



Genetic Improvement Programming

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GISMOE: Genetic Improvement of Software for Multiple Objectives

Genetic Programming to Improve Existing Software

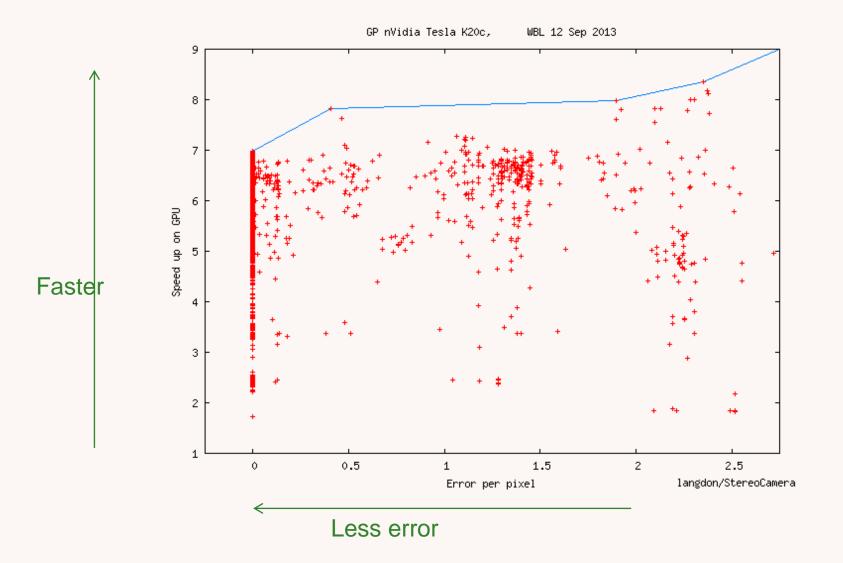
- Why
- Background
 - GP to write software
 - GP to improve human written programs
- Examples
 - Demonstration systems, automatic bug fixing
 - Evolving code for a new environment (gzip)
 - Improving non-functional properties <u>IEEE TEC</u>
 - Faster parallel code for stereo imaging
- Implications



When to Automatically Improve Software

- When to use GP to create source code
 - Small. E.g. glue between systems "mashup"
 - Hard problems. Many skills needed.
 - Multiple conflicting ill specified non-functional requirements
- Genetic programming as tool. GP tries many possible options. Leave software designer to choose between best.

Tradeoff 2 objectives Pareto front



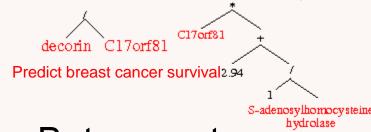
Some applications of Genetic Programming

- Most GP generates solutions, e.g.:
 - data modelling,
 - chemical industry: soft sensors,

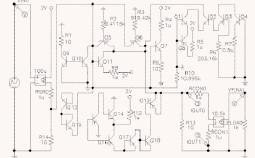


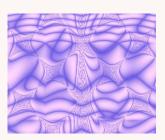
- image processing,
- predicting steel hardness,

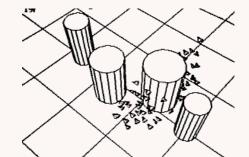
cinema "boids", Cliff hanger, Batman returns













Genetic Programming to Create Software

- GP has created real programs
 - domain specific hash functions
 - cache management
 - heap management, garbage collection
 - evolving communications protocols
- These can do better than existing standard approach by GP not only creating code but also tailoring it for specific use

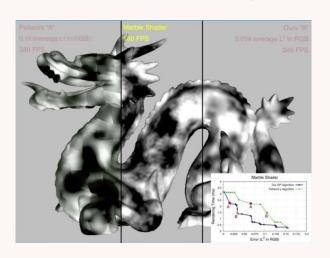
Genetic Programming to Improve Human written Programs

- Gluing together existing programs to create new functionality
 - combining object files
 - web services, mashup



GP to Improve human written programs

- Finch: evolve Java byte code
 - no compilation errors, 6 benchmarks
- Improving GPU shaders
- Functionality v speed or battery life

