

Announcements

- HW3 is out, due next Friday
 - Covers indexing
- Piazza Poll which advanced topics?
 - Information Retrieval
 - Data Mining
 - Distributed Databases
 - Information Integration
 - Semistructured Data



Review

- Why are we learning algorithms for implementing RA operators?
- Why are we learning multiple ways to implement them?
- What is a "full relational unary operator"?



Review

- Why are we interested in the "iterator model" for our operators?
- What is the "buffer pool"?
- How do we measure the cost of operators?
- What is a "table scan"?



Review

- How did we implement the "tuple at a time" operators?
- What is a "nested loop join"?
- What is a "two phase multiway merge sort"?



One-Pass/NBLJ Summary

Operator	M required	I/O Cost
σ, π	1	В
δ, γ	В	В
U ,∩ , -, ⋈, ×	min(B(R),B(S))	B(R)+B(S)
\bowtie	M≥2	B(R)B(S)/M



Sort-Based Algorithms

- Can base implementation of operations on merge sort
 - Special implementation for large data
 - Called *Two-Phase Multiway Merge-Sort*
 - TPMMS

TPMMS

- Has two phases
 - Phase 1: sort memory sized sublists
 - Phase 2: merge sorted sublists
- Requires $\sqrt{B(R)} \leq M$

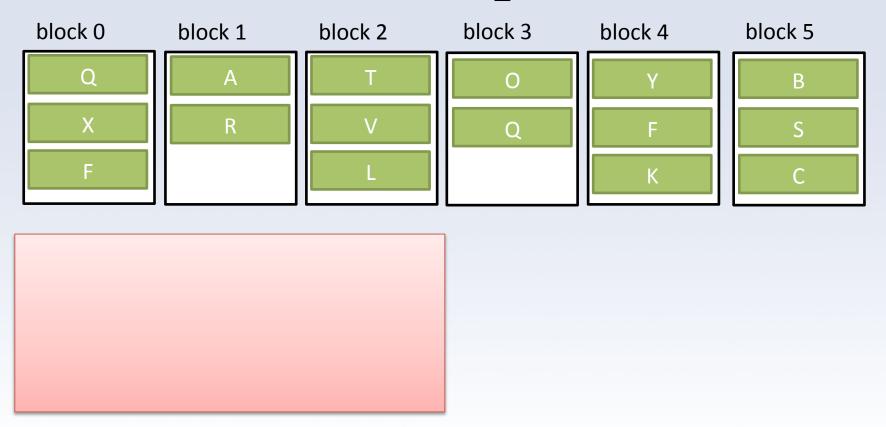


TPMMS

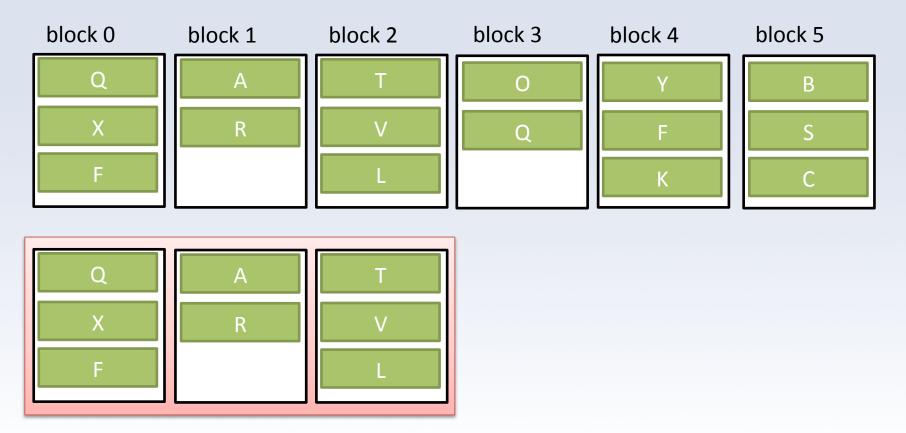
- Phase 1: Repeat until R is exhausted:
 - Fill up all M buffers with blocks from R
 - Sort using quicksort
 - Write resulting sublist to disk
- Phase 2: Repeat until sublists exhausted:
 - Load buffers with smallest block for each sublist
 - Identify smallest elements, move to output block
 - When output block full, write to disk
 - If list buffer is empty, get next one from disk



