

Toward GPUs being mainstream in analytic processing

An initial argument using simple scanaggregate queries

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DaMoN 2015

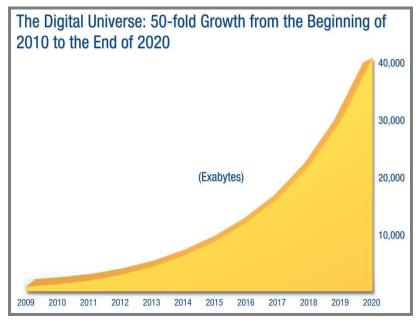
Summary

- GPUs are energy efficient
 - Discrete GPUs unpopular for DBMS
 - New integrated GPUs solve the problems
- Scan-aggregate GPU implementation
 - Wide bit-parallel scan
 - Fine-grained aggregate GPU offload
- Up to 70% energy savings over multicore CPU
 - Even more in the future



Analytic Data is Growing

- Data is growing rapidly
- Analytic DBs increasingly important



Source: IDC's Digital Universe Study. 2012.

Want: High performance

Need: Low energy

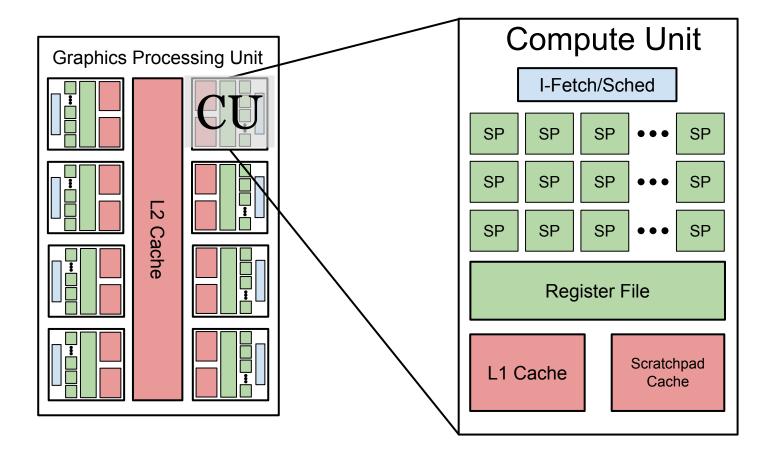


GPUs to the Rescue?

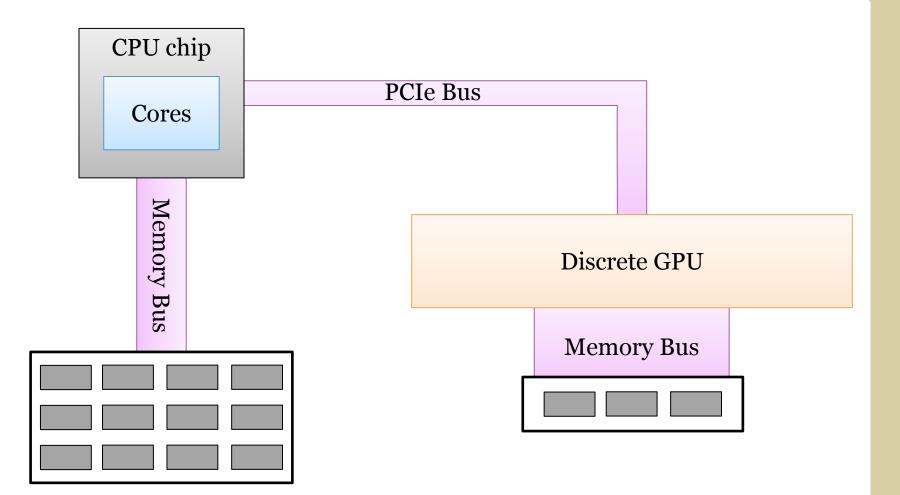
- GPUs are becoming more general
 - Easier to program
 - Integrated GPUs are everywhere
- GPUs show great promise [Govindaraju '04, He '14, He '14, Kaldewey '12, Satish '10, and many others]
 - Higher performance than CPUs
 - Better energy efficiency
- Analytic DBs look like GPU workloads