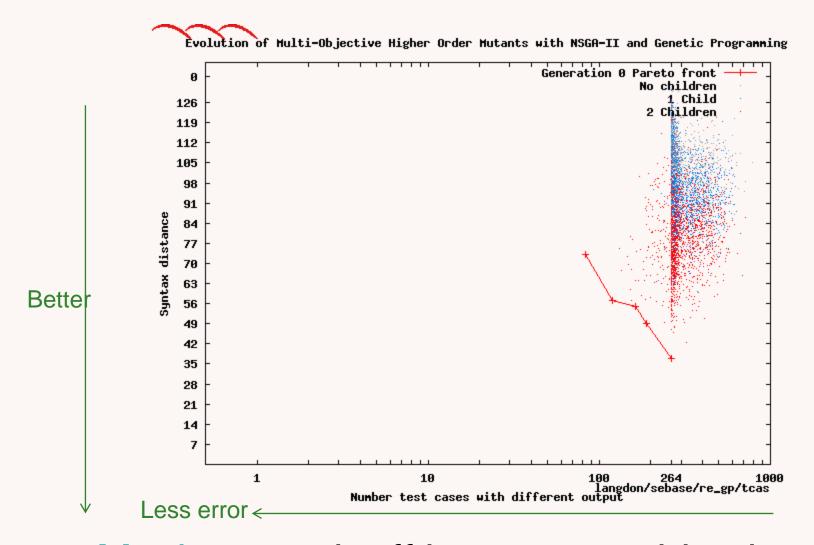


```
int Factorial(int a)
{
  if (a <= 0)
    return 1;
  else
    return (a * Factorial(a-1));
}
Factorial source code,</pre>
```

87% reduction in instructions, [white,2011]



GP Evolving Pareto Trade-Off



Movie to tradeoff between 2 objectives



GP Automatic Bug Fixing

- Run code: example to reproduce bug, a few tests to show fixed code still works.
- Search for replacement C statement within program which fixes bug.
- Real bugs in real C programs.
 - 1st prize Human-Competitive GECCO 2009
 Gold Humie

GP Automatic Coding

- Show a machine optimising existing human written code to trade-off functional and nonfunctional properties.
 - E.g. performance versus:
 Speed or memory or battery life.
- Trade off may be specific to particular use.
 For another use case re-optimise
- Use existing code as test "Oracle".
 (Program is its own functional specification)

GP Automatic Coding 2

- Target non-trivial open source system:
 - Bowtie2 state-of-the-art DNA lookup tool
- Tailor existing system for specific use:
 - nextgen DNA from 1000 genomes project
- Use existing system as test "Oracle"
 - Smith-Waterman exact algorithm (slow)
- Use inputs & answer to train GP.
- Clean up new code

Problems with BLAST

- BLAST contains biologists heuristics and approximations for mutation rates. It is the "gold standard" answer.
 - A few minutes per look up
- "Next Gen" DNA sequencing machines generate 100s millions short noisy DNA sequences in about a day.
- BLAST originally designed for longer sequences. Expects perfect data. Human genome database too big for PC memory.

Human Generated Solutions

- More than 140 bioinformatic sequence tools
- All human generated (man years)
- Many inspired by BLAST but tailored to
 - DNA or Proteins
 - Short or long sequences. Any species v man.
 - Noise tolerance. Etc. etc.
- Manual trade-off lose accuracy for speed
 - Bowtie 35million matches/hour but no indels
 - Bowtie2 more functionality but slower

Why Bowtie 2?

- Target Bowtie2 DNA sequencing tool
 - 50000 line C++, 50 .cpp 67 .h files, scripts, makefile, data files, examples, documentation
 - SourceForge
 - New rewrite by author of successful C Bowtie
- Aim to tailor existing system for specific (important data source)
- 1000 genomes project
 - Project aims to map all human mutations
 - 100s millions of short human DNA sequences
 - Download raw data via FTP

