

# Scientific GPU computing with Go

A novel approach to highly reliable CUDA HPC

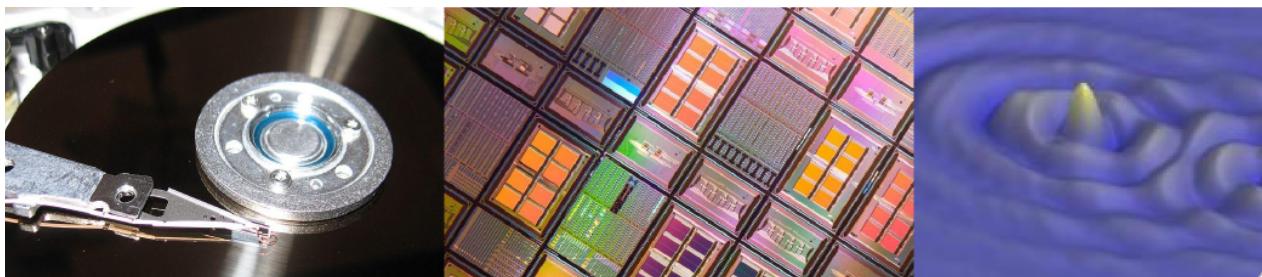
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Arne Vansteenkiste  
Ghent University

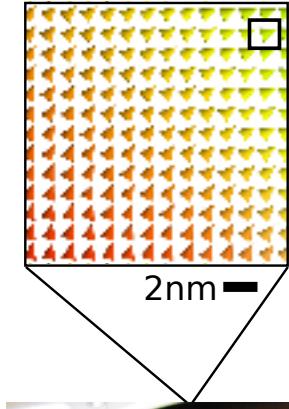
## Real-world example (micromagnetism)

DyNaMat LAB @ UGent: Microscale Magnetic Modeling:

- Hard Disks
- Magnetic RAM
- Microwave components
- ...



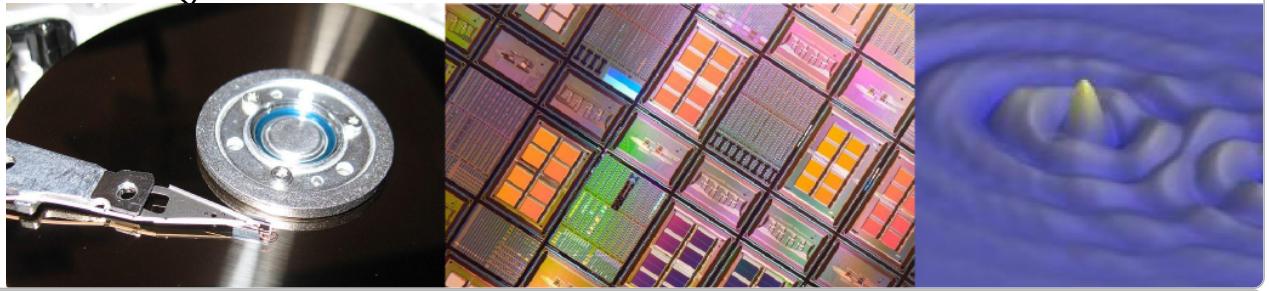
## Real-world example (micromagnetism)



