A Virtual Machine Model for Accelerating Relational Database Joins using a General Purpose GPU

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Databases Are Used Everywhere

SQLite: Most Deployed DB

Virtual Machine-based Database Engine

- 300 million copies of Firefox
- 20 million Apple computers
- 500 million iPhones*
- 1 billion Android Devices**
- 450 million registered Skype users
- 10 million Solaris 10 installations

Adapted from: https://www.sqlite.org/mostdeployed.html

*http://onforb.es/1gpk4Fs

**http://www.androidcentral.com/android-passes-1-billion-activations



G2 instance: 65¢/hr



G2 instance: 65¢/hr





G2 instance: 65¢/hr





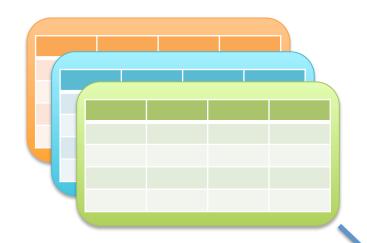
Why don't we combine SQLite and GPUs to improve performance?

Focus on JOINs

Overview

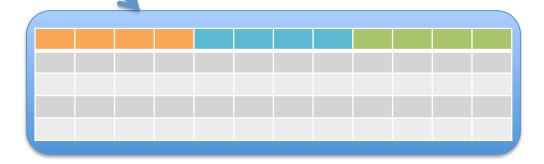
- Review of Database JOINs
- Why GPUs a good fit?
- Virtual Machine Implementation
 - Query Workflow
 - SQL Queries as Programs
 - Memory Concerns
 - VM Design
- Experimental Results

Database JOIN



Predefined, Static Relationship

JOIN



Cross-Join Example

T₁ and T₂ share attribute c₂

1 1						
c_1	c_2					
1	W					
2	Z					
3	Z					

\mathbf{T}_2						
c_2	c_3					
X	5					
Z	6					
W	7					

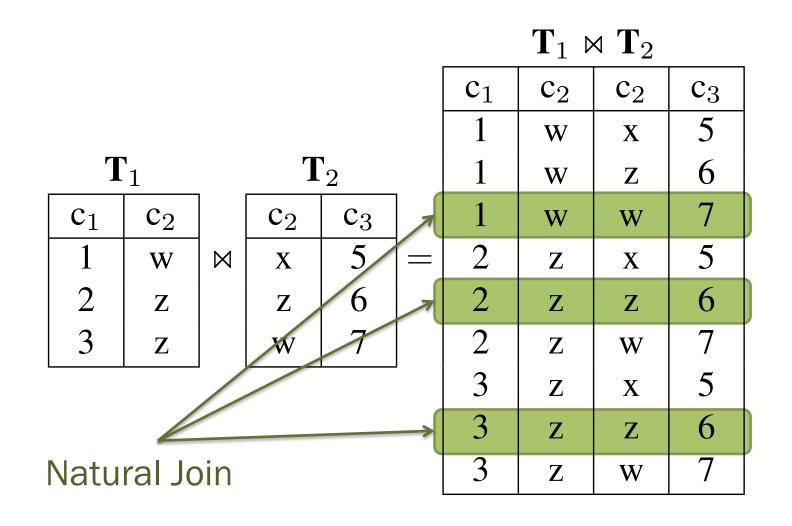
Cross-Join Example

T	, 1	
c_1	c_2	
1	W	M
2	Z	
3	Z	

_	Z	
c_2	c_3	
X	5	=
Z	6	
W	7	

$\mathbf{T}_1 \bowtie \mathbf{T}_2$							
c_1	c_2	c_2	c_3				
1	W	X	5				
1	W	Z	6				
1	W	\mathbf{W}	7				
2	Z	X	5				
2	Z	Z	6				
2	Z	\mathbf{W}	7				
3	Z	X	5				
3	Z	Z	6				
3	Z	\mathbf{W}	7				

