

## Eden: Parallel Processes, Patterns and Skeletons

## Eden

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## Parallel

Functional
Programming

## Contents

(1) The Language Eden (in a nutshell)
(2) Skeleton-Based Programming
(3) Small-Scale Skeletons: Map and Reduce

4 Process Topologies as Skeletons
(5) Algorithm-Oriented Skeletons

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Learning Goals:

- Writing programs in the parallel Haskell dialect Eden;
- Reasoning about the behaviour of Eden programs;
- Applying the idea of skeleton-based programming;
- Applying and implementing parallel skeletons in Eden.


## Parallel Dialects of Haskell

- Data-Parallel Haskell ${ }^{\ddagger}$ (pure)

Type-driven parallel operations (on parallel arrays), sophisticated compilation (vectorisation, fusion, ...)

- Glasgow Parallel Haskell ${ }^{\ddagger *}$ (pure)
par, seq annotations for evaluation control, Evaluation Strategies


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- Eden* ("pragmatically impure") explicit process notion (mostly functional semantics), Distributed Memory (per process), implicit/explicit message passing


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- Glasgow Parallel Haskell ${ }^{\ddagger *}$ (pure) par, seq annotations for evaluation control, Evaluation Strategies
- Eden* ("pragmatically impure") explicit process notion (mostly functional semantics), Distributed Memory (per process), implicit/explicit message passing
- Concurrent Haskell ${ }^{\ddagger}$, Eden implementation* (monadic) explicit thread control and communication, full programmer control and responsibility
- Par Monad ${ }^{\ddagger}$, Cloud Haskell* (monadic) newer explicit variants, similar to Eden implementation


## Eden Constructs in a Nutshell

- Developed since 1996 in Marburg and Madrid
- Haskell, extended by communicating processes for coordination


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## Eden constructs for Process abstraction and instantiation

```
process ::(Trans a, Trans b)=> (a -> b) -> Process a b
(# ) :: (Trans a, Trans b) => (Process a b) -> a -> b
spawn :: (Trans a, Trans b) => [ Process a b ] -> [a] -> [b]
```

- Distributed Memory (Processes do not share data)
- Data sent through (hidden) 1:1 channels
- Type class Trans:
- stream communication for lists
- concurrent evaluation of tuple components
- Full evaluation of process output (if any result demanded)
- Non-functional features: explicit communication, $n$ : 1 channels


## Quick Sidestep: WHNF, NFData and Evaluation

- Weak Head Normal Form (WHNF):

Evaluation up to the top level constructor

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- Weak Head Normal Form (WHNF): Evaluation up to the top level constructor
- Normal Form (NF):

Full evaluation (recursively in sub-structures)

```
class NFData a where From Control.DeepSeq
    rnf :: a -> () -- This was a _Strategy_ in 1998
    rnf a = a 'seq' () -- returning unit ()
instance NFData Int
instance NFData Double
instance (NFData a) => NFData [a] where
    rnf [] = ()
    rnf (x:xs) = rnf x 'seq' rnf xs
```

instance (NFData a, NFData b) => NFData (a,b) where
$\operatorname{rnf}(a, b)=\operatorname{rnf} a \quad ' s e q ' \operatorname{rnf} b$

## Essential Eden: Process Abstraction/Instantiation

## Process Abstraction: process ::... (a -> b) -> Process a b <br> ```multproc = process (\x -> [ x*k | k <- [1,2..]])```

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## Essential Eden: Process Abstraction/Instantiation

## Process Abstraction: process ::... (a -> b) -> Process a b multproc $=$ process ( $\backslash \mathrm{x}$-> [ $\mathrm{x} * \mathrm{k} \mid \mathrm{k}<-[1,2 .$.$] )$ <br> Process Instantiation: (\#) ::... Process a b -> a -> b multiple5 = multproc \# 5 <br> 

- Full evaluation of argument (concurrent) and result (parallel)
- Stream communication for lists


## Spawning many processes: spawn ::.... [Process a b] -> [a] -> [b]

multiples $=$ spawn (replicate 10 multproc) [1..10]