$$\begin{aligned}
 \frac{\partial \vec{m}}{\partial t} &= \gamma_0 (\vec{\tau}_{\text{LL}} + \vec{\tau}_{\text{STT}}) \\
 \vec{\tau}_{\text{LL}} &= \frac{1}{1+\alpha^2} (\vec{m} \times \vec{B}_{\text{eff}} + \alpha (\vec{m} \times (\vec{m} \times \vec{B}_{\text{eff}}))) \\
 \alpha &= \alpha(\vec{r}, t) \\
 \vec{m} &= \frac{\vec{M}(\vec{r}, t)}{M_s} \\
 M_s &= |\vec{M}(\vec{r}, t)| \\
 \vec{B}_{\text{eff}} &= \vec{B}_d + \vec{B}_a + \vec{B}_s + \vec{B}_n + \vec{B}_{\text{in}} \\
 \vec{B}_d(\vec{r}, t) &= \iiint_V \vec{K}(\vec{r} - \vec{r}') \cdot \mu_0 \vec{M}(\vec{r}', t) d^3 \vec{r}' \\
 &= \mathcal{F}^{-1} \left( \mathcal{F} \left( \vec{K}(\vec{r}) \right) \cdot \mathcal{F} \left( \mu_0 \vec{M}(\vec{r}, t) \right) \right) \\
 \vec{K}_i(\vec{r}) &= \frac{1}{4\pi} \left( \frac{3(\vec{e}_i \cdot \hat{r})\hat{r} - \vec{e}_i}{r^3} \right) + \frac{2}{3} \vec{e}_i \delta^3(\vec{r}) \\
 \vec{B}_a &= \frac{2A}{M_s} \Delta \vec{m} + \frac{2D}{M_s} \left( \frac{\partial m_x}{\partial x} \frac{\partial m_z}{\partial y} - \frac{\partial m_x}{\partial x} \frac{\partial m_y}{\partial y} \right) \\
 A &= A(\vec{r}, t) \\
 \vec{B}_s &= \vec{B}_s(\vec{r}, t) \\
 \vec{B}_n &= \vec{B}_n + \vec{B}_c \\
 K_{nl} &= K_{nl}(\vec{r}, t)
 \end{aligned}$$

$$\begin{aligned}
 \vec{u} &= \vec{u}(\vec{r}, t) \\
 \vec{B}_{\text{el}} &= (A_{10}c_{10} + A_{19}c_{21} + A_{23}c_{30}) \\
 \vec{A}_c &= K_{c1} (a_1(a_2^2 + a_3^2), a_2(a_1^2 + a_3^2), a_3(a_1^2 + a_2^2)) \\
 a_i &= \vec{e}_i \cdot \vec{m} \\
 \vec{e}_1 &= \vec{e}_1(\vec{r}, t) \\
 \vec{e}_2 &= \vec{e}_2(\vec{r}, t) \\
 \vec{e}_3 &= \vec{e}_1 \times \vec{e}_2 \\
 \vec{B}_{\text{th}}(\vec{r}, t) &= \eta(\vec{r}, t) \sqrt{\frac{k_B^2 T}{\mu_0 \gamma_0 M_s \Delta V \Delta t}} \\
 \vec{\tau}_{\text{STT}} &= \vec{\tau}_{\text{ZL}} + \vec{\tau}_{\text{SL}} \\
 &= \frac{1}{1+\alpha^2} ((1+\xi\alpha) \vec{m} \times (\vec{m} \times (\vec{u} \cdot \nabla) \vec{m}) + (\xi-\alpha) \vec{m} \times (\vec{u} \cdot \nabla) \vec{m}) \\
 \vec{u} &= \frac{\mu_0 \eta_0}{2e \gamma_0 B_0 (1+\xi^2)} \vec{j} \\
 \vec{j} &= \vec{j}(\vec{r}, t) \\
 \vec{\tau}_{\text{SL}} &= \beta \epsilon (\vec{m} \times \vec{m}_P \times \vec{m}) - \beta \epsilon' \vec{m} \times \vec{m}_P \\
 \beta &= \frac{j_z \hbar}{M_s e d} \\
 \epsilon &= \frac{P(\vec{r}, t) \Lambda^2}{(\Lambda^2 + 1) + (\Lambda^2 - 1)(\vec{m} \cdot \vec{m}_P)} \\
 \epsilon' &= \epsilon'(\vec{r}, t)
 \end{aligned}$$

$$\begin{aligned}
 \frac{\partial m_x}{\partial y} \Big|_{\partial V} &= -\frac{D}{2A} m_z \\
 \frac{\partial m_y}{\partial x} \Big|_{\partial V} &= 0 \\
 \frac{\partial m_z}{\partial x} \Big|_{\partial V} &= \frac{D}{2A} m_x \\
 \frac{\partial m_x}{\partial z} \Big|_{\partial V} &= 0 \\
 \frac{\partial m_y}{\partial y} \Big|_{\partial V} &= -\frac{D}{2A} m_z \\
 \frac{\partial m_z}{\partial y} \Big|_{\partial V} &= \frac{D}{2A} m_y \\
 \frac{\partial m_x}{\partial z} \Big|_{\partial V} &= 0 \\
 \frac{\partial m_y}{\partial z} \Big|_{\partial V} &= 0 \\
 \frac{\partial m_z}{\partial z} \Big|_{\partial V} &= 0
 \end{aligned}$$