Cross-Join Example



Restrict Result with Predicate

• SELECT * FROM T_1,T_2 WHERE $T_1.c_2 = T_2.c_2$ $T_1 \bowtie T_2$

						c_1	C_2	$\mid c_2 \mid$	C_3	
						1	W	X	5	
\mathbf{T}_1 \mathbf{T}_2						1	\mathbf{W}	Z	6	
c_1	c_2		c_2	c_3		1	\mathbf{W}	\mathbf{W}	7	
1	W	\bowtie	X	5		2	Z	X	5	
2	Z		Z	6		2	Z	Z	6	
3	Z		W	7		2	Z	\mathbf{W}	7	
		•				3	Z	X	5	
						3	Z	Z	6	
						3	7.	\mathbf{w}	7	

Restrict Result with Predicate

• SELECT * FROM T_1,T_2 WHERE $T_1.c_2 = T_2.c_2$ $T_1 \bowtie T_2$

						c_1	c_2	c_2	c_3	
					_	1	***	***	5	LV
						1	vv	Λ		
T	1		T	$\overline{2}$	_	1	***		6	LV
_	· T	1			1	1	VV	L		^
c_1	c_2		c_2	c_3		1	W	W	7	
1	W	\bowtie	X	5	l	<u> </u>		₹7	5	LV
1	VV		Λ			<u>_</u>	L	Λ		
2	Z		Z	6		2	Z	Z	6	
3	7		**7	7		<u> </u>		***	7	
			W	/	_	<u> </u>	L	VV	'	
	•			•	_	2		₹7	5	
					_		L	Λ		
						3	Z	Z	6	
						2		~~~	7	I V
						J	L	VV	/	

Restrict Result with Predicate

• SELECT * FROM T_1,T_2 WHERE $T_1.c_2 = T_2.c_2$ $T_1 \bowtie T_2$

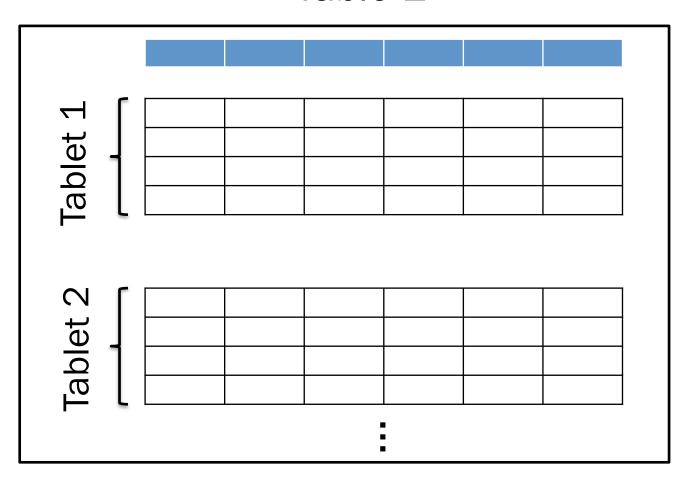
$\begin{vmatrix} c_1 & c_2 & c_2 \end{vmatrix}$	
$\frac{1}{1}$ $\frac{1}{W}$ $\frac{5}{\Lambda}$	$\exists \mathbf{v}$
\mathbf{T}_1 \mathbf{T}_2 \mathbf{T}_2 \mathbf{T}_3	\perp \vee
	$\perp \wedge$
$\begin{array}{ c c c c c c c c c c c c c c c c c c c$	
$\begin{array}{ c c c c c c c c c c c c c c c c c c c$	\perp
$ 1 w M X 5 = \frac{2}{2} \frac{ z }{ z } \frac{ x }{ z } \frac{5}{ z }$	$\top \wedge$
2 z 6 2 z 6	
3 z w 7 2 z w 7	\perp
3 Z W / Z Z W /	$\top \wedge$
$\frac{3}{2}$	
SIMD Execution 3 z z 6	
	$\perp X$

