Genetic improvement: Taking real-world source code and improving it using genetic programming

Sæmundur Ó. Haraldsson, Alexander Brownlee

John R. Woodward,

Markus Wagner Bradley Alexander







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Research Cou DE160100850 EP/S005803/1

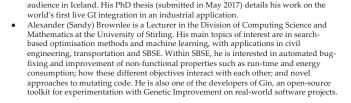
GECCO 2021

Instructors

UNIVERSITY of STIRLING







Saemundur O. Haraldsson is a Lecturer at the University of Stirling. He co-organised every version of this tutorial. He has multiple publications on Genetic Improvement, including two that have received best paper awards. Additionally, he co-authored the

first comprehensive survey on GI 1 which was published in 2017. He has been invited to give talks on the subject in two Crest Open Workshops and for an industrial



Instructors



• John R. Woodward is a lecturer at the Queen Mary University of London. Formerly he was a lecturer at the University of Stirling, and was employed on the DAASE project (http://daase.cs.ucl.ac.uk/). Before that he was a lecturer for four years at the University of Nottingham. He holds a BSc in Theoretical Physics, an MSc in Cognitive Science and a PhD in Computer Science, all from the University of Birmingham. His research interests include Automated Software Engineering, particularly Search Based Software Engineering, Artificial Intelligence/Machine Learning and in particular Genetic Programming. He has over 50 publications in Computer Science, Operations Research and Engineering which include both theoretical and empirical contributions, and given over 50 talks at International Conferences and as an invited speaker at Universities. He has worked in industrial, military, educational and academic settings, and been employed by EDS, CERN and RAF and three UK Universities.





Engineering and Physical Sciences Research Council Exploiting Defect Prediction for Automatic Software Repair (Fixie), grant EP/S005803/1

Instructors



- Markus Wagner is a Senior Lecturer at the School of Computer Science, University of Adelaide, Australia. He has done his PhD studies at the Max Planck Institute for Informatics in Saarbruecken, Germany and at the University of Adelaide, Australia. For the outcomes of his studies, he has received the university's Doctoral Research Medal the first for his school and three best paper awards. His research topics range from mathematical runtime analysis of heuristic optimisation algorithms and theory-guided algorithm design to applications of heuristic methods to renewable energy production, professional team cycling and software engineering. So far, he has been a program committee member 60+ times, and he has written 150+ articles with 150+ different co-authors. He is on SIGEVO's Executive Board and serves as the first ever Sustainability Officer. He has contributed to GECCOs as Workshop Chair and Competition Chair, and he has chaired several education-related committees within the IEEE CIS.
- Bradley (Brad) Alexander is a Senior Lecturer and Director of Teaching at the School of Computer
 Science, University of Adelaide, Australia. Brad's research interests include program optimisation,
 rewriting, genetic-programming (GP) especially the discovery of recurrences and search-basedsoftware-engineering. He has also supervised successful projects in the evolution of control algorithms
 for robots, the evolution of three-dimensional geological models, and the synthesis of artificial water
 distribution networks, and using background optimisation to improve the performance of instruction
 set simulators (ISS)'s. He has also worked on improving algorithms for the analysis of water distribution





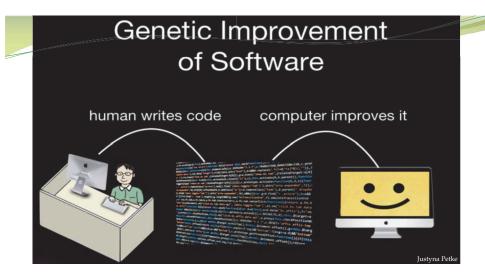
Australian Government

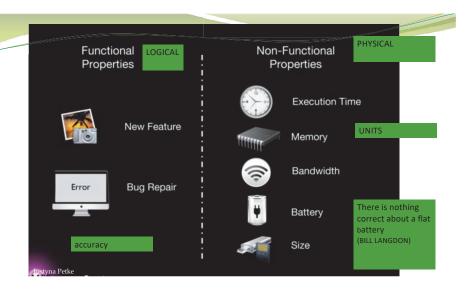
Australian Research Council

DE160100850. http://purl.org/auresearch/grants/arc/DE160100850

Overview

- Introduction
- Fixing Bugs and other examples
- Noteworthy papers and issues
- Getting involved
- Summary and Q&A





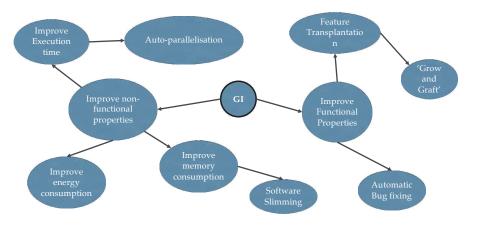
What is Genetic Improvement

A wordy definition:

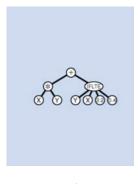
Genetic Improvement is the application of search-based (typically evolutionary) techniques to modify software with respect to some user-defined fitness measure.

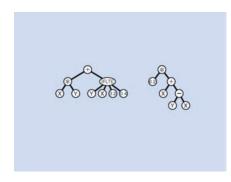


WHAT IS GENETIC IMPROVEMENT



GENETIC PROGRAMMING OVERVIEW





mutation

crossover

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Genetic Programming: GI's ROOTS

- **1. Aim** *to discover new programs by telling the computer what we want it to do, but not how we want it to do it* John Koza
- **2. How** we evolve computer programs using natural selection.
- **3. Starts** from scratch (empty program)
- 4. Choose **primitives** (terminal set/FEATURES and function set)
- 5. Choose **representation** (tree based, graph based, linear e.g. CGP)
- **6.** Choose fitness function, parameters, genetic operators.

GI forces "the full capabilities of programming languages"- side effects, ADFs, LOOPS

GP vs GI: if you can't beat them, join them.

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ABSTRACT

Genetic Programming (GP) has been criticized for targeting irrelevant problems [12], and is true of the wider machine (procedures, methods, macros, routines), and so GI has to deal with the reality of existing software systems. However, most of the GP literature is not concerned with Tur-

Popular Science

• easy to digest articles for non-specialists.

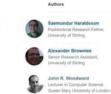
https://theconversation.com/computers-will-soon-be-able-to-fix-themselves-are-it-departments-for-the-chop-85632

Computers will soon be able to fix themselves – are IT departments for the chop?