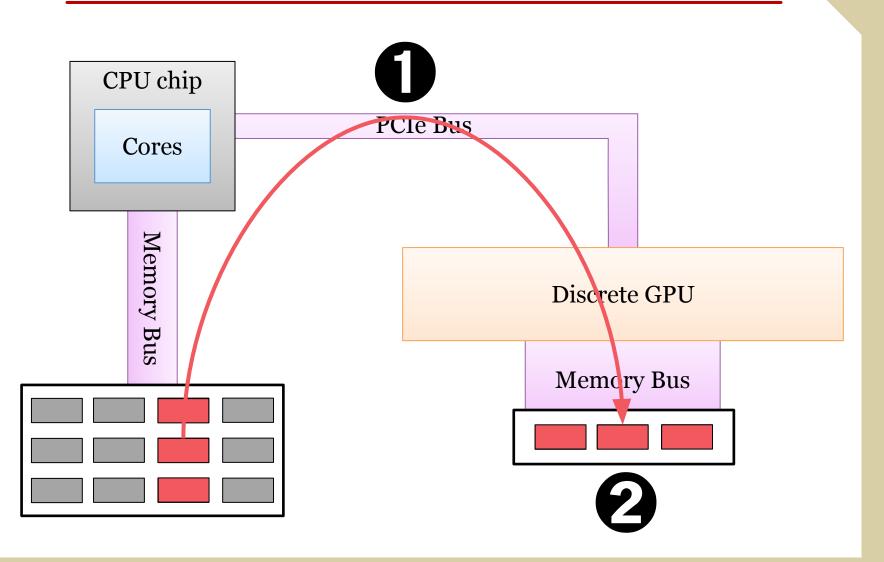


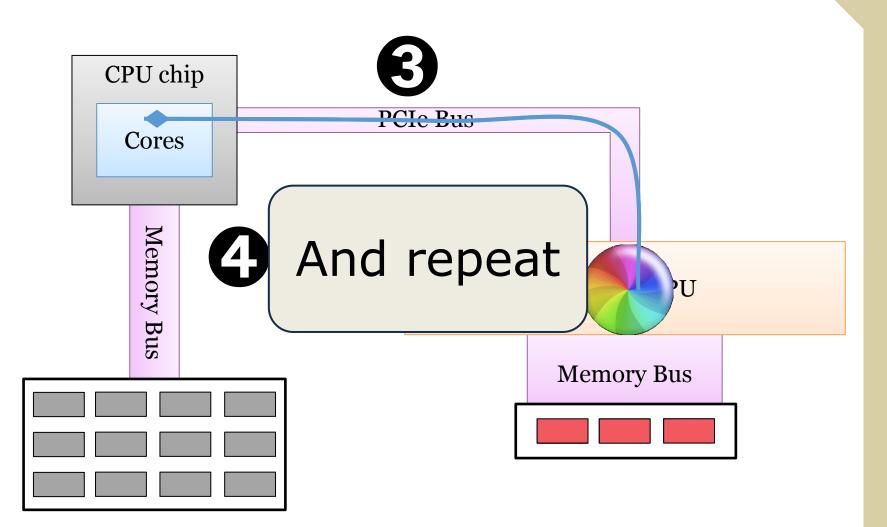
GPU Microarchitecture













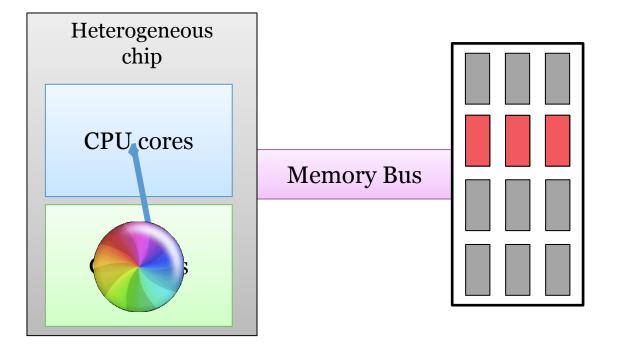
Copy data over PCIe

- Low bandwidth
- High latency
- **2**Small working memory
- \bigcirc High latency user \rightarrow kernel calls
- A Repeated many times

98% of time spent not computing



Integrated GPUs





Heterogeneous System Arch.