IMPLEMENTATION

Virginian Database

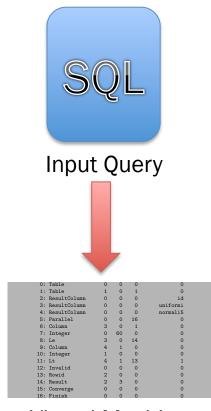
- Written by Peter Bakkum, NEC Labs, New Jersey
- Virtual Machine-based Implementation
 - Similar to SQLite
- Demonstrated speedup for single-table queries
- Presented "Tablet" data structure for efficient processing on CPU/GPU

Table 1

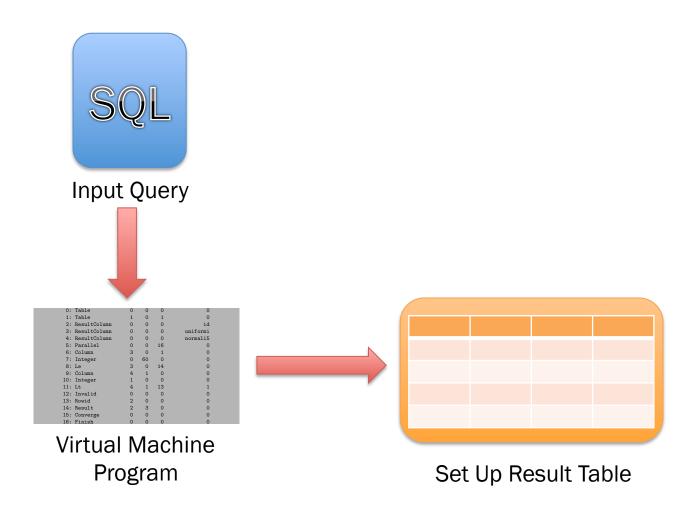


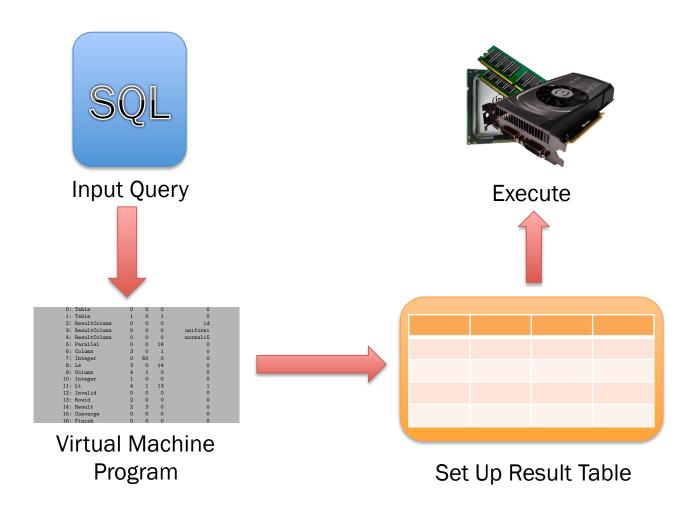






Virtual Machine Program





SELECT test.id, test1.uniformi, test.normali5 FROM test,test1
WHERE test1.uniformi > 60 AND test.normali5 < 0;

SELECT test.id, test1.uniformi, test.normali5 FROM test,test1
WHERE test1.uniformi > 60 AND test.normali5 < 0;

0:	Table	0	0	0	0
1:	Table	1	0	1	0
2:	ResultColumn	0	0	0	id
3:	ResultColumn	0	0	0	uniformi
4:	ResultColumn	0	0	0	normali5
5:	Parallel	0	0	16	0
6:	Column	3	0	1	0
7:	Integer	0	60	0	0
8:	Le	3	0	14	0
9:	Column	4	1	0	0
10:	Integer	1	0	0	0
11:	Lt	4	1	13	1
12:	Invalid	0	0	0	0
13:	Rowid	2	0	0	0
14:	Result	2	3	0	0
15:	Converge	0	0	0	0
16:	Finish	0	0	0	0