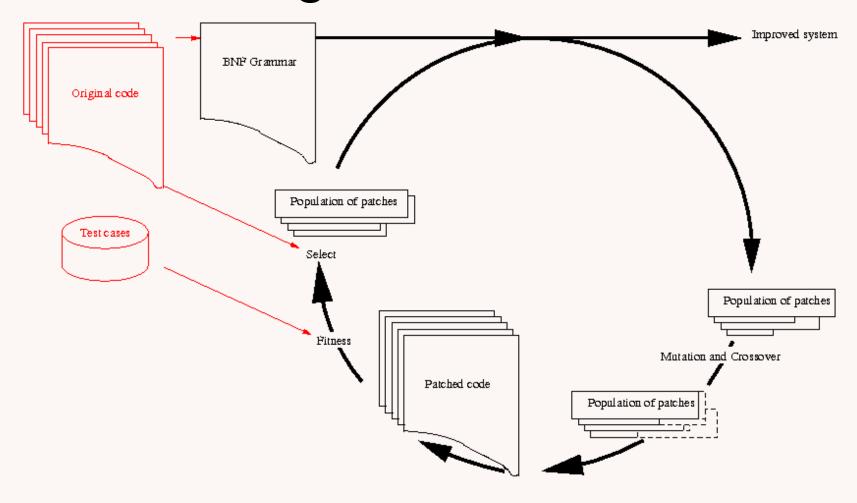


Evolving Bowtie2

- Convert code to grammar
- Grammar used to both instrument code and control modifications to code
- Genetic programming manipulates patches
 - Small movement/deletion of existing code
 - New program source is syntactically correct
 - Compilation errors mostly variable out-ofscope



GP Evolving Patches to Bowtie2



BNF Grammar

```
vhi = _mm_cmpeq_epi16(vhi, vhi); // all elts = 0xffff
vlo = _mm_xor_si128(vlo, vlo); // all elts = 0
vmax = vlo;
```

Lines 363-365 aligner_swsse_ee_u8.cpp

Fragment of Grammar (Total 28765 rules)

7 Types of grammar rule

- Type indicated by rule name
- Replace rule only by another of same type
- 5792 statement (eg assignment, Not declaration)
- 2252 IF

```
    <pe_118> ::= "{Log_count64++;/*20254*/} if" <IF_pe_118> " {\n"
    <IF_pe_118> ::= "(!olap)"
```

• 272 for1, for, for3

```
• <sam_36> ::= "for(" <for1_sam_36> ";" <for2_sam_36> ";" <for3_sam_36> ") {\n"
```

106 WHILE

```
<pat_731> ::= "while" <WHILE_pat_731> " {\n"<WHILE_pat_731> ::= "(true)"
```

24 ELSE

```
    <aln_sink_951> ::= "else {" <ELSE_aln_sink_951> " {Log_count64++;/*21439*/}};\n"
    <ELSE_aln_sink_951> ::= "met.nunp_0++;"
```



Representation

- GP evolves patches. Patches are lists of changes to the grammar.
- Append crossover adds one list to another
- Mutation adds one randomly chosen change
- 3 possible changes:
 - Delete line of source code (or replace by "", 0)
 - Replace with line of Bowtie2 (same type)
 - Insert a copy of another Bowtie2 line

Example Mutating Grammar

```
<_aligner_swsse_ee_u8_707> ::= "vh = _mm_max_epu8(vh, vf);"
<_aligner_swsse_ee_u8_365> ::= "vmax = vlo;"
```

2 lines from grammar

```
<_aligner_swsse_ee_u8_707><_aligner_swsse_ee_u8_365>
```

Fragment of list of mutations

Says replace line 707 of file aligner_swsse_ee_u8.cpp by line 365

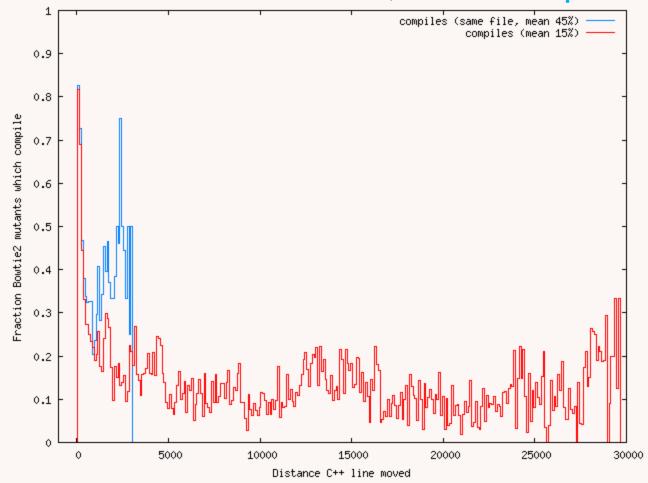
```
vh = _{mm_max_epu8(vh, vf); \{Log_count64++; /*28919*/\}}
```

