

# Software Testing



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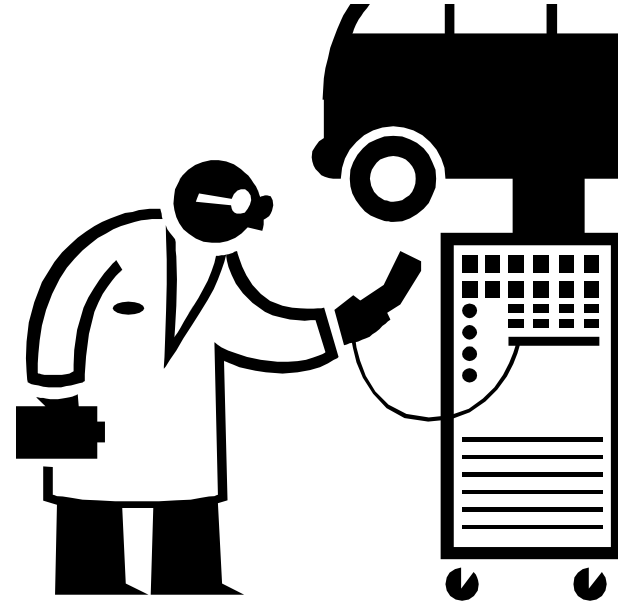
# Outline

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1. Introduction
2. Theory
3. Testing Strategies
  1. Black Box testing
  2. White Box Testing
4. Mutation testing

# Reality

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Even after verifying the design and code  
we will still need to test.

# Some Terminology

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- Reliability: probability that a program runs for a given time without software error.
- Validation: determination that a program is consistent with its requirements.
- Verification: determination that a program is correct with respect to its (formal) spec.
- Testing: examination of the behavior of a program by running it on selected sample data sets (inputs).

# Terminology (2)

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- Unit testing: testing a procedure, function, or class.
- Integration testing: testing connection between units and components.
- System testing: test entire system.
- Acceptance testing: testing to decide whether to purchase the software.

# Terminology (3)

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- Alpha testing: system testing by a user group within the developing organization.
- Beta testing: system testing by select customers.
- Regression testing: retesting after a software modification.

# Dynamic Fault Classification

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- Logic faults: omission or commission.
- Overload: data fields are too small.
- Timing: events are not synchronized.
- Performance: response is too slow.
- Environment: error caused by a change in the external environment.

# Test Harnesses

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Allows us to test incomplete systems.

- Test drivers: test components.
- Stubs: test a system when some components it uses are not yet implemented.

Often a short, dummy program --- a method with an empty body.

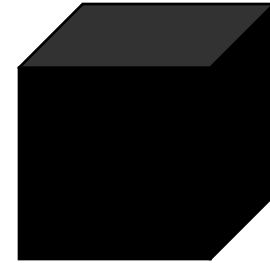


# Test Oracles



- Determine whether a test run completed with or without errors.
- Often a person, who monitors output.
  - Not a reliable method.
- Automatic oracles check output using another program.
  - Requires some kind of executable specification.

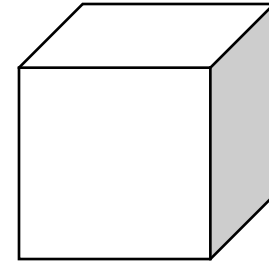
# Testing Strategies: Black Box Testing



- Test data derived solely from specifications.  
Also called “functional testing”.
- Statistical testing.  
Used for reliability measurement and prediction.

# Testing Strategies:

## White Box Testing



- Internal program structure used to derive test cases.
- Often test cases are selected to exercise particular program paths.
- Ideal white box test is to execute “all paths”.
- Also called “structural testing”.

# Testing Theory:

## Why Is Testing So Difficult?

- Theory often tells us what we can't do.
- Testing theory main result: *perfect testing* is impossible.

# An Abstract View of Testing

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- Let program  $P$  be a function with an input domain  $D$  (i.e., set of all integers).
- We seek test data  $T$ , which will include selected inputs of type  $D$ .
  - $T$  is a subset of  $D$ .
  - $T$  must be of finite size.

