Performance prediction on heterogeneous systems using statistical methods

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Purpose of this thesis

- Predicting performance on heterogeneous systems
- Creating the prediction model
- Facilitating the mapping of applications on various computational resources

Our Approach

- Collection of hardware events
 - o PAPI
 - Nvprof
- Use of Statistical methods
 - Linear Regression
 - Neural Networks
 - Random Forests

Background Issues

Architectural differences

CPUs:

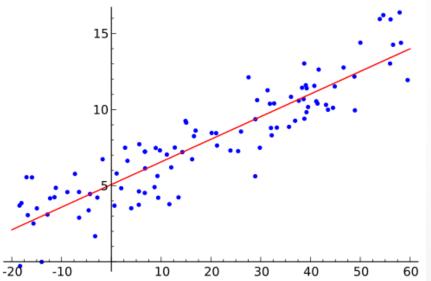
- few processors supporting ~1 hardware thread
- execute one stream of instructions as fast as possible
- latency: large caches and branch prediction hardware
- a few hundreds GFLOPS

• GPUs:

- many processors supporting many hardware threads
- o execute many parallel streams of instructions as fast as possible
- o latency: supporting thousands of threads at once
- thousands of GFLOPS

Linear Regression

- Linear relationship
- Equation form: $y = X\beta + \epsilon$

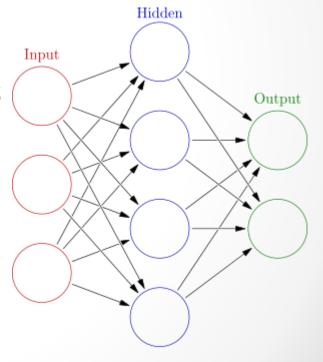


Neural Networks

- Interconnected nodes, forming layers
- Each node performs a mathematical function, such as

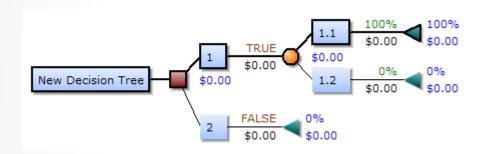
$$S(t) = \frac{1}{(1+e^{-t})}$$

Black box



Random Forests

A combination of decision trees



- Node → Test on an attribute
- Branch → Result of a test
- Leaf → Decision taken

Data Collection

Benchmarks

- Rodinia benchmark suite 2.4
- The Scalable HeterOgeneous Computing (SHOC)
 Benchmark Suite

Intel Xeon CPUs

- Use of OpenCL
- Get number of compute units
- Use of low level PAPI Interface for events
- OpenCL built in functions for time

Event measurement example

NVIDIA GPUs

- NVprof: not supporting OpenCL
- Used CUDA
- CL Interface

Model Formation

 Find correlation between target value and measured variables

Intel to NVIDIA		NVIDIA to Intel	
Counter	Correlation	Counter	Correlation
BR_TKN	0,8874	ldst_executed	0,88
TOT_CYC	0,868	inst_issued	0,88
BR_PRC	0,862	issue_slots	0,88
TOT_INS	0,861	inst_executed	0,86
LD_INS	0,8553	ldst_issued	0,86
BR_CN	0,8543	cf_issued	0,85
LST_INS	0,8503	cf_executed	0,85
BR_INS	0,8473	12_read_trans	0,83

- Linear regression
- Tested different number of hardware events.
 Calculated RMSE

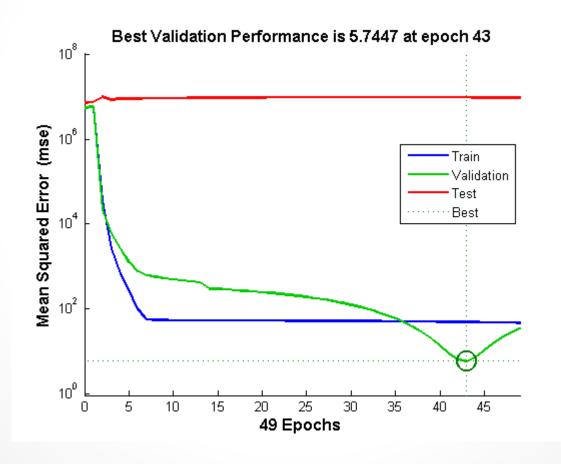
Number of Events Used	RMSE	
9	5,12	
8	7,29	
7	7,17	
6	8,86	
5	229	

Linear regression formula:

