# Generics in Data Parallel Haskell

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# Programming in DPH

### Multiply sparse vector with dense vector

$$svvm :: [:(Int,\ Float):] \rightarrow [:Float:] \rightarrow Float$$
$$svvm\ v\ w = sumP\ [:x * (w\ !:\ i) \mid (i, x) \leftarrow v:]$$

# Programming in DPH

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$$svvm :: [:(Int, Float):] \rightarrow [:Float:] \rightarrow Float$$
$$svvm\ v\ w = sumP\ [:x * (w\ !:\ i) \mid (i, x) \leftarrow v:]$$

## Multiply sparse matrix with dense vector

$$smvm :: [:[:(Int, Float):]:] \rightarrow [:Float:] \rightarrow Float$$
$$smvm\ m\ w = [:svvm\ v\ w \mid v \leftarrow m:]$$

# Programming in DPH

## Quicksort

$$
\begin{aligned}
&qsort\ xs \mid lengthP\ xs \leq 1 = xs \\
&\qquad\quad\ \mid otherwise \qquad\quad = yss\ !:\ 0\ +:+\ eqs\ +:+\ yss\ !:\ 1 \\
&\quad \mathbf{where} \\
&\qquad m \ \ = xs\ !:\ 0 \\
&\qquad lts = [:y \mid y \leftarrow xs,\ y < m:] \\
&\qquad eqs = [:y \mid y \leftarrow xs,\ y == m:] \\
&\qquad gts = [:y \mid y \leftarrow xs,\ y > m:] \\
&\qquad yss = [:\ qsort\ ys \mid ys \leftarrow [:lts,\ gts:]:]
\end{aligned}
$$

# Nested data parallelism

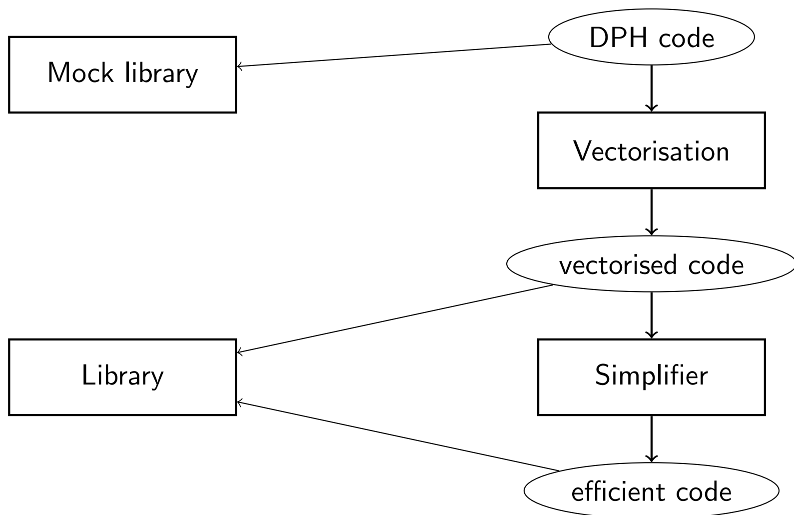
## Great for programmers

- single control flow
- implicit synchronisation and communication
- high-level specification of parallelism
- transparent integration into Haskell

## Bad for implementors

- extensive program transformations
- needs a very good optimiser

# Architecture



# Design

Library | Compiler

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Type families

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Parallel RTS

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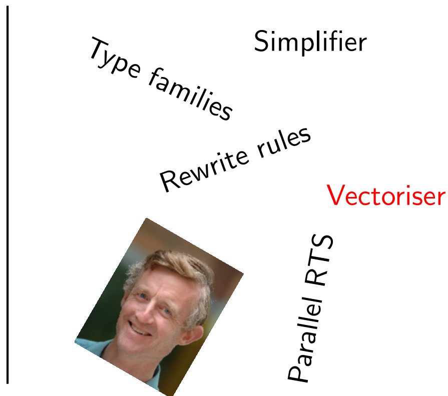

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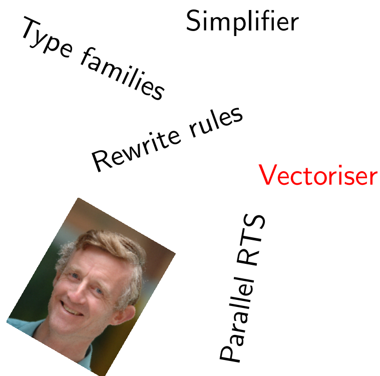

