CSE/ECE 848 Introduction to Evolutionary Computation

Module 5 - Lecture 23 - Part 1b (formerly 3)

Genetic Improvement

Wolfgang Banzhaf, CSE
John R. Koza Chair in Genetic Programming



- Genetic Improvement (GI)
- ARJA: GI for Repairing Bugs in Java

Genetic Improvement - The Idea

- Genetic Improvement (GI) is related to the original goal of GP (i.e., automatic programming), rather than symbolic regression.
- GP generally cannot start from scratch to create a real-world program
- GI offers a different slant: why not take human-written program as a starting point and improve it

Genetic Improvement - Types

- There are two types of improvements
 - Functional properties: from a buggy program to a correct program (bug repair)
 - Non-Functional properties: e.g., from a slow program to a faster program (software optimization)
- GI = Software improvement through GP

Original software



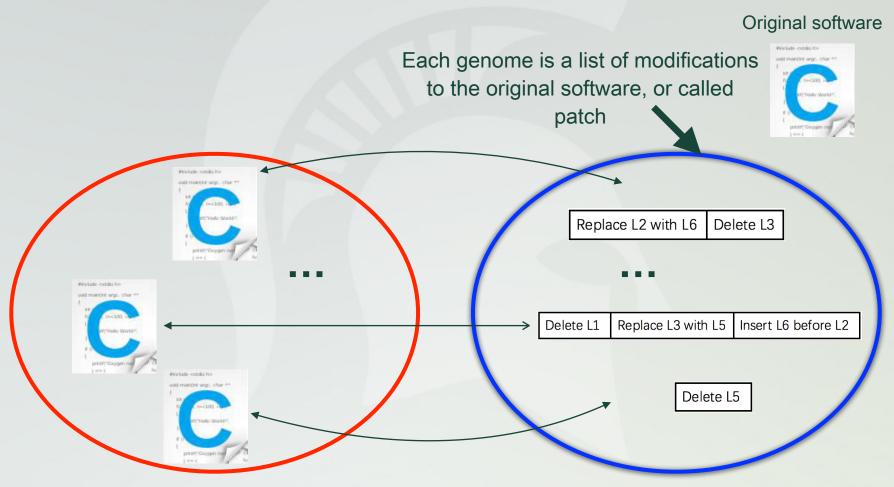
Here each solution is really a (large-scale) software, but it is very similar to the original software



Solution space

Encoded space

Genetic Improvement - Patches



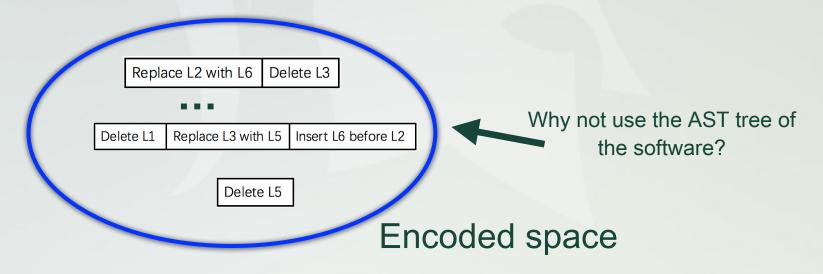
Solution space

CSE/ECE 848 Introduction to **Evolutionary Computation**

Encoded space

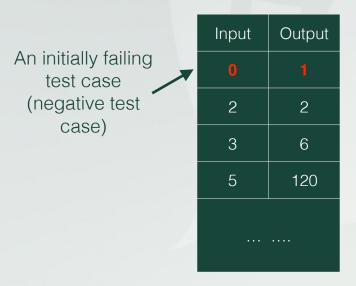
Genetic Improvement - How?

- AST tree structure does not scale to large software
- A population of AST trees for large software cannot fit in computer memory
- The trees are normally very similar to the tree of the original software
- Representing each genome as a list of modifications can avoid storing redundant copies of untouched code



Genetic Improvement - Bug Repair

 GI for bug repair: given a buggy program and a set of test cases (with at least one test case failing), conduct a biased GP search to improve the buggy software, in order to make all the test cases passing.



```
int Factorial(int a) {
   if (a <= 0)
      return 0;
   else
      return (a * Factorial(a - 1));
}</pre>
```

