PPar Lunch 2014/12/10

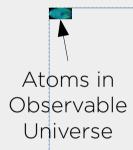
Iterative Compilation 1.01

Compiler optimisations

- •Modern Compilers:
 - -100s of possible code transformations
- •GCC 4.8.1:
 - -205 optimisation options
 - -138 parameters affecting optimisation

Compiler optimisations

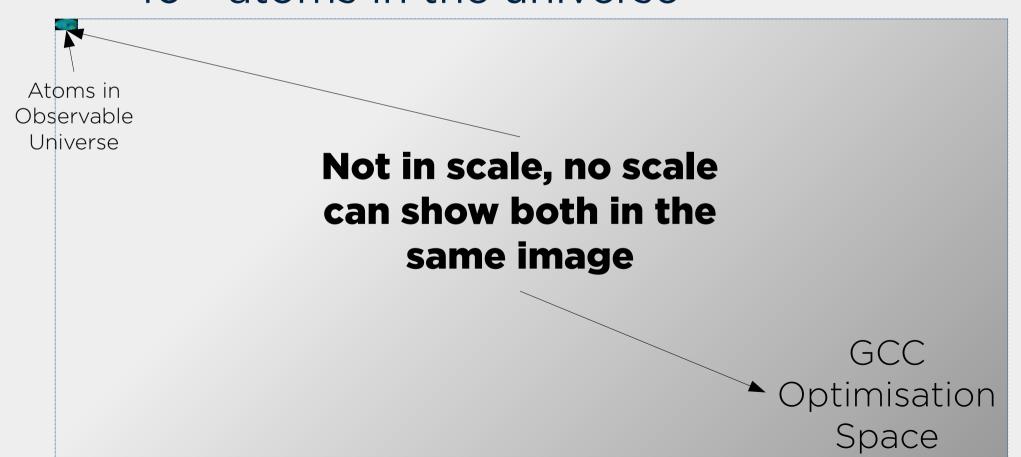
- -~10³⁴⁰ combinations!
- -~1082 atoms in the universe



GCC Optimisation Space

Compiler optimisations

- -~10³⁴⁰ combinations!
- -~1082 atoms in the universe





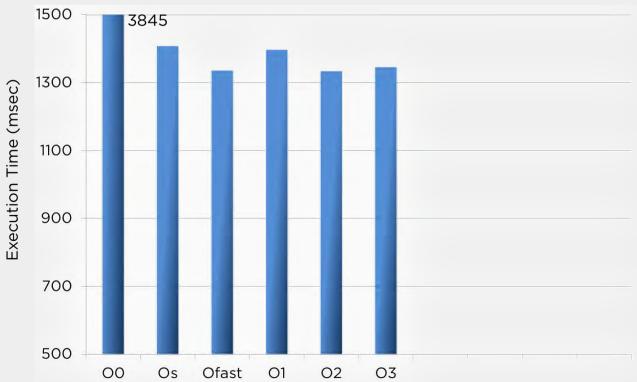
How do we choose optimisations?

How do we choose optimisations?

- Try each one on its own
 - -Transformations affect each other
- Use an expert to find the best combo
 - -Experts are few
 - -One optimisation set doesn't fit all
- Use analytical models
 - -Could help, but far from perfect
 - -Architecture dependent

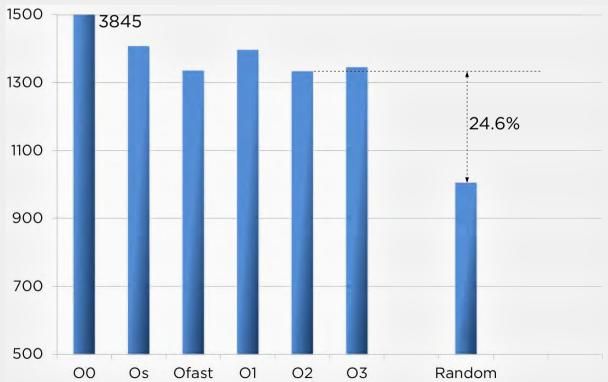
Choosing options - Example

- *Usual approach: Let's pretend -O2 is the optimal choice
 - -Enables transformations which usually improve performance
 - -But not the optimal set of transformations for each program
- •libquantum benchmark:



Choosing options - Example

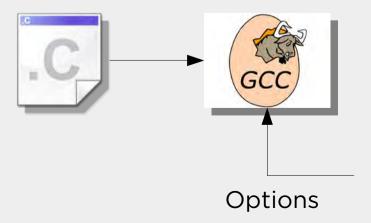
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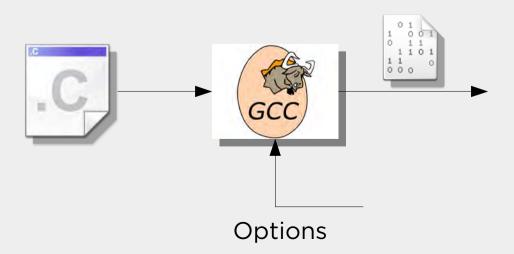
Choosing options

- Performance is lost because of inefficient optimisation strategies
- •How do we harvest all the potential performance?

