

# Multi- and Many-core Architectures for Scientific Programming

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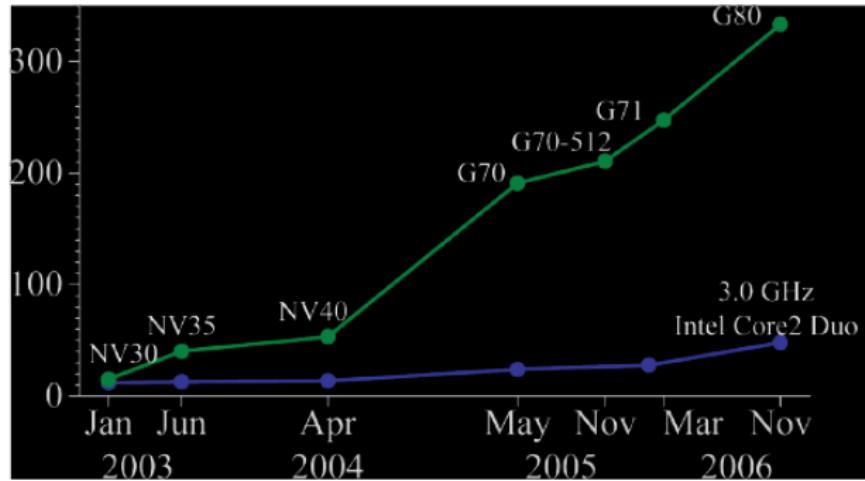
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Workshop Commissione Calcolo e Reti INFN 2009  
May 11 – 15 2009, Palau Olbia

- 1 A Look Inside NVIDIA GT200 GPU
- 2 Cell and Nehalem Multi-core Architectures
- 3 Where are we going ?

# Motivation: GPU Evolution

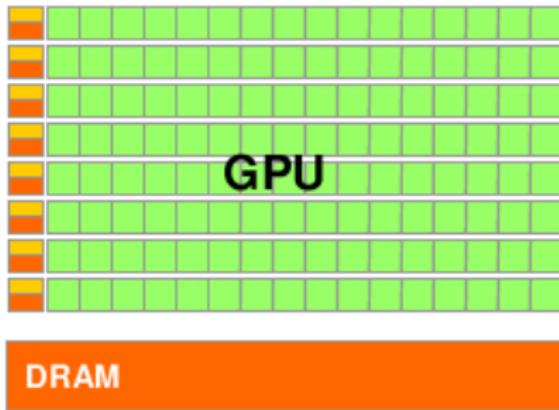
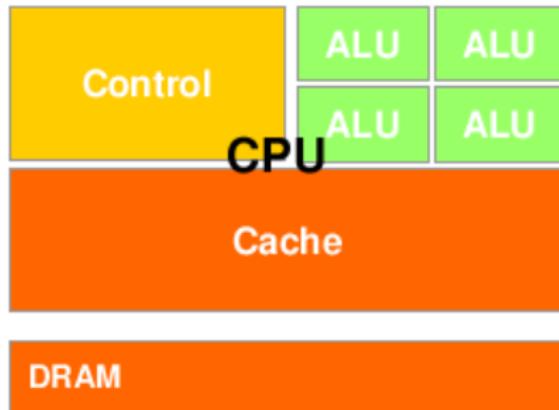
GFLOPS



G80 = GeForce 8800 GTX  
G71 = GeForce 7900 GTX  
G70 = GeForce 7800 GTX  
NV40 = GeForce 6800 Ultra  
NV35 = GeForce FX 5950 Ultra  
NV30 = GeForce FX 5800

- GPUs evolve much faster in terms of raw-computing power
- Fast-growing video-game market forces innovation

# GPUs vs CPUs



- GPUs specialized for highly data-parallel and intensive computation (exactly what rendering is about)
- more transistors devoted to **data-processing** rather than **data caching** and **flow-control**

