



A Dynamically-typed Language for Prototyping High-Performance Data Parallel Programs



Hidehiko Masuhara and Tomoyuki Aotani (Tokyo Tech)

Ikra is

- an extended Ruby impl.
- w/data-parallel exec.
- (map & reduce over arrays)
- targeting GPGPU
- (& TSUBAME in future)

example: 2D diffusion + visualization

```
require 'sdl' # user requested library
# array initialization
a = Array.new_(SIZE,SIZE){ |x,y|
  ...initialization...
}
while true
  a = a.neighbor9.map{ |n|
    [n[-1,-1]+n[-1, 0]+n[-1, 1]+
     n[ 0,-1]+n[ 0, 0]+n[ 0, 1]+
     n[ 1,-1]+n[ 1, 0]+n[ 1, 1]]/9
  }
  # visualize the array
  p = a.map{|v| colorHSB(v,1,0.5)}
  show(SIZE,SIZE,p.pack("I!*"))
end
```

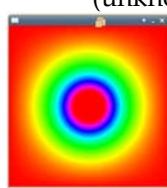
Goal of Ikra

- quick prototyping of data-parallel programs
- integration with Ruby
 - straightforward APIs
 - minimal annotations
 - compatibility with existing libraries

type inference
supports integer,
floating, bool, array,
tuple, neighbor & proc.
(plus user defined
objects in future)

array(2,float)
int × int → float
array(2,array(2,float))
array(2,float) → float

Object × Object × Object → Object
(unknown library calls are dynamic)



visualize

Rubyv
Rubyv
Ruby
libs.

sequential code (Ruby)

```
require('parray') #parallel array class
require('kern') #compiled kernel code
require('sdl') #user requested library
...
a = (PArray.new_(SIZE,SIZE){ |ptr,s0,s1|
  Kern.init0(ptr,s0,s1)
})
while true do
  a = a.neighbor9().map{ |pin,z,pout,s0,s1|
    Kern.map1(pin,z,pout,s0,s1)
  }
  p = a.map(){ |pin,pout,s0,s1|
    Kern.map2(pin,pout,s0,s1)
  }
  show(SIZE,SIZE,p.pack("I!*"))
end
```

type inference

```
require 'sdl' # user requested library
# array initialization
a = Array.new_(SIZE,SIZE){ |x,y|
  ...initialization...
}
while true
  a = a.neighbor9.map{ |n|
    [n[-1,-1]+n[-1, 0]+n[-1, 1]+
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  }
  # visualize the array
  p = a.map{|v| colorHSB(v,1,0.5)}
  show(SIZE,SIZE,p.pack("I!*"))
end
```

criteria: operations on arrays
where no Object type appears

+ parallel skeletons

kernel selection

kernel & wrapper (CUDA)

```
device
inline float body_map1(float n_1, float n_10, float n_11, float n_0_1,
                      float n_0, float n_01, float n_1, float n_10_1, float n_11_1)
return (n_1_1+n_10+n_11+n_0_1+n_10+n_11)/9;
}
global_ void body_map1wrap(float* in, float z, float* out,
int offset=blockIdx.x*blockDim.x+threadIdx.x;
int io=(offset+*s1);
int id=(offset+(s0*s1))/s1, id1=offset*s1;
out[offset]=
body_map1(aref(in,i0-1,i1-1,s0,s1,z),
aref(in,i0-1,i1, s0,s1,z),
...rest 7 neighbors...);
}
inline void kernel_map1(float* in, float z, float* out,
int s0, int s1);
body_map1wrap<<<(s0*s1+NUM_THREADS-1)/NUM_THREADS>>>(in,z,out,s0,s1);
```

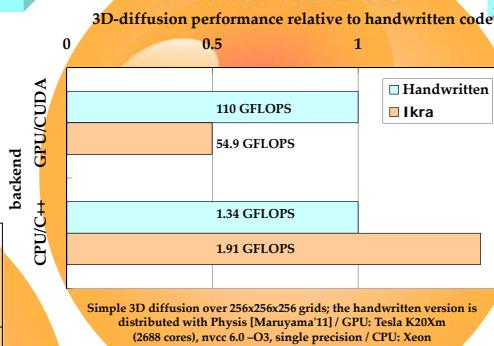
kernel call

wrapper

stock Ruby VM (CPU)

native method call

Performance



GPU

API

`PArray.new(s1,...,sN)`
`{ |i1,..,iN| e }` constructs
 N-dim. array initialized by e
`a.map{ |v| e }` constructs a
 new array from elements in a
`a.reduce(z){ |x,y| e }`
 reduces all elements in a by e
`a1.zip(a2,..,aN)` virtually
 constructs an array of N-tuples
`a.neighborN()` virtually
 constructs an array of N-neighbors

extended to
support stencil
computation

compatible w/Ruby's Array

~50% of handwritten CUDA code due to
different implementation of boundary checking

~ 1000x faster than
Ruby interpreter

Optimizations (plan)

- separating boundary comp.
- spatial/temporal blocking
- loop fusion
- communication/computation overlapping

designing a modular
framework