- API for tightly-integrated accelerators
- Industry support
 - Initial hardware support today
 - HSA foundation (AMD, ARM, Qualcomm, others)
- No need for data copies
 - Cache coherence and shared address space
- No OS kernel interaction
 - User-mode queues





Outline



Background

Algorithms

- Scan
- Aggregate
- Results



Analytic DBs

- Resident in main-memory
- Column-based layout
- WideTable & BitWeaving [Li and Patel '13 & '14]
 - Convert queries to mostly scans by pre-joining tables
 - Fast scan by using sub-word parallelism
 - Similar to industry proposals [SAP Hana, Oracle Exalytics, IBM DB2 BLU]
- Scan-aggregate queries

Running Example

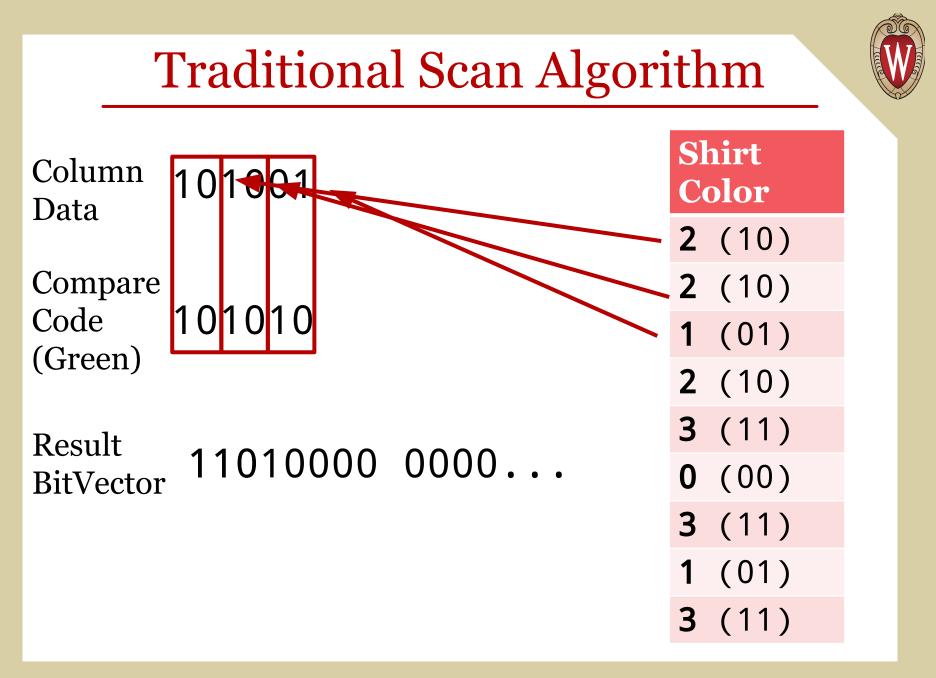
Shirt Color	Shirt Amount
2	1
2	3
1	1
2	5
3	7
0	2
3	1
1	4
3	2

Color	Code
Red	0
Blue	1
Green	2
Yellow	3

Running Example



Shirt Color	Shirt Amount		Count the number of
2	1		green shirts in the
2	3		inventory
1	1		
2	5		Scan the color
3	7	Û	column for green (2)
0	2		
3	1		Aggregate amount
1	4	0	Aggregate amount where there is a match
3	2		where there is a match



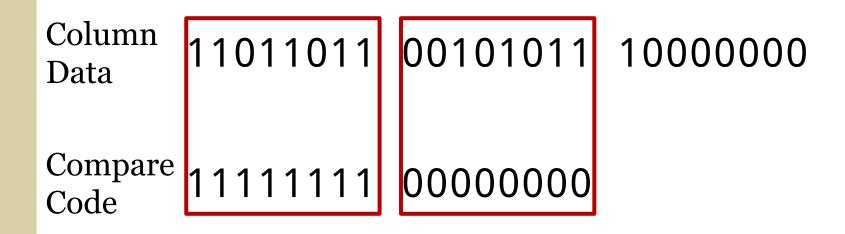


Vertical Layout

	Color		word	co	C1	c2	c3	c 4	c 5	c6	c 7
co	2 (10)	_	\rightarrow								
c1	2 (10)		wo	1	1	0	1	1	0	1	0
c2	1 (01)		W1	0	0	1	0	1	0	1	1
c3	2 (10)			c8	c9						
c4	3 (11)		w2	1	0						
c5	0 (00)		w3	1	0						
c6	3 (11)										
c 7	1 (01)	1 ·	110110110 00101011 10000000								
c8	3 (11)										
c9	0 (00)										