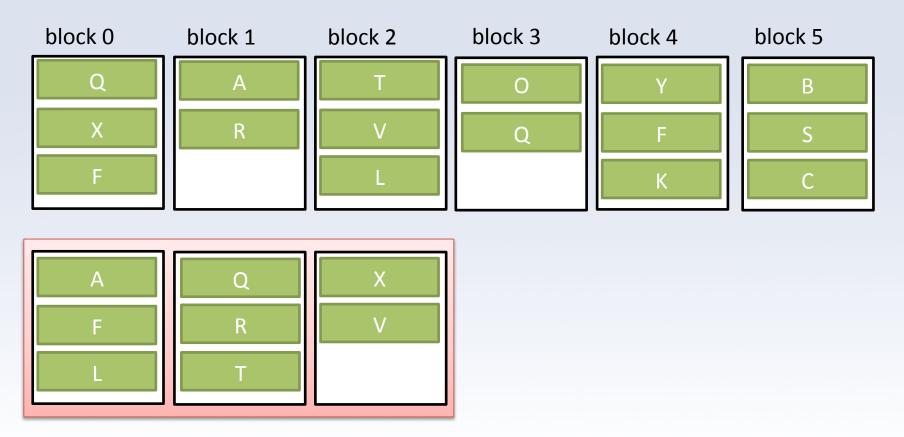
Example



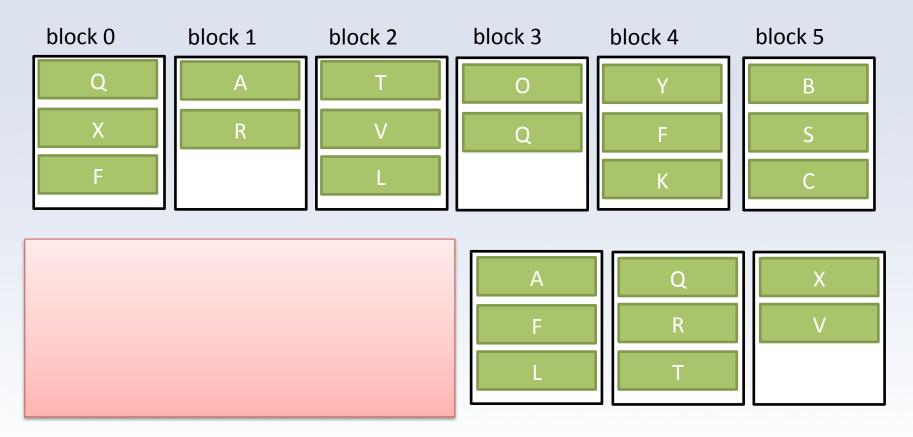




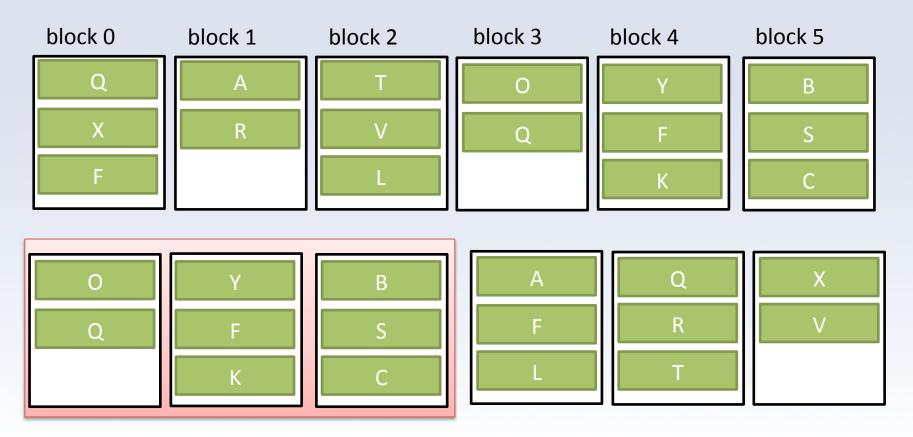




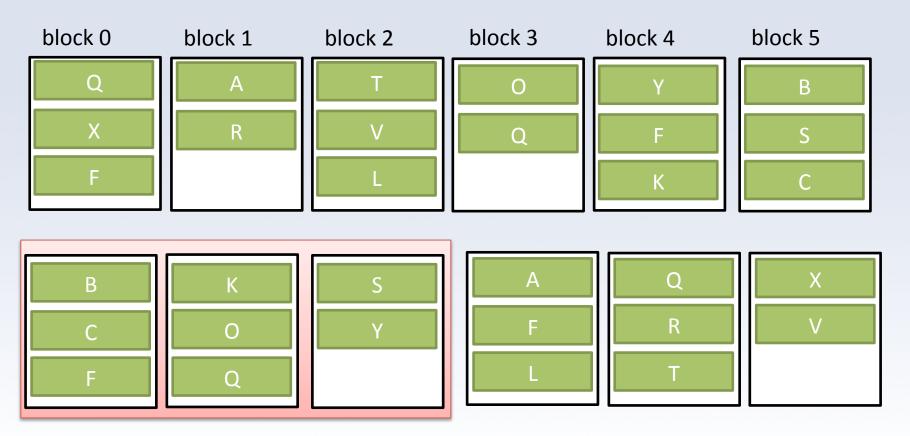




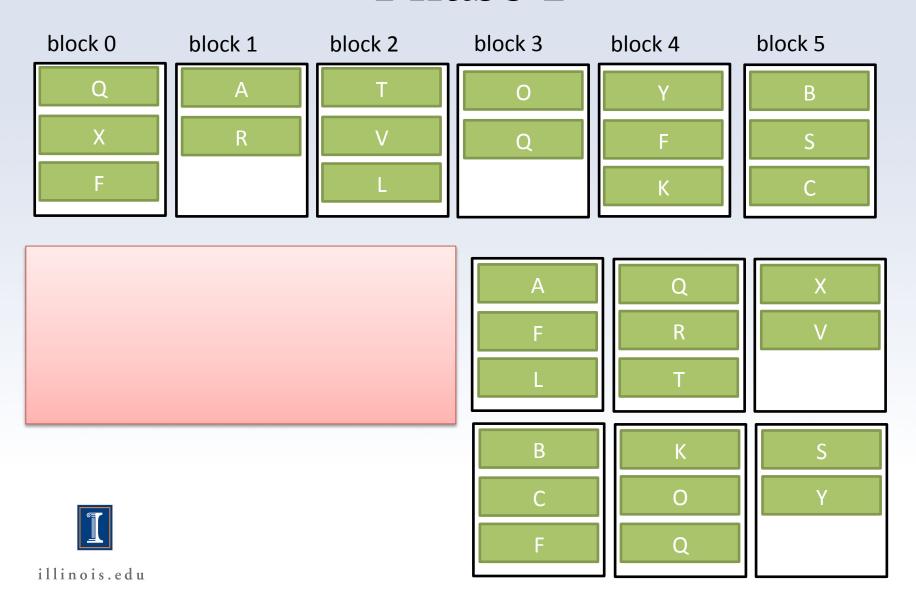


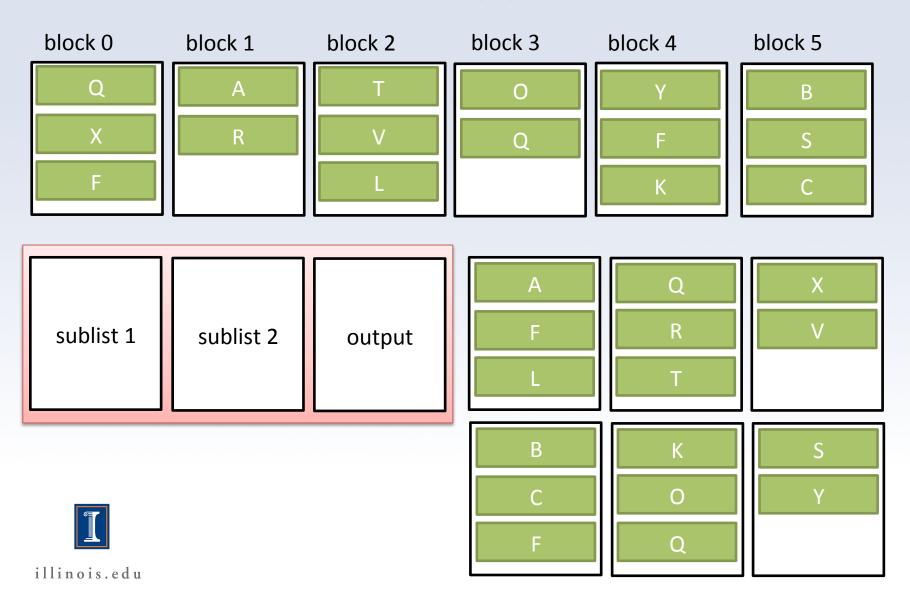


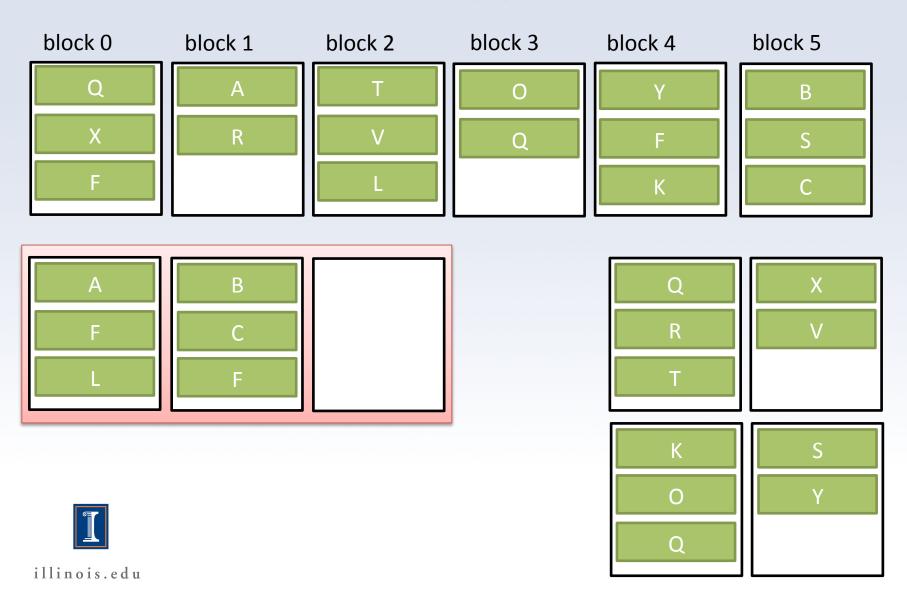


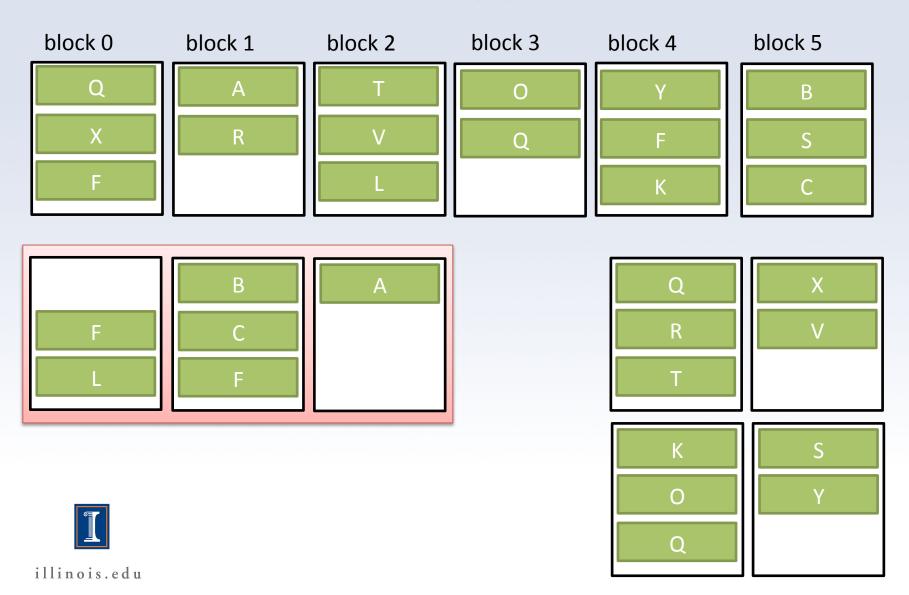


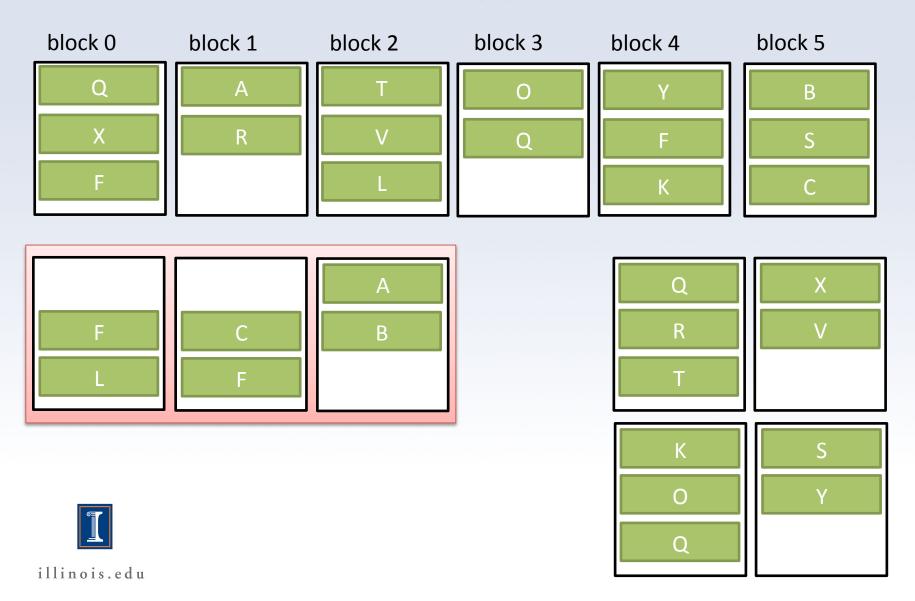


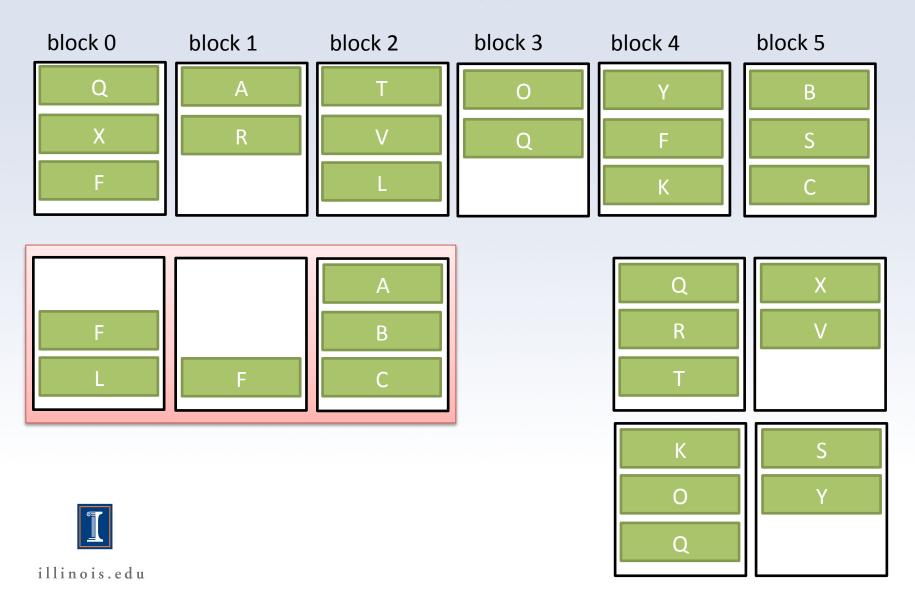


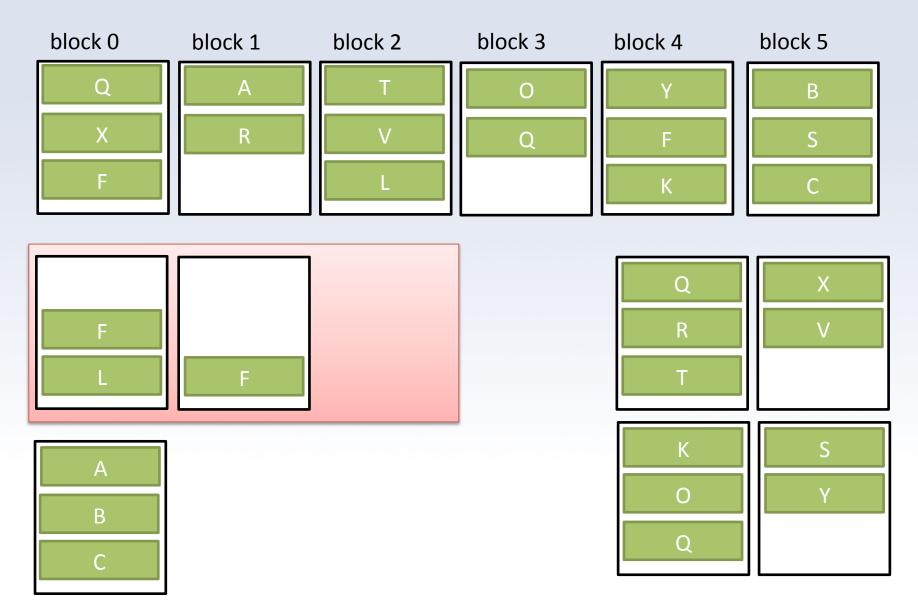


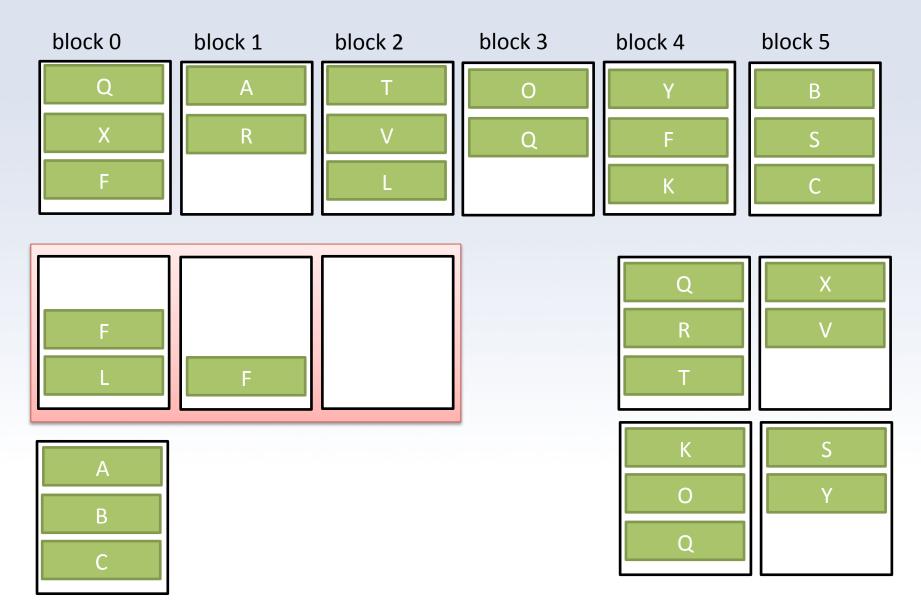


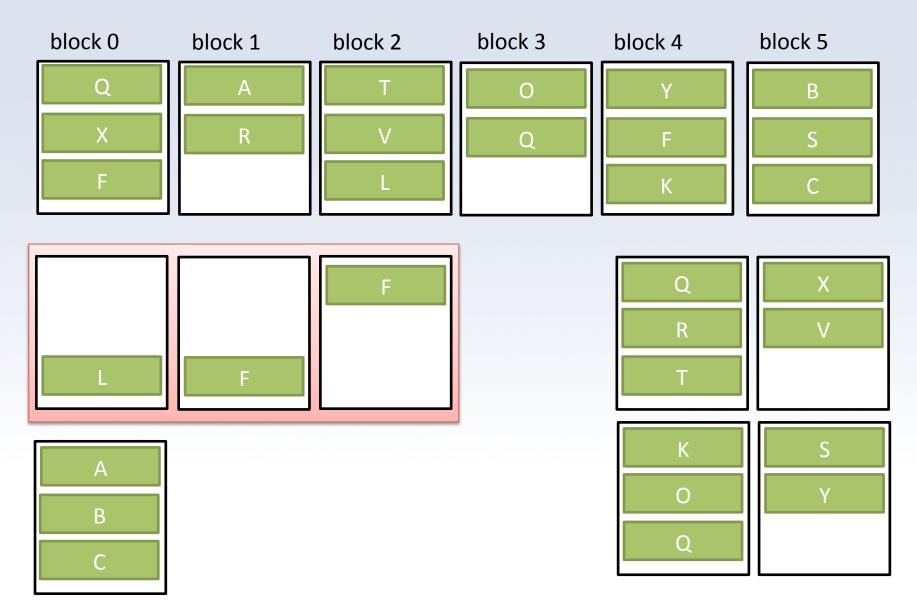


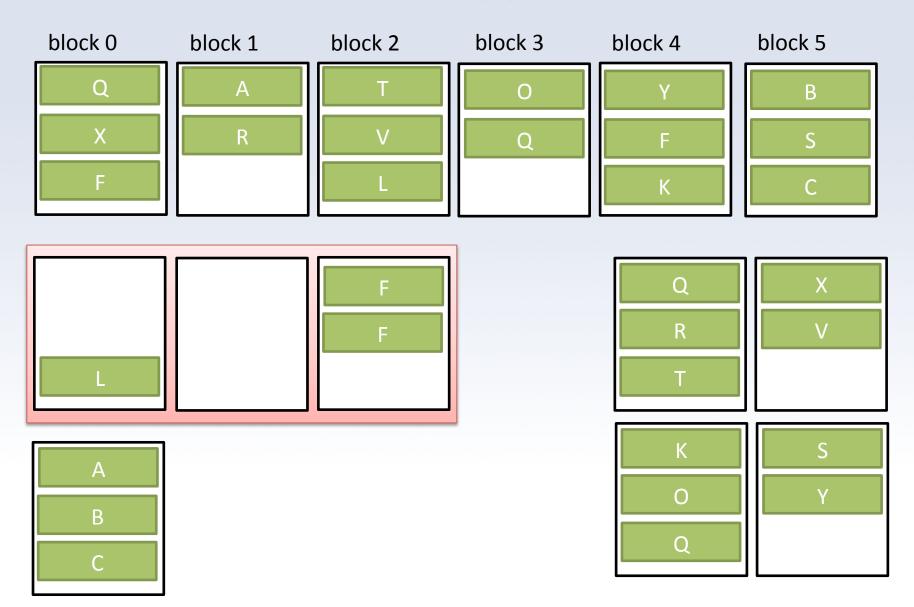


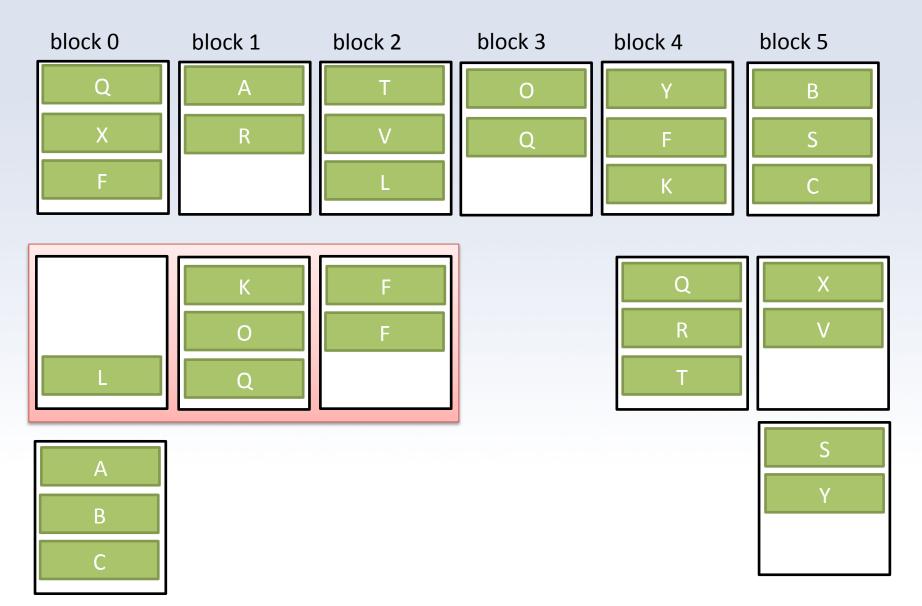


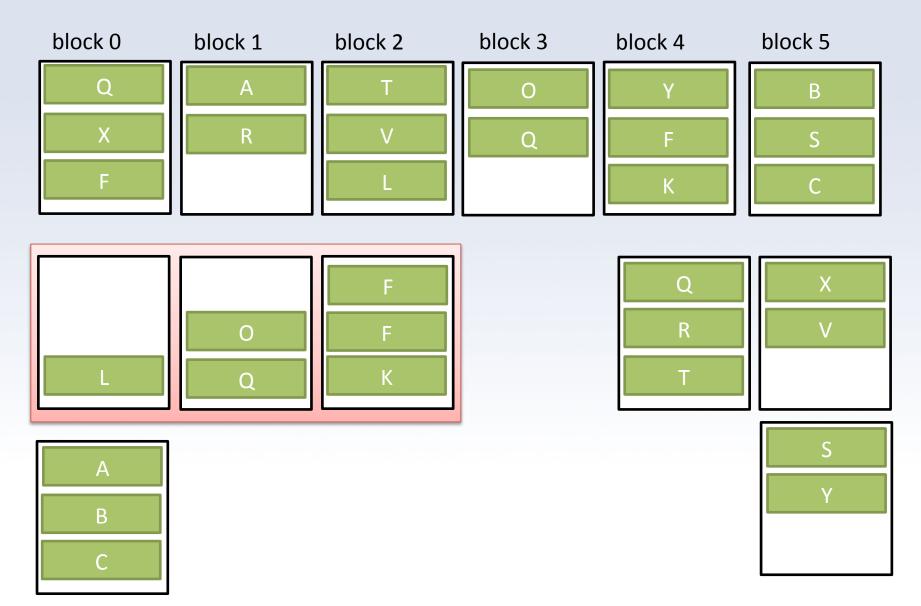


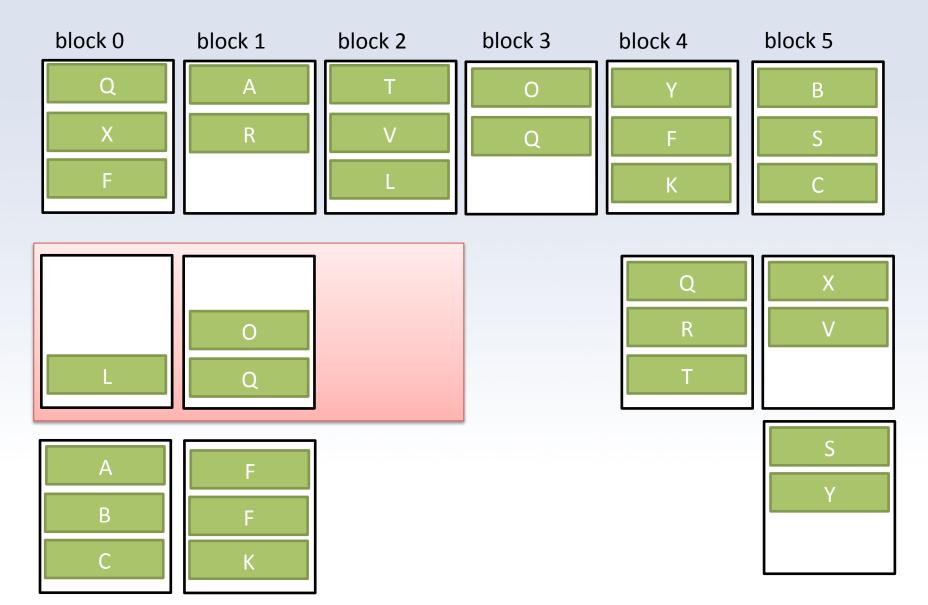


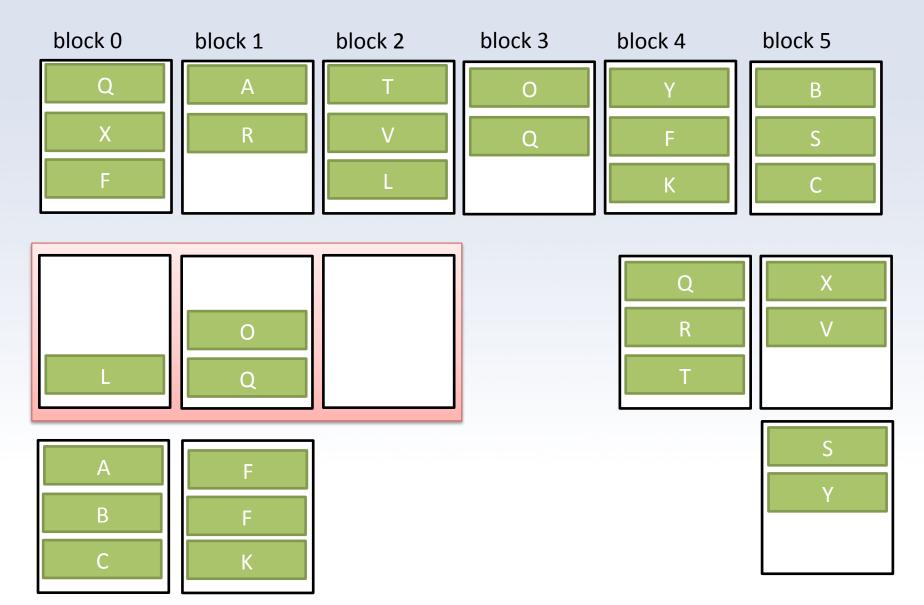


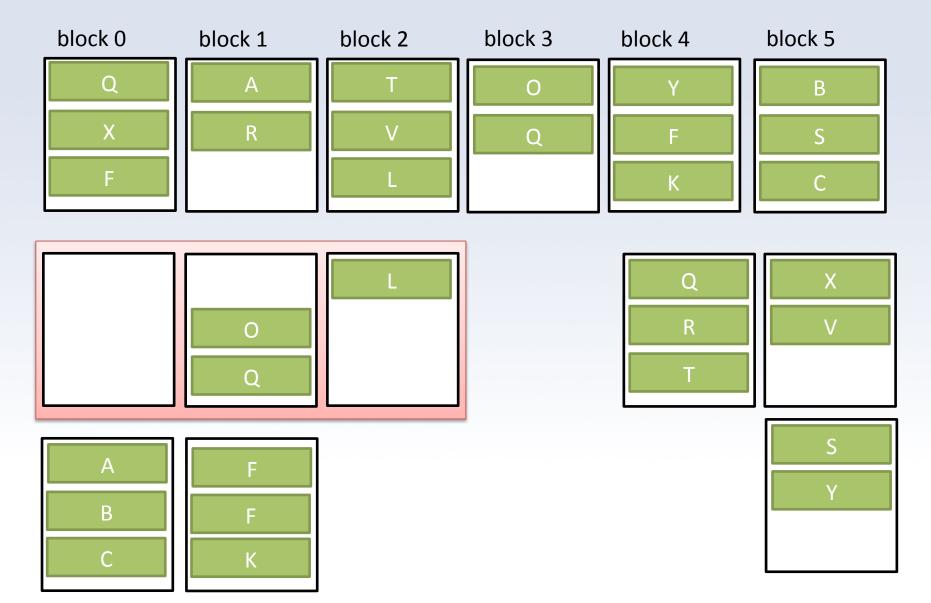


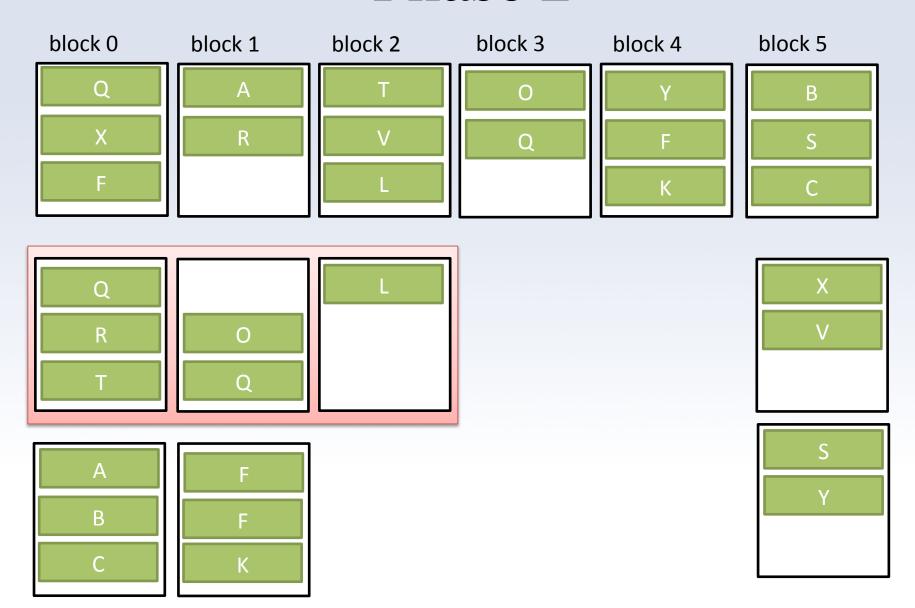


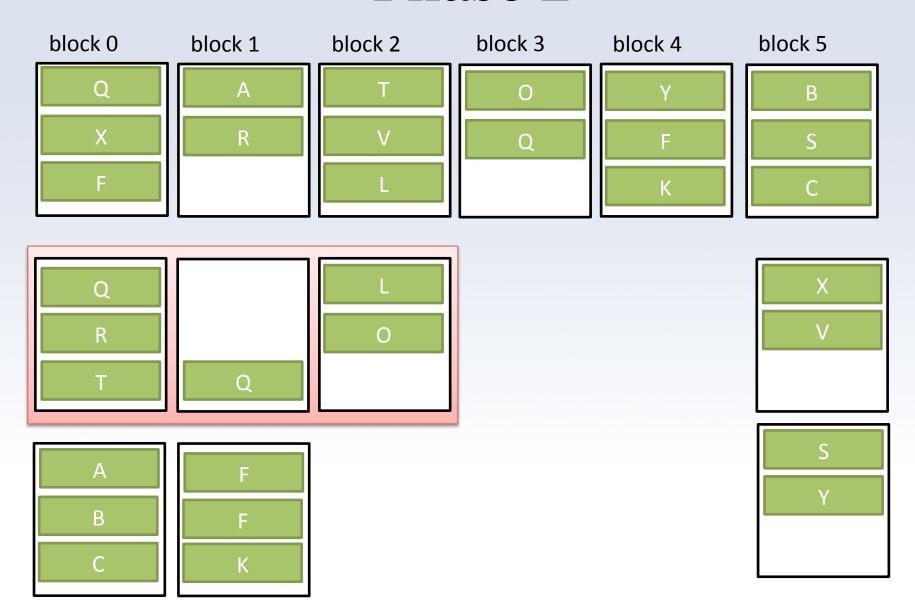


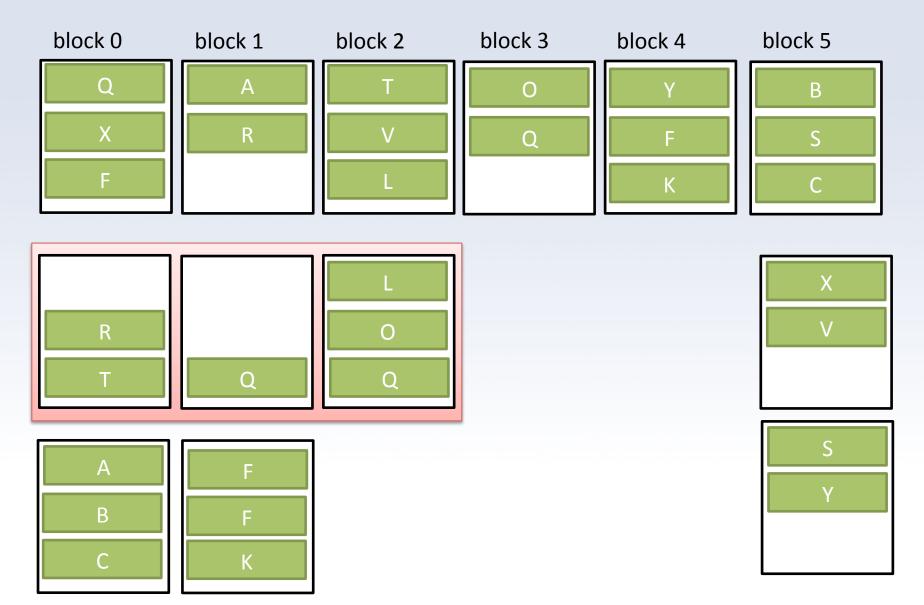


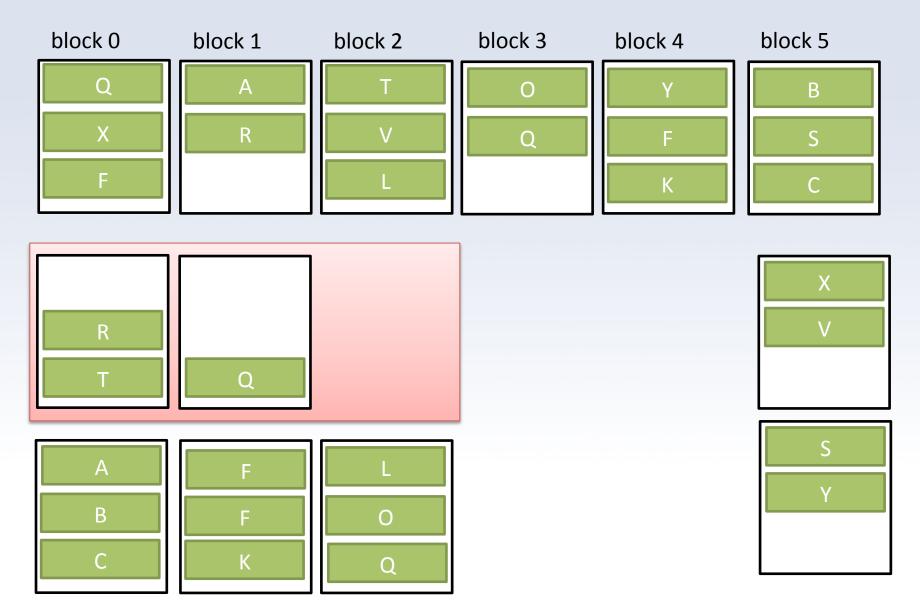


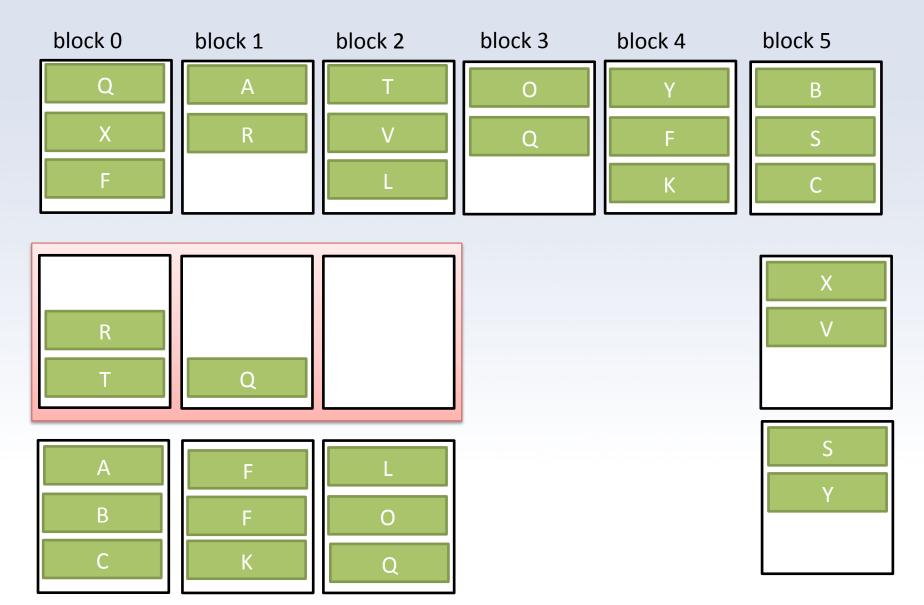


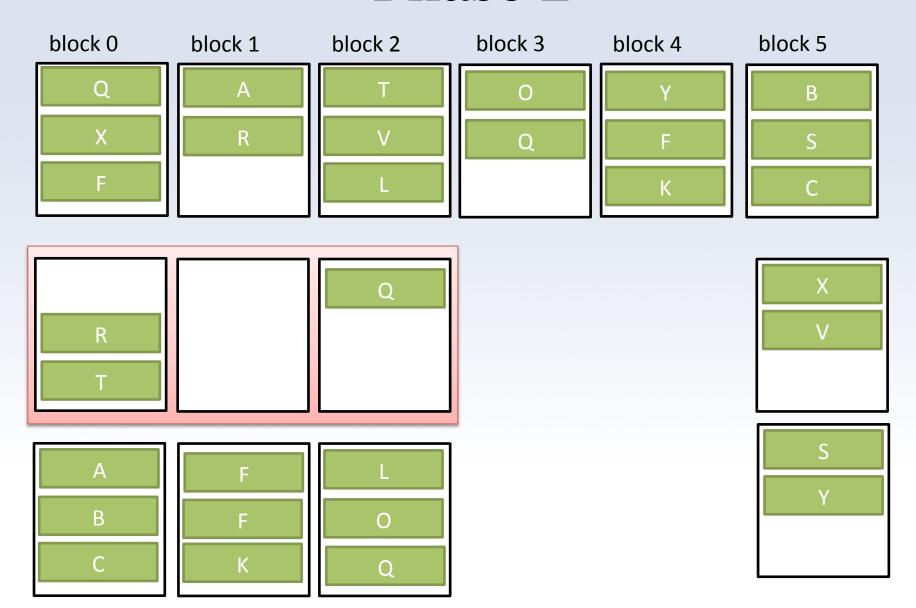


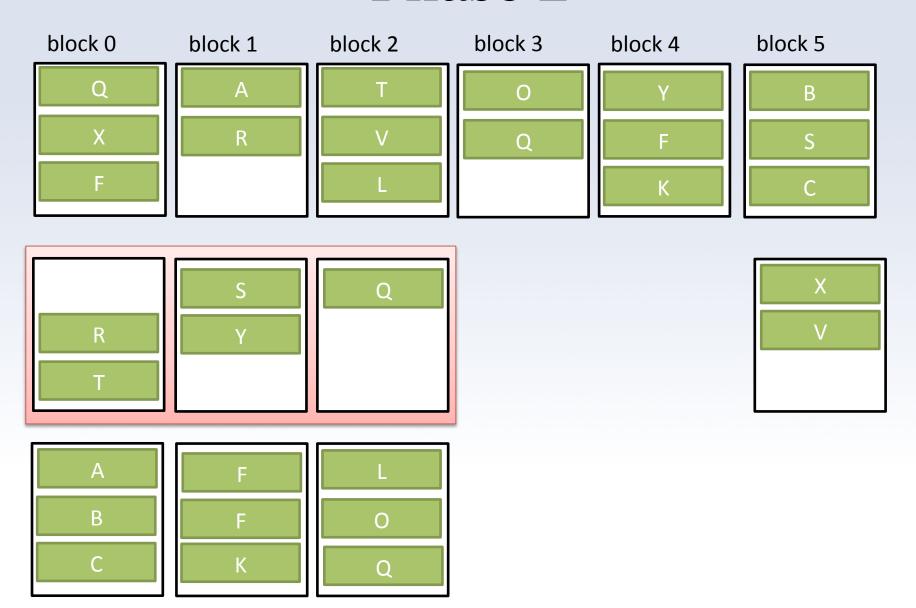


