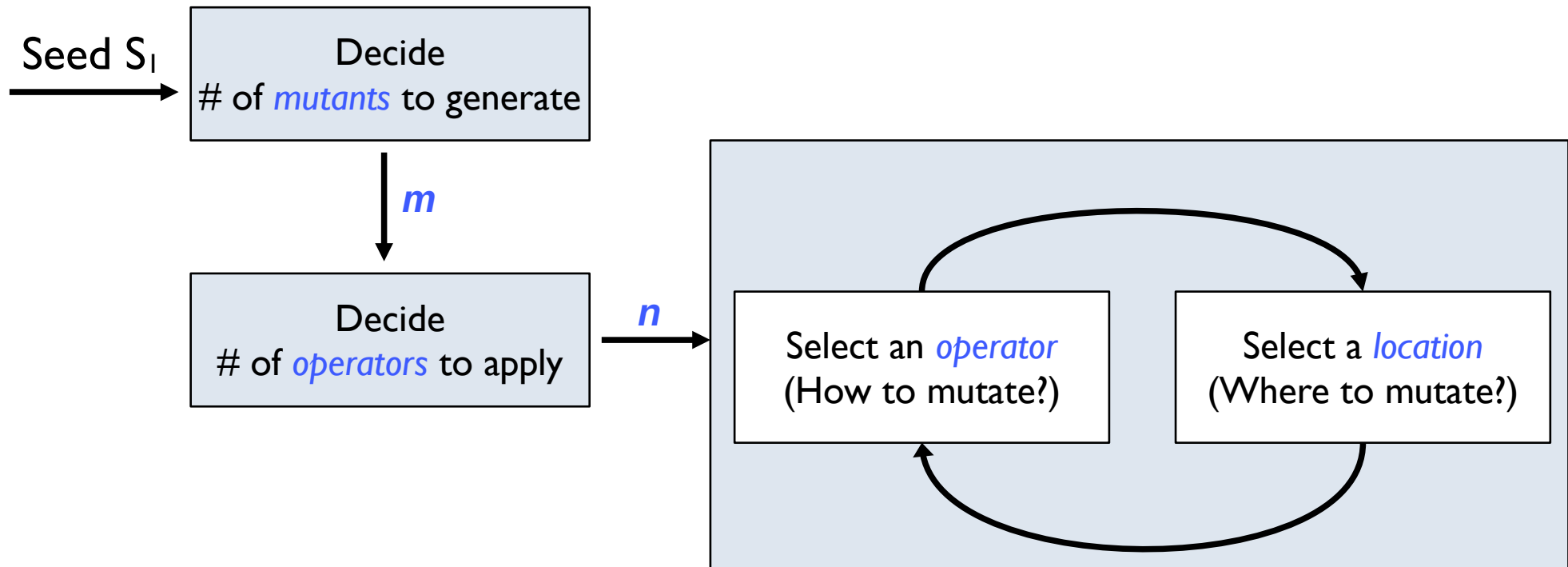
Seed Mutation

- Deterministic stage
 - Apply each mutation operator to the input from the start to the end
 - Perform all mutation operators in order



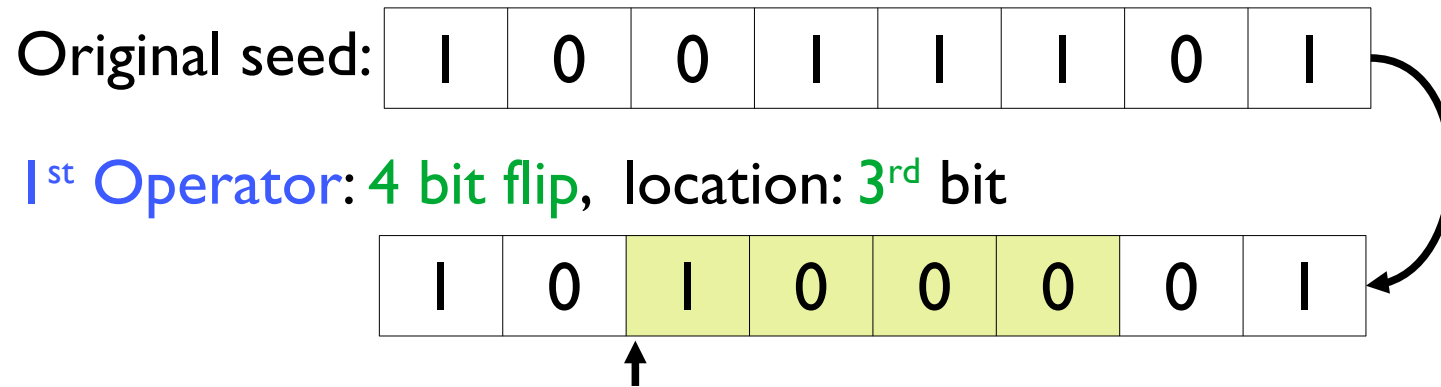
Seed Mutation

- Havoc stage
 - Apply n mutation operators to mutate the seed input.
 - Repeat the above to produce m mutants.



Seed Mutation

- Havoc stage
 - Apply **3** mutation operators to mutate the seed input.



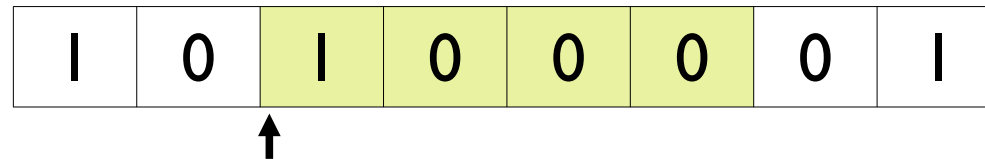
Seed Mutation

- Havoc stage
 - Apply **3** mutation operators to mutate the seed input.

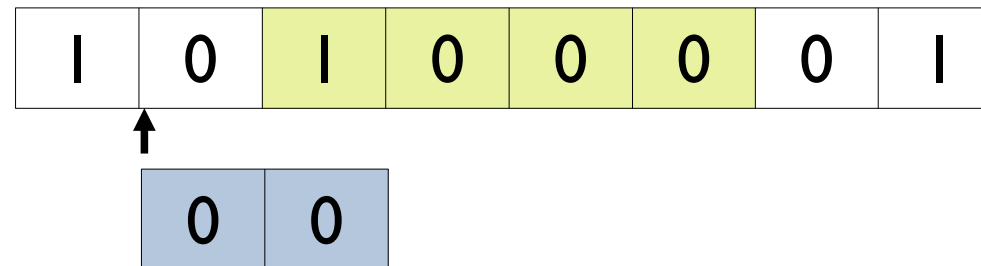
Original seed:

1	0	0	1	1	1	0	1
---	---	---	---	---	---	---	---

1st Operator: **4 bit flip**, location: **3rd** bit



2nd Operator: **Insert**, value: **00**, location: **2nd** bit



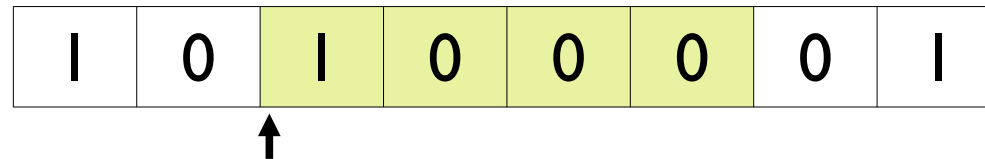
Seed Mutation

- Havoc stage
 - Apply **3** mutation operators to mutate the seed input.

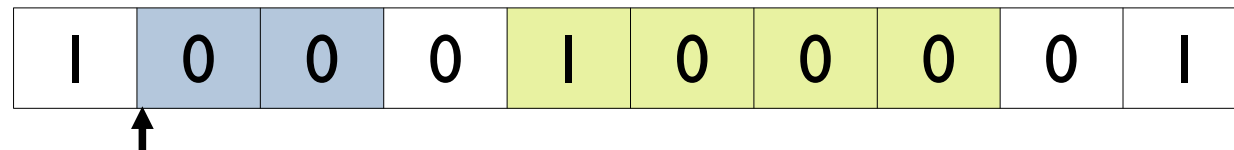
Original seed:

1	0	0	1	1	1	0	1
---	---	---	---	---	---	---	---

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Seed Mutation

- Havoc stage
 - Apply **3** mutation operators to mutate the seed input.

Original seed:

1	0	0	1	1	1	0	1
---	---	---	---	---	---	---	---

1st Operator: **4 bit flip**, location: **3rd** bit

1	0	1	0	0	0	0	1
---	---	---	---	---	---	---	---

↑

2nd Operator: **Insert**, value: **00**, location: **2nd** bit

1	0	0	0	1	0	0	0	0	1
---	---	---	---	---	---	---	---	---	---

↑

3rd Operator: **delete**, size: **1 bit**, location: **7th** bit

1	0	0	0	1	0	0	0	0	1
---	---	---	---	---	---	---	---	---	---

↑

Seed Mutation

- Havoc stage
 - Apply **3** mutation operators to mutate the seed input.

Original seed:

1	0	0	1	1	1	0	1
---	---	---	---	---	---	---	---

1st Operator: **4 bit flip**, location: **3rd** bit

1	0	1	0	0	0	0	1
---	---	---	---	---	---	---	---

↑

2nd Operator: **Insert**, value: **00**, location: **2nd** bit

1	0	0	0	1	0	0	0	0	1
---	---	---	---	---	---	---	---	---	---

↑

3rd Operator: **delete**, size: **1 bit**, location: **7rd** bit

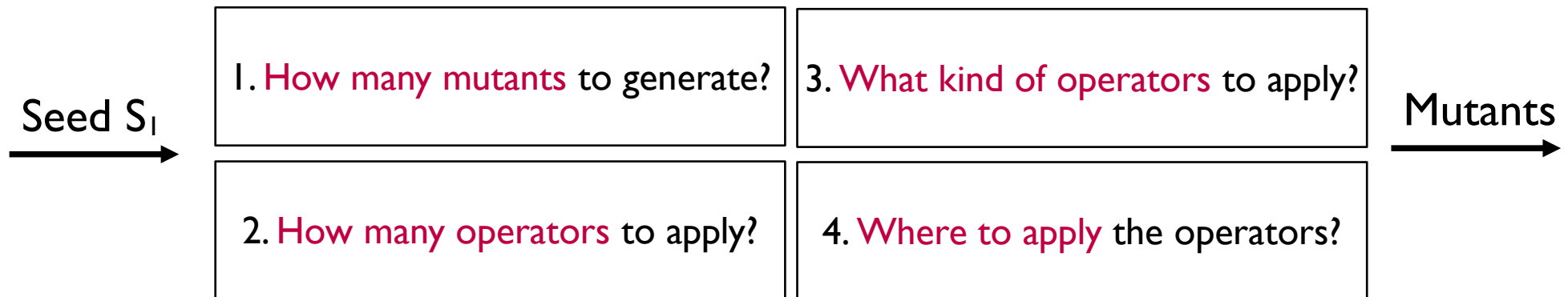
Mutant 1:

1	0	0	0	1	0	0	0	1
---	---	---	---	---	---	---	---	---

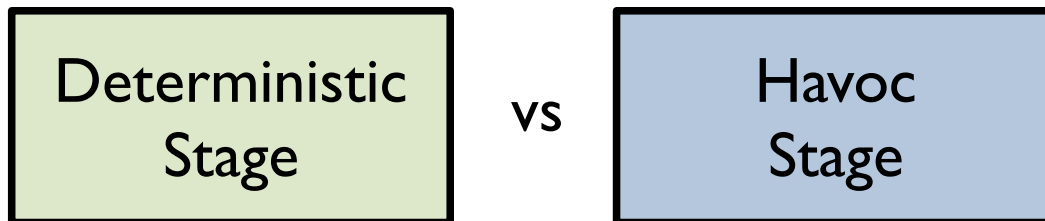
↑

Seed Mutation

- Havoc stage
 - **Totally rely on randomness** to generate mutants.