October 12, 2017 3.29pm BST





https://theconversation.com/how-computersare-learning-to-make-human-software-workmore-efficiently-43798

How computers are learning to make human software work more efficiently

June 25, 2015 10.08am BST



http://www.davidrwhite.co.uk/2014/11/27/ge netic-programming-has-gone-backwards/



Genetic Programming has gone Backwards

When Genetic Programming (GP) first arose in the late 80s and early 90s, there was one very defining characteristic of its application, which was so widely accepted as to be left unsaid:

GP always starts from scratch

http://www.davidrwhite.co.uk/tag/ genetic-programming/





Competent Programmers Hypothesis

- 1. programmers write programs that are <u>almost</u> perfect.
- 1. program faults are syntactically small (slip of finger, T/F)
- 1. corrected with a few keystrokes. (e.g. < for <=)
- 1. GI can find small patches.
- 1. Small changes are non-unique (write 7 lines code, or utter 7 words before they're unique)

Plastic Surgery Hypothesis.

the content of new code can often be assembled out of fragments of code that already exist.

Barr et al. [71] showed that changes are 43% graftable from the exact version of the software being changed.

The Plastic Surgery Hypothesis: Changes to a codebase contain snippets that already exist in the codebase at the time of the change, and these snippets can be efficiently found and exploited.

THE CODE CONTAINS SOLUTIONS – CANDIDATE PATCHES

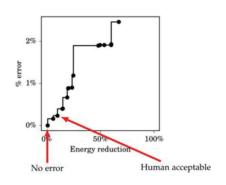
Representations of PROGRAMS

Natural Representation of CODE

- 1. Text files e.g. Program.java is a text file. Saemi.
- 2. Abstract syntax tree (AST) Genprog, Genofix.
- 3. Java byte code (also C binaries) [102]
- 1. Errors, compile, halting (Langdon discard)

Multi-Objective

- Seems be convex
- – simple argument (see pic)
- Can provide a set of programs
- weighted sum of objectives?
- •weight has meaning to user.
- Will there be elbow/knee points?



Objectives

- Functional (**logical properties**)
 - Accuracy e.g. as in machine learning FLOAT
 - Number of bugs as measured against a set of test cases. BOOLEAN
 - New functionality e.g.
- Non-functional (*physical* properties)
 - Execution time
 - Energy (power consumption peak/average)
 - Memory
 - Bandwidth
- Multi-objective
 - Trade-offs, convex, a set of programs = a single tuneable program

Slow connections.



Loading Gmail



Loading standard view | Load basic HTML (for slow connections)

GISMOE

The GISMOE challenge:

to create an automated program development environment in which the Pareto program surface is automatically constructed to support dialog with and decision making by the software designer concerning the trade offs present it the solution space of programs for a specific programming problem.

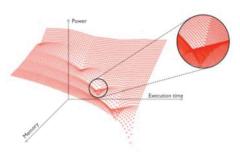


Figure 1: The GISMOE Pareto Program Surface

EDIT Operators – changes to programs

- Line level
- •Single Character level
- Function/module level.
- AST GIN, Gen-0-fix, genprog,
- Java machine code java byte code.
- •LIST OF EDITS IS A PATCH.

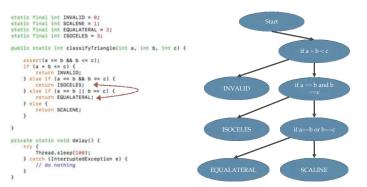
CI: AN EXAMPLE OF EXECUTION TIME OPTIMISATION

```
static final int INVALID = 0;
static final int SCALENE = 1;
static final int SCALENE = 2;
static final int EDALATERAR = 2;
static final int EDALATERAR = 2;
static final int INVALID = 0;
delay();

assert(a <= b &b b c c);
if (a + b c c) f
return INVALID = 0;
} else int (contact final b = c) f
return INVALID = 0;
return INVALID = 0;
return SCALENE;
}

private static void delay() f
try f
Thread.sleep(180);
cathor (InterruptedException e) f
// do nothing
}
```

GI: AN EXAMPLE OF AUTOMATED BUG FIXING



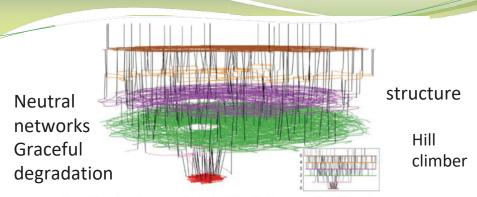
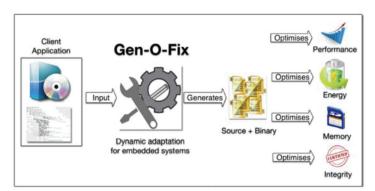


Fig. 1. Local optima network of the Triangle Program using 100 random starts (see Section 4.4). Edges are coloured if they start and end at the same fitness. Insert shows fitness levels edge on. Best (bottom) red 0 (pass all tests), pink 1 (fail only one test), green 2, purple 3, orange 4, brown 5.

System Diagram for Gen-O-Fix



Gen-O-Fix: Abstract Syntax Trees

Main features of framework are

- 1. **Embedded** adaptively.
- 2. Minimal end-user requirements.
 - 1. Initial source code: **location** of Scala source code file containing a function
 - 2. Fitness function: providing a means of **evaluating the quality** of system
- 3. Source to source transformations
- 4. Operates on ASTs (i.e. arbitrarily fine).

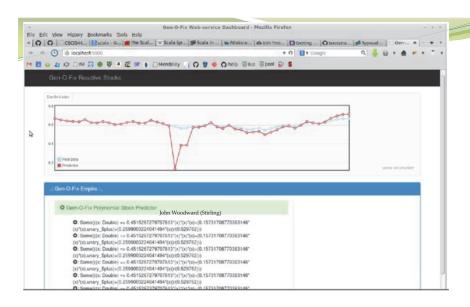
AST - scala

Code as data, data as code.

```
// code to data:
var m = 2; var x = 3; var c = 4
val expr = reify( ( m * x ) + c )
println( "AST = " + showRaw( expr. tree ) )

// output:
AST = Apply(Select(Apply(Select(Select(Ident("m"), "elem"), "$times"), List(Select(Ident("x")), "elem"))), "$plus"), List(Select(Ident("c"), "elem")))
```

```
// run AST datatype as code:
println( "eval = " + expr.tree.eval() )
// output:
eval = 10
```



Gt Hashcode tuning

- **1. Hadoop** provides a mapReduce implementation in Java.
- 2. Equals method has to obey **contract** (Reflective, Symmetric, Transitive, ...)
- 3. x.equals(y) **implies** hashCode(x)== hashCode(y).
- 4. hashCode method is an integer function of a subset of an object's fields

Some GP Settings

- **1. Terminal set** is
 - 1. Field values
 - 2. Random integers [0, 100]
- 2. Function set is
 - 1. {+, *, XOR, AND}
- **3. Fitness function**: close to uniform distribution of hashes (uniform distribution is the ideal), over 10,000 instances.

Distribution of Hashcodes

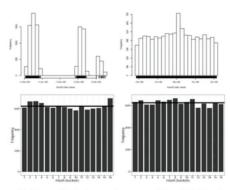


Fig. 1: The distribution of the hashcode values (top) and the distribution of the created objects in hash buckets (bottom), generated by the Apache commons (left) and the evolved function (right).

