

# Toward GPUs being mainstream in analytic processing

### An initial argument using simple scanaggregate queries

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DaMoN 2015

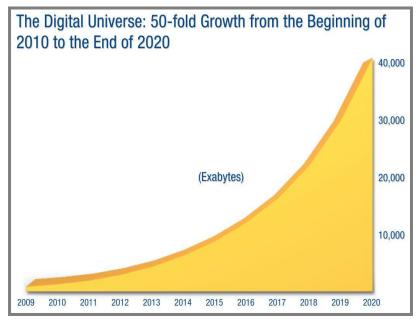
Summary

- 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

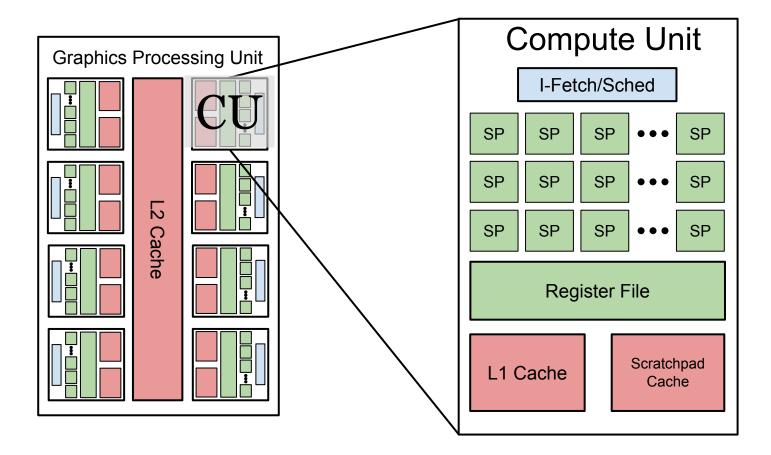


### GPUs to the Rescue?

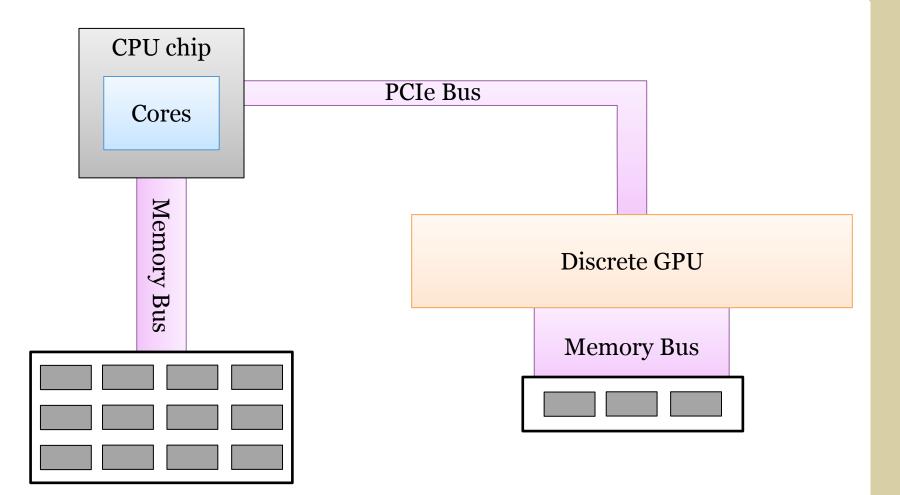
- GPUs are becoming more general
  - Easier to program
  - Integrated GPUs are everywhere
- GPUs show great promise [Govindaraju '04, He '14, He '14, Kaldewey '12, Satish '10, and many others]
  - Higher performance than CPUs
  - Better energy efficiency
- Analytic DBs look like GPU workloads