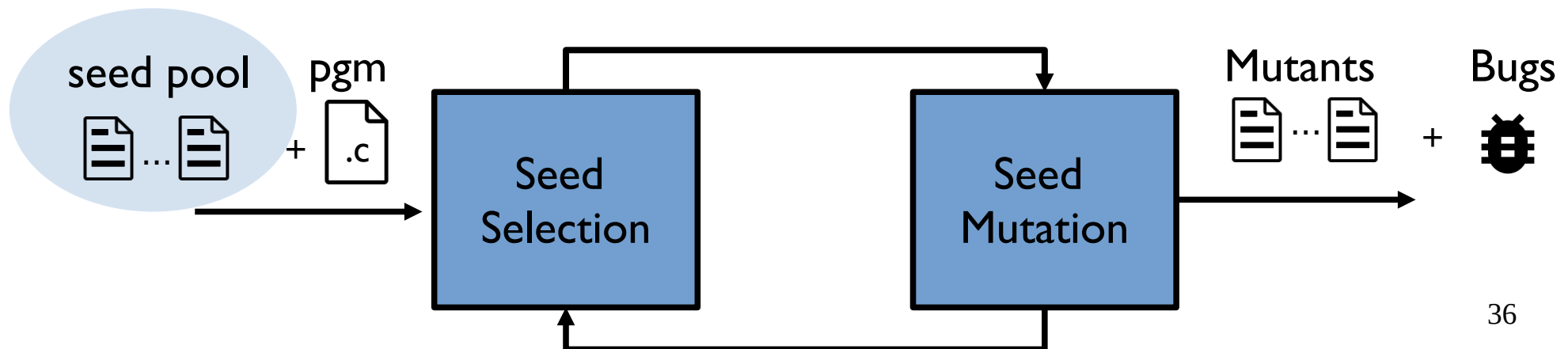


- Which one is more effective at finding bugs?



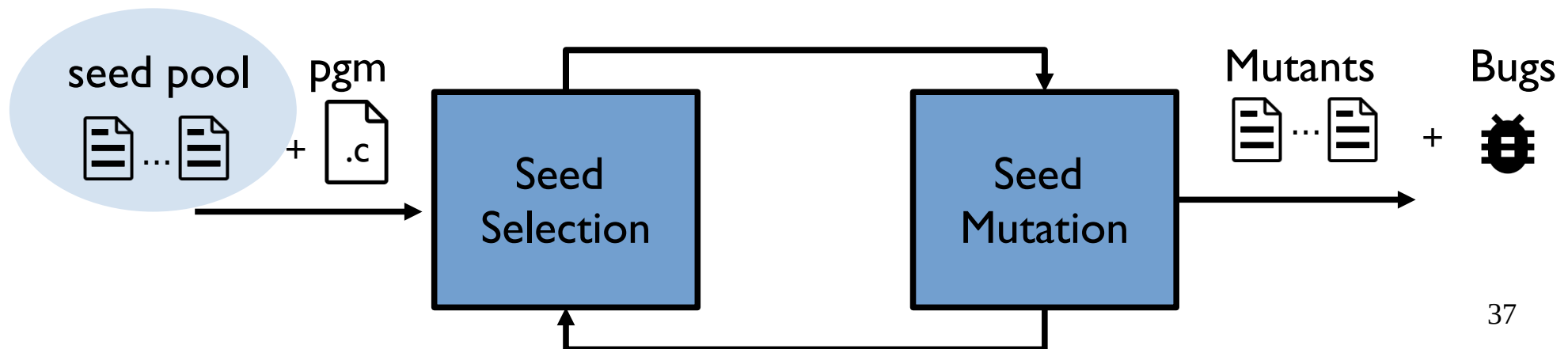
Topics in Mutation-based Fuzzing

- Seed Pool Generation
 - **How to effectively build** an initial seed pool?



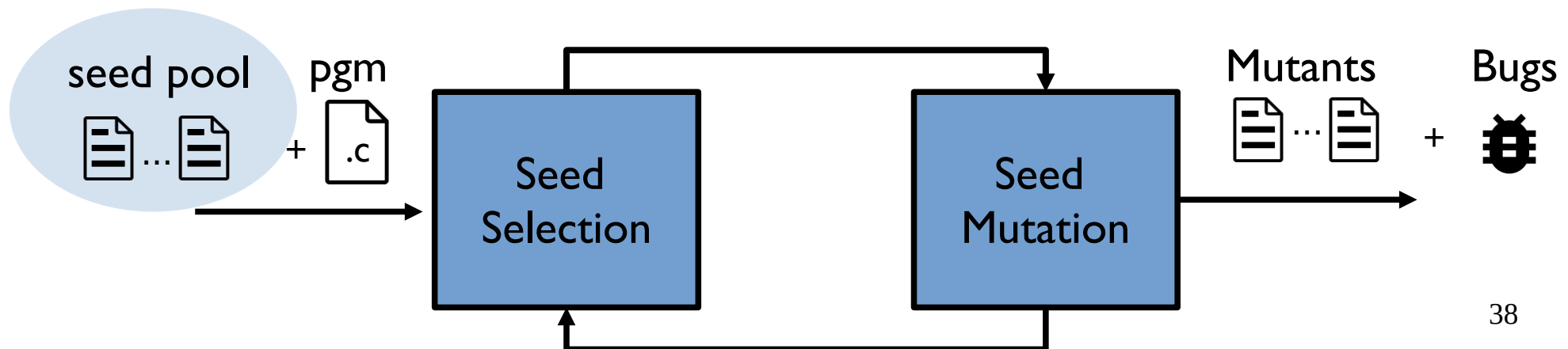
Topics in Mutation-based Fuzzing

- Seed Pool Generation
 - How to effectively build an initial seed pool?
- Seed Selection
 - Which seed from the pool to select first?



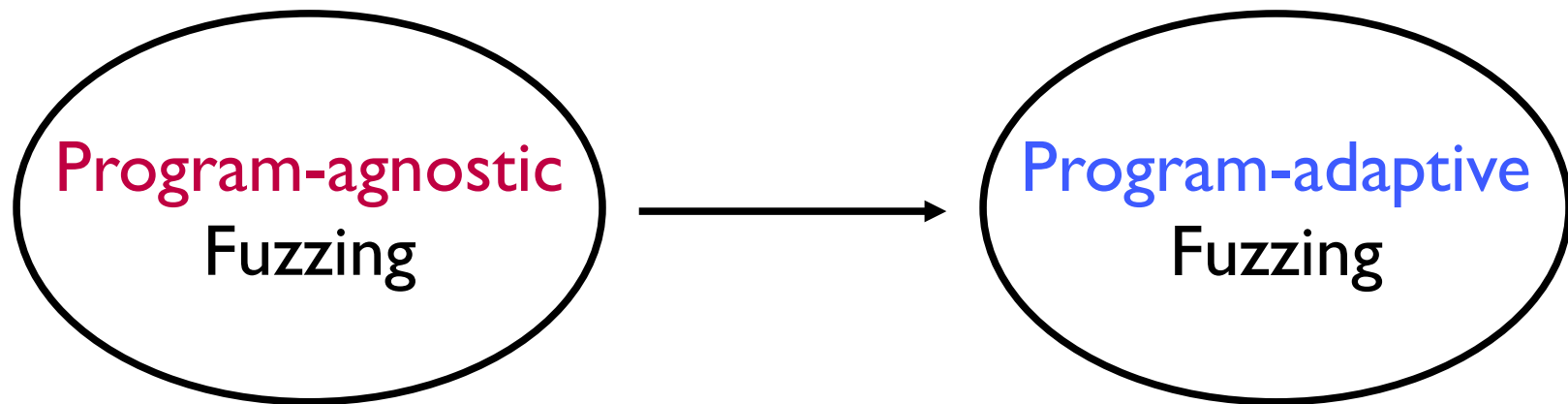
Topics in Mutation-based Fuzzing

- Seed Pool Generation
 - How to effectively build an initial seed pool?
- Seed Selection
 - Which seed from the pool to select first?
- Seed Mutation
 - How to mutate the selected seed?



Recent Trend

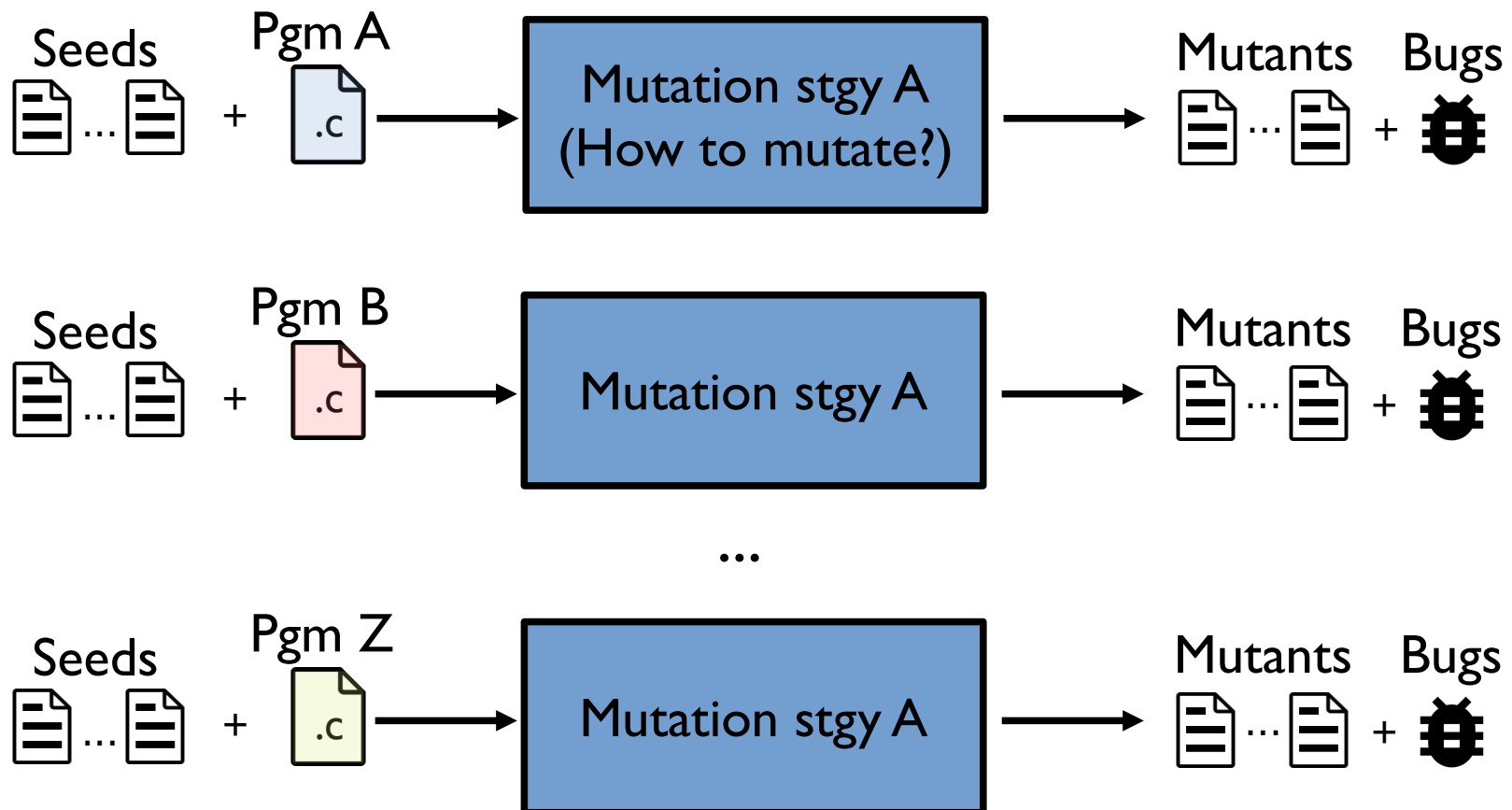
- Recent trend in mutation-based fuzzing