Instrumented original code

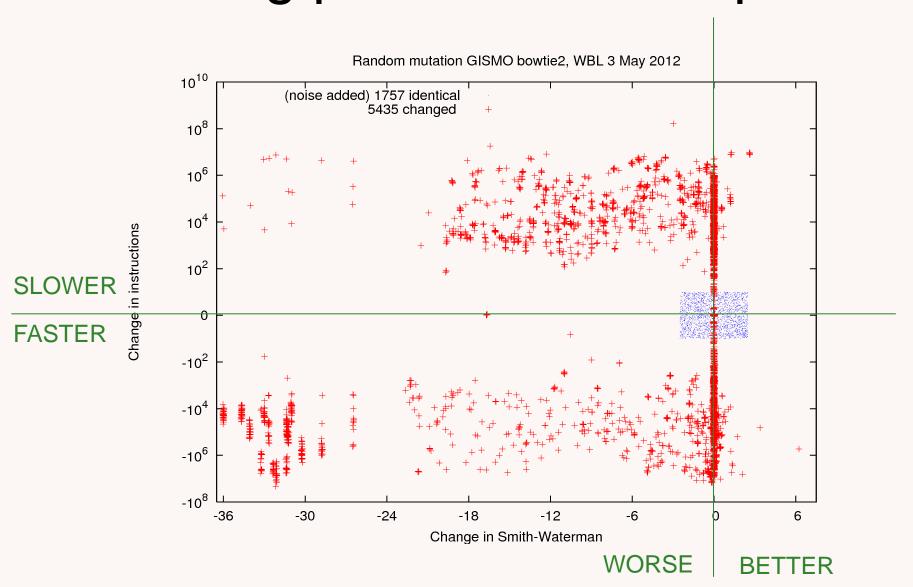


Compilation Errors

- Use grammar to replace random line, only 15% compile. But if move <100 lines 82% compile.
- Restrict moves to same file, 45% compile



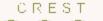
C++ is not fragile Trading performance v speed



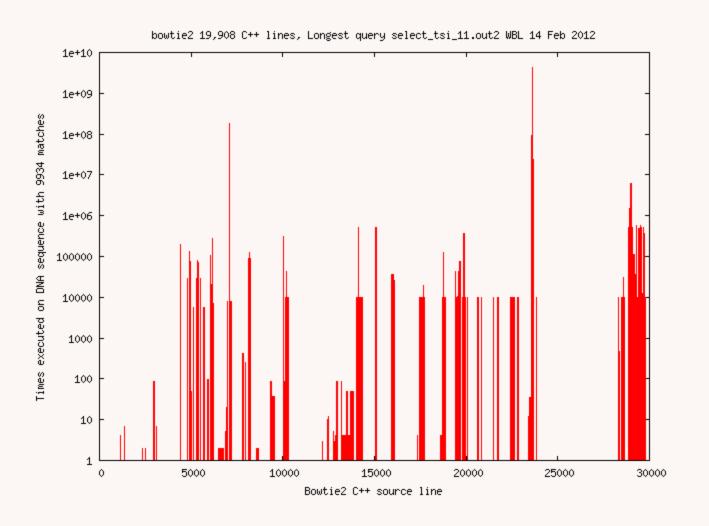


Recap

- Representation
 - List of changes (delete, replace, insert). New rule must be of same type
- Genetic operations
 - Mutation (append one random change)
 - Crossover (append other parent)
- Apply change to grammar then use it to generate new C++ source code.