## A Small Eden Example ${ }^{1}$

- Subexpressions evaluated in parallel
- ... in different processes with separate heaps

```
main = do args <- getArgs
    let first_stuff = (process f_expensive) # (args!!0)
        other_stuff = g_expensive $# (args!!1) -- syntax variant
    putStrLn (show first_stuff ++ '\n':show other_stuff)
```

[^0]
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... which will not produce any speedup!

```
main = do args <- getArgs
    let [first_stuff,other_stuff]
            = spawnF [f_expensive, g_expensive] args
    putStrLn (show first_stuff ++ '\n':show other_stuff)
```

- Processes are created when there is demand for the result!
- Spawn both processes at the same time using special function.
${ }^{1}$ (compiled with option -parcp or -parmpi)


## Basic Eden Exercise: Hamming Numbers

The Hamming Numbers are defined as the ascending sequence of numbers:

$$
\left\{2^{i} \cdot 3^{j} \cdot 5^{k} \mid i, j, k \in \mathbb{N}\right\}
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Dijkstra:
The first Hamming number is 1 . Each following Hamming number $H$ can be written as $H=2 K$, $H=3 K$, or $H=5 K$; where $K$ is a Hamming smaller than $H$.

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- Write an Eden program that produces Hamming numbers using parallel processes. The program should take one argument $n$ and produce the first $n$ Hamming numbers.

- Observe the parallel behaviour of your program using EdenTV.


## Non-Functional Eden Constructs for Optimisation

Location-Awareness:

```
noPe, selfPe :: Int
spawnAt :: (Trans a, Trans b) => [Int] -> [Process a b] -> [a] -> [b]
instantiateAt :: (Trans a, Trans b) =>
    Int -> Process a b -> a -> IO b
```


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Explicit communication using primitive operations (monadic)

```
data ChanName = Comm (Channel a -> a -> IO ())
createC :: IO (Channel a , a)
class NFData a => Trans a where
    write :: a -> IO ()
    write x = rdeepseq x 'pseq' sendData Data x
    createComm :: IO (ChanName a, a)
    createComm = do (cx,x) <- createC
        return (Comm (sendVia cx) , x)
```

Nondeterminism!

```
merge :: [[a]] -> [a]
```

Hidden inside a Haskell module, only for the library implementation.

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## The Idea of Skeleton-Basked Parallelism

You have already seen one example in the homework:

- Divide and Conquer, as a higher-order function

```
divConq :: (prob -> Bool) -- is the problem indivisible?
    -> (prob -> [prob]) -- split
    -> ([sol] -> sol) -- join
    -> (prob -> sol) -- solve a sub-problem
    -> (prob -> sol)
divConq indiv divide combine basecase = ...
```

(this is just one version, more later...)

- Parallel structure (rose tree) exploited for parallelism
- Abstracted from concrete problem


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parMap :: (a->b) -> [a] -> Par? [b]

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parMap :: (a->b) -> [a] -> Eval? [b]

## Algorithmic Skeletons for Parallel Programming

## Iteration:


divide \& conquer (fixed degree):


Algorithmic Skeletons [Cole 1989]: Boxes and lines - executable!

- Abstraction of algorithmic structure as a higher-order function


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- Embedded "worker" functions (by application programmer)
- Hidden parallel library implementation (by system programmer)


## Algorithmic Skeletons for Parallel Programming



## Google Map-Reduce:



Algorithmic Skeletons [Cole 1989]: Boxes and lines - executable!

- Abstraction of algorithmic structure as a higher-order function
- Embedded "worker" functions (by application programmer)
- Hidden parallel library implementation (by system programmer)
- Different kinds of skeletons: topological, small-scale, algorithmic


## Algorithmic Skeletons for Parallel Programming

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- Hidden parallel library implementation (by system programmer)
- Different kinds of skeletons: topological, small-scale, algorithmic

Explicit parallelism control and functional paradigm are a good setting to implement and use skeletons for parallel programming.