- Program-agnostic fuzzing
 - Using a fixed seed mutation (or selection) strategy regardless of the target program. (e.g., AFL)
- Program-adaptive fuzzing
 - Using an adaptive seed mutation (or seed selection) strategy depending on the target program. (e.g., MOPT)

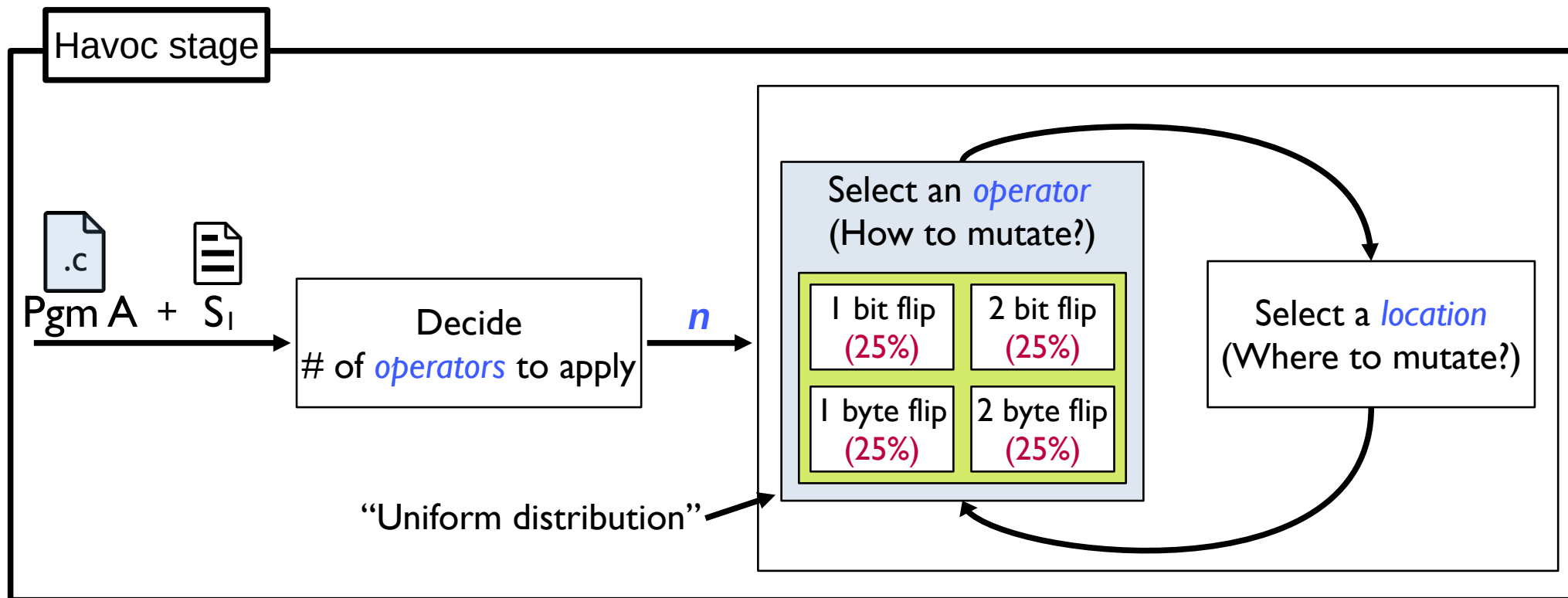
Recent Trend

- **Program-agnostic** seed mutation strategy
 - **Unchanging** the seed mutation strategy.



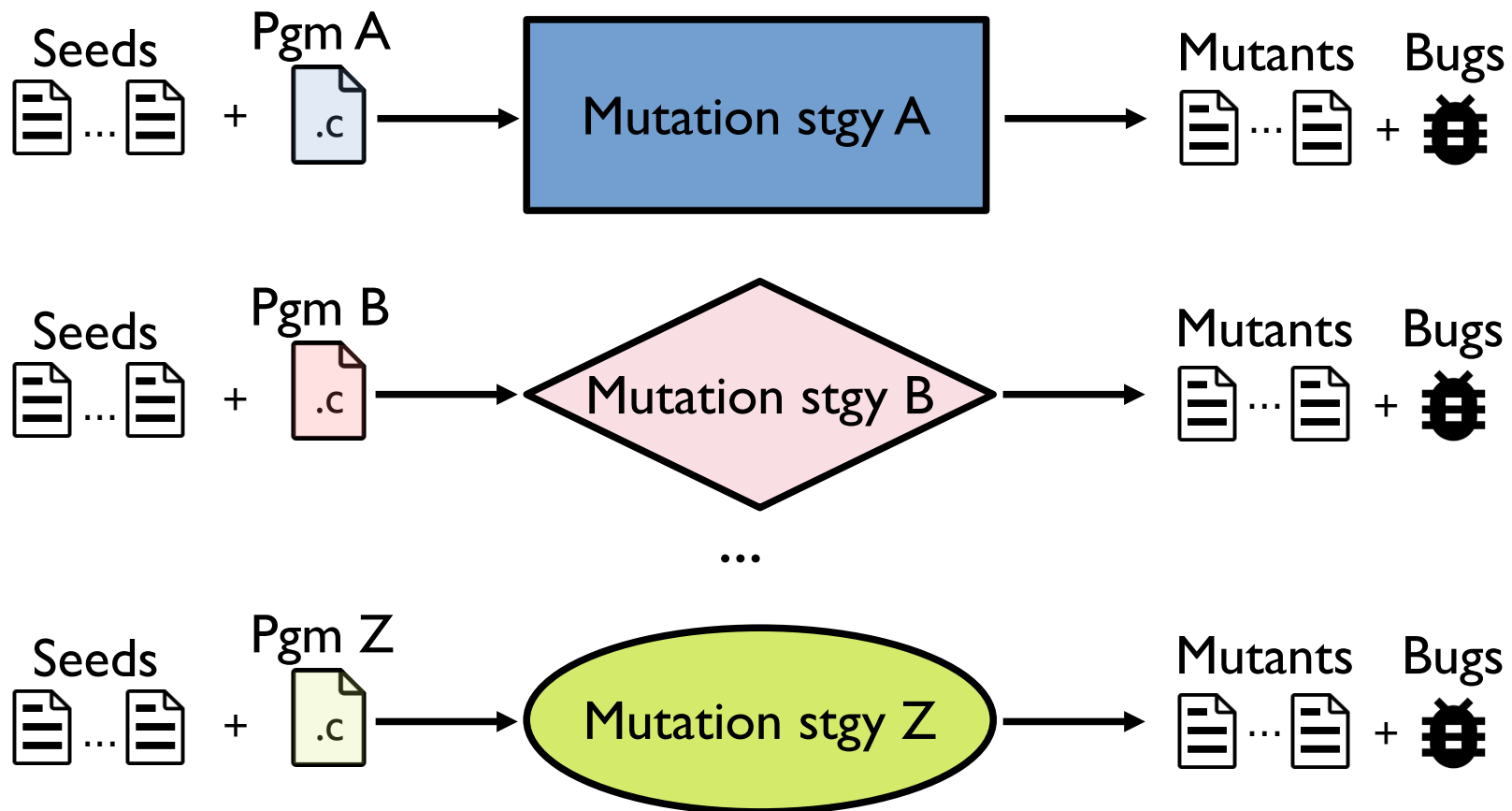
Program-Agnostic Mutation Strategy

- Seed mutation strategy (in AFL)
 - Randomly select the predefined mutation operators.



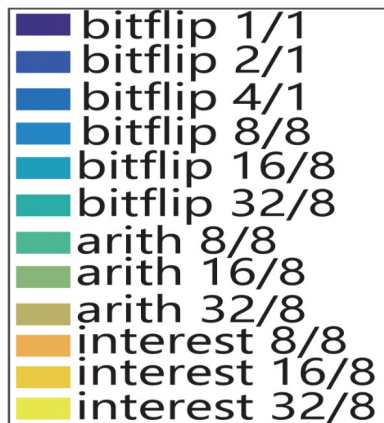
Recent Trend

- **Program-adaptive** seed mutation strategy
 - Changing the seed mutation strategy depending on the program.

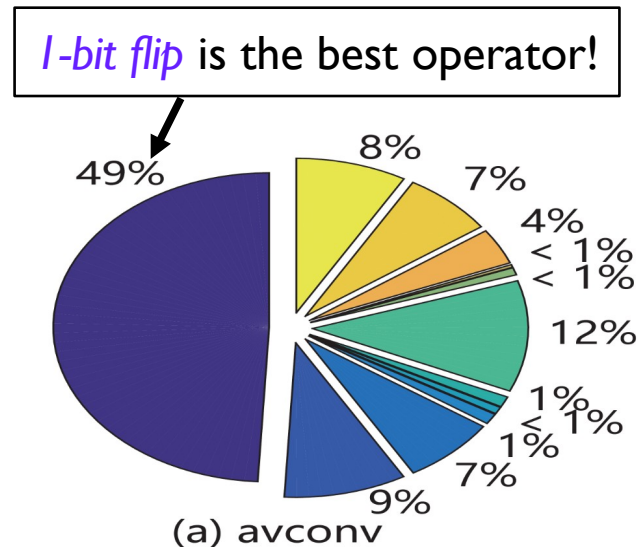


MOPT

- Motivation (Key observation)
 - Effective mutation operators are different from the target program.¹



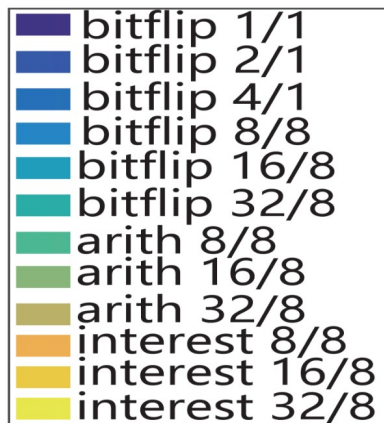
Mutation operator types.



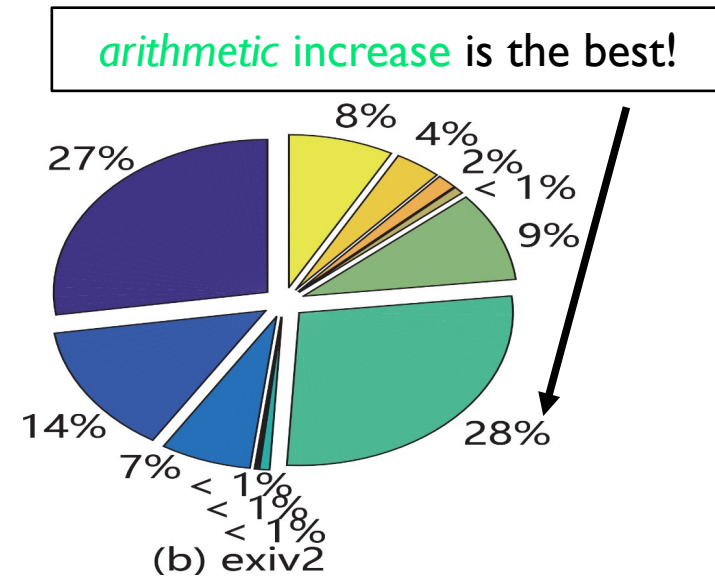
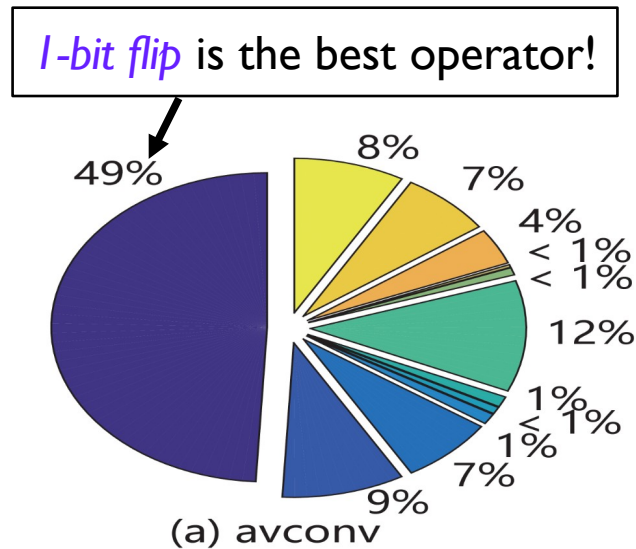
Percentages of *interesting* mutants generated by each mutation operator

MOPT

- Motivation (Key observation)
 - Effective mutation operators are different from the target program.¹



Mutation operator types.