Why?

# We Need a Test Oracle

- Assume the best possible oracle --- the specification  $S$ , which is function with input domain  $D$ .
- On a single test input  $i$ , our program passes the test when

$$P(i) = S(i)$$

Or if we think of a spec as a Boolean function that compares the input to the output:  $S(i, P(i))$

# Requirement For Perfect Testing

[Howden 76]

1. If all of our tests pass, then the program is correct.

$$\forall x[x \in T \Rightarrow P(x) = S(x)]$$

$$\Rightarrow \forall y[y \in D \Rightarrow P(y) = S(y)]$$

- If for all tests  $t$  in test set  $T$ ,  $P(t) = S(t)$ , then we are sure that the program will work correctly for all elements in  $D$ .
- If any tests fail we look for a program fault.

# Requirement For Perfect Testing

2. We can tell whether the program will eventually halt and give a result for any  $t$  in our test set  $T$ .

$\forall x[x \in T \Rightarrow \text{“}\exists \text{ a computable procedure for determining if } P \text{ halts on input } x\text{”}]$



## But, Both Requirements Are Impossible to Satisfy.

- 1<sup>st</sup> requirement can be satisfied only if  $T = D$ .

We test all elements of the input domain.

- 2<sup>nd</sup> requirement depends on a solution to the ***halting problem***, which has no solution.

We can demonstrate the problem with Requirement 1 [Howden 78].

# Comments



- Key here is the need for finite sized test sets.
- Program domains are not usually finite.
- We seek to determine the behavior of programs with (effectively) infinite domains, using finite sets.
- We try to do the impossible.

# Other Undecidable Testing Problems

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- Is a control path feasible?

Can I find data to execute a program control path?

- Is some specified code reachable by any input data?

These questions cannot, in general, be answered.

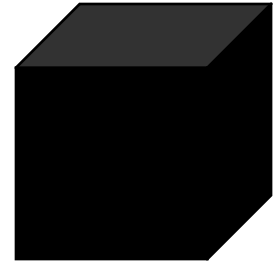
# Software Testing Limitations



- There is no perfect software testing.
- Testing can show defects, but can never show correctness.

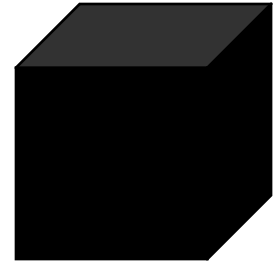
We may never find all of the program errors during testing.

# Black Box Testing: Equivalence Partitioning



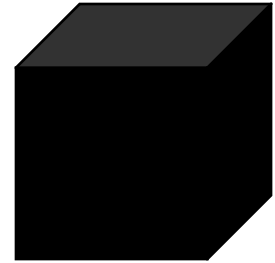
- Partition input domain of a program into a finite number of equivalence classes.
  - With respect to formal specifications.
  - Requires heuristics.
- Select at least one input from each class.

# One Approach to Partitioning



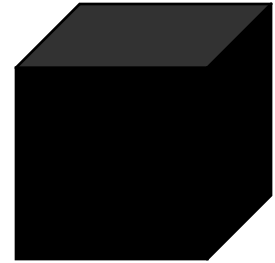
- Valid equivalence classes: valid inputs to the program.
- Invalid equivalence classes: invalid inputs to the program.

# Black Box Testing: Boundary Value Analysis



- Select test cases of inputs just above and below boundaries.  
Edges of equivalence classes.
- Select the ends of ranges of values.  
Both just above and below legal values.

# Black Box Testing: Cause-Effect Analysis

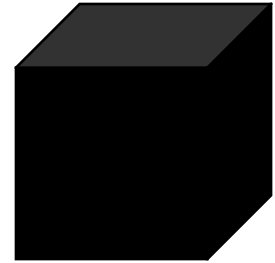


- Rely on pre-conditions and post-conditions and dream up cases.
- Identify impossible combinations.
- Construct decision table between input and output conditions.