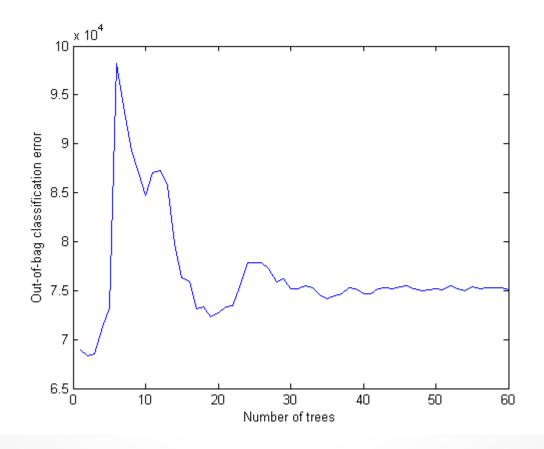
Execution time =

$$\begin{bmatrix} -2.9 \times 10^{-2} & -1.24 \times 10^{-3} & -1.12 & 7.865 \times 10^{-3} & -4.81 \times 10^{-3} & 1.08 \end{bmatrix} \times \begin{bmatrix} BR_IKN \\ TOT_CYC \\ BR_PRC \\ TOT_INS \\ LD_INS \\ BR_CN \end{bmatrix} \times 10^{-6} + 2.0829$$

Neural Networks



Random forests



Linear regression

Number of Events Used	RMSE
10	92,6
9	99,1
8	110
7	110
6	109
5	115

Linear regression formula:

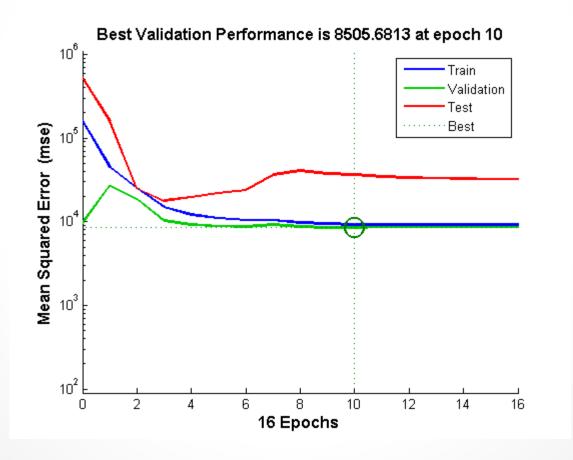
Execution time =

$$\begin{bmatrix} -2.12 & 65.06 & -65.9 & 1.07 & 0.4 & 7.96 & -7.89 & 0.37 & -35.8 & 43.4 \end{bmatrix} \times \begin{bmatrix} inst_issued \\ issue_slots \\ inst_executed \\ ldst_issued \\ cf_issued \\ cf_executed \\ l2_read_trans \\ l2_write_trans \\ dram_write_trans \end{bmatrix} \times 10^{-5}$$

