Machine Learning-Driven Program Transformation to Increase Performance in Heterogeneous Architectures
S. Tamarit, G. Vigueras, M. Carro, J. Mariño

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## Context and Motivation.

## Heterogeneous architectures

- Different computing elements.
- Each of them better suited for a type of computation $\Rightarrow$ high-performance.
- But programming paradigms differ in each component.
- Increased complexity of development, maintenance.
- More bugs.
- Less widespread.
- Programming heterogeneous architectures restricted to a few experts.
- Hinders widespread adoption.


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## Scientific code

- Performance is a must.
- Non-trivial algorithms, code very optimized for target architecture.
- Many existing algorithms, implementations.
- Deal with them in a cost-effective, safe way $\Rightarrow$ mechanize.


## Goal.

Develop a framework for sound, semantics-based program transformation of scientific code in order to improve non-functional characteristics.

Transformation example: from $\mathbf{c}=a \mathbf{v}+b \mathbf{v}$ to $\mathbf{c}=(a+b) \mathbf{v}$.

```
        IṄİİİL̆ COODE
float c[N], v[N], a, b;
for (int i = 0; i < N; i++)
    c[i] = a * v[i];
for (int i = 0; i < N;i++)
    c[i] += b * v[i];
```

Fiñà CöD̈è

```
float k = a + b;
for(i = 0; i < N; i++)
    c[i] = k * v[i];
```


## Step by Step.



## Goal (Cont.).

## Capture and respect common properties

```
INITIAL CODE
    FinAl CODE
Complex c[N], v[N], a, b; Complex c[N], v[N], a, b;
for (int i = 0; i < N; i++) \ Complex k;
        cmp_mult(v[i], a, c[i]);
Complex aux;
for (int i = 0; i < N;i++) {
        cmp_mult(b, v[i], aux);
        cmp_add(aux, c[i], c[i]);
    }
```


## Points to Address.

- Sound.
- Respect functional properties.
- Sound.
- Respect functional properties.
- Extensible.
- Different architectures.
- Different domains.
- Sound.
- Respect functional properties.
- Extensible.
- Different architectures.
- Different domains.
- Targeted.
- Improve (certain) non-functional properties (which depend on architecture, aims).


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Search among applicable rules

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## Points to Address.

## How to...

- Sound.
- Respect functional properties.


## Use syntactic and semantic

 propertiesinfer / capture these properties?

Use (sets of) rules defining transformations

- Extensible.
- Different architectures.
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## How to...

- Sound.
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infer / capture these properties?

Use (sets of) rules defining transformations
. . . express and apply these transformations in an extensible way?

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## How to...

- Sound.
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Use syntactic and semantic properties
infer / capture these properties?

Use (sets of) rules defining transformations
. . . express and apply these transformations in an extensible way?

- Targeted.
- Improve (certain) non-functional properties Search Select applicable rule (which depend on architecture, aims). identify the right rule(s)?


## Rules: Extensible Program Transformations

## Example: Undo Distributive Property.

| ' for (int i=0;i<N;i++) | for (int i=0;i<N;i++) |
| :---: | :---: |
| $\mathrm{c}[\mathrm{i}]=\mathrm{a} * \mathrm{v}[\mathrm{i}]+\mathrm{b} * \mathrm{v}$ [i]; | $\mathrm{c}[\mathrm{i}]=(\mathrm{a}+\mathrm{b}) * \mathrm{v}[\mathrm{i}]$; |

## Transformation rule:

```
undo_distributive {
    pattern: {
        (cexpr(e1) * cexpr(e2)) + (cexpr(e1) * cexpr(e3));
    }
    condition: {
    pure(cexpr(e1));
    pure(cexpr(e2));
    pure(cexpr(e3));
    }
    generate: {
        cexpr(e1) * (cexpr(e2) + cexpr(e3));
    }
}
```


## Example: Undo Distributive Property.

| ' for (int i=0;i<N;i++) | ' for (int i=0;i<N; i++) |
| :---: | :---: |
| $\mathrm{c}[\mathrm{i}]=\mathrm{a} * \mathrm{v}[\mathrm{i}]+\mathrm{b} * \mathrm{v}[\mathrm{i}]$; | । $\mathrm{c}[\mathrm{i}]=(\mathrm{a}+\mathrm{b}) * \mathrm{v}[\mathrm{i}]$; |