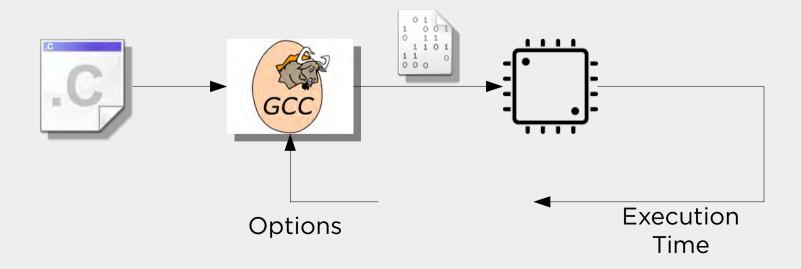
Choose options



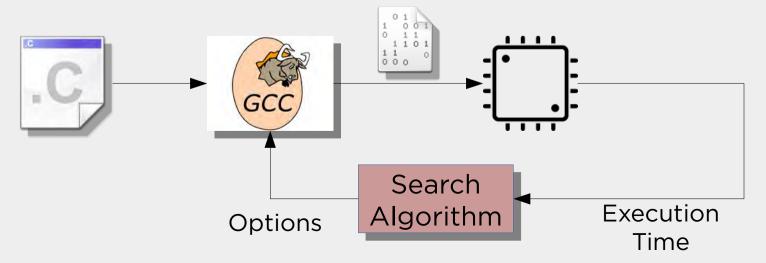
- Choose options
- Build the executables



- Choose options
- Build the executables
- Get their runtime



- Choose options
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- Get their runtime
- Based on the results, choose the next round of optimisation options and repeat