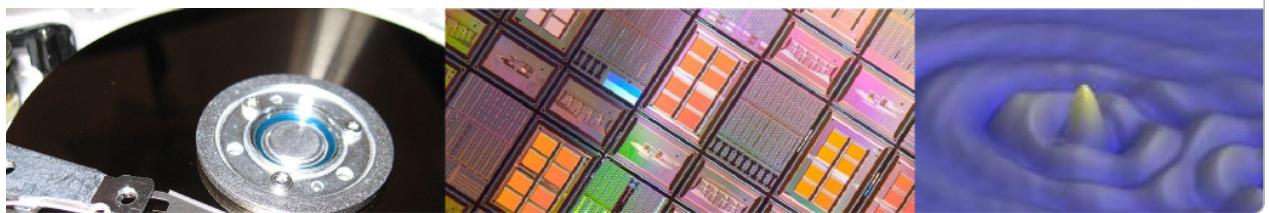


## Real-world example (micromagnetism)

**MuMax3** (GPU, script + GUI): ~ 11,000 lines CUDA, Go  
(<http://mumax.github.io>)

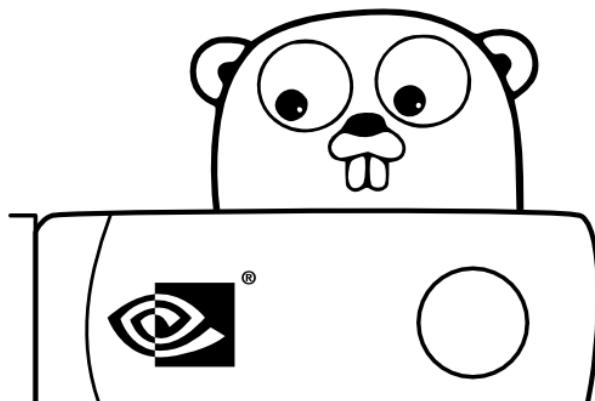
Compare to:

- **OOMMF** (script + GUI): ~100,000 lines C++, tcl
- **Magnum** (GPU, script only): ~ 30,000 lines CUDA, C++, Python



## How suitable is Go for HPC?

- Pure Go number crunching
- Go plus {C, C++, CUDA} number crunching
- Concurrency



### Go is

- compiled
- statically typed

### but also

- garbage collected
- memory safe
- dynamic

## Hello, math!

```
func main() {
    fmt.Println("(1+1e-100)-1 =", (1+1e-100)-1)
    fmt.Println("√-1      =", cmplx.Sqrt(-1))
    fmt.Println("J₁(0.3)      =", math.J1(0.3))
    fmt.Println("Bi(666, 333) =", big.NewInt(0).Binomial(666, 333))
}
```

Run

### Go math features:

- precise compile-time constants
- complex numbers
- special functions
- big numbers.

```
(1+1e-100)-1 = 1e-100
√-1      = (0+1i)
J₁(0.3)      = 0.148318816273104
Bi(666, 333) = 94627427937349739136904337970206130
```

Program exited.

### But missing:

- matrices
- matrix libraries (BLAS, FFT, ...)

Run Kill Close

## Performance

### Example: dot product

```
func Dot(A, B []float64) float64{
    dot := 0.0
    for i := range A{
        dot += A[i] * B[i]
    }
    return dot
}
```

## Performance

```
func Dot(A, B []float64) float64{
    dot := 0.0
    for i := range A{
        dot += A[i] * B[i]
    }
    return dot
}

func BenchmarkDot(b *testing.B) {
    A, B := make([]float64, 1024), make([]float64, 1024)

    sum := 0.0
    for i:=0; i<b.N; i++{
        sum += Dot(A, B)
    }
    fmt.Fprintln(DevNull, sum) // use result
}
```

go test -bench .