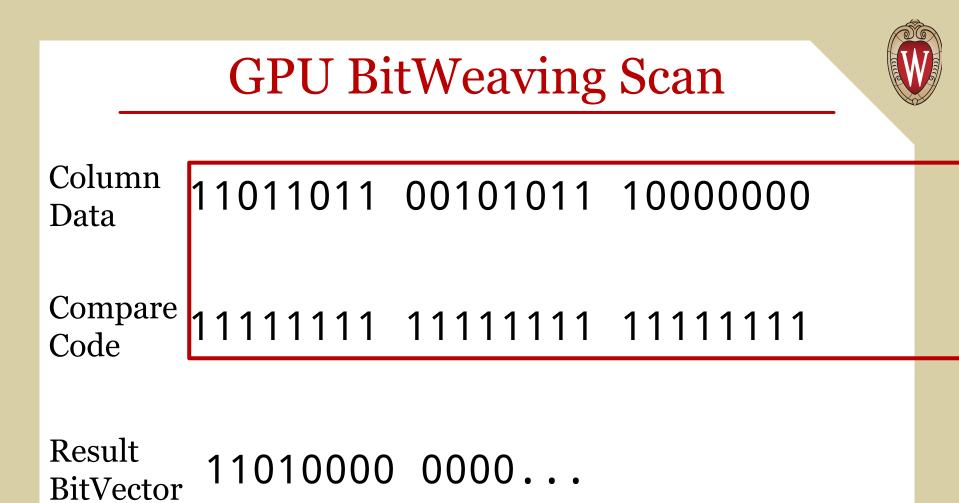


CPU BitWeaving Scan



Result BitVector 11010000 0000...

CPU width: 64-bits, up to 256-bit SIMD



GPU width: 16,384-bit SIMD

GPU Scan Algorithm

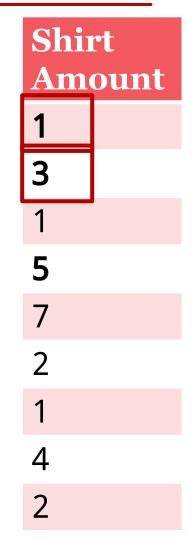
- GPU uses very wide "words"
 - CPU: 64-bits or 256-bits with SIMD
 - GPU: 16,384 bits (256 lanes × 64-bits)
- Memory and caches optimized for bandwidth
- HSA programming model
 - No data copies
 - Low CPU-GPU interaction overhead



CPU Aggregate Algorithm

Result BitVector 11010000 0000...

Result 1+3+5+...