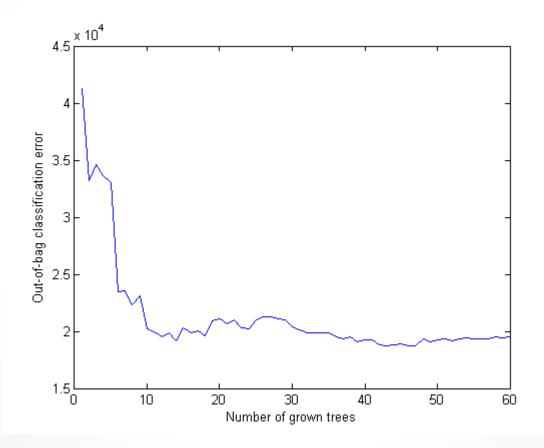
ldst_executed

+14.791

Neural Networks

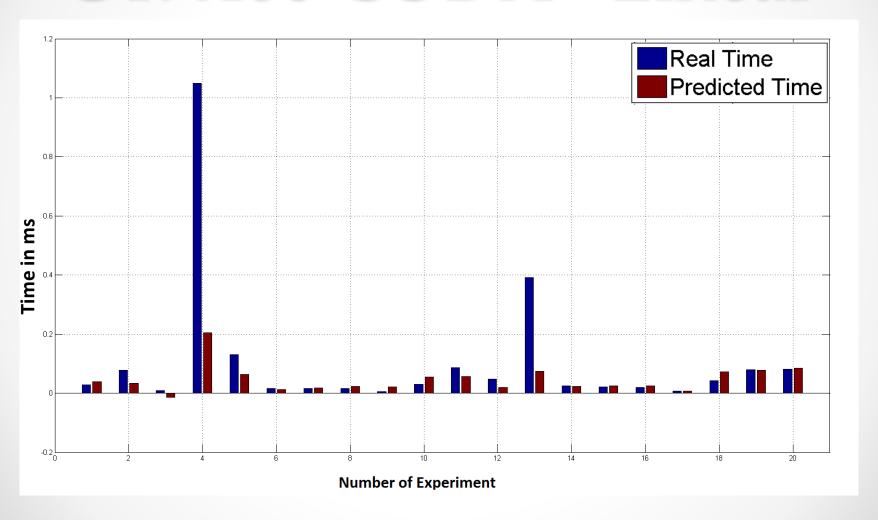


Random Forests

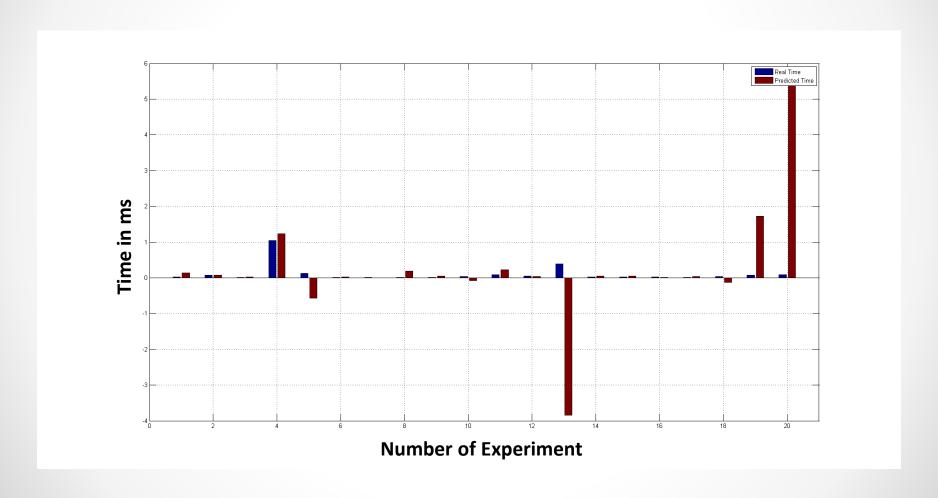


Evaluation

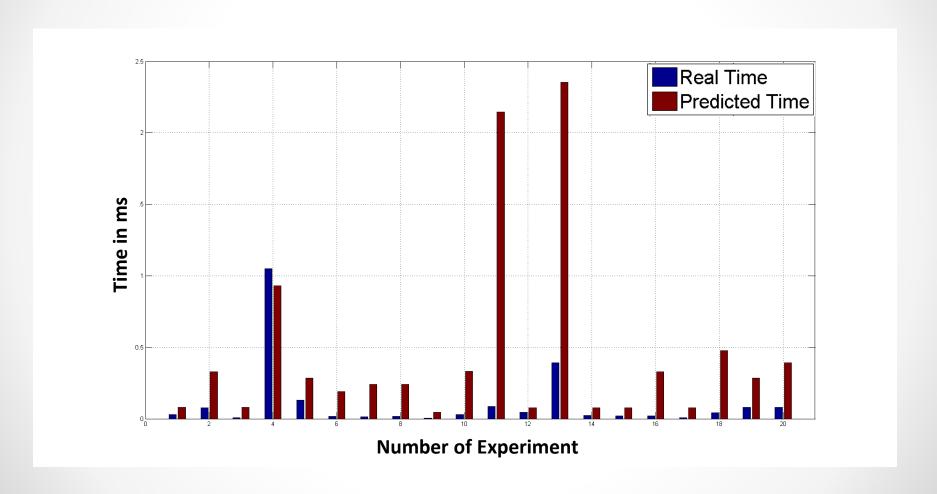
Intel Xeon OpenCL to GTX480 CUDA - Linear