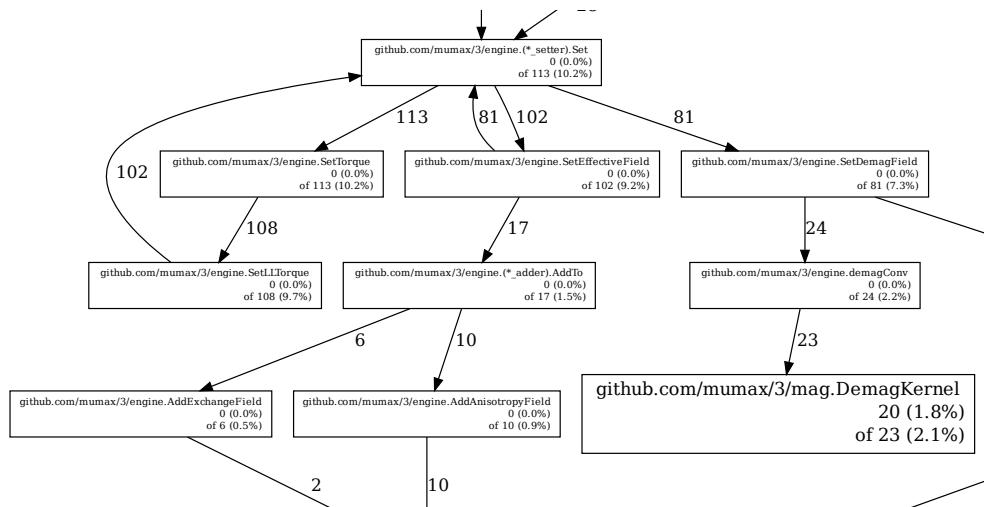
times all BenchmarkXXX functions

## Profiling

Go has built-in profiling

```
go tool pprof
```

outputs your program's call graph with time spent per function



## Performance

Dot product example

Go (gc)	1 980 ns/op
Go (gcc -O3)	1 570 ns/op
C (gcc -O3)	1 460 ns/op
C (gcc -march=native)	760 ns/op
Java	2 030 ns/op
Python	200 180 ns/op

- Typically, Go is ~10% slower than optimized, portable C
- But can be 2x - 3x slower than machine-tuned C

## Pure Go number crunching

On the up side

- Good standard math library
- Built-in testing, benchmarking & profiling
- Managed memory

On the down side

- Still slower than machine-tuned C
- No matrix libraries etc.

## How suitable is Go for HPC?

- Pure Go number crunching
- Go plus {C, C++, CUDA} number crunching
- Concurrency

## Hello, GPU!

Go can call C/C++ libs

```
//#include <cuda.h>
//#cgo LDFLAGS: -lcuda
import "C"
import "fmt"

func main() {
    buf := C.CString(string(make([]byte, 256)))
    C.cudaDeviceGetName(buf, 256, C.CUdevice(0))
    fmt.Println("Hello, your GPU is:", C.GoString(buf))
}
```

Program exited.

Building:

```
go build
```

All build information is in the source

## Hello, GPU! (wrappers)

```
import(
    "github.com/barnex/cuda5/cu"
    "fmt"
)

func main(){
    fmt.Println("Hello, your GPU is:", cu.Device(0).Name())
}
```

Run

Hello, your GPU is: GeForce GT 650M

Installing 3rd party code:

Program exited.

```
go get github.com/user/repo
```

(dependencies are compiled-in)

Run Kill Close

## Calling CUDA kernels (the C way)

GPU (code for one element)

```
__global__ void add(float *a, float *b, float *c, N) {
    int i = blockIdx.x * blockDim.x + threadIdx.x;
    if (i < N)
        c[i] = a[i] + b[i];
}
```

CPU wrapper (divide and launch)

```
void gpu_add(float *a, float *b, float *c, int N){
    dim3 block = ...
    add<<<N/BLOCK, BLOCK>>>(a, b, c);
}
```

Go wrapper wrapper

```
func Add(a, b, c []float32){
    C.gpu_add(unsafe.Pointer(&a[0]), unsafe.Pointer(&b[0]),
              unsafe.Pointer(&c[0]), C.int(len(a)))
}
```

## Calling CUDA kernels (cuda2go)

- CUDA kernel to Go wrapper (calling `nvcc` once).
- Further deployment without `nvcc` or CUDA libs.