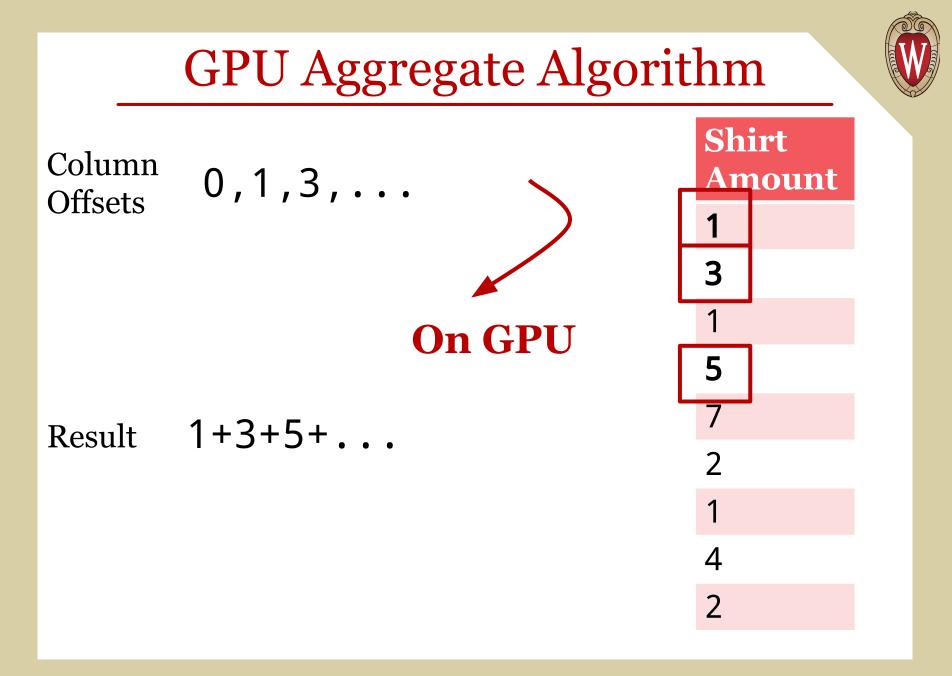


GPU Aggregate Algorithm

Result BitVector

Column 0,1,3,...







Aggregate Algorithm

- Two phases
 - Convert from BitVector to offsets (on CPU)
 - Materialize data and compute (offload to GPU)
- Two group-by algorithms (see paper)
- HSA programming model
 - Fine-grained sharing
 - Can offload subset of computation

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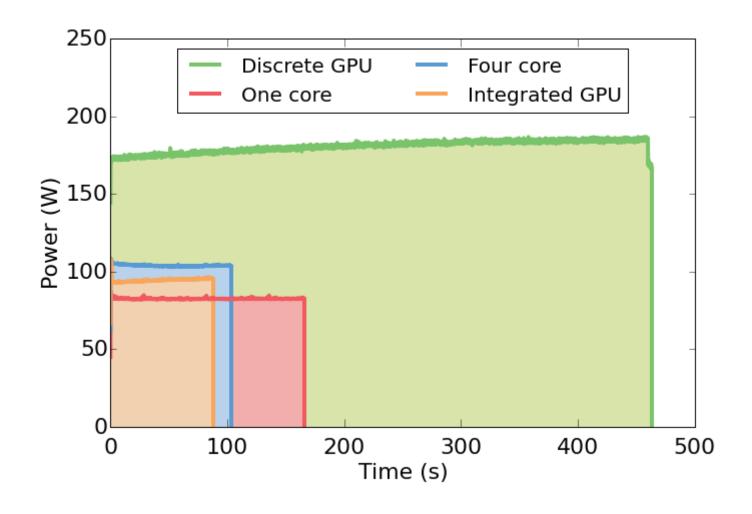


Experimental Methods

- AMD A10-7850
 - 4-core CPU
 - 8-compute unit GPU
 - 16GB capacity, 21 GB/s DDR3 memory
 - Separate discrete GPU
- Watts-Up meter for full-system power
- TPC-H @ scale-factor 10