Transformation rule:

```
undo_distributive {
    pattern: {
        (cexpr(e1) * cexpr(e2)) + (cexpr(e1) * cexpr(e3)); Syntactical pattern
    }
    condition: {
        pure(cexpr(e1));
        pure(cexpr(e2));
        pure(cexpr(e3));
    }
    generate: {
        cexpr(e1) * (cexpr(e2) + cexpr(e3));
    }
}
```


## Example: Undo Distributive Property.

| ' for (int i=0;i<N;i++) | ' for (int i=0;i<N; i++) |
| :---: | :---: |
| $\mathrm{c}[\mathrm{i}]=\mathrm{a} * \mathrm{v}[\mathrm{i}]+\mathrm{b} * \mathrm{v}[\mathrm{i}]$; | । $\mathrm{c}[\mathrm{i}]=(\mathrm{a}+\mathrm{b}) * \mathrm{v}[\mathrm{i}]$; |

Transformation rule:

```
undo_distributive {
    pattern: {
        [(expr(e1)] * cexpr(e2)) + (cexpr(e1) * cexpr(e3)); Syntactical pattern
    }
    condition: {
        pure(cexpr(e1));
        pure(cexpr(e2));
        pure(cexpr(e3));
    }
    generate: {
        cexpr(e1) * (cexpr(e2) + cexpr(e3));
    }
}
```


## Example: Undo Distributive Property.

| for (int i=0;i<N;i++) | ' for (int i=0;i<N; i++) |
| :---: | :---: |
| $\mathrm{c}[\mathrm{i}]=\mathrm{a} * \mathrm{v}[\mathrm{i}]+\mathrm{b} * \mathrm{v}[\mathrm{i}]$; | $\mathrm{c}[\mathrm{i}]=(\mathrm{a}+\mathrm{b}) * \mathrm{v}[\mathrm{i}]$; |

Transformation rule:

```
undo_distributive {
    pattern: {
        (cexpr(e1) * cexpr(e2)) + (cexpr(e1) * cexpr(e3)); Syntactical pattern
    condit-on: {
    pure (cexpr(e1));
    pure (cexpr(e2));
    pure.cexpr(e3));
    }
    generate.{ (cexpr(e1)] * (cexpr(e2) + cexpr(e3)); Resulting code
    }
}
```


## Example: Invariant Code Motion.

```
for(int i=0;i<N;i++) float k = a + b;
    c[i] = (a+b) * v[i];
```

```
for(int i=0;i<N;i++)
```

for(int i=0;i<N;i++)
c[i] = k *_v[i];
c[i] = k *_v[i];

Transformation rule (simplified):

```
loop_inv_code_motion{
    pattern:{
        for (cexpr(ind) = cexpr(ini); cexpr(cond); cexpr(mod)){ Syntactical
        cexpr(e3) = cexpr(e1) * cexpr(e2); }
    }
    condition:{
        pure(cexpr(e2));
        no_reads(cexpr(e2), cexpr(ind));
        no_reads(cexpr(e2), cexpr(e3));
        no_reads_in_written(cexpr(e2), cexpr(mod));
        no_reads_in_written(cexpr(e2), cexpr(e1));
        no_reads_in_written(cexpr(e2), cexpr(e3));
    }
    generate:{
        cdecl(ctype(cexpr(e2)), cexpr(aux));
        cexpr(aux) = cexpr(e2);
        for (cexpr(ind) = cexpr(ini); cexpr(cond); cexpr(mod)){
            cexpr(e3) = cexpr(e1) * cexpr(aux);}
    }
}
```

Semantic conditions (uses predefined properties)

Resulting code

## The STML Rule Language.

```
rule_name {
    pattern: {...}
    condition: {...}
    generate: {...}
}
```

- Inspired by CML [BrownLukKelly 2005], an evolution of CTT
[BoekholdKarkowskiCorporaal 1999].
- STML rules expressed in a subset of C.
- Sections:
- pattern:
$\star$ Matches (localizes) code.
औ Meta-expressions (cexpr, cstmt, ...) substituted by actual symbols before transformation.
- condition:
* Code properties required by rule (soundness).
- generate: New code, replaces matched section.
- Not shown: (new) properties of generated code.