Others to fetch your CUDA project the usual way:

```
go get github.com/user/my-go-cuda-project
```

```
// THIS FILE IS GENERATED BY CUDA2GO, EDITING IS FUTILE
func Add(a, b, c unsafe.Pointer, N int, cfg *config) {
    args := add_args_t{a, b, c, N}
    cu.LaunchKernel(add_code, cfg.Grid.X, cfg.Grid.Y, cfg.Grid.Z, cfg.Block.X, cfg.Block.Y, cfg.Blo
}

// PTX assembly
const add_ptx_20 = `

.version 3.1
.target sm_20
.address_size 64
```

## A note on memory (CPU)

Go is memory-safe, garbage collected.

Your typical C library is not.

Fortunately:

- Go is aware of C memory (no accidental garbage collection).
- Go properly aligns memory (needed by some HPC libraries)

Allocate in Go, pass to C, let Go garbage collect

## A note on memory (GPU)

GPU memory still needs to be managed manually.  
But a **GPU memory pool** is trivial to implement in Go.

```
var pool = make(chan cu.DevicePtr, 16)

func initPool(){
    for i:=0; i<16; i++{
        pool <- cu.MemAlloc(BUFSIZE)
    }
}

func recycle(buf cu.DevicePtr){
    pool <- buf
}

func main(){
    initPool()

    GPU_data := <- pool
    defer recycle(GPU_data)
    // ...
}
```

Run Run

## Vector add example

Adding two vectors on GPU (example from nvidia)

```
#include "../common/book.h"
#define N 10

int main( void ) {
    int a[N], b[N], c[N];
    int *dev_a, *dev_b, *dev_c;
    // allocate the memory on the GPU
    HANDLE_ERROR( cudaMalloc( (void**)&dev_a, N * sizeof(int) ) );
    HANDLE_ERROR( cudaMalloc( (void**)&dev_b, N * sizeof(int) ) );
    HANDLE_ERROR( cudaMalloc( (void**)&dev_c, N * sizeof(int) ) );

    // fill the arrays 'a' and 'b' on the CPU
    for (int i=0; i<N; i++) {
        a[i] = -i;
        b[i] = i * i;
    }
    // copy the arrays 'a' and 'b' to the GPU
    HANDLE_ERROR( cudaMemcpy( dev_a, a, N * sizeof(int), cudaMemcpyHostToDevice ) );
    HANDLE_ERROR( cudaMemcpy( dev_b, b, N * sizeof(int), cudaMemcpyHostToDevice ) );
    add<<<N,1>>>( dev_a, dev_b, dev_c ); // copy the array 'c' back from the GPU to the CPU
    HANDLE_ERROR( cudaMemcpy( c, dev_c, N * sizeof(int), cudaMemcpyDeviceToHost ) );
    // display the results
    for (int i=0; i<N; i++) r
```

## Vector add example

Adding two vectors on GPU (Go)

```
package main

import "github.com/mumax/3/cuda"

func main(){

    N := 3
    a := cuda.NewSlice(N)
    b := cuda.NewSlice(N)
    c := cuda.NewSlice(N)
    defer a.Free()
    defer b.Free()
    defer c.Free()

    a.CopyHtoD([]float32{0, -1, -2})
    b.CopyHtoD([]float32{0, 1, 4})

    cfg := Make1DConfig(N)
    add_kernel(a.Ptr(), b.Ptr(), c.Ptr(), cfg)

    fmt.Println("result:", a.HostCopy())
}
```