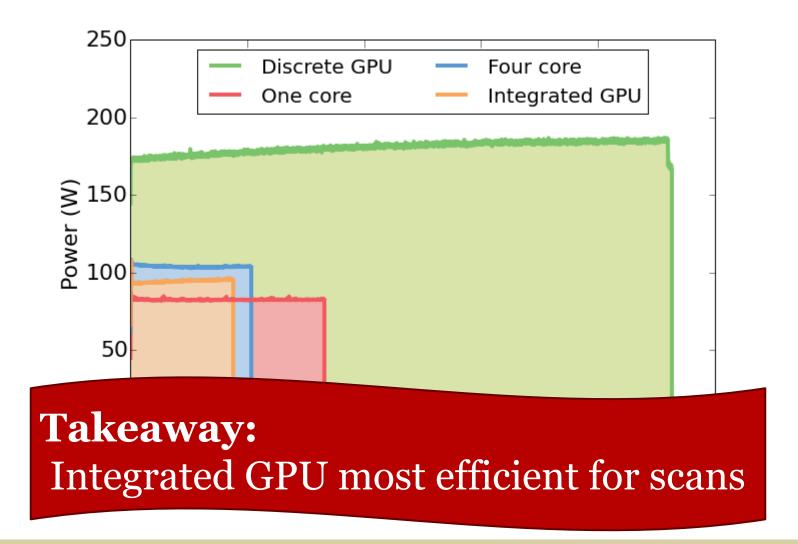


Scan Performance & Energy



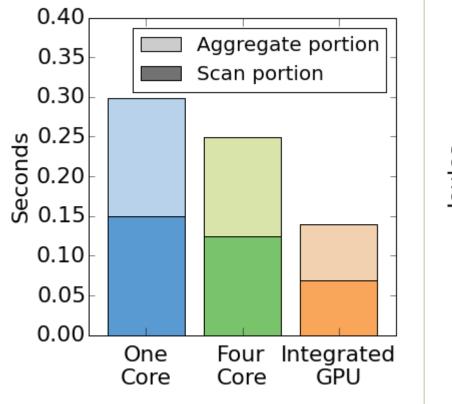


Scan Performance & Energy

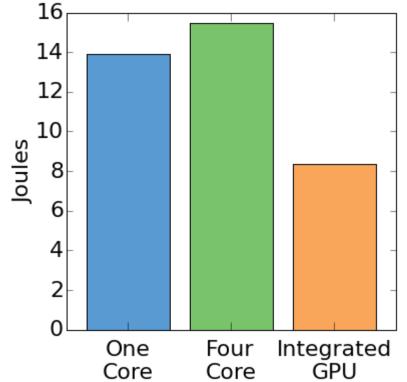




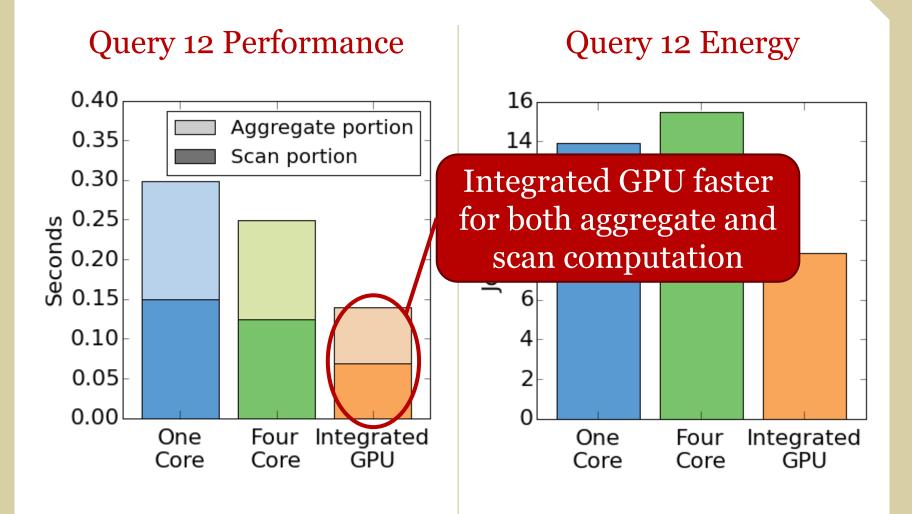
Query 12 Performance



Query 12 Energy

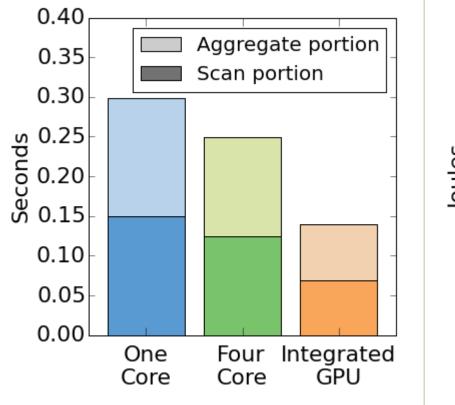




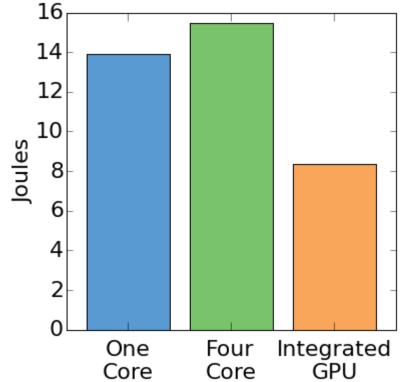




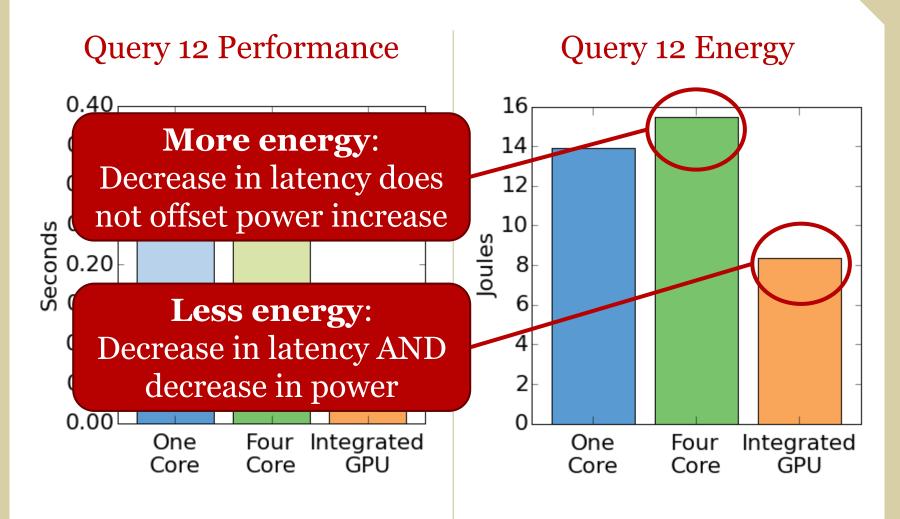
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Query 12 Energy



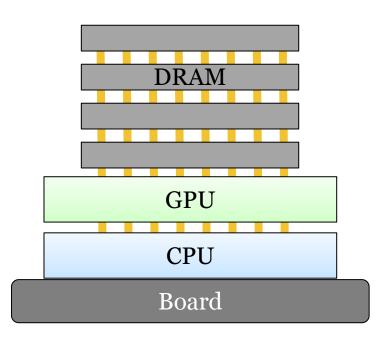






Future Die Stacked GPUs

- 3D die stacking
- Same physical & logical integration
- Increased compute
- Increased bandwidth