Each column corresponds to a test case.



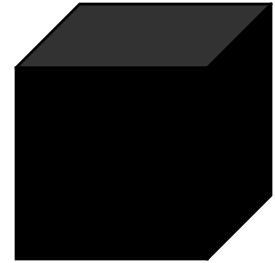
# Black Box Testing: Error Guessing



- “Some people have a knack for ‘smelling out’ errors” [Meyers].
- Enumerate a list of possible errors or error prone situations.
- Write test cases based on the list.

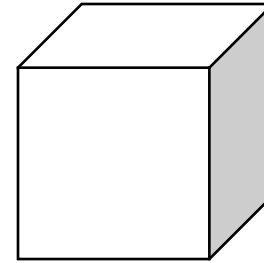
Depends upon having *fault models*, theories on the causes and effects of program faults.

# Black Box Testing: Random Testing



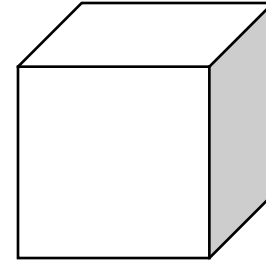
- Generate tests randomly.
- “Poorest methodology of all” [Meyers].
- Promoted by others.
- Statistical testing:
  - Test inputs from an operational profile.
  - Based on reliability theory.
  - Adds discipline to random testing.

# Structural White Box Testing



- Assume that a “path” through a program represents one relevant partition of the input domain.
- Choose one input from each path.
- Alas, a program with a loop has potentially an infinite number of paths.

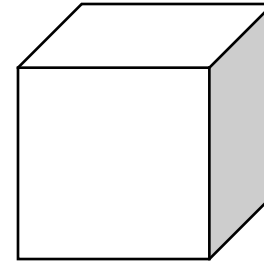
# We Need to Find



1. Finite subsets  $FS(P)$  of the set of all paths through program  $P$ .
2. Choose Data to test each  $p \in FS(P)$ .

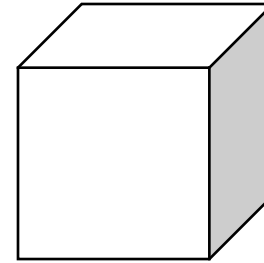
Does this guarantee that every time path  $p$  executes, it gives the correct output?

# Choosing Data To Exercise Paths



- Symbolic execution.
  - Solve Boolean equations to find input data.
  - Some paths are infeasible.
  - Satisfiability problem: NP-complete for simple Boolean expressions. Undecidable for expressions containing ints or reals.
- Generate tests, then see how many paths are covered.

# One Abstraction for Program Analysis



- Flowgraph directed graph:

$$G = (N, E, s, t)$$

$N$ : set of nodes.

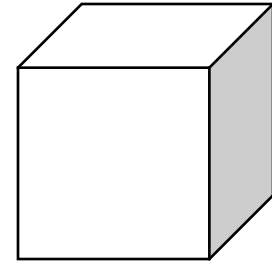
$E$ : set of edges  $E \subseteq N \times N$

$s$ : start node  $s \in N$

$t$ : terminal node  $t \in N$

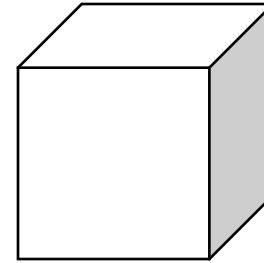
Invariants:  $N \neq \emptyset$ ,  $E \neq \emptyset$ , every  $n \in N$  lies on a path from  $s$  to  $t$ .