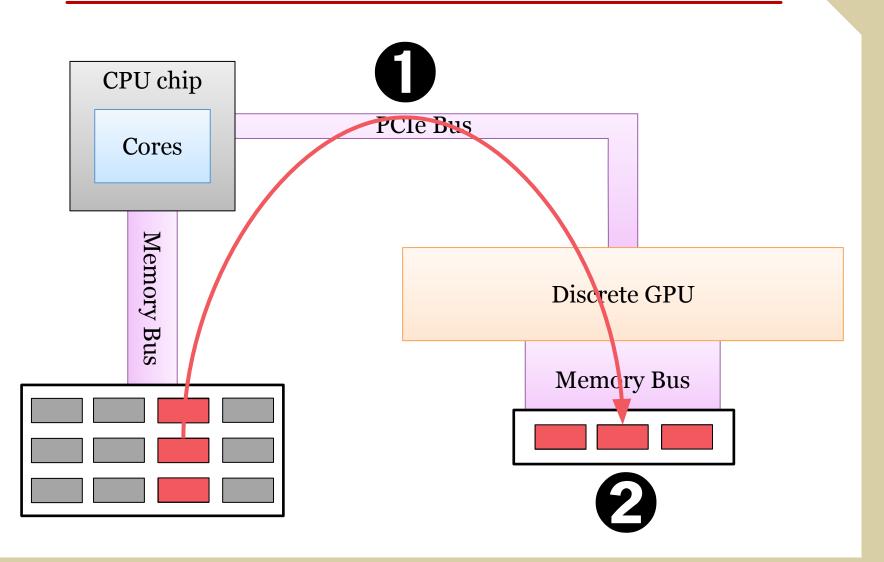


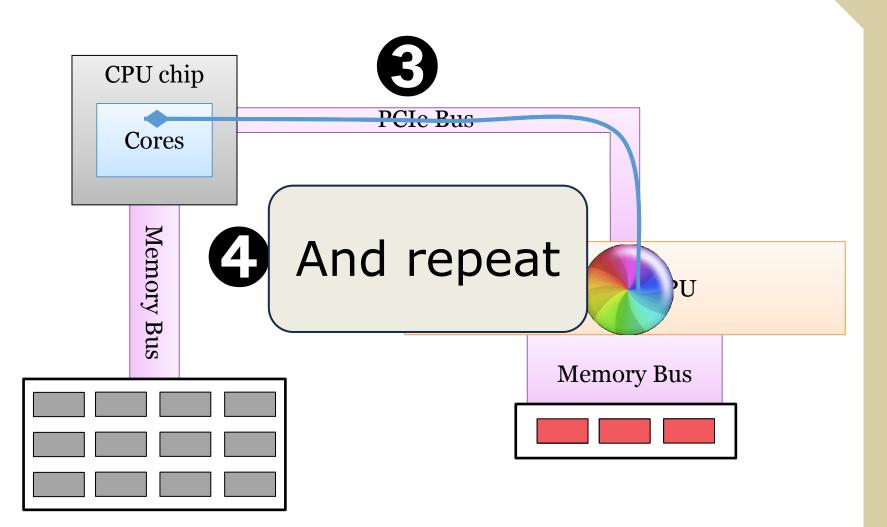
### **GPU Microarchitecture**













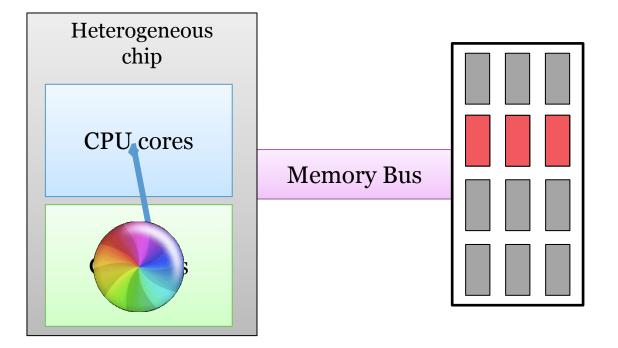
### Copy data over PCIe

- 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

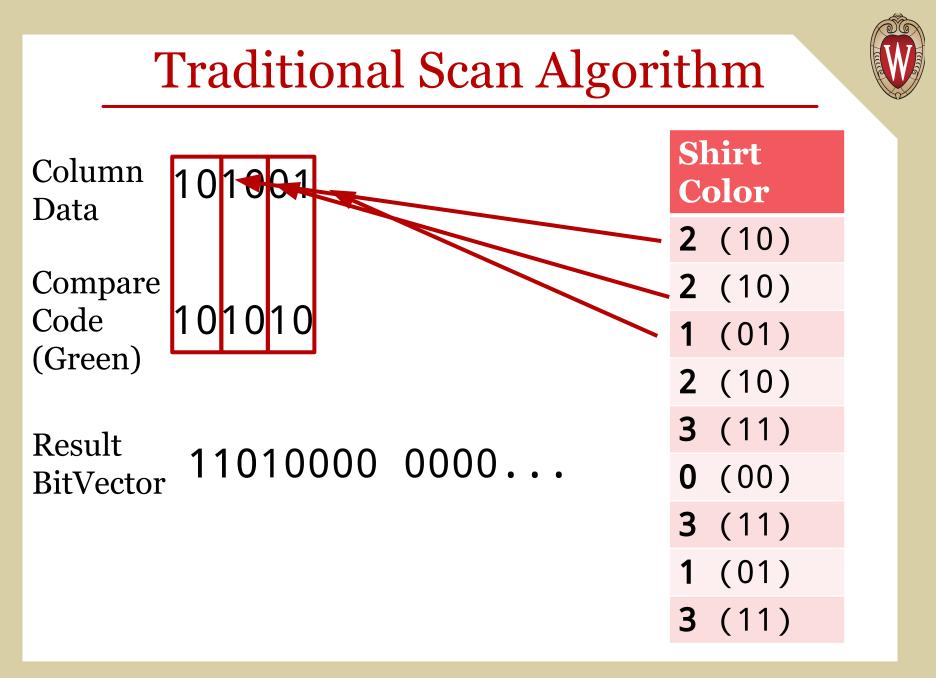
Shirt Color	Shirt Amount
2	1
2	3
1	1
2	5
3	7
0	2
3	1
1	4
