

# Mutational Robustness and Automatic Program Repair

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YOU KNOW THIS METAL  
RECTANGLE FULL OF  
LITTLE LIGHTS?

YEAH.



I SPEND MOST OF MY LIFE  
PRESSING BUTTONS TO MAKE  
THE PATTERN OF LIGHTS  
CHANGE HOWEVER I WANT.

SOUNDS  
GOOD.



BUT TODAY, THE PATTERN  
OF LIGHTS IS *ALL WRONG!*

OH GOD! TRY  
PRESSING MORE  
BUTTONS!  
*IT'S NOT  
HELPING!*

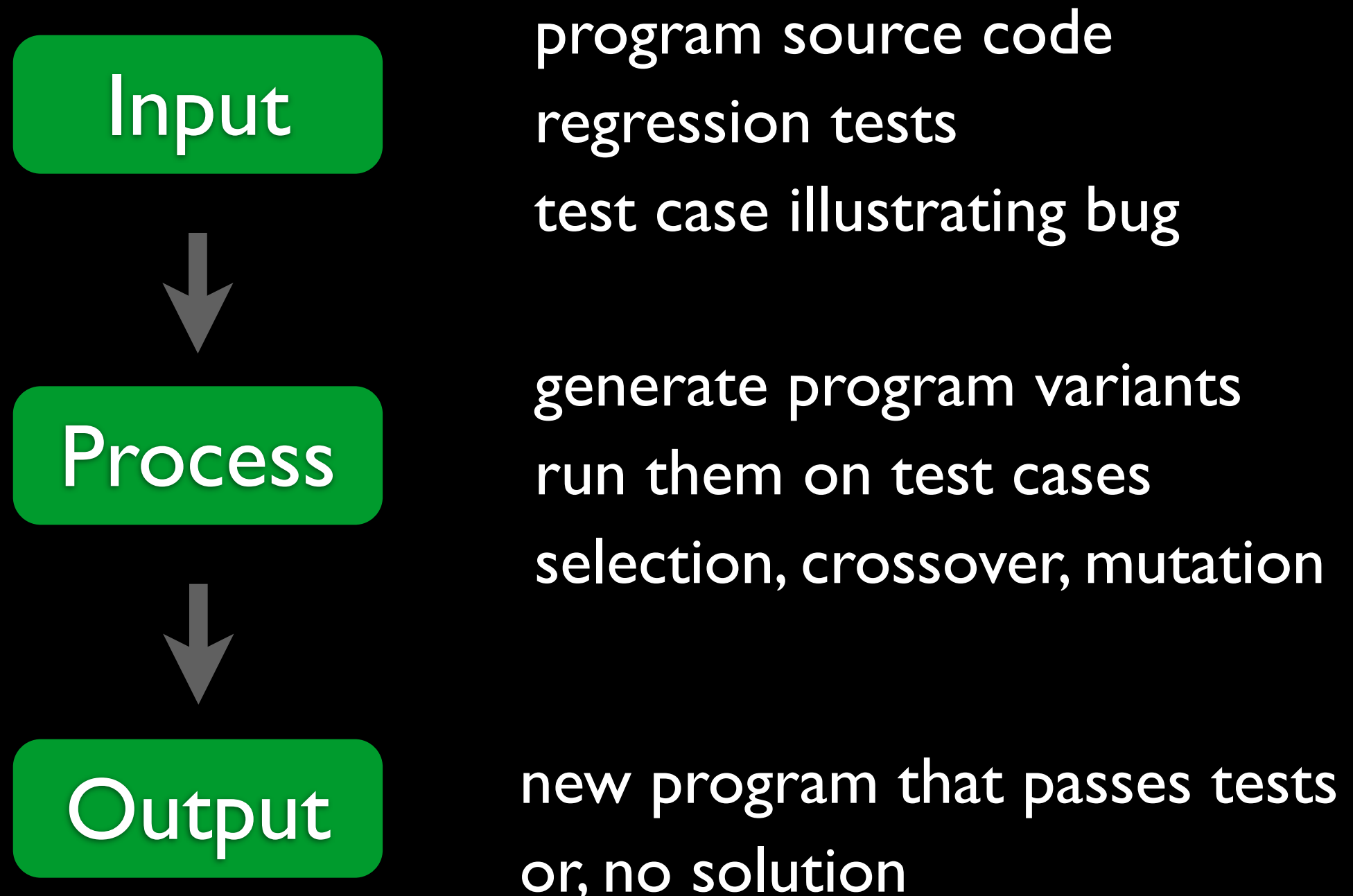


# Automatic Program Repair via Genetic Programming

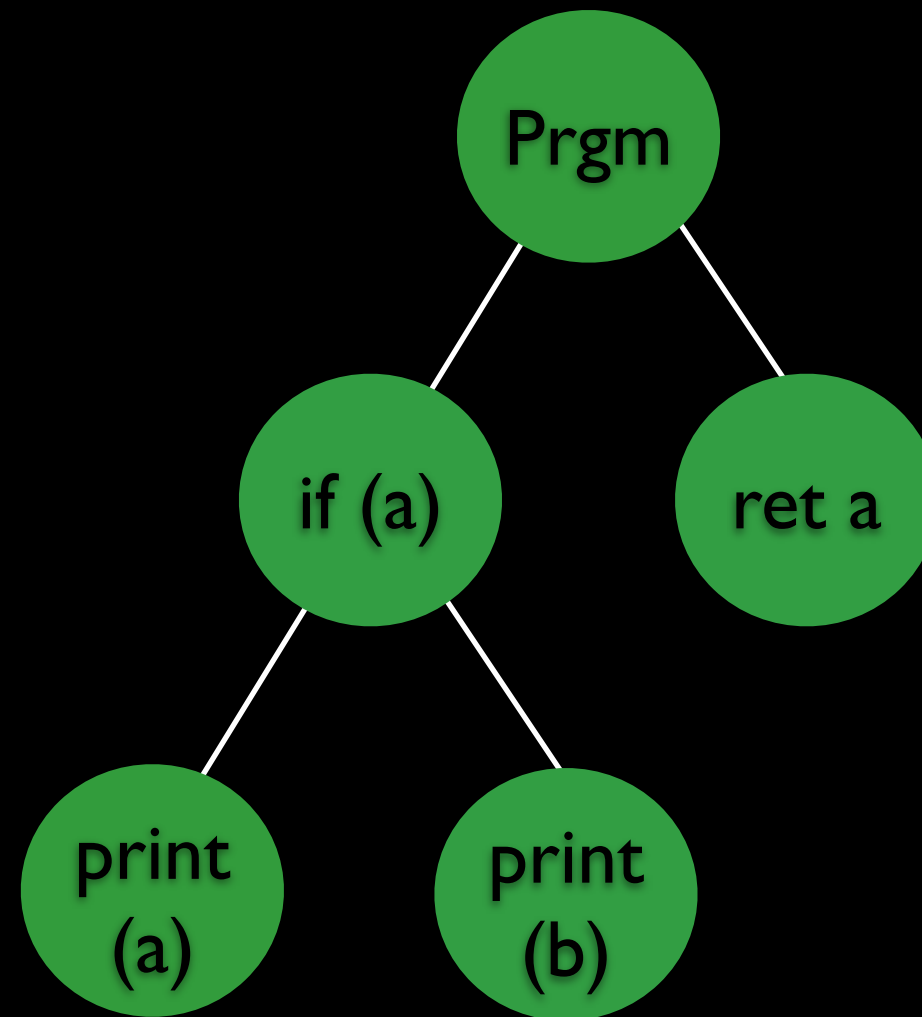
Weimer and Forrest

An optimization technique inspired by evolution

# GP Program Repair



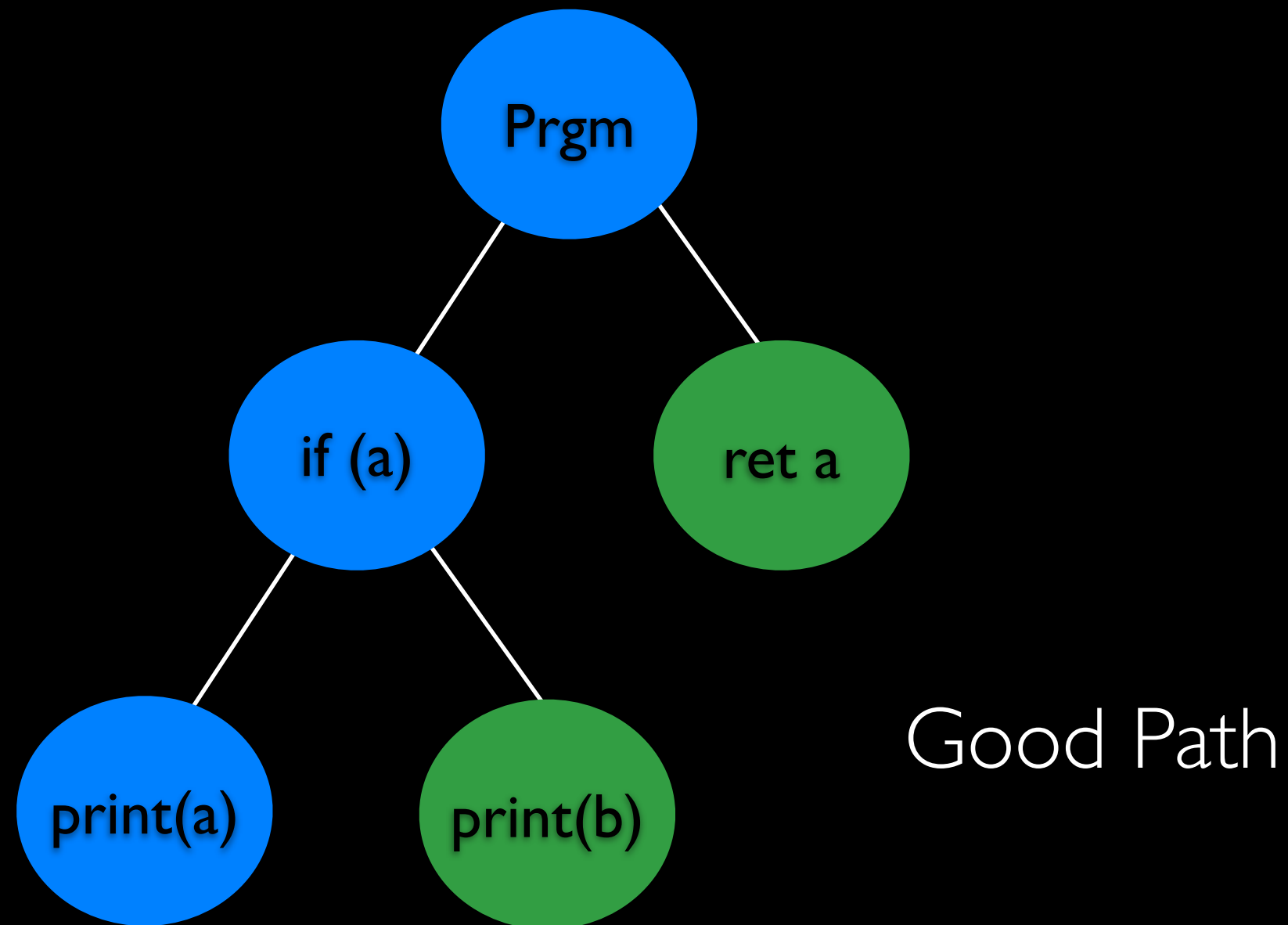
# Representation



Individuals represented as ASTs

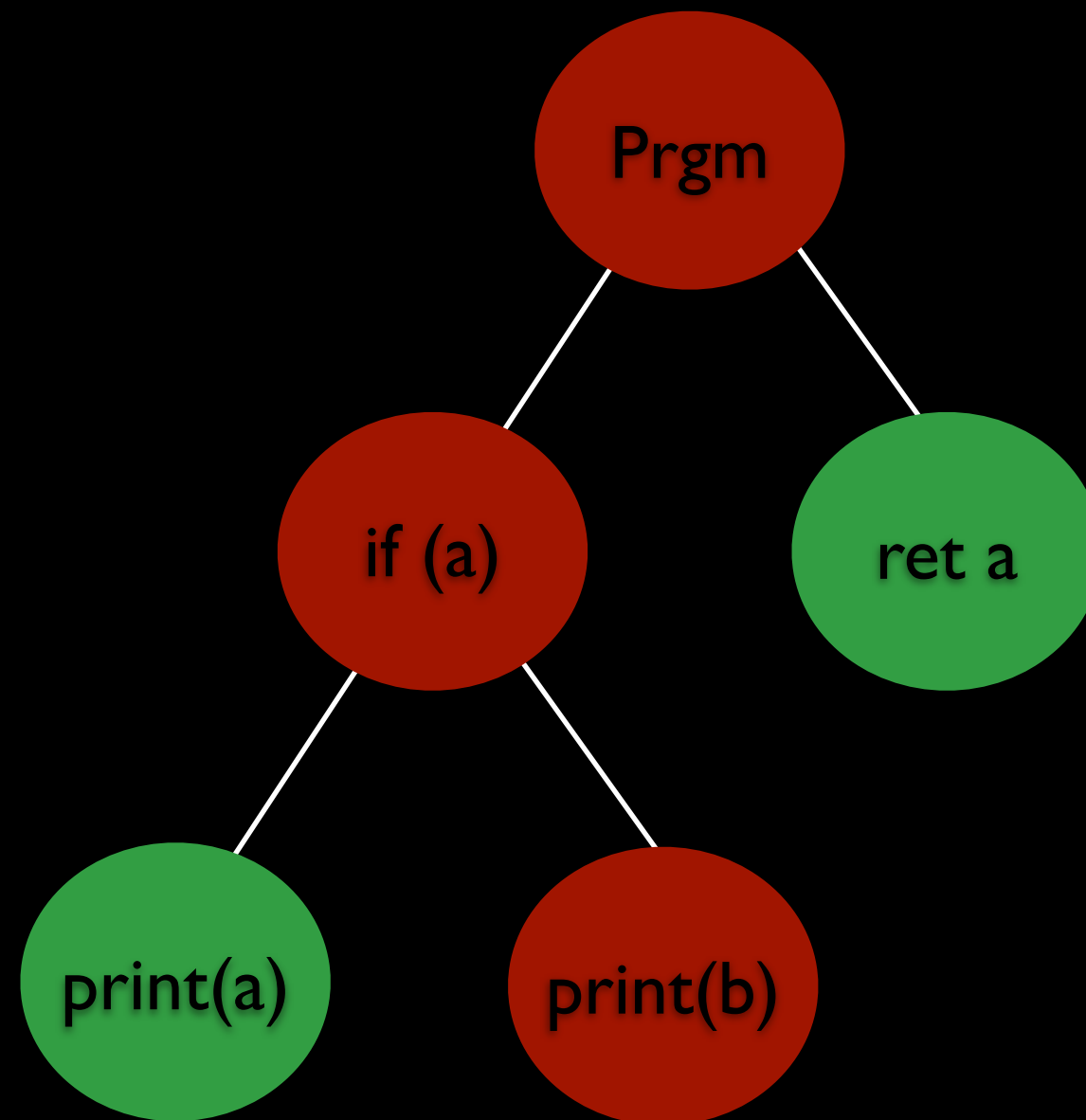
# Weighted Path

A means of fault localization



# Weighted Path

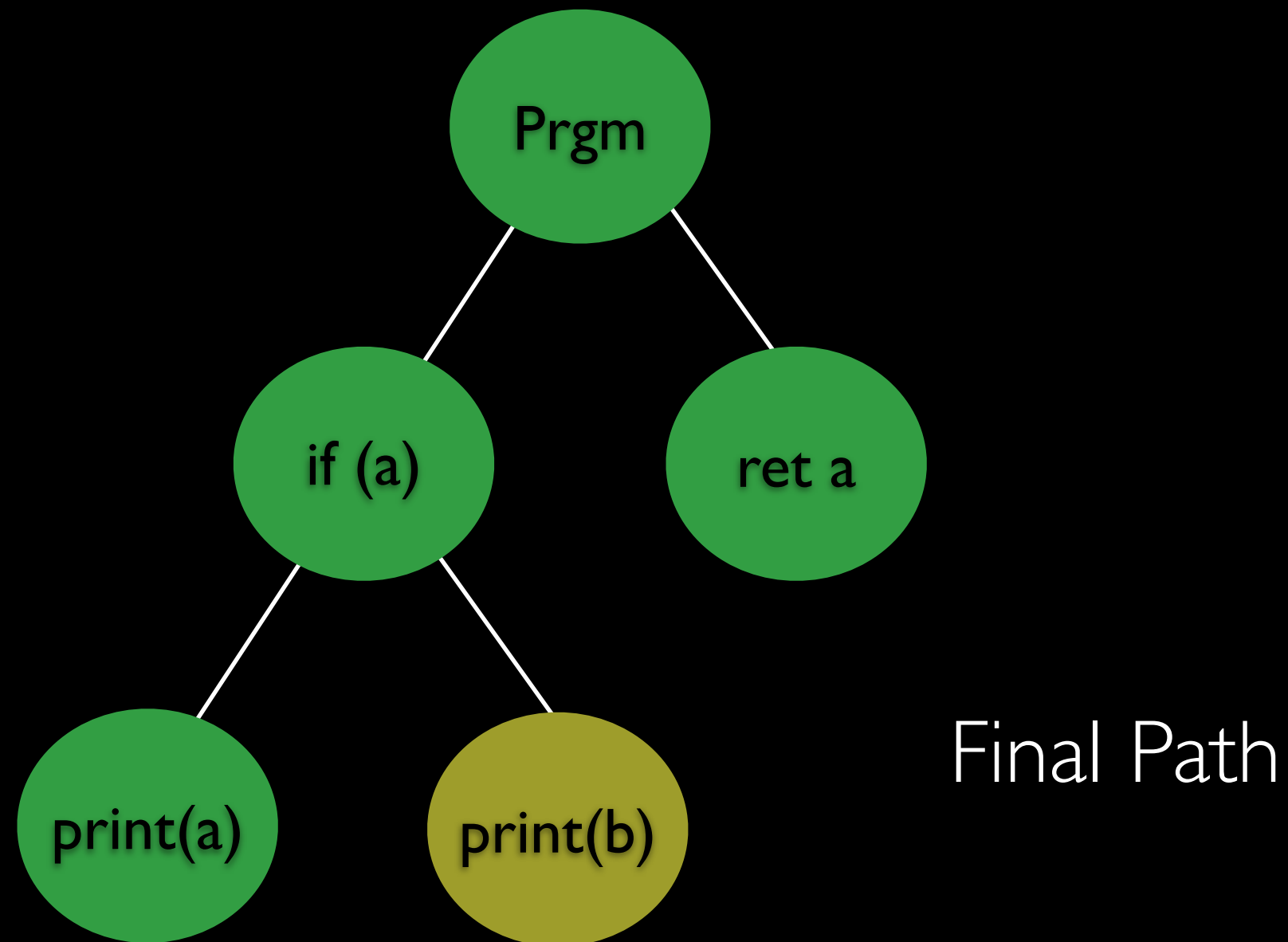