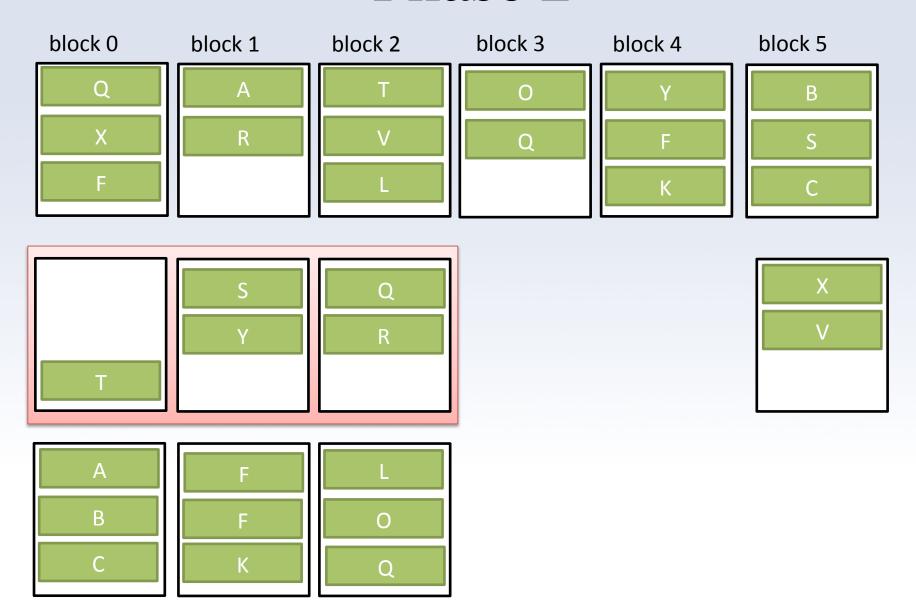


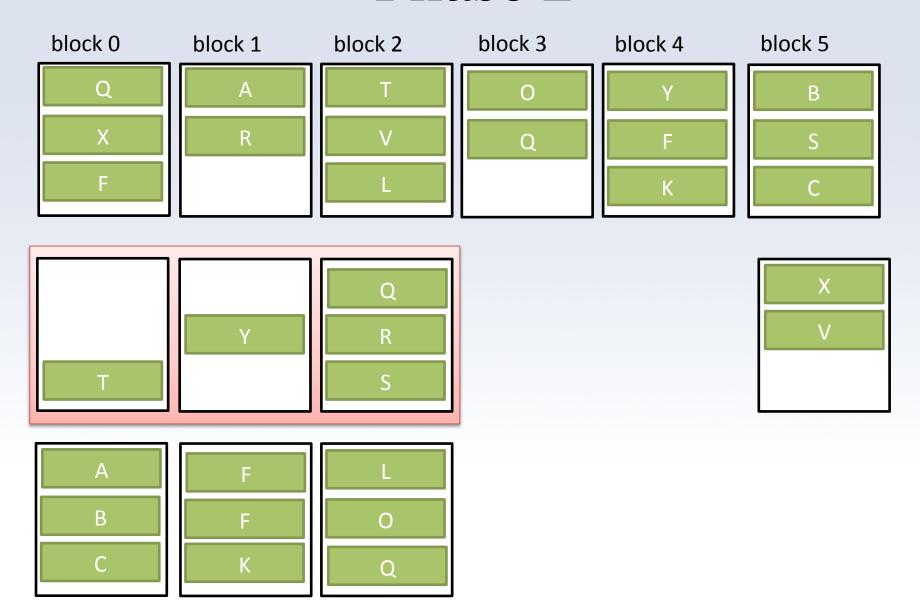


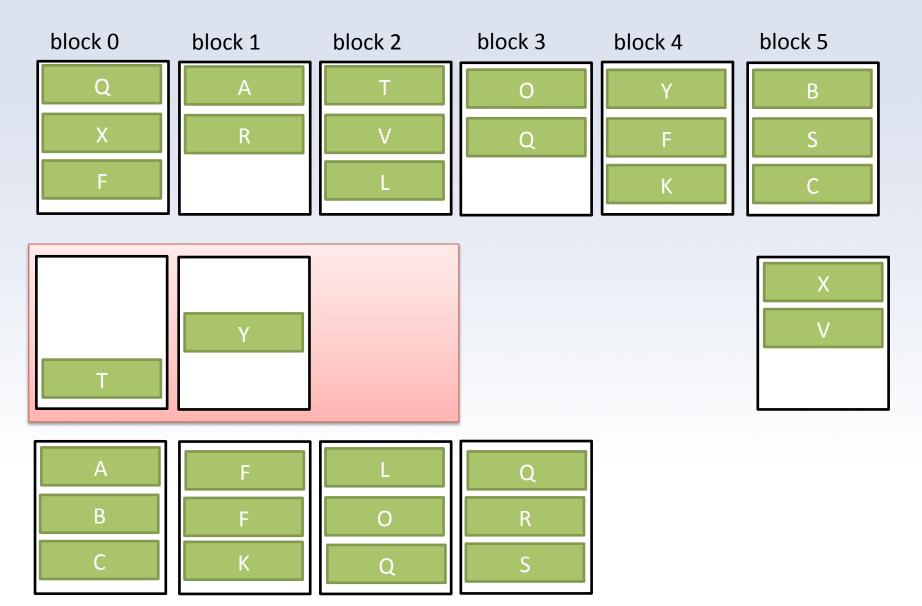


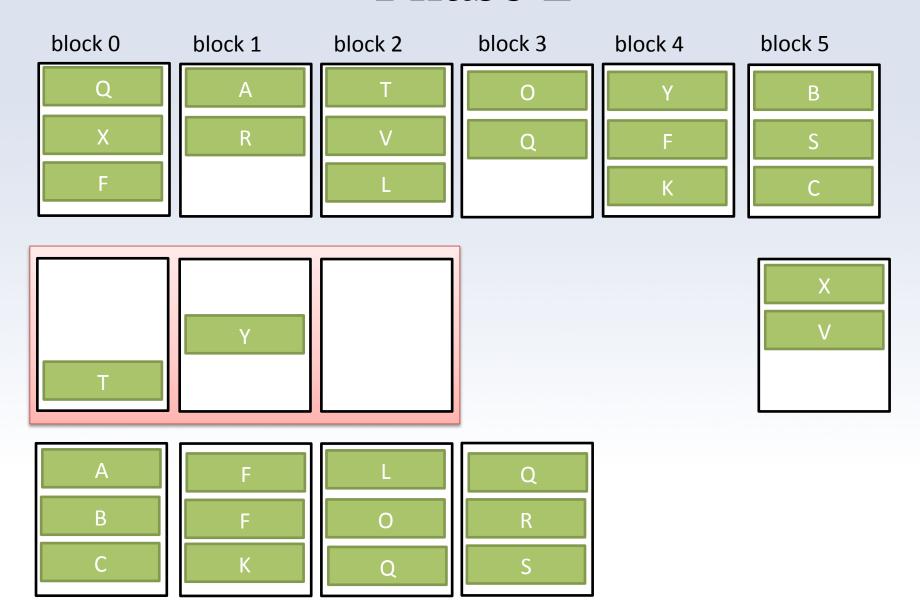


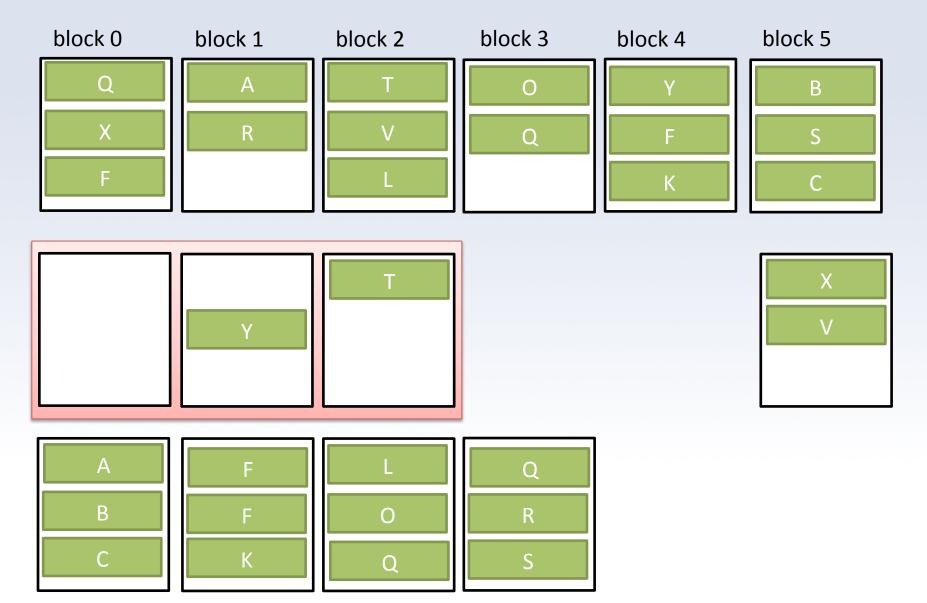


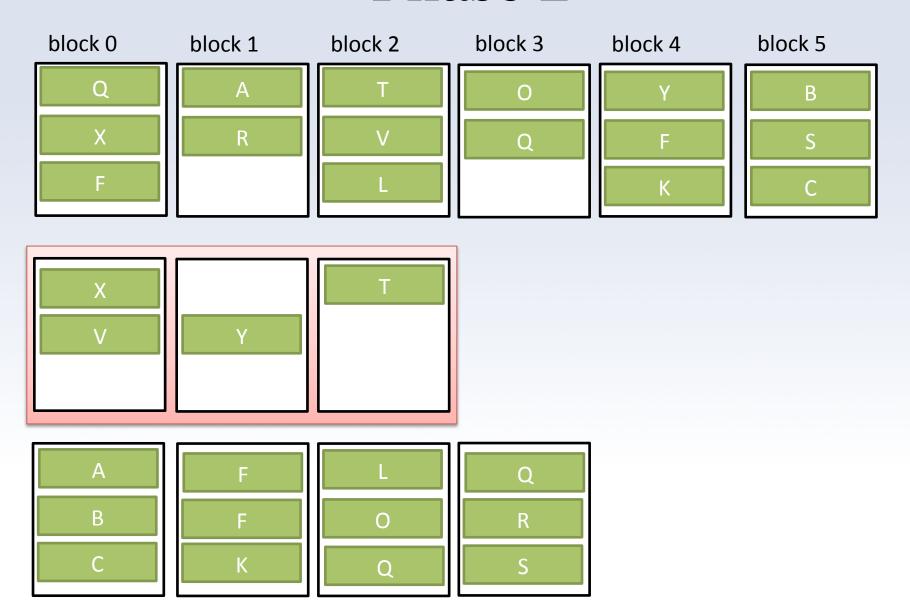


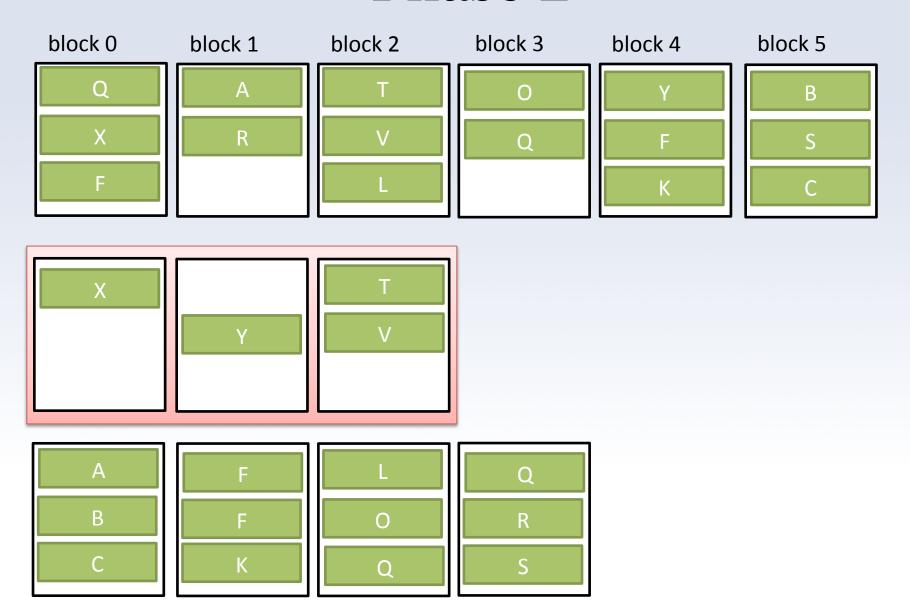


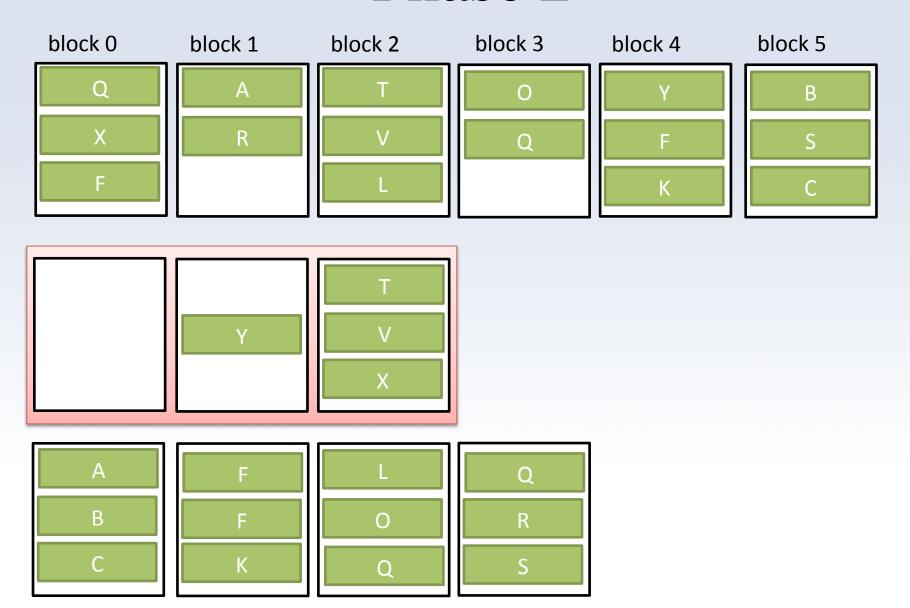


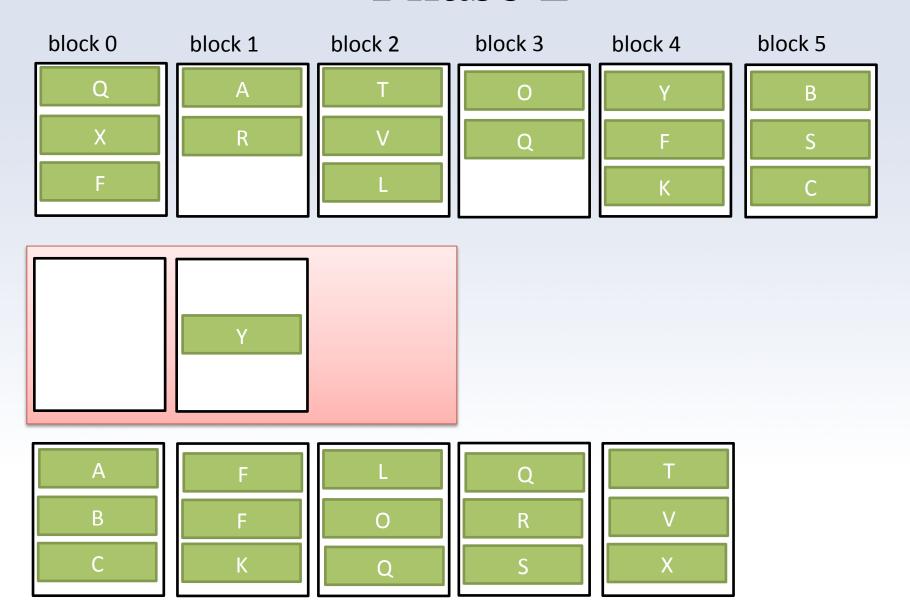


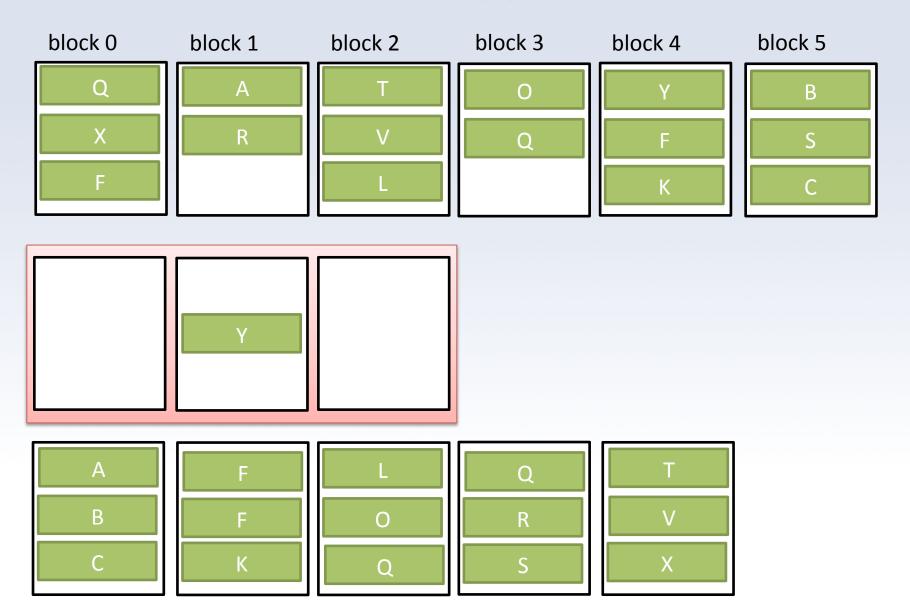


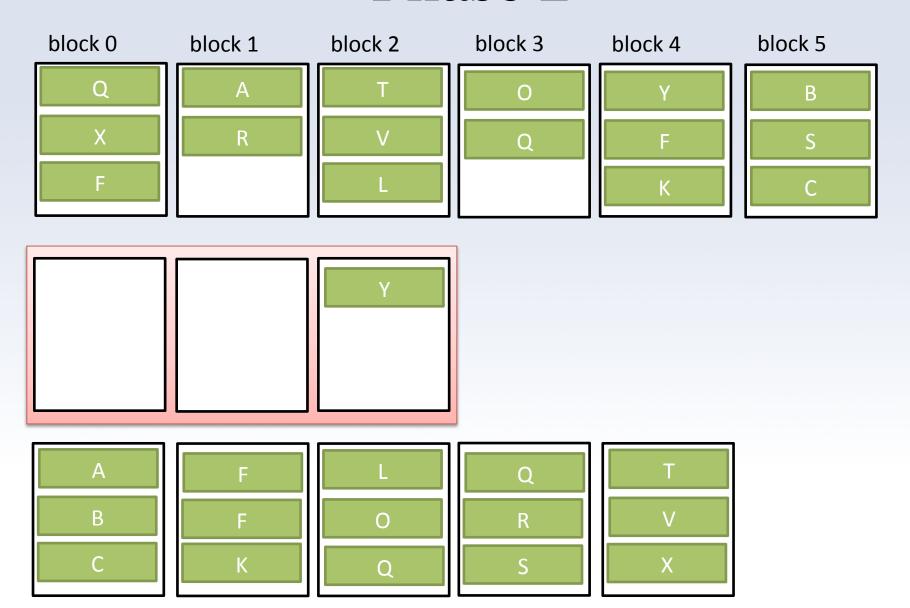


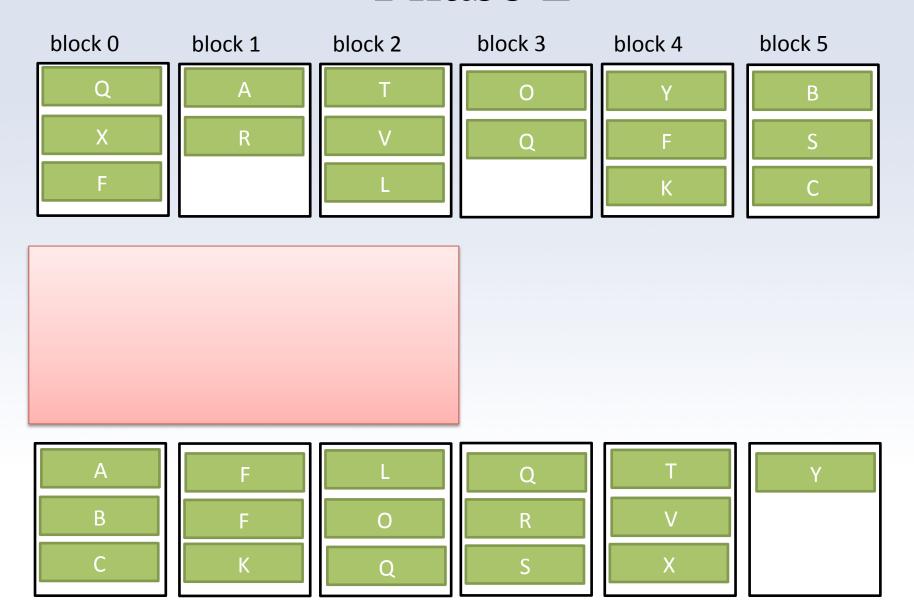












- Number of sublists is B/M
- To fit one block for each sublist, B/M≤M-1
- Rewritten, $B \le M(M-1)$
- Requires that $\sqrt{B(R)} \leq M$



- Read B(R) blocks in Phase 1
- Write B(R) blocks in Phase 1
- Read B(R) blocks in Phase 2
- Write B(R) blocks in Phase 2



- Read B(R) blocks in Phase 1
- Write B(R) blocks in Phase 1
- Read B(R) blocks in Phase 2
- Write B(R) blocks in Phase 2 Output



- Read B(R) blocks in Phase 1
- Write B(R) blocks in Phase 1
- Read B(R) blocks in Phase 2
- Write B(R) blocks in Phase 2 Output
- Cost is therefore 3B(R)



