# Abstract, declarative control over partitioning in parallel functional programs: experiences with Caliban

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## Introduction

Slogan:

# "Parallel programming by programmed partitioning"

- Suppose automatic parallelisation isn't good enough
- The programmer has to decide how the data and computation is distributed across the processing elements
  - $\star$  This is a software engineering problem
  - $\star$  It needs a good programming language
  - ★ We need to support abstraction and re-use

# The fully-dynamic approach

- Futures (lazy task creation etc)
- par, "sparking"
- threads, fork
- ParAlfl's "**\$on**' operator

Processes are generated one-at-a-time as the computation progresses

Higher-level operators (pipelines, farms, arrays of processes etc) can easily be programmed, but the structure is not evident to the implementation

- ★ If we can separate the partitioning from the program execution, we can exploit broader knowledge when allocating resources
- ★ Start with fully static
- ★ Necessary precursor to semi-static

# Compile-time vs run-time

Programmed partitioning depends on:

- the target machine: number of processors, communication vs computation performance, memory capacity
- the problem instance: size, shape of key data structures
- the evolution of the computation: e.g. where eddies, collisions or precipitation occur

Some information available at compile-time, some only later

#### Introducing Caliban



• This is expressed as an annotation (in fact a Haskell data structure):

Bundle [a,b] And Bundle [c,d,e] And Bundle [f] And (Arc a d) And (Arc a f)

- The **Bundle** assertions specify that
  - expressions a and b are allocated to the same processor
  - c, d and e to the same, presumably different processor
  - and **f** to a third.
  - The Arc a d assertion specifies a link in the task graph, either because a consumes d or *vice versa*.

#### The process placement rule

- The **moreover** assertion specifies the placement of named expressions onto the task graph.
- The "process placement rule" (analogous to "owner computes" in HPF) specifies where the computation of these expressions takes place.
- ★ In the absence of any annotations, every processor executes the entire program.
   Some arbitrarily-chosen processor's result expression is output.
- ★ The expression Bundle [x] asserts that x is computed on one processor only, and all non-local references to x involve communication.

# Communication

• For simplicity, we assume that placed expressions such as x are of list type, and our implementation serialises such lists by evaluating and sending each element in its entirety in turn.

Explanation:

- Caliban is based on the model of a static network of processes communicating via streams of messages.
- To support this, we require that in an annotation such as **Bundle** [x], the name x should refer to an object which can be transmitted as a stream of messages.

This raises various issues...

- Isn't it unnecessarily restrictive?
- Doesn't it interfere with the semantics?

# Compute-ahead and strictness

- Evaluation proceeds in anticipation of demand, so that the producer of a stream can operate in parallel with its consumer.
- This "compute-ahead" is restricted by the availability of buffer space in the consumer (and, of course, by availability of operands).

Streams are easy to compute ahead, because the consumer cannot choose what to demand next

# Threads: "bundling"

- In an assertion such as Bundle [a,b], two expressions are assigned to the same processor.
- We create a thread for each expression, each charged with computing elements ahead of demand and sending the values to each of the consumer processors.

In principle, these threads should be pre-emptively scheduled, so that evaluation of all the expressions allocated to a processor proceeds even if one of the threads loops or blocks.

#### Strictness and semantics

• "Bundle" assigns a thread to evaluate the elements of the specified stream in order

An annotation may stop a working program from producing all its outputs:

- Threads are evaluated in advance of demand. If the stream turns out not to be needed, may loop – but a consumer thread oughtn't be interfered with
- 2. Stream elements are evaluated in order. If one element turns out not to be needed may loop evaluating unwanted stream element, so subsequent elements are never reached
- 3. If threads are not pre-emptively scheduled, one expression may wait forever for evaluation of another expression on the same processor

## Strictness and semantics - cont'd

One more reason an annotation may stop a working program from producing all its outputs:

4 To evaluate the annotation, the compiler may have to evaluate expressions which aren't needed by the computation by itself – the compiler may not terminate

# Example: ray tracing

A simple ray-tracer can be reduced to the following Haskell program:

ray\_trace scene viewpoint

```
= map impact rays
where
rays = generate_rays viewpoint
impact ray = fold earlier impacts
where
impacts = map (hit ray) scene
```

- hit ray obj tests whether a ray strikes a given object in the scene, and if so, returns details of the impact
- earlier impact<sub>1</sub> impact<sub>2</sub> returns the first impact struck by the ray

The closest impact to the viewpoint determines the colour of the pixel in the output.