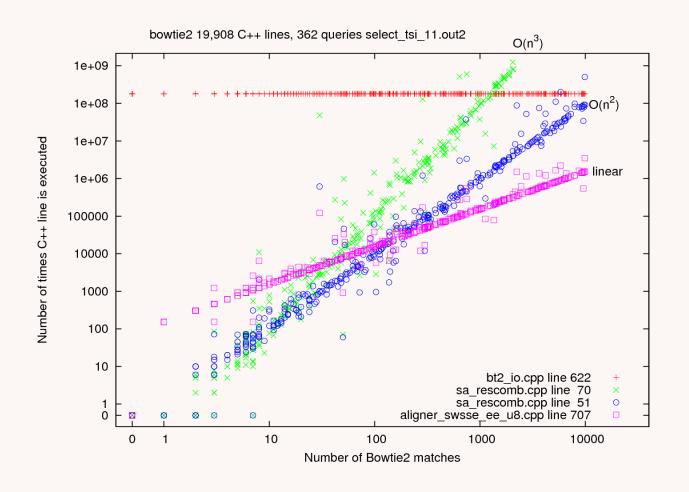


Which Parts of Bowtie2 are Used

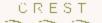




Scaling of Parts of Bowtie2

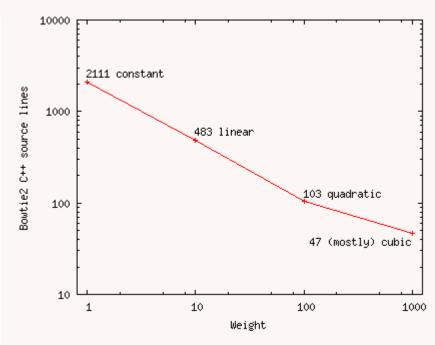


4 Heavily used Bowtie2 lines which scale differently



Focusing Search

C++ Lines	Files	Bowtie2	
50745	50 .cpp, 67 .h	All C++ source files	
19908	40 .cpp	no conditional compilation no header files.	
2744	21 .cpp	no unused lines	
		Weights target high usage	
39	6 .cpp	evolve	
7	3 .cpp	clean up	





Testing Bowtie2 variants

- Apply patch generated by GP to instrumented version of Bowtie2
- "make" only compiles patched code
 - precompile headers, no gcc optimise
- Run on small but diverse random sample of test cases from 1000 genomes project
- Calculate fitness
- Each generation select best from population of patched Bowtie2



Fitness

- Multiple objective fitness
 - Compiles? No→no children
 - Run patched Bowtie2 on 5 example DNA sequences from The 1000 Genomes Project
 - Compare results with ideal answer (Smith-Waterman)
 - Sort population by
 - Number of DNA which don't fail or timeout
 - Average Smith-Waterman score
 - Number of instrumented C++ lines executed (minimise)
 - Select top half of population.
- Mutate, crossover to give 2 children per parent.
- Repeat 200 generations



Run time errors

- During evolution 74% compile
- 6% fail at run time
 - 3% segfault
 - 2% cpulimit expired
 - 0.6% heap corruption, floating point (e.g. divide by zero) or Bowtie2 internal checks
- 68% run ok



GP Evolution Parameters

- Pop 10, 200 generations
- 50% append crossover
- 50% mutation (3 types delete, replace, insert)
- Truncation selection
- 5 test examples, reselected every generation
- ≈25 hours



Clean up evolved patch

- Allowed GP solution to grow big
- Use fixed subset (441 DNA sequences) of training data
- Remove each part of evolved patch one at time
- If makes new bowtie2 (more than a little) worse restore it else remove it permanently
- 39 changes reduced to 7
- Took just over an hour (1:08:38)



Patch

Wei ght	Mutati on	Source file	line	type	Original Code	New Code
999	replaced	bt2_io.cpp	622	for2	i < offsLenSampled	i < this->_nPat
1000	replaced	sa_rescomb	50	for2	i < satup>offs.size()	0
1000	disabled	.cpp	69	for2	j < satup>offs.size()	
100	replaced	aligner_sws	707	vh	= _mm_max_epu8(vh, vf);	vmax = vlo;
1000	deleted		766		pvFStore += 4;	
1000	replaced	se_ee _u8.cpp	772	_mm_:	store_si128(pvHStore, vh);	vh = _mm_max_epu8(vh, vf);
1000	deleted		778	ve :	= _mm_max_epu8(ve, vh);	

- Evolved patch 39 changes in 6 .cpp files
- Cleaned up 7 changes in 3 .cpp files
- 70+ times faster



Results

- Patched code (no instrument) run on 200 DNA sequences (randomly chosen from same scanner but different people)
- Runtime 4 hours v. 12.2 days
- Quality of output
 - 89% identical
 - 9% output better (higher mean Smith-Waterman score). Median improvement 0.1
 - -0.5% same
 - 1.5% worse (in 4th and 6th decimal place).



Results

- Wanted to trade-off performance v. speed:
 - On "1000 genomes" nextgen DNA sequences
 - 70+ faster on average
 - Very small improvement in Bowtie2 results

Conclusions

- Genetic programming can automatically re-engineer source code. E.g.
 - hash algorithm
 - Random numbers which take less power, etc.
 - mini-SAT
- fix bugs (>10⁶ lines of code, 16 programs)
- create new code in a new environment (graphics card) for existing program,gzipwcci 110
- speed up GPU image processing
- speed up 50000 lines of code