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Fixing Bugs and other examples

Saemundur O. Haraldsson

- Fixing bugs
- Making software faster
- Making software more accurate



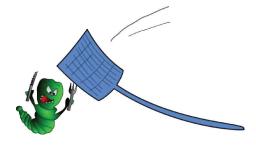
Fixing bugs

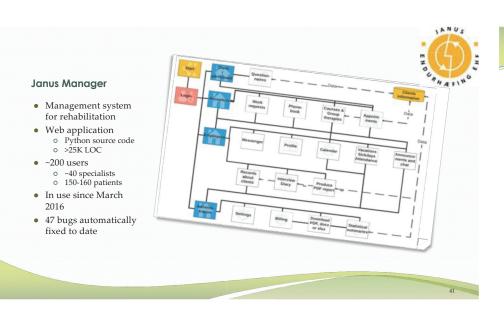
A real world example of GI in action

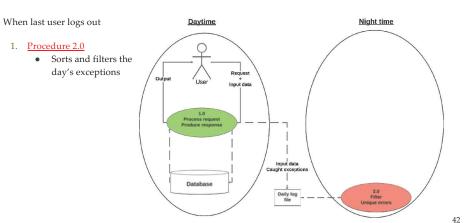
Saemundur O. Haraldsson, John R. Woodward, Alexander E. I. Brownlee, and Kristin Siggeirsdottir. 2017. Fixing bugs in your sleep: how genetic improvement became an overnight success. In Proceedings of the Genetic and Evolutionary Computation Conference Companion (GECCO '17). ACM, New York, NY, USA, 1513-1520. DOI: https://doi.org/10.1145/3067698.30082517

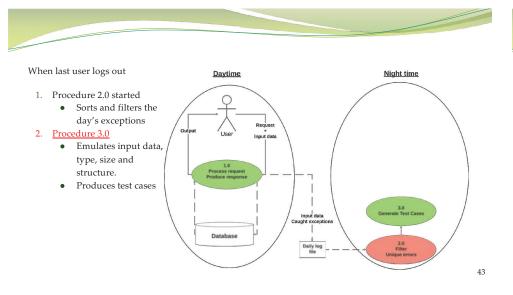
S. O. Haraldsson, J. R. Woodward and A. I. E. Brownlee, "The Use of Automatic Test Data Generation for Genetic Improvement in a Live System," 2017 IEEE/ACM 10th International Workshop on Search-Based Software Testing (SBST), Buenos Aires, 2017, pp. 28-31. DOI: https://10.1109/SBST.2017.10

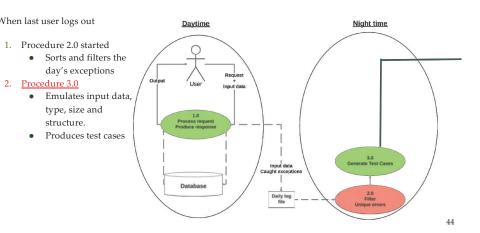
S.O. Haraldsson, 2017. 'Genetic Improvement of Software: From Program Landscapes to the Automatic Improvement of a Live System', PhD thesis, University of Stirling, Stirling, http://hdl.handle.net/1893/26007







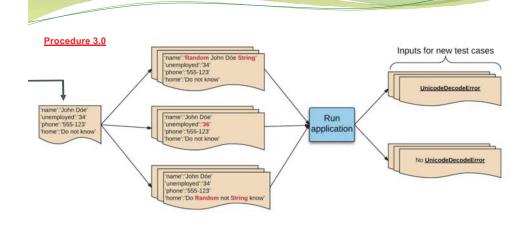


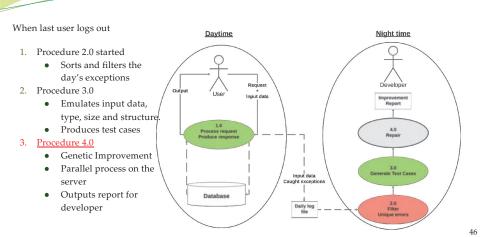


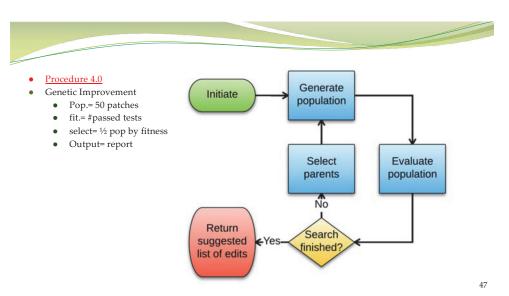
When last user logs out

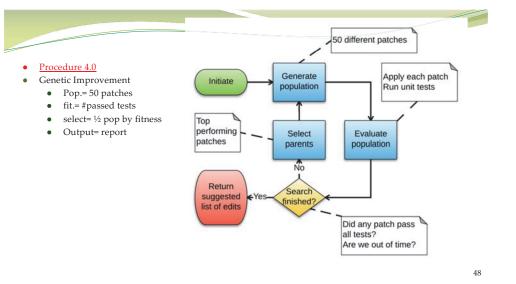
2. Procedure 3.0

structure.









4 different types of implemented Edits

Primitive types:

Copy

 Equivalent to: CTRL+C -> CTRL+V

Delete

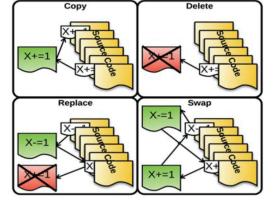
Almost what you think

Composite types:

Replace

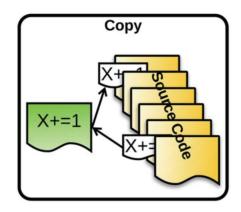
Copy + Delete

• Swap • 2x Copy + 2x Delete



Сору

- CTRL+C => CTRL+V
- Applied to whole lines
- Some restrictions on what lines can be copied
 - Identified with regular expressions

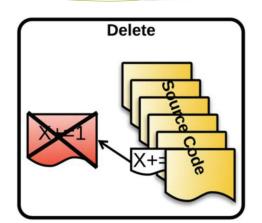


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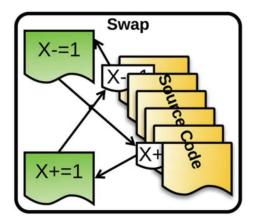
Delete

- Adds "#" to beginning of line
 - "Comment"
- Applied to whole lines
- Some restrictions on what lines can be commented out
 - Identified with regular expressions
- Can be reversed for previously deleted lines
 - "Uncomment"



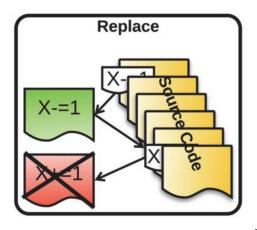
Swap

- Copies both lines above each other
- Then deletes the originals
- Applied to whole lines
- Like for like



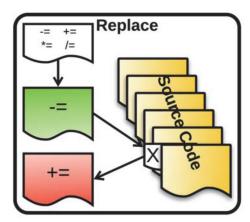
Replace

- Copies one line above another
- Then deletes that line



Replace -- extra

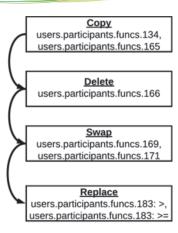
- Deep parameter tuning
- Operator specific replacement
- and numbers too
- From a list of equivalent operators.