A means of fault localization



Bad Path

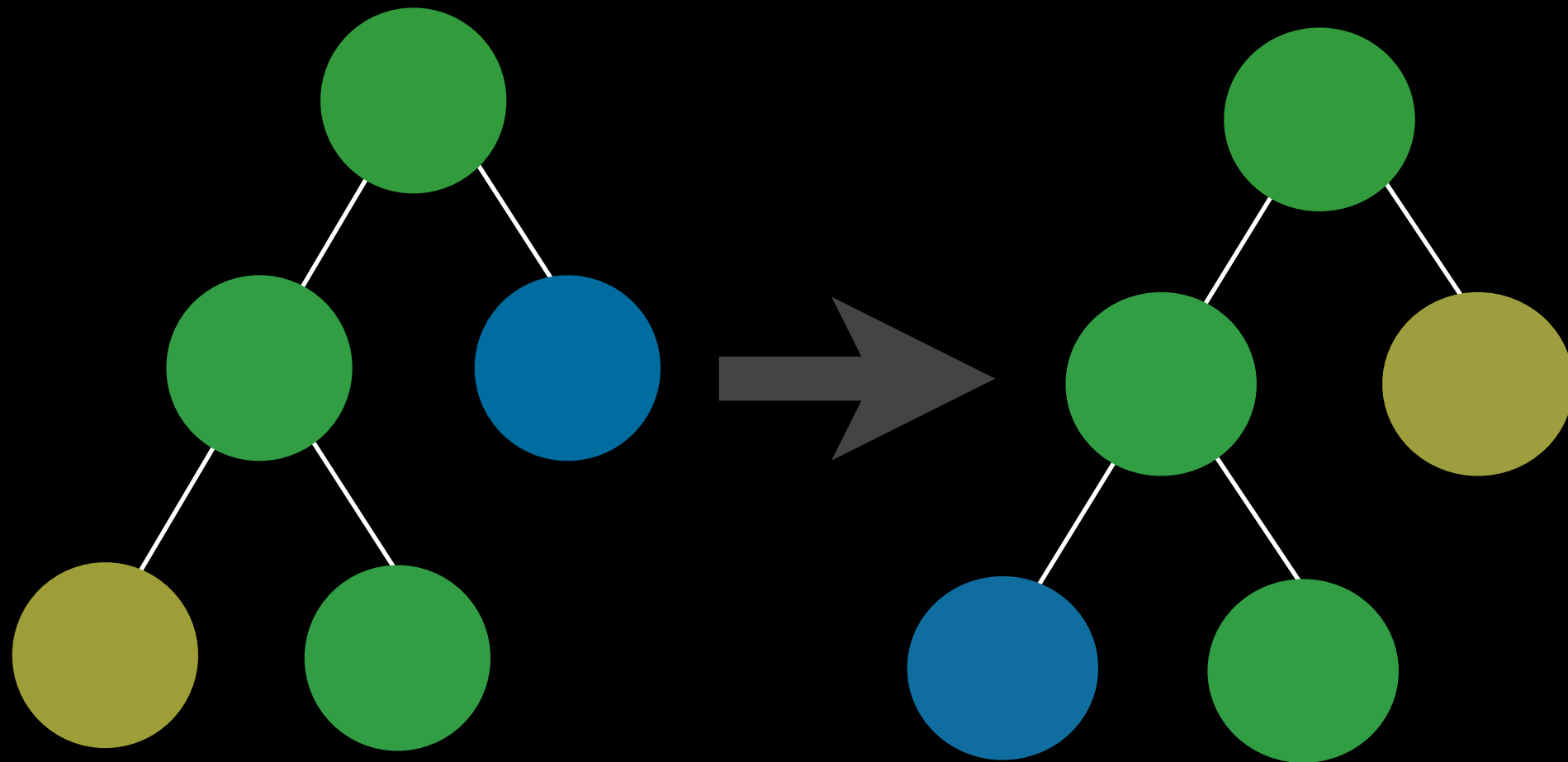
# Weighted Path

A means of fault localization



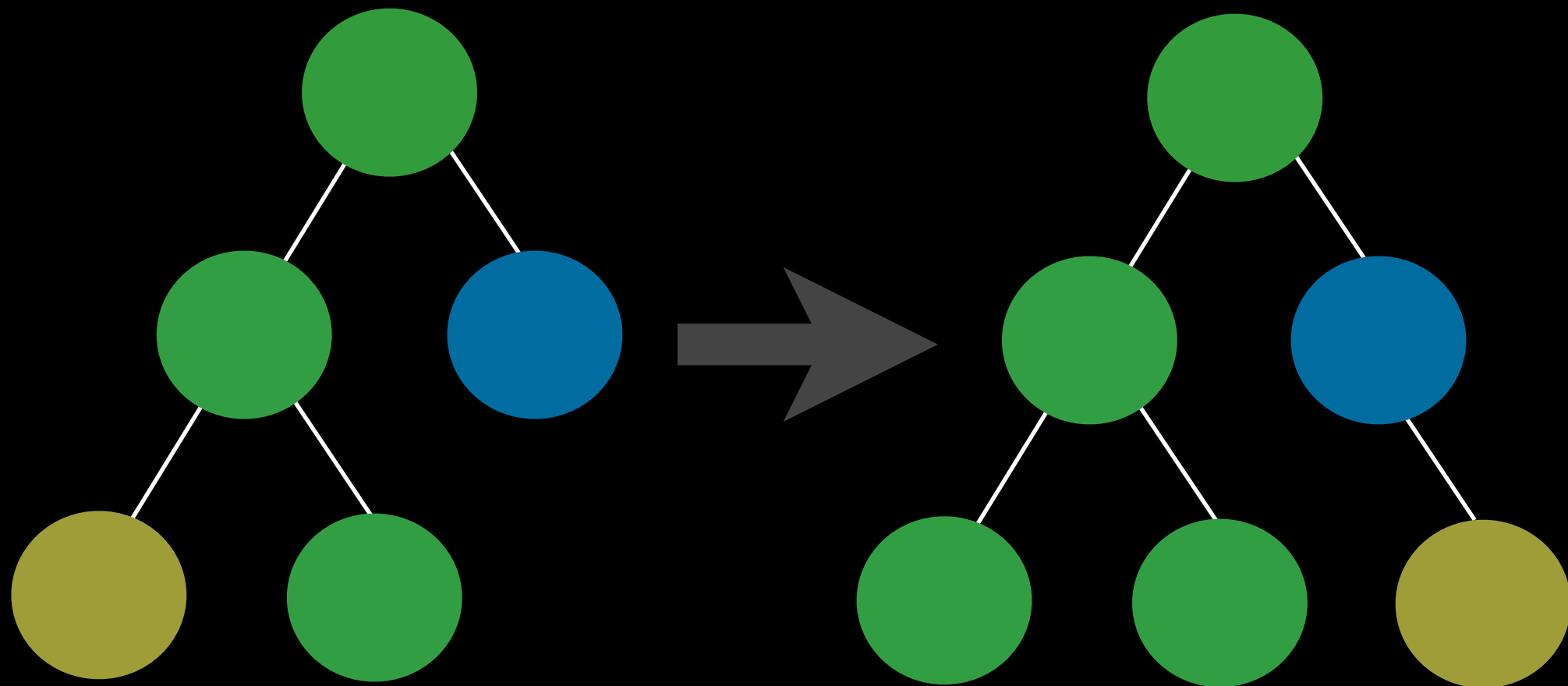


# Mutation: Swap



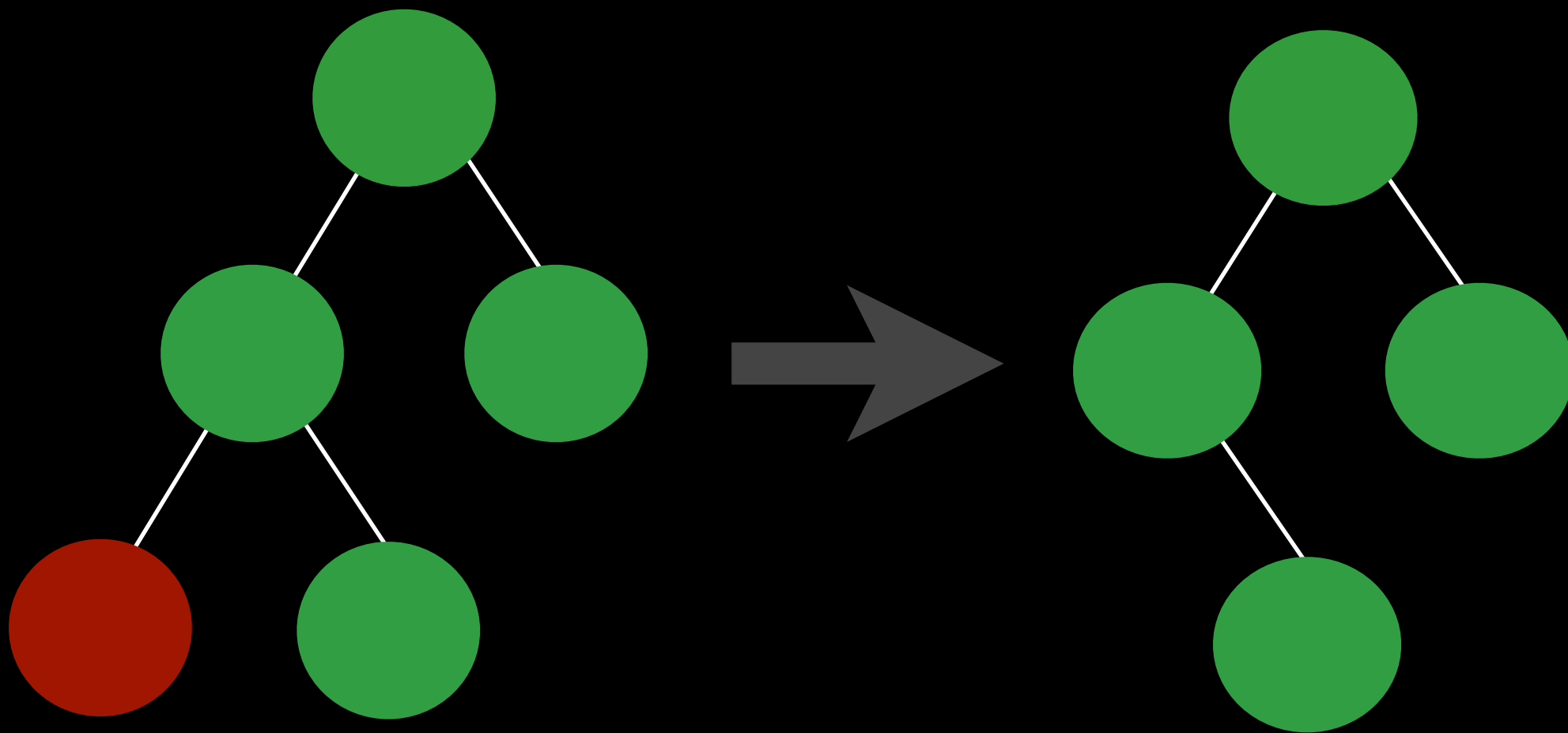
Exchange two nodes on the tree

# Mutation: Append



Copy a node to elsewhere on the tree

# Mutation: Delete



Delete a node from the tree

# GP Program Repair Details

To compute fitness, compile a variant

If it fails to compile, then  $\text{fitness} = 0$

Otherwise, run test cases

Now,  $\text{fitness} = \# \text{ tests passed}$

Negative test case(s) more heavily weighted

# Does it actually work?

deroff	gcd	look
indent	uniq	zune
atris	leukocyte	imagemagick
tiff	nullhttpd	python
php	lighttpd	openldap

A few repaired programs

So what about robustness?