Parallel RTS

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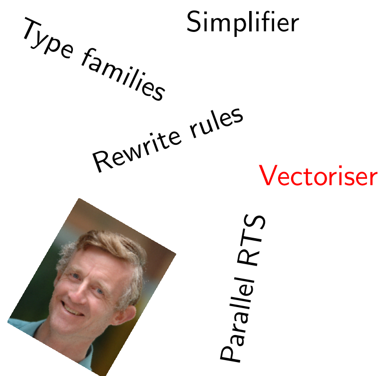

Parallel RTS

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## Library

Generic operations

Gang parallelism

Fusion

Type transformation

## Compiler

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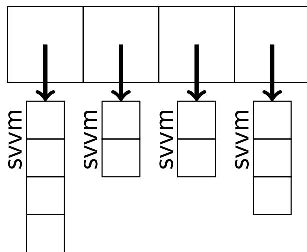
Parallel RTS

# Vectorisation – Basic idea

**Flattening transformation (Blelloch 1995)**

Convert a nested data parallel program into a flat data parallel program

- store nested arrays as flat arrays + segmenting information
- modify computations accordingly



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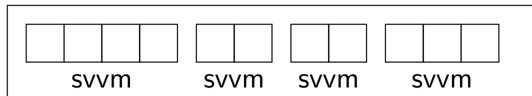

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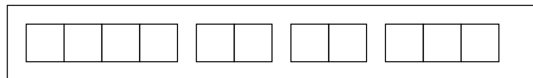

svvm$^{\uparrow}$

# Representing arrays

## Non-parametric representation

$$\begin{array}{lcl}\textbf{data family } [:a:] & & \\ \textbf{data instance } [:Double:] & = & \mathit{PDouble\ ByteArray} \\ \textbf{data instance } [:(a,b):] & = & \mathit{PPair}\ [:a:]\ [:b:] \\ \textbf{data instance } [:[:a:]:] & = & \mathit{PNest\ Segd}\ [:a:]\end{array}$$

- arrays of primitive types $\Longrightarrow$ memory blocks, no boxing!
- arrays of tuples $\Longrightarrow$ tuples of arrays
- nested arrays $\Longrightarrow$ flat data array + segmenting information
- arrays of trees $\Longrightarrow$ trees of arrays

## Sparse matrices

$[:[:(0,1.3),(2,0.5):],[:(1,3.4):],[::],[:(0,4.3),(1,2.2):]:]$
$\Longrightarrow$ $\mathit{PNest}\ \langle 2,1,0,2\rangle\ (\mathit{PPair}\ [:0,2,1,0,1:]\ [:1.3,0.5,3.4,4.3,2.2:])$

# Operations on parallel arrays

How do we implement $indexP :: [:a:] \rightarrow Int \rightarrow a$?

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How do we implement $indexP :: [:a:] \to Int \to a$?

**class** $PR\ a$ **where**
  $indexPR :: [:a:] \to Int \to a$
**instance** $PR\ Double$ **where**
  $indexPR\ (PDouble\ bytes)\ i = indexU\ bytes\ i$
**instance** $(PR\ a,\ PR\ b) \Rightarrow PR\ (a, b)$ **where**
  $indexPR\ (PPair\ xs\ ys)\ i = (indexPR\ xs\ i,\ indexPR\ ys\ i)$

$indexP_V :: PR\ a \Rightarrow [:a:] \to Int \to a$
$indexP_V = indexPR$

# Operations on parallel arrays

How do we implement $indexP :: [{:}a{:}] \rightarrow Int \rightarrow a$?

$$\textbf{data}\ PR\ a = PR\ \{\ indexPR :: [{:}a{:}] \rightarrow Int \rightarrow a\ \}$$

$$dPR_Double :: PR\ Double$$
$$dPR_Double = PR\ \{\ indexPR = \lambda(PDouble\ bs)\ i \rightarrow indexU\ bs\ i\ \}$$

$$dPR_Pair :: PR\ a \rightarrow PR\ b \rightarrow PR\ (a, b)$$
$$dPR_Pair\ pra\ prb = PR\ \{\ indexPR = \lambda(PPair\ xs\ ys)\ i \rightarrow (indexPR\ pra\ xs\ i,\ indexPR\ prb\ ys\ i)\ \}$$

$$indexP_V :: PR\ a \rightarrow [{:}a{:}] \rightarrow Int \rightarrow a$$
$$indexP_V = indexPR$$

# Handling user-defined types

$$\textbf{data}\ \mathit{Complex}\ a \ = \ \mathit{Complex}\ a\ a$$
$$\textbf{data instance}\ [{:}\mathit{Complex}\ a{:}] \ = \ \mathit{PComplex}\ [{:}a{:}]\ [{:}a{:}]$$

