### Compiling a Gesture Recognition Application for an Ultra Low-Power Architecture

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### **Classification Applications**







W







0.5 mW!

### GreenArrays: Ultra Low Power Processor



• 32 instructions

#### **Programming challenges**

- Manually partition a program across 144 cores
- Explicitly manage communication between cores
- Program in assembly-like stack-based language

How to program and compile for highly-constrained multicore processors with very small distributed memory?

**Key idea:** partition a program smartly to fit the code in tiny cores and achieve fast execution time by balancing **communication** and **code replication**.

In the context of partitioning both:

1. data & computation

2. control flow statements

# Existing Solution

[Phothilimthana et al. PLDI'14]

**Input:** imperative sequential program

**Output:** per-core assembly programs

**Compilation strategy:** similar to SPMD

- Output per-core programs have the same structure (control flow).
- Data and computation are distributed across cores.

#### Extra input:

partial constraints on data and computation partitioning

### Spatial programming model

int x[12];
for(i from 1 to 12)
 x[i] += x[i-1];

### Spatial programming model

# int[4]@{1,2,3} x[12]; for(i from 1 to 12) x[i] += x[i-1];

### **Partition Type**

pins data and operators to specific partitions (logical cores)

Similar to [Chandra et al. PPoPP'08]



### Spatial programming model

# int[4]@{1,2,3} x[12]; for(i from 1 to 12) x[i] +=@loc(x[i]) x[i-1];

### **Partition Type**

pins data and operators to specific partitions (logical cores)

Similar to [Chandra et al. PPoPP'08]



### Incomplete Annotations

int[4] x[12];
for(i from 1 to 12)
 x[i] += x[i-1];

### Incomplete Annotations

int[4]@?? x[12];
for(i from 1 to 12)
 x[i] +=@?? x[i-1];

#### Hard constraint:

Code fits in each logical core (partition).

#### **Objective:**

Minimize number of messages sent between partitions.

Program + some partition annotations

#### Partition Type Inference

#### Complete partition annotations

### Problems

#### Problem I Generated code is too large.

Cause Control flow statements partitioning strategy exploits code replication but not enough communication.

#### Problem II Slow execution time (no parallelism)

Cause Data & computations partitioning strategy does not exploit code replication.

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# Control Flow Partitioning



Given a program with a complete partitioning assignment, how to generate code for each partition?

- Which control flow constructs need to be replicated in each partition?
- Which partition is a "requestor" of a function?
- Which partitions are "actors" of a function?





#### Control Dependence Graph (CDG)



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#### Relevant control flow slice of partition 21



### SPMD Strategy



### SPMD + Actor Strategy



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Cause Data & computations partitioning strategy does not exploit code replication.

## Original: No Task Parallelism

```
fix1_t@1 classify(fix1_t@1 acc[3], fix1_t@2 model[N]) {...}
```

```
prob1 = classify(acc, model1);
prob2 = classify(acc, model2);
```



### Work Around: Manual Replication

```
fix1_t@1 classify1(fix1_t@1 acc[3], fix1_t@2 model[N]) {...}
fix1_t@5 classify2(fix1_t@5 acc[3], fix1_t@6 model[N]) {...}
prob1 = classify1(acc, model1);
prob2 = classify2(acc, model2);
```



## Solution: Automatic Replication



```
C2_classify(acc);
```

### Solution Summary

# Balance the use of **communication** and **code replication** to partition

- program control flow statements
- data and computation

### Hand Gesture Recognition



### Hand Gesture Recognition



### Implementation



- Use mixed partitioning strategy to make the application fit on GA144.
   orange = actor cores
- 2. Use **parallel module** to classify circle and flip-roll gestures in parallel.

### Result

Can we use Chlorophyll with our extensions to generate code for the gesture recognition application for GA144?

Partitioning strategy	Number of cores	Overflowed cores	Size of largest core (words)
SPMD	90	12	87
SPMD + Actor	82	0	64

Note: each core can contain up to 64 words.

Code occupies 82 out of 144 cores.

Prediction accuracy = 80-91% (similar to Wigee [Schlomer et al. 08])

### GA144 vs. MSP430

# How much energy consumption can we reduce by being able to compile for GA144?

\*per one round of accelerometer reading

Processor	Execution	Energy consumption (uJ)			
	time (ms)	Accelerometer	Computation	Total	
GA144	2.6	1.7	0.6	2.2	
MSP430	61.3	0.8	41.2	41.9	
	22x faster	10 X	more energy-e	fficient	

### Demo







### Summary

Partition a program smartly to fit the code in tiny cores and achieve fast execution time by balancing communication and code replication.







### 0.5 mW!