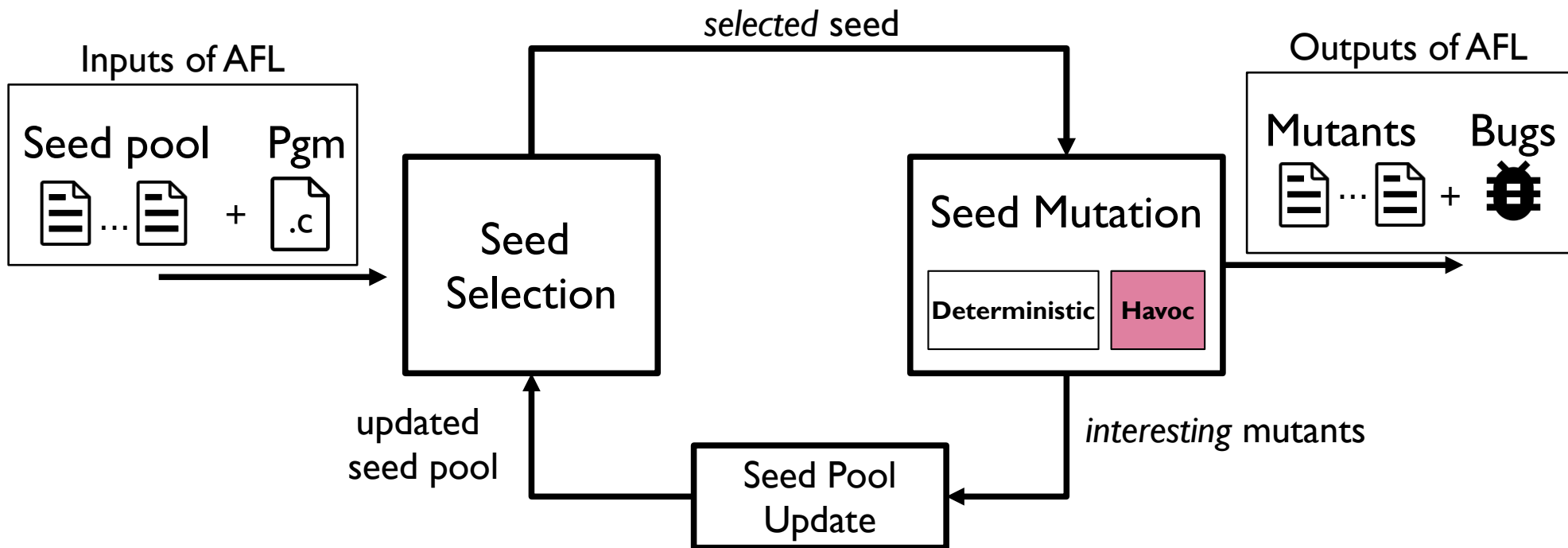


Percentages of *interesting* mutants generated by each mutation operator

MOPT

- Goal

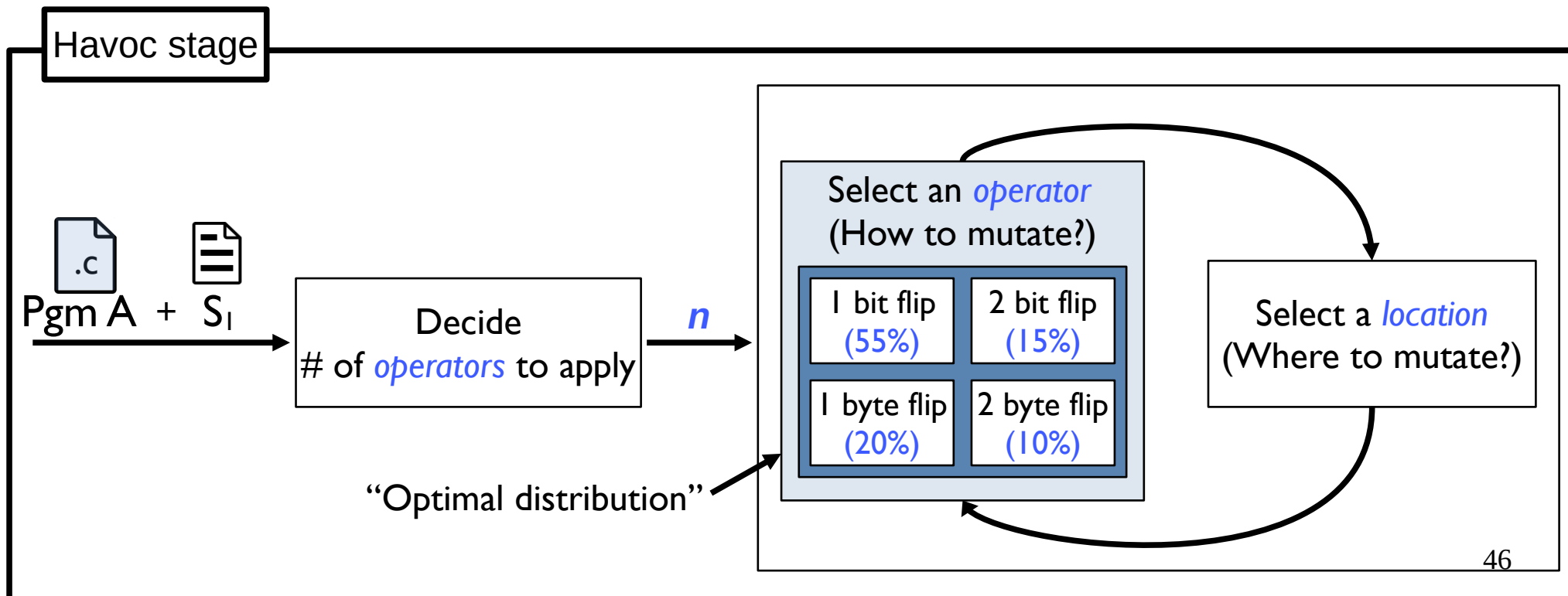
- Finding an optimal mutation strategy in havoc stage for the program.



MOPT

- Goal

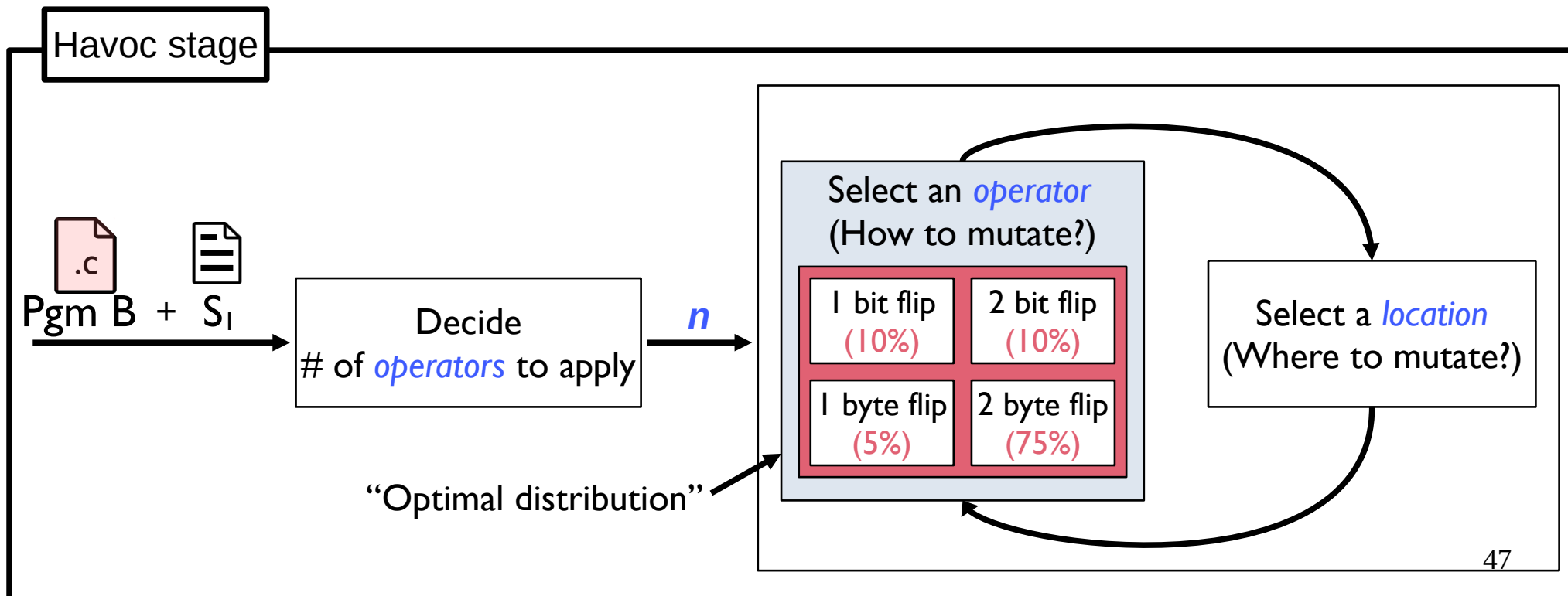
- Finding an optimal mutation strategy in havoc stage for the program.
= Finding an optimal probabilistic distribution of selecting operators.



MOPT

- Goal

- Finding an optimal mutation strategy in havoc stage for the program.
= Finding an optimal probabilistic distribution of selecting operators.



MOPT

- Key Idea: online learning algorithm
 - Present a customized particle swarm optimization algorithm.

MOPT

- Key Idea: online learning algorithm
 - Present a customized particle swarm optimization algorithm.
 - A particle = A probability of selecting a single operator.
 - A swarm = A probabilistic distribution of selecting operators.