SELECT test.id, test1.uniformi, test.normali5 FROM test,test1
WHERE test1.uniformi > 60 AND test.normali5 < 0;

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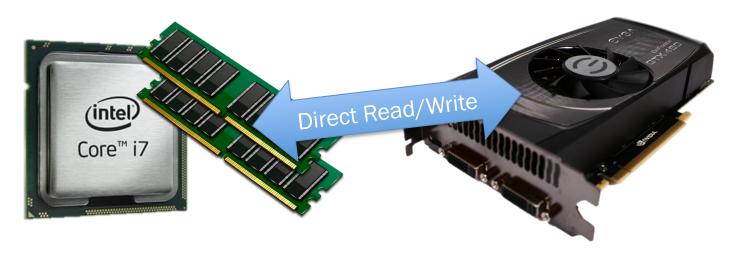
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	14:	Result	2	3	0	0
	15:	Converge	0	0	0	0
	16:	Finish	0	0	0	0

Compute for each row

Allocating Result Table

- How much memory must be allocated? $||TableA|| \cdot ||TableB||$
- If each table has 3500 rows, then we have... $12\ 250\ 000\ \mathrm{rows}$
- Memory limits of GPU: use mapped memory!



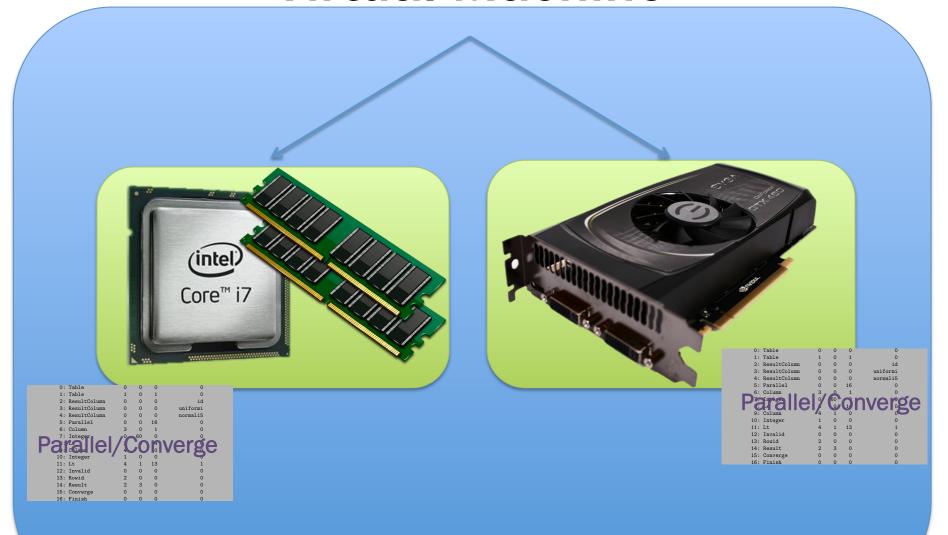
Virtual Machine

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Virtual Machine



Virtual Machine

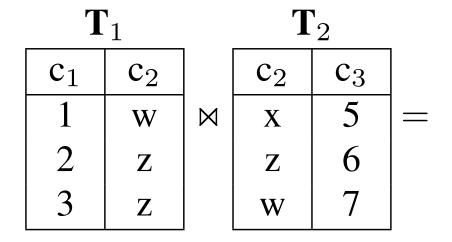


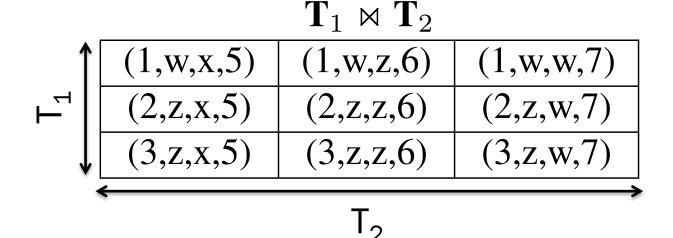


Query Processing

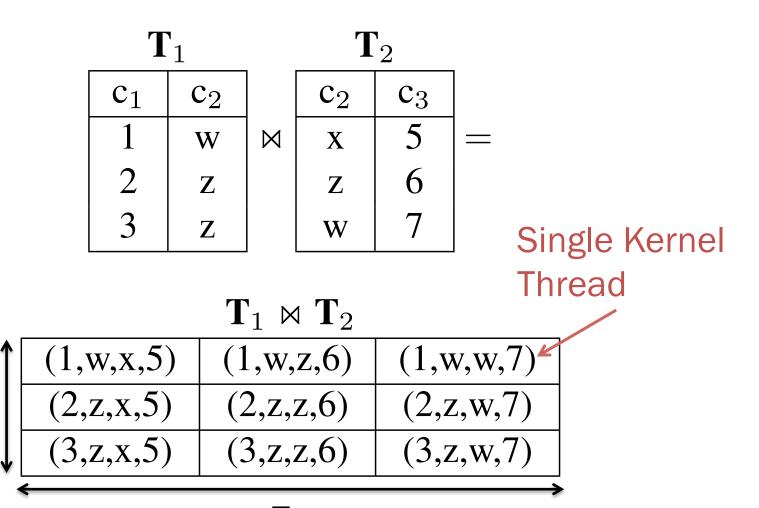
- CPU: Nested Loop Join (NLJ) a la SQLite
- GPU: exploit 3D topological structure of CUDA threads
 - Assign source table to each thread dimension
 - Each thread represents a single entry in the cross-product
 - Write entries satisfying predicate to result table