## Types of Skeletons

## Common Small-scale Skeletons

- encapsulate common parallelisable operations or patterns
- parallel behaviour (concrete parallelisation) hidden

Structure-oriented: Topology Skeletons

- describe interaction between execution units
- explicitly model parallelism

Proper Algorithmic Skeletons

- capture a more complex algorithm-specific structure
- sometimes domain-specific


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## Basic Skeletons: Higher-Order Functions

- Parallel transformation: Map

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map :: (a -> b) -> [a] -> [b]
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independent elementwise transformation
... probably the most common example of parallel functional programming (called "embarrassingly parallel")

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- Parallel Reduction: Fold

```
fold :: (a -> a -> a) -> a -> [a] -> a
```

with commutative and associative operation.

- Parallel (left) Scan:

```
parScanL :: (a -> a -> a) -> [a] -> [a]
```

reduction keeping the intermediate results.

- Parallel Map-Reduce:
combining transformation and reduction.


## Embarrassingly Parallel: map

map: apply transformation to all elements of a list

```
Straight-forward element-wise parallelisation
```

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parmap :: (Trans a, Trans b) => (a -> b) -> [a] -> [b]
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parmap = spawn . repeat . process
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Much too fine-grained!

```
Group-wise processing: Farm of processes
farm :: (Trans a, Trans b) => (a -> b) -> [a] -> [b]
farm \(f\) xs \(=\) join results
    where results = spawn (repeat (process (map f))) parts
    parts = distrib noPe xs -- noPe, so use all nodes
    join :: [[a]] -> [a]
    join \(=\)...
    distrib :: Int -> [a] -> [[a]]
    distrib n = ... -- join . distrib n == id
```


## Example application

Mandelbrot set visualisation $z_{n+1}=z_{n}^{2}+c$ for $c \in \mathbb{C}$

## Mandelbrot (Pseudocode)

```
pic :: ..picture-parameters.. -> PPMAscii
pic threshold ul lr dimx np s = ppmheader ++ concat (parMap computeRow rows)
    where rows = ...dimx..ul..lr..
        parMap = ...np..s.. -- different options to distribute rows
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- Fractal properties (self-similarity)
- Colours indicate speed of divergence
- Far out: diverges rapidly
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- Different rows expose different complexity


## Example Application: Chunked Tasks

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pic :: ..picture-parameters.. -> PPMAscii
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Simple chunking leads to load imbalance (task complexities differ)

## Example Application: Round-robin Tasks

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## Mandelbrot (Pseudocode)

```
pic :: ..picture-parameters.. -> PPMAscii
pic threshold ul lr dimx np s = ppmheader ++ concat (parMap computeRow rows)
    where rows = ...dimx..ul..lr..
        parMap = ..distributing round-robin..
```




Better: round-robin distribution, but still not well-balanced.

## Dynamic load-balancing: Master-Worker Skeleton

Worker nodes transform elementwise:

```
worker :: task -> result
```

Master node manages task pool

```
mw :: Int -> Int ->
    ( a -> b ) -> [a] -> [b]
```

mw np prefetch f tasks =


Parameters: no. of workers, prefetch

- Master sends a new task each time a result is returned (needs many-to-one communication)
- Initial workload of prefetch tasks for each worker: Higher prefetch $\Rightarrow$ more and more static task distribution Lower prefetch $\Rightarrow$ dynamic load balance


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- Result order needs to be reestablished!


## Master-Worker: An Implementation

```
Master-Worker Skeleton Code
mw np prefetch f tasks = results
    where
        fromWorkers = spawn workerProcs toWorkers
        workerProcs = [process (zip [n,n..] . map f) | n<-[1..np]]
        toWorkers = distribute tasks requests
```

- Workers tag results with their ID (between 1 and np ).


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## Master-Worker Skeleton Code

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    where
        fromWorkers = spawn workerProcs toWorkers
        workerProcs = [process (zip [n,n..] . map f) | n<-[1..np]]
        toWorkers = distribute tasks requests
        (newReqs, results) = (unzip . merge) fromWorkers
        requests = initialReqs ++ newReqs
        initialReqs = concat (replicate prefetch [1..np])
```