Sorting-based δ

- We can implement δ operation with our TPMMS as follows:
 - Create all of the sorted sublists as usual
 - 2. In phase 2, don't output duplicate entries (keep track of last record output and skip identical records)



Sorting-based y

- We can implement γ operation with our TPMMS as follows:
 - Create all of the sorted sublists as usual
 - 2. In phase 2, don't output duplicate entries. Instead, compute aggregation functions on duplicate entries



Sorting-based U

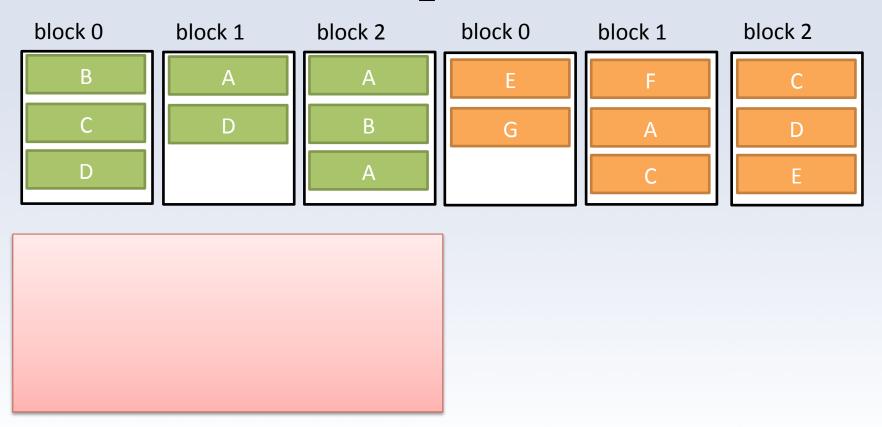
- We can implement ∪ operation with our TPMMS as follows:
 - 1. Create all of the sorted sublists for both R and S
 - 2. In phase 2, bring sorted lists for BOTH relations into buffers. Don't output duplicate entries.
- Essentially, treat R and S as one relation



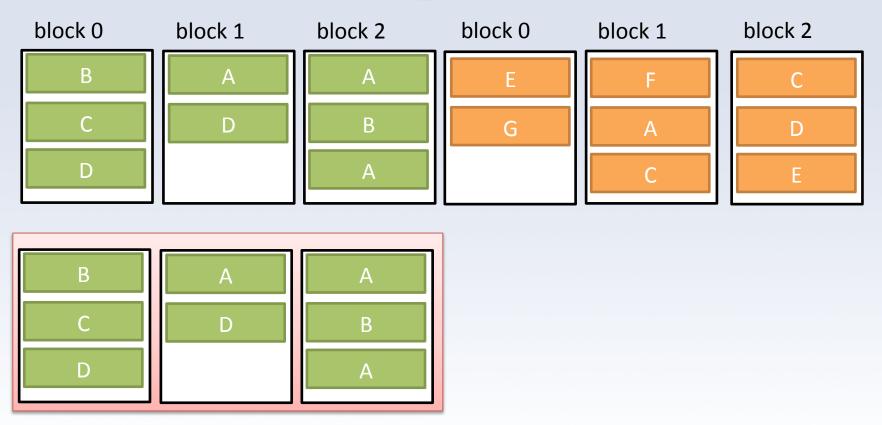
Sorting-based ∩

- We can implement ∩ operation with our TPMMS as follows:
 - 1. Create all of the sorted sublists for both R and S
 - 2. In phase 2, bring sorted lists for both relations into buffers, but keep them separate
 - 3. Only output when both relations contain the same value