# Mapping Programs To Flowgraphs



- Basic Block:
  - Maximal length sequential block of command-level code.
  - All code in a block is always executed sequentially.
- Map each basic block to one  $n \in N$ .
- *Each edge  $(nx, ny) \in E$  represents a possible transfer of control from the end of the  $nx$  block to the start of the  $ny$  block.*

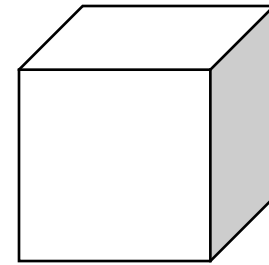
# Mapping (2)



- $s$ ,  $t$  do not necessarily represent a basic block in code.
- Any path from  $s$  to  $t$  in a flowgraph  $G$  represents a potential execution path.
- $EP(G)$ : set of all paths from  $s$  to  $t$ .
- Some paths from  $s$  to  $t$  are infeasible.

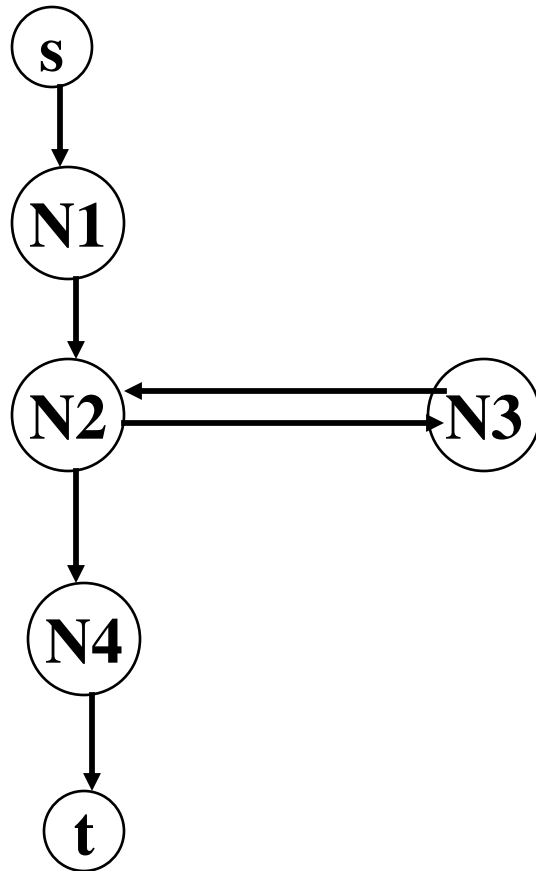
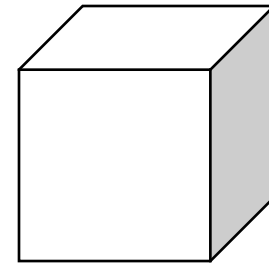


# Consider Method sum



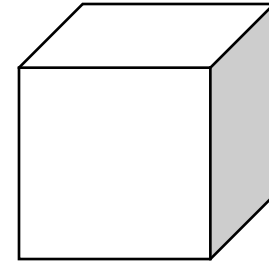
```
int sum(int a[]) {  
    int i=0; int total = 0; /* N1 */  
    while (i < a.length) /* N2 */  
    {  
        total = total + a[i]; i++; /* N3 */  
    }  
    return total; /* N4 */  
}
```

# Flowgraph for sum



- $|Paths(P)| = \infty$
- $\langle s, N1, N2, N4, t \rangle$  is infeasible.
- **But there are an infinite number of feasible paths.**

# Structural Testing Strategies

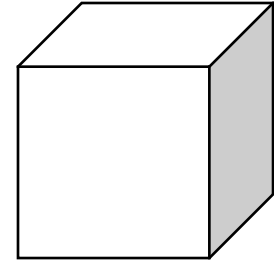


- To specify a structural testing strategy (flowgraph based) we specify criteria for finite sets of execution paths:

$$FS(G) \subseteq EP(G)$$

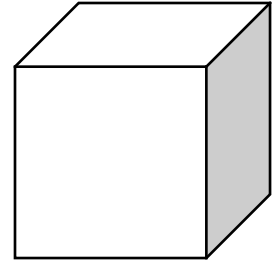
- $|FS(G)|$  is finite.