# Some Definitions

**mutational robustness:** the probability of a change in genotype affecting a change in phenotype

**neutral fitness landscape:** described by region of differing genotypes assigned the same fitness value

# Motivation

High mutational robustness seems to support the idea of **evolving** software

Robustness and neutral fitness may be key ideas for repairing more **complicated bugs**



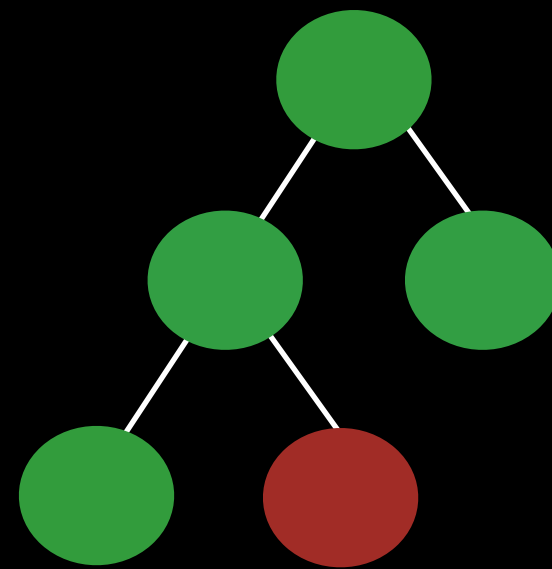
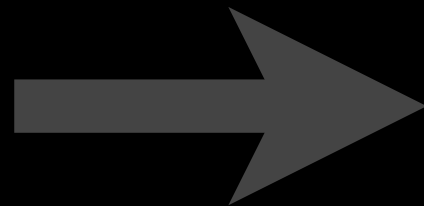
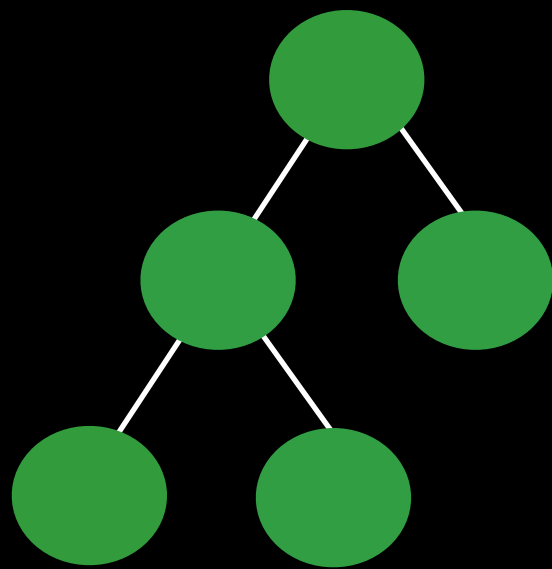
# Questions

How do we measure robustness?

Given a metric, how mutationally robust are typical programs?

How does robustness affect automatic program repair?

# Measuring Robustness



Original Program

Apply Mutation (x1000)

Metrics:

Average distance in fitness

Percent of mutations that are neutral

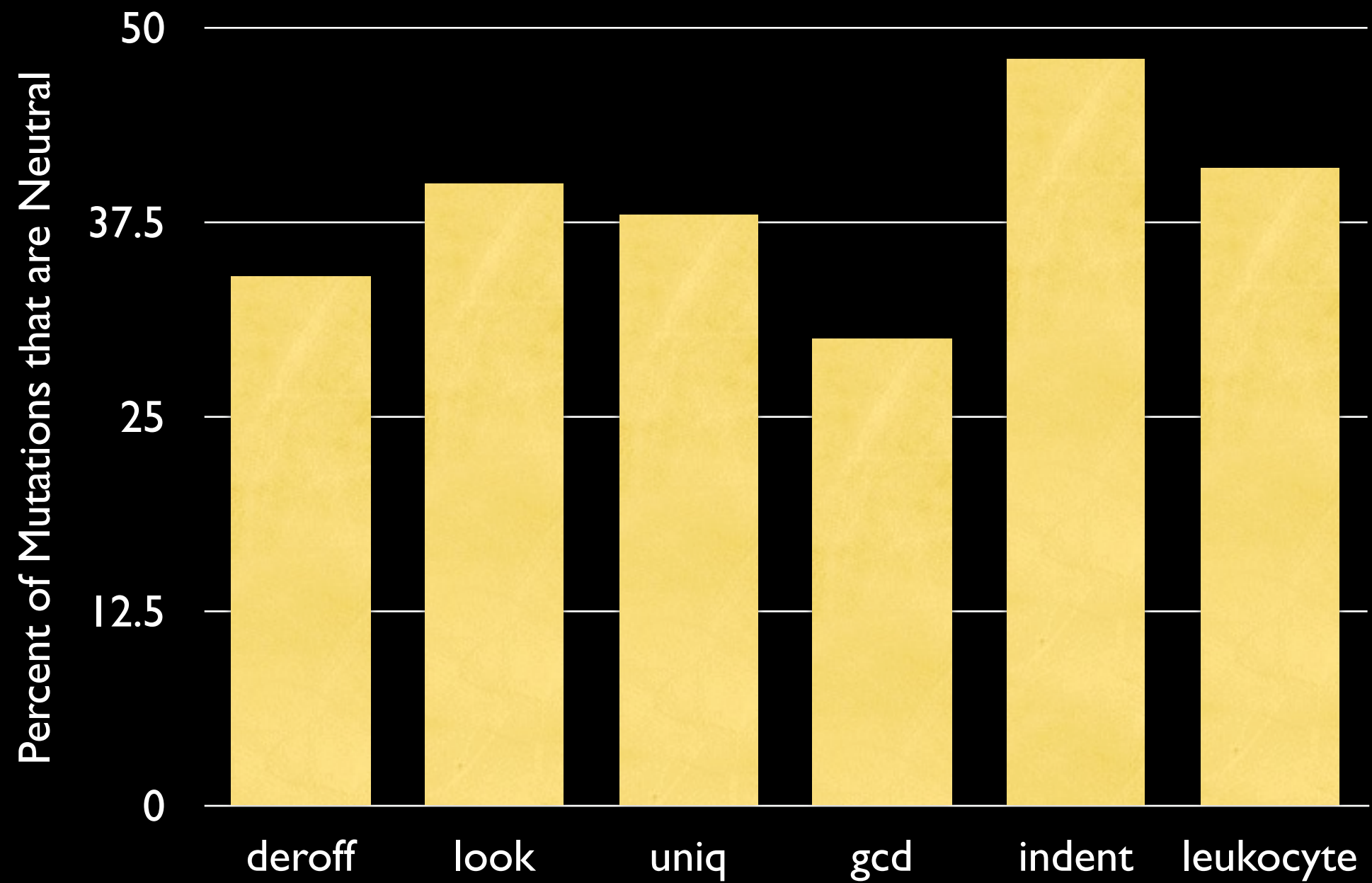
# Your Intuition

(A walk down the garden path)

Suppose that we make a single mutation to  
some arbitrary program.