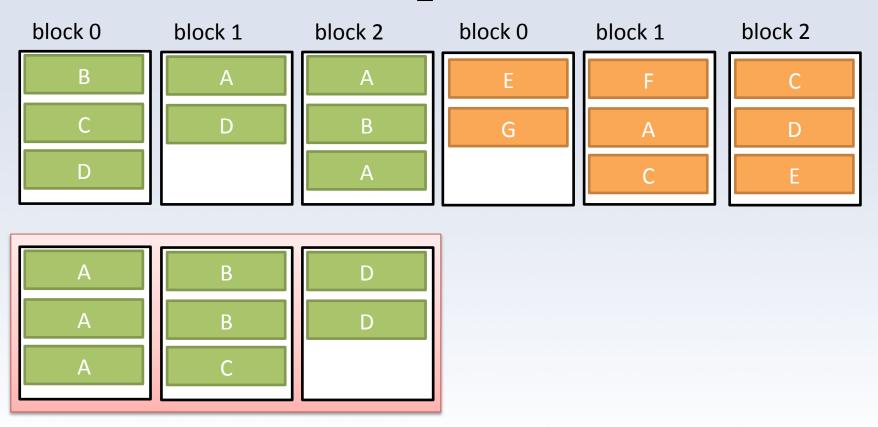




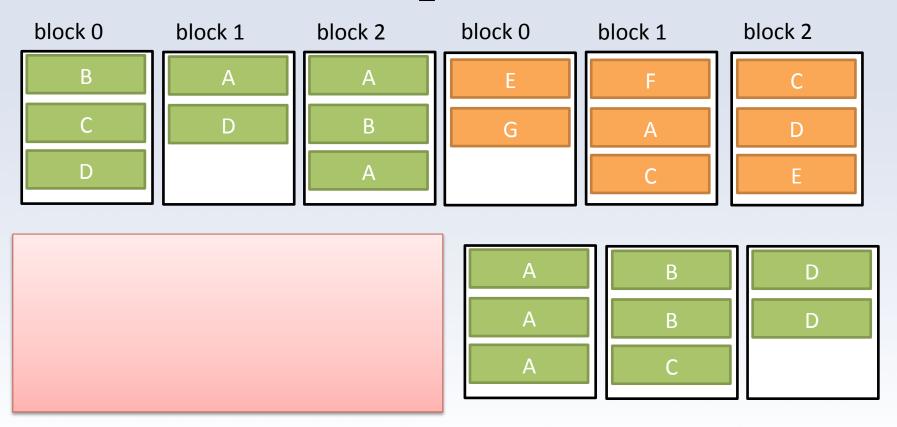




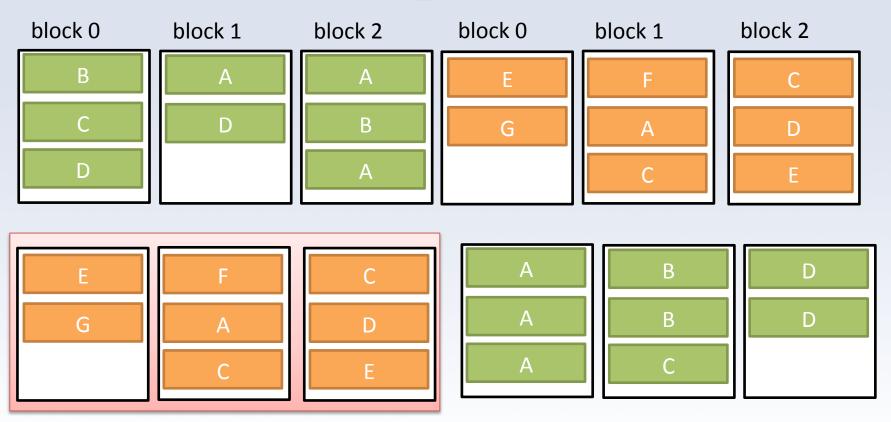




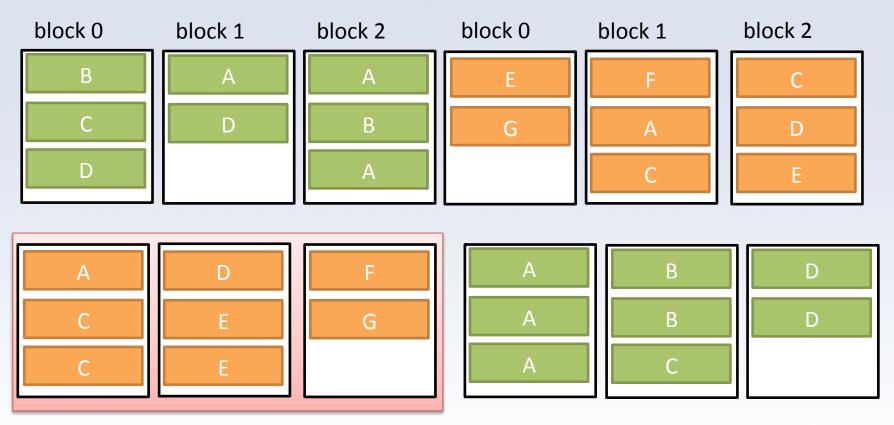




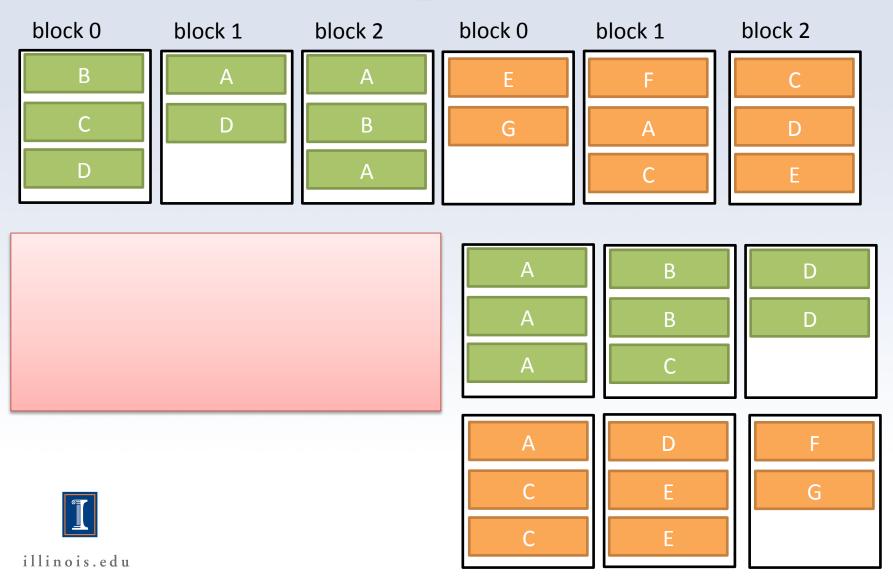


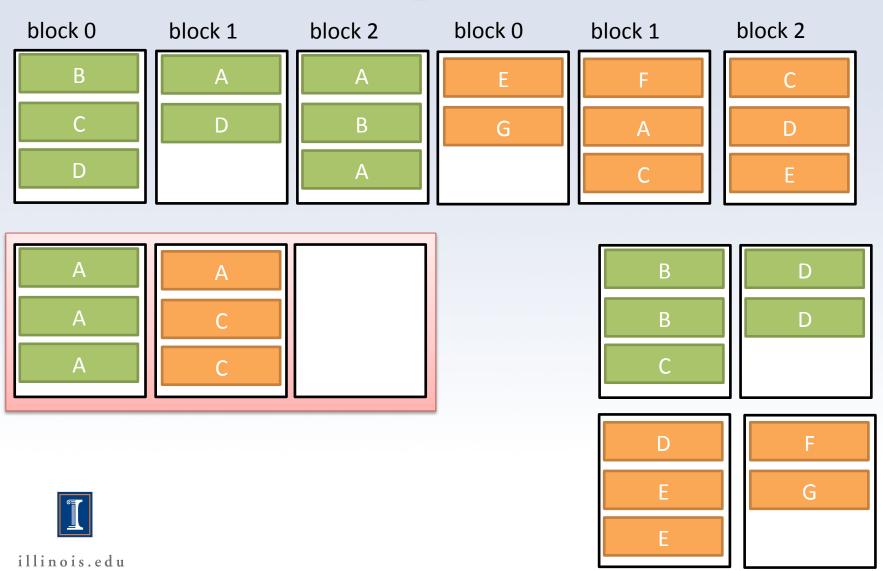


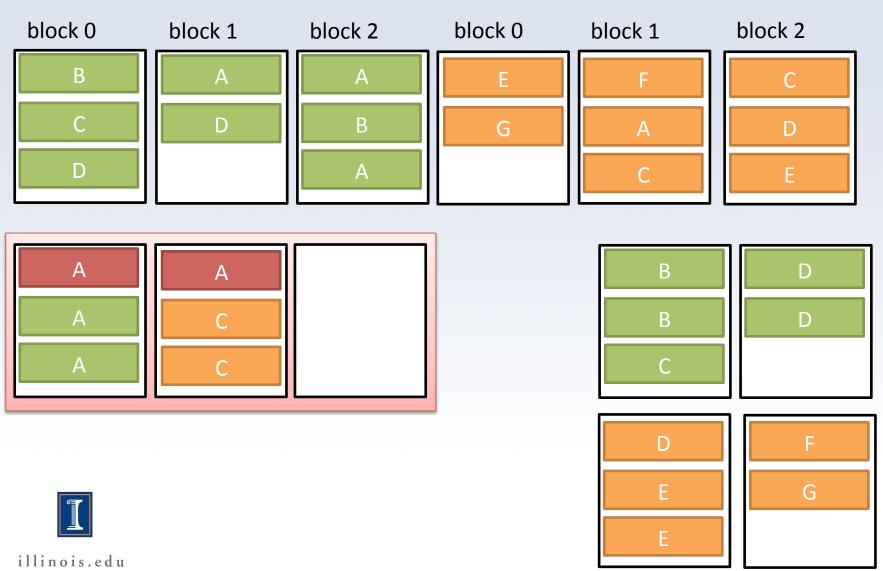


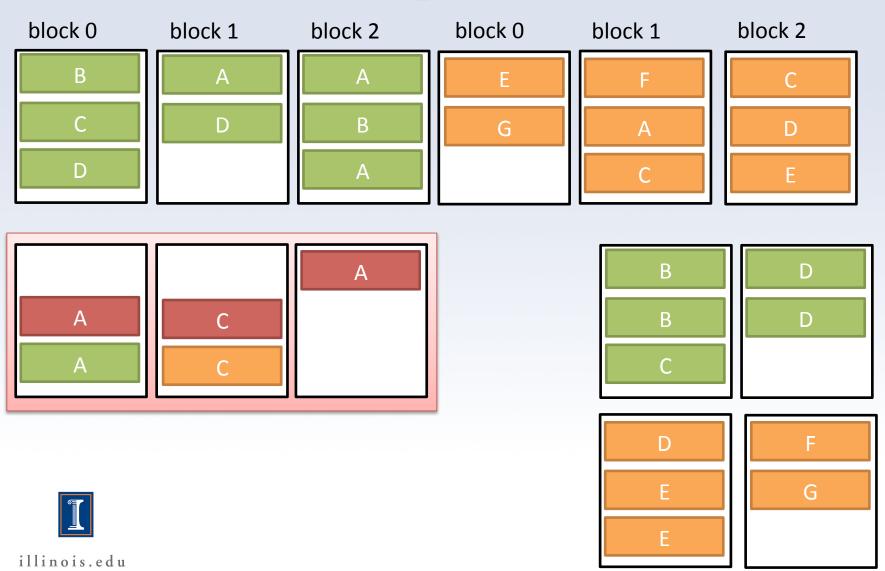


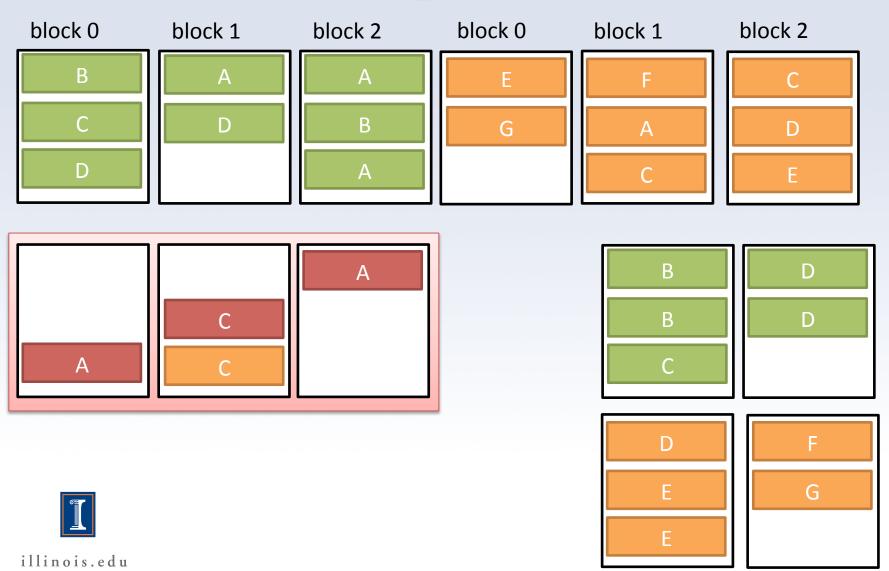


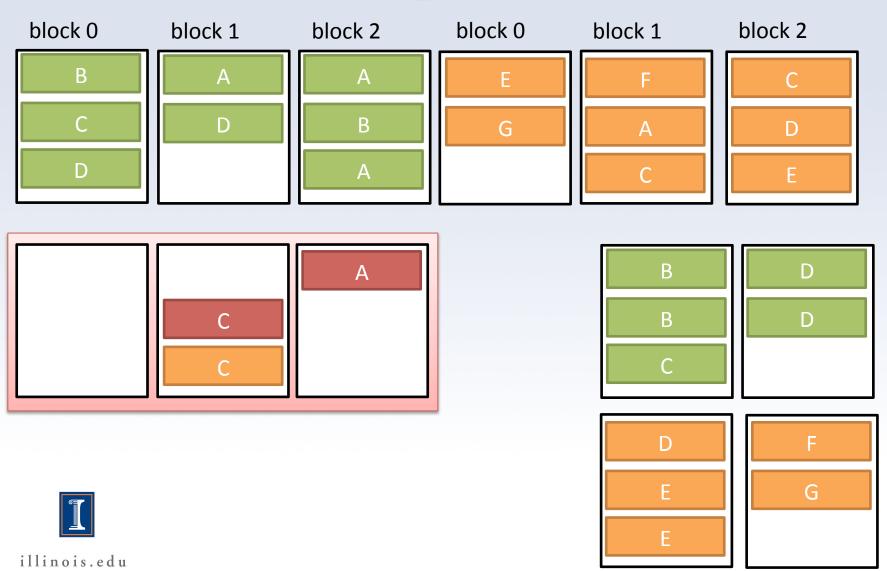


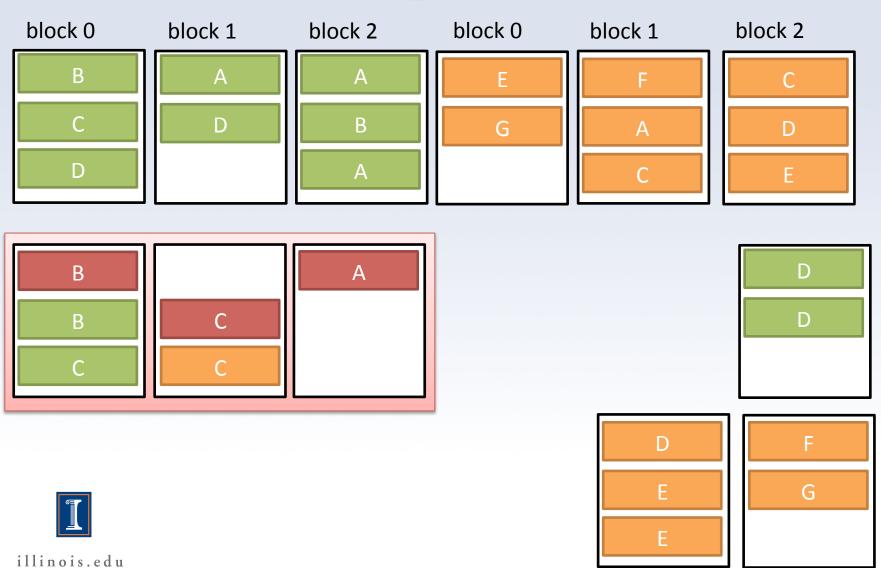


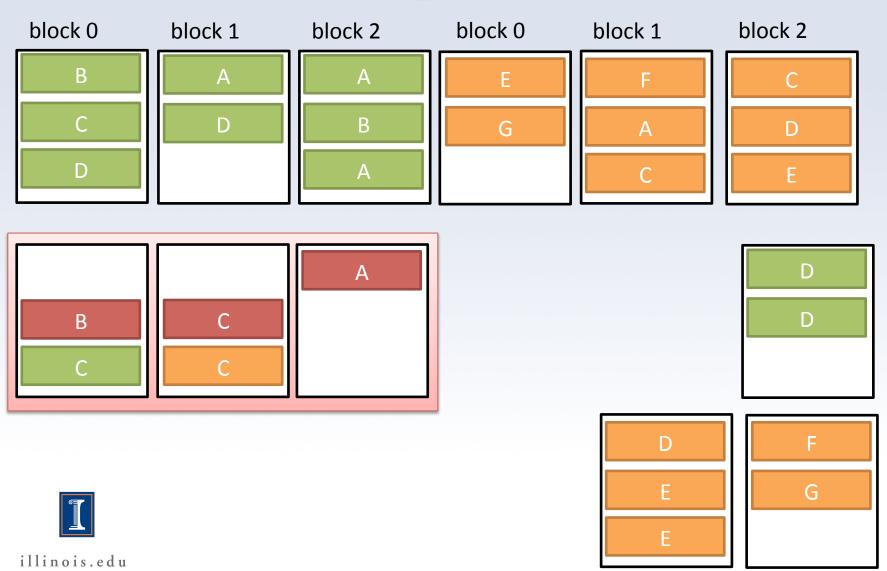


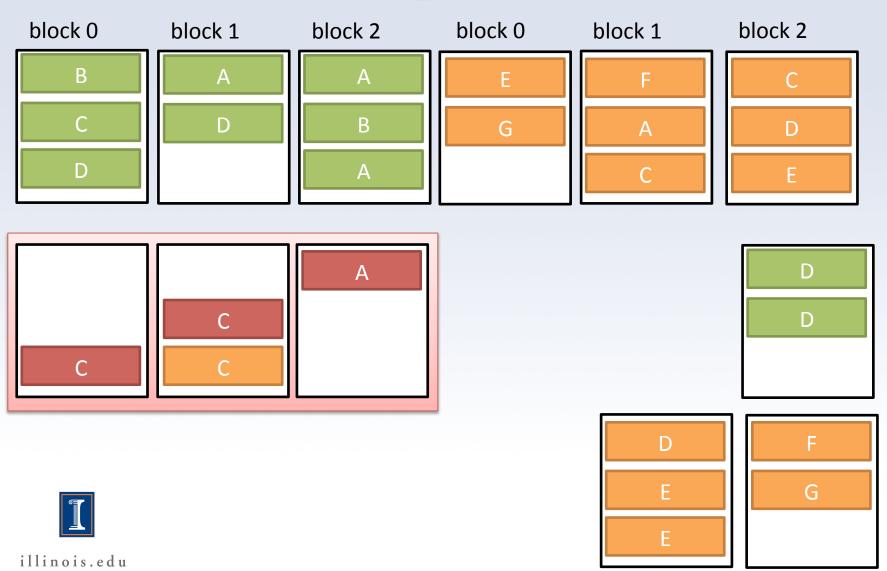


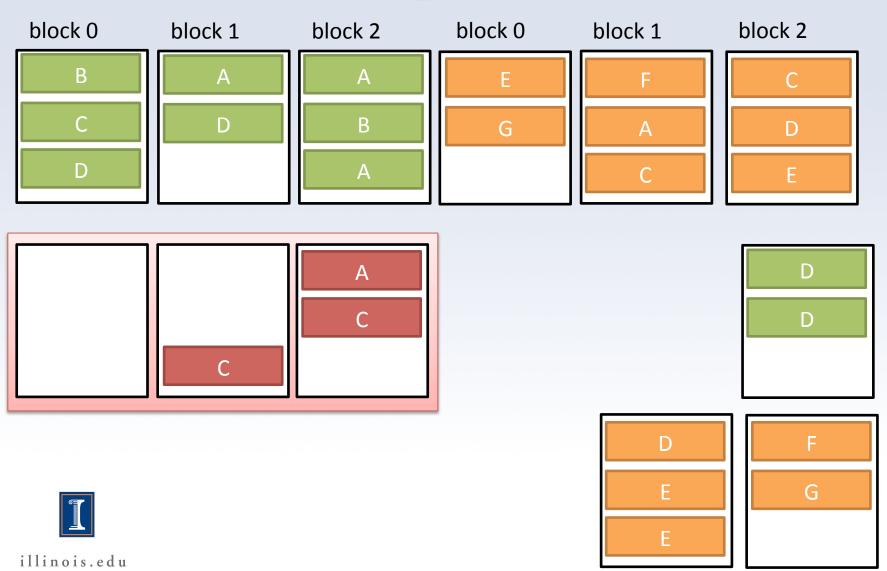


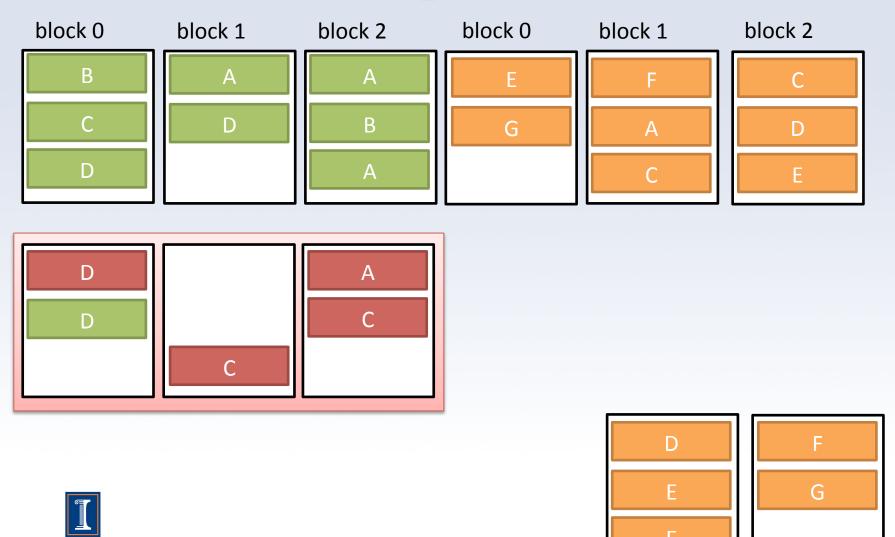




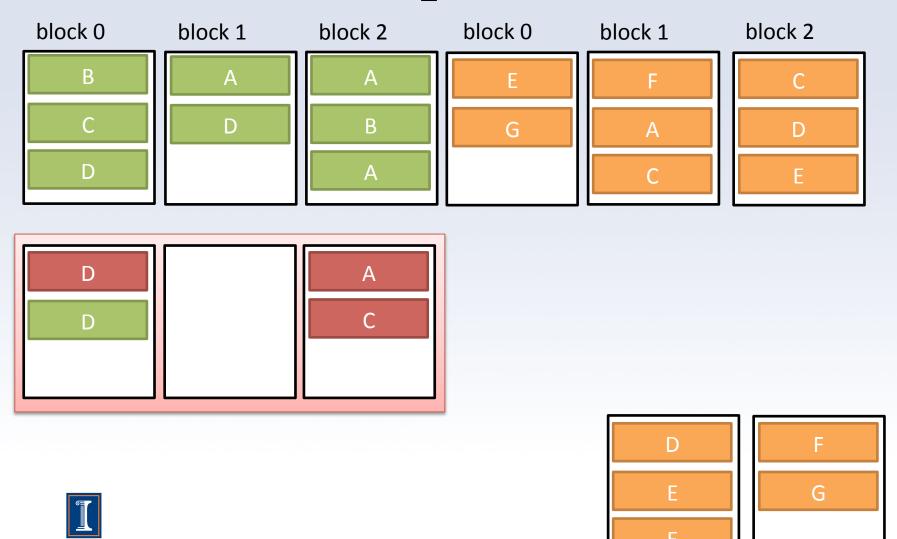




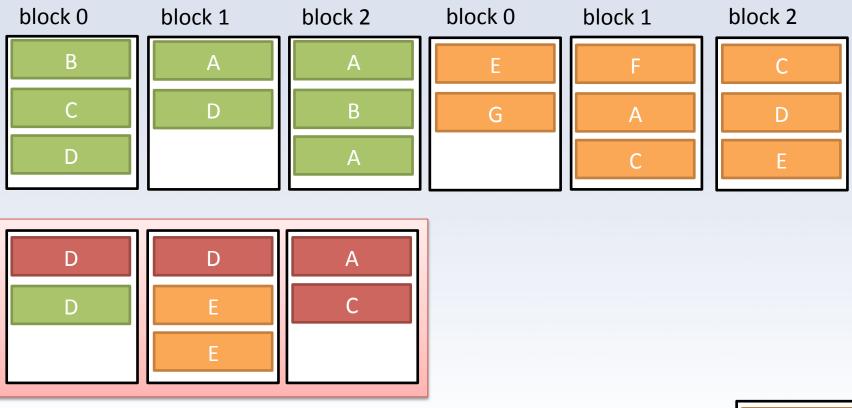




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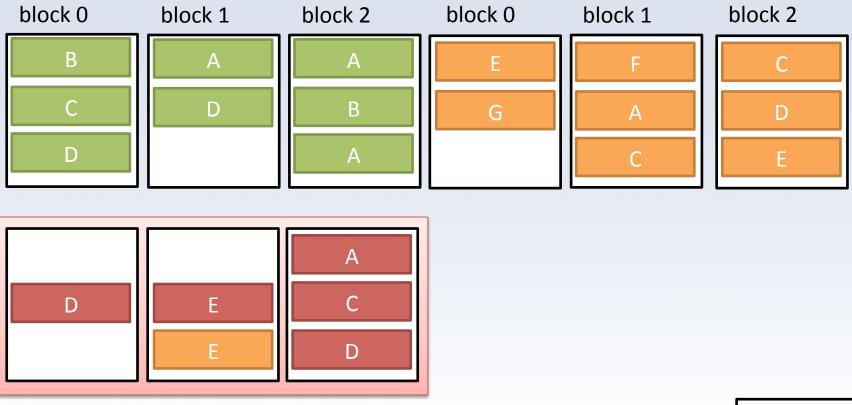


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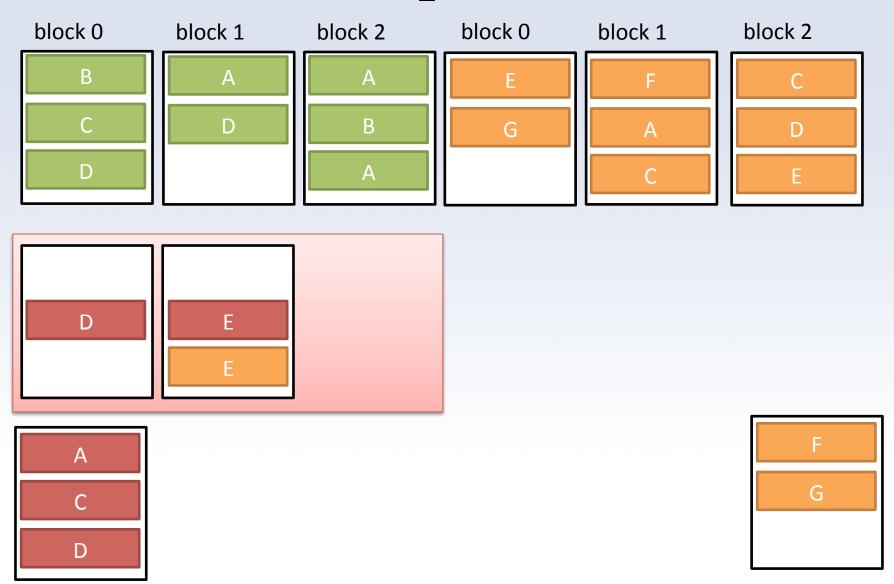


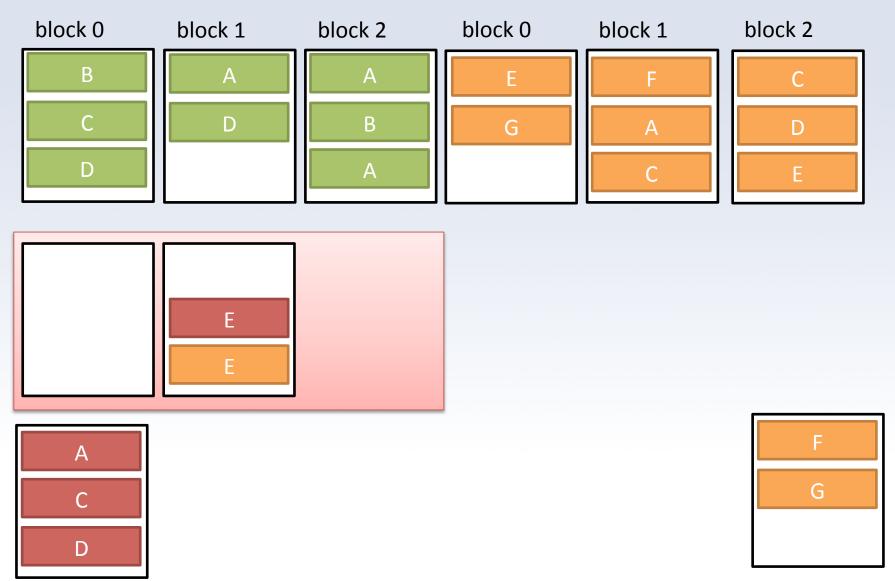