IEEE TEC





GECCO 2014, Vancouver 12-16 July

Abstract submission: **January 15, 2014**

Full papers: January 29, 2014

http://www.sigevo.org/gecco-2014

END

http://www.cs.ucl.ac.uk/staff/W.Langdon/

http://www.epsrc.ac.uk/ EPSRC



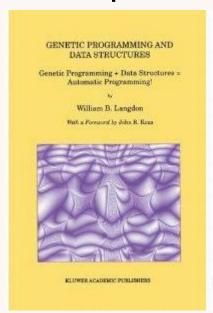


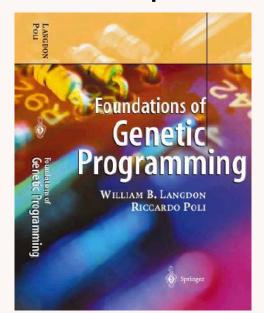
Genetic Improvement Programming

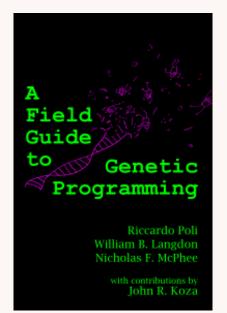


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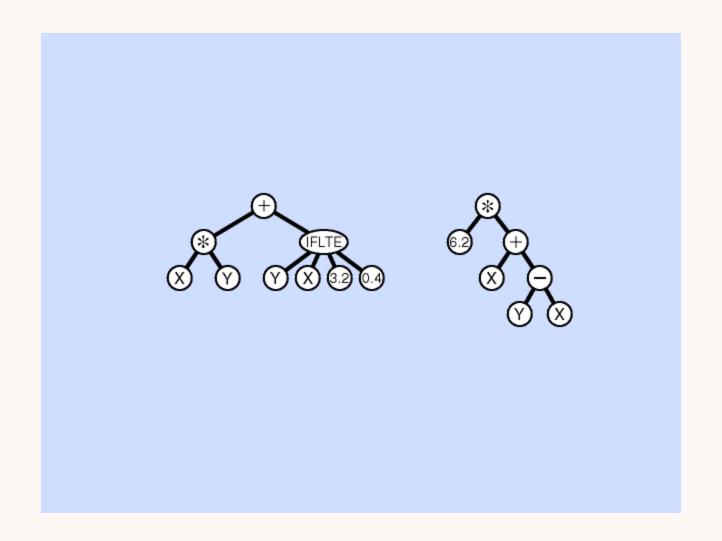




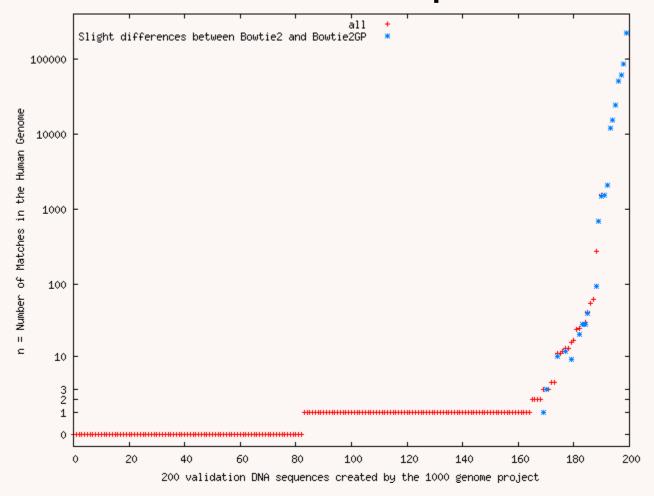




Creating new programs - Crossover



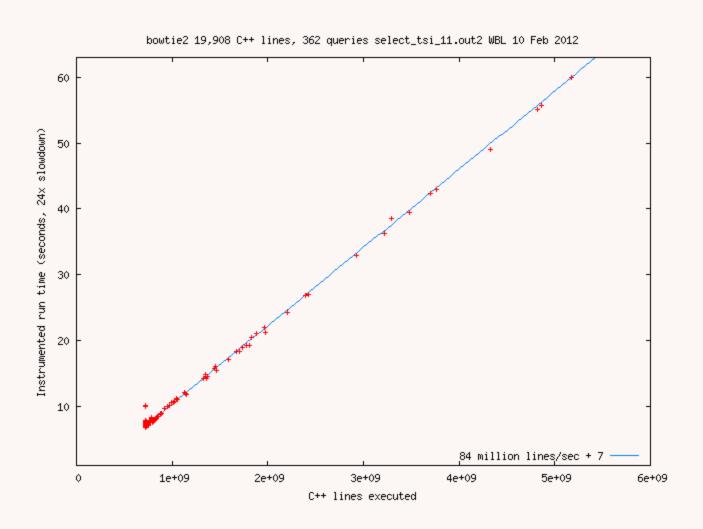
Where does Bowtie2^{GP} improvement arise