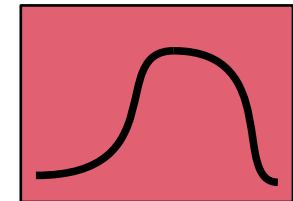
“Particle”

1 bit flip (24%)

“Swarm”

1 bit flip (24%)	2 bit flip (26%)
1 byte flip (22%)	2 byte flip (28%)

=

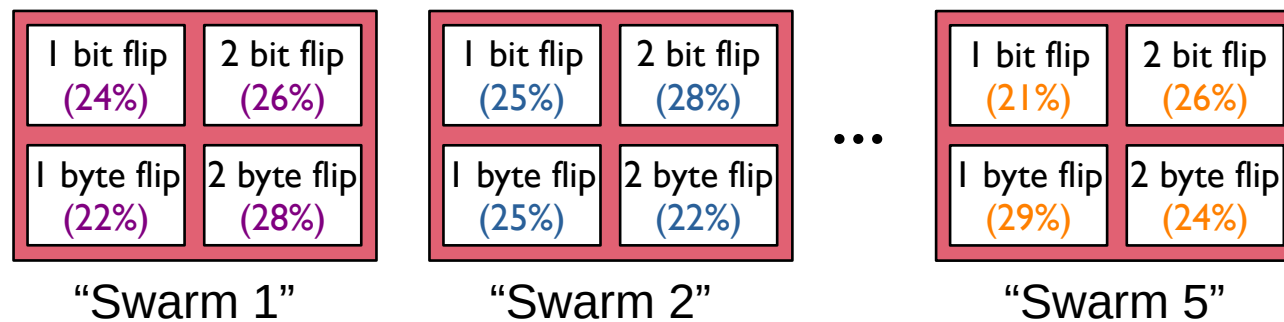


MOPT

- Key Idea

(1) Initialize k swarms. ($k=5$)

- An initial swarm = a random distribution of mutation operators



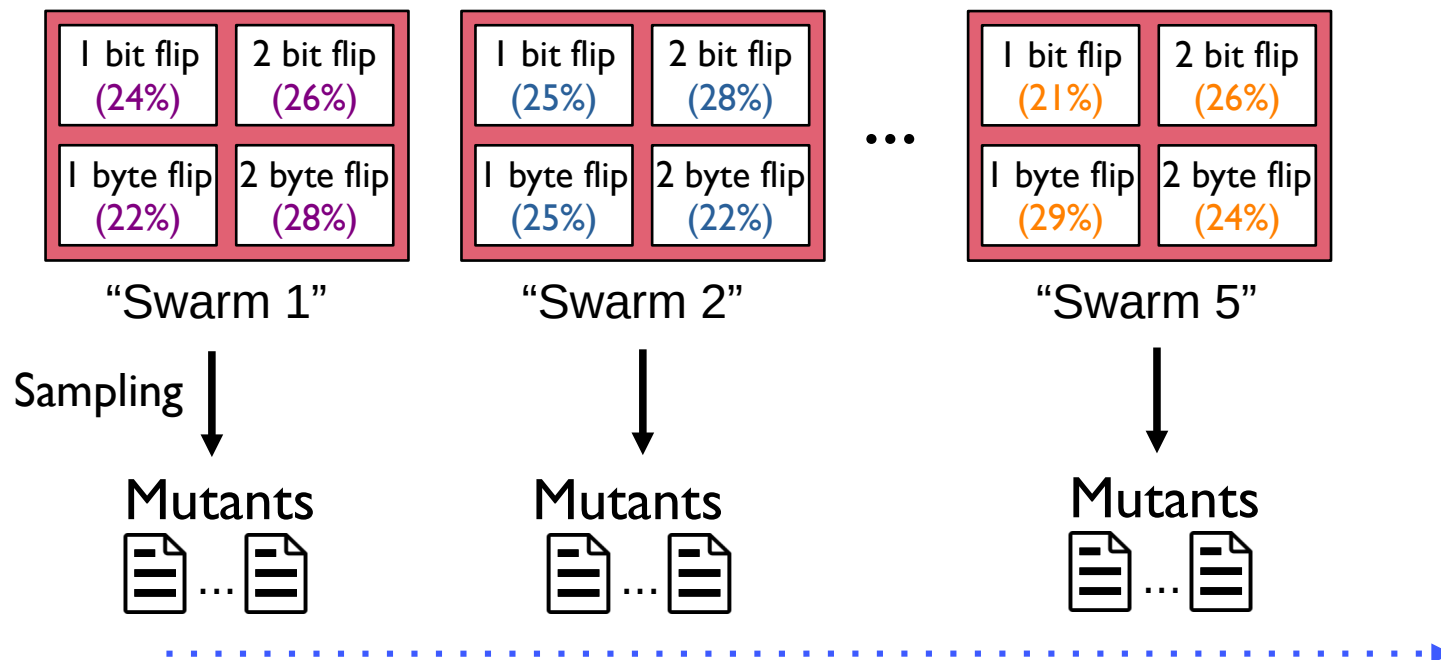
MOPT

- Key Idea

(1) Initialize k swarms. ($k=5$)

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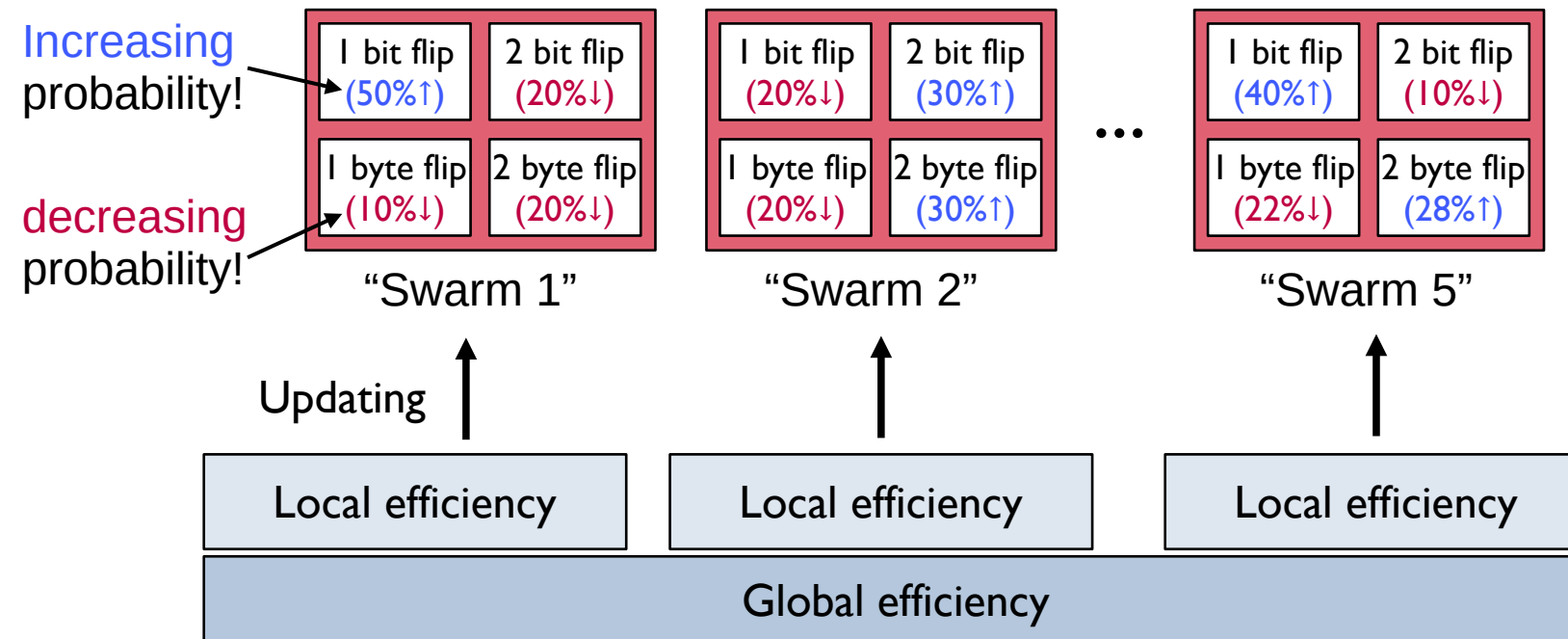
(2) Generate mutants by sampling from k swarms, respectively.



MOPT

• Key Idea

- (3). Update k swarms based on local and global efficiency.
- Intuition for the efficiency
 - How effective is each particle in generating interesting mutants?



MOPT

- Key Idea

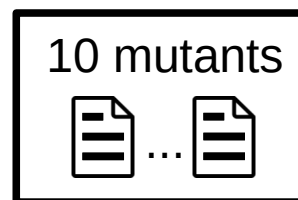
- (2). Updating k swarms based on local and global efficiency.

- Local efficiency of a particle in a swarm

(ex) Local efficiency of the particle ($op_{1\text{bit-flip}}$) for swarm 1

1 bit flip (24%)	2 bit flip (26%)
1 byte flip (22%)	2 byte flip (28%)

“Swarm 1”



3 *Interesting* mutants



MOPT

- Key Idea

- (2). Updating k swarms based on local and global efficiency.

- Local efficiency of a particle in a swarm

(ex) Local efficiency of the particle ($op_{1\text{bit-flip}}$) for swarm 1

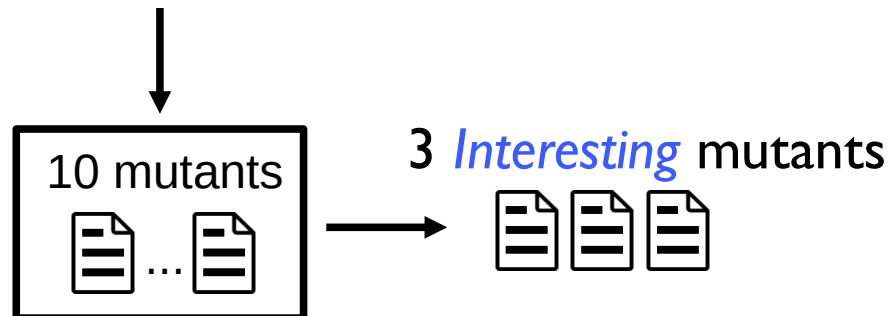
1 bit flip (24%)	2 bit flip (26%)
1 byte flip (22%)	2 byte flip (28%)

“Swarm 1”

1. # of total sampled times for $op_{1\text{bit-flip}}$: 5

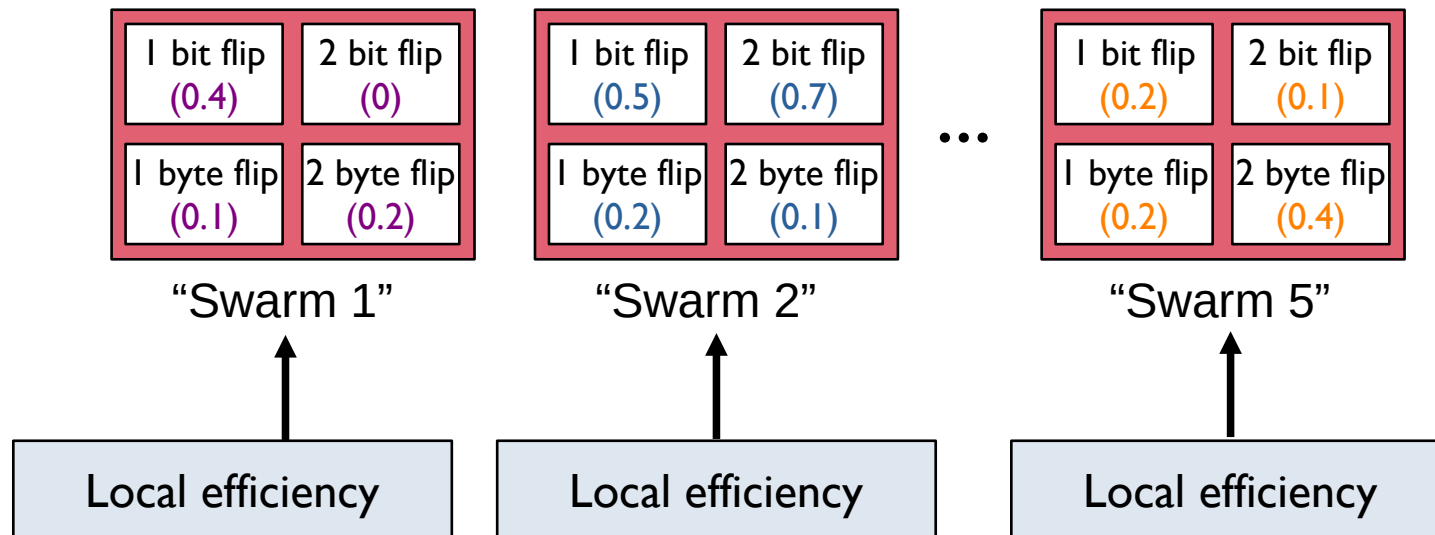
2. # of interesting mutants contributed by $op_{1\text{bit-flip}}$: 2