- Workers tag results with their ID (between 1 and $n p$ ).
- Result streams are non-deterministically merged into one stream


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## Master-Worker Skeleton Code

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        toWorkers = distribute tasks requests
        (newReqs, results) = (unzip . merge) fromWorkers
        requests = initialReqs ++ newReqs
        initialReqs = concat (replicate prefetch [1..np])
    distribute :: [t] -> [Int] -> [[t]]
    distribute tasks reqs = [taskList reqs tasks n | n<-[1..np]]
        where taskList (r:rs) (t:ts) pe | pe == r = t:(taskList rs ts pe)
        | otherwise = taskList rs ts pe
        taskList _ _ _ = []
```

- Workers tag results with their ID (between 1 and np).
- Result streams are non-deterministically merged into one stream
- The distribute function supplies new tasks according to requests.


## Parallel Reduction, Map-Reduce

Reduction (fold) usually has a direction

- foldl :: (b $\rightarrow$ a $\rightarrow$ b) $\rightarrow$ b $\rightarrow$ [a] $\rightarrow$ b
foldr :: (a $\rightarrow$ b $\rightarrow$ b) $\rightarrow$ b $\rightarrow$ [a] $\rightarrow$ b
Starting from left or right, implying different reduction function.
- To parallelise: break into sublists and pre-reduce in parallel.
$\Rightarrow$ needs to drop direction and narrow type
- Better options if order does not matter.


## Parallel Reduction, Map-Reduce

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- foldl :: (b $\rightarrow$ a $\rightarrow$ b) $\rightarrow$ b $\rightarrow$ [a] $\rightarrow$ b
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Starting from left or right, implying different reduction function.
- To parallelise: break into sublists and pre-reduce in parallel. $\Rightarrow$ needs to drop direction and narrow type
- Better options if order does not matter.

$$
\text { Example: } \sum_{k=1}^{n} \varphi(k)=\sum_{k=1}^{n}|\{j<k \mid \operatorname{gcd}(k, j)=1\}|
$$

## sumEuler

```
result = foldl (+) O (map phi [1..n])
phi k = length (filter (\ n -> gcd n k == 1) [1..(k-1)])
```


## Parallel Map-Reduce: Restrictions

## Map-Reduce skeleton

```
parmapReduce :: Int ->
    (a -> b) -> (b -> b -> b) -> b ->
    [a] -> b
parmapReduce np mapF redF neutral list = foldl redF neutral subRs
    where sublists = distrib np list
        subFold = process (foldl' redF neutral . (map mapF))
        subRs = spawn (replicate np subFold) sublists
```

- need to narrow type of the reduce parameter function!
- Associativity and neutral element (essential).


## Parallel Map-Reduce: Restrictions

## Map-Reduce skeleton

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```

- need to narrow type of the reduce parameter function!
- Associativity and neutral element (essential).
- commutativity (desired, more liberal distribution)
- distrib function may distribute in any order if redF commutative $\Rightarrow$ input consumed incrementally as a stream.
- Otherwise: need to know list length ahead of time (inefficient).


## Google Map-Reduce: Grouping Before Reduction

gMapRed : : (k1 -> v1 -> [(k2,v2)]) -- mapF
gMapRed : : (k1 -> v1 -> [(k2,v2)]) -- mapF
-> (k2 -> [v2] -> Maybe v3) -- reduceF
-> Map k1 v1 -> Map k2 v3 -- input / output

(1) Input: key-value pairs ( $\mathrm{k} 1, \mathrm{v} 1$ ), many or no outputs ( $\mathrm{k} 2, \mathrm{v} 2$ )
(2) Intermediate grouping by key k2
(3) Reduction per (intermediate) key k2 (maybe without result)
(3) Input and output: Finite mappings

## Google Map-Reduce: Grouping Before Reduction



## Word Occurrence