#### First approach: using a processor farm



Idea: define a function to generate this pattern:

 $\mathsf{fan} :: \mathsf{Stream} \to [\mathsf{Stream}] \to \mathsf{Placement}$ 

fan s [] = NoPlace

fan s (a:as) = (Bundle [a]) And (Arc a s) And (fan s as)

the expression fan collector [slave0, slave1, slave2, slave3] yields the annotation

Bundle [slave0] And (Arc slave0 collector) And Bundle [slave1] And (Arc slave1 collector) And Bundle [slave2] And (Arc slave2 collector) And Bundle [slave3] And (Arc slave3 collector) And NoPlace

("NoPlace" is the null annotation). "fan" is called an NFO - a network forming operator.

## Skeletons

We can define a reusable function which encapsulates the processor farm behaviour; we use the fan operator to build its annotation:

farm ::  $(a \rightarrow a) \rightarrow [a] \rightarrow [a]$ 

farm func operands

- = farmed moreover fan farmed farmed
  where
  farmed = map func operands
- The assertion says that each element of the list farmed is evaluated on a separate processor, and the results (the actual list farmed) are collected onto a single processor for output
- The assertion is evaluated by the compiler to yield an annotation which places each of the slaves on a separate processor.
- number of **operands** must be known

#### Using the parallel operator

ray\_trace scene viewpoint

= farm impact rays where

rays = generate\_rays viewpoint

impact ray = fold earlier impacts

#### where

impacts = map (hit ray) scene

This is unfolded by the compiler:

```
ray_trace scene viewpoint
= farmed moreover fan farmed farmed
where
farmed = map impact rays
rays = generate_rays viewpoint
impact ray = fold earlier impacts
where
impacts = map (hit ray) scene
```

If the viewpoint is known at compile-time, we can calculate the list of rays:

```
ray_trace scene viewpoint
= farmed moreover fan farmed farmed
where
farmed = map impact rays
rays = generate_rays viewpoint = [r_0, r_1, r_2, r_3]
....
```

Using the definition of map the compiler can construct the list of unevaluated processes:

```
ray_trace scene viewpoint
```

. . .

= farmed moreover fan farmed farmed
where
[farmed<sub>0</sub>, farmed<sub>1</sub>, farmed<sub>2</sub>, farmed<sub>3</sub>] = farmed
farmed
= [impact r<sub>0</sub>, impact r<sub>1</sub>, impact r<sub>2</sub>, impact r<sub>3</sub>]
rays = generate\_rays viewpoint = [r<sub>0</sub>, r<sub>1</sub>, r<sub>2</sub>, r<sub>3</sub>]

The compiler can expand fan farmed farmed and build the static process network.... ray\_trace scene viewpoint

= farmed moreover

Bundle [farmed<sub>0</sub>] And (Arc farmed<sub>0</sub> farmed) And Bundle [farmed<sub>1</sub>] And (Arc farmed<sub>1</sub> farmed) And Bundle [farmed<sub>2</sub>] And (Arc farmed<sub>2</sub> farmed) And Bundle [farmed<sub>3</sub>] And (Arc farmed<sub>3</sub> farmed) And NoPlace

where

```
[farmed<sub>0</sub>, farmed<sub>1</sub>, farmed<sub>2</sub>, farmed<sub>3</sub>]
```

```
= [impact r_0, impact r_1, impact r_2, impact r_3]
rays = generate_rays viewpoint = [r_1, r_2, r_3, r_4]
impact ray = fold earlier impacts
```

```
where
impacts = map (hit ray) scene
```

- You could have written the annotation above manually
- Caliban's compile-time symbolic evaluation allowed a more concise annotation

# Aggregation by bundling

- The ray tracer above assigns one processor per ray
- With many (or an unknown number of) rays, we want to allocate many rays to each processor
- It seems natural to use **Bundle**:

```
\begin{array}{l} \mbox{farm}::(a{\rightarrow}a){\rightarrow}[a]{\rightarrow}[a] \\ \mbox{farm} \mbox{func} \mbox{operands} \\ = \mbox{farmed} \mbox{moreover} \mbox{fan} \mbox{farmed} \mbox{slaves} \\ \mbox{where} \\ \mbox{farmed} = \mbox{map} \mbox{func} \mbox{operands} \\ \mbox{slaves} = \mbox{partition} \mbox{ noOfProcessors} \mbox{farmed} \end{array}
```

where partition  $n \times s$  simply splits a list  $\times s$  into n sublists as equally as possible.