## STML Constructs.

For all sections

| Construct | Description |  |
| :--- | :--- | :--- |
| bin_op $\left(E_{o p}, E_{1}, E_{2}\right)$ | represents the binary operation $E_{1}$ | $E_{o p}$ |
| $E_{2}$ |  |  |
| una_op $\left(E_{o p}, E\right)$ | represents the unary operation $E_{o p}$ | $E$ |

## In the condition section

| Construct | Description |
| :--- | :--- |
| is_identity $\left(E_{o p}, E\right)$ | $E$ is identity element for $E_{o p}$ |
| no_writes $\left(E_{v},(S\|[S]\| E)\right)$ | $E_{\mathrm{v}}$ is not written onto in $(S\|[S]\| E)$ |
| no_reads $\left(E_{\mathrm{v}},(S\|[S]\| E)\right)$ | $E_{\mathrm{v}}$ is not read from in $(S\|[S]\| E)$ |
| no_rw $\left(E_{\mathrm{v}},(S\|[S]\| E)\right)$ | $E_{\mathrm{v}}$ is neither read from nor written onto in (S\| [S]|E) |
| pure $(S\|[S]\| E))$ | There are no assignments in (S\| [S]|E) |
| is_const $(E)$ | There are variables inside $E$ |
| is_block $(S)$ | $S$ is a block of statements |
| is_commutative $\left(E_{o p}\right)$ | Operation $E_{o p}$ is commutative |
| is_associative $\left(E_{o p}\right)$ | Operation $E_{o p}$ is associative |
| not $\left(E_{\text {cond }}\right)$ | $E_{\text {cond }}$ is false |

## STML Constructs (Cont.).

## In the generate section

```
Construct
subs((S| [S]|E), Ef , Et)
if_then:{E ( cond ;(S|[S]|E);}
if_then_else:{E Emond
    (S|[S]|E)
    gen_list:{[(S|[S]|E)];}
```


## Description

Replace each occurrence of $E_{f}$ in (S| [S] |E) for $E_{t}$
If $\mathrm{E}_{\text {cond }}$ is true, then generate ( $\mathrm{S}|[\mathrm{S}]| \mathrm{E}$ )
If $E_{\text {cond }}$ is true, then generate $(S|[S]| E)_{t}$, else generate $(S|[S]| E)_{e}$
Each statement/expression in [(S|[S]|E)] produces a different rule consequent

## Extracting Code Properties

## Inferring and Expressing Properties.

```
loop_inv_code_motion{
    pattern:{...}
    condition:{
        pure(cexpr(e2));
        no_reads(cexpr(e2), cexpr(ind));
        no_reads(cexpr(e2), cexpr(e3));
        no_reads_in_written(cexpr(e2), cexpr(mod));
        no_reads_in_written(cexpr(e2), cexpr(e1));
        no_reads_in_written(cexpr(e2), cexpr(e3));
    }
    generate:{...}
}
```

(1) Low-level properties, capture characteristics of imperative languages.

- Destructive assignment \& its effects.
- Aliasing.
- Memory management
(2) Obtained from:
- Automatic program analysis.
- User-provided pragmas.


## Program Analysis.

- Cetus: source-to-source C compiler written in Java.
- Extensive set of compiler passes working on a high-level IR.
- Analyses and transformations:
- Dependence analysis.
- Loop parallelizer.
- Source program in canonical form.
- Loop outlining (procedural abstraction of loops).
- Read and Write pragmas (OpenMP and STML - explained later)
- Modifications to Cetus:
- Generation of STML annotations.
- Rewriting pass to adapt input code to Cetus (C99).
- Modified Artistic License $\Longrightarrow$ change Cetus without restrictions.


## STML Pragmas.

- Not all properties always automatically inferred.
- \#pragmas in code.

```
#pragma stml writes c in {0}
for (i = 0; i < N; i++)
    c[i] = i*2;
#pragma stml writes c in {-1,0}
for (i = 1; i < N; i++) {
    c[i-1] = i;
    c[i] = c[i-1] * 2;
}
```

```
#pragma stml reads c in {-1,0,+1}
for (i = 0; i < N; i++)
    a += c[i-1]+c[i+1]-2*c[i];
```

```
#pragma stml iteration_space 0 N-1
for (i = 0; i < N; i++)
    c[i] = i*2;
```