# Some Common Criteria



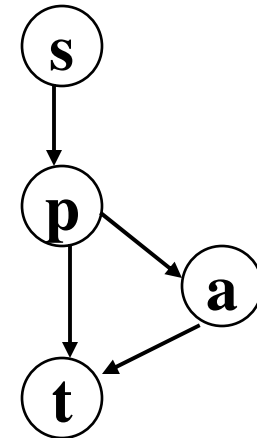
- Control flow based criteria:
  - Node or statement coverage.
  - Edge or branch coverage.
$$|\text{FS}(G)_{\text{NodeCov}}| \leq |\text{FS}(G)_{\text{EdgeCov}}|$$
- Paths to test can be determined from the control flow graph alone.

# Example: Edge Coverage is Stronger than Node Coverage

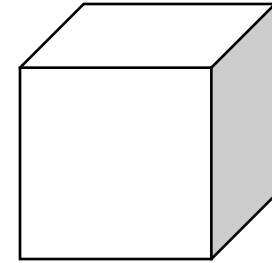


- Code: if (p) a;
- Node coverage:  
 $Pnodes = \{ \langle s, p, a, t \rangle \}$
- Edge coverage:  
 $Pedges = \{ \langle s, p, a, t \rangle, \langle s, p, t \rangle \}$

## Flowgraph



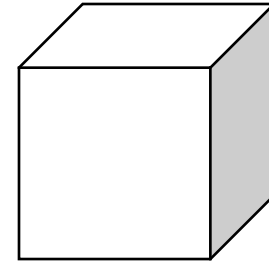
# Other Control Flow Criteria



- All acyclic paths.
- Go through each “loop” 0 & 1 time.  
Acyclic paths + each “loop” (or rather cycle) exercised once.
- McCabe’s criteria: test each “linearly independent path”.

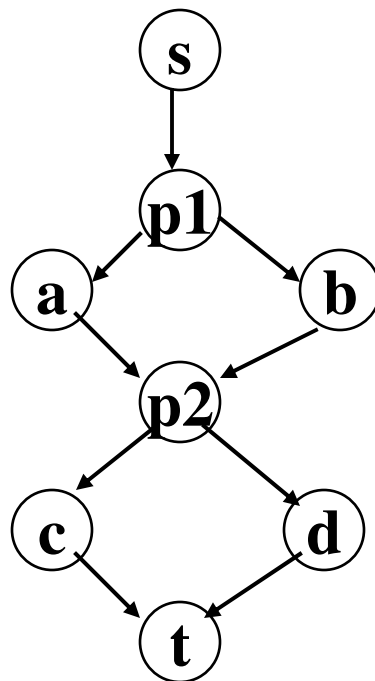
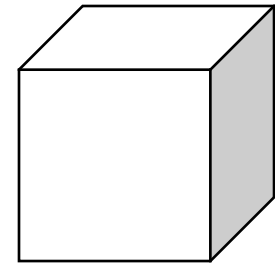
# McCabe's Criteria

## Basis Path Testing



- Each path in the finite set contains one edge not on any other path & the set is an edge covering.
- Cyclomatic number of paths to test.  
Number of decisions + 1 paths.
- Might not include all acyclic paths.

# McCabe's Criteria Misses Acyclic Paths

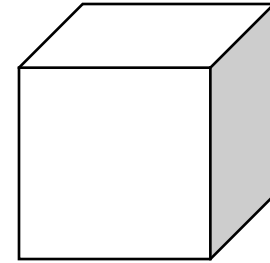


**BPaths** = { $\langle s, p1, a, p2, c, t \rangle$ ,  
 $\langle s, p1, b, p2, c, t \rangle$ ,  
 $\langle s, p1, a, p2, d, t \rangle$ }  
meets McCabe's criteria.