# NVIDIA GT200 Specs

## Processor:

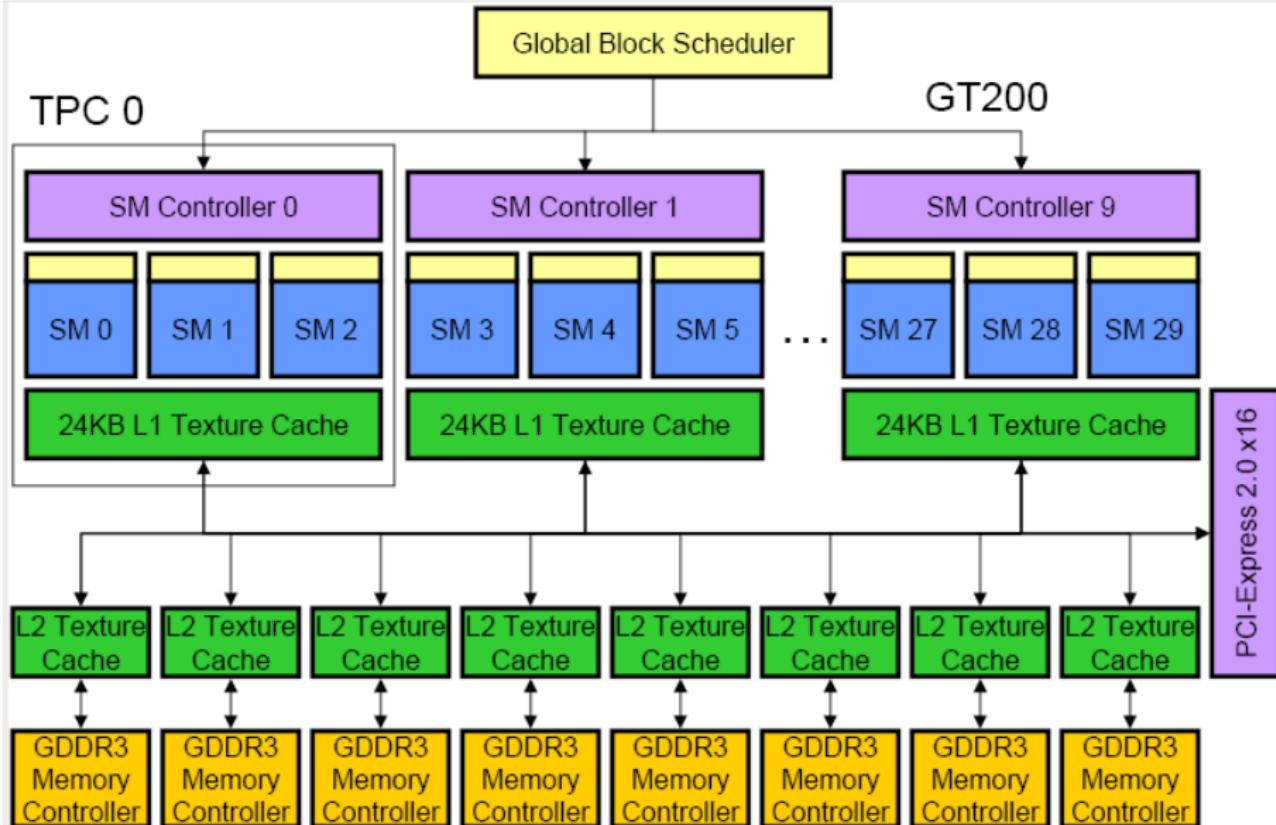
- 1.3 GHz, **936 Gflops SP**, 78 Gflops DP
- 10 **Texture Processor Cluster** (TPC)
- 1 TPC includes 3 **Streaming Multiprocessor** (SM) + 1 Texture Memory
- 1 SM include 8 **Streaming Processor** (SP)  
loosely corresponding to a modern CPU-core with 8-way SIMD computing-unit
- a total of  $10 \times 3 \times 8 = 240$  **threads**

## Memory:

- 4GB Global Memory, 512-bit, 102.4 GB/s, 400 ... 600 cycles of latency
- Constant memory 64 KB (RO)
- Texture memory 256 KB (RO, two dimensional locality)