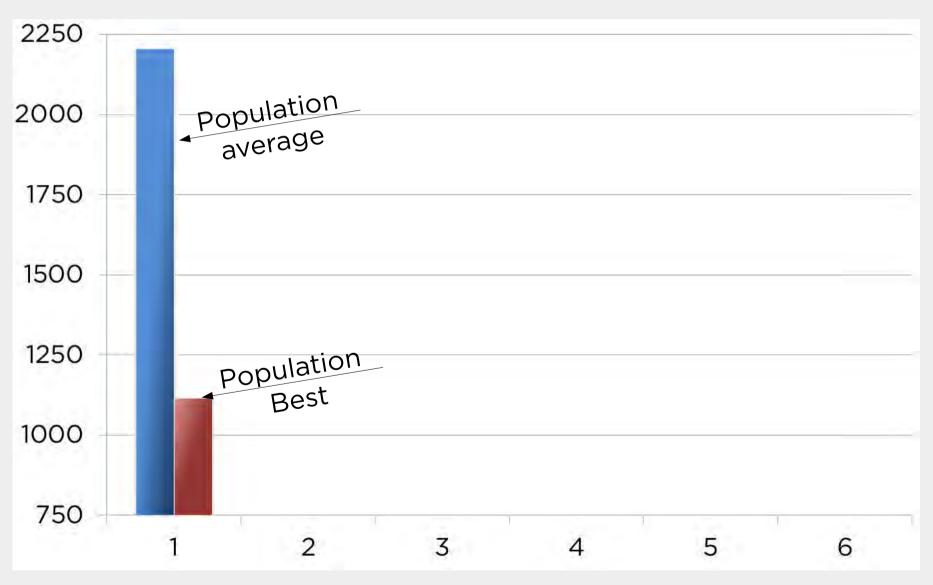


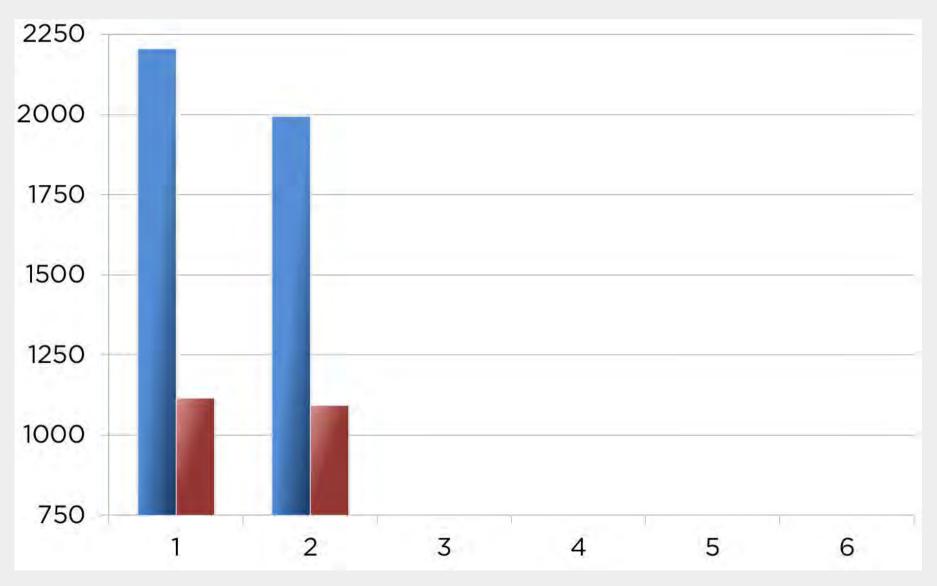
Search Algorithms

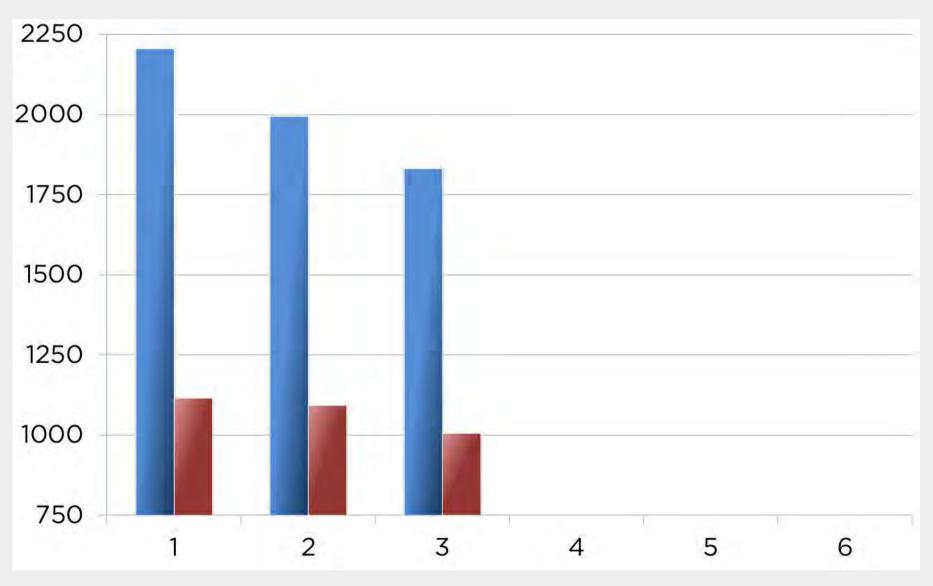
- Search algorithm choice → critical parameter
 - -Genetic Algorithm, Simulated Annealing, Grid Search, Window Search, etc.
- •General Idea:
 - -Find high performing areas of the space
 - Improve your knowledge of these areas