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$$\textbf{data}\ Complex\ a\ =\ Complex\ a\ a$$
$$\textbf{data instance}\ [{:}Complex\ a{:}]\ =\ PComplex\ [{:}a{:}]\ [{:}a{:}]$$

## Easy?

$$dPR_Complex\ ::\ PR\ a\ \rightarrow\ PR\ (Complex\ a)$$
$$dPR_Complex\ pa\ =\ PR\ \{\ indexPR\ =\ \lambda(PComplex\ xs\ ys)\ i\ \rightarrow$$
$$\quad Complex\ (indexPR\ pa\ xs\ i)\ (indexPR\ pa\ ys\ i)\ \}$$

# Handling user-defined types

$$\textbf{data}\ Complex\ a\ =\ Complex\ a\ a$$
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**Easy?**

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**Problem**

- *PR* has 14 operations at the moment
- there will be more in the future
- compiler has to know how to generate them

# Handling user-defined types

$$\textbf{data}\ Complex\ a = Complex\ a\ a$$
$$\textbf{data instance}\ [{:}Complex\ a{:}] = PComplex\ [{:}a{:}]\ [{:}a{:}]$$

**Easy with generic programming!**

- associate user-defined types with a generic product-sum representation
- fixed set of representation types
- vectoriser generates conversion functions
- "real" operations are implemented in the library

# Generic representations

**type family** $PRepr\ a$
**type instance** $PRepr\ (a, b)\ =\ (a, b)$
**type instance** $PRepr\ (Complex\ a\ a)\ =\ (a, a)$
**type instance** $PRepr\ (Maybe\ a)\ =\ Sum_2\ a\ ()$

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**type instance** $PRepr\ (Maybe\ a)\ =\ Sum_2\ a\ ()$

**data** $PA\ a\ =\ PA\ \{$
$toPRepr\ ::\ a \rightarrow PRepr\ a$
$fromPRepr\ ::\ PRepr\ a \rightarrow a$
$toArrPRepr\ ::\ [{:}a{:}] \rightarrow [{:}PRepr\ a{:}]$
$fromArrPRepr\ ::\ [{:}PRepr\ a{:}] \rightarrow [{:}a{:}]$
$dictPRepr\ ::\ PR\ (PRepr\ a)\}$

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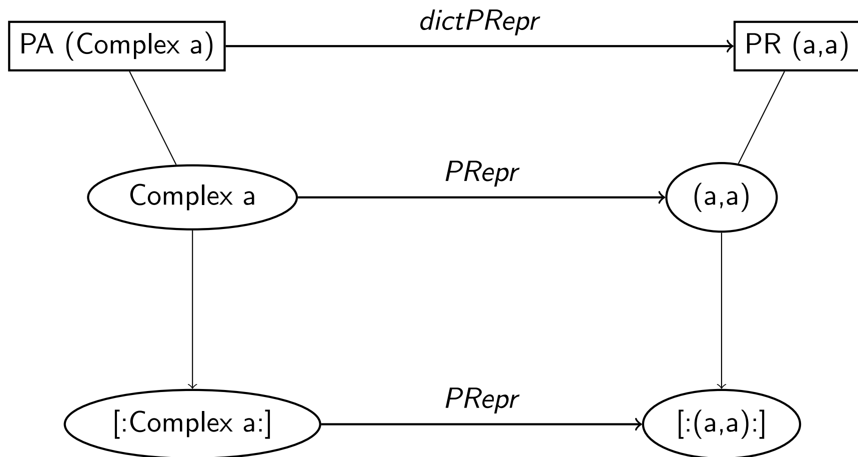
$\quad fromArrPRepr\ ::\ [{:}PRepr\ a{:}]\ \rightarrow\ [{:}a{:}]$
$\quad dictPRepr\ ::\ PR\ (PRepr\ a)\}$

$dPA_Complex\ ::\ PA\ a\ \rightarrow\ PA\ (Complex\ a)$
$dPA_Complex\ pa\ =\ PA\ \{\ toPRepr\ =\ \lambda(Complex\ x\ y)\ \rightarrow\ (x, y),\ \ldots\ \}$