160 Watt (typical, w/Memory), < 6 Gflops/Watt (SP, w/MUL), 1.5 K €

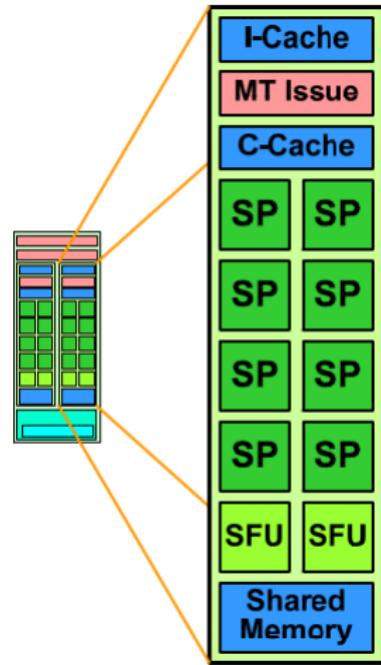
# Architecture of GP-GPU: NVIDIA GT200



# NVIDIA GT200 SM Specs

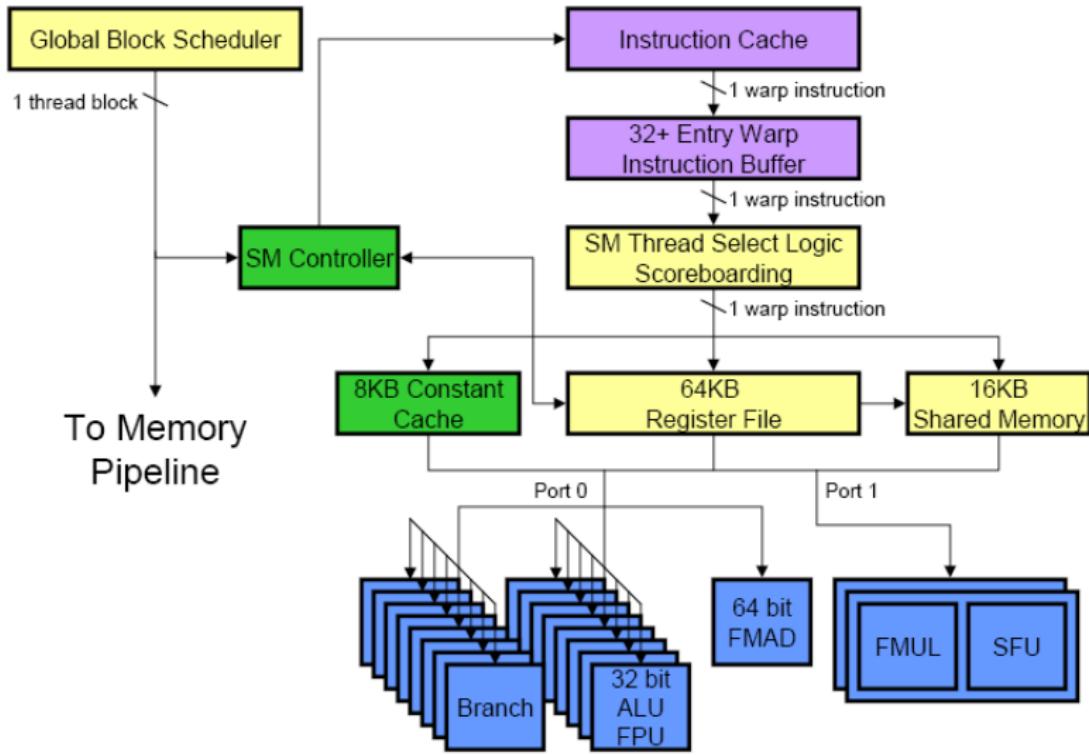
Each SM includes:

- 64KB register file (2KB for each SP)
- 16KB shared memory
- 8KB constant cache
- 8 32-bit ALU/FPU
- 1 64-bit FMAD
- 8 branch units