Thread Mapping Scheme

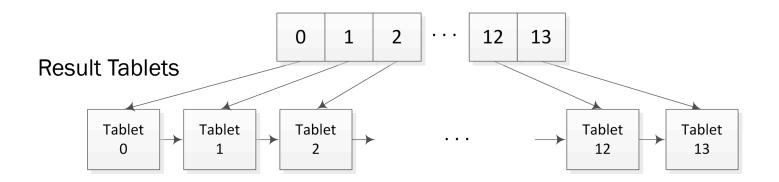




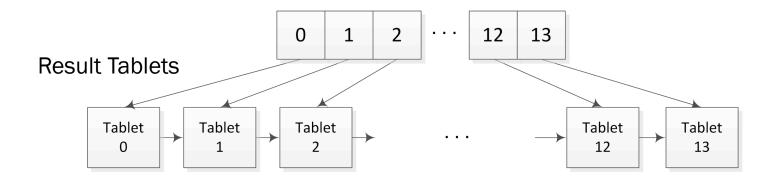
Thread Mapping Scheme

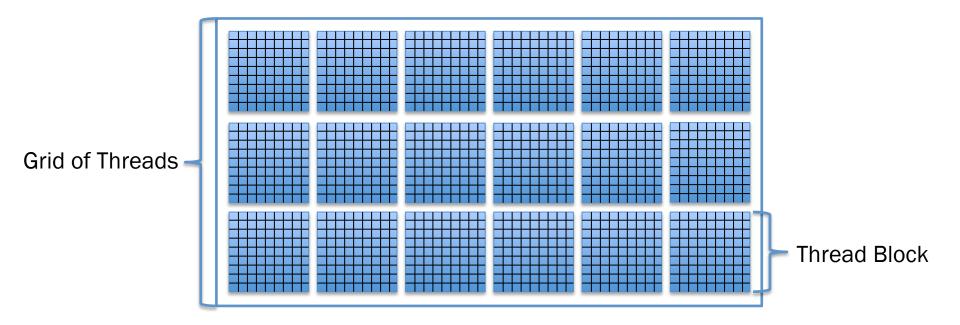


Writing Results

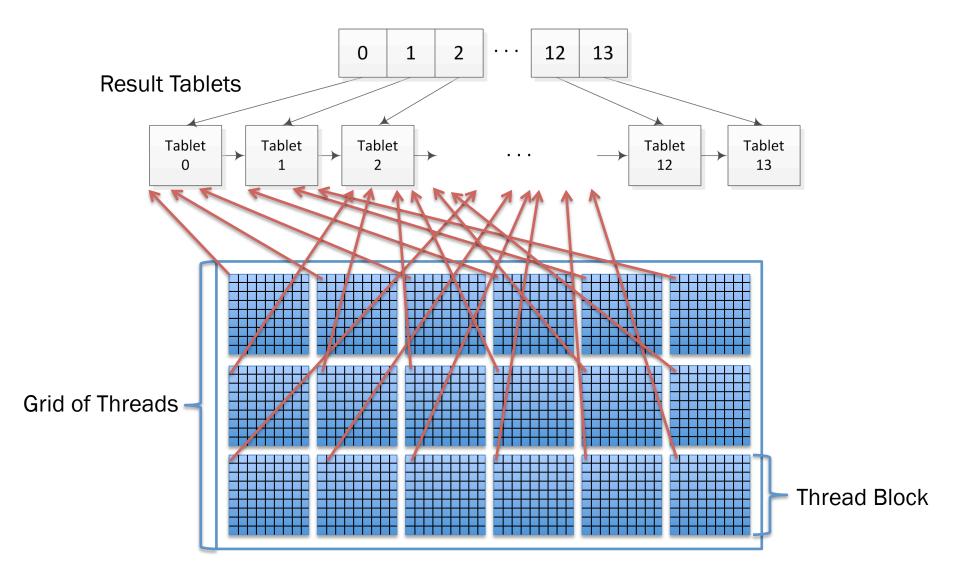


Writing Results

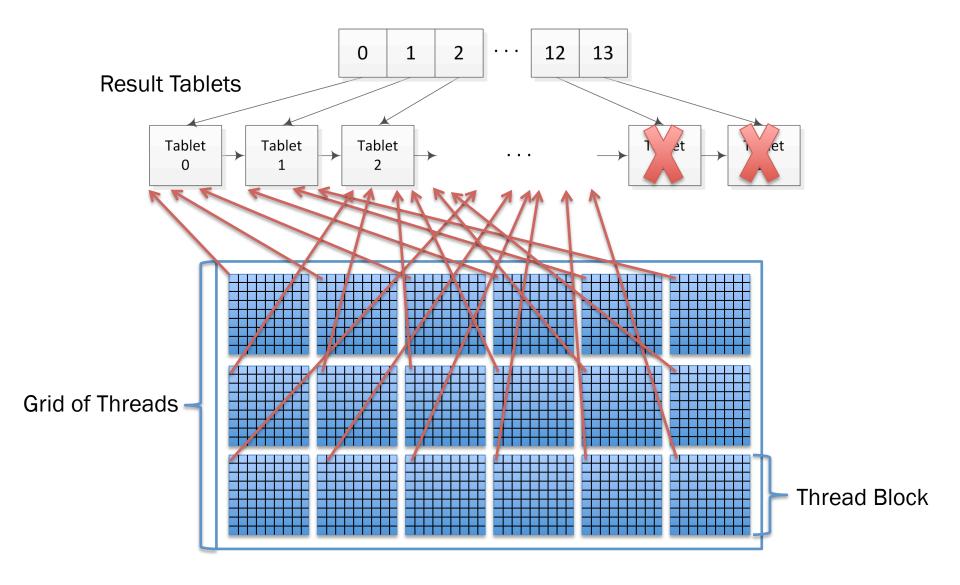