## Go plus {C, C++, CUDA} number crunching

On the downside

- Have to write C wrappers

On the upside

- You *can* call C
- Have Go manage your C memory

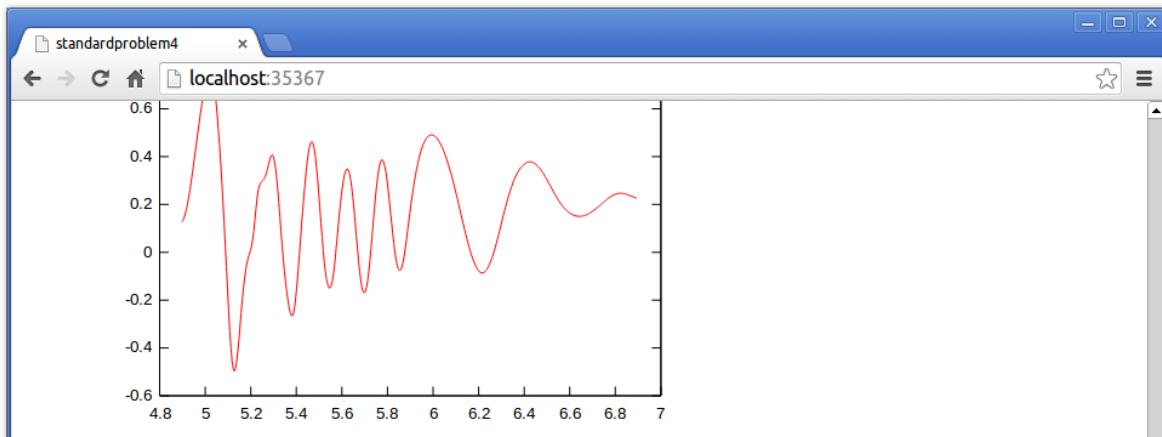
## How suitable is Go for HPC?

- Pure Go number crunching
- Go plus {C, C++, CUDA} number crunching
- **Concurrency**

## Real-world concurrency (MuMax3)

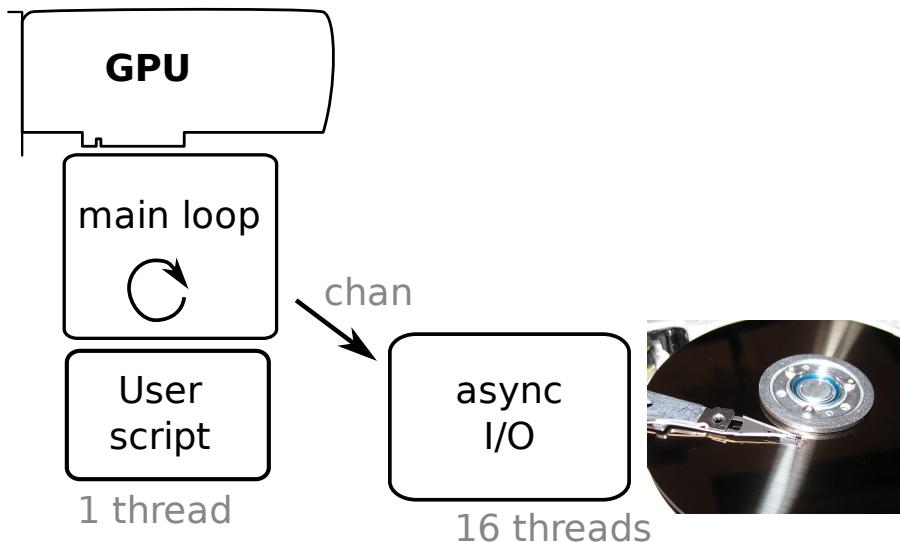
There's more to HPC than number crunching and memory management

- I/O
- Interactive supercomputing
- ...