3	2

Color	Code
Red	0
Blue	1
Green	2
Yellow	3

# Running Example



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



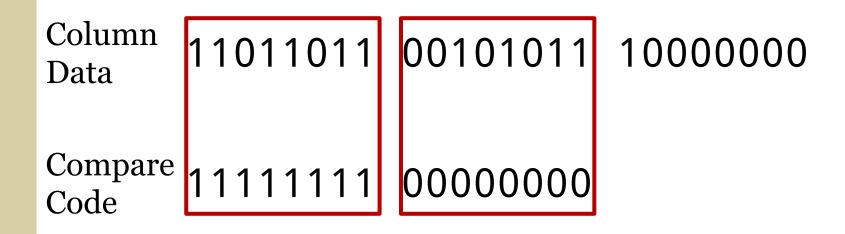


### Vertical Layout

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

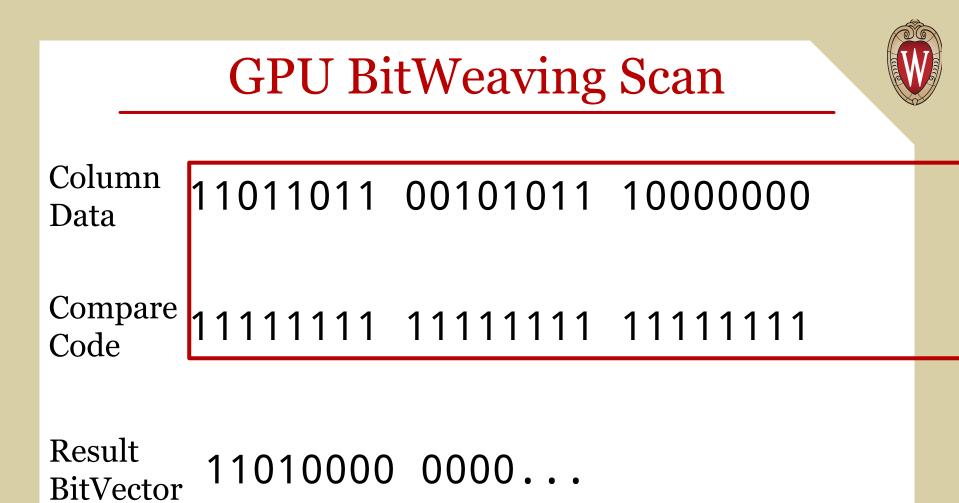


### **CPU BitWeaving Scan**



Result BitVector 11010000 0000...