Mostly identical. Improvement with DNA which makes Bowtie2 work hard. NB nonlinear Y-scale



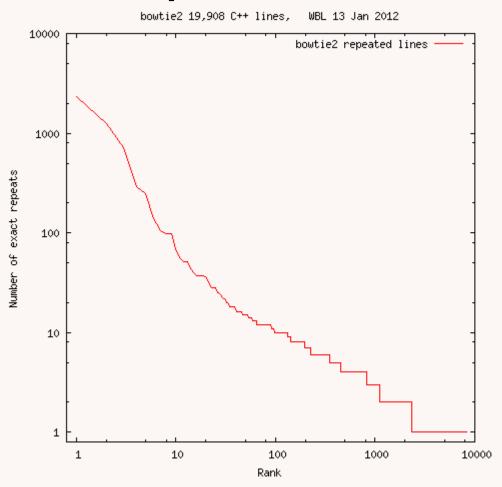
Instrumented Bowtie2



counter increments added to instrument Bowtie2



Zipf's Law

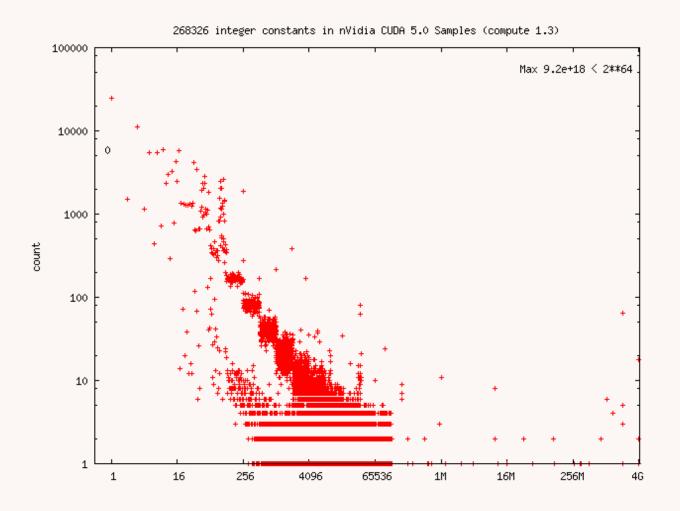


Distribution of exactly repeated Bowtie2 C++ lines of code after macro expansion, follows Zipf's law, which predicts straight line with slope -1.

44

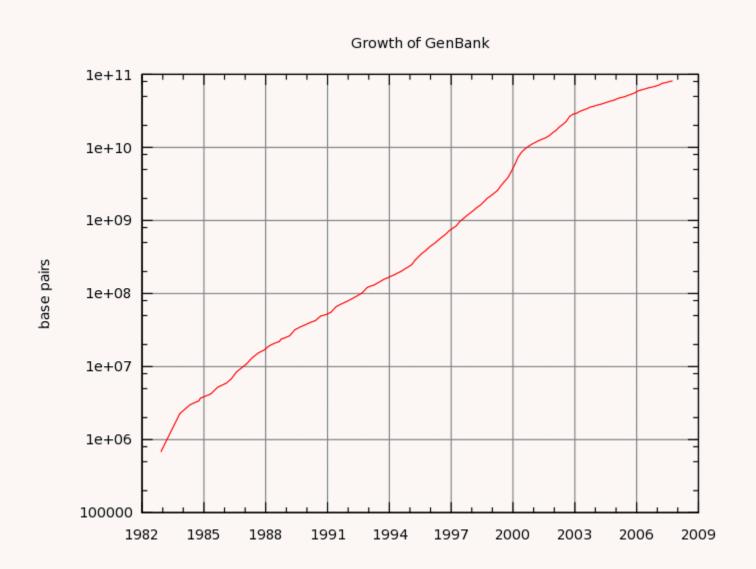


What my favourite number?





"Moore's Law" in Sequences



The Genetic Programming Bibliography

http://www.cs.bham.ac.uk/~wbl/biblio/

9018 references and 8614 online publications



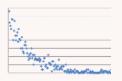
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