**All acyclic paths =**  
**BPaths  $\cup$  { $\langle s, p1, b, p2, d, t \rangle$ }**



# McCabe's Anomoly



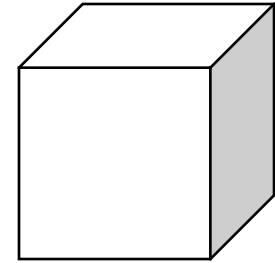
- Number of acyclic paths in 2 or more adjacent sequential flowgraph components is computed by multiplying the number of paths in each component.

$$|\text{AcyclicP}(F)| = |\text{AcyclicP}(I)| \times |\text{AcyclicP}(II)|$$

- Number of basis paths is a linear function on the number of decisions:

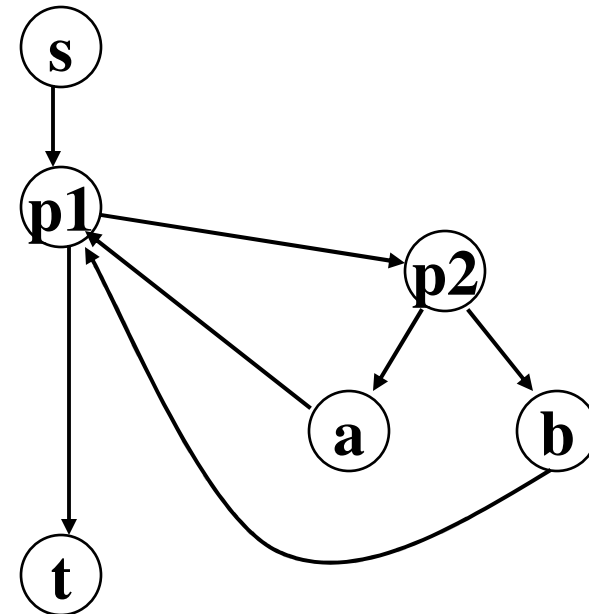
$$|\text{Bpaths}(F)| = \# \text{ decisions} + 1$$

# All Acyclic Paths + Once Through Each Cycle

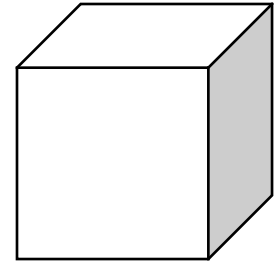


- if-then-else in a while loop.
- $FS(G) = \{ \langle s, p1, t \rangle, \langle s, p1, p2, a, p1, t \rangle, \langle s, p1, p2, b, p1, t \rangle \}$
- Alternative  $FS(G) = \{ \langle s, p1, t \rangle, \langle s, p1, p2, a, p1, p2, b, p1, t \rangle \}$

## Flowgraph G:

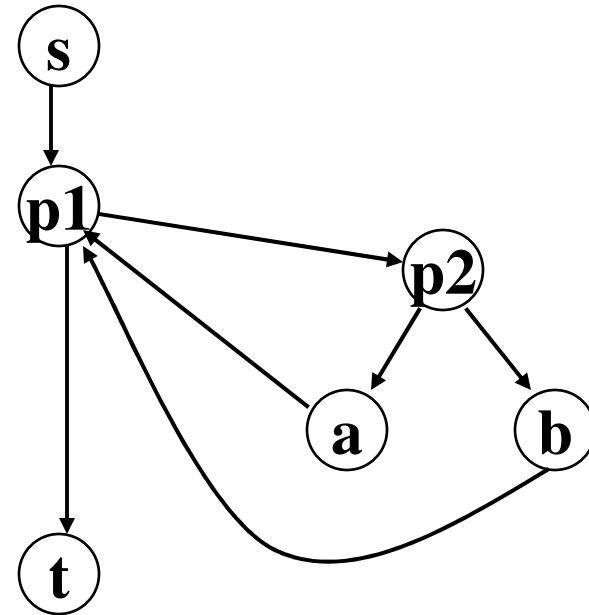


# Note that

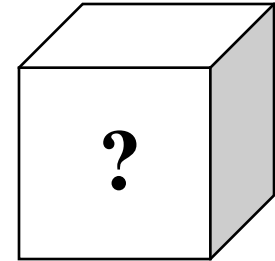


- $b \in \text{succ}(a) \wedge$   
 $a \in \text{succ}(b) \wedge$   
 $a \in \text{succ}(a) \wedge$   
 $b \in \text{succ}(b)$
- Errors can lurk in these cases.
- Should we include all such possibilities?