## High-Level Annotations.

- Stemming from POLCA project
- Functional programming flavor.
- Capture algorithmic skeletons:
- Summarily capture properties of underlying code.
- Also, can help determine transformation strategies.
\#pragma polca map F v w
\#pragma polca fold F INI v e
\#pragma polca itn F INI n w
\#pragma polca zipWith F u v w
\#pragma polca scanl F INI v w

```
#pragma polca map B v c
for(int i=O;i<N;i++)
#pragma polca def B
#pragma polca input v[i]
#pragma polca output c[i]
    c[i] = a*v[i];
```


## Information from Annotations.

## What map tells us

```
#pragma polca map B v c
for(int i=0;i<N;i++)
#pragma polca def B
#pragma polca input v[i]
#pragma polca output c[i]
    c[i] = a*v[i];
```

- $v$ is input to map, $c$ is output.
- For every v[i], c [i] is produced using only v[i].
- No global variables, no dependencies across iterations.
- Computation of c[i] enclosed inside B.
$\Rightarrow$ Pragma as summary of simpler properties.
$\Rightarrow$ These are dealt with by rules in program transformation tool.


## Translating High-level Annotations to STML.

STML properties can be inferred for high-level annotations.

```
#pragma polca map B v c
for(int i=0;i<N;i++)
#pragma polca def B
#pragma polca input v[i]
#pragma polca output c[i]
    c[i] = a*v[i];
```

```
    #pragma stml reads v in {0}
    #pragma stml writes c in {0}
    #pragma stml same_length v c
    #pragma stml pure B
    #pragma stml iteration_space 0 length(v)
    #pragma stml iteration_independent
    for(int i = 0; i < N; i++)
    #pragma polca def B
    #pragma polca input v[i]
    #pragma polca output c[i]
    c[i] = a*v[i];
```


## Components of the Transformation Tool.



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## Rule Testing.

- How to ensure rules are correct?
- Formal proofs possible (and ideal), but also resource-consuming.
- A set of automatic testers has been developed to ease the checking of the rules behavior.
- Perform several tests over the resulting code of each transformation step.
- Several strategies at the moment:
- $n$ random steps.
- Random order, using each rule at most once.
- Similar to last one, backtracking to previous states that could enable non-applied rules.
- Output:
- Applied rules.
- Not applied rules:
* Because patterns or conditions not met.
* Because they were not chosen (but could be applied).
- In case of error: The failing rule and the concrete transformation step.


## Rule Selection.

- Generally, several rules can be applied at a number of code locations.
- At any point, candidate rules can be:
- applicable,
- maybe (not) applicable,
- definitely not applicable
- Selected rule(s) should improve code.


## Interactive Rule Selection

- User chooses the transformation steps to apply.
- Meld and other specific tools (e.g. POGADE) help in this process.
- Useful to refine rules, perform specific transformations, do aggressive program refactoring.
- Scalability an issue.


## Rule Selection.

- We need a way to mechanize rule selection.
- Select at each step the rule which reduces some metric $\rightarrow$ local minima
- Explore a bounded number of possible rule applications
$\star$ Too small $\rightarrow$ local minima
$\star$ Too big $\rightarrow$ exponential explosion
- Additional issues
- Inverse/involutive rules $\rightarrow$ infinite loops
- Rules that duplicate code (e.g. unfolding) $\rightarrow$ same rule/different states


## External Oracle to Select Rules

- Given code and set of applicable transformation steps, return which transformation step should be applied.
- Given code, return whether transformation can be assumed successfully finishd.


## Source-to-Source Tool Interface.



- All the communications are done using JSON.
- The s2s tool interface is also used by POGADE.
- A serialization of the AST, the applicable transformations and other relevant data is done when the s2s interfaces are used.

Machine Learning: Deciding Strategy

## Problems to be Addressed.

- Rule-based transformation $\Longrightarrow$ state-space exploration problem.
- Efficiently explore (prune) search space.
- Define a stop criteria.
- Improve quality of code.


## Possible solution: heuristics to drive the search



- Heuristics which monotonically increase some metric would disallow sequences which temporarily generate code of lesser quality.
- Heuristics should "plan for" whole sequences or at least series of steps.


## Non-Local Heuristics and Machine Learning.

- Don't come up with heuristics: likely complex and brittle.
- Synthesize (learn) them from existing, good examples.
- ML generally operates on descriptions of real world.
- Program abstractions as descriptions of actual programs.
- Should capture features of states (codes) in the search space.
- Should reflect the changes performed by actions.
- Based on code features related with:
- Control flow,
- Data layout,
- Data dependencies...
- Also on code annotations (externally provided.)
- Note: previous works apply ML to e.g. compilation.