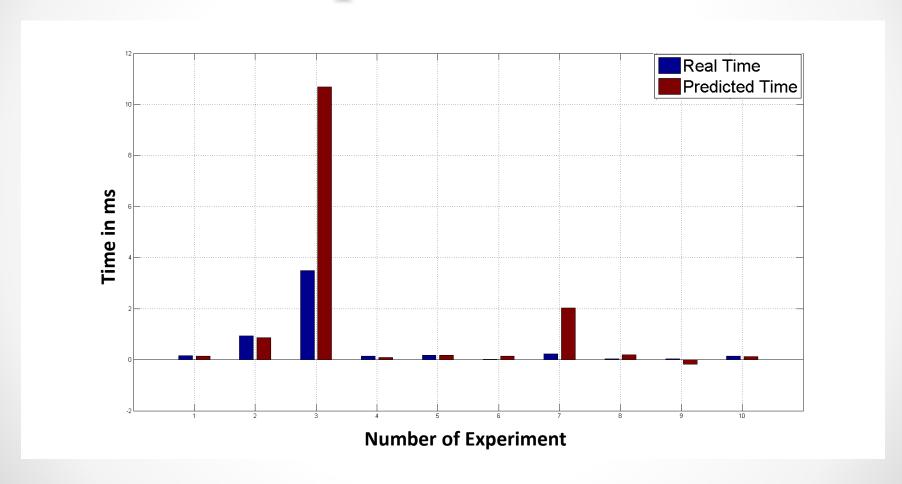
Intel Xeon OpenCL to GTX480 CUDA - NN



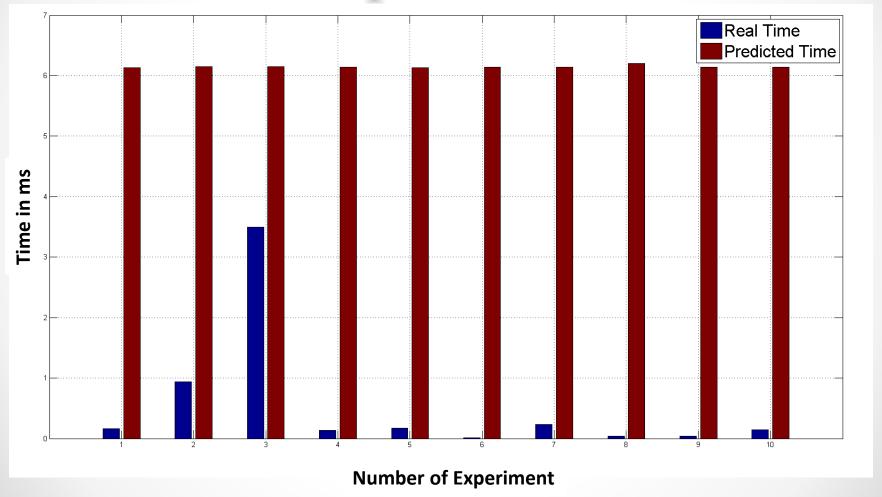
Intel Xeon OpenCL to GTX480 CUDA - RF



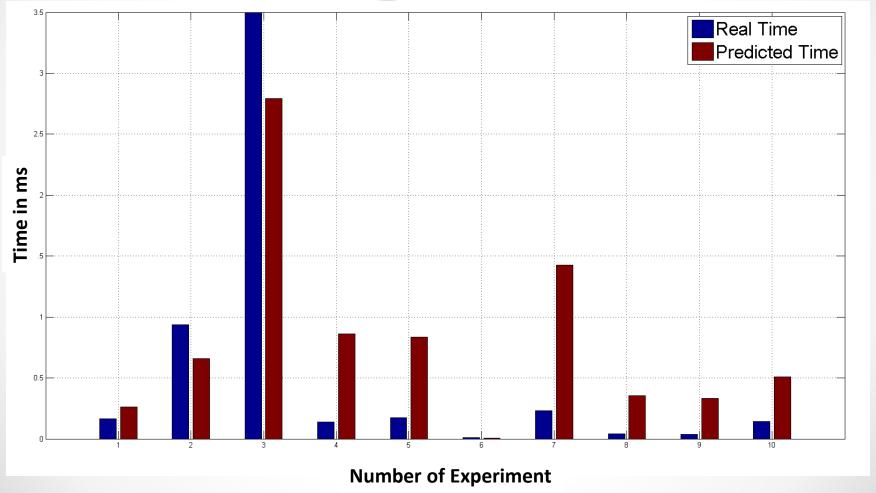
GTX480 CUDA to Intel Xeon OpenCL - Linear



GTX480 CUDA to Intel Xeon OpenCL - NN



GTX480 CUDA to Intel Xeon OpenCL - RF



Future work

- Predict execution time on ATI GPUs
- Increase size of training set
- Predict power
- Use our models with a run-time system

Questions