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



### GPU width: 16,384-bit SIMD

# GPU Scan Algorithm

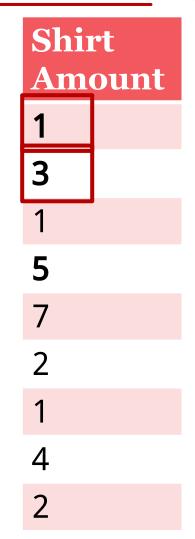
- GPU uses very wide "words"
  - CPU: 64-bits or 256-bits with SIMD
  - GPU: 16,384 bits (256 lanes × 64-bits)
- Memory and caches optimized for bandwidth
- HSA programming model
  - No data copies
  - Low CPU-GPU interaction overhead



# CPU Aggregate Algorithm

#### Result BitVector 11010000 0000...

#### Result 1+3+5+...



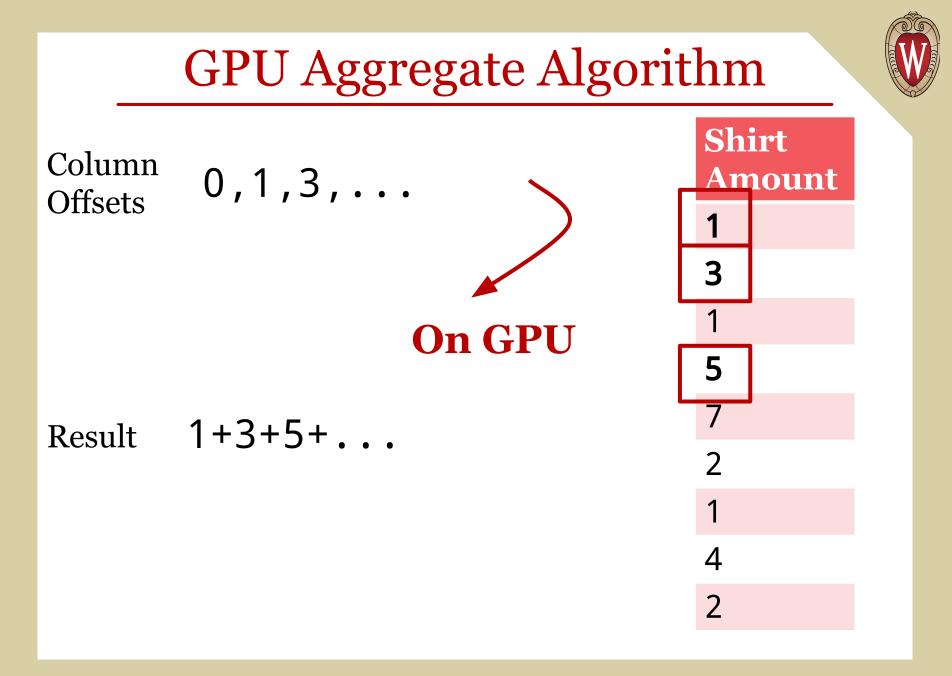


### GPU Aggregate Algorithm

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

### Outline



- Background
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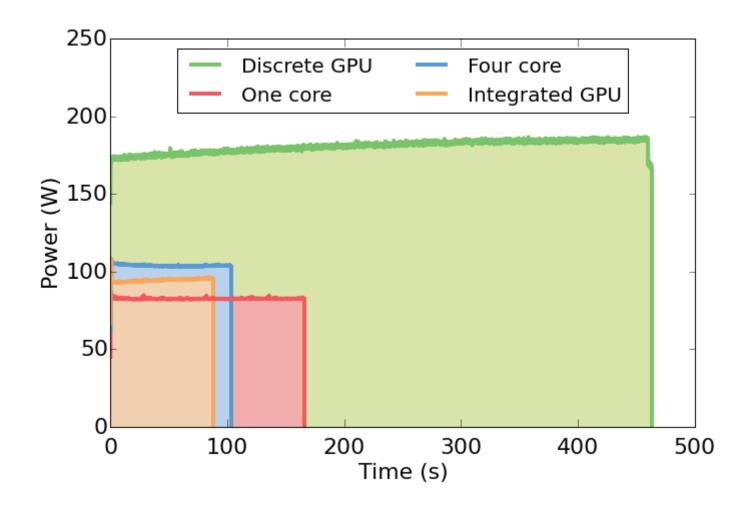