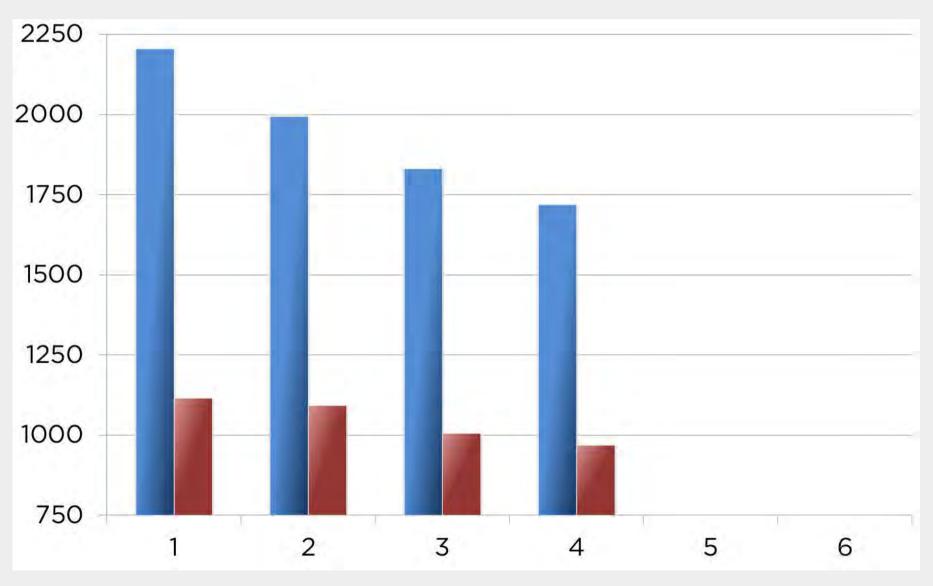


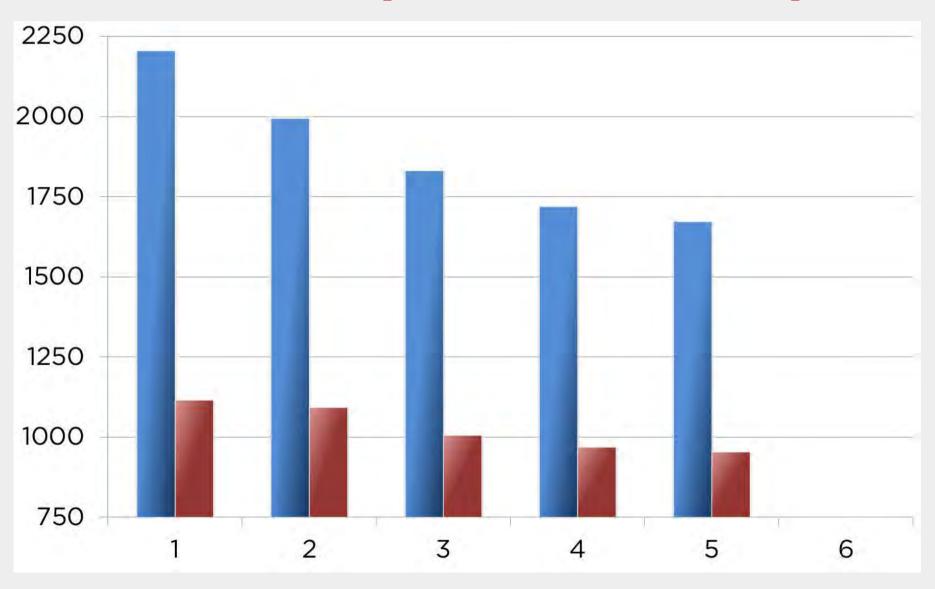
•libquantum + Genetic Algorithm

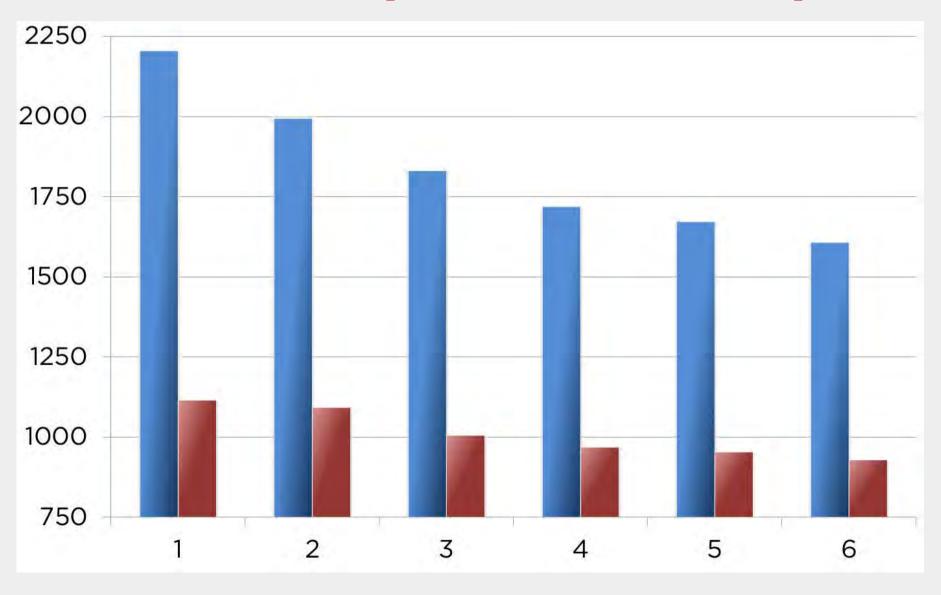


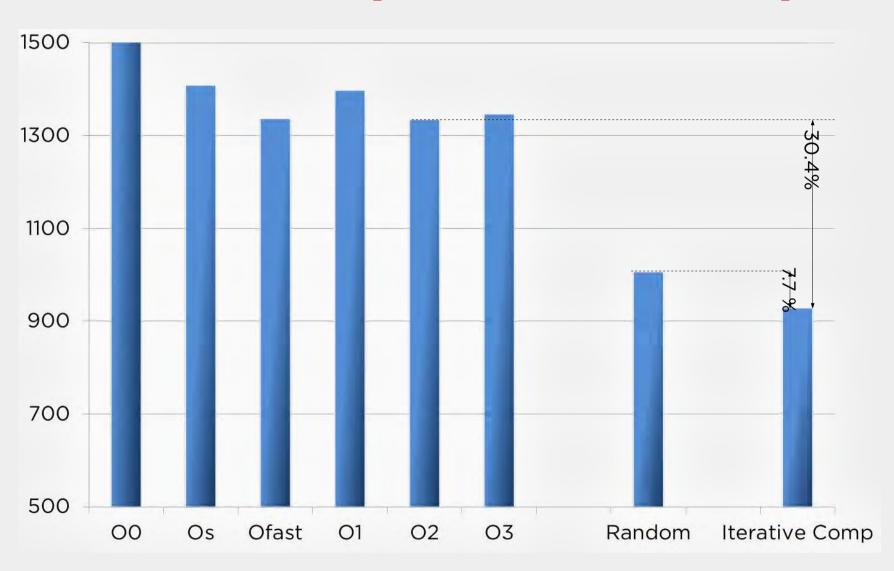












Caveats

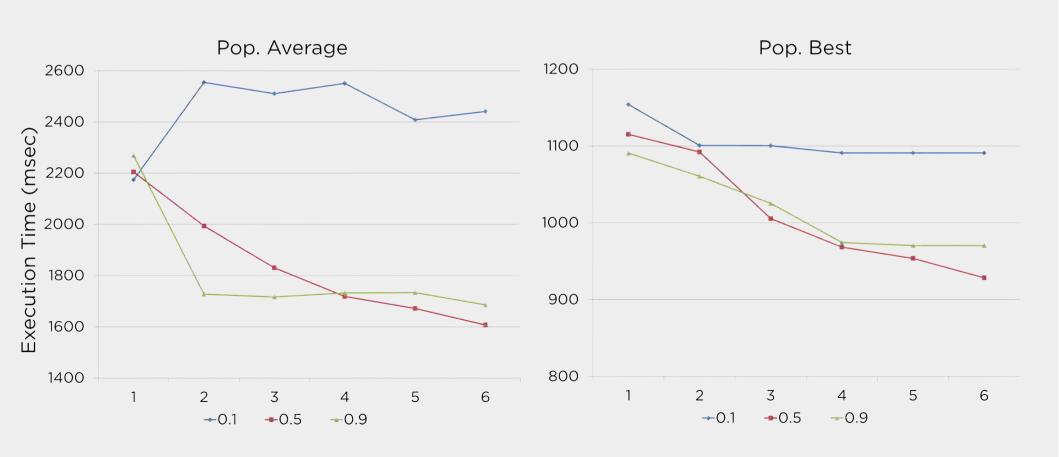
- Local maxima
- Noise
- Compiler errors

Local Maxima

- Wide & shallow vs narrow & deep search
 - -Wide and shallow search not bothered by local maxima, but slow