# Generic operations



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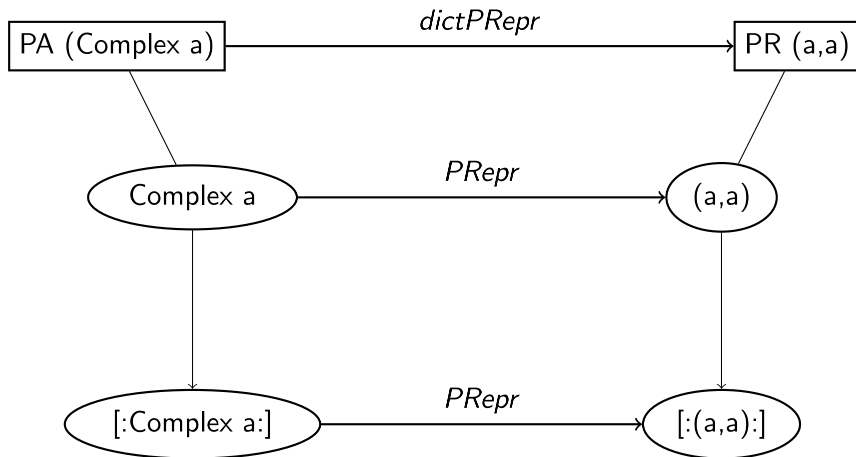


**type instance** *PRepr* [:*a*:] = [:*PRepr a*:]

# Implementing generic operations

## Library

$$
\begin{aligned}
&dPR_Pair \;::\; PR\;a \to PR\;b \to PR\;(a,b)\\
&dPR_Pair\;pra\;prb \;=\; PR\,\{\,indexPR \;=\; \lambda(PPair\;xs\;ys)\;i \to\\
&\qquad\qquad (indexPR\;pra\;xs\;i,\;indexPR\;prb\;ys\;i)\,\}
\end{aligned}
$$

$$
\begin{aligned}
&indexP_V \;::\; PA\;a \to [:a:] \to Int \to a\\
&indexP_V\;pa\;xs\;i \;=\; fromPRepr\;pa\\
&\qquad\qquad (indexPR\;(dictPRepr\;pa)\;(toArrPRepr\;pa\;xs)\;i)
\end{aligned}
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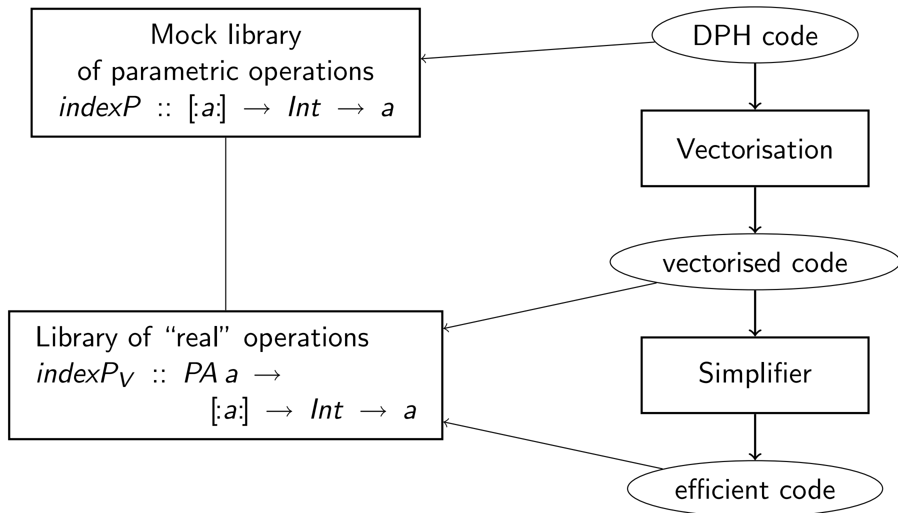
&indexP_V :: PA\ a \rightarrow [{:}a{:}] \rightarrow Int \rightarrow a\\
&indexP_V\ pa\ xs\ i = fromPRepr\ pa\\
&\qquad (indexPR\ (dictPRepr\ pa)\ (toArrPRepr\ pa\ xs)\ i)
\end{aligned}
$$

## Vectoriser

$$
\begin{aligned}
&\textbf{data instance}\ [{:}Complex\ a{:}] = Complex\ [{:}a{:}]\ [{:}a{:}]\\
&\textbf{type instance}\ PRepr\ (Complex\ a) = (a, a)
\end{aligned}
$$

$$
\begin{aligned}
&dPA_Complex :: PA\ a \rightarrow PA\ (Complex\ a)\\
&dPA_Complex\ pa = PA\ \{\ toPRepr = \lambda(Complex\ x\ y) \rightarrow (x, y)\ \ldots\ \}
\end{aligned}
$$

# Architecture – now with one more line!



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### ... for DPH

- minimise knowledge built into the vectoriser
- keep library code in the library
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## ... for everything else

- new approach to implementing generics
- design pattern directly supported by the language
- no performance penalty
- we can write papers about it!