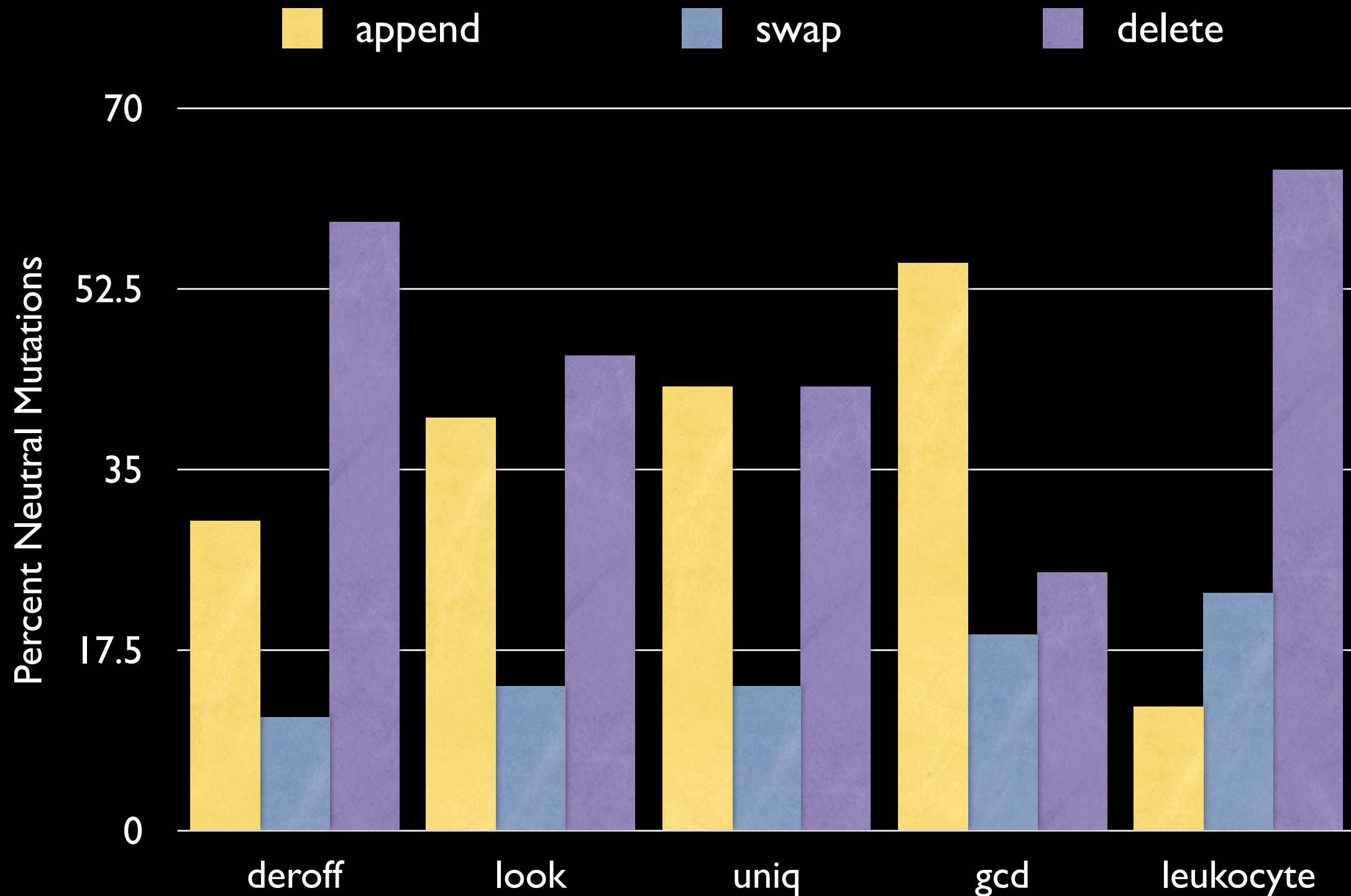
How often will its behavior change?

# Neutral Mutations



What **mutation operators**  
are likely to result in  
neutral mutations?

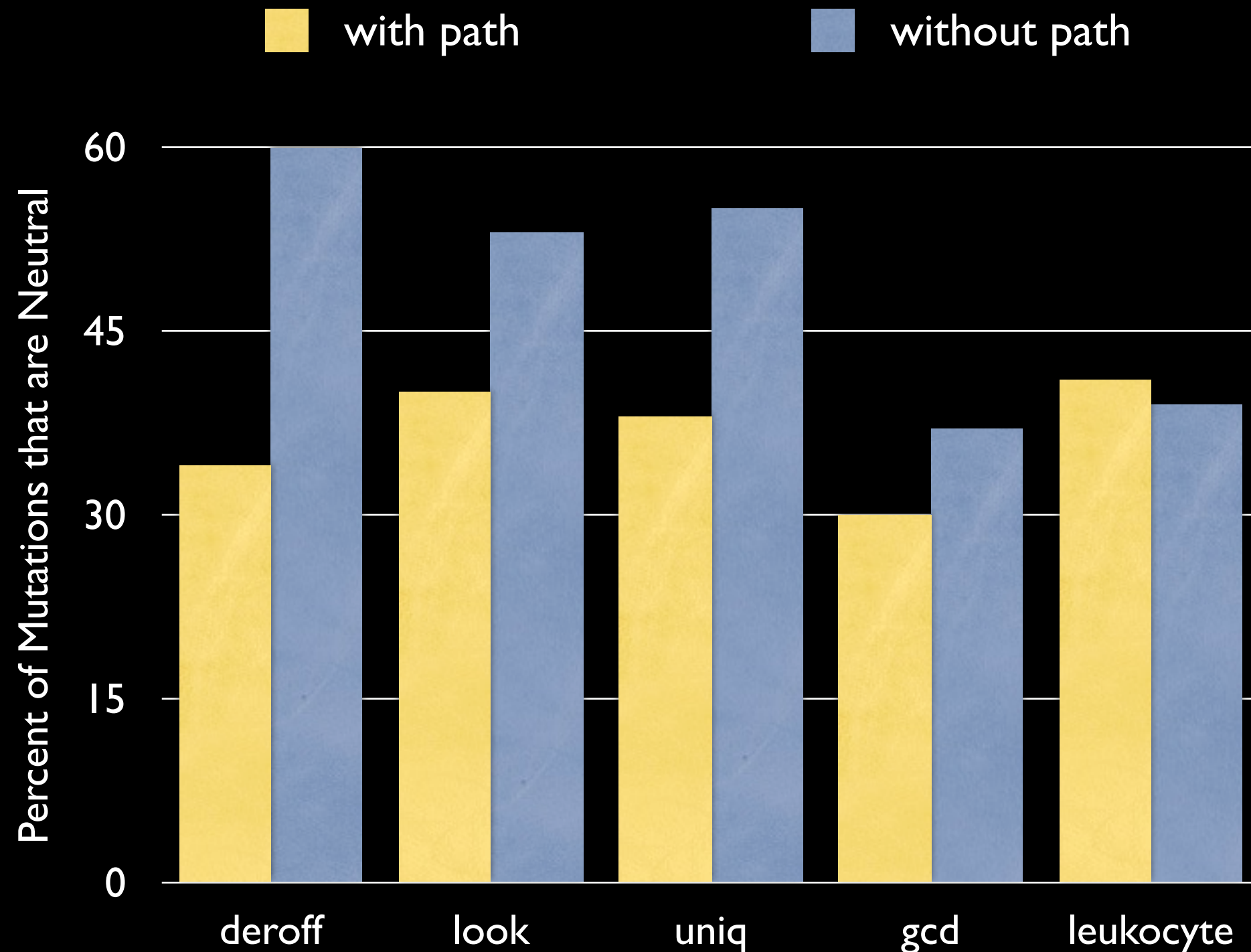
# By Mutation Operators



So what about that  
weighted path?

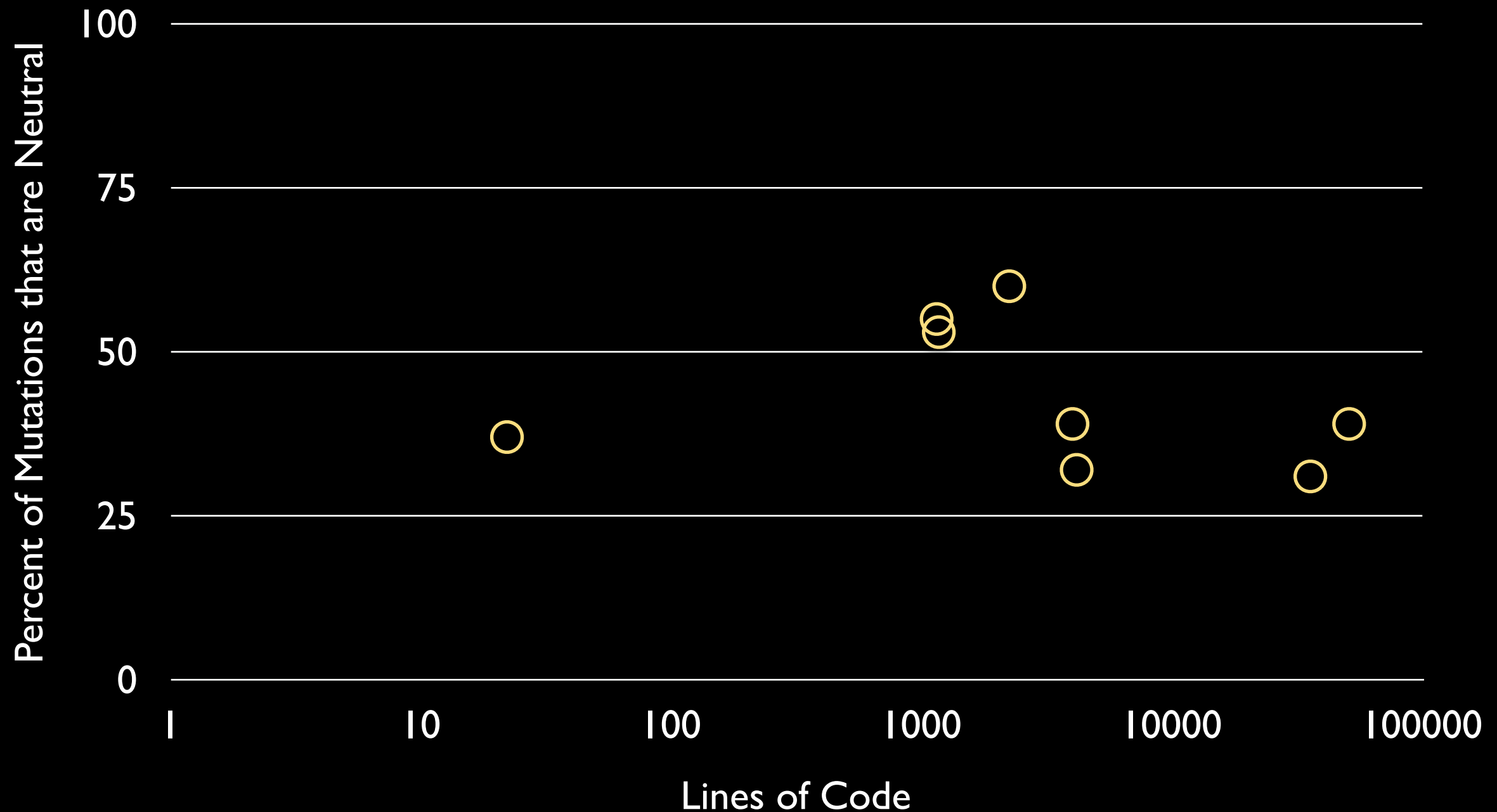
Shouldn't one look at programs more generally?