Writing Results



Writing Results



EXPERIMENTAL RESULTS

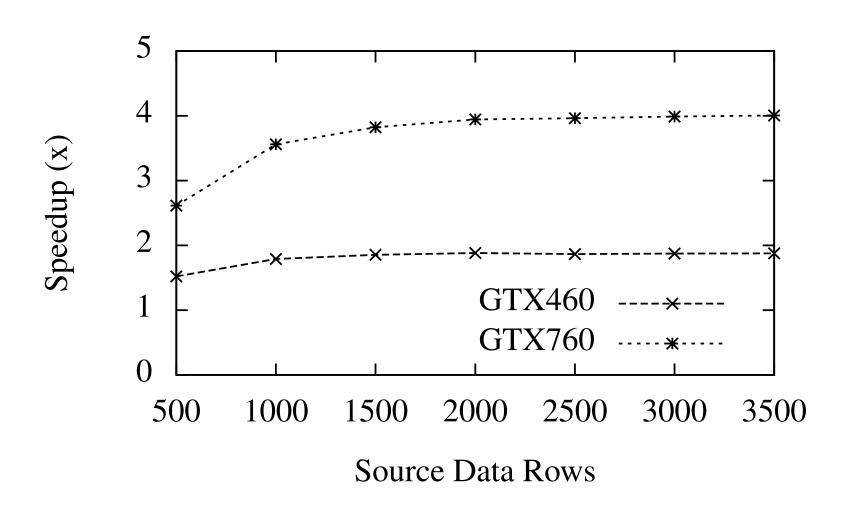
Test Machine

- Intel Core® i7 920 CPU @ 2.66 GHz
- NVIDIA GTX460 GPU
 - Fermi Microarchitecture
 - 336 CUDA Cores
 - 1 GB Memory
- NVIDIA GTX760 GPU
 - Kepler Microarchitecture
 - 1152 CUDA Cores