## Program Abstractions.

Abstraction: vector of code features (to be further enriched)

| $\mathbb{N}$ | $\mathbb{N}$ | $\mathbb{N}$ | $\mathbb{B}$ | $\mathbb{B}$ | $\mathbb{N}$ | $\mathbb{B}$ | $\mathbb{N}$ | $\mathbb{B}$ | $\mathbb{N}$ | $\mathbb{N}$ | $\mathbb{N}$ | $\mathbb{N}$ | $\mathbb{N}$ | $\mathbb{N}$ |
| :--- | :--- | :--- | :--- | :--- | :--- | :--- | :--- | :--- | :--- | :--- | :--- | :--- | :--- | :--- |

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## Reinforcement Learning.

- Large area of machine learning.
- Learn from sequences of states and transformations which may contain "counterproductive" steps.
- Reinforcement learning (RL):
- Well-suited for problems with long-term (vs. short-term) rewards.
- Used by the transformation engine:
- Reduce search space by applying "good" transformation sequences.
- Handle sequences with intermediate "bad" steps (delayed reward).



## RL Basics.

- RL learns from examples.
- Sequences of stepwise changes from some initial state to some final state.
- Result: action-value matrix $Q\left(s_{t}, a_{t}\right)$
- Determines, for each state, the relative profit of applying every action.
- Action to apply: the one with highest profit.
- $Q$ matrix filled with learning model
- $\alpha$ : learning rate (new information)

- $\gamma$ : discount factor (future rewards)
$Q\left(s_{t}, a_{t}\right)= \begin{cases}Q\left(s_{t}, a_{t}\right)+\alpha \cdot\left(r_{t+1}+\gamma \cdot Q\left(s_{t+1}, a_{t+1}\right)-Q\left(s_{t}, a_{t}\right)\right) & \text { if } s_{t} \text { not final } \\ Q_{\text {init }}\left(s_{t}, a_{t}\right) & \text { otherwise }\end{cases}$


## Code, State/Action Table, Rule Selection.



## Search Space: Stopping.

- Stopping encoded in state-action table.
- Determine stop criteria for training sequences:
- Is the current code ready to be translated into the target architecture?
- If so, is the current code of quality enough?
- One possibility: at each step, check if transformed code can be used to generate correct platform-specific code.
- Involves calling translator, platform-specific compiler.
- Not practical - might take an exceedingly long time (e.g., in FPGAs).
- Can not determine / measure optimality.
- Classification methods learned to classify good states to stop transforming.
- Not a canonical form, but a shape which looks good enough.


## Learning Example.

- A transformation sequence on a simple example.
- Initial state: 2D convolution kernel.
- Final code:
* Well-suited to generate MaxJ (FPGA/DFE) code.
* Learned by a classifier as a good final state for the translation phase.
- Abstraction (vector of features) shown for each code state.
- Transformation steps provided by expert programmer.
- Code transformed in each step highlighted as:
* before transformation
* after transformation


## Learning Example (Cont.).

- Initial code

```
// ABSTR: [3, 0, 0, 0, 0, 0, 1, 0, 0, 1, 1, 3, 2, 4, 0]
int dead_rows = K / 2;
int dead_cols = K / 2;
int normal_factor = K * K;
for (r = 0; r < N - K + 1; r++) {
    for (c = 0; c < N - K + 1; c++) {
        sum = 0;
        for (i = 0; i < K; i++) {
            for (j = 0; j < K; j++) {
                sum += input_image[r+i][c+j] * kernel[i][j];
            }
        }
        output_image[r+dead_rows][c+dead_cols] = (sum / normal_factor);
    }
}
```


## Learning Example (Cont.).

- Step 1 (Rule: ArrayFlatten)

```
// ABSTR: [3, 0, 0, 0, 0, 0, 1, 0, 0, 1, 1, 2, 2, 4, 0]
int dead_rows = K / 2;
int dead_cols = K / 2;
int normal_factor = K * K;
for (r = 0; r < N - K + 1; r++) {
    for (c = 0; c < N - K + 1; c++) {
        sum = 0;
        for (i = 0; i < K; i++) {
            for (j = 0; j < K; j++) {
                sum += input_image[(r+i)*(N - K + 1) + (c+j)] * kernel[i][j];
            }
        }
        output_image[r+dead_rows][c+dead_cols] = (sum / normal_factor);
    }
}
```