## Flowgraph G:

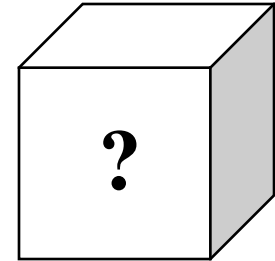


# Mutation Testing



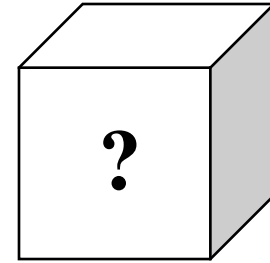
- Not a testing technique.
- Used to evaluate test data adequacy.
- Let  $P$  be a program,  $S$  is a specification, &  $T = \{t1, \dots, tn\}$  be a set of test input data.
- Assume  $\forall t [t \in T \Rightarrow (P(t) = S(t))]$   
so  $P$  appears correct according to  $T$ .

# Is $T$ a Good Test Set?



- We can create a set of mutants  $M$  of  $P$ .
- $m \in M \Rightarrow m$  “differs from  $P$ ”.  
We inject a fault in  $P$  to create a mutant.
- We run each mutant  $m \in M$  on test data  $T$ .
- A mutant  $m$  may live or die:  
$$Lives(m) \equiv \forall t [t \in T \Rightarrow S(t) = m(t)]$$
$$Died(m) \equiv \neg Lives(m)$$

# Mutants Live or Die

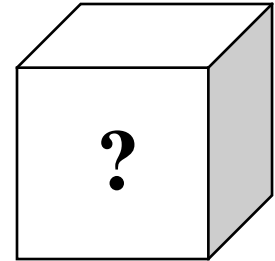


- The more mutants to live, the lower our confidence that  $P$  is actually correct.
- Why? Many mutants appear as good as  $P$ .

A live mutant survived the tests as well as  $P$ .

- We assume that mutation is harmful.

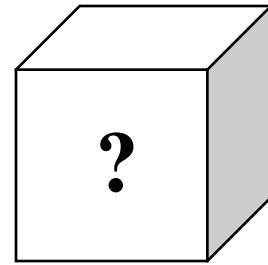
# What if Mutants Survive?



- $T$  can be expanded to kill the mutants.
- The key to mutation analysis:  
Generating some *plausible mutants* ---  
mutants with some possibility of survival.

*Mutate:*

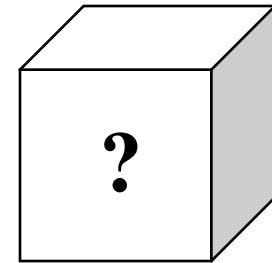
*program*  $\rightarrow$  *set of programs*



- Each mutant  $m \in \text{Mutate}(m)$  must be syntactically correct or it is not a program.
- Syntax preserving transformations must be used.

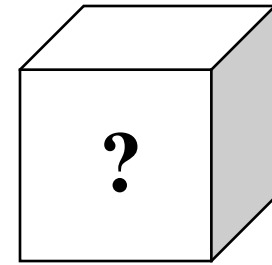


# Mutant Operators --- Mutate Functions



- Data objects --- constants, scalar variables, array references:
  - At each reference, generate mutants referencing all other accessible data objects.
  - Change values of a constant (a small change).
  - Change array referenced to another array.

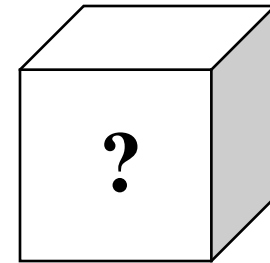
# Mutant Operators (2)



- Operators

- Change operators to different ones of the same type (arithmetic, relational, logical).
- Replace logical expression with true or false.
- Delete subexpressions.
- Delete unary operators.