# With and Without Path





# Robustness vs. Code Size



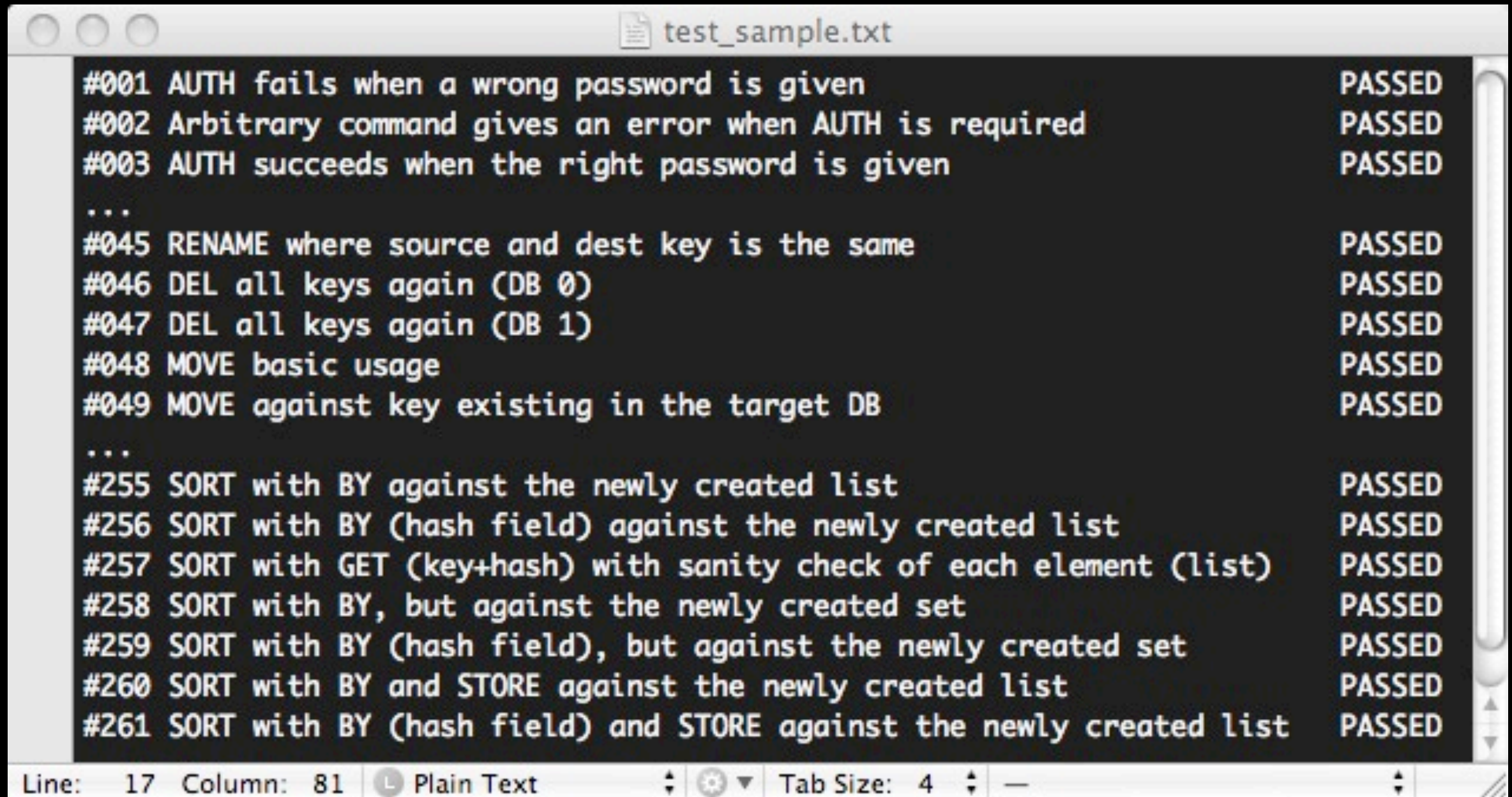
But perhaps my tests suites  
are simply quite terrible?

Do these results actually generalize?

# Neutral Mutations on Large Suites



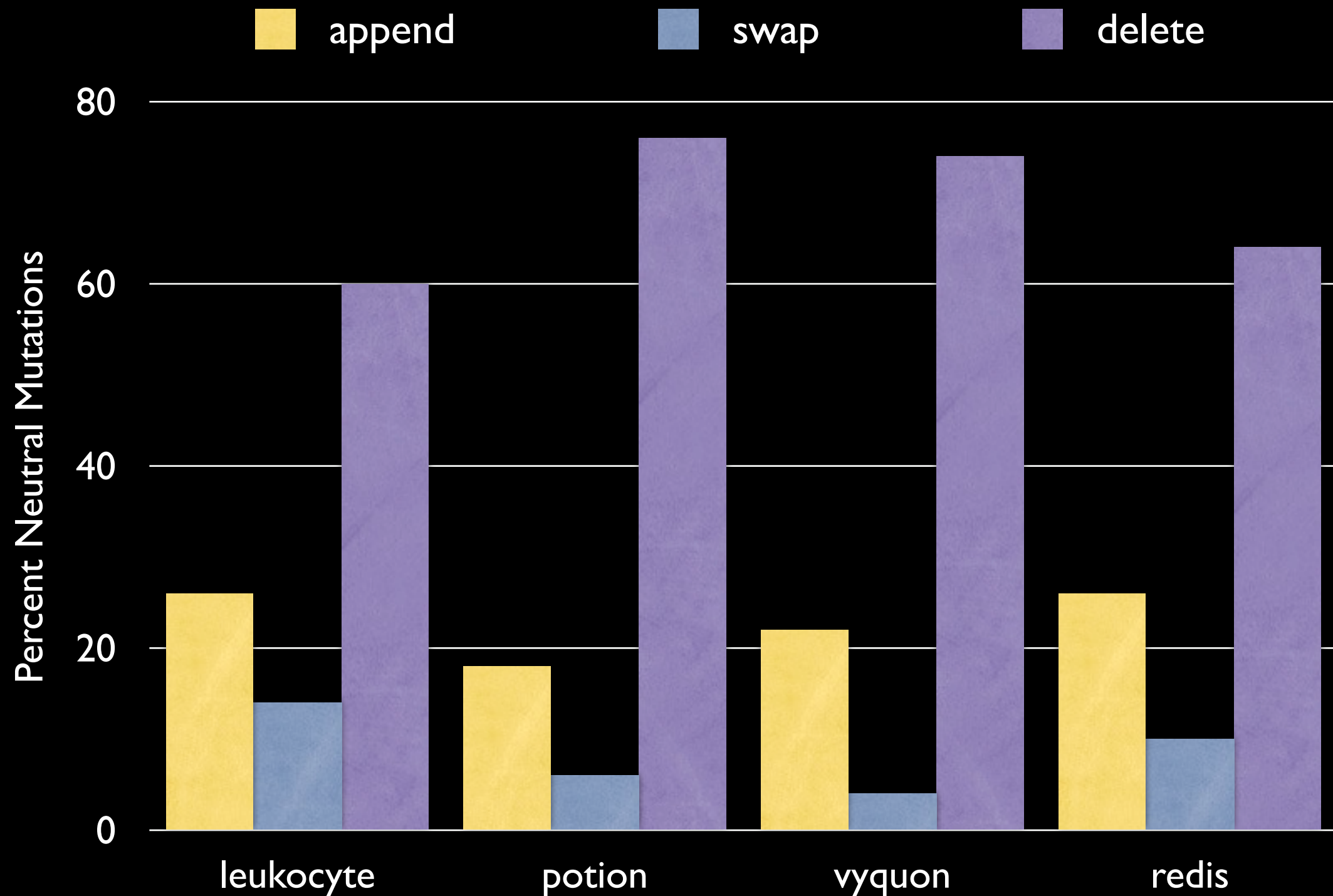
# A Non-Trivial Test Suite



```
#001 AUTH fails when a wrong password is given PASSED
#002 Arbitrary command gives an error when AUTH is required PASSED
#003 AUTH succeeds when the right password is given PASSED
...
#045 RENAME where source and dest key is the same PASSED
#046 DEL all keys again (DB 0) PASSED
#047 DEL all keys again (DB 1) PASSED
#048 MOVE basic usage PASSED
#049 MOVE against key existing in the target DB PASSED
...
#255 SORT with BY against the newly created list PASSED
#256 SORT with BY (hash field) against the newly created list PASSED
#257 SORT with GET (key+hash) with sanity check of each element (list) PASSED
#258 SORT with BY, but against the newly created set PASSED
#259 SORT with BY (hash field), but against the newly created set PASSED
#260 SORT with BY and STORE against the newly created list PASSED
#261 SORT with BY (hash field) and STORE against the newly created list PASSED
```

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# By Mutation Operators





# Stepping Back

Surprising to see such **high** levels of **mutational robustness**, at this level of representation

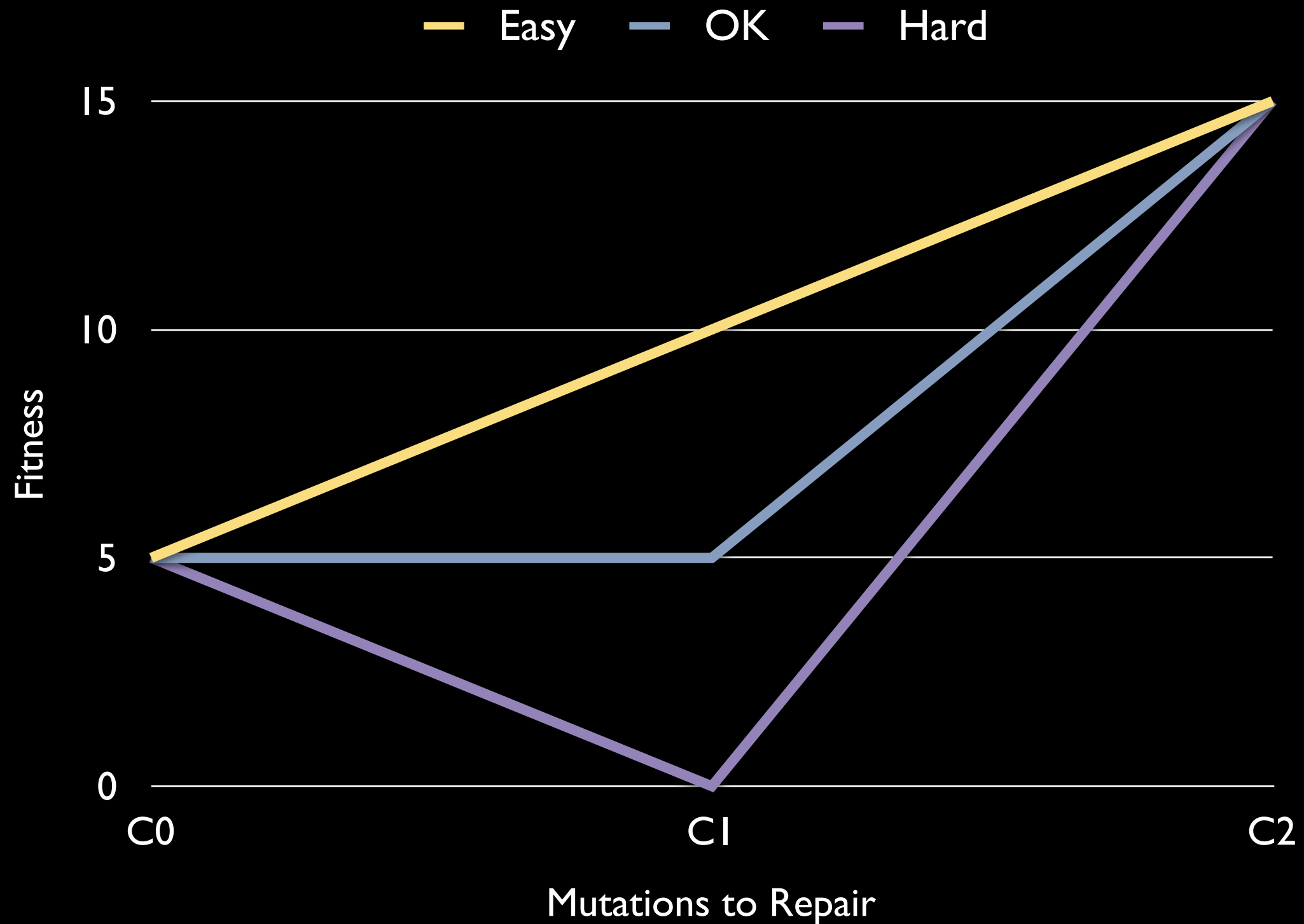
Possibly **contributes** \* to the success of Program Repair via GP

