

GPU computing and R

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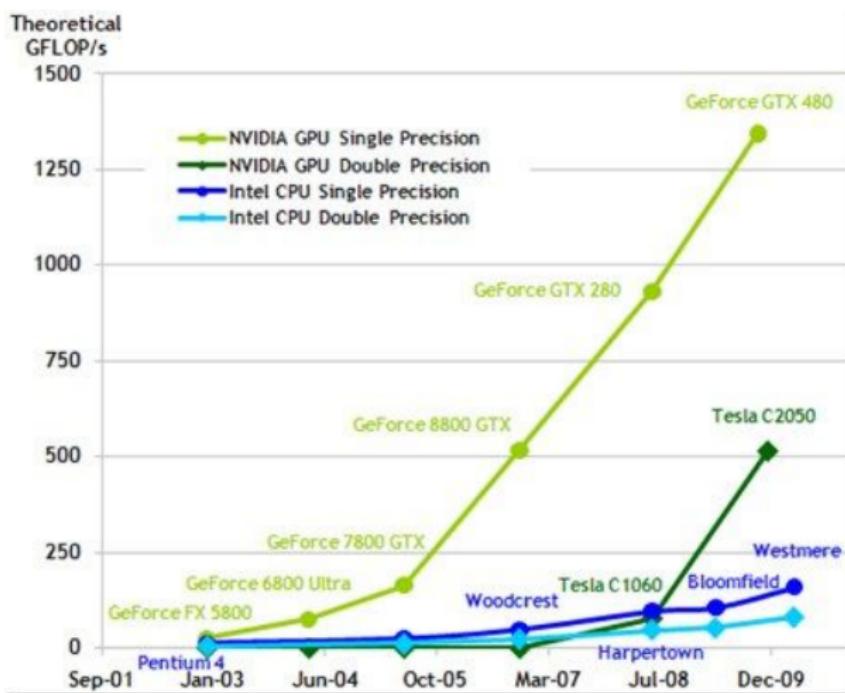
Introduction to GPU computing

GPU computing and R

Introducing ROpenCL

ROpenCL example

Why GPU computing?



<http://theinf2.informatik.uni-jena.de/Lectures/Programming+with+CUDA/SS+2011.html>

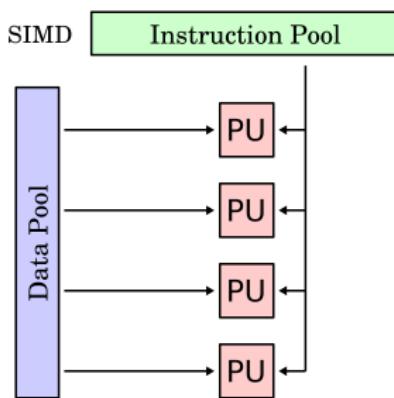


When to use GPU computing

GPU's are specifically well suited for:

- ▶ (Large) matrix operations

Streaming processors are SIMD (Same Instruction Multiple Data)



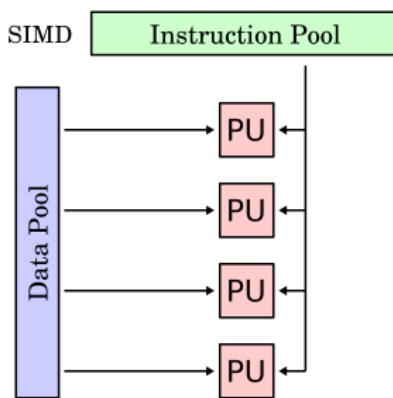
<http://en.wikipedia.org/wiki/File:SIMD.svg>

When to use GPU computing

GPU's are specifically well suited for:

- ▶ (Large) matrix operations
- ▶ (Embarrassingly) parallel operations

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Introduction to GPU computing and OpenCL

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- ▶ CUDA and FireStream were developed to aid developers.
- ▶ Problem they are brand specific
- ▶ Enter OpenCL

What is OpenCL

OpenCL is an industry standard framework for programming computers composed of a combination of CPU's, GPU's and other processors.

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- ▶ Write code that runs on multiple (multi core) platforms e.g. GPU and CPU.

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- ▶ cudaBayesreg

gputools

Provides GPU implementations of various statistical algorithms.
Restricted to NVidia cards, because it uses CUDA.

rgpu

R/GPU is a user-friendly package that can evaluate any given R expression by making transparent use of an NVIDIA Graphics Processing Unit (GPU) through CUDA.
Not actively developed anymore.

cudaBayesreg

Implements the `rhierLinearModel` from the `bayesm` package using nVidia's CUDA language and tools to provide high-performance statistical analysis of fMRI voxels.

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- ▶ Execute kernels in the right order on the right components
- ▶ Collect the results

Introducing ROpenCL

ROpenCL is an R library which provides a **user friendly** interface to the **OpenCL** library.

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- ▶ And I like it!