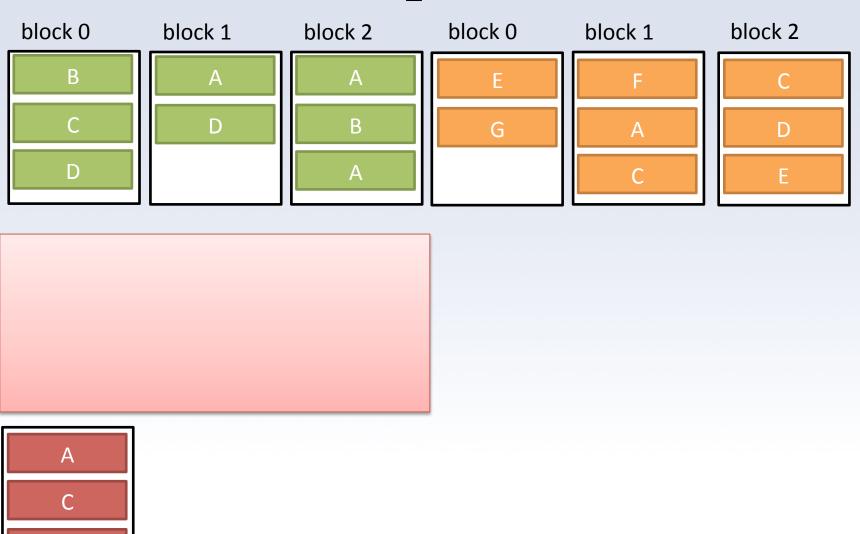












Sorting-based Set Difference

- We can implement R-S similar to the way we implemented $R \cap S$
- Instead of only outputting when value appears in both relations, we only output when it appears in R and not S



Simple Sort Join

- Implementing ⋈ is more complex
- Problem: we have to combine *all* of the records that share the join attributes
 - Those tuples might be bigger than memory!
 - We'd have to use nested loop join
- Simple sort join makes as much memory available for join as possible
- Similar to sort based intersection

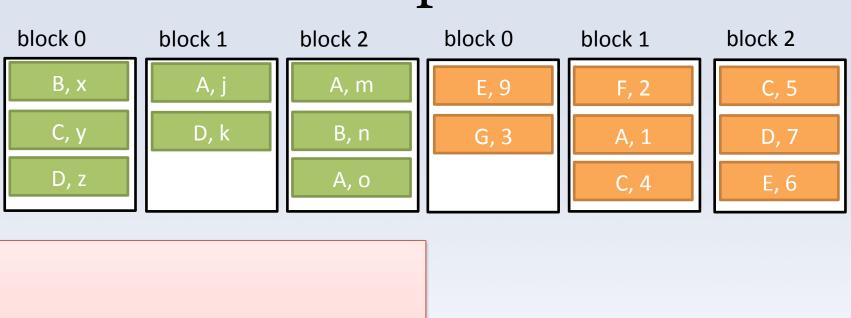


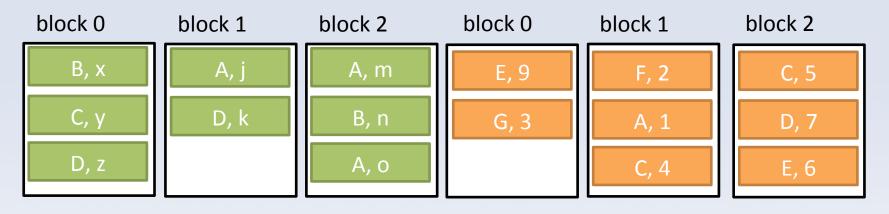
Simple Sort Join

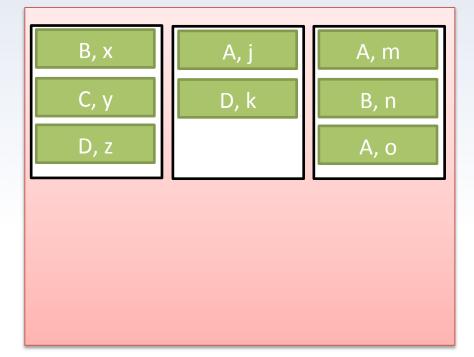
Joining R and S on attributes Y

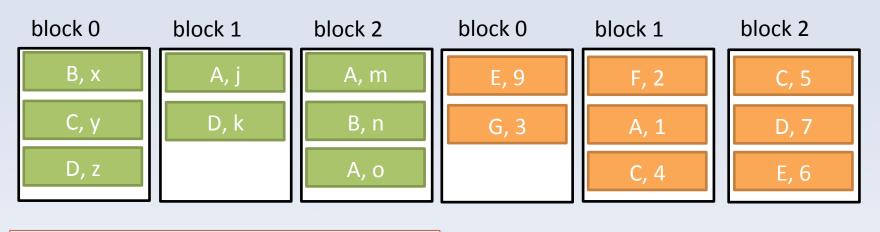
- 1. Sort R using TPMMS using Y as key
- 2. Sort S using TPMMS using Y as key
- 3. Use two buffers to merge sorted lists, outputting when tuples can be joined
 - expand these two buffers until sure we have all values