3. Local efficiency of $op_{1\text{bit-flip}}$: 0.4 (2/5)



MOPT

- Key Idea
 - (2). Updating k swarms based on local and global efficiency.
 - Local efficiency of a particle in a swarm
- (ex) Local efficiency of the particle ($op_{1\text{bit-flip}}$) for swarm 1



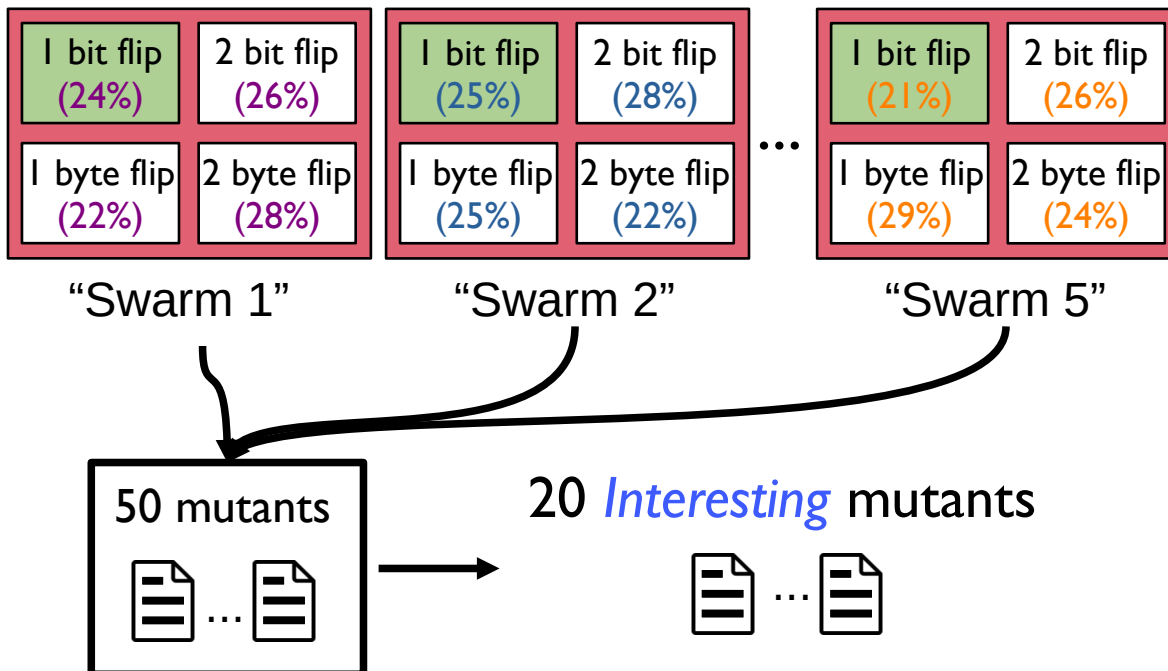
MOPT

- Key Idea

- (2). Updating k swarms based on local and global efficiency.

- Global efficiency of a particle in 5 swarms

(ex) Global efficiency of the particle ($op_{1\text{ bit-flip}}$) for 5 swarms



MOPT

- Key Idea

- (2). Updating k swarms based on local and global efficiency.

- Global efficiency of a particle in 5 swarms

(ex) Global efficiency of the particle ($op_{1\text{bit-flip}}$) for 5 swarms

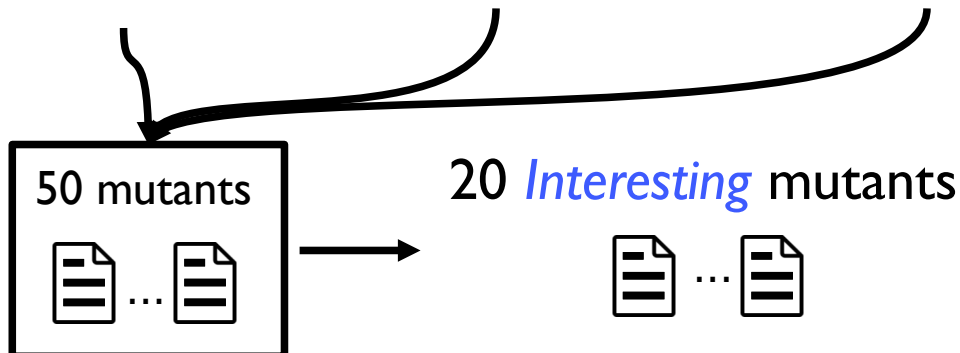
1 bit flip (24%)	2 bit flip (26%)	1 bit flip (25%)	2 bit flip (28%)	...	1 bit flip (21%)	2 bit flip (26%)
1 byte flip (22%)	2 byte flip (28%)	1 byte flip (25%)	2 byte flip (22%)		1 byte flip (29%)	2 byte flip (24%)

“Swarm 1”

“Swarm 2”

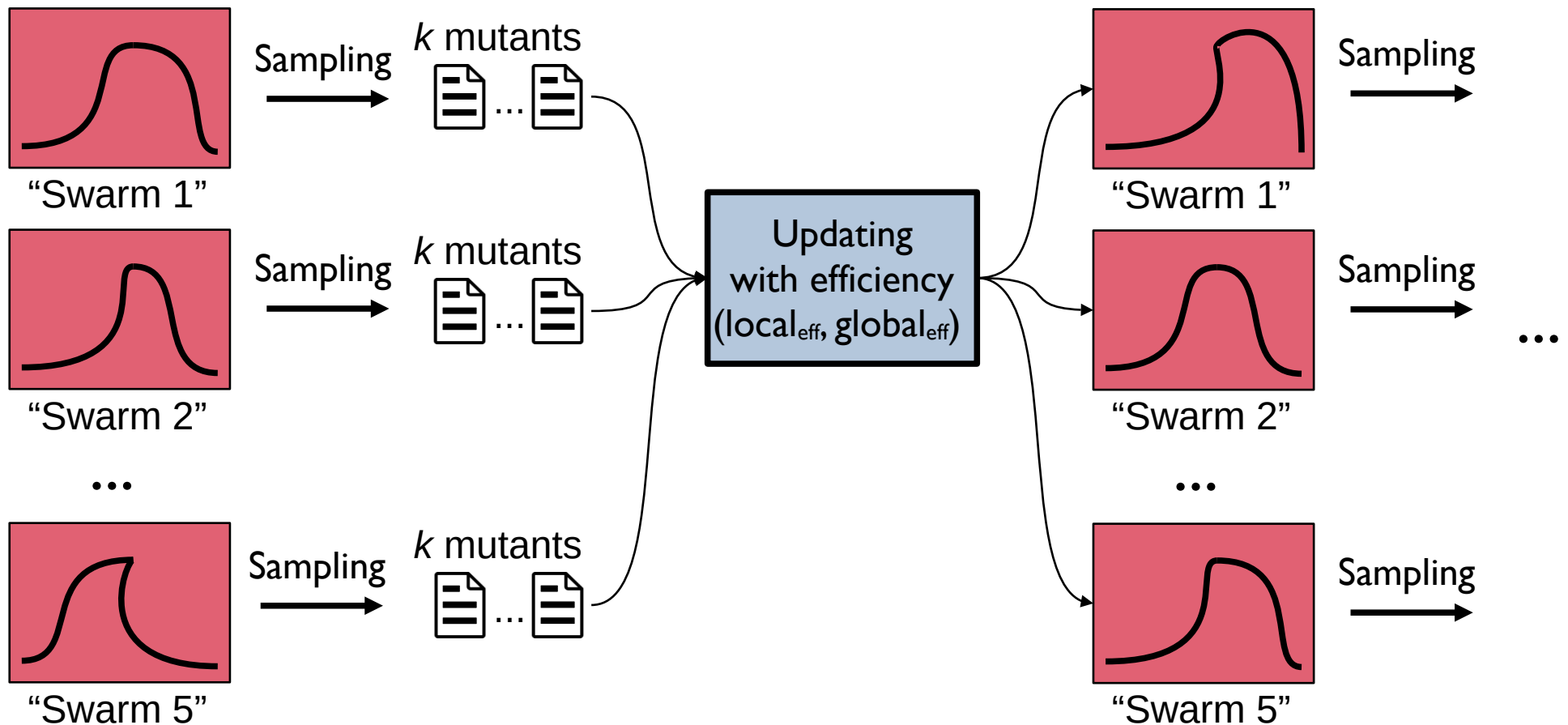
“Swarm 5”

1. # of total times for $op_{1\text{bit-flip}}$: 20
2. # of interesting mutants by $op_{1\text{bit-flip}}$: 10
3. Global efficiency of $op_{1\text{bit-flip}}$: 0.5 (10/20)



MOPT

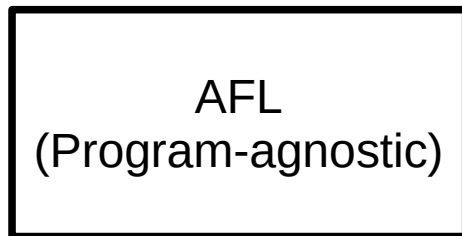
- Repeat the sampling and updating process.



MOPT

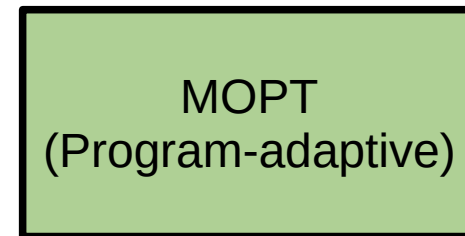
- Effectiveness

- Total benchmarks : 13 open-source linux programs.
- Testing budget: 10 days.