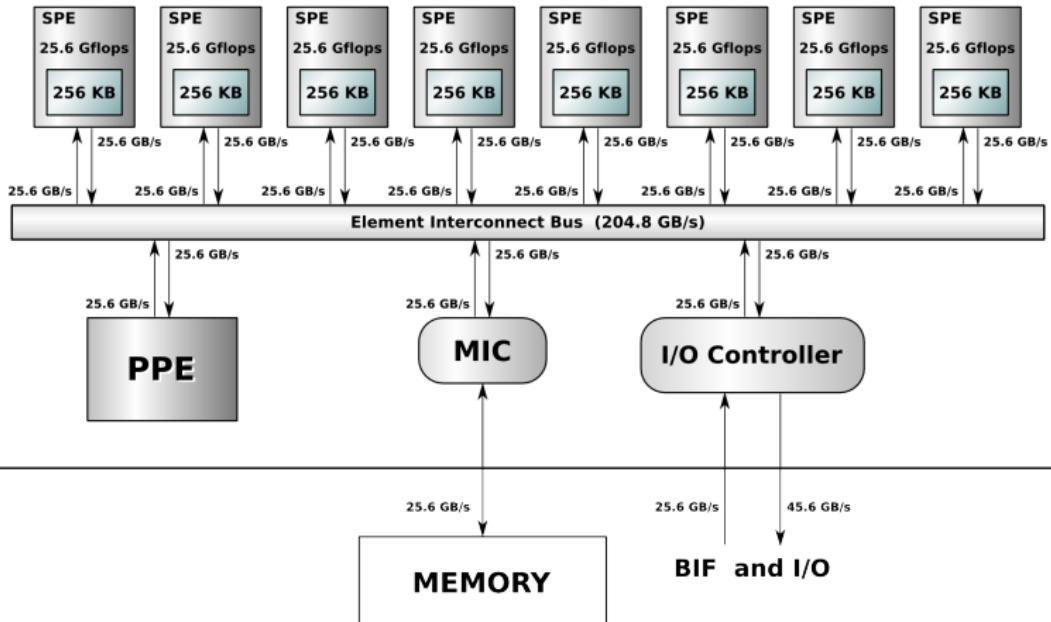


1 SM executes in parallel up to 8 thread and manages up to 1024 threads

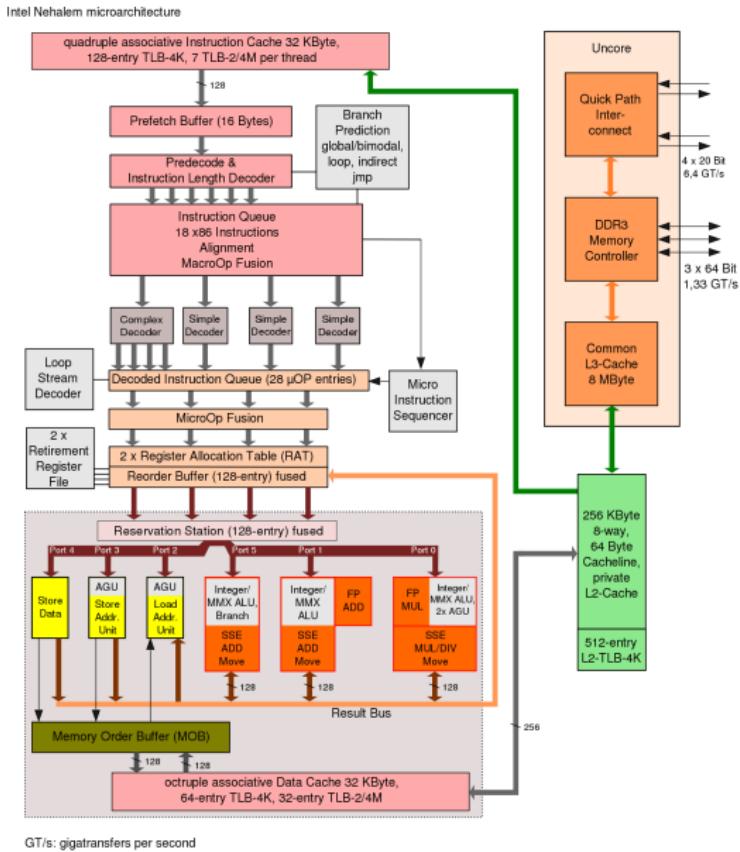
# The SM Architecture



# The Multicore Cell-BE Architecture



# The Multicore Nehalem Architecture

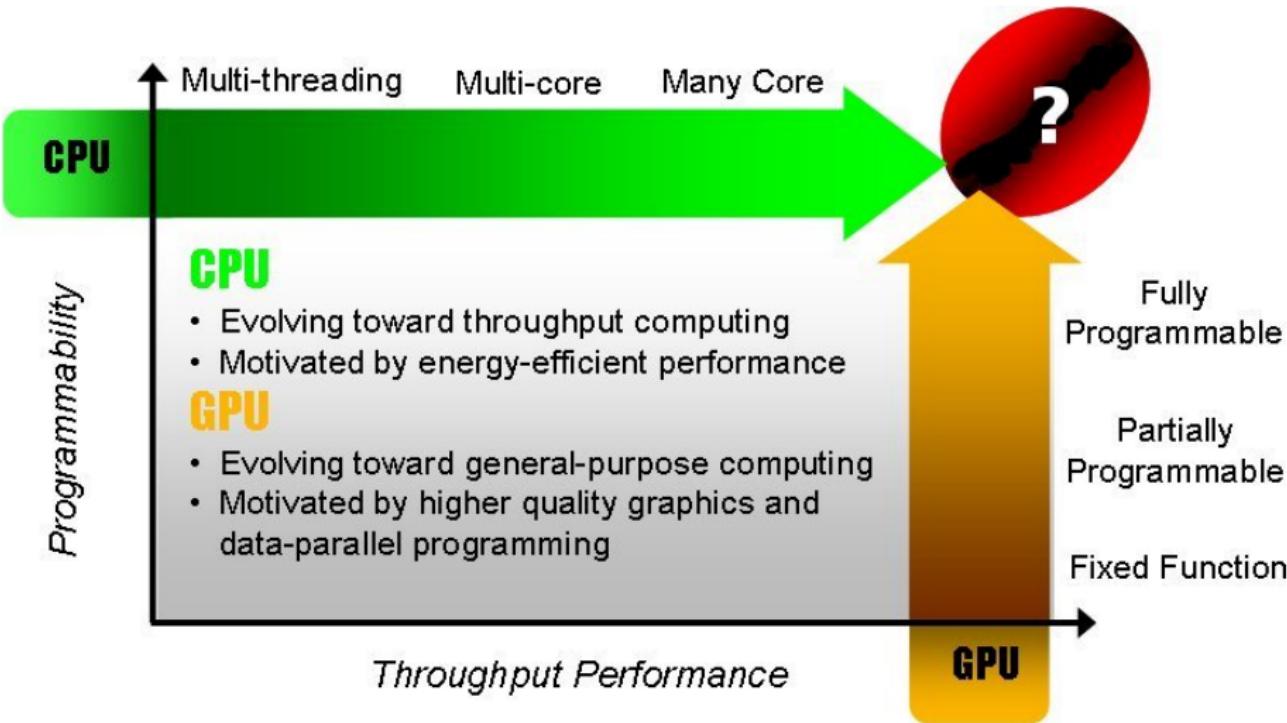


# Multicore Performance for Spin-Glass Simulations

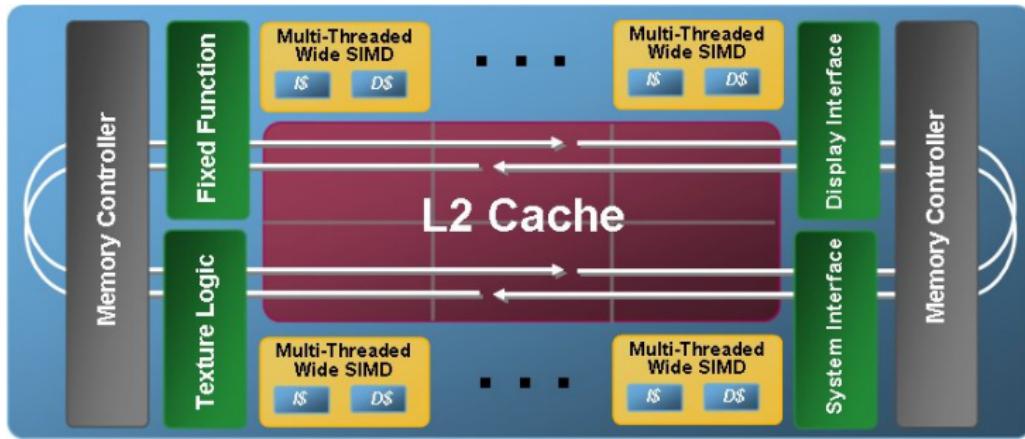
		Binary model				Gaussian model						
<i>L</i>	<i>C</i>	<i>V</i>	<i>w</i>	Mem	SUT	GUT	<i>V</i>	<i>w</i>	Mem	SUT	GUT	
16	8	8	16	local	0.83	0.052	4	1	local	1.00	0.250	
16	16		16	local	1.17	0.073	4	1	local	1.34	0.335	
32	8		16	local	0.40	0.050	4	1	local	0.65	0.162	
32	16		16	local	0.26	0.032	4	1	local	0.42	0.105	
48	8		8	local	0.48	0.030	4	1	global	1.65	0.412	
48	16		8	local	0.25	0.016	4	1	local	0.42	0.105	
64	8		32	4	local	0.29	0.072	4	1	global	1.71	0.427
64	16		32	4	local	0.15	0.037	4	1	global	2.23	0.557
80	8		8	local	0.82	0.051	4	1	global	1.59	0.397	
80	16		8	local	1.03	0.064	4	1	global	2.10	0.525	
96	8		16	8	global	0.42	0.052	-	-	-	-	
96	12		16	8	global	0.41	0.051	-	-	-	-	
128	8		64	2	global	0.24	0.120	-	-	-	-	
128	16		64	2	local	0.12	0.060	-	-	-	-	

Performance are 1 order of magnitude far from that of the Janus system for which we have a SUT of 0.016 ns/spin

# Where are we going ?



# The Larrabee Processor



- **in-order** cores, SIMD-16
- 512-bit SIMD vector unit, 32-, 64-bit operations
- 32-bit vector registers
- 8 MB L2 cache shared across the cores

**2 Tflops SP, 2GHz, 32-cores, 300 W (?)**

# How to go Beyond Petaflops ?

- currently scientific project attack Petaflop performance connecting commodity processors by a 3D-torus custom-network (a la APE):
  - ▶ QPACE: German project based on Cell
  - ▶ Aurora: Italian project based on Nehalem
- for the future one possible option would be to use 2-4 Tflops GP-GPU interconnected by a high-scalable 3D-torus network