## PFC Opencl based machine learning labeling of biomecical datasets.

## Project goals

- Learn Opence
- Test Opence
- Improve performance (redice execution time) of a Medical inaging program


## Project scope

## NVIDIA architecture



ATl architecture

## Project environment

- Adaboost algorithmifor classifying datasets
- Medical inaging techniques to visualize body datasets


PHOTORESEARCHERS

## Algorithm

* affaclass - rciji]
- polaritat $\mathrm{gc}[\mathrm{jij}$;

4 thresh bocitit255:
4
4- validáa fabs (polaritat) 000001

- pasitui (polaritat $>0.5$ );

4 mesgran = is thresh;
 talfaclass);

## OpenCL: NDRange space

NDRange space

| Work Group 0,0 |  |  |  | Work Group 1,0 |  |  |  |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
| WI 0,0 | WI 1,0 | WI 2,0 | WI 3,0 | WI 0.0 | WI 1,0 | WI 2,0 | WI 3,0 |
| WI | WI | WI | WI | WI | WI | WI | WI |
| 0,1 | 1,1 | 2,1 | 3,1 | 0,1 | 1,1 | 2,1 | 3,1 |
| WI | WI | WI | WI | WI | WI | WI | WI |
| 0,2 | 1,2 | 2,2 | 3,2 | 0,2 | 1,2 | 2,2 | 3,2 |
| WI | WI | WI | WI | WI | WI | WI | WI |
| 0,3 | 1,3 | 2,3 | 3,3 | 0,3 | 1,3 | 2,3 | 3,3 |


| Work Group 0,1 |  |  |  |
| :---: | :---: | :---: | :---: |
| WI | WI | WI | WI |
| 0,0 | 1,0 | 2,0 | 3,0 |
| WI | WI | WI | WI |
| 0,1 | 1,1 | 2,1 | 3,1 |
| WI | WI | WI | WI |
| 0,2 | 1,2 | 2,2 | 3,2 |
| WI | WI | WI | WI |
| 0,3 | 1,3 | 2,3 | 3,3 |


| Work Group 1,1 |  |  |  |
| :---: | :---: | :---: | :---: |
| WI | WI | WI | WI |
| 0,0 | 1,0 | 2,0 | 3,0 |
| WI | WI | WI | WI |
| 0,1 | 1,1 | 2,1 | 3,1 |
| WI | WI | WI | WI |
| 0,2 | 1,2 | 2,2 | 3,2 |
| WI | WI | WI | WI |
| 0,3 | 1,3 | 2,3 | 3,3 |

## GPGPU architecture

## Global Memory



## NVIDIA Fermi architecture



## NVIDIA Fermi architecture



## ATI architecture



## ATI architecture



## OpenCL: code example

## ATI thread and

 instruction level parallelism codeNVDIA thread level parallelism code
kernel void operation( __global float4 input, _global float4 output K
output[get_local_id(0)]= input[get_local_id(0)]+input[get_local_id(0)];
kernel void operation( __global float input, _global float output ) \{
output[get_local_id(0)]= input[get_local_id(0)]+input[get local_id(0)];

# OpenCl medical imaging implementation 

- Reducing PGle usage to $1 / 64$ th
- Adding an extra kernel
- First step to test and validate


## OpenCL implementation



## OpenCL implementation

## \#define BLOCK_DIMX 32 <br> \#define BLOCK_DIMY 4

- Gradient Kernel

kernel void kernelGradient( ( _ global unsigned char* values, _global float* gradients, local float* local_gradients, int dimx, int dimy, int dimz) \{
__local float tile[BLOCK_DIMY + 2 ][BLOCK_DIMX + 2 ]; float infront;
float behind;
//fill tile and infront
for ( $k=1$; $k<\operatorname{dimz-1}$; $k++$ )
\{
behind = tile[get_local_id(1)][get_local_id(0)-1];
tile[get_local_id(1)][get_local_id(0)+1] = infront;
// calculate gradients
// copy results from registers
// to local memory (4 transfers per thread 128x4, 1024 lost cycles)
// copy results from local to global memory
// (4 transfers each 16 threads $8 \times 4,19200$ lost cycles)
// we loose 20.224 cycles instead of $\mathbf{3 0 7 . 2 0 0}$ cycles
\}
\}


## OpenCL implementation

- Gradient Kernel



## OpenCL implementation

* Classification kernel
_kernel void kernelClassif( __global unsigned char* values,
global float* gradients,
_global unsigned char* voxelsOut,
_global float* rc, global float* gc,
_global float* bc, constant int* var, __local float* sumArr, _local float* sumalfaArr, __local float* thisVoxel) \{
// Read alfaclass, polaritat, thresh
// Fill thisVoxel first 2 values
for $(z=1 ; z<(\operatorname{var}[4]-1) ;++z)\{$
// Fill the next 6 thisVoxel values
// Copy the first 8 thisVoxel values to the last 8 thisVoxel positions
// Calculate two matrix of $\mathbf{2 4 0}$ values and do reduction for each one
voxelsOut[get_group_id(0)+(get_num_groups(0)*get_group_id(1))+(get_num_groups(0)*get_num_groups(1)*(z-1))]= // Compare the two final values to obtain 1 or 0
\}


## OpenCL implementation

- Classification kernel



## OpenCL implementation

- Results


Times for different cards


## OpenCL analysis

* Stencil method for wave propagation simulations



## OpenCL analysis

* Based on Micikevicius GPU algorithm



## OpenCL performance

* Comparing CUDA with GPU Openc



## OpenCL portability

- Testing a working and tuned code for NVIDA in an ATI card



## OpenCL for ATI

* Exploring ATI architecture



## Architectural trends

- CPU and GPU integration
- Inte Sandy Bridge: CPU
- AMD AP (Accelerated processing Unit)
- Lots of stantups Tilera, Zil Smooth Stone...
- Low power consumption for HPC (GPUS,ARM, Godson)


## Conclusions

- OpenCE for GPUs is a good option for data parallel
- GPU's for HPG have stil a way to walle
- Possible convergence of GPU and GPU architectures
- What is sure is that future looks parallel