610 unique crashes
47,618 unique paths

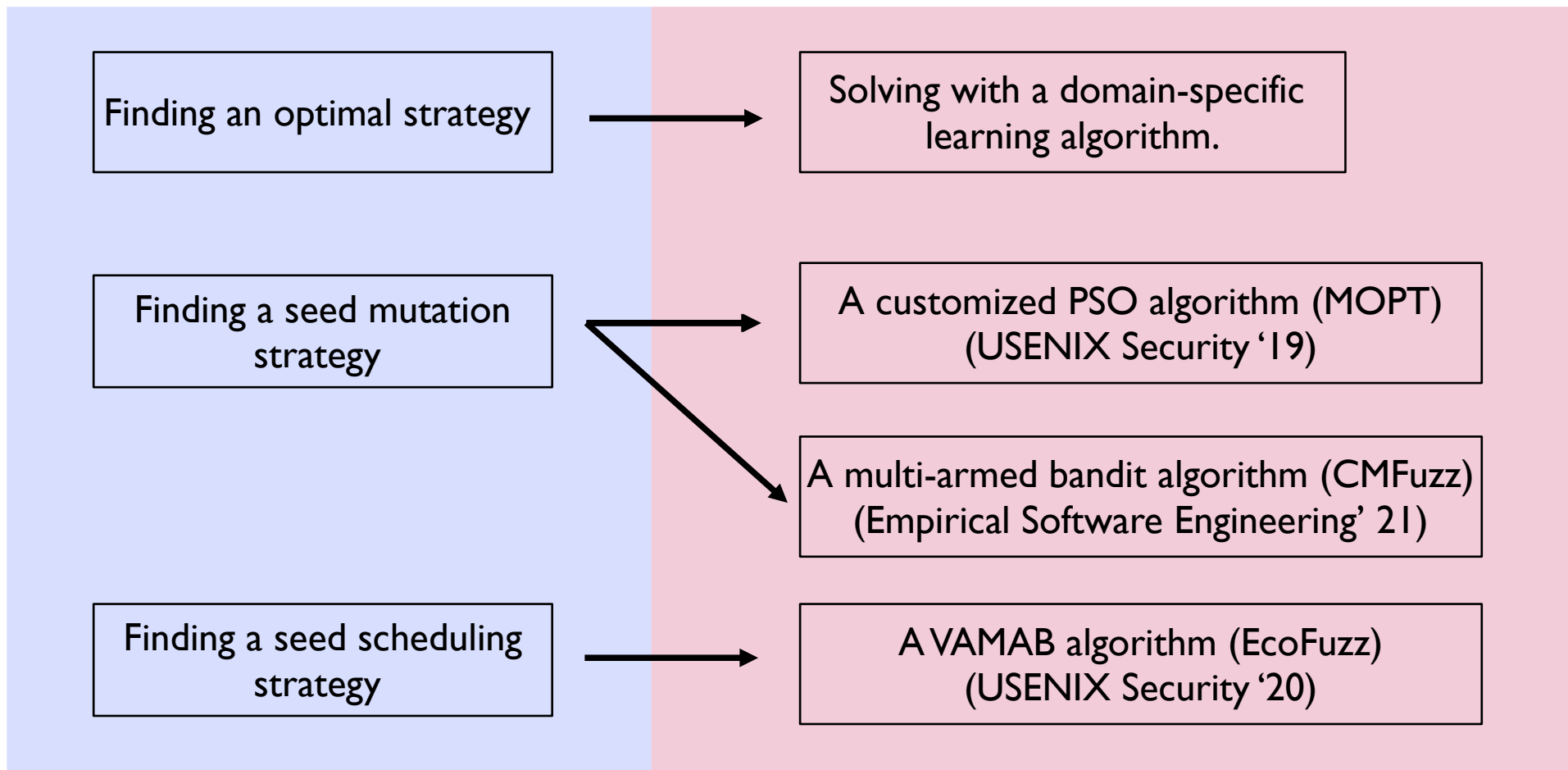
<



2,944 unique crashes (x4.8)
104,133 unique paths (x2.2)

Recent Mutation-based Fuzzing

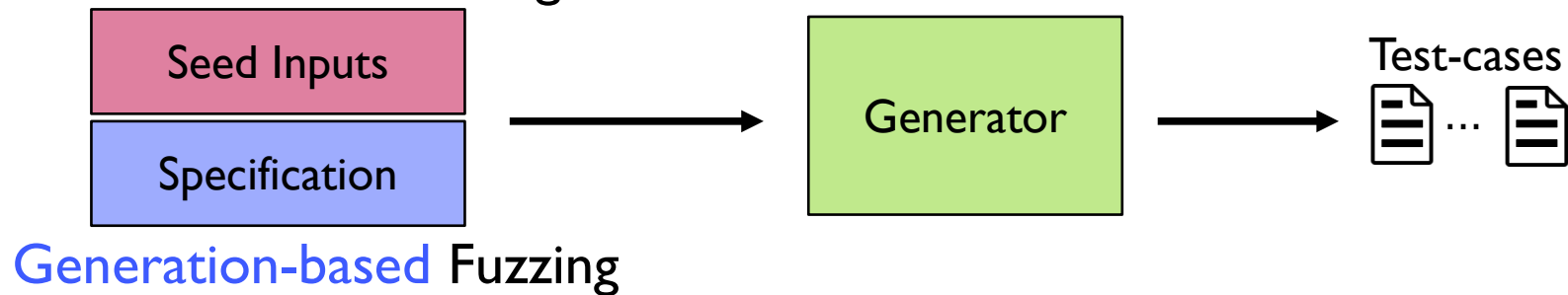
- Mutation-based Fuzzing + Learning



Generation-based Fuzzing

- Generation-based fuzzing vs Mutation-based fuzzing

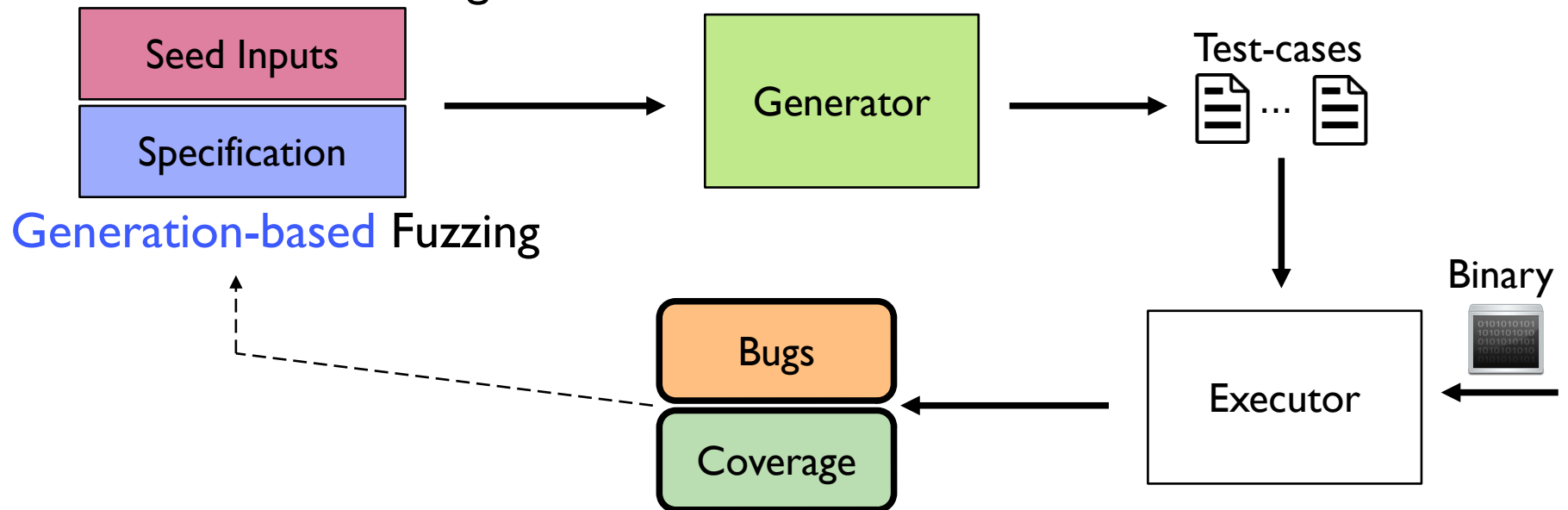
Mutation-based Fuzzing



Generation-based Fuzzing

- Generation-based fuzzing vs Mutation-based fuzzing

Mutation-based Fuzzing

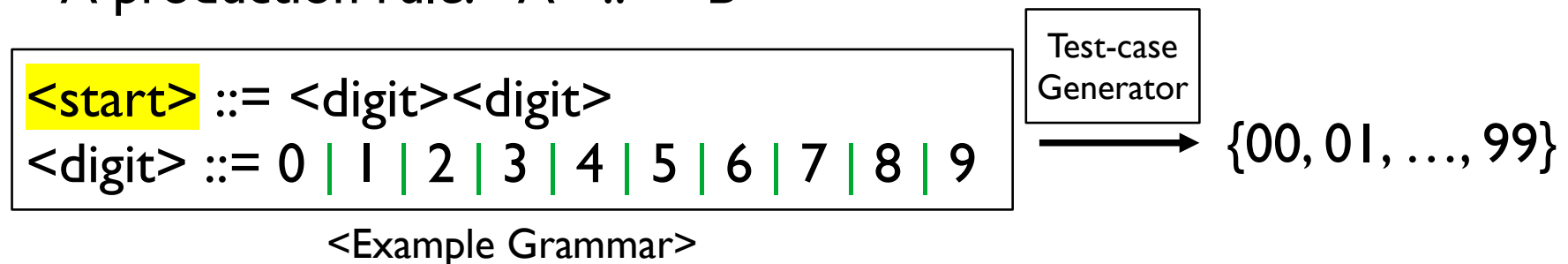


Generation-based Fuzzing

- Generate test-cases by using a specification.¹
 - ex) a specification: **grammars** of valid inputs to a target program.
- A Grammar
 - Express **the syntactical structure** of an input. (e.g., javascript)
 - Consists of **a start symbol** and a set of **production rules**.
 - A start symbol: $\langle \text{start} \rangle$
 - A production rule: $\langle A \rangle ::= \langle B \rangle$

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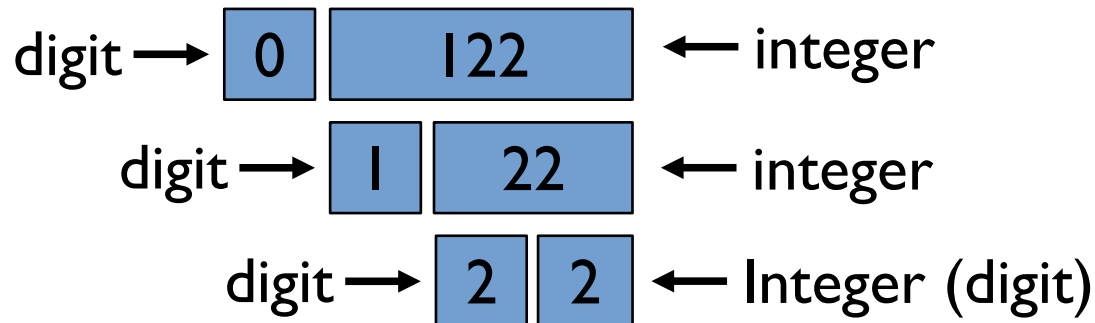
Grammars

- Grammars can be recursive.¹

```
<start>    ::= <integer>
<integer>  ::= <digit> | <digit><integer>
<digit>    ::= 0 | 1 | 2 | 3 | 4 | 5 | 6 | 7 | 8 | 9
```

<Example Grammar>

ex) A test-case: 0122



Grammars

- More complex grammars for arithmetic expressions
 - Cover full arithmetic expressions