_

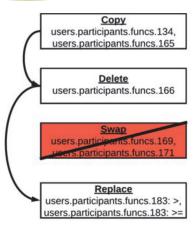
A list of edits makes a suggestion

- Reads like a recipe
- Step-by-step
- Automatically reduced
- Delta debugging
- Scrutinised by the developer
- Might change the recipe



A list of edits makes a suggestion

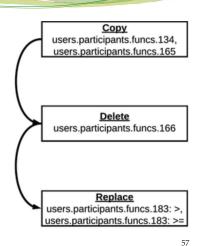
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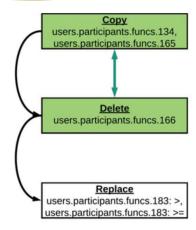
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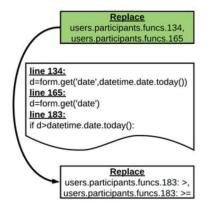
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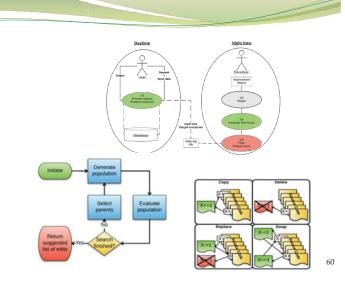
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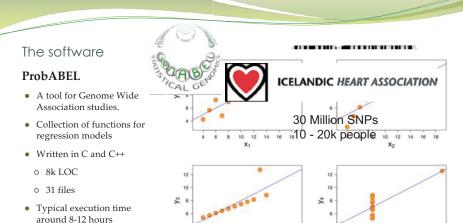


Summary

- Real-world example
- Catches inputs that produce crashes
- Line(-ish) based GI
- 4 types of edits
- Overnight repair
- Developers are the gatekeepers



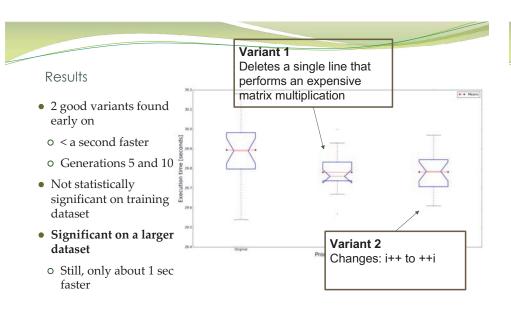
Faster Another example of GI in action Saemundur O, Haraldsson, John R. Woodward, Alexander E. I. Browniee, Albert V. Smith, and Vilmundur Gudnason. 2017. Genetic improvement of untime and its fitness landscape in a binformatics application. In Proceedings of the Genetic and Evolutionary Computation Conference Companion (GECCO 17), ACM, New York, NY, USA, 1621-1628. DOI: https://doi.org/10.1145/30676853.3082528 S.O. Haraldsson, 2017. 'Genetic improvement of Software: From Program Landscapes to the Automatic Improvement of a Live System', PhD thesis, University of Stirling, Stirling, http://hdl.handle.net/1633/26607

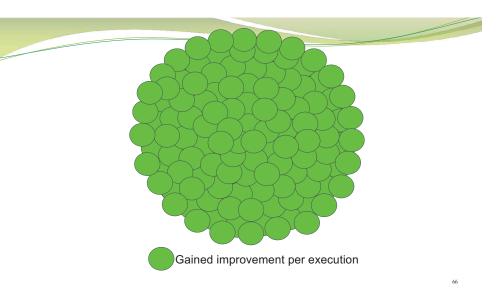


http://www.genabel.org/packages/ProbABEL

The GI setup Same as before Except for the evaluation Mean CPU time from 20 executions None compiling and failing variants are not discarded

2 good variants found early on < a second faster Generations 5 and 10 years Not statistically significant on training dataset Ougust Ougust Program variants Verset 3 Program variants





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Improving CUDA DNA Analysis Software with Genetic Programming (2015)
W.B. Langdon , B.Y.H. Lam , J. Petke & M. Harman



- 1. DNA sequencing
- 2. consisting of **8,000+** lines of code.
- 3. improved version is up to 3x faster
- 4. downloaded 1,000 times.
- **5. Ported by IBM** to one of their super computers

Optimising Existing Software with Genetic Programming

William B. Langdon and Mark Harman

- •Bowtie2, a **DNA sequence** alignment/sequence analysis tool
- •Using Genetic Improvement, Harman and Langdon were capable of increasing performance 70x.

A Systematic Study of Automated Program Repair; Fixing 55 out of 105 Bugs for \$8 Each

Stephanie Forrest

Computer Science Department

(2012)Cited ~400 times

University of Virginia University of New Mexico Charlottesville, VA Albuquerque, NM {legoues,mkd5m}@cs.virginia.edu forrest@cs.unm.edu Fitness = number of passed test cases Converted to AST CPP Mutation Operators: Deletion fault localisation Replace (e.g. LOC visited in Copy only a buggy case → high weight)

Michael Dewey-Vogt

Computer Science Department

Claire Le Goues

Test cases

Westley Weimer Computer Science Department University of Virginia Charlottesville, VA weimer@cs.virginia.edu

- Where an adequate test suite is provided, GenProg has been shown to fix realworld bugs
- It has inspired a variety of alternative frameworks, most of which claim to outperform GenProg

Mark Harman Yue Jia Alexandru Marginean Justyna Petke CREST, University College London, Malet Place, London, WC1E 6BT, UK {e.barr,m.harman,yue.jia,alexandru.marginean.13,j.petke}@ucl.ac.uk Featured in: CPP muScalpel Donor Host

Automated Software Transplantation

Babel Pidgin: SBSE Can Grow and Graft Entirely New Functionality into a Real World System

> Mark Harman, Yue Jia, and William B. Langdon University College London, CREST centre, UK

> > English to Korean; English to Portuguese



(2015)





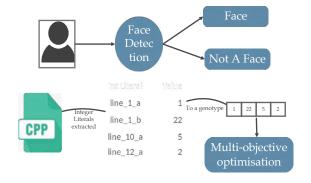
Deep Parameter Optimisation for Face Detection Using the Viola-Jones Algorithm in OpenCV

Bobby R. Bruce^{1(⊠)}, Jonathan M. Aitken^{2(□)}, and Justyna Petke^{1(□)}



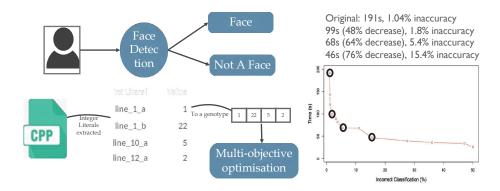
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Deep Parameter Optimisation for Face Detection Using the Viola-Jones Algorithm in OpenCV

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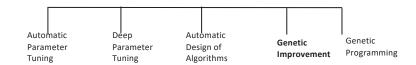


PHD THESES

- David R. White
- Andrea Accuri
- Bobby R. Bruce
- Sæmundur Ó. Haraldsson
- Mahmoud R. Bokhari
- And many more to come...