Quite counter-intuitive (so we assert)

\* robustness  $\neq$  good (tradeoff with **evolvability**)

# Relating Robustness to Repair Difficulty

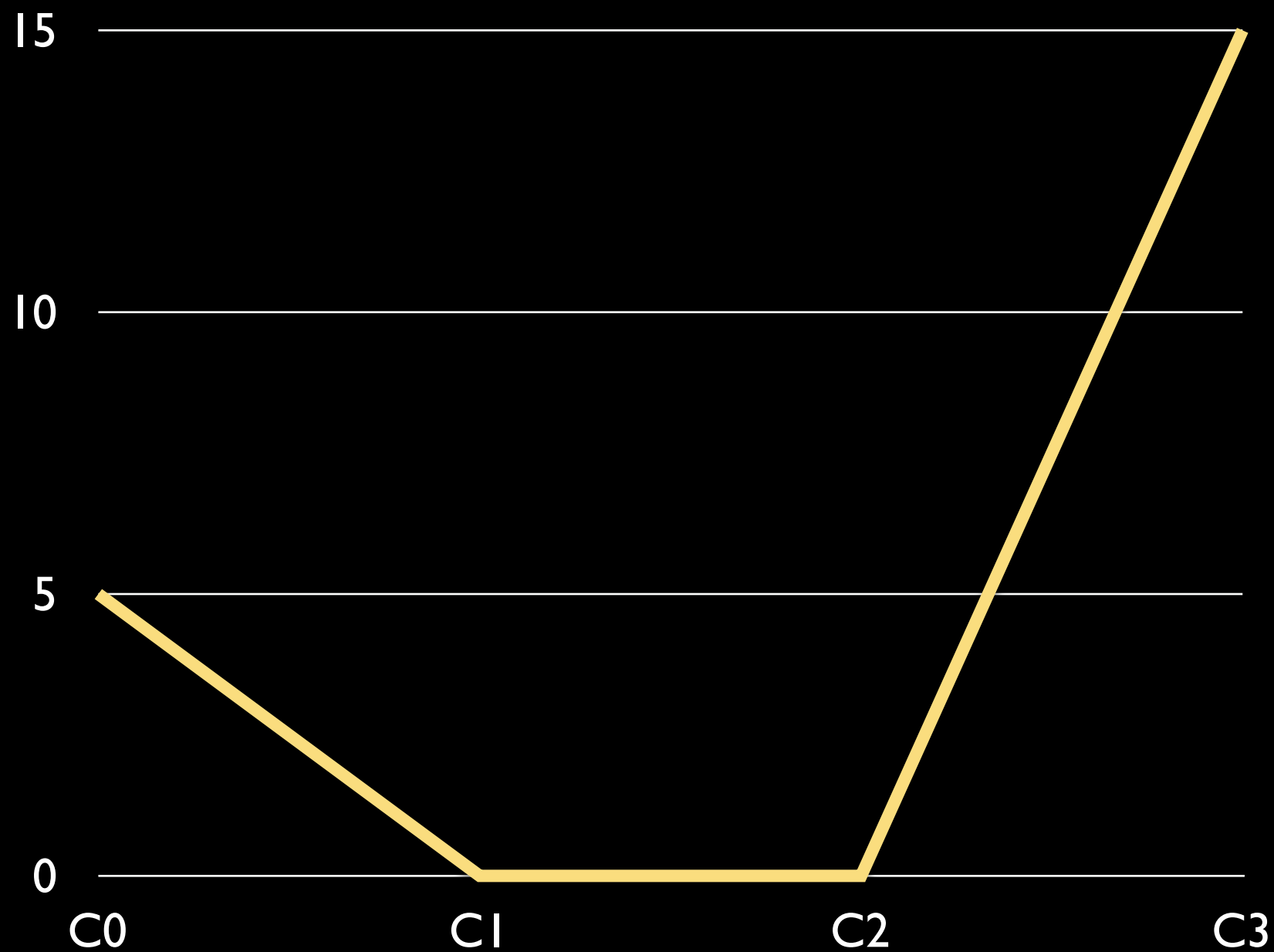
# A Problem?





# Three-Step Repair

A pathological case study



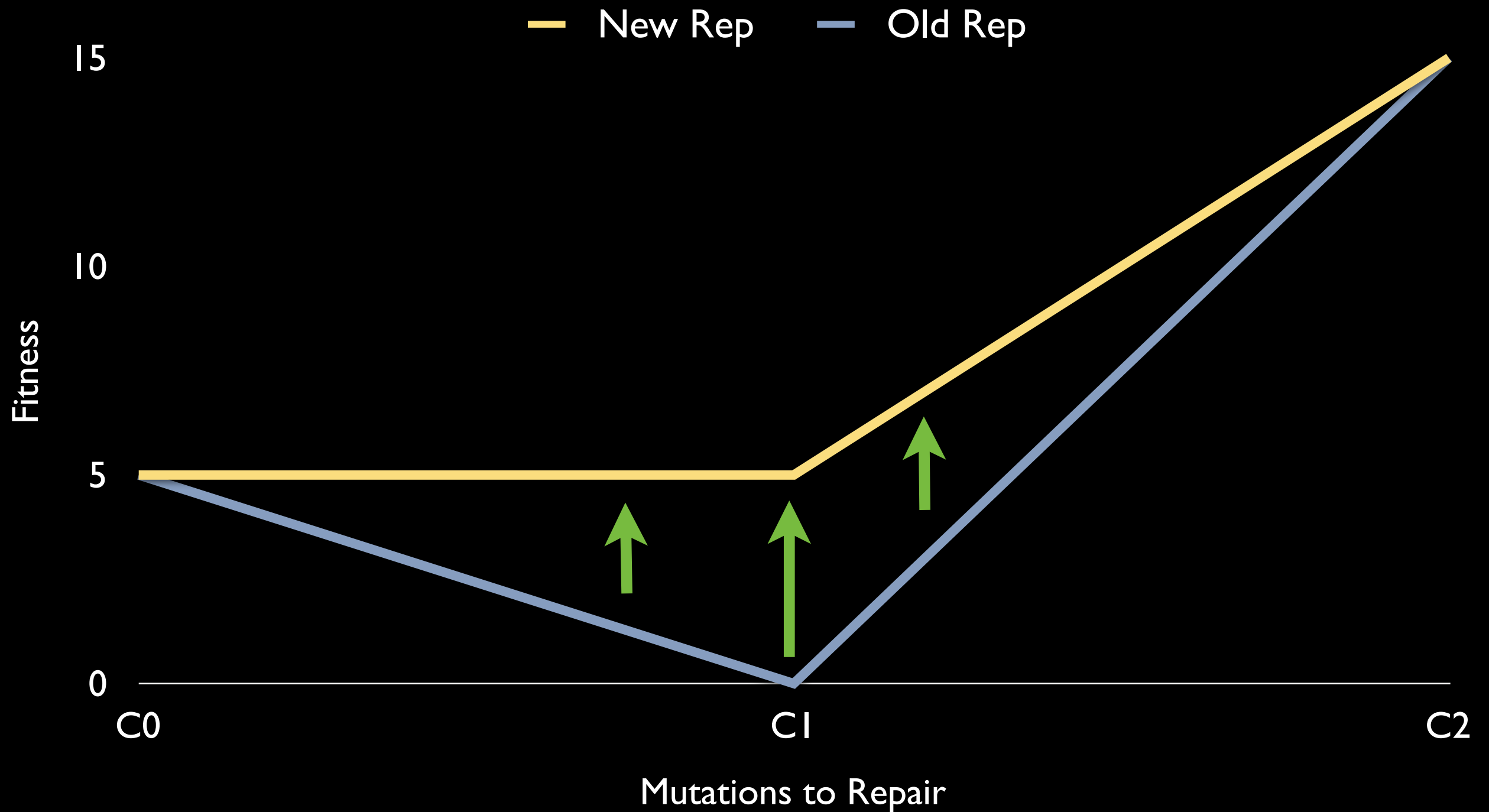
# How might we solve this?

Use a **new** representation, with a **higher** degree of **mutational robustness**

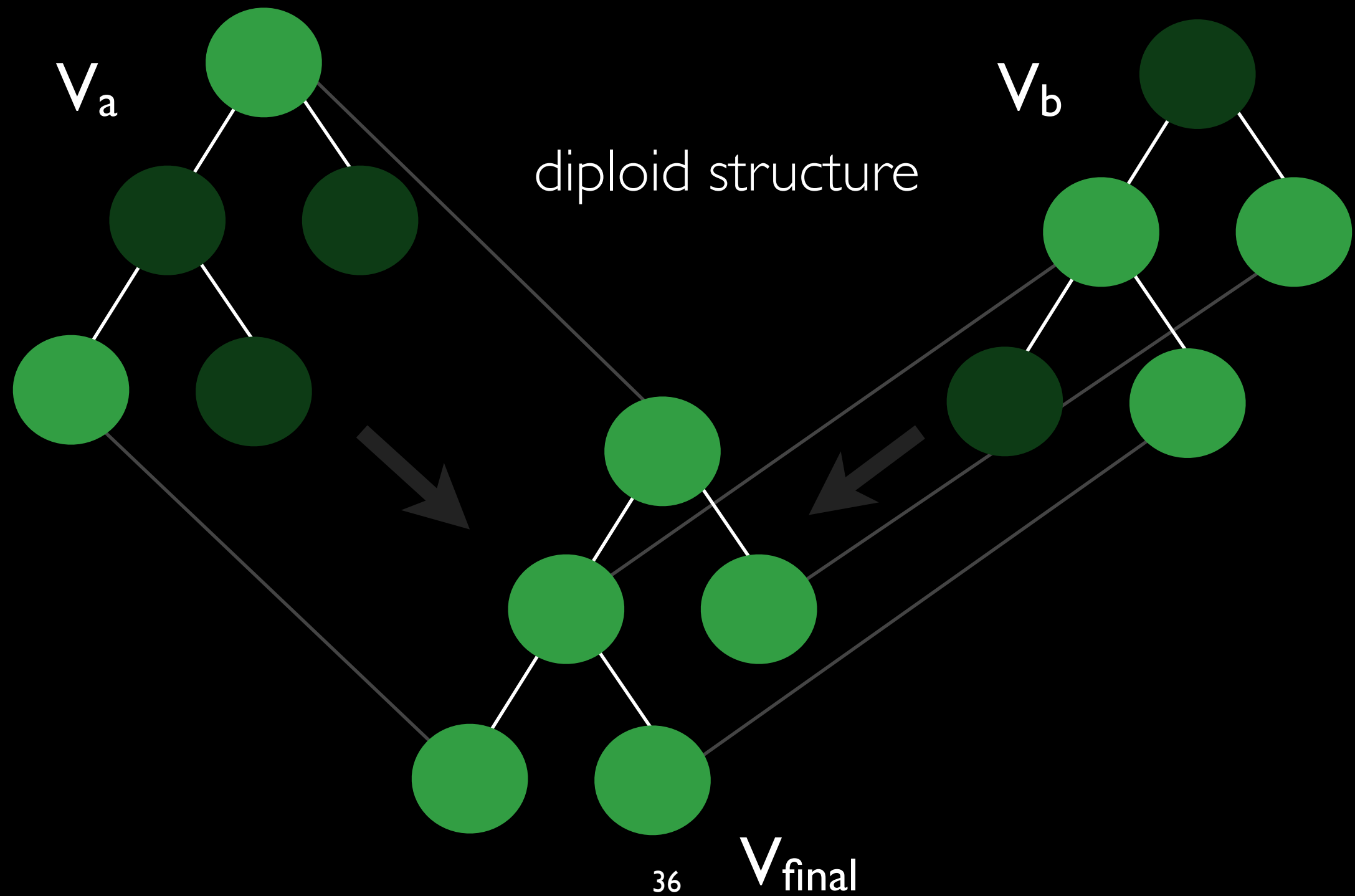
Inspiration: a **diploid** chromosomal structure

