# Generating optimal HPC code with ML 

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## Outline

Intro

What is code generation? Not ChatGPT!

Representation? + HPC $=$ Polyhedral

Democratise Polyhedral: a polyhedral mini-tutorial

Current status: Tadashi

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## Motivation/Overview



## What is code generations?

The ultimate is goal: "Hey AI, optimise this code!"

- Source to source transformations
- Targeting high-level optimisations.
- Here High-level optimisations are code transformations which exploit deeper insight, an overview of the overall structure/context of the application
- This is in contrast to low-level, local transformations performed by compilers.
- Something with practical use/impact.

Fundamental requirement: the transformations need correct/legal.

## What is not code generation? (at least in this context)

Code generations (by ML) is very popular:

- ChatGPT, Co-pilot, generative models
- Deepmind's "new" sort algorithm ${ }^{1}$ and "new" matrix multiplication ${ }^{2}$
- NLP: code from human languages/commit messages

Code generation in general:

- Compilers: compiler pass

[^0]
## The (potential) problem with LLMs/generative models

Deepmind ${ }^{1}$ found algorithms using unittests, however tests don't guarantee correctness/legality.

- Primary purpose of testing is to check human code.
- Writing tests is hard, especially ones that ensure full coverage.
- Not universal: each program needs new unittests.
- Writing an AI to write unittests is just moving the goalpost (how do we know tests writen by AI ensure correctness/legality).


## Example of bad unittests

## Original

```
double gold(double input[N]) {
    double result = 0;
    for (int i = 0; i < N; i++)
        result += input[i];
    return result;
}
```


## Unittest

```
void unittest(int kpass) {
    srand((unsigned int)time(NULL));
    double input[N];
    for (int k = 0; k < kpass; k++) {
        for (int i = 0; i < N; i++)
            input[i] = (double)rand();
        compare(input);
    }
}
```

Equal? yes; delta: 0.00000000000000000000 ; gold: 2243918836.000000 ; cgpt: 2243918836.000000 ;
Equal? yes; delta: 0.00000000000000000000 ; gold: 4117298770.000000 ; cgpt: 4117298770.000000;
Equal? yes; delta: 0.0000000000000000000 ; gold: 3835775724.000000 ; cgpt: 3835775724.000000;
And now for some tricky input:
Equal? no ; delta: 0.00000005960464477539 ; gold: 400000000.000000 ; cgpt: 400000000.000000 ;

## Transformed

```
double cgpt(double input[N]) {
    double result = 0;
    for (int i = N - 1; i >= 0; i--)
        result += input[i];
    return result;
}
```


## Main

```
int main(int argc, char *argv[]) {
    unittest(10);
    double tricky_input[] = {400000000, 9e-8, 9e-8};
    printf("And now for some tricky input:\n");
    compare(tricky_input);
    return 0;
}
```


## Representation

One of the first questions we have was: When training the ML model, which representation(s) do we use?

Representations at different compiler passes:

1. Source code
2. Abstract Syntax Tree (AST)
3. Intermediate Represation(s) (IR), e.g. LLVM IR
4. Assembly code
5. Binary code

Other representations:

1. Graphical representations ${ }^{3}$ (call flow data flow graph)
2. Polyhedral model
[^1]
## HPC codes, just the right ratio of difficult

The next question: How to constrain the problemspace, to make it more feasible while still keeping it relevant/impactful?

We target HPC/scientific codes (e.g. stencils, simulations) because:

- The plethora of research papers describing optimisations of HPC codes is evidence that this is not a solved problem.
- HPC codes usually contain deep and complex nested loops, but each loop separately is regular (regular memory accesses and boundaries).
- We have experience with optimising such codes.


## Polyhedral model

## Why polyhedral? "Best bang for the buck."

- Reasonable restrictions.
- Mathematically provable correctness/legality.
- Compact way to express optimisation opportunities (e.g. parallelism)
- Compact way to express big transformations (e.g. schedule of the tile)


## Reasonable restrictions

SCoP/SANA ${ }^{4}$ : Most is true for HPC codes

- Static control: control does not depend on input data
- Affine: all relevant expressions are (quasi-)affine
- No Aliasing: essentially no pointer manipulations

These restrictions can be relaxed if care is taken.

[^2]
## Working example

Dpendecy in the outer loop, inner loop can be parallel:

$$
\begin{aligned}
& \text { for(int } i=1 ; i<N ; i++) \\
& \quad \text { for }(\text { int } j=0 ; j<M ; j++) \\
& \text { S1: } a[i][j]+=a[i-1][j] ;
\end{aligned}
$$

Components of polyhedral compilation

- SCoP extraction
- Dependency analysis
- Find a schedule $\theta$
- Legality check
- Generate the new source code


## Polyhedral basics

Everything can be represented as a matrix

- Statements: $S_{1}\left(S_{1}\right.$ is a label). S1: a[i][j] += a[i-1][j];
- Statement instances $S_{1}(i, j)$ ( $i, j$ are symbols for integer variables)
- Domain of $S_{1}:\left\{S_{1}(i, j): 1 \leq i \leq N-1, \quad 0 \leq j \leq M-1\right\}$ ( $N$ is a symbolic constant, unknown but not changing)
- Dependency graph: $e_{1}: S_{1}\left(i_{s}, j_{s}\right) \rightarrow S_{1}\left(i_{t}, j_{t}\right)$ (between statement instances)
- Notation: $\mathrm{s}=$ source (before), $\mathrm{t}=$ target (after)
- Dependency polyhedron: $P_{e}=\left\{S_{1}\left(i_{s}, j_{s}, i_{t}, j_{t}\right): i_{s}=i_{t}-1, \quad j_{s}=j_{t}\right\}$