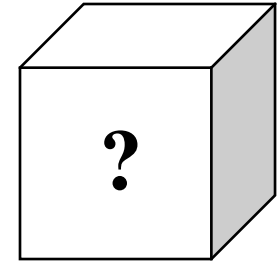
# Mutant Operators (3)



- Statements
  - Delete statements.
  - Change to return statements.
  - Change order of assignments.
- Scramble labels.

Many mutants can be generated.

# Mutations that Focus on Possible Errors



- Consider statement:  
if (  $i * j + 3 > k$  ) blah;
- Mutated predicates:

(  $(i + 1) * j + 3 > k$  )

(  $(i - 1) * j + 3 > k$  )

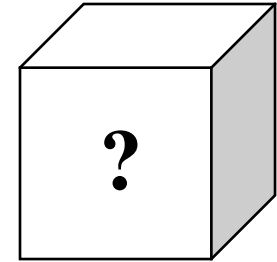
(  $i * (j + 1) + 3 > k$  )

(  $i * j + 2 > k$  )

(  $i * j + 4 > k$  )

Move predicate boundary slightly --- test correctness of predicate boundary.

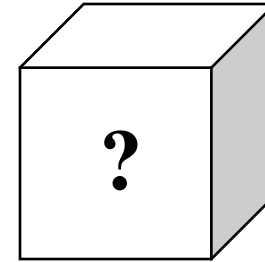
# More Mutations



- Consider statement:  
 $a = b + c;$
- Mutants:  
 $a = \text{abs}(b) + c;$   
 $a = b + \text{abs}(c);$   
 $a = \text{abs}(b + c);$
- To kill these mutants we must include tests where:
  - $b$  is negative
  - $c$  is negative
  - $b + c$  is negative.

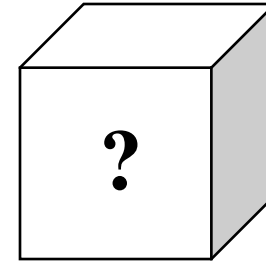
# Comments

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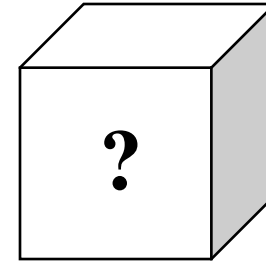
- Mutation testing is not infallible.
- MOTHRA: a current mutation system in experimental use at Purdue [DeMillo et al] used to assist in performing structural & functional testing.
- TDS for testing distributed CORBA systems [Ghosh].

# Bebugging [Mills]



- Error seeding to generate error statistics.
- Analogy from ecology:
  - 1000 fish caught in lake, marked & released.
  - Later catch a new batch of 1000 fish.
  - We find 100 fish are marked.
  - We are justified in saying that the true number of fish in the lake is between 8,500 and 12,000.

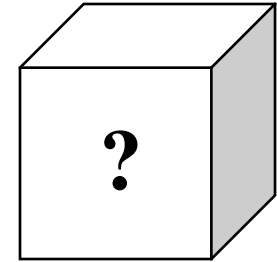
# Error Seeding



- Marked fish  $\leftarrow$  inserted errors.
- Insert errors and use a similar analysis to discover the actual number of errors.



# Problems



- Need a large number of errors in the project for the statistics to work.  
Need on the order of 10,000 errors.
- Errors are not uniformly distributed like fish are.
- Tests may discover artificial errors easier than natural ones.
- Seeded errors may differ from natural ones.

# Competent Programmer Hypothesis



- Programmers are not adversaries.
- Most errors are:
  - Simple in form.
  - Well understood.
  - Classifiable.
- Hypothesis is necessary for any effective testing strategy.

# Summary

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- Testing is very difficult.
  - Theory: testing for correctness is impossible.
  - All methods fail to find all errors.
- Random testing for evaluating reliability.
- Black box testing: use the specification to define tests.
- White box testing: use the internal structure to define tests.
- Fault injection for test evaluation.