 - 2 GB Memory
- Linux 3.13.0-39-generic kernel
- CUDA 6.5, NVIDIA 340.29 Driver

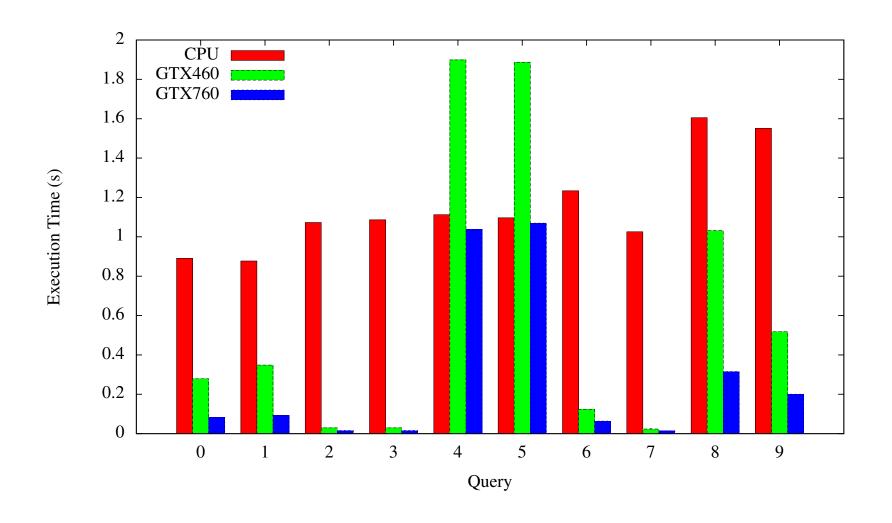
Query Test Suite

- 10 Queries
 - 5 32-bit Integer / 5 32-bit Floating Point
- Random data (Both uniformly, normally distributed)
- All results based of 10 executions