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PyOpenCL for R

- ▶ R deserves a library like that
- ▶ Like Rcpp for OpenCL
- ▶ No need to worry about memory management, let ROpenCL manage it for you

Code or didn't happen : Vector addition

```
// set and log Global and
Local work size dimensions
szLocalWorkSize = 256;
szGlobalWorkSize =
    shrRoundUp((int)szLocalWorkSize, iNumElements);

// Allocate and initialize host arrays
srcA = (void *)malloc(sizeof(
    cl_float) * szGlobalWorkSize);
srcB = (void *)malloc(sizeof(
    cl_float) * szGlobalWorkSize);
dst = (void *)malloc(sizeof(
    cl_float) * szGlobalWorkSize);
shrFillArray((float*)srcA, iNumElements);
shrFillArray((float*)srcB, iNumElements);
```

```
library(ROpenCL)

a <- seq(11444777)/10
b <- seq(11444777)
out <- rep(0, length(a))
localWorkSize = 256
globalWorkSize =
    ceiling(length(a)/localWorkSize)*localWorkSize
```

Code or didn't happen

```
//Get an OpenCL platform
ciErr1 = clGetPlatformIDs(1, &cpPlatform, NULL);
//Get the devices
ciErr1 = clGetDeviceIDs(cpPlatform,
    CL_DEVICE_TYPE_GPU, 1, &cdDevice, NULL);
//Create the context
cxGPUContext = clCreateContext(0, 1, &cdDevice,
    NULL, NULL, &ciErr1);
// Create a command-queue
cqCommandQueue = clCreateCommandQueue(cxGPUContext,
    cdDevice, 0, &ciErr1);
// Allocate the OpenCL buffer memory objects
// for source and result on the device GMEM
cmDevSrcA = clCreateBuffer(cxGPUContext,
    CL_MEM_READ_ONLY, sizeof(cl_float)
    * szGlobalWorkSize, NULL, &ciErr1);
cmDevSrcB = clCreateBuffer(cxGPUContext,
    CL_MEM_READ_ONLY, sizeof(cl_float)
    * szGlobalWorkSize, NULL, &ciErr2);
cmDevDst = clCreateBuffer(cxGPUContext,
    CL_MEM_WRITE_ONLY, sizeof(cl_float)
    * szGlobalWorkSize, NULL, &ciErr2);
```

```
#Get an OpenCL platform
platformIDs <- getPlatformIDs()
#Get the devices
deviceIDs <- getDeviceIDs(platformIDs[[1]])
#Create the context
context <- createContext(deviceIDs[[1]])
#Create a command-queue
queue <- createCommandQueue(context, deviceIDs[[1]])
#Allocate the OpenCL buffer memory objects
    for source and result on the device GMEM
inputBuf1 <- createBuffer(context,
    "CL_MEM_READ_ONLY", globalWorkSize, a)
inputBuf2 <- createBuffer(context,
    "CL_MEM_READ_ONLY", globalWorkSize, b)
outputBuf1 <- createBufferFloatVector(context,
    "CL_MEM_WRITE_ONLY", globalWorkSize)
```

Code or didn't happen

```
// Create the program
cpProgram = clCreateProgramWithSource(cxGPUContext,
    1, (const char **)&cSourceCL, &szKernelLength,
    &ciErr1);
ciErr1 = clBuildProgram(cpProgram, 0, NULL,
    NULL, NULL, NULL);
// Create the kernel
ckKernel = clCreateKernel(cpProgram, "VectorAdd",
    &ciErr1);
// Set the Argument values
ciErr1 = clSetKernelArg(ckKernel, 0,
    sizeof(cl_mem), (void*)&cmDevSrcA);
ciErr1 |= clSetKernelArg(ckKernel, 1,
    sizeof(cl_mem), (void*)&cmDevSrcB);
ciErr1 |= clSetKernelArg(ckKernel, 2,
    sizeof(cl_mem), (void*)&cmDevDst);
ciErr1 |= clSetKernelArg(ckKernel, 3,
    sizeof(cl_int), (void*)&iNumElements);
```

```
kernel <- "__kernel void VectorAdd(__global const
    float* a, __global const int* b,
    __global float* c, int iNumElements)
{
    // get index into global data array
    int iGID = get_global_id(0);
    // bound check (equivalent to the limit on a
    // 'for' loop for standard/serial C code
    if (iGID >= iNumElements)
    {
        return;
    }
    // add the vector elements
    c[iGID] = a[iGID] + b[iGID];
}"
kernel <- createProgram(context, kernel,
    "VectorAdd", inputBuf1, inputBuf2,
    outputBuf1, length(out))
```

Code or didn't happen

```
// Asynchronous write of data to GPU device
ciErr1 = clEnqueueWriteBuffer(cqCommandQueue,
    cmDevSrcA, CL_FALSE, 0, sizeof(cl_float)
    * szGlobalWorkSize, srcA, 0, NULL, NULL);
ciErr1 |= clEnqueueWriteBuffer(cqCommandQueue,
    cmDevSrcB, CL_FALSE, 0, sizeof(cl_float)
    * szGlobalWorkSize, srcB, 0, NULL, NULL);
// Launch kernel
ciErr1 = clEnqueueNDRangeKernel(cqCommandQueue,
    ckKernel, 1, NULL, &szGlobalWorkSize,
    &szLocalWorkSize, 0, NULL, NULL);
// Synchronous/blocking read of results, and
// check accumulated errors
ciErr1 = clEnqueueReadBuffer(cqCommandQueue,
    cmDevDst, CL_TRUE, 0, sizeof(cl_float)
    * szGlobalWorkSize, dst, 0, NULL, NULL);
```

```
enqueueWriteBuffer(queue, inputBuf1,
    globalWorkSize, a)
enqueueWriteBuffer(queue, inputBuf2,
    globalWorkSize, b)
enqueueNDRangeKernel(queue, kernel,
    globalWorkSize, localWorkSize)
result <- enqueueReadBuffer(queue,
    outputBuf1, globalWorkSize, out)
```

OpenCL

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- ▶ A little over a week ago the OpenCL package has been published on CRAN by Simon Urbanek.
- ▶ And currently it seems to be a very thin layer around OpenCL.
- ▶ The goal of ROpenCL is to abstract a little more, like PyOpenCL.

Contact details



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```
install.packages("ROpenCL", repos = "http://repos.openanalytics.eu", type = "source")
```