 - Narrow and deep search focuses on local maxima, but might miss the optimal
- Algorithms and their parameters chosen to control this trade-off

Local Maxima - Example

- GA with tournament selection
 - -Various tournament selection probabilities



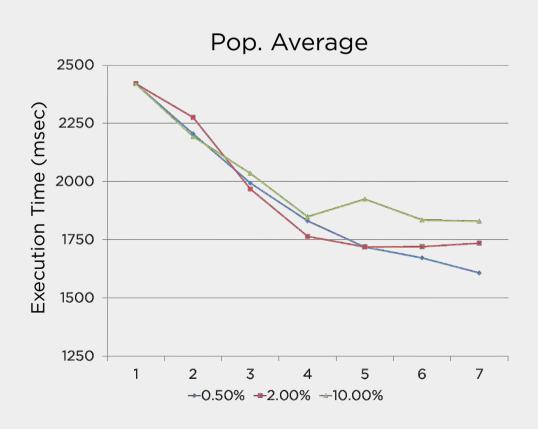


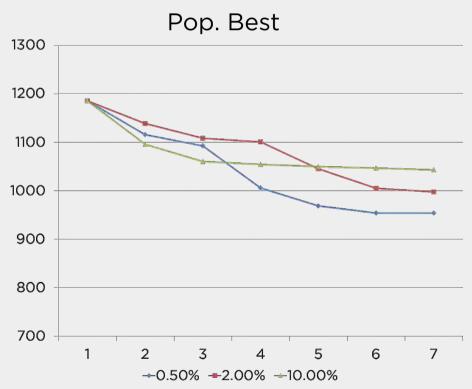
 Noise makes iterative compilation less efficient