## Real-world concurrency (MuMax3)

Output: GPU does not wait for hard disk



## Real-world concurrency (MuMax3)

Go channels are like type-safe UNIX pipes between threads.

```
var pipe = make(chan []float64, BUFSIZE)

func runIO(){
    for{
        data := <- pipe // receive data from main
        save(data)
    }
}

func main() {
    go runIO()           // start I/O worker
    pipe <- data       // send data to worker
}
```

Real example: 60 lines Go, ~2x I/O speed-up

## Real-world concurrency (MuMax3)

You can send function **closures** over channels.

```
var pipe = make(chan func())           // channel of functions

func main() {
    for {
        select{
            case f := <- pipe:          // execute function if in pipe
                f()
            default: doCalculation() // nothing in pipe, crunch on
        }
    }
}

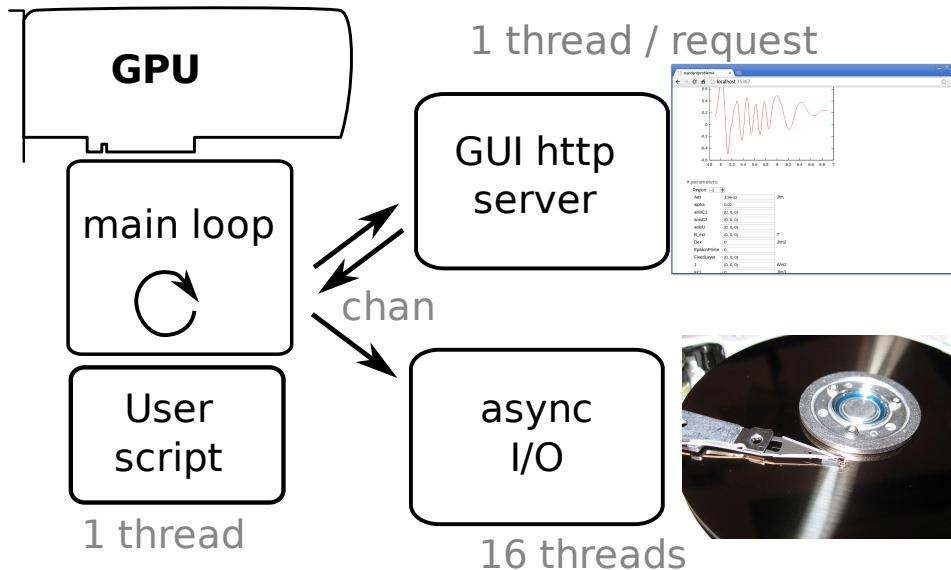
func serveHttp(){
    pipe <- func(){ value = 2 }      // send function to main loop
    ...
}
```

Run Run

Concurrency without mutex locking/unlocking.

## Real-world concurrency (MuMax3)

GUI: change parameters while running,  
without race conditions



## And we can prove it's thread-safe

Go has built-in testing for race conditions

```
go build -race
```

enables race testing. Output if things go wrong:

```
=====
WARNING: DATA RACE
Write by goroutine 3:
    main.func·001()
        /home/billgates/buggycode/race.go:10 +0x38

Previous read by main goroutine:
    main.main()
        /home/billgates/buggycode/race.go:21 +0x9c

Goroutine 3 (running) created at:
    main.main()
        /home/billgates/buggycode/race.go:12 +0x33
=====
```

## Go concurrency

### On the up side

- Easy, safe, built-in concurrency

### On the down side

- There is no downside

## Demonstration

### Input script

```
setgridsize(512, 256, 1)
setcellsize(5e-9, 5e-9, 5e-9)
ext_makegrains(40e-9, 256, 0)

Aex      = 10e-12    // J/m
Msat    = 600e3     // A/m
alpha   = 0.1
m       = uniform(0, 0, 1)

// set random parameters per grain
for i:=0; i<256; i+>{
    AnisU.SetRegion(i, vector(0.1*(rand()-0.5), 0.1*(rand()-0.5), 1))

    for j:=i+1; j<256; j+{
        ext_scaleExchange(i, j, rand())
    }
}

// Write field
f      := 0.5e9      // Hz
B_ext  = sin(2*pi*f*t)

// spin UP and write
```