### **Experimental Methods**

- 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

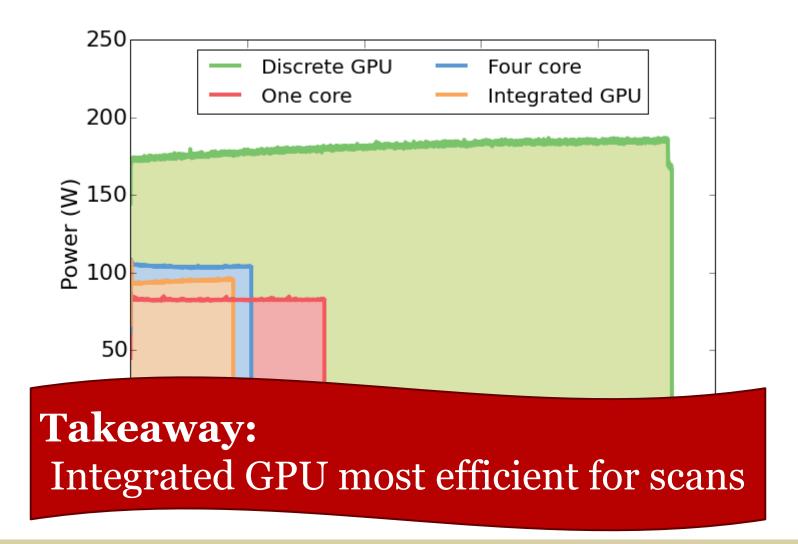


### Scan Performance & Energy



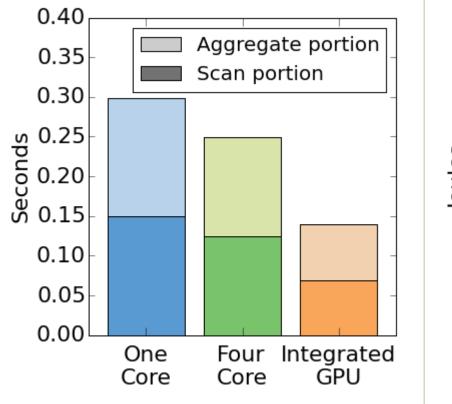


### Scan Performance & Energy

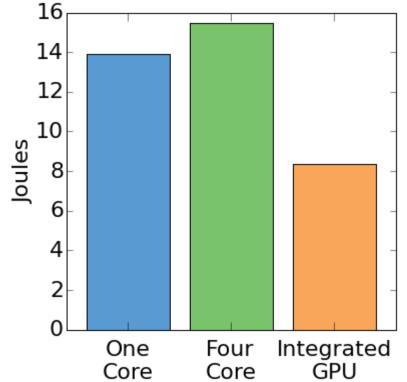




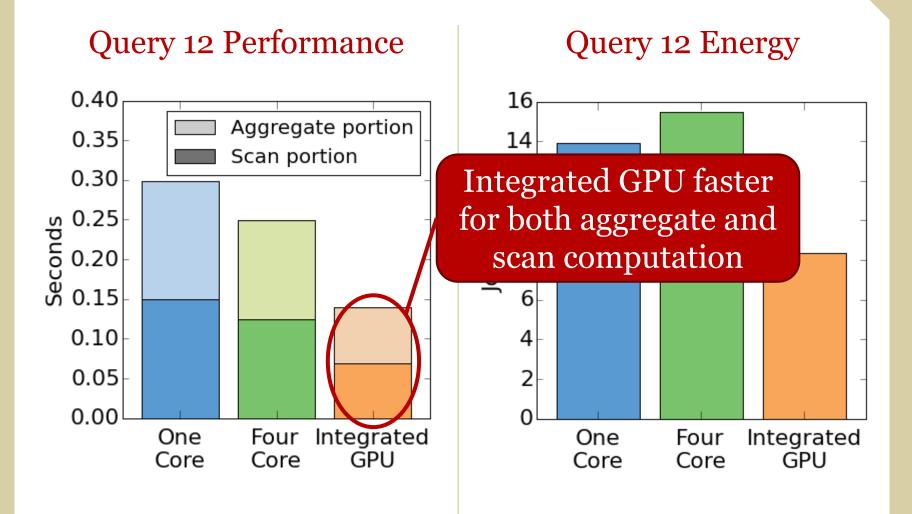