- Noise makes iterative compilation less efficient
- If noise comparable to the performance variation in the population
 - → search algorithm becomes random

Noise - Example

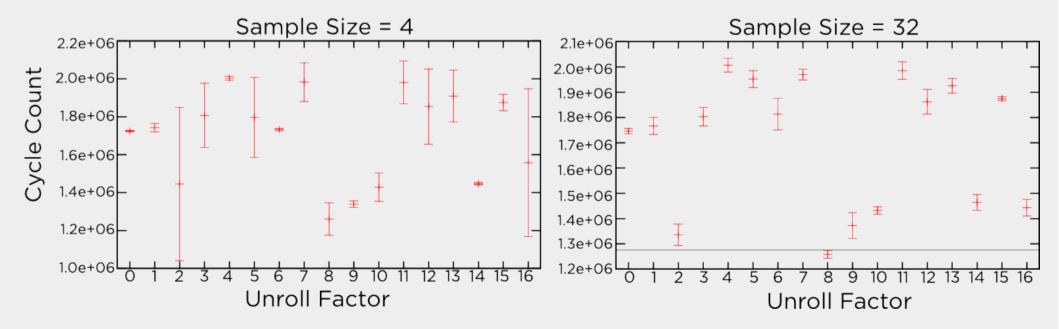
•libquantum + gaussian noise



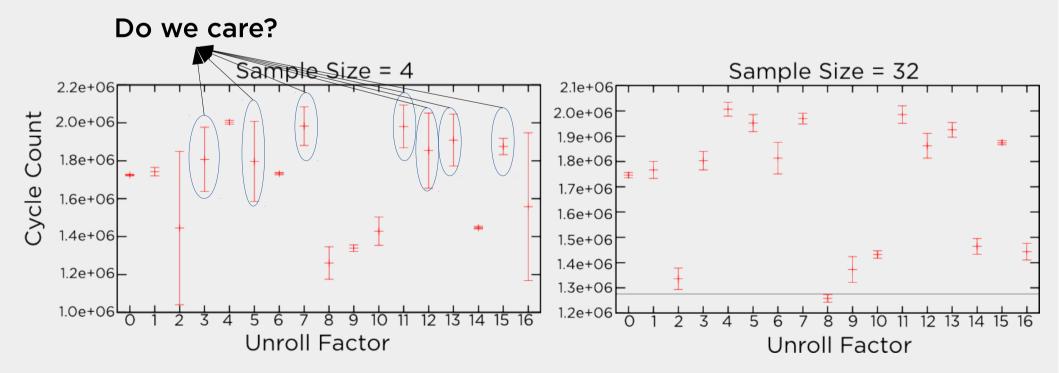


- Multiple observations to establish with some certainty the runtime of an executable
 - → Higher evaluation overhead
 - → slow

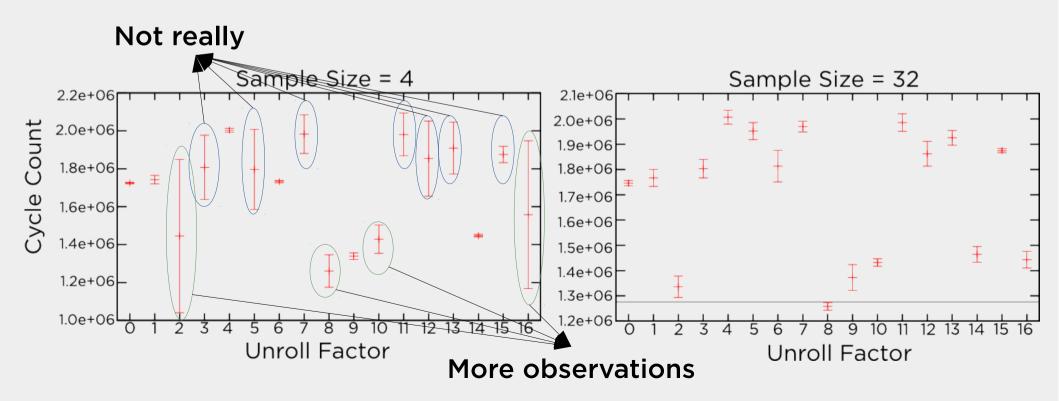
Do we need high certainty?