## Demonstration

standardproblem4 Paused

▼ console

▼ mesh

gridsize:	256	x	128	x	1	cells
cellsize:	3.90625e-9	x	3.90625e-9	x	5e-9	m <sup>3</sup>
PBC:	0	x	0	x	0	repetitions
worldsize:	1000	x	500	x	5	nm <sup>3</sup>

mesh up to date

▼ geometry

▼ initial m

▼ solver

Type: heun

Run	1e-9	s	step:	17797	fixdt:	0	s
Steps	1000		time:	6.892992e-09 s	mindt:	0	s
Break			dt:	1.937301e-13 s	maxdt:	0	s
			err/step:	5.378718e-05	maxerr:	0.0001	/step
			MaxTorque:	3.517232e-02 T			

▼ display

m Reduced magnetization (unit length)



## Demonstration

standardproblem4

localhost:35367

Steps: 1000      dt: 1.937301e-13 s      mindt: 0 s  
Break      maxdt: 0 s      MaxTorque: 3.517232e-02 T      maxerr: 0.0001 /step

▼ display

m Reduced magnetization (unit length)

z-slice: [ ] zoom out: [ ]



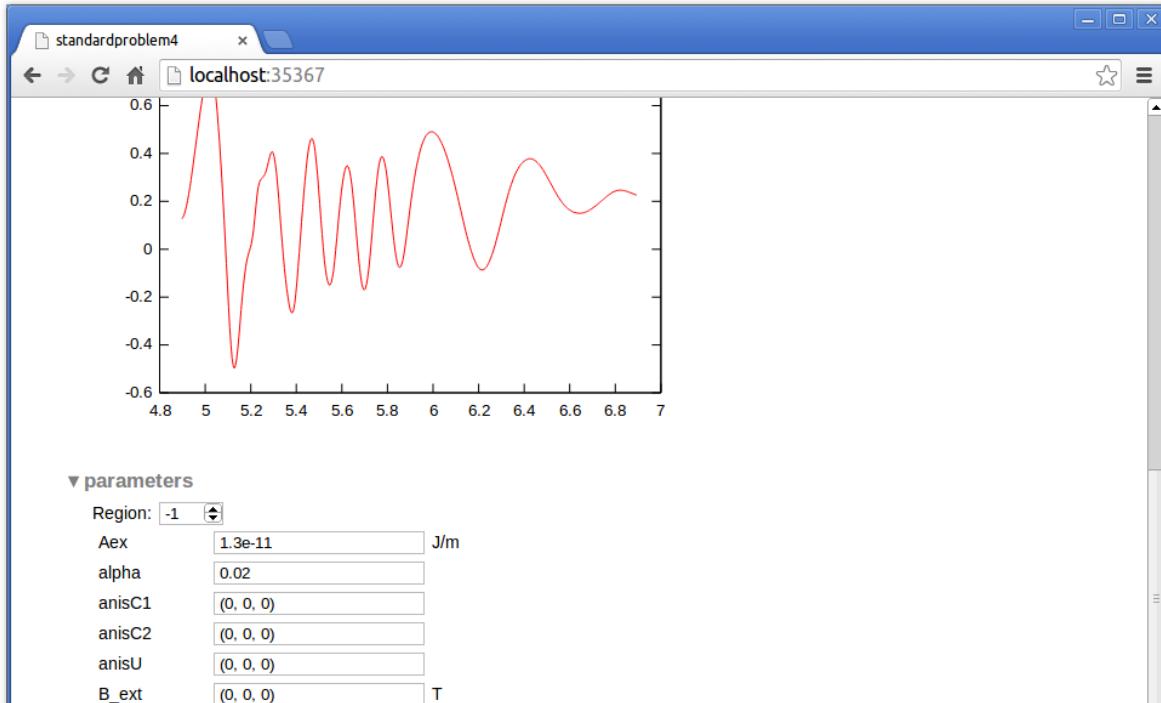
▼ gnuplot

Plot of "table.txt", provided table is being autosaved and gnuplot installed.

plot "table.txt" using  with lines



## Demonstration



## Demonstration

Hard disk magnetization (white = up = 1, black = down = 0)

## Thank you

Arne Vansteenkiste

Ghent University

[Arne.Vansteenkiste@Ugent.be](mailto:Arne.Vansteenkiste@Ugent.be) (<mailto:Arne.Vansteenkiste@Ugent.be>)

<http://mumax.github.io> (<http://mumax.github.io/>)