```
mapF :: URL -> String -> [(String,Int)]
mapF _ content = [(word,1) | word <- words content ]
reduceF :: String -> [Int] -> Maybe Int
reduceF word counts = Just (sum counts)
```


## Google Map-Reduce (parallel)


R.Lämmel,

Google's
Map-Reduce
Programming
Model
Revisited.
In: SCP 2008

```
gMapRed :: Int -> (k2-> Int) -> Int -> (v1->Int) -- parameters
    (k1 -> v1 -> [(k2,v2)]) -- mapper
    -> (k2 -> [v2] -> Maybe v3) -- pre-reducer
    -> (k2 -> [v3] -> Maybe v4) -- final reducer
    -> Map k1 v1 -> Map k2 v4 -- input / output
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## Examples / Exercise

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Describe how to compute the following in Google Map-Reduce:

- Reverse Web-Link Graph:

For a set of web pages, compute a dictionary to look up the pages that link to each page.

- URL Access Frequencies:

Compute access counts for URLs from a set of web server log files.

## Examples / Exercise

```
gMapRed :: Int -> (k2->Int) -> Int -> (v1->Int) -- parameters
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- Reverse Web-Link Graph:

For a set of web pages, compute a dictionary to look up the pages that link to each page.

## Reverse Link

Input are all URLs and page contents of the set. The map function outputs pairs (link target, source URL) for each link found in the source URL contents. The (pre-)reduce function joins the source URLs to the pair (target, list(source)) (removing duplicates).

- URL Access Frequencies:

Compute access counts for URLs from a set of web server log files.

## URL Access Frequency

Input are all log entries, stating the requested URLs. As in word-occurrence: The map function emits (URL,1) pairs for requested URLs, the reduce functions sum the counts.

## Outline

(1) The Language Eden (in a nutshell)
(2) Skeleton-Based Programming
(3) Small-Scale Skeletons: Map and Reduce
(4) Process Topologies as Skeletons
(5) Algorithm-Oriented Skeletons

## Process Topologies as Skeletons: Explicit Parallelism

- describe typical patterns of parallel interaction structure
- (where node behaviour is the function argument)
- to structure parallel computations


## Examples:



## Process Topologies as Skeletons: Explicit Parallelism

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- (where node behaviour is the function argument)
- to structure parallel computations


## Examples:


$\Rightarrow$ well-suited for functional languages (with explicit parallelism). Skeletons can be implemented and applied in Eden.

## Process Topologies as Skeletons: Ring



```
type RingSkel i o a b r = Int \(\rightarrow\) (Int \(\rightarrow\) i \(\rightarrow\) [a]) -> ([b] \(\rightarrow\) o) \(\rightarrow\)
    \(((a,[r]) \rightarrow(b,[r])) \rightarrow i \rightarrow 0\)
ring size makeInput processOutput ringWorker input =...
```

- Good for exchanging (updated) global data between nodes
- Ring processes connect to parent to receive input/send output
- Parameters: functions for
- decomposing input, combining output, ring worker


## Process Topologies as Skeletons: Torus

```
torus ::
    -- node behaviour
    (c-> [a]-> [b] -> (d, [a],[b])) ->
    -- input (truncated to shortest)
    [[c]] -> [[d]] -- result
```

- Initialisation data [[c]]

- Ring-shaped neighbour communication in two dimensions


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| $\mathrm{A}_{11}$ | $\mathrm{A}_{12}$ | $B_{11} B_{12}$ |  |
| :---: | :---: | :---: | :---: |
| $\mathrm{A}_{21}$ | $\mathrm{A}_{22}$ | $B_{21} B^{2}$ | $\begin{aligned} & \mathrm{A}_{2} \mathrm{~B}_{11} \mathrm{~A}_{2} \mathrm{~B}_{21} \mathrm{~B}_{2} \mathrm{~A}_{22} \mathrm{~B}_{21}+\mathrm{A}_{22} \mathrm{~B}_{22} \end{aligned}$ |

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## Algorithm-oriented Skeletons

## Divide and conquer

