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An Operational Semantics for Skeletons

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Outline

Skeletons
Semantics – motivations
The schema of semantics
Axioms – rules
Example
Concluding remarks

Skeletons

Skeletons are language constructs
 well-defined input-output behavior
 parallelism exploitation patterns
 (sometimes) can be nested
 several prepackaged implementations

Two main families
 Data Parallel (*map, reduce, scan* ...)
 Task & Stream parallel (*farm, pipeline,* ...)

Motivations

Usually formal functional semantics, informal parallel behavior

- Describe skeletons
 - in-out relationship (functional behavior)
 - parallel behavior
 - in uniform and precise way (non steady state)
 - in structural way
- Theoretical work motivated by concrete needs
 - Enable and automate performance-driven source-to-source optimizations
 - same in/out different parallel behaviors
 - Compare different skeleton sets expressive power



sequential source code just plugged in data items arrives in sequence, we cannot assume data is already distributed, data distribution cost is large, several farm scheduling policies are possible, as well as several data mappings
DipInf; 31/08/2003

Slide 5

DI1

pipe (map fc (seq f1) fd) (map gc (seq f2) gd)



Running example language: Lithium

Stream and Data Parallel

- 🖙 farm, pipe
- map, reduce, D&C, ...
- Can be freely nested
- All skeletons have a stream as in/out
- Java-based (skeletons are Java classes)
- Implemented and running [FGCS 19(5):2003] <u>http://www.di.unipi.it/~marcod/Lithium/</u> or sourceforge
- Macro data-flow run-time
- Support heterogeneous COWs
- Includes parallel structure optimization
 - performance-driven, source-to-source

The schema of semantics

Axioms, three kind per skeleton:
1. Describe skeletons within the steady state
2. Mark the begin of stream *

3. Manage the end of stream *

Six rules:

Ø

- 1. Two describing parallel execution (SP, DP)
 - Have a cost
- 2. Four to navigate in the program structure
 - No cost, ensure strict execution order

Look to SP/DP rules only to figure out program performance

The meaning of labels

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Label represent an enumeration of PEsTwo kind of labels:

On streams represent data mapping:

 $\left\langle x \right\rangle_{3}$ means x is available on PE₃

On arrows represent computation mapping

means such computation is performed by PE₄

Re-label O(l,x) a stream means communicate it
 Semantics may embed an user-defined policy O(l,x)
 Cost depend on label (topology) and data item x (size)

Axioms (steady state)

Skel params $\langle x, \tau \rangle_{\ell_1} \xrightarrow{\ell_1} \mathcal{F} \langle x \rangle_{\ell_2} :: Skel params \langle \tau \rangle_{\ell_3}$

1. Apply inner skeleton $F \in param$ to the stream head x

2. Recur on the tail of the stream

3. Expressions 1 & 2 are joined by :: operator

a. The arrow label gets left-hand side stream label

b. Labels in the right-hand side may change *(stream items may be bounced elsewhere)*

Lithium axioms (for stream par skeletons)

seq
$$f\langle x,\tau\rangle_{\ell} \xrightarrow{\iota} (fx)_{\ell} :: \text{seq } f\langle \tau\rangle_{\ell}$$

Embed seq code Stream unfolded, Labels unchanged

pipe $\Delta_1 \Delta_2 \langle x, \tau \rangle_{\ell} \xrightarrow{\ell} \Delta_2 \mathcal{R}_{\mathcal{O}(\ell, \mathbf{x})} \ \Delta_1 \langle x \rangle_{\ell} :: \text{pipe } \Delta_1 \Delta_2 \langle \tau \rangle_{\ell}$

- a. arrow label gets stream one happens locally
- b. label doesn't change keep 1st stage Δ_1 locally
- c. re-label R inserted in between 1st & 2nd stage it will map 2nd stage elsewhere
- d. tail is expected from the same source

farm
$$\Delta \langle x, \tau \rangle_{\ell} \xrightarrow{\ell} \Delta \langle x \rangle_{\mathcal{O}(\ell,\mathbf{x})} :: \text{ farm } \Delta \langle \tau \rangle_{\mathcal{O}(\ell,\mathbf{x})}$$

a. stream item is distributed accordingly *O* policyb. a reference of tail of the stream follows the head

Lithium axioms (DP skeletons)

 $\mathsf{map}\ f_{\mathsf{c}}\Delta\ f_{\mathsf{d}}\langle x,\tau\rangle_{\ell} \stackrel{\ell}{\longrightarrow} f_{\mathsf{c}}(\alpha\Delta)f_{\mathsf{d}}\langle x\rangle_{\ell}::\ \mathsf{map}\ f_{\mathsf{c}}\Delta\ f_{\mathsf{d}}\langle \tau\rangle_{\ell}$