Relationship to other fields

- Optimization/machine learning OVERFITTING (or: specialisation?) ("Is the cure worse than the disease?" Smith et al. FSE 2015)
- Genetic Programming and Metaheuristics
- the automatic design of algorithms
- Automatic parameter tuning/deep parameter tuning/GI



GI & Benchmarking





- 1. GP suffered a "midlife crisis"
- 2. Toy problem e.g. lawnmower
- 3. Genetic Programming Needs Better Benchmarks [White et al.]
- 4. Machine Learning that Matter [Wagstaff 2012] what is 1% meaning
- 5. Is Software Engineering the best benchmark for GP?
- 6. Do we have a stable set of benchmarks for GI? (for program repair: http://program-repair.org/benchmarks.html))
- 7. Benchmarking is more complex (noise, hardware, prog lang, ...)

Measuring Energy

• computational energy consumption growing importance, particularly at the extremes (i.e., mobile devices and datacentres).

one line = one unit

simulate (runtime/system calls/) Tools Opacitor, PowerGauge

read battery indicator

physically measure and validate(e.g. see Bokhari et al.) GECCO '20 Towards Rigorous Validation of Energy Optimisation Experiments

Deep Parameter Optimisation on Android Smartphones for GI@GECCO'17 Energy Minimisation - A Tale of Woe and a Proof-of-Concept

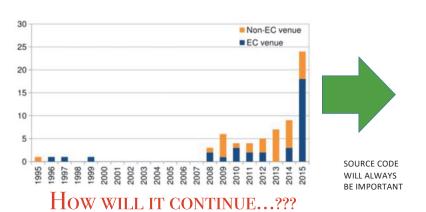
CEC 2019 Mind the gap - a distributed framework for enabling energy optimisation on modern smart-phones in the presence of noise, drift, and statistical insignificance [#19776]



GENETIC PROGRAMMING AND EVOLVABLE MACHINES e 62nd CREST Open Workshop - Au ogram Repair and Genetic Improve video

GI @ ICSE 2020 International Workshop on Automatic Software Optimisation International Workshop on Automatic Software
Optimisation activities of GI @ GECCO 2019 in Prague, Czech Republ The 6th International Workshop on Genetic Improvement @ICSE January 2018 SCHLOSS DAGSTUHL

GROWTH OF PAPERS



Source of Genetic Material

- 1. the program being improved,
- 1. a different program written in the same language (Petke: MiniSAT competition),

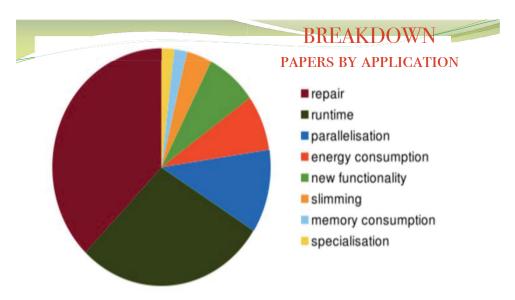
J. Petke, M. Harman, W. B. Langdon, and W. Weimer, "Using genetic improvement and code transplants to specialise a C++ program to a problem class," in *European Conf. on Genetic Programming EuroGP*, ser. LNCS, vol. 8599. Springer, 2014, pp. 137–149.

Leibniz-Zentrum für Informatik

- 1. a piece of code generated from scratch (GP),
- 1. different programming language other than the software to be improved.

THEORY

- Hard!
- NFL not really valid for GP, and therefore GI.
 - Why because many programs share same functionality.
 - => GI will remain empirical for years to come



Grant Writing

• A grant about GP (0%)

VS

• A grant about GI. (100%)



Websites Genetic Improvement Workshop

An International Workshop on the Repair and Optimisation of Software using Computational Search

- http://geneticimprovementotsottware.com/
- https://en.wikipedia.org/wiki/Genetic improvement (com puter science)
- http://www.davidrwhite.co.uk/

Google Scholar

label:genetic_improvement

- http://daase.cs.ucl.ac.uk/
- http://crest.cs.ucl.ac.uk/publications/
- https://clairelegoues.com/blog/
- https://cs.adelaide.edu.au/~optlog/research/software.php

Starting point - POP science, GIN, Survey

IEEE TRANSACTIONS ON EVOLUTIONARY COMPUTATION

Genetic Improvement of Software: a Comprehensive Survey

Justyna Petke, Saemundur O. Haraldsson, Mark Harman, William B. Langdon, David R. White, and John R. Woodward

A Survey of Genetic Improvement Search Spaces

GI@GECCO'19

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Overview

- Introduction
- Fixing Bugs and other examples
- Noteworthy papers and issues
- Getting involved
- Summary and Q&A

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GI in No Time

David R. White UCL, London, UK david.r.white@ucl.ac.uk

(Inaugural paper at GI@GECCO 2017)

Get involved with GI in No time - or GIN

Available at https://github.com/gintool/gin



http://www.davidrwhite.co.uk/



v2.0 published in June 2019 "Gin: Genetic Improvement Research Made Easy" (GECCO 2019)

02

The inaugural paper

official V2.0 released on 12 June 2019: https://github.com/gintool/gin/releases

Gin: Genetic Improvement Research Made Easy

Alexander E.I. Brownlee Computing Science and Mathematic University of Stirling Stirling, UK sbr@es.stir.ac.uk

Earl T. Barr epartment of Computer Sc

ABSTRACT

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David R. White Department of Computer Science The University of Sheffield Sheffield, UK

1

Genetic improvement (GI) is a young field of research on the cupof transforming software development. GI uses search to improve existing software. Researchers have already shown that GI can improve human-written code, ranging from program repair to optinising run-time, from reducing energy-consumption to the transplantation of new functionality. Much remains to be done. The cost of re-implementing GI to investigate new approaches is hindering progress. Therefore, we present GIn, an extensible and modifiable 1 INTRODUCTION

Genetic improvement (GD) is a young field of software engineering research that uses search to improve existing software. GI aims to improve both functional, notably bug fixing, and non-functional properties of software, such as runtime or energy consumption. The intersection of automated program repair (APR) and GI has had the greatest impact to date, from the release of the GI-based tool GenProg [27] to successful integration of APR into commercial development processes [19, 20]. Non-functional improvement (NB) is



Bradley Alexander



Justyna Petke



Earl T.