## Learning Example (Cont.).

- Step 2 (Rule: ArrayFlatten)

```
// ABSTR: [3, 0, 0, 0, 0, 0, 1, 0, 0, 1, 1, 1, 2, 4, 0]
int dead_rows = K / 2;
int dead_cols = K / 2;
int normal_factor = K * K;
for (r = 0; r < N - K + 1; r++) {
    for (c = 0; c < N - K + 1; c++) {
        sum = 0;
        for (i = 0; i < K; i++) {
            for (j = 0; j < K; j++) {
                sum += input_image[(r+i)*(N - K + 1) + (c+j)] * kernel[i*K+j];
            }
        }
        output_image[r+dead_rows][c+dead_cols] = (sum / normal_factor);
    }
}
```


## Learning Example (Cont.).

- Step 3 (Rule: ArrayFlatten)

```
// ABSTR: [3, 0, 0, 0, 0, 0, 1, 0, 0, 1, 1, 0, 2, 4, 0]
int dead_rows = K / 2;
int dead_cols = K / 2;
int normal_factor = K * K;
for(r = 0; r < N - K + 1; r++) {
    for(c = 0; c < N - K + 1; c++) {
        sum = 0;
        for (i = 0; i < K; i++) {
            for (j = 0; j < K; j++) {
            sum += input_image[(r+i)*(N - K + 1) + (c+j)] * kernel[i*K+j];
            }
        }
        output_image[(r+dead_rows)*(N-K+1) + (c+dead_cols)] = (sum / normal_factor);
    }
}
```


## Learning Example (Cont.).

- Step 4 (Rule: CollapseFoorLoops)

```
// ABSTR: [2, 0, 0, 0, 0, 0, 1, 1, 0, 1, 1, 0, 2, 3, 0]
int dead_rows = K / 2;
int dead_cols = K / 2;
int normal_factor = K * K;
for (z = 0; z < (N - K + 1)*(N - K + 1); z++) {
    sum = 0;
    for (i = 0; i < K; i++) {
        for (j = 0; j < K; j++) {
            sum += input_image[((z / (N - K + 1)) +i)*(N - K + 1) + ((z % (N - K +
                        1))+j)] * kernel[i*K+j];
        }
    }
    output_image[((z / (N - K + 1))+dead_rows)*(N - K + 1) +
        ((z % (N - K + 1))+dead_cols)] = (sum / normal_factor);
}
```


## Learning Example (Cont.).

- Mapping of states $(S)$ to abstractions $(A)$

$$
\begin{array}{ll}
S_{0}=A\left(C_{0}\right): & {[3,0,0,0,0,0,1,0,0,1,1,3,2,4,0] \text { (Initial) }} \\
S_{1}=A\left(C_{1}\right): & {[3,0,0,0,0,0,1,0,0,1,1,2,2,4,0]} \\
S_{2}=A\left(C_{2}\right): & {[3,0,0,0,0,0,1,0,0,1,1,1,2,4,0]} \\
S_{3}=A\left(C_{3}\right): & {[3,0,0,0,0,0,1,0,0,1,1,0,2,4,0]} \\
S_{4}=A\left(C_{4}\right): & {[2,0,0,0,0,0,1,1,0,1,1,0,2,3,0] \text { (Goal) }}
\end{array}
$$

- Expert sequence (transition matrix)
- Initial action-value table $(Q)$

|  | $R_{0}$ | $R_{1}$ |
| :--- | :--- | :--- |
| $S_{0}$ | $S_{1}$ | $S_{0}$ |
| $S_{1}$ | $S_{2}$ | $S_{1}$ |
| $S_{2}$ | $S_{3}$ | $S_{2}$ |
| $S_{3}$ | $S_{3}$ | $S_{4}$ |
| $S_{4}$ | $S_{4}$ | $S_{4}$ |


|  | $R_{0}$ | $R_{1}$ | Best |
| :---: | :---: | :---: | :---: |
| $C_{0}$ | 1.0 | 1.0 | - |
| $C_{1}$ | 1.0 | 1.0 | - |
| $C_{2}$ | 1.0 | 1.0 | - |
| $C_{3}$ | 1.0 | 1.0 | - |
| $C_{4}$ | 1.0 | 1.0 | - |