Power et al. *Implications of 3D GPUs on the Scan Primitive* SIGMOD Record. Volume 44, Issue 1. March 2015

Conclusions



	Discrete GPUs	Integrated GPUs	3D Stacked GPUs			
Performance	High 😳	Moderate	High 😳			
Memory Bandwidth	High 😳	Low 🙁	High 😳			
Overhead	High 😫	Low 😳	Low 😳			
Memory Capacity	Low 🙁	High 😳	Moderate			



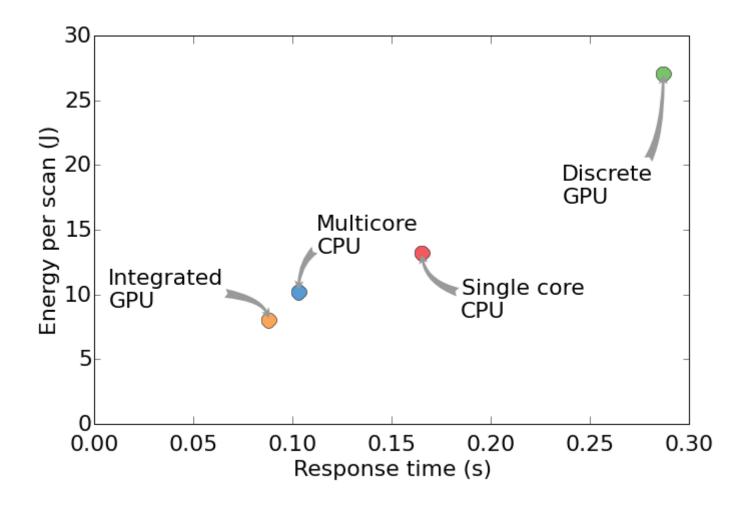


HSA vs CUDA/OpenCL

- HSA defines a heterogeneous architecture
 - Cache coherence
 - Shared virtual addresses
 - Architected queuing
 - Intermediate language
- CUDA/OpenCL are a level above HSA
 - Come with baggage
 - Not as flexible
 - May not be able to take advantage of all features

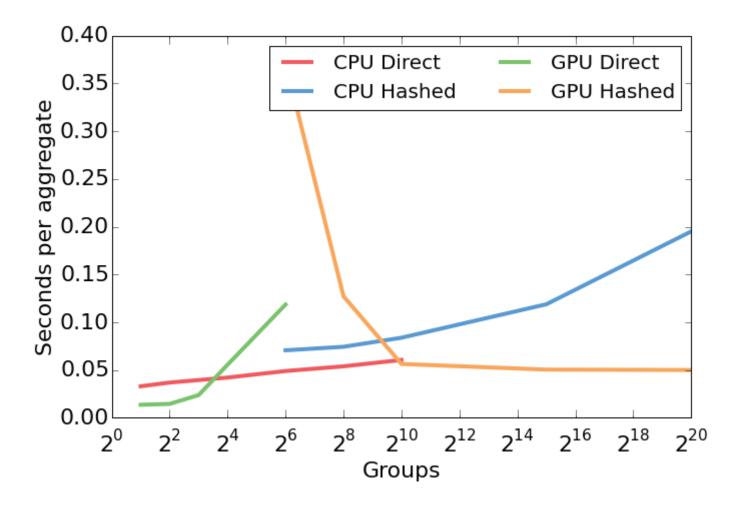


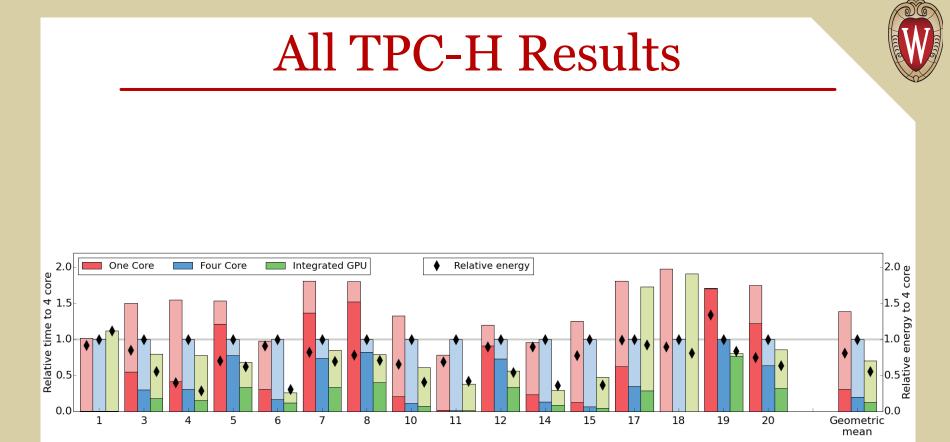
Scan Performance & Energy





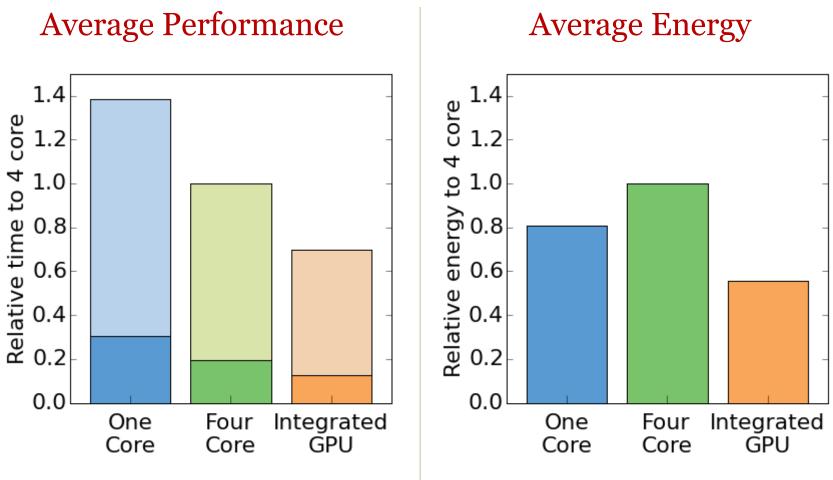
Group-by Algorithms







Average TPC-H Results



6/1/2015

What's Next?

- Developing cost model for GPU
 - Using the GPU is just another algorithm to choose
 - Evaluate exactly when the GPU is more efficient
- Future "database machines"
 - GPUs are a good tradeoff between specialization and commodity

Conclusions

- Integrated GPUs viable for DBMS?
 - Solve problems with discrete GPUs
 - (Somewhat) better performance and energy

- Looking toward the future...
 - CPUs cannot keep up with bandwidth
 - GPUs perfectly designed for these workloads