```
divCon :: (a -> Bool) -> (a -> b) -- trivial? / then solve
    -> (a -> [a]) -> (a -> [b] -> b) -- split / combine
    -> a -> b -- input / result
```


## Iteration

```
iterateUntil :: (inp -> ([ws],[t],ms)) -> -- split/init
    (t -> State ws r) -> -- worker
    ([r] -> State ms (Either out [t])) -- manager
    -> inp -> out
```


## Backtracking (Tree search)

```
backtrack :: (a -> (Maybe b, [a]) -- maybe solve problem, refine problem
    -> a -> [b] -- start problem / solutions
```


## Divide and Conquer Skeletons

- Mary, slide 66 in strategies lecture: binary divide\&conquer

```
divConq indiv split join f prob = undefined
divCon :: (a -> b) -> a -- base case fct., input
    -> (a -> Bool) -- parallel threshold
    -> (b -> b -> b) -- combine
    -> (a -> Maybe (a,a)) -- divide
    -> b
```

- Simon Marlow: slide 53, with a more general version

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```

- Simon Marlow: slide 53, with a more general version
divConq : : (prob -> Bool) -- is the problem indivisible?
-> (prob -> [prob]) -- split
-> ([sol] -> sol) -- join
-> (prob -> sol) -- solve a sub-problem
-> (prob -> sol)

```
so here is my version:
```

```
divCon :: Int ->
```

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(a -> Bool) -> (a -> b) -- trivial? / then solve
(a -> Bool) -> (a -> b) -- trivial? / then solve
-> (a -> [a]) -> (a -> [b] -> b) -- split / combine
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```


## Simple Divide \& Conquer

```
Divide \& Conquer Skeleton (simple general version)
dc_c depth trivial solve split combine x
    \(=\) if depth < 1 then seqDC \(x\)
        else if trivial \(x\) then solve \(x\)
            else childRs 'seq' -- early demand on children results
                                combine \(x\) (myR : childRs)
    where myself = dc_c (depth - 1) trivial solve split combine
    seqDC \(x=\) if trivial \(x\) then solve \(x\)
                                else combine \(x\) (map seqDC (split \(x\) ))
    (mine:rest) = split \(x\)
        myR = myself mine
        childRs = parMapEden myself rest
```


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        (mine:rest) = split x
        myR = myself mine
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```

Room for optimisation:

- Number of sub-problems often fixed by the algorithm
- Processes should be placed evenly on all machines

The Eden skeleton library contains many variants.
http://hackage.haskell.org/package/edenskel/

## Parallel iteration (an algorithmic skeleton)

## Iterated parallel map on tasks

```
iterateUntil ::
    (inp -> Int -> ([ws],[t],ms)) -> -- split/init
    (t -> State ws r) -> -- worker
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```



Worker: compute result r from task t using and updating a local state ws
Manager: decide whether to continue, based on master state ms and worker results [r]. produce tasks [ t ] for all workers

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Worker: compute result r from task t using and updating a local state ws
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Applications: N -body, K-means clustering, genetic algorithms...

## Backtracking: A Dynamically Growing Task Pool

- We use the master-worker skeleton with a small modification:

```
worker :: task -> (Maybe result,[task])
```

- New tasks enqueued in dynamically growing task pool.
- Backtracking: Test decision alternatives until reaching a result.


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## Parallel SAT Solver

- Can a given logic formula be satisfied?
- Task pool starting with just one task (no variable assigned).



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## Parallel SAT Solver

- Can a given logic formula be satisfied?
- Task pool starting with just one task (no variable assigned).
- Stateful master with task counter:
- consumes output of all workers

- add new tasks to task list
- shutdown when counter reaches zero


## Summary

- Eden: Explicit parallel processes, mostly functional face
- Two levels of Eden: Skeleton implementation and skeleton use
- Skeletons: High-level specification exposes parallel structure
- and enables programmers to think in parallel patterns.
- Different skeleton categories (increasing abstraction)
- Small-scale skeletons (map, fold, map-reduce, ...)
- Process topology skeletons (ring, torus...)
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- More information on Eden:
http://www.mathematik.uni-marburg.de/ eden
http://hackage.haskell.org/package/edenskel/ http://hackage.haskell.org/package/edenmodules/
http://hackage.haskell.org/package/edentv/