Do we need high certainty?



Do we need high certainty?



Compiler Errors

- By randomly combining optimisations we enter an uncharted area
- •Some combinations crash the compiler:
 - They don't work together and we know it
 - They don't work together and we don't know it
 - They trigger unknown bugs in the compiler code
- Crashes affect the speed of iterative compilation
 - Time wasted on unsuccessful builds
 - Benefit if we find problematic combinations early

Compiler Errors

- Others affect the program correctness
 - -Due to bugs or unsafe transformations
- •How do we handle them?
 - -Fully deterministic behaviour
 - Output sensitive on changes in the control flow
 - Validation of the output after each execution

Improving iterative compilation

- Focus searching on areas predicted to be near optimal
 - -Human expert input
 - -Analytical models
- Pruning of irrelevant/harmful transformations
- Reduce evaluation overhead
 - Iterative compilation on the function call level





Iterative compilation - Drawbacks



Iterative compilation - Drawbacks

- Takes too long
 - -Hours to month for the search to terminate
- Search parameters might need to be hand tuned
- Still, it's worth the effort for:
 - -embedded systems
 - **-**libraries
 - -and others

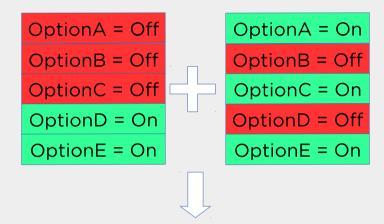
Beyond iterative compilation

- Most of the cost → evaluation
 - -1000s of evaluations
 - -To the garbage bin when we complete the search
- •Can we somehow reuse the evaluation results for different programs?





Genetic Algorithm - Example



OptionA = Off

OptionB = Off

OptionC = On

OptionD = On

OptionE = On

OptionA = On

OptionB = Off

OptionC = Off

OptionD = On

OptionE = On

OptionA = Off

OptionB = Off

OptionC = Off

OptionD = Off

OptionE = On

OptionA = On

OptionB = Off

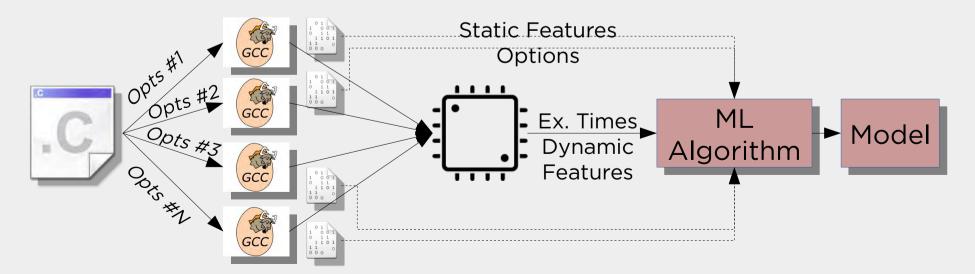
OptionC = On

OptionD = On

OptionE = On

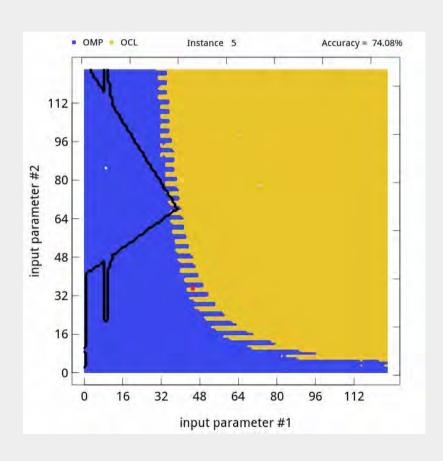
ML-based optimising compiler

- Choose options
- Build
- Evaluate
- Train



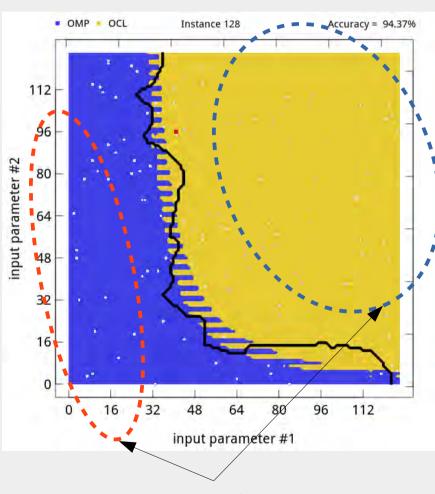
Example

- Best device (cpu or gpu) for an app based on the input size
 - -Hotspot benchmark, 2d input
 - Random input sizes evaluated
 - -Model built with the nearest neighbour algorithm