## Dependency check

Original: $\theta_{0}: S_{1}(i, j) \rightarrow(i, j)$

- Dependencies are maps between event instances: $S_{1}(i-1, j) \rightarrow S_{1}(i, j)$
- Schedules are maps from statement instances to (multidimensional) time

Apply the schedule to the range and domain

- Dependency: $S_{1}(i-1, j) \rightarrow S_{1}(i, j)$
- Map to time: $(i-1, j) \prec(i, j)(\prec$ is the lexicographic order) or
- $(i, j)-(i-1, j)=(1,0) \succ 0$ OK!
- $\theta(\vec{s}) \prec \theta(\vec{t})$ for the dependency $\vec{s} \rightarrow \vec{t}$
- $\delta(i, j) \succ 0$ where $\delta(i, j)=\theta(i, j)-\theta(i-1, j)$


## Expressing Transformations

Swap loops $\theta_{1}: S_{1}(i, j) \rightarrow(j, i)$

- Check: $(j, i)-(j, i-1)=(0,1) \succ 0$ OK!
- You can start get $\theta_{1}$ from scratch, but you can also modify $\theta_{0}$ : in this case $\theta_{1}=T \circ \theta_{0}$ where $T=(i, j) \mapsto(j, i) . \theta_{1}: S_{1}(i, j) \xrightarrow{\theta_{0}}(i, j) \xrightarrow{T}(j, i)$
- The zero in $\delta=(0,1)$ we can parallelise the $j$ loop

Reverse j $\theta_{2}: S_{1}(i, j) \rightarrow(i,-j)$

- Check: $(i,-j)-(i-1,-j)=(1,0) \succ 0$ OK!

Reverse i $\theta_{3}: S_{1}(i, j) \rightarrow(-i, j)$

- Check: $(-i, j)-(-(i-1), j)=(-1,0) \nsucc 0$ ILLEGAL!


## More transformations

Diagonal from $(0,0) \theta_{4}: S_{1}(i, j) \rightarrow(i+j, j)$ :

- Check: $(i+j, j)-(i-1+j, j)=(1,0)$ : OK!

Alternative diagonal from $(0,0) \theta_{5}: S_{1}(i, j) \rightarrow(i+j, i)$


- Check: $(i+j, i)-(i-1+j, i-1)=(1,1):$ OK!

Tiling: $\theta(i, j)=(\lfloor i / T\rfloor,\lfloor j / T\rfloor, i \bmod T, j \bmod T)$

- $(\lfloor i / T\rfloor,\lfloor j / T\rfloor, i \bmod T, j \bmod T)-(\lfloor(i-1) / T\rfloor,\lfloor j / T\rfloor,(i-$ 1) $\bmod T, j \bmod T$ )
- The delta: $\left(q_{i}, 0, r_{i}, 0\right)$ where $q_{i}=\lfloor i / T\rfloor-\lfloor(i-1) / T\rfloor$,
- here $q_{i}=1$ if $i \mid T$ and $q_{i}=0$ when when $i \nmid T$
- $r_{i}=1-q_{i} T$ which is $1-T<0$ if $i \mid T$
- when $i \mid T:(1,0,1-T, 0) \succ 0$; when $i \nmid T:(0,0,1,0) \succ 0$ : OK!


## Tools

## A slide from Verdoolaege, "Polyhedral compilation without polyhedra".

isl and Related Libraries and Tools

isl: manipulates parametric affine sets and relations
barvinok: counts elements in parametric affine sets and relations
pet: extracts polyhedral model from clang AST
PPCG: Polyhedral Parallel Code Generator
iscc: interactive calculator
isa: prototype tool set including derivation of process networks and equivalence checker

## Tadashi

## Ultimate goal: legality check

- Ask <random LLM/generative model> to optimise your code, and have a tool to check the legality of the output the model produced!
- Very difficult: which original statement corresponds to which transformed statement?

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- Uses polyhedral.
- Checks the legal of any schedule.
- Quite easy to do with the ISL library.
- Generates ${ }^{5}$ the transformed code (if the transformation is legal).

[^3]
## Restrictions and relaxations

## The restrictions

1. Polyhedral is oblivious to the statements
2. Polyhedral is oblivious to the hardware
3. Bending the SANA/SCoP rules

And how to bend them

1. More involved data flow analysis
2. The $\delta$ encodes info about parallelism and data locality

- Transformations in and after polyhedral

3. Approximations and/or pw_qpolynomial

## A framework to automate the process




[^0]:    ${ }^{1}$ Mankowitz et al, Faster sorting algorithms discovered using deep reinforcement learning
    ${ }^{2}$ Fawzi et al, Discovering faster matrix multiplication algorithms with reinforcement learning $\bar{\equiv}$

[^1]:    ${ }^{3}$ Cummins at al, ProGraML: Graph-based Deep Learning for Program Optimization and Analysis

[^2]:    ${ }^{4}$ Verdoolaege, Polyhedral compilation without polyhedra

[^3]:    ${ }^{5}$ work in progress.