So, if noOfProcessors is 2, farm f [x<sub>0</sub>, x<sub>1</sub>, x<sub>2</sub>, x<sub>3</sub>] expands to

 $\begin{array}{l} \mbox{farmed} = \mbox{ map f } [x_0, \, x_1, \, x_2, \, x_3] \\ \mbox{moreover} & \mbox{Bundle } p_1 \mbox{ And } (\mbox{Arc } p_1 \mbox{ farmed}) \mbox{ And} \\ & \mbox{Bundle } p_2 \mbox{ And } (\mbox{Arc } p_2 \mbox{ farmed}) \mbox{ And} \\ & \mbox{NoPlace} \end{array}$ 

where

 $p_1 = [x_0, x_1]$  $p_2 = [x_2, x_3]$ 

• This is nice in that the code for the computation itself is completely unchanged

# Problems with aggregation by bundling

- Unfortunately, using bundling alone for aggregation has some serious practical problems
  - → The compiler has to elaborate the entire list of rays, rather than the list of partitions
  - → There are many communication messages, one for each ray – we have aggregated computation but not communication
- Related to this:
  - → The compiler has to know when to stop evaluating  $p_1$  and  $p_2$  – our implementation always evaluates placed expressions to WHNF
  - → So the first element of each list x<sub>0</sub>, x<sub>1</sub>,
     x<sub>2</sub> and x<sub>3</sub> is computed at compile-time (unless further input is required)

# Aggregation by restructuring

We can fix most of the the problems by modifying the computation to partition explicitly:

rayTrace scene viewpoint
= farm noOfProcessors impact rays
where
rays = generateRays viewpoint

impact ray = fold earlier impacts

#### where

impacts = map (hit ray) scene

farm :: Int  $\rightarrow$  (a $\rightarrow$ a)  $\rightarrow$  [a]  $\rightarrow$  [a]

farm n func input

= farmed moreover fan farmed slaves
where
farmed = unpartition slaves

slaves = map (map func) jobs

jobs = partition n input

#### Farmed ray tracer - performance



- The scene consisted on 20 simple objects
- There were  $100 \times 100$  rays
- MPI on the Fujitsu AP1000 at Imperial College (25MHz Sparc processors)
- Implementation built on the Haskell<sup>-</sup> compiler and a version of the FCG project back-end (with thanks to Pieter Hartel and others)



The graphs show execution time and speedup for the explicitly-partitioned ray tracer:

- A small further adjustment was made to avoid compile-time ray tracing
- Compute-ahead was adjusted to avoid excessive blocking
- Rays were partitioned into groups of 40 for maximum performance

# Summary

• Motivation

Controlling how a program is executed in parallel - declaratively

• Promise

Use the power of the functional language, use powerful generic operators, re-use operators from the computation

• Implementation

Use partial evaluation to specialise the program for each target machine, and perhaps even each problem instance

- Re-use, composition
- Problems

Compilation time, controlling partial evaluation, surprising effects of demand-driven execution

#### Discussion: dynamic vs static

- $\rightarrow$  Can Caliban be extended to be more dynamic?
  - Sure, we could interpret Bundle as a "spark"
  - We lose the opportunity to schedule aggregate process networks to processors

(eg to ensure processes in a pipeline aren't allocated to the same processor)

 We lose control over where communication happens
 (so where a given expression is computed may depend on a non-deterministic race between evaluation and sparking)

#### Directions for further work

- $\rightarrow$  Partial evaluation has lots of potential
- $\rightarrow$  Hybrid dynamic/static resource allocation is an interesting area
- $\rightarrow$  Caliban did not provide enough control over evaluation order and communication