#### Query 12 Performance



#### Query 12 Energy

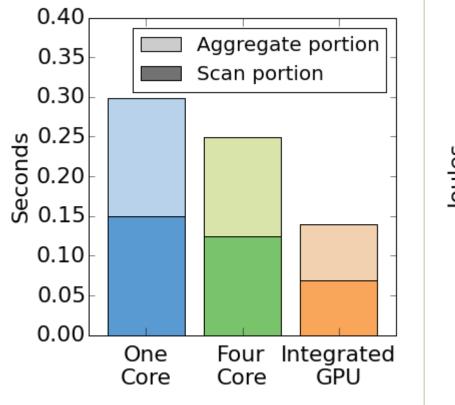




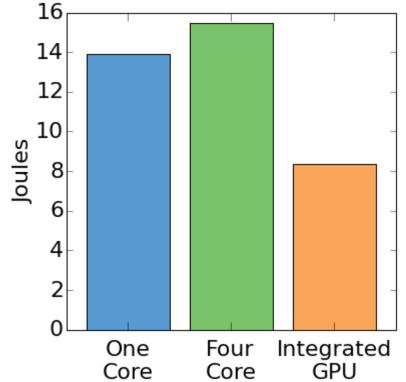




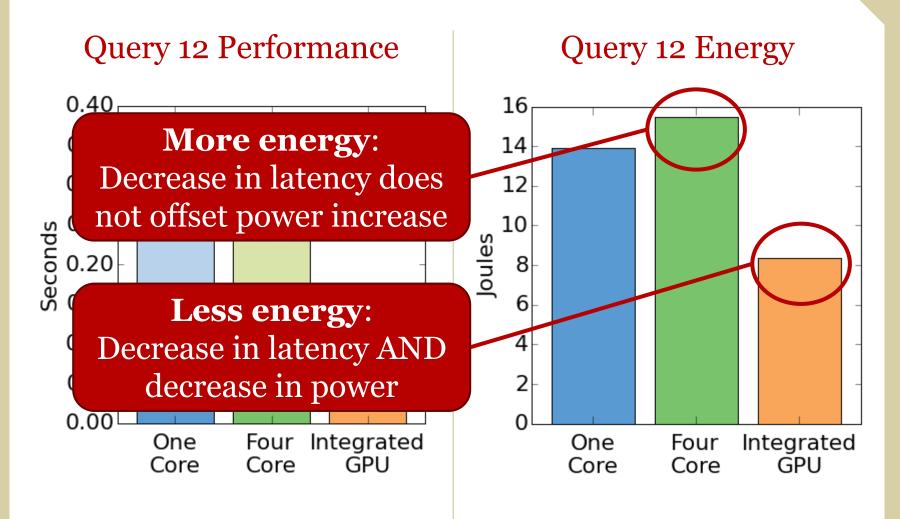
#### Query 12 Performance



#### Query 12 Energy



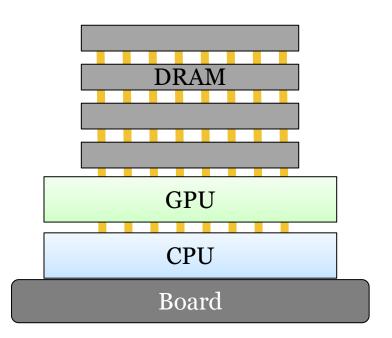






### **Future Die Stacked GPUs**

- 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			