Change the **gradient** of the fitness landscape leading to repair

# The Basic Idea



# New Representation



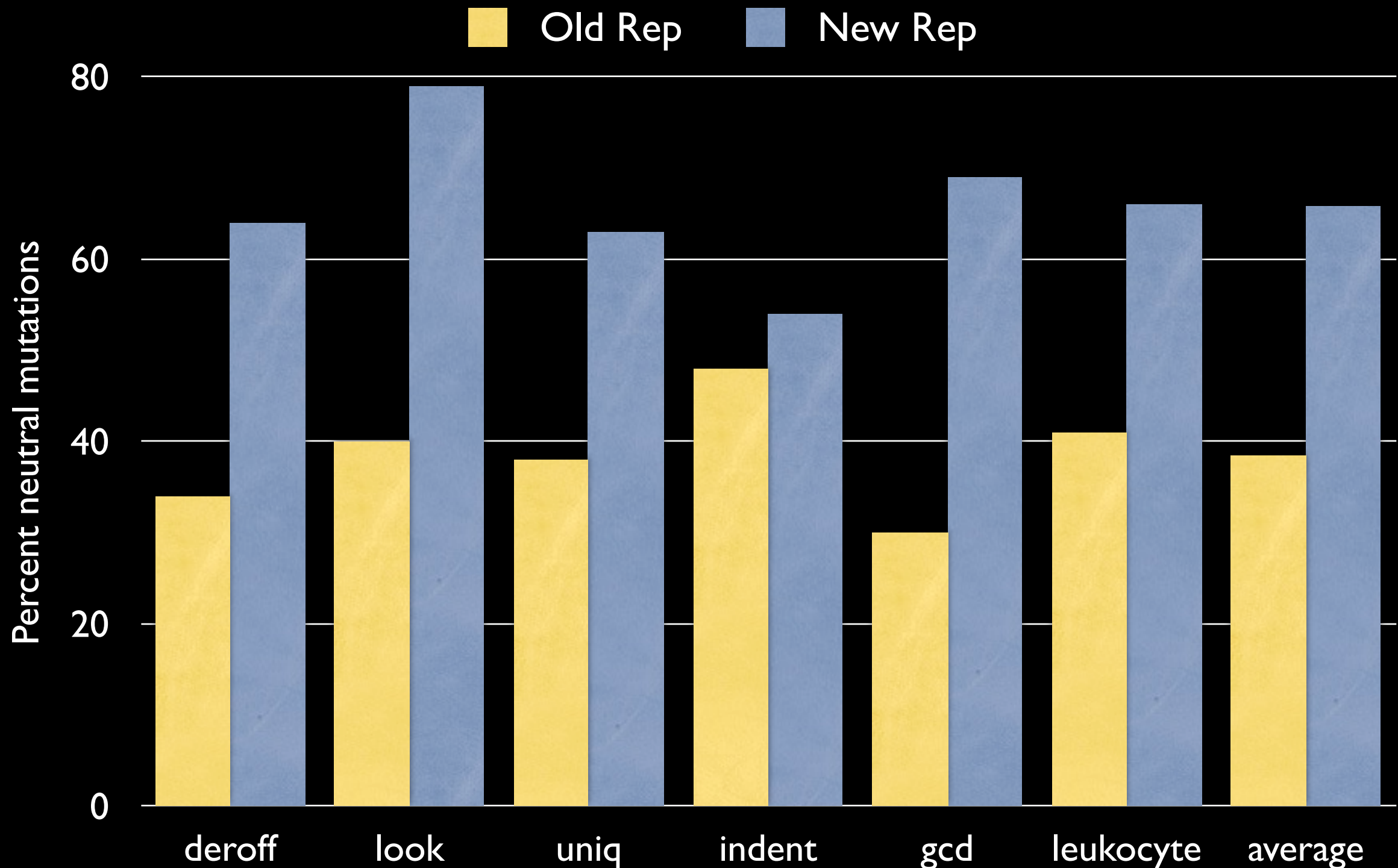
# Upshot

Mutations can be made to program segments that are **not applied functionally**

A **smoother** fitness gradient to repair

**Innovation:** Occasionally these non-functional mutations will be transformed into functional mutations

# New Rep More Robust?



# Preliminary Results

Of a Mixed Nature

Two-step repair found 3x as often

Three-step repair never found

Working on Additional Strategies

Different representations

Fitness function

# Conclusions

Programs are **surprisingly robust**

Result holds for **large and complicated** programs and **test suites**

But more **robust representations** may help in repairing certain kinds of bugs



# Questions?

Suggestions are also welcome

```
int main(int argv, char * argc[]){
    int x = atoi(argc[1]);
    int p1 = 0;
    int p2 = 0;
    int p3 = 0;
    //p1 = 7;
    //p2 = 3;
    //p3 = 4;
    int now = p1+p2+p3;
    if( x == 1 ){
        printf("%d:%d:%d\n",x,p1-p2-p3,now==p1+p2+p3);
    }
    if( x == 2 ){
        printf("%d:%d:%d\n",x,p1-p2-p3,now==p1+p2+p3);
    }
    if( x == 3 ){
        printf("%d:%d:%d\n",x,p1-p2-p3,now==p1+p2+p3);
    }
    if( x == 4 ){
        printf("%d:%d:%d\n",x,p1-p2-p3,now==p1+p2+p3);
    }
    if( x == 5 ){
        printf("%d:%d:%d\n",x,p1-p2-p3,now==p1+p2+p3);
    }
    if( x == 666) {
        printf("%d:%d:%d:%d\n",x,p1+p2+p3,p1-p2-p3,now==p1+p2+p3);
    }
    p1 = 7;
    p2 = 3;
    p3 = 4;
}
```

# Robustness Benchmark

Program	MR *	Neutral **
deroff	20%	34%
look	20%	40%
uniq	24%	38%
indent	16%	48%
gcd	23%	30%
leukocyte	19%	41%

\* measured average change in test case fitness

\*\* percent of mutations that do not affect fitness

# With Mutation Operators

Program	MR *	Neutral **	Append	Swap	Delete
deroff	20%	31%	30%	11%	59%
look	20%	43%	40%	14%	46%
uniq	24%	34%	43%	14%	43%
gcd	23%	34%	55%	19%	25%
leukocyte	19%	39%	12%	23%	64%

\* measured average change in test case fitness

\*\* percent of mutations that do not affect fitness

# With No Path Weights

Program	Neutral	Append	Swap	Delete
deroff	60%	28%	20%	52%
look	53%	34%	15%	51%
uniq	55%	27%	17%	56%
gcd	37%	61%	11%	28%
leukocyte	39%	32%	13%	56%

Even **more robust** to random mutations

# For Larger Test Suites?

Program	Neutral	Append	Swap	Delete
leukocyte	35%	26%	14%	60%
potion	39%	18%	6%	76%
vyquon	32%	22%	4%	74%
redis	31%	26%	10%	64%

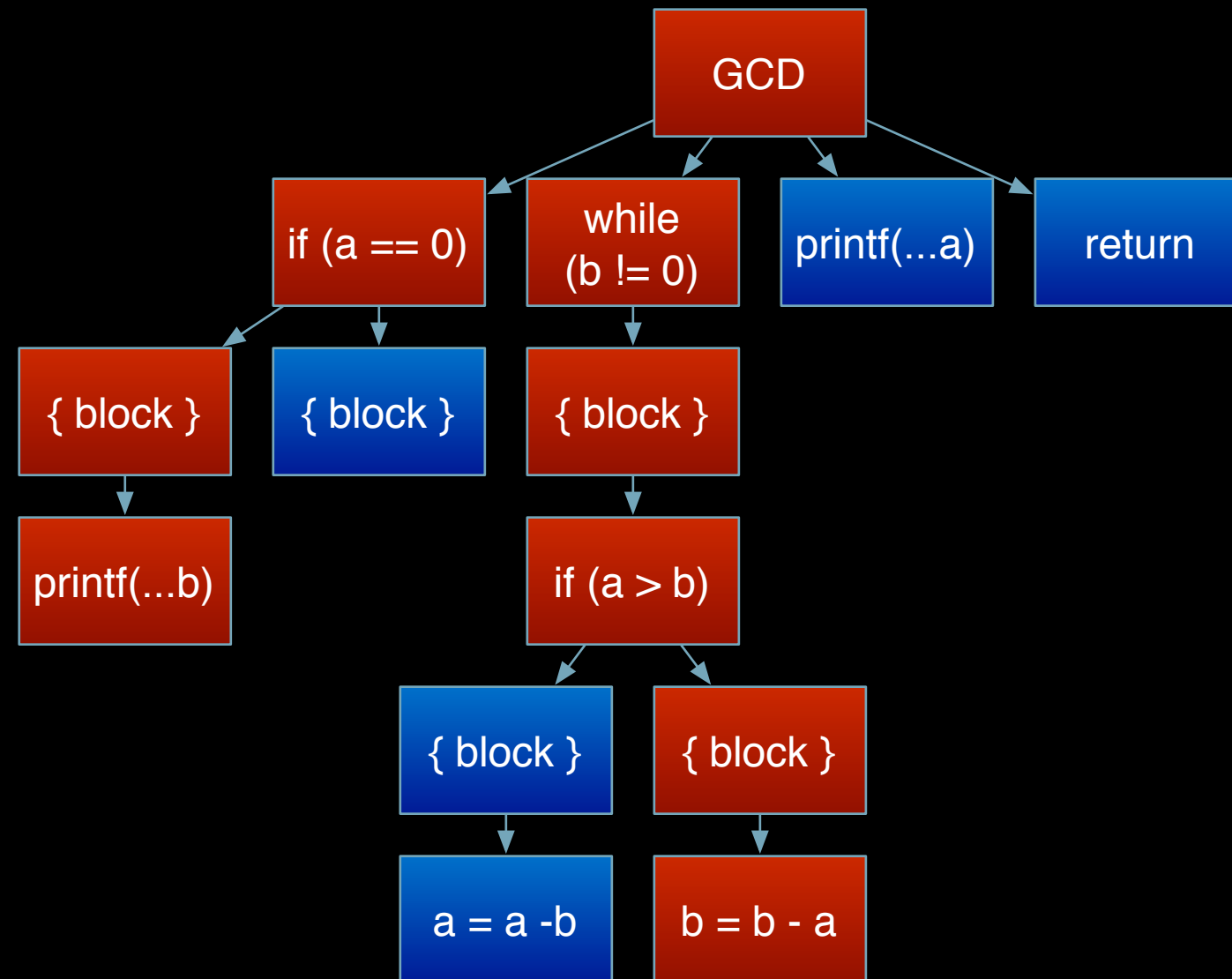
Seems not to be artifact of small test suites

# New Rep More Robust?

Program	Old Rep *	New Rep *
deroff	34%	64%
look	40%	79%
uniq	38%	63%
indent	48%	54%
gcd	30%	69%
leukocyte	41%	66%
Average	38.5%	65.8%