A buggy program

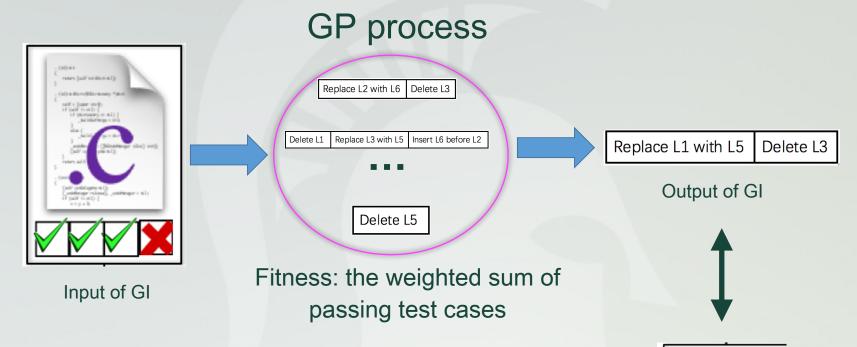
A set of test cases



- Fitness function (GI for bug repair)
 - Give a genome/patch (i.e., a list of modifications), how to compute the its fitness
 - Fitness: the weighted sum of the number of test cases that are passed



Run it on the given test cases to see how many test cases it passes



Overall procedure of GI for bug repair A repaired software





- Search space of patches in GI for bug repair
 - Where (e.g., which lines of code) we want to modify?



A buggy software with test cases

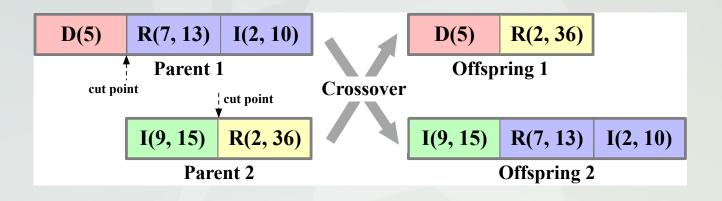
may change

The lines we

- Search space of patches in GI for bug repair
 - Which kind of modifications (of a line) we want to conduct?
 - A. Delete a line
 - B. Replace a line with another code
 - C. Inert another code before a line
 - Another code?: assume that it just comes from the code in the current buggy software
 - Redundancy assumption: we can usually fix a bug only using the existing code in the buggy software

Le Goues, C., Nguyen, T., Forrest, S., & Weimer, W. (2011). GenProg: A generic method for automatic software repair. *IEEE Transactions on Software Engineering*, 38(1), 54-72.

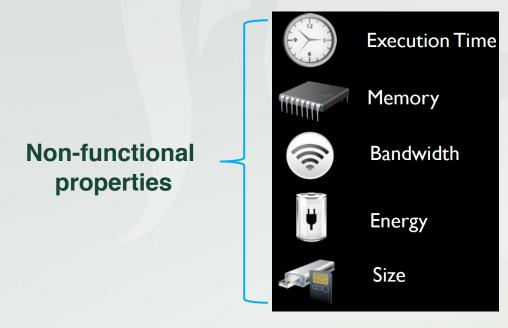
- Genetic Operators (GI for bug repair)
 - Crossover and mutation





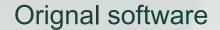
Genetic Improvement - Optimization

 GI for Software Optimization: improve non-functional properties (e.g., running time, memory consumption, power/energy consumption) of a software



Delete L3

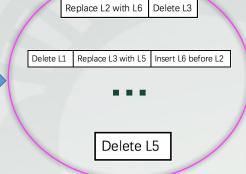
Genetic Improvement





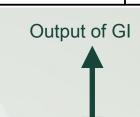
Input of GI

GP process



Fitness?