## Example: All Pairs Shortest Paths (Floyd-Warshall)

Adjacency Matrix
Distance Matrix

$$
\left(\begin{array}{ccccc}
0 & w_{1,2} & w_{1,3} & \ldots & w_{1, n} \\
w_{2,1} & 0 & w_{2,3} & \ldots & w_{2, n} \\
w_{3,1} & w_{3,2} & 0 & \ldots & w_{3, n} \\
\vdots & \vdots & \vdots & \vdots & \vdots \\
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\end{array}\right)
$$

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w_{n, 1} & w_{n, 2} & w_{n, 3} & \ldots & 0
\end{array}\right) \Rightarrow\left(\begin{array}{ccccc}
0 & d_{1,2} & d_{1,3} & \ldots & d_{1, n} \\
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d_{3,1} & d_{3,2} & 0 & \ldots & d_{3, n} \\
\vdots & \vdots & \vdots & \vdots & \vdots \\
d_{n, 1} & d_{n, 2} & d_{n, 3} & \ldots & 0
\end{array}\right)
$$

## Floyd-Warshall: Update all rows k in parallel

```
ring_iterate :: Int -> Int >> Int -> [Int] -> [[Int]] -> ([Int],[[Int]])
ring_iterate size k i rowk rows
    | i > size = (rowk, []) -- finished
    | i == k = (result, rowk:rest) -- send own row
    | otherwise = (result, rowi:rest)
    where rowi:xs = rows
        (result, rest) = ring_iterate size k (i+1) nextrowk xs
        nextrowk | i == k = rowk -- no update for own row
        | otherwise = updaterow rowk rowi distki
        distki = rowk!!(i-1)
```


## Trace of Warshall Program

First version:


## Trace of Warshall Program

First version:


with additional demand

## Differential Evolution [Price/Storn] with iteration skeleton

Worker: Mutate and select

- Randomly choose $C$ as either the best known (50\%) or a random candidate
- Add weighted difference of 2 other candidates: $C_{i}^{\prime}=C+\gamma\left(C_{r_{1}}-C_{r_{2}}\right)$
- Using fitness function $f$ : retain $C_{i}$ if better (minimising: $\left.f\left(C_{i}\right)<f\left(C_{i}^{\prime}\right)\right)$.
x NP Parameter vectors from generation $G$
- Newly generated parameter vector $\underline{v}$

- State: random gen., local candidates


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- Newly generated parameter vector $\underline{v}$

- State: random gen., local candidates

Manager: Collect/redistribute, identify best

- Termination:
- when best/all cand. good enough,
- or after $n$ iteration steps
- State: Iteration counter



## Eden usage example

```
Compile example, (with tracing - eventlog):
berthold@bwlf01$ COMPILER -parcp -eventlog -02 -rtsopts --make mandel.hs
[1 of 2] Compiling ParMap
    ( ParMap.hs, ParMap.o )
[2 of 2] Compiling Main ( mandel.hs, mandel.o )
Linking mandel ...
```


## Run, second run with tracing:

```
berthold@bwlf01$ ./mandel 0 200 1 -out +RTS -qp4 > out.ppm
==== Starting parallel execution on 4 processors ...
berthold@bwlf01$ ./mandel 0 50 1 +RTS -qp4 -l
==== Starting parallel execution on 4 processors ...
Done (no output)
Trace post-processing...
    adding: berthold=mandel#1.eventlog (deflated 65%)
    adding: berthold=mandel#2.eventlog (deflated 59%)
    adding: berthold=mandel#3.eventlog (deflated 58%)
    adding: berthold=mandel#4.eventlog (deflated 58%)
berthold@bwlf01$ edentv berthold\=mandel_0_50_1_+RTS_-qp4_-1.parevents
```


[^0]:    ${ }^{1}$ (compiled with option -parcp or -parmpi)

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