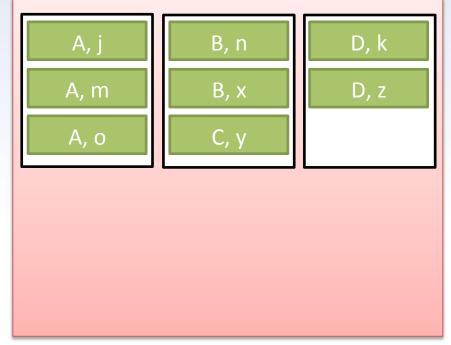


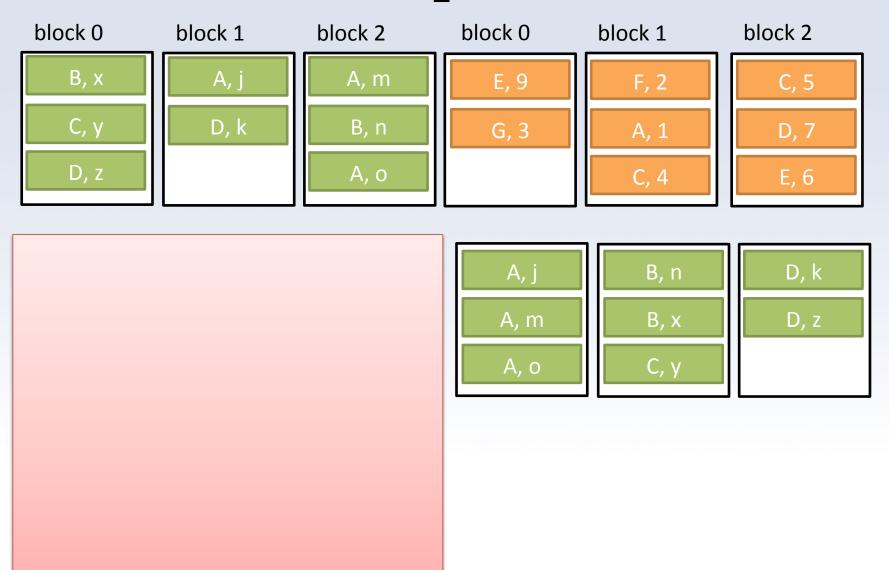


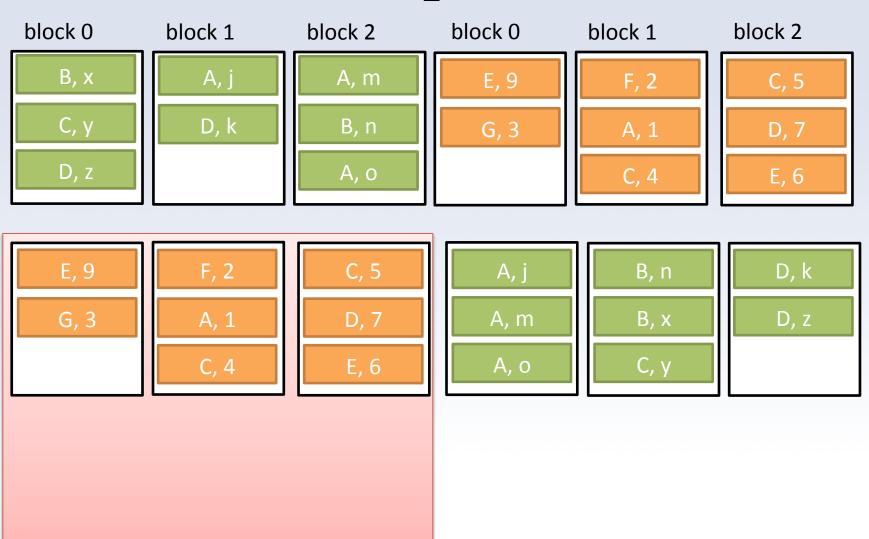


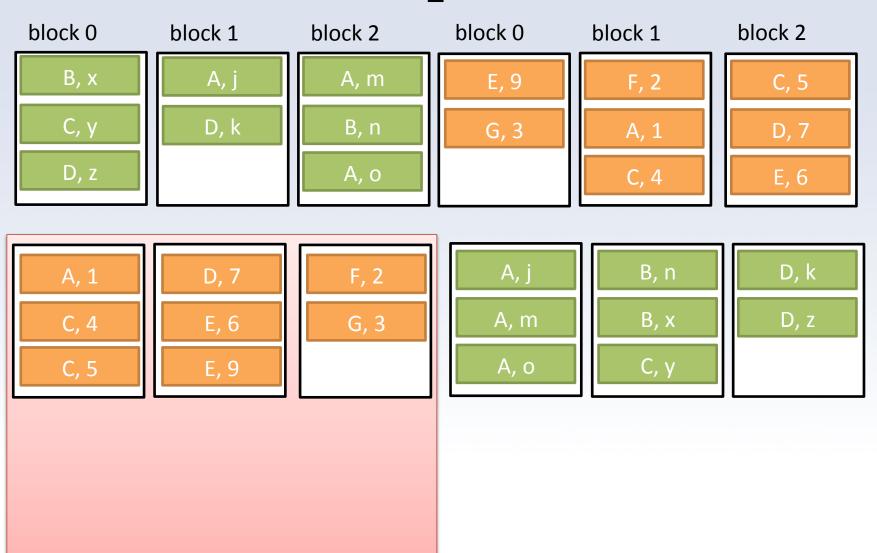


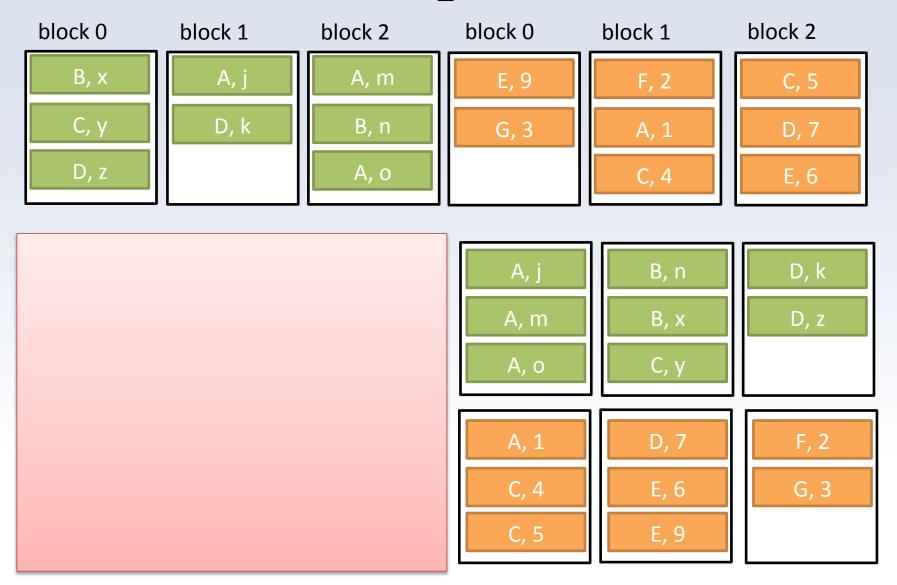






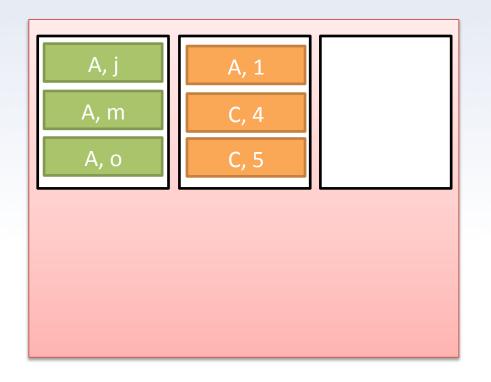


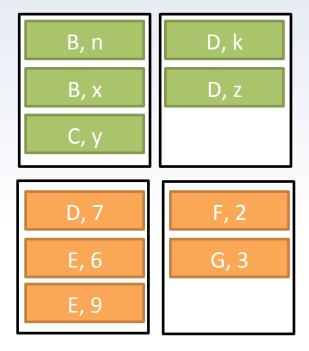


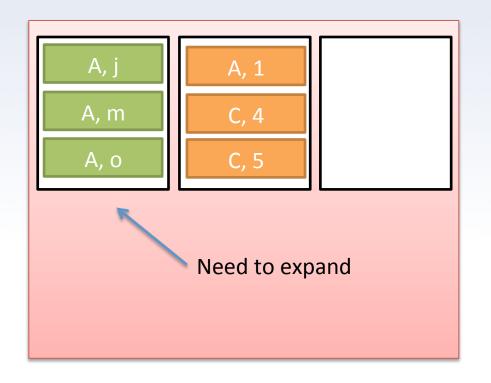


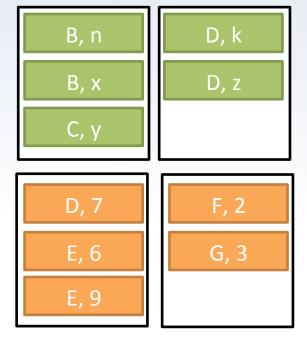
Example $R\bowtie S$



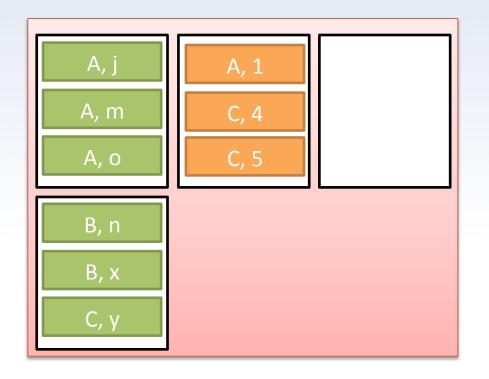


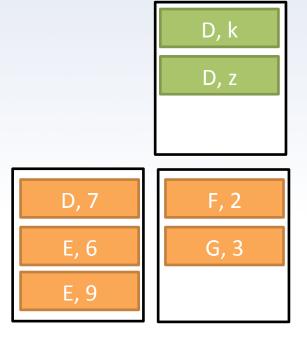


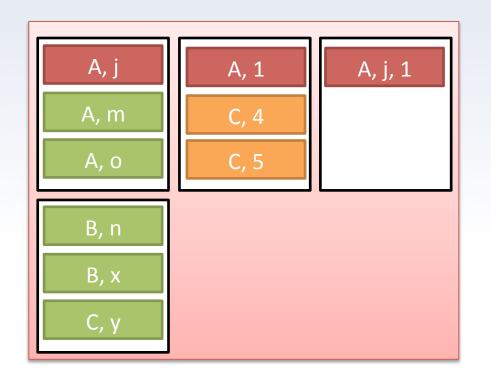




Example R S

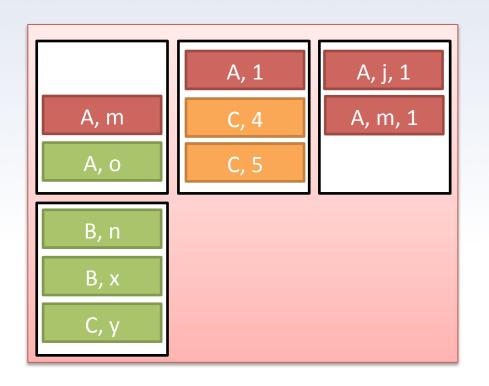






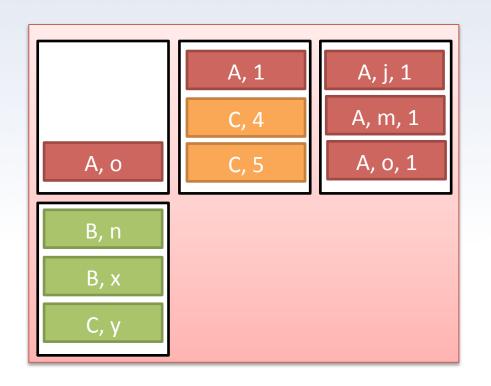
D, k
D, z

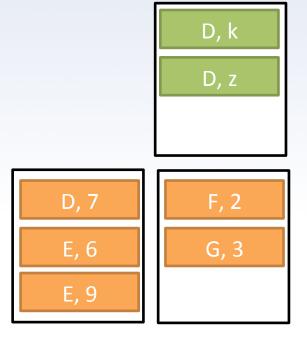
D, 7
F, 2
E, 6
G, 3
E, 9



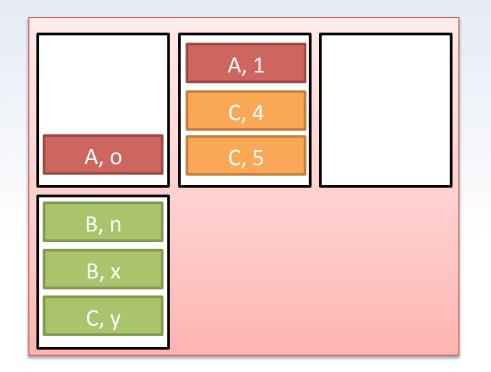
D, k
D, z

D, 7
F, 2
E, 6
G, 3
E, 9





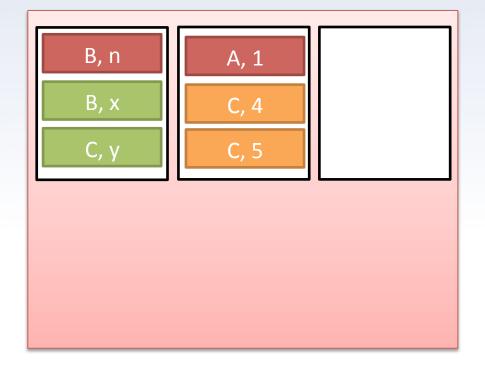
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A, m, 1
A, o, 1

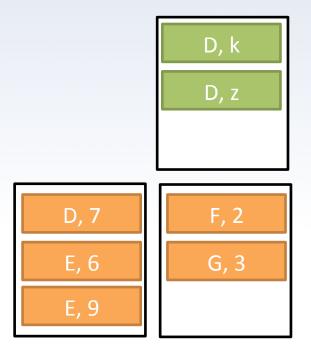


D, k
D, z

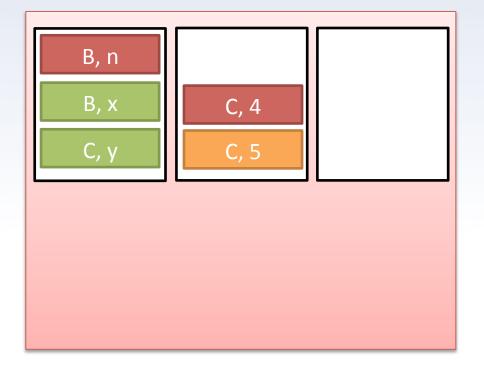
D, 7
F, 2
E, 6
E, 9

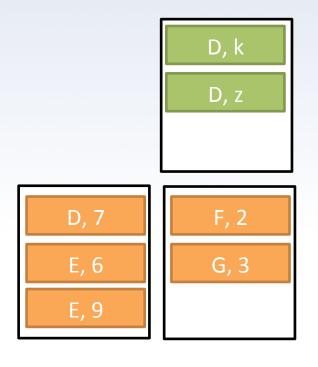
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A, m, 1
A, o, 1



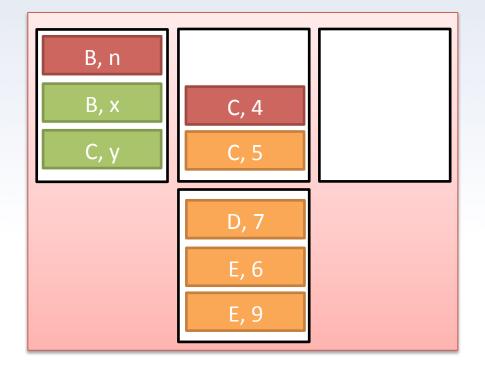


A, j, 1
A, m, 1
A, o, 1





A, j, 1
A, m, 1
A, o, 1



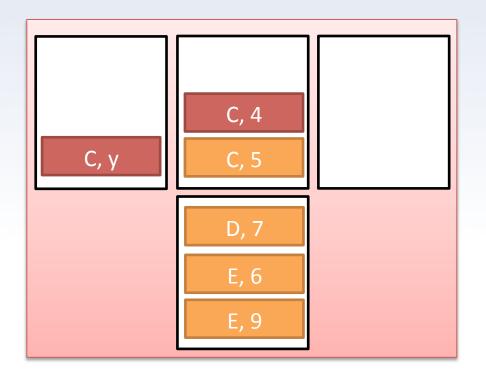
D, k
D, z
F, 2
G, 3

Example R S

A, j, 1

A, m, 1

A, o, 1



D, k

D, z

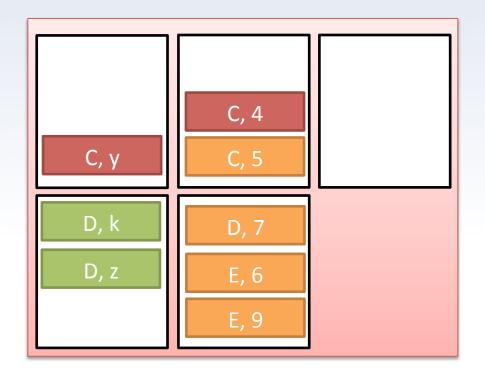
F, 2

G, 3

A, j, 1

A, m, 1

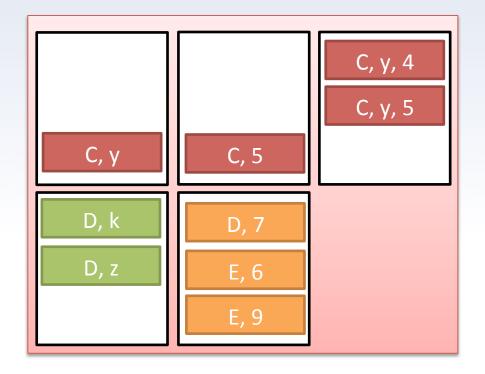
A, o, 1



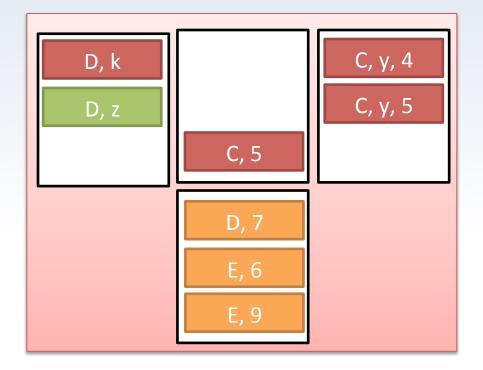
F, 2

G, 3

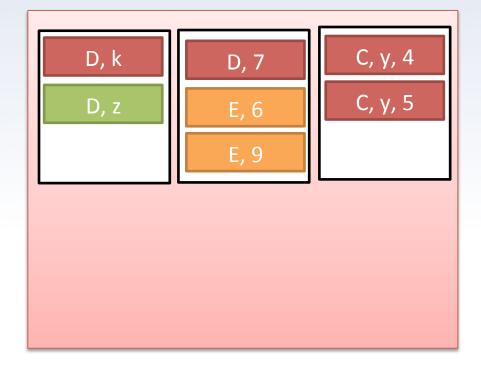
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A, m, 1
A, o, 1



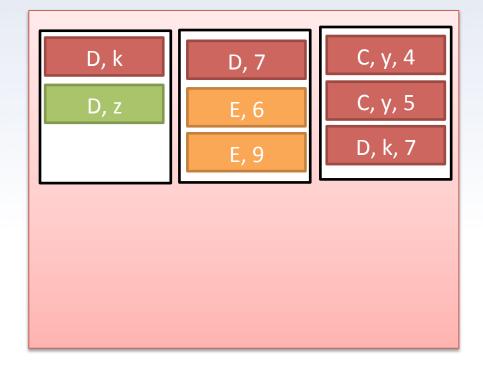
A, j, 1
A, m, 1
A, o, 1



A, j, 1
A, m, 1
A, o, 1



A, j, 1
A, m, 1
A, o, 1



A, j, 1

A, m, 1

A, o, 1

C, y, 4

C, y, 5

D, k, 7

D, k

D, z

D, 7

E, 6

E, 9

F, 2

G, 3

Example R\sim S

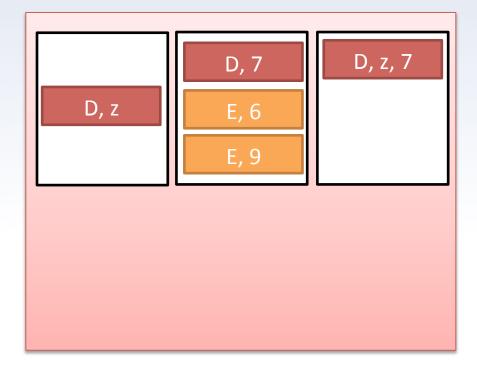
A, j, 1

A, m, 1

A, o, 1

C, y, 4 C, y, 5

D, k, 7



F, 2

G, 3

Example R\sim S

A, j, 1

A, m, 1

A, o, 1

C, y, 4

C, y, 5

D, k, 7

D, 7

D, z, 7

E, 6

E, 9

F, 2 G, 3

Example R\sim S

A, j, 1

A, m, 1

A, o, 1

C, y, 4

C, y, 5

D, k, 7

D, z, 7

Analysis

- Requires both B(R)<M² and B(S)<M²
- Shared join attributes must fit in memory, so max(B(R),B(S))<M²
- Requires 5(B(R)+B(S)) I/O operations
 - Two reads and writes for first pass
 - Two reads and writes for second pass
 - One read and write for joining relations



Sort Join

- We can do better:
 - Assume all records with common join attributes fit in memory
 - Build the join into the second phase
 - Create sorted sublists using join attributes
 - Bring smallest block for each sublist into memory and perform joins



Sort Join

- Requires all records with same join attribute fit in memory
 - $-(B(R)+B(S))< M^2$
- Only requires two passes:
 - -3(B(R)+B(S)) I/O operations

Sorting-Based Summary

Operator	M required	I/O Cost
δ, γ	٧B	3B
U ,	V(B(R)+B(S))	3(B(R)+B(S))
\bowtie	√(max(B(R),B(S))	5(B(R)+B(S))
\bowtie	V(B(R)+B(S))	3(B(R)+B(S))



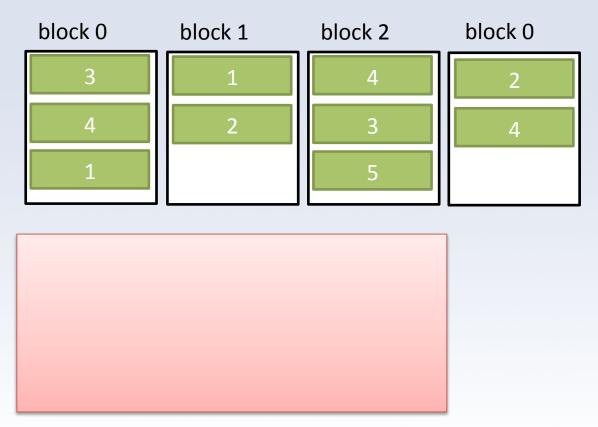
2 Pass Hash Based Algorithms

- We can also design two pass algorithms based on hashing
- First pass: hash the relation into M-1 buckets
- Second pass: process each bucket using one pass algorithms
 - requires that each bucket fits in memory