Improved software

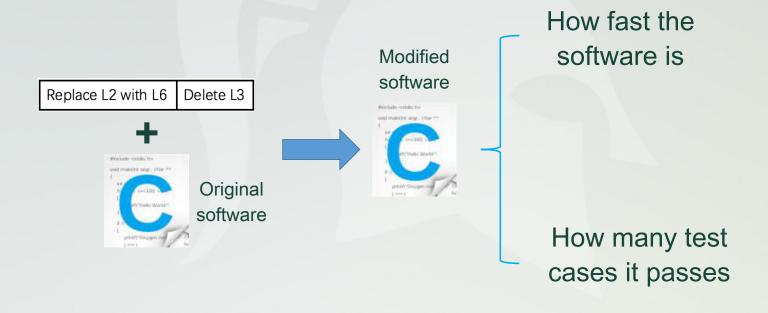


Replace L1 with L5

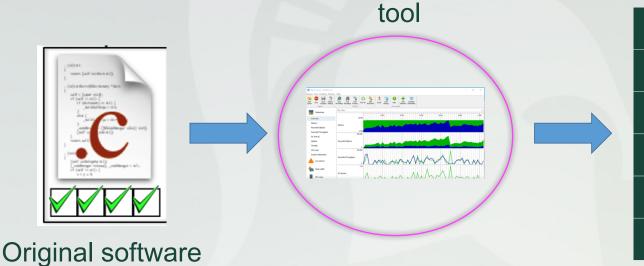


Overall procedure of GI for software optimization

GI for software optimization: fitness function





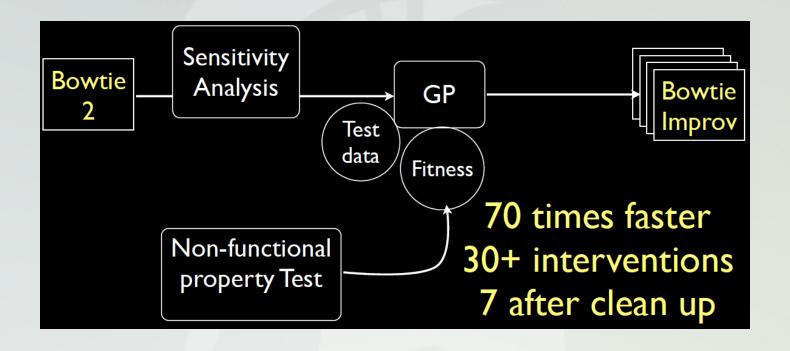


Performance profiling

Bottleneck lines L12 L20 L87 L201

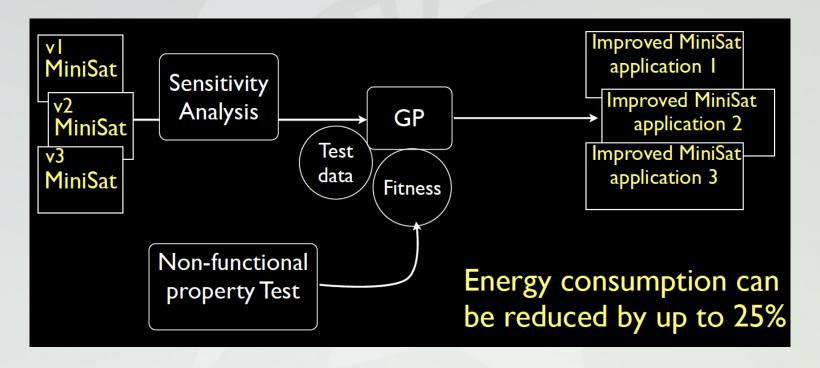
The lines we may change

Genetic Improvement - Application



William B. Langdon and Mark Harman, Optimizing Existing Programs with Genetic Programming, IEEE

Transactions on Evolutionary Computation, 19 (1), 2015.



Bobby R. Bruce, Justyna Petke, Mark Harman, and Earl T. Barr, Approximate Oracles and Synergy in Software Energy Search Spaces, IEEE Transactions on Software Engineering, in press

Genetic Improvement - A Problem

- Overfitting in GP/GI
 - The output program/software may be incorrect beyond passing the given test cases
 - The underlying reason is that the test suite is usually an incomplete program specification

Input	Output
0	1
2	2
3	6
5	120
6	720

Test cases

```
int Factorial(int a)
{
   if (a <= 0)
      return 1;
   else
      return (a * Factorial(a-1));
}</pre>
```

Correct program (very likely)

CSE/ECE 848 Introduction to Evolutionary Computation

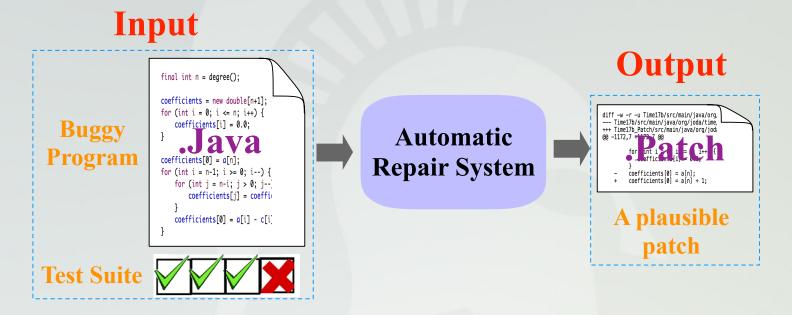
```
int Factorial(int a) {
    switch (a) {
    case 0: return 1;
    case 2: return 2;
    case 3: return 6;
    case 5: return 120;
    case 6: return 720;
    default: return 0;
    }
}
```

Overfitting!