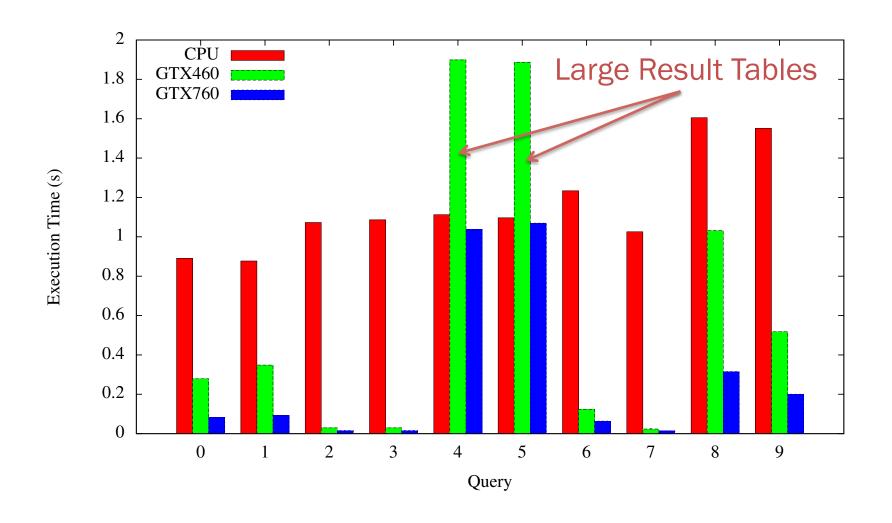
Speedup for Increasing Table Size



Query Execution Time

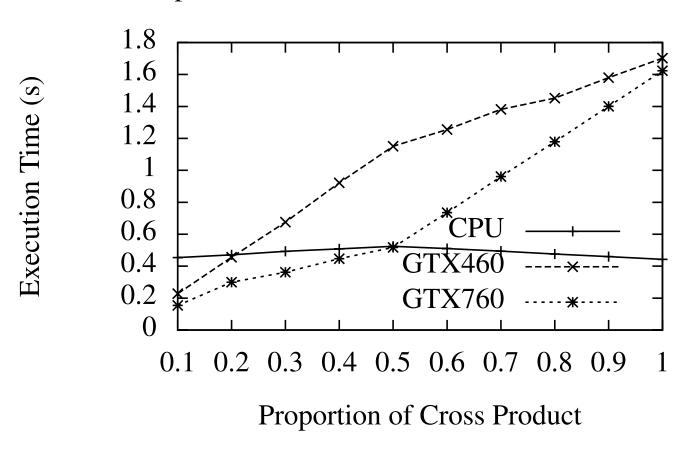


Query Execution Time



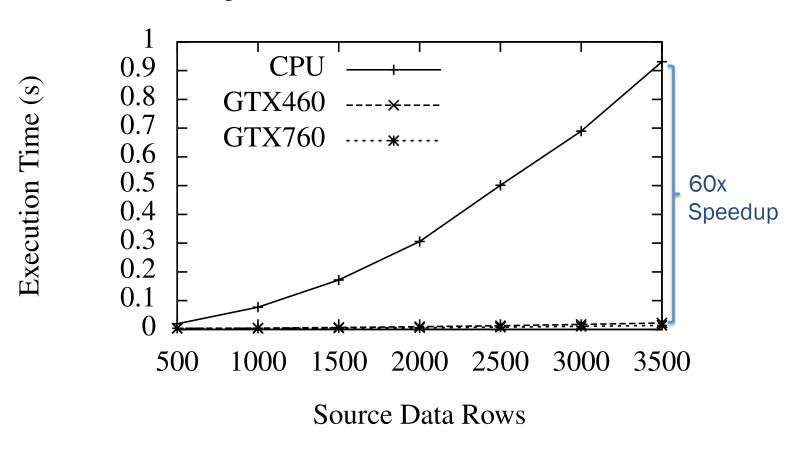
PCIe Bottleneck

Proportion of Rows Returned vs. Execution Time



Natural Join

Running Time Growth with Restrictive Predicate



Conclusions / Future Directions

- GPU implementation of VM-based query processor can be used to accelerate relational database joins
 - 2x-4x on average; 20x-60x in many common cases
- Scalability of a VM-based approach?
- Multiple GPUs to process queries
- Dynamically choose between CPU and GPU

Thank You!

Questions