\* percent of mutations that do not affect fitness

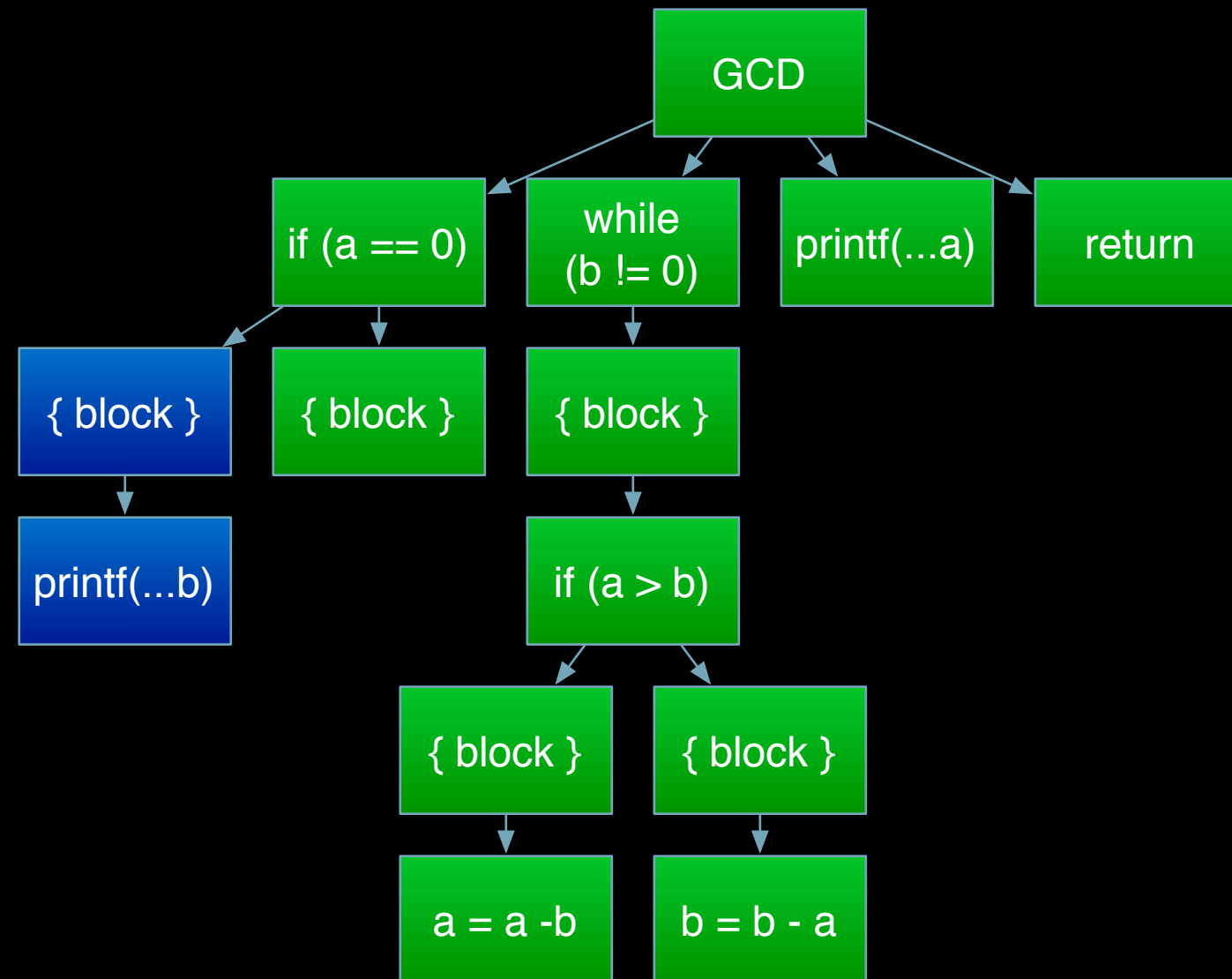
# Weighted Path



Negative Test Case

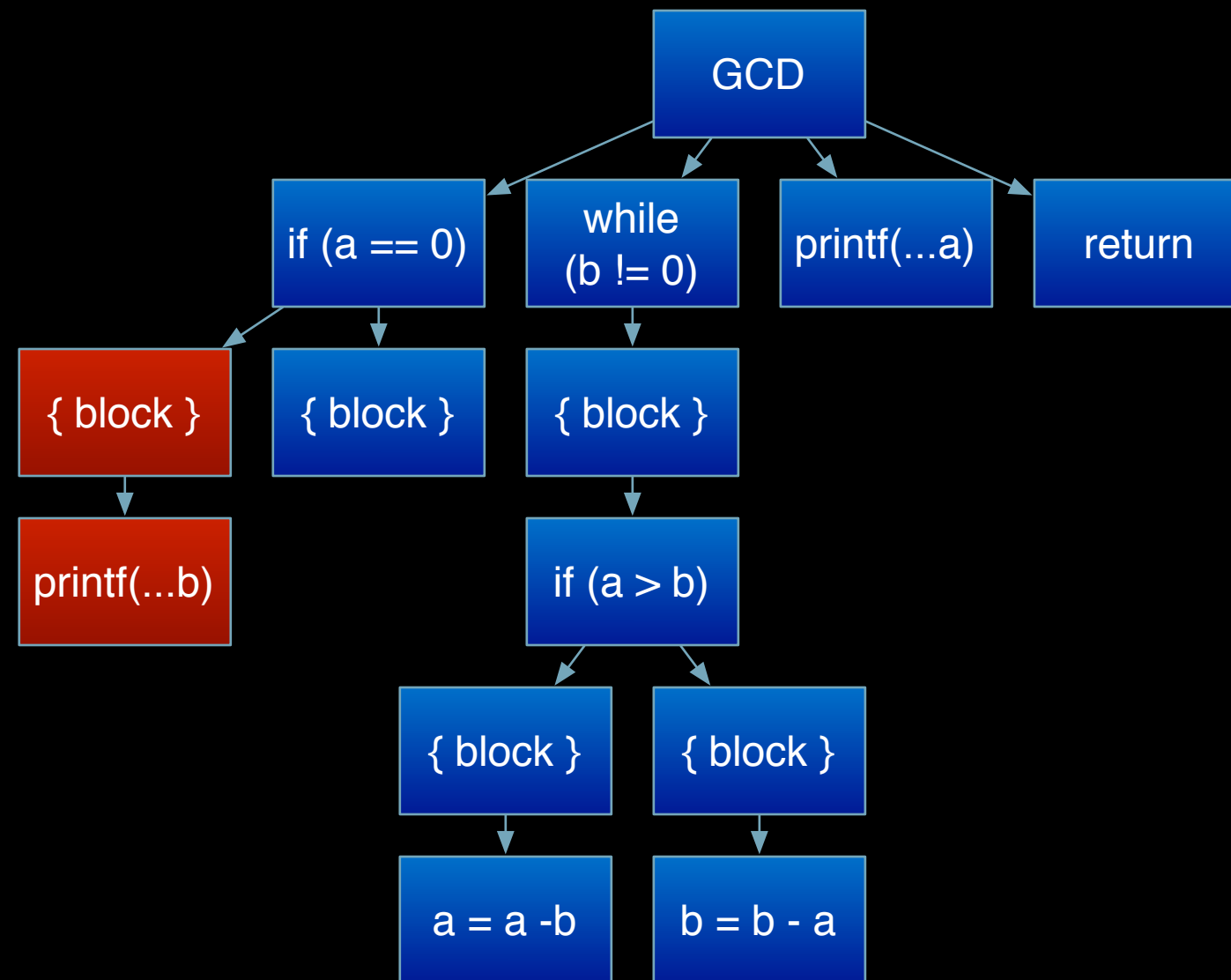


# Weighted Path



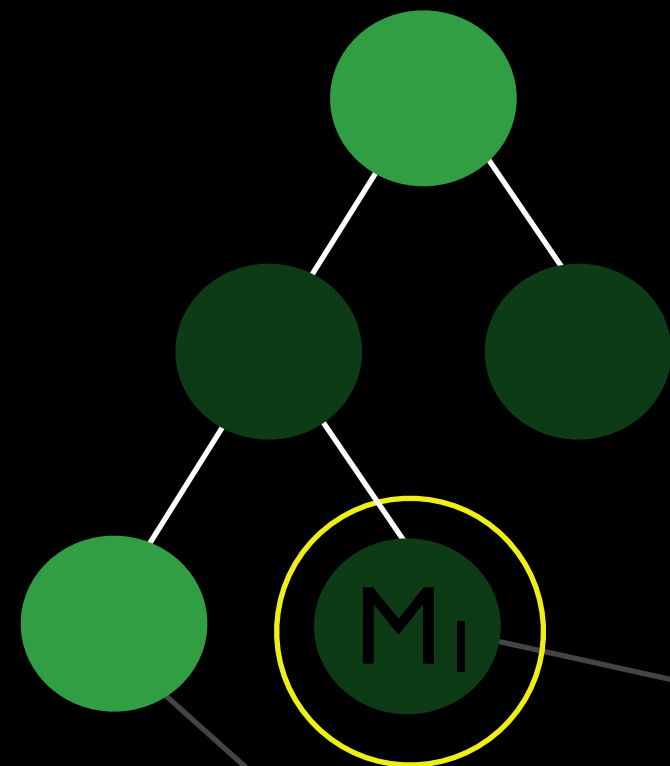
Positive Test Case

# Weighted Path

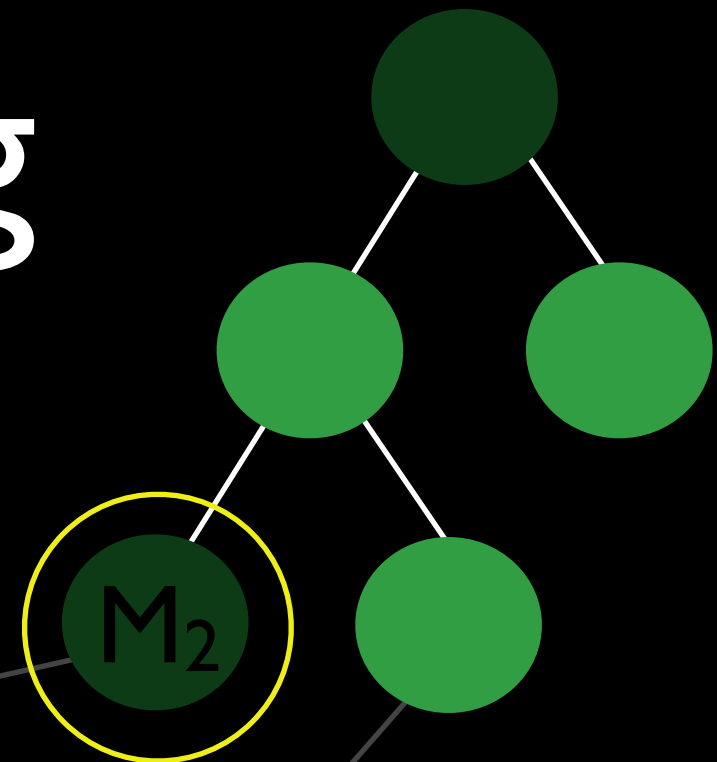


Final Path

# Swapping



$$F(M_1 \text{ or } M_2) = 0$$

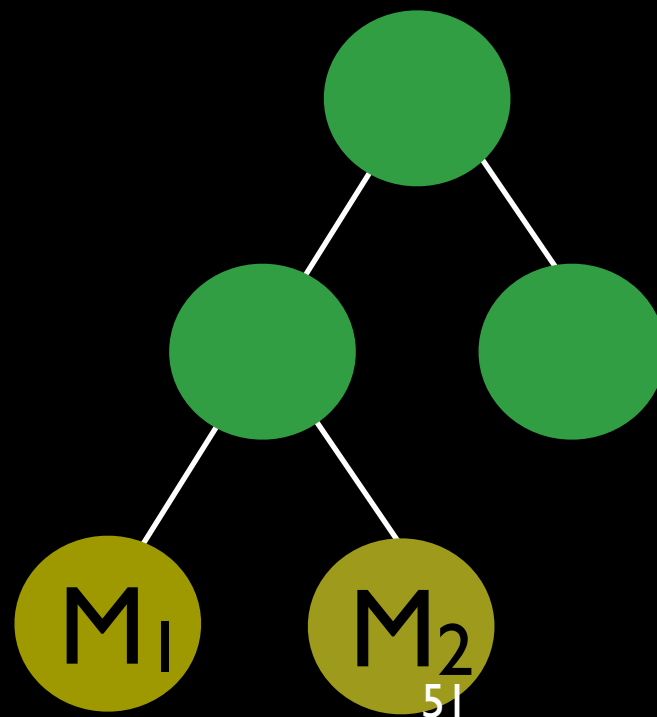


swap 2

swap 1

$$F_a = 5$$

$$F_b = 5$$



$$F_{\text{final}} = 15$$