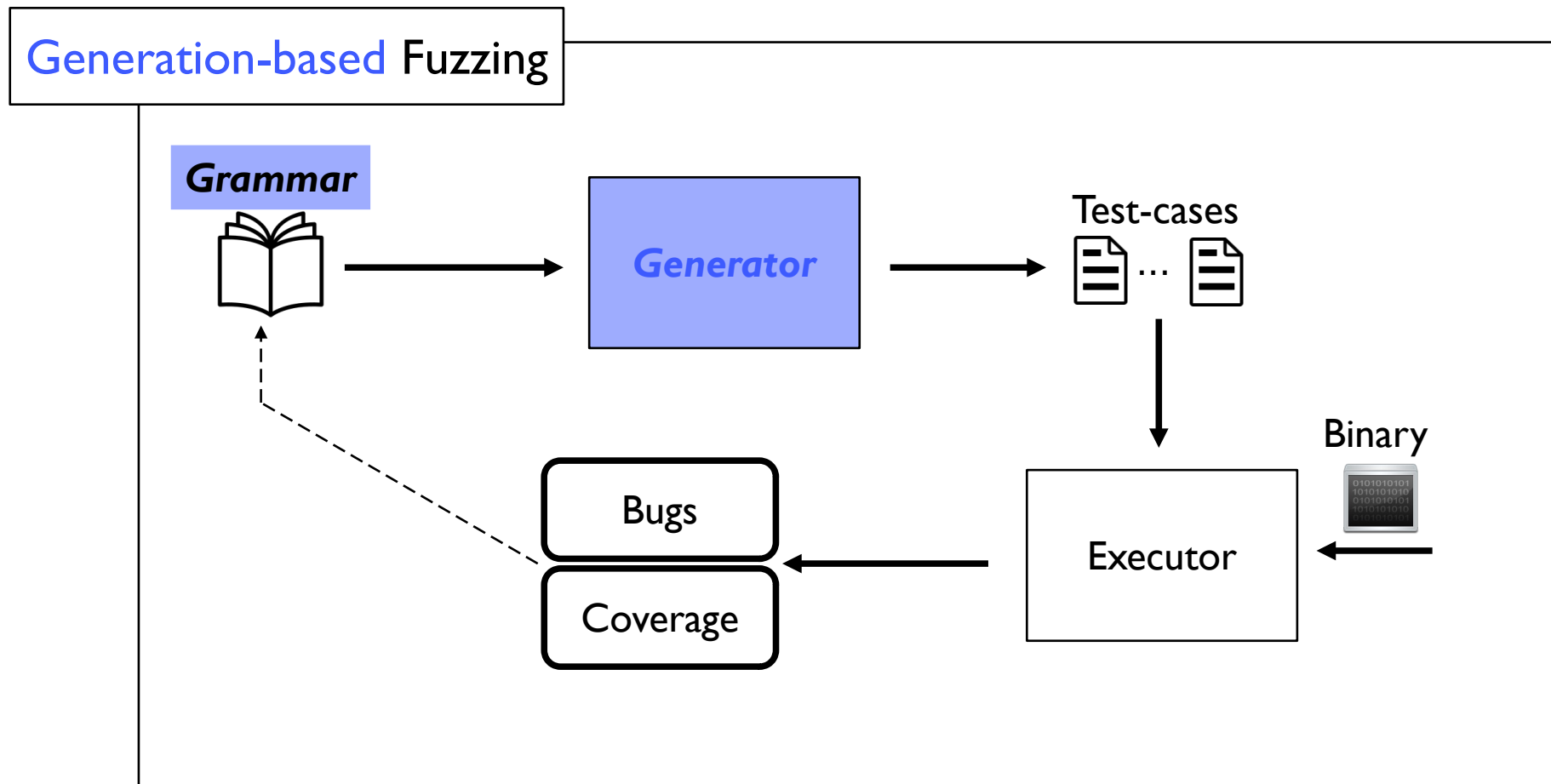
```
<start>    ::= <expr>
<expr>     ::= <term> + <expr> | <term> - <expr> | <term>
<term>      ::= <term> * <factor> | <term> / <factor> | <factor>
<factor>    ::= +<factor> | -<factor> | (<expr>) | <integer> | <integer>.<integer>
<integer>   ::= <digit><integer> | <digit>
<digit>     ::= 0 | 1 | 2 | 3 | 4 | 5 | 6 | 7 | 8 | 9
```

ex) A test-case: $(1 + 2) * (3.4 / 5.6)$

```
<start> → <expr> → <term> → <term> * <factor> → <factor> * <factor> →
(<expr>)*<factor> → (<term> + <expr>)*<factor> → (<factor> + <expr>)*<factor> →
(<integer> + <expr>)*<factor> → (<digit> + <expr>)*<factor> → (1 + <expr>)*<factor> →
(1 + <term>)*<factor> → (1 + <factor>)*<factor> → (1 + <integer>)*<factor> →
(1 + <digit>)*<factor> → (1 + 2)*<factor> → ... → (1 + 2)* (3.4 / 5.6)
```

Topics in Generation-based Fuzzing

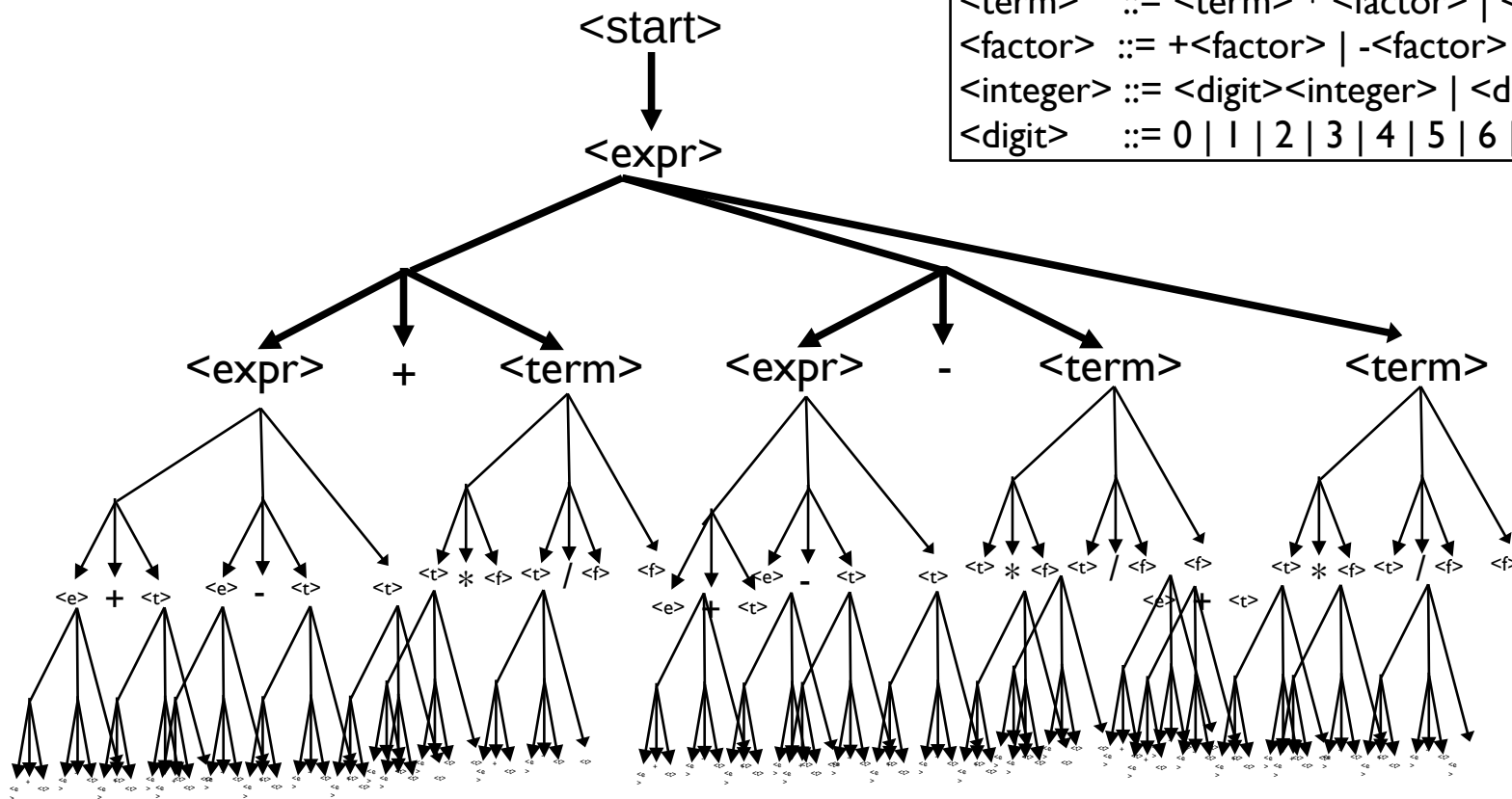
- How to build a good grammar?
- How to expand from a given grammar?



Generator

- How to expand from a given grammar?

<start>	::= <expr>
<expr>	::= <expr> + <term> <expr> - <term> <term>
<term>	::= <term> * <factor> <term> / <factor> <factor>
<factor>	::= +<factor> -<factor> (<expr>) <integer>
<integer>	::= <digit><integer> <digit>
<digit>	::= 0 1 2 3 4 5 6 7 8 9



Generator

- Smart generation from the grammar is very important to the performance of generation-based fuzzing.

```
program
  : HashBangLine? sourceElements? EOF
  ;

sourceElement
  : statement
  ;

statement
  : block
  | variableStatement
  | importStatement
  | exportStatement
  | emptyStatement
  | classDeclaration
  | expressionStatement
  | ifStatement
  | iterationStatement
  | continueStatement
  | breakStatement
  | returnStatement
  ...
```

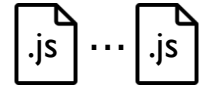
grammar



A grammar-based fuzzer
(Grammarinator)

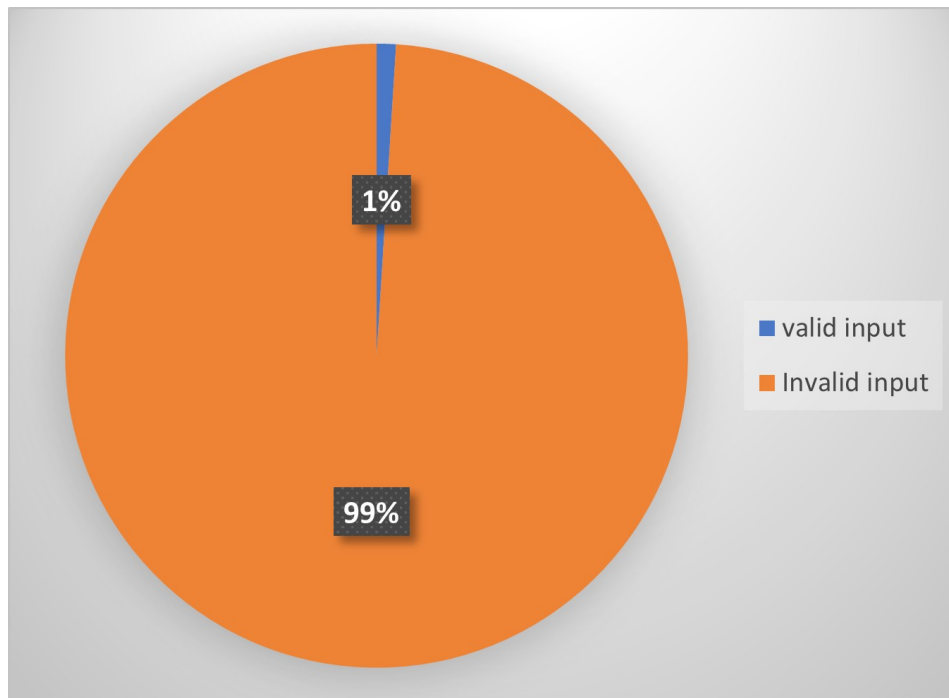
Random
derivation
→

Test-cases
(e.g., javascript files)



Generator

- Random generation from the grammar **does not** generate **valid** JavaScript files.



Examples for 1% valid Inputs

(ex1) `class async{ ; }`

(ex2) `function async() { }`

(ex3) `try{ } catch{ }`

(ex4) `try{ } finally{ }`

Grammar

- Building a good grammar is **also essential**.

```
program
: HashBangLine? sourceElements? EOF
;

sourceElement
: statement
;

statement
: block
| variableStatement
| importStatement
| exportStatement
| emptyStatement
| classDeclaration
| expressionStatement
| ifStatement
| iterationStatement
| continueStatement
| breakStatement
| returnStatement
...

```

Naive grammar



An example for 99% invalid Inputs

(exl). return var l = 4 + 3

(Exception: Syntax Error: Return statements are only valid inside functions)

Summary

- Grey-box testing is classified into **two methods**.
 - **Mutation-based** fuzzing / **Generation-based** fuzzing
- Mutation-based fuzzing has **three key components**.
 - Seed pool generation / Seed selection / **Seed mutation**
- Recent trend in mutation-based fuzzing is ...
 - **Program-agnostic** Fuzzing → **Program-adaptive** Fuzzing
- Generation-based Fuzzing
 - How to build a good grammar?
 - How to expand from a given grammar?

Thank You