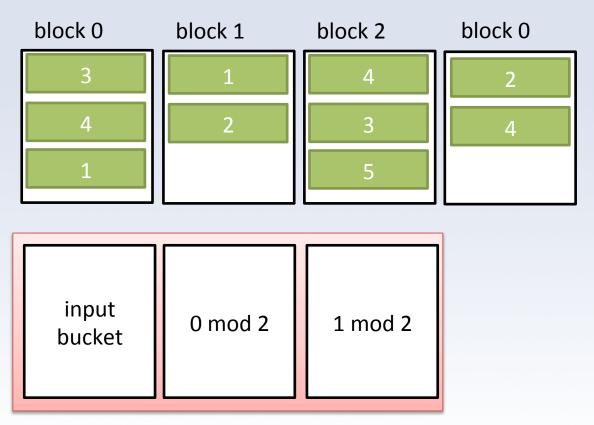
- Hash the relation into M-1 buckets
- Identical records should hash to the same bucket
- Bring each bucket into memory and output each tuple only once



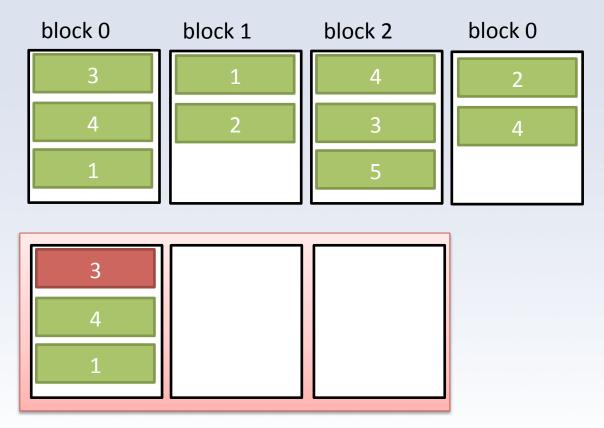
Two Pass Hashing for δ



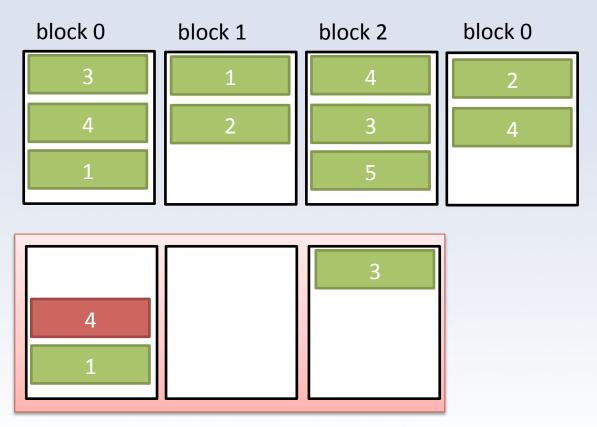




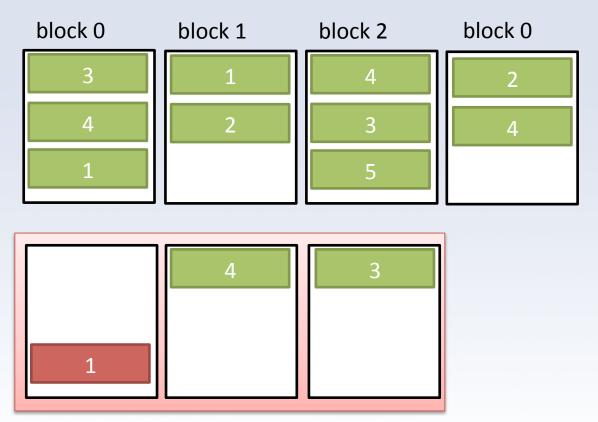




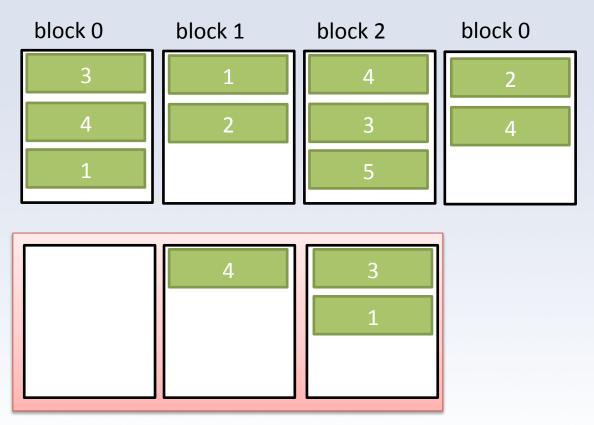




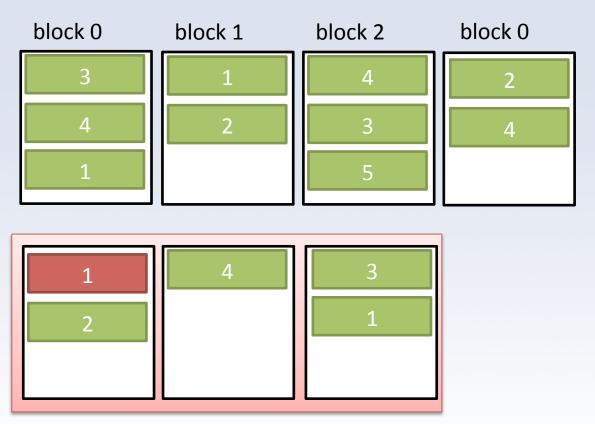




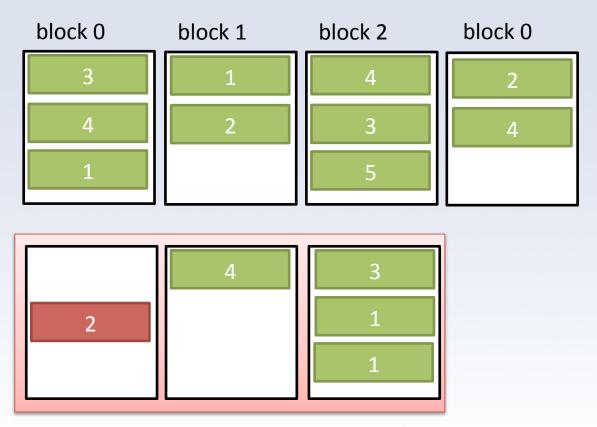




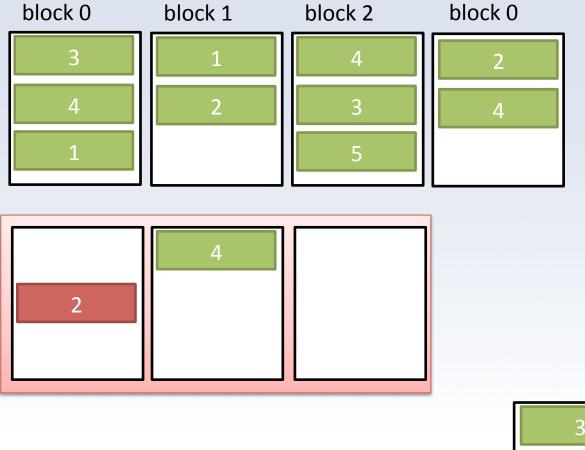




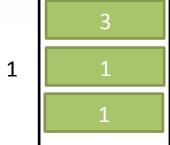


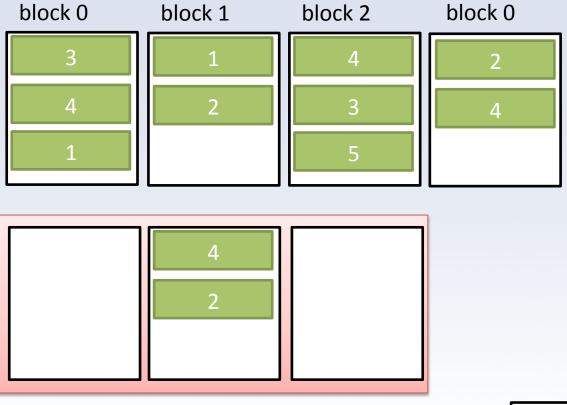






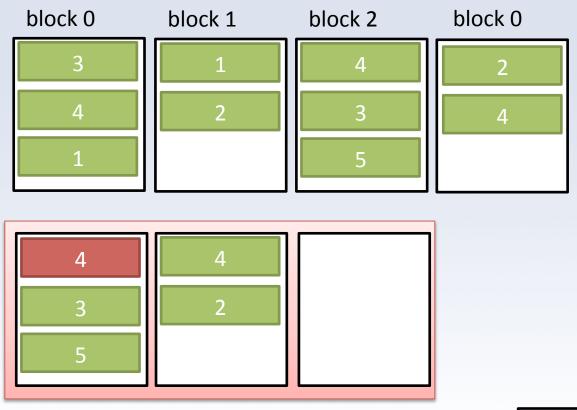






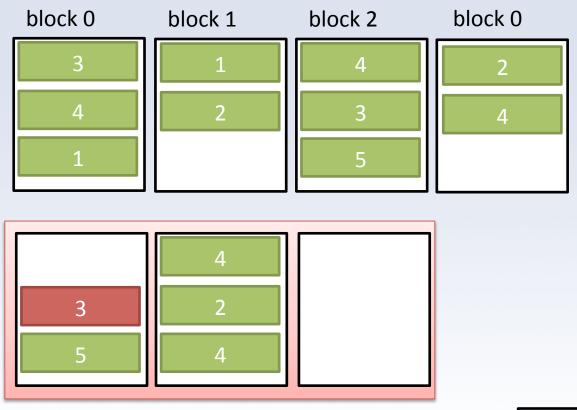






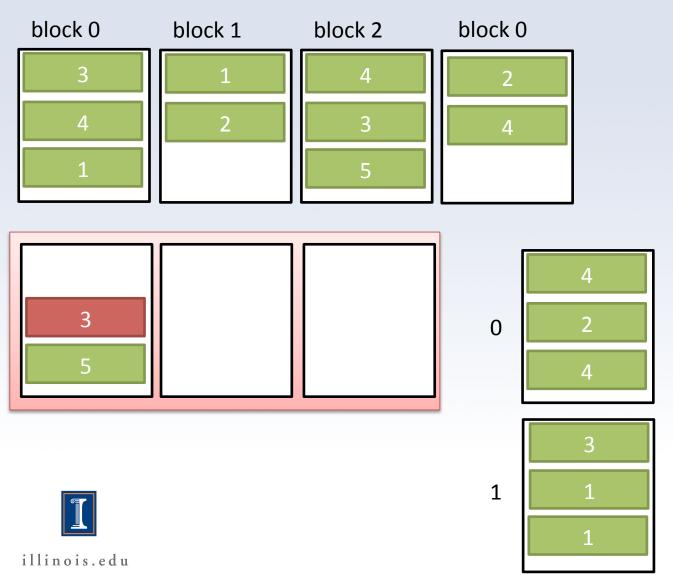


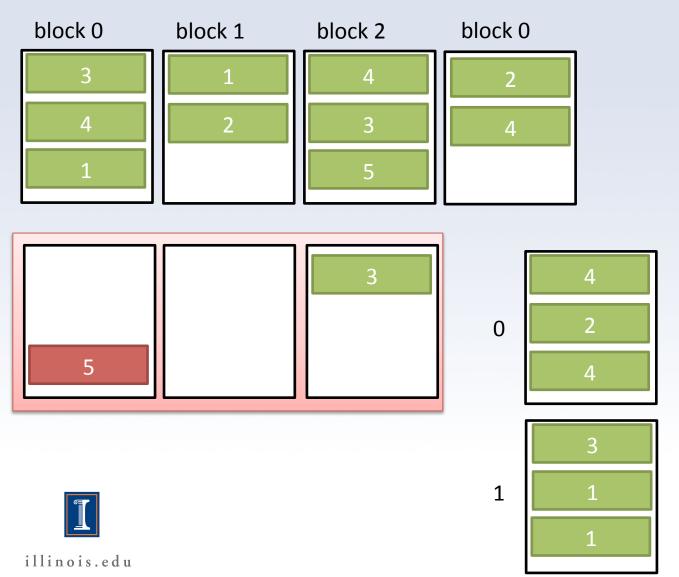


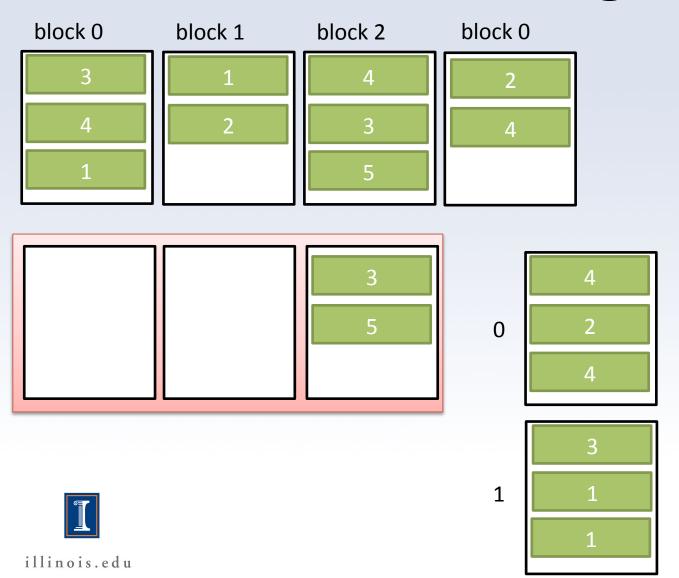


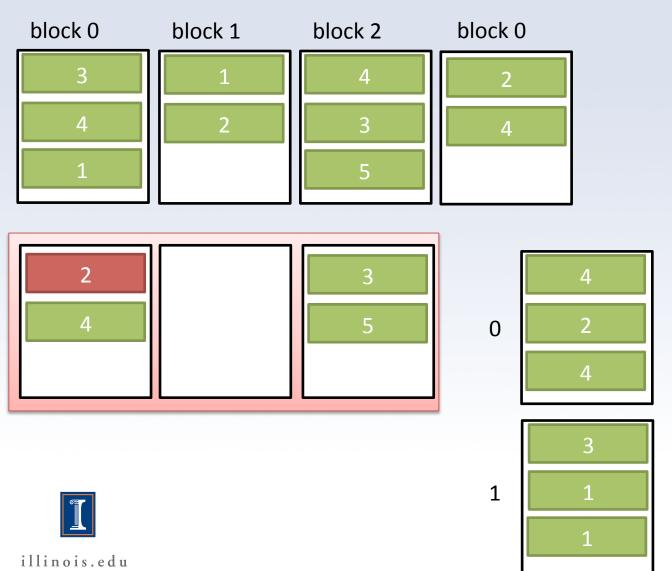


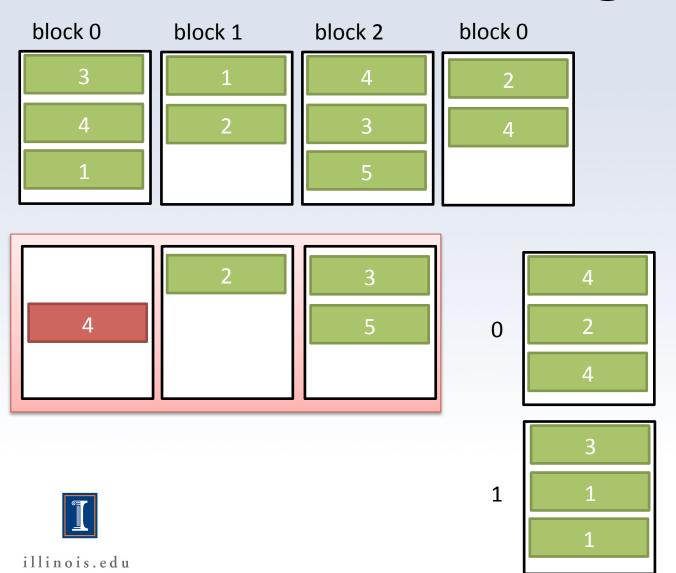


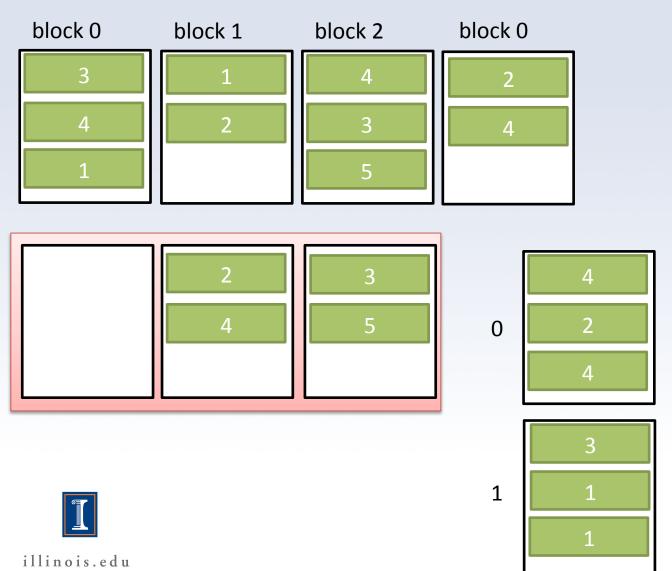


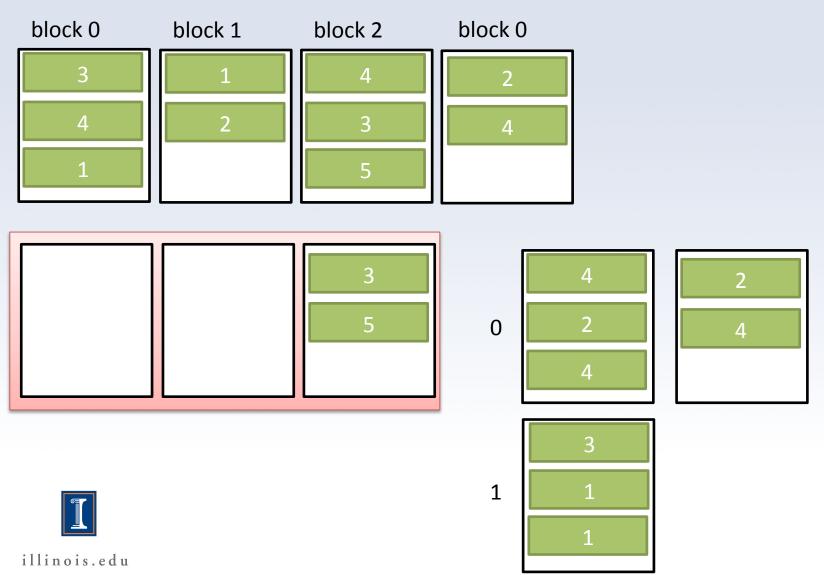