## Learning Example (Cont.).

- Simple training: 10 iterations, 3 interactions/iteration
- Learning params: $\alpha=0.5, \gamma=0.5, r_{t+1}=1$ (final) $\mid 0$ (not final) Iteration 1 Iteration 3

|  | $R_{0}$ | $R_{1}$ | Best |
| :---: | :---: | :---: | :---: |
| $C_{0}$ | 0.9955 | 1.0 | $R_{1}$ |
| $C_{1}$ | 0.995 | 1.0 | $R_{1}$ |
| $C_{2}$ | 1.0 | 1.0 | - |
| $C_{3}$ | 1.0 | 1.0 | - |
| $C_{4}$ | 1.0 | 1.0 | - |

Iteration 5

|  | $R_{0}$ | $R_{1}$ | Best |
| :---: | :---: | :---: | :---: |
| $C_{0}$ | 0.9955 | 0.9881 | $R_{0}$ |
| $C_{1}$ | 0.9925 | 0.9955 | $R_{1}$ |
| $C_{2}$ | 1.2145 | 1.0910 | $R_{0}$ |
| $C_{3}$ | 1.0 | 1.4434 | $R_{1}$ |
| $C_{4}$ | 1.0 | 1.0 | - |


|  | $R_{0}$ | $R_{1}$ | Best |
| :---: | :---: | :---: | :---: |
| $C_{0}$ | 0.9955 | 0.9881 | $R_{0}$ |
| $C_{1}$ | 0.995 | 1.0 | $R_{1}$ |
| $C_{2}$ | 1.0 | 1.0 | - |
| $C_{3}$ | 1.0 | 1.4434 | $R_{1}$ |
| $C_{4}$ | 1.0 | 1.0 | - |

Iteration 10

|  | $R_{0}$ | $R_{1}$ | Best |
| :---: | :---: | :---: | :---: |
| $C_{0}$ | 1.1490 | 0.9881 | $R_{0}$ |
| $C_{1}$ | 1.2216 | 0.9905 | $R_{0}$ |
| $C_{2}$ | 1.5942 | 1.0910 | $R_{0}$ |
| $C_{3}$ | 0.995 | 1.9466 | $R_{1}$ |
| $C_{4}$ | 1.0 | 1.0 | - |

## Implementation Details.

- Implemented in Python using two packages
- Scikit-learn: classification methods
- Several machine learning algorithms
- Good support and ample documentation
- Widely used by the scientific community.
- PyBrain: reinforcement learning
- Modular structure in classes (environment, actions, ...)
- Extension of PyBrain classes to implement our approach


## Preliminary Evaluation.

- Using cases from POLCA and UTDSP benchmark suite as training set.
- Featuring different algorithmic patterns (image processing):
- Compress (JPG compression).
- Edge detection (convolution).
- RGB filter (image filter).
- Threshold (image filter).
- Manually identified transformation sequences leading to OpenCL
- Simple reward scheme linked to efficiency (much higher for the best code).
- Evaluation using different examples, sharing patterns with training set:
- 3D rotation.
- Image difference.
- Brightness change.
- Different aspects were satisfactorily evaluated:
- Transformation process terminates, "good" final states reached.
- RL learned different sequences which might share intermediate steps.
- Transformations can start at any point in the learned sequences


## Preliminary Evaluation.

- Results to show the non-monotonic behavior of transformation sequences




## Preliminary Evaluation.

- Evaluation of OpenCL code mechanically generated from learnt transformation sequences

SpeedUp for all use cases



## Conclusions.

- Extensible, flexible rule-based framework for program transformation.
- Interface to external tools to select transformations to be enacted.
- Using machine learning-based oracle to guide rule selection.
- Preliminary evaluation (more around the corner!) satisfactory.


## Future Steps.

- Add more (complex) transformation rules, evaluate on larger examples.
- Add more interfaces to external analysis tools.
- Dependence analysis (e.g., polytope-based analyzers).
- Reasoning over heap pointers (e.g., separation logic).
- Integrate profiling techniques $\rightarrow$ ease evaluation and give feedback.
- Increase training set, enhance code abstractions as needed.
- Use richer, non-functional measurements for RL.
- Measures of optimality to better discriminate between possible final states based on performance.
- Explore using ML to decide best platform for a piece of code
- E.g., use a single action-value table with several goals, select the best platform among the most reinforced sequences

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