Markus Wagner
Also uses GIN in teaching since 2017
https://tinyurl.com/giassignment



Sandy Brownlee



David R. White

"Stupidly simple"

GIN





https://cs.gmu.edu/~eclab/projects/ecj/

Genetic Improvement

- Many success stories
- •...however, these typically need at GI expert in the loop
- •What's needed is a more systematic approach
- A toolkit to enable experimentation

Gin's Goals

- Remove incidental difficulties of GI for research and teaching
- Enable focus on general questions
- Provide a central tool for the community
- Support more than bug-fixing: non-functional properties
- \bullet Work on open-source software projects out-of-the-box

Gin Design



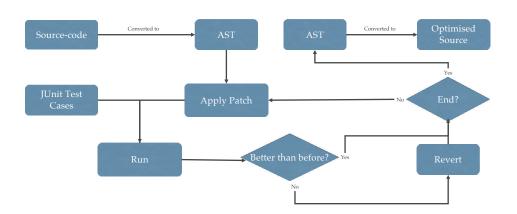
What's in Gin?

- Implementations of edits for source code
- Evaluate edits: compile and run JUnit tests
- Searches and Samplers
- Test generation (EvoSuite)
- Profiler to identify hot methods (hprof)
- Build tool integration (Maven, Gradle)

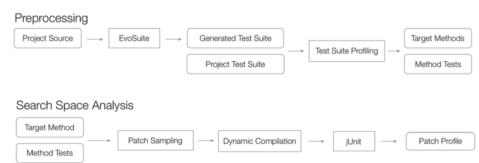
Let's see those in more detail...

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VANILLA GIN: NEIGHBOURHOOD SEARCH



Gin Pipelines

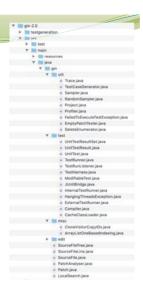


Vanilla GIN Version 1.0

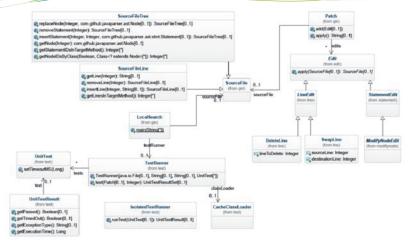
▼ ■ gin-1.0 ▼ ■ usc ▼ ■ test ▼ ■ main ▼ ■ pris ▼ ■ pris ▼ ■ pris ▼ ■ pris □ restflamer java □ restflamer java □ restflamer java □ cochroida. ouder java ▼ ■ dit ■ ■ modifynode □ blooderflamer java □ cochroida. ouder java ■ cochroida. ouder java ■ blooderflamer java □ blooderflamer java □ cochroida. ouder java □ blooderflamer java □ cochroida. ouder java □ cochroida.

Vanilla GIN Version 2.0:

GRADLE/MAVEN SUPPORT,
VARIOUS TYPES OF EDITS,
PROFILER TO FIND "HOT" METHODS,
VARIOUS SAMPLERS. ...



Gin v2 Core Classes



Edits

- Edits are single changes to source code
 - Building blocks of a repair
 - Combined into Patches
 - Question: actually, what scale might an *edit* be?
- •Gin supports edits at:
 - line level (Langdon) delete/replace/copy/swap/move
 - statement level (GenProg) delete/replace/copy/swap/move
 - constrained (matched) statement replace/swap
 - micro edits
 - \bullet binary & unary operator replacement (OR $\Leftrightarrow\! AND)$ (++ $\Leftrightarrow\!$ --)
 - ullet reorder Boolean expressions (X && Y \Leftrightarrow Y && X)
 - loop and method shortcuts (insert return/break/continue)

Edits

- •We provide many wrappers to make your life easier, so that you can focus on higher-level tasks:
 - "Tell me which lines are eligible for deletion in this method"
 - "Delete this line"
 - "Give me all the for loop conditions in this method"
 - And many more...

Example edits

```
1 public class ReplaceStatement extends StatementEdit {
    public int sourceID;
    public int destinationID;
    public ReplaceStatement(SourceFileTree sf, Random r) {
      sourceID = sf.getRandomStatementID(false, r);
      destinationID = sf.getRandomStatementID(true, r);
9
10
11
    public SourceFile apply(SourceFileTree sf) {
      Statement source = sf.getStatement(sourceID);
12
      Statement dest = sf.getStatement(destinationID);
13
14
      return sf.replaceNode(dest, source.clone());
15
16
17 }
```

Disclaimer: this was an old version. Today, it is a little bit longer, e.g., to prevent us from replacing statements within the same parent node.

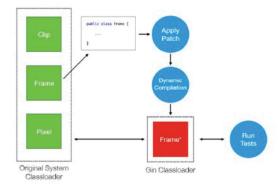
Patch Evaluation

Gin invokes test cases via Junit...

Tracks:

- •compile success;
- •run-time errors, exception types
- actual & expected outcomes
- •timing: wallclock and CPU time

An analogy: video editing. Here: Gin Compiles and Reloads onthe-fly



Note: If you prefer to use the more "traditional" way of writing the file to disk first - e.g., due to integration of Gin into other pipelines - then you can use a command-line flag to do so.

Sampling and Searching

- Included samplers:
 - EmptyPatchTester
 - RandomSampler
 - DeleteEnumerator
- Searches: LocalSearch, GP
- Possible Questions:
 - What is the effectiveness of a given edit type for fixing a category of bug?
 - How robust is the space of single-line edits, modulo the given test suite?

• ..