- Overfitting in GP/GI
 - This is a very tricky problem
 - Usually, we use GP/GI to output a number of programs, and then use certain kind of heuristic to rank the output programs (ranked higher, more likely to be correct).

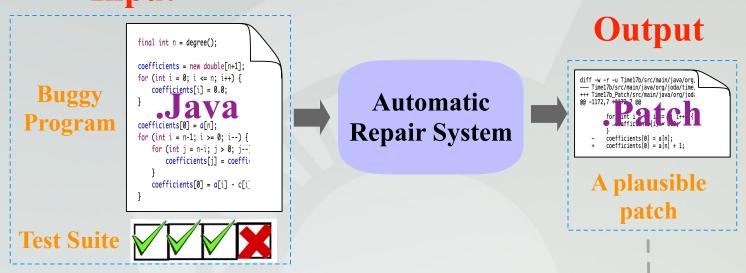
Problem Statement (Test-Suite Based Bug Repair)



Note:

- 1. Test suite should contain a number of initially passing (i.e., positive) test cases and at least one initially failing (i.e., negative) test case
- 2. Sometimes the number of plausible patches obtained can be more than one.

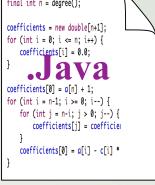
Problem Statement Input



 A plausible patch: a patch that can make the entire test suite pass.

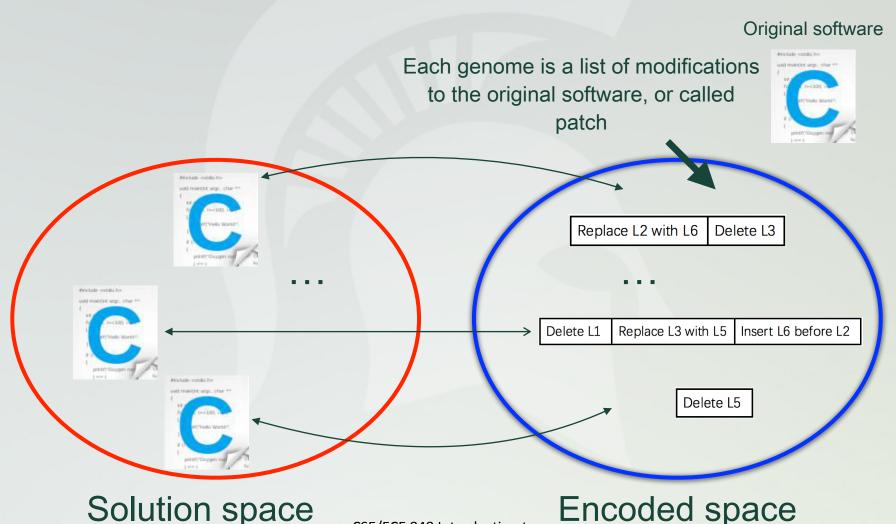
Note: A plausible patch is a patch that can make the modified program pass the entire test suite.

Modified Program



Test Suite





Module 5 Lecture 23 Part 1b

CSE/ECE 848 Introduction to **Evolutionary Computation**

Encoded space

- Components in a GI System for Bug Repair
 - Search space (which patches we want to consider)
 - Search algorithm
 - Fitness function (how to define the goodness of a patch)
 - Find best patches using GP (over a genetic representation)
 - Handling the overfitting issue

A detailed introduction can be found in our papers:

[1] Yuan Yuan, Wolfgang Banzhaf, ARJA: Automated repair of Java programs via multi-objective genetic programming. *IEEE Transactions on Software Engineering, Vol 46, 1040—1067 (2020)* [This paper introduces the original ARJA system]

[2] Yuan Yuan, Wolfgang Banzhaf. Towards better evolutionary program repair: An integrated approach. *ACM Transactions on Software Engineering and Methodology, Vol 29, 5:1—5:53 (2020)* [This paper introduces an enhanced ARJA version called ARJA-e]