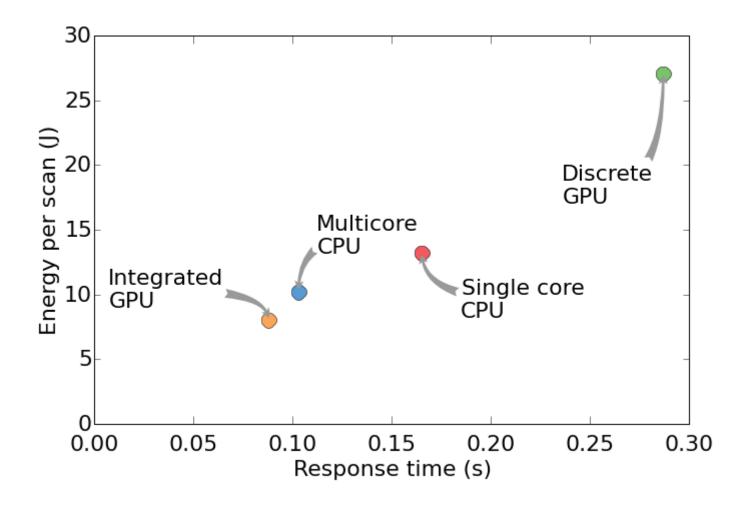


# HSA vs CUDA/OpenCL

- 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

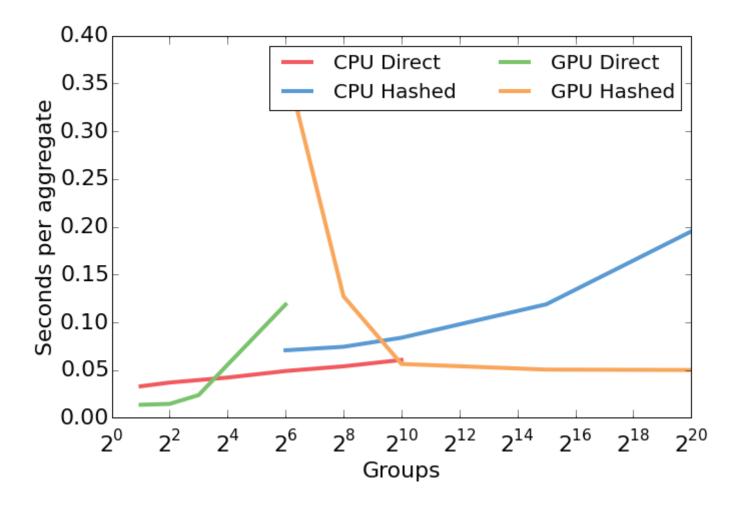


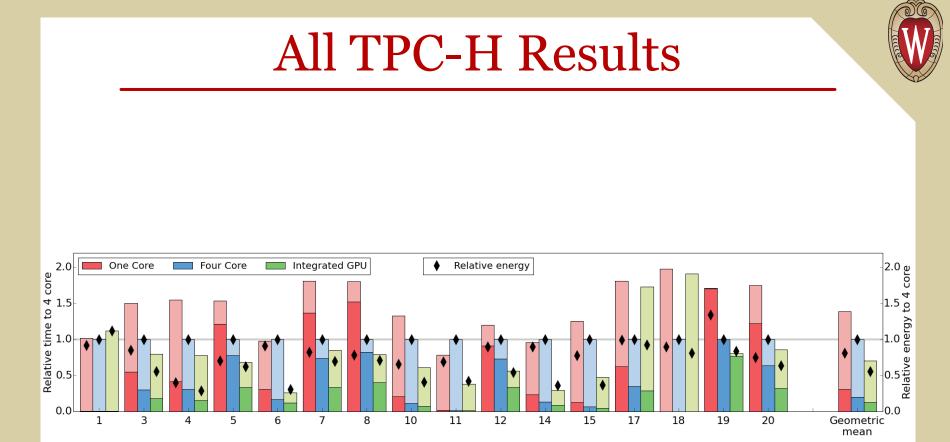
### Scan Performance & Energy





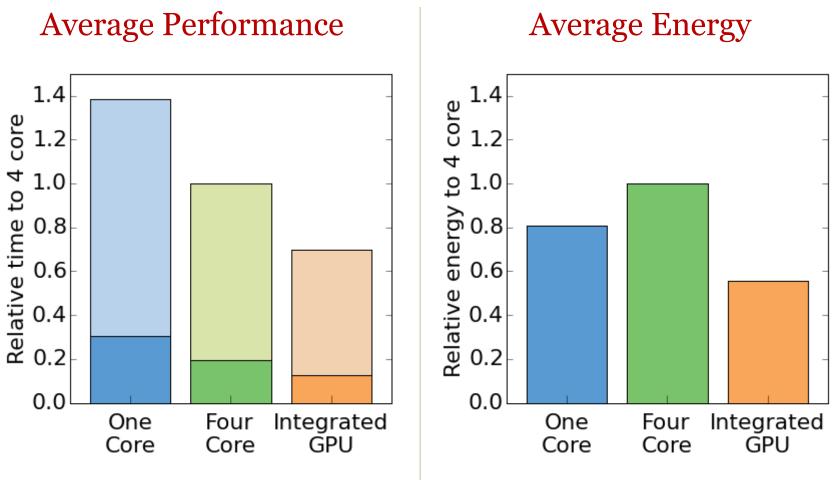
# Group-by Algorithms







### **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