DeleteEnumerator

```
public static void main(String[] args) {
 UnitTest[] ut = {
  new UnitTest("TriangleTest", "testInvalidTriangles")
 int reps = 1:
 SourceFileTree sf = new SourceFileTree(
       "examples/simple/Triangle.java".
     Collections.singletonList(
           "classifyTriangle(int,int,int)"));
 TestRunner tr = new TestRunner(
     new File("examples/simple"), "Triangle"
      "examples/simple", Arrays.asList(ut));
// Start with the empty patch
 // Run empty patch and log
 UnitTestResultSet rs = tr.test(patch, reps);
writeResults(rs, 0);
 for (int id : sf.getStatementIDsInTargetMethod()) {
  patchCount++:
  patch.add(new DeleteStatement(sf.getFilename().id)):
  rs = tr.test(patch, reps):
  writeResults(rs. patchCount);
```

Sampling

The following is one really wide output file - here of RandomSampler:

PatchIndex PatchSize Patch

1		gm.edit.statement.swapstatement./src/maii/java/org/jcodec/codecs/vpx/vrxbitstream.java./52 <->./src/maii/java/org/jcodec/codecs/vpx/vrxbitstream.java.864
2	1	$ \ gin.edit.statement. Replace Statement . /src/main/java/org/jcodec/codecs/prores/ProresEncoder.java: 2310 -> . /src/main/java/org/jcodec/codecs/prores/ProresEncoder.java: 1185 -> . /src/main/java/org/jcodec/codecs/prores/prores/Prores/Prores/Prores/Prores/Prores/Prores/Prores/Prores/Prores/P$

1 | gin edit. statement. CopyStatement /src/main/java/org/jcodec/containers/mp4/boxes/Box.java-S14-> /src/main/java/org/jcodec/containers/mp4/boxes/Box.java-S10-110 |

TestTimedOut	TestExceptionType	TestExceptionMessage	AssertionExpectedValue	AssertionActualValue	
FALSE	java.lang.AssertionError	expected:<255> but was:<207>	255	207	
FALSE	N/A	N/A	N/A	N/A	
FALSE	N/A	N/A	N/A	N/A	

MethodIndex Test	RepNumber	PatchValid	PatchCompiled	TestPassed	TestExecutionTime(ns)	TestCPUTime(ns)	
152	1 org.jcodec.codecs.vpx.TestCoeffEncoder.testCoeffDCTU []	(TRUE	TRUE	FALSE	2853708	1535633
189	1 org.jcodec.codecs.prores.ProresEncoderTest.testWholeThing []		TRUE	FALSE	FALSE	0	0
184	1 org.jcodec.containers.mp4.boxes.TrunBoxTest.testReadWriteCreate []		TRUE	FALSE	FALSE	0	0

Local search, output

-bash-4.1\$ java -jar build/gin.jar gin.LocalSearch -filename examples/triangle/Triangle.java -m "classifyTriangle(int, int, int)"

Local search

```
1 private Patch search() {
      // start with the empty patch
      Patch_bestPatch = new Patch(sourceFile);
long bestTime = testRunner.test(bestPatch, 10).
            totalExecutionTime();
       for (int step = 1; step <= NUM_STEPS; step++) {</pre>
          Patch neighbour = neighbour(bestPatch, rng);
          UnitTestResultSet rs = testRunner.test(neighbour
          if (rs.getValidPatch() && rs.getCleanCompile() &&
              rs.allTestsSuccessful() &&
              rs.totalExecutionTime() < bestTime) {
            bestPatch = neighbour;
bestTime = rs.totalExecutionTime();
       return bestPatch;
20 public Patch neighbour(Patch patch, Random rng) {
      Patch neighbour = patch.clone();
      if (neighbour.size() > 0 && rng.nextFloat() > 0.5) {
          neighbour.remove(rng.nextInt(neighbour.size()));
          neighbour.addRandomEdit(rng, allowableEditTypes);
      return neighbour;
```

Local search, output

-bash-4.15 java -jar build/gin.jar gin.localSearch-filename examples/friangle.java -a "classifyTriangle(int, int, int)"
2020-04-10 06:36:41 gin.localSearch.escrio() INFO: localsearch of file examples/friangle/friangle/friangle/friangle/gin.
2020-04-10 06:36:44 gin.localSearch.escrio() INFO: localsearch of file examples/friangle/friangle/gin.

Local search, output

-bash-4.13 java -jar build/gin.jar gin.localSearch -filename examples/triangle/Triangle.java -m "classifyTriangle(int, int, int)"
2002-04-10 04:16:14 gin.localSearch.search() INFO: Localsearch on file: examples/triangle/Triangle/Triangle/int int) classifyTriangle(int, int, int)
2002-04-10 04:36:59 gin.localSearch.search() INFO: Original execution time: 164697123]-gin.localSearch.search() INFO: Step: 1, Patch: | gin.edit.line.ReplaceLine examples/triangle/Triangle.java:23 |, Falled to compile

Local search, output

```
bash-4.15 java -jar build/gin.jar gin.LocalBearch -filename examples/triangle/Triangle.java -m "classifyfriangle(int, int, int)"
2020-04-10 04:36:44 gin.LocalBearch.search() INFO: Localsearch on file: examples/triangle/Triangle.java method: classifyfriangle(int, int, int)
2020-04-10 04:36:36 gin.LocalBearch.search() INFO: Decalsearch on file: examples/triangle/Triangle.java method: classifyfriangle(int, int, int)
2020-04-10 04:36:39 gin.LocalBearch.search() INFO: Original execution time: 106971218ms
2020-04-10 04:36:39 gin.LocalBearch.search() INFO: Step: , Patch: | gin.edit.line.PelsteLine examples/triangle/Triangle.java:5 ->
2020-04-10 04:36:39 gin.LocalBearch.search() INFO: Step: 2, Patch: | gin.edit.line.DelsteLine examples/triangle/Triangle.java:16 , Failed to compile
2020-04-10 04:36:36:39 gin.LocalBearch.search() INFO: Step: 3, Patch: | gin.edit.line.DelsteLine examples/triangle/Triangle.java:18 , Failed to compile
2020-04-10 04:36:36:39 gin.LocalBearch.search() INFO: Step: 5, Patch: | gin.edit.line.DelsteLine examples/triangle/Triangle.java:18 , Failed to compile
2020-04-10 04:36:36:39 gin.LocalBearch.search() INFO: Step: 5, Patch: | gin.edit.line.DelsteLine examples/triangle/Triangle.java:31 ->
2020-04-10 04:36:36:39 gin.LocalBearch.search() INFO: Step: 5, Patch: | gin.edit.line.DelsteLine examples/triangle/Triangle.java:31 ->
2020-04-10 04:37:00 gin.test.linernalTeatCompile
20
```