Example

 Most points offer little to no information → don't improve the model

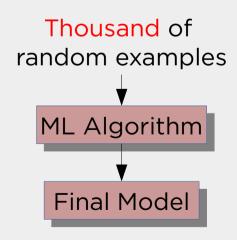


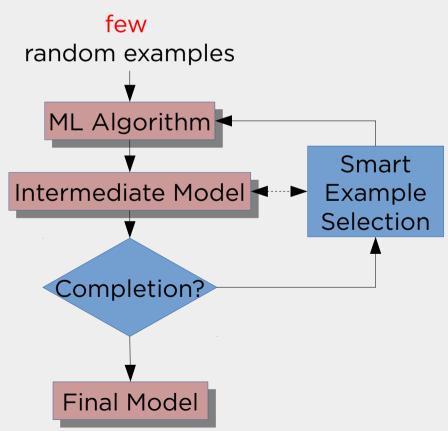
Waste of time

Active Learning

Random training points selection wasteful

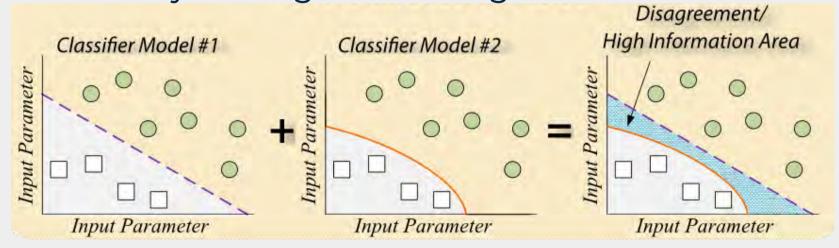
Active Learning



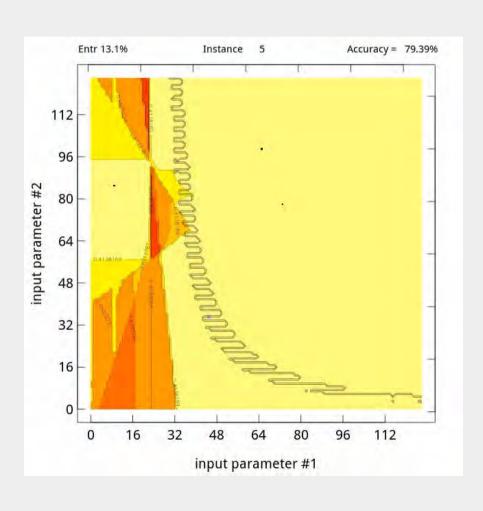


Active Learning

- Best points → High information points
 - -close to the cpu-gpu border
 - -max uncertainty about the best device
- •Quantifying uncertainty:
 - -Multiple ML algorithms trained with the same points
 - -Uncertainty == degree of disagreement



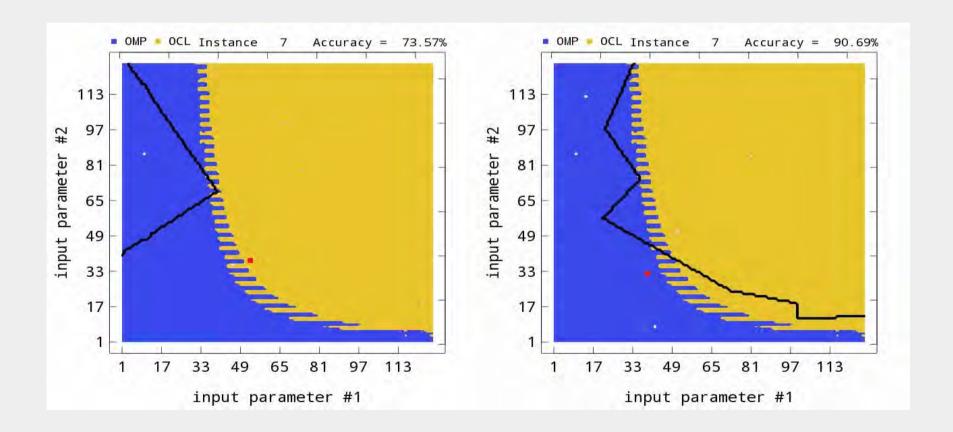
Disagreement



Comparison

Random

Active Learning



130 training points for 95% accuracy

85 training points for 95% accuracy





Iterative compilation

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