 $\begin{aligned} \mathsf{d\&c} \ f_{\mathsf{tc}} f_{\mathsf{c}} \Delta \ f_{\mathsf{d}} \langle x, \tau \rangle_{\ell} & \stackrel{\ell}{\longrightarrow} \mathsf{d\&c} \ f_{\mathsf{tc}} f_{\mathsf{c}} \Delta \ f_{\mathsf{d}} \langle x \rangle_{\ell} :: \ \mathsf{d\&c} \ f_{\mathsf{tc}} f_{\mathsf{c}} \Delta \ f_{\mathsf{d}} \langle \tau \rangle_{\ell} \\ \mathsf{d\&c} \ f_{\mathsf{tc}} f_{\mathsf{c}} \Delta \ f_{\mathsf{d}} \langle y \rangle_{\ell} & \inf f_{\mathsf{c}} \langle f_{\mathsf{c}} \Delta \ f_{\mathsf{d}} \rangle_{\ell} \\ f_{\mathsf{c}} \left(\alpha \left(\mathsf{d\&c} \ f_{\mathsf{tc}} \ f_{\mathsf{c}} \Delta \ f_{\mathsf{d}} \right) \right) \ f_{\mathsf{d}} \ \langle y \rangle_{\ell} & otherwise \end{aligned}$

Lithium rules overview

 $\frac{f_{\mathsf{d}} \langle x \rangle_{\ell} \stackrel{\ell}{\longrightarrow} \langle \langle y_{1} \rangle_{\ell}, \cdots \langle y_{n} \rangle_{\ell} \ \diamond \quad \Delta \langle y_{i} \rangle_{\ell} \stackrel{\ell}{\longrightarrow} \langle z_{i} \rangle_{\ell} \quad f_{\mathsf{c}} \langle \langle z_{1} \rangle_{\ell}, \cdots \langle z_{n} \rangle_{\ell} \ \diamond \stackrel{\ell}{\longrightarrow} \langle z \rangle_{\ell}, \ i = 1..n}{f_{\mathsf{c}} (\alpha \Delta) f_{\mathsf{d}} \langle x \rangle_{\ell} \stackrel{\ell}{\longrightarrow} \langle z \rangle_{\ell}} dp$

$$\frac{\Delta \langle x \rangle_{\ell_1} \stackrel{\ell_2}{\longrightarrow} \langle y \rangle_{\ell_3}}{\mathcal{R}_{\ell} \ \Delta \langle x \rangle_{\ell_1} \stackrel{\ell_2}{\longrightarrow} \langle y \rangle_{\ell}} \ relabel$$

$$\frac{E_1 \stackrel{\ell}{\longrightarrow} E_2}{\Delta \ E_1 \stackrel{\ell}{\longrightarrow} \Delta \ E_2} \ context$$

 $\begin{array}{ll} \langle \sigma \rangle_{\ell_1} :: \langle \tau \rangle_{\ell_2} \xrightarrow{\perp} \langle \sigma, \tau \rangle_{\perp} & join \\ \langle \epsilon \rangle_{\perp} :: \langle \tau \rangle_{\ell_2} \xrightarrow{\perp} \langle \tau \rangle_{\ell_2} & join_e \end{array}$

sp rules details

Many semantics for each program

- *i=j=1* always possible, i.e. no stream parallelism is exploited
- All of them are "functionally confluent", describe the same in-out relationship
- All of them describe the same parallel behavior, but with different degrees of parallelism

Example (2-ways-2-stages pipeline)

farm(pipe (seq f_1) (seq f_2)) $\langle x_1, x_2, x_3, x_4, x_5, x_6, x_7 \rangle$

 $\langle \epsilon \rangle_{\perp} :: \mathsf{farm} \ (\mathsf{pipe} \ (\mathsf{seq} \ f_1) \ (\mathsf{seq} \ f_2)) \ \langle x_1, x_2, x_3, x_4, x_5, x_6, x_7 \rangle_{\perp}$

- $\langle \epsilon \rangle_{\perp}$: Iterate the same operation on the whole stream
 - pipe now disappeared
- $\langle \epsilon \rangle_{\perp}$:: pipe (• two different labels on streams: 0 and 1
 - two different labels on R: 02, 12

 $\begin{array}{l} \langle \epsilon \rangle_{\perp} :: \operatorname{seq} f_2 \mathcal{R}_{1} \underbrace{\operatorname{seq}} f_2 \langle t_1 \rangle_{\mathfrak{f}} \operatorname{same} \operatorname{same} \operatorname{same} \mathfrak{same} \mathfrak{same}$

 $x_5, x_6, x_7
angle_1$

 $\frac{1}{2} f_2 \langle x_3 \rangle_0 ::$

 $\langle x_7 \rangle_0$

Example (continued)

 $\begin{array}{l} \langle \epsilon \rangle_{\perp} :: \operatorname{seq} f_2 \, \mathcal{R}_{02} \operatorname{seq} f_1 \, \langle x_1 \rangle_0 :: \operatorname{seq} f_2 \, \mathcal{R}_{12} \operatorname{seq} f_1 \, \langle x_2 \rangle_1 :: \operatorname{seq} f_2 \, \mathcal{R}_{02} \operatorname{seq} f_1 \, \langle x_3 \rangle_0 :: \operatorname{seq} f_2 \, \mathcal{R}_{12} \operatorname{seq} f_1 \, \langle x_4 \rangle_1 :: \\ \operatorname{seq} f_2 \, \mathcal{R}_{02} \operatorname{seq} f_1 \, \langle x_5 \rangle_0 :: \operatorname{seq} f_2 \, \mathcal{R}_{12} \operatorname{seq} f_1 \, \langle x_6 \rangle_1 :: \operatorname{seq} f_2 \, \mathcal{R}_{02} \operatorname{seq} f_1 \, \langle x_7 \rangle_0 \end{array}$

- This formula no longer can be reduced by axioms
- sp rule can be applied:

"Any rightmost sequence of expressions can be reduced provided streams exploits different labels"

 In this case the longest sequence includes two expressions, i.e. the max. par degree is 2 (matching the double-pipeline startup phase)

Example (continued)

$$\begin{split} &\langle \epsilon \rangle_{\perp} :: \operatorname{seq} f_2 \langle f_1 \, x_1 \rangle_{02} :: \operatorname{seq} f_2 \langle f_1 \, x_2 \rangle_{12} :: \operatorname{seq} f_2 \, \mathcal{R}_{02} \operatorname{seq} f_1 \langle x_3 \rangle_0 :: \operatorname{seq} f_2 \, \mathcal{R}_{12} \operatorname{seq} f_1 \langle x_4 \rangle_1 :: \\ &\operatorname{seq} f_2 \, \mathcal{R}_{02} \operatorname{seq} f_1 \, \langle x_5 \rangle_0 :: \operatorname{seq} f_2 \, \mathcal{R}_{12} \operatorname{seq} f_1 \, \langle x_6 \rangle_1 :: \operatorname{seq} f_2 \, \mathcal{R}_{02} \operatorname{seq} f_1 \, \langle x_7 \rangle_0 \end{split}$$

Due to the re-labeling we have 4 adjacent expressions exploiting different labels: 02, 12, 0, 1 – i.e. a max. parallelism degree of 4

- The step can be iterated up to the end of stream
- Max parallelism degree 4 since no more than 4 different labels appear adjacently (easy to prove)

Example (continued)

By iterating SP rule we eventually get $\langle \epsilon \rangle_{\perp} :: \langle f_2(f_1 x_1) \rangle_{02} :: \langle f_2(f_1 x_2) \rangle_{12} :: \langle f_2(f_1 x_3) \rangle_{02} :: \langle f_2(f_1 x_4) \rangle_{12} :: \langle f_2(f_1 x_5) \rangle_{02} :: \langle f_2(f_1 x_5) \rangle$

Count parallelism
Count communications
or reason about it

Summary

Operational semantics for skeletons Describes both functional and parallel behavior User-defined mapping/scheduling User-defined comm/comp costs - General, easy to extend No similar results within the skeleton community Enable performance reasoning Skeleton normal-form [PDCS99, FGCS03, web] Provably correct automatic optimizations Formally describe your brand new skeleton and its performance





Questions ?

www.di.unipi.it/~aldinuc

Stream skeletons

farm \Rightarrow functionally the identity ! -a.k.a. parameter sweeping, embarrassingly parallel, replica manager ... instead for some other group it is apply-to-all pipe parallel functional composition = pipe $f_1 f_2 < x >$ computes $f_2 (f_1 x)$ $= f_1$, f_2 executed in parallel on different data items

Describe skeletons

Usually functional behavior only described
 Parallel behavior does matter for performance
 Usually performance described by cost formulas

$$T(\operatorname{scan} Op) = \left(\frac{n}{p} - 1\right) t_{Op} + g(p-1)\operatorname{comm_size} + l + \left(\frac{n}{p} + p - 2\right) t_{Op}$$

Doesn't describe the behavior just the cost
 What happens if Op is parallel ?
 Not compositional
 handmade for each architecture
 Data layout not described

Axioms (begin/end of the stream)

Begin of stream marking:

Skel params $\langle x, \tau \rangle \xrightarrow{\perp} \langle \epsilon \rangle_{\perp} ::$ Skel params $\langle x, \tau \rangle_{\perp}$

Skel params $\langle x \rangle_{\ell_1} \xrightarrow{\ell_1} \mathcal{F} \langle x \rangle_{\ell_2}$

An example of reduction