Local search, output

```
Pash-4.15 java -jar build/gin.jar gin.LocalSearch -filename examples/triangle/Triangle.java = "classif/Triangle.java (int. int.)"
2020-04-10 04:36:44 gin.LocalSearch.search() INFO: Localsearch on file: examples/triangle.java method: classif/Triangle.java int. int.)
2020-04-10 04:36:44 gin.LocalSearch.search() INFO: Localsearch on file: examples/triangle.java method: classif/Triangle.java int. int.)
2020-04-10 04:36:49 gin.LocalSearch.search() INFO: Original execution time: 104697:12168
2020-04-10 04:36:39 gin.LocalSearch.search() INFO: Step: 2, Patch: | gin.edit.line.ReplaceLine examples/triangle/Triangle.java:36 |, Failed to compile
2020-04-10 04:36:39 gin.LocalSearch.search() INFO: Step: 2, Patch: | gin.edit.line.DeleteLine examples/triangle/Triangle.java:38 |, Failed to compile
2020-04-10 04:36:39 gin.LocalSearch.search() INFO: Step: 3, Patch: | gin.edit.line.DeleteLine examples/triangle/Triangle.java:18 |, Failed to compile
2020-04-10 04:36:39 gin.LocalSearch.search() INFO: Step: 5, Patch: | gin.edit.line.DeleteLine examples/triangle/Triangle.java:18 |, Failed to compile
2020-04-10 04:36:39 gin.LocalSearch.search() INFO: Step: 5, Fatch: | gin.edit.line.DeleteLine examples/triangle.java:19 |, Failed to compile
2020-04-10 04:36:39 gin.LocalSearch.search() INFO: Step: 7, Fatch: | gin.edit.line.DeleteLine examples/triangle.java:19 |, Failed to compile
2020-04-10 04:36:39 gin.LocalSearch.search() INFO: Step: 7, Fatch: | gin.edit.line.DeleteLine examples/triangle.java:17 |, Failed to compile
2020-04-10 04:36:39 gin.LocalSearch.search() INFO: Step: 7, Fatch: | gin.edit.line.DeleteLine examples/triangle.java:34 ->
examples/triangle/Triangle.java:17 |, Failed to compile
2020-04-10 04:36:39 gin.LocalSearch.search() INFO: Step: 7, Fatch: | gin.edit.line.DeleteLine examples/triangle.java:34 ->
examples/triangle/Triangle.java:10 |, Failed to compile
2020-04-10 04:36:30 gin.LocalSearch.search() INFO: Step: 96, Fatch: | gin.edit.line.DeleteLine examples/triangle/Triangle.java:10 | gin.edit.line.SwapLine
examples/triangle/
```

Local search: What did we actually optimise here?

Generating tests and Profiling

Generate new test cases

java -cp build/gin.jar gin.util.TestCaseGenerator
-projectDir examples/maven-simple -projectName my-app
-classNames com.mycompany.app.App -generateTests

Profile a test suite

java -cp build/gin.jar gin.util.Profiler -p my-app -d examples/maven-simple/ .

Results written to profiler_output.csv.

Examples with jCodec (maven project)

• Profiler

projectnameforgin='jcodec';

java -Dtinylog.level=trace -cp ../../ginfork/build/gin.jar gin.util.Profiler -h ~/.sdkman/candidates/maven/current/ -p \$projectnameforgin -d . -o \$projectnameforgin.Profiler_output.csv -r 1

Build tool integration

- Maven and Gradle API documentation is sparse!
 - And many projects seem to break conventions about paths, resources etc.
- Project class wraps most of what we have learned
 - provide the classpath for a project
 - find a particular source file within a project's file hierarchy
 - provide a standard method signature for a given method
 - provide a list of project tests
 - run a unit test given its name
- •Gin can infer the necessary classpath and dependencies for running unit tests from a Maven or Gradle project, or these can be specified manually
- Maven projects can be updated automatically with new unit tests from EvoSuite

Examples with jCodec (maven project)

Profile

projectnameforgin='jcodec';

java -Dtinylog.level=trace -cp ../../ginfork/build/gin.jar gin.util.Profiler h ~/.sdkman/candidates/maven/current/ -p \$projectnameforgin -d -o \$projectnameforgin.Profiler_output.csv -r 1

Examples with jCodec (maven project)

projectnameforgin='jcodec';

java -Dtinylog.level=trace -cp ../../ginfork/build/gin.jar gin.util.Profiler -h ~/.sdkman/candidates/maven/current/ -p \$projectnameforgin -d . -o \$projectnameforgin.Profiler_output.csv -r 1

• EmptyPatchTester

projectnameforgin='jcodec';

java -Dtinylog.level=trace -cp ../../ginfork/build/gin.jar gin.util.EmptyPatchTester -h ~/.sdkman/candidates/maven/current/ -p \$projectnameforgin -d .
-m \$projectnameforgin.Profiler_output.csv

-o \$projectnameforgin.EmptyPatchTester_output.csv

Examples with jCodec (maven project)

projectnameforgin='jcodec';

java -Dtinylog.level=trace -cp ../../ginfork/build/gin.jar gin.util.Profiler -h ~/.sdkman/candidates/maven/current/ -p \$projectnameforgin -d -o \$projectnameforgin.Profiler_output.csv - 1

• EmptyPatchTester

projectnameforgin='jcodec';

java -Dtinylog.level=trace -cp ../../ginfork/build/gin.jar gin.util.EmptyPatchTester -h ~/.sdkman/candidates/maven/current/ -p \$projectnameforgin -d . -m \$projectnameforgin.Profiler_output.csv

-o \$projectnameforgin.EmptyPatchTester_output.csv

• PatchSampler

projectnameforgin='jcodec';

java -Dtinylog.level=trace -cp ../../ginfork/build/gin.jar gin.util.PatchSampler h. c/.sdkman/candidates/maven/current/ -p \$projectnameforgin -d . -m \$projectnameforgin.Profiler_output.csv

-o \$projectnameforgin.PatchSampler_LINE_output.csv -editType LINE -patchNo 100

Gin: Genetic Improvement Research Made Easy

• Available at https://github.com/gintool/gill

- The team actively uses Gin to push the GI boundaries, and quite a few papers are in the works.
- Open for contributions!
 - Particularly new edits and tools
 - https://github.com/gintool/gin
 - we'd like this to become the MiniSAT of GI





Injecting Shortcuts for Faster Running Java Code

Analysing Program Transformation Spaces for Genetic Improvement using Gin

lustyna Petke, Brad Alexander, Earl T. Barr, Alexander E.I. Brownlee, Markus Wagner, David R.White

Software Improvement with Gin: A Case Study

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University of Stirling, Stirling, UK sbr@cs.stir.ac.uk

Overview

- Introduction
- Fixing Bugs and other examples
- Noteworthy papers and issues
- Getting involved
- Summary and O&A

Comments/questions: Sandy (Alexander E.I. Brownlee) sbr@cs.stir.ac.uk

Genetic Improvement vs Genetic Programming

- 1. Start from an existing program
- 2. BLOAT? interpretation?
- 3. NO function / terminal set
- 4. Improvement of non-functional properties.
- 5. Easier to write grants
- 6. Different benchmarks.
- 7. Population of edits **NOT programs**.

Questions?

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Brad Alexander

bradley.alexander@adelaide.edu.au>

PUTTING IT ALL TOGETHER

- •Let's start with **existing programs**. Not like standard GP.
- Python vs C vs Java? Amenable to GI? Most popular
- •Benchmarking???
- Population of edits, not programs
- •GP applied to real software
 - Large, loops, side-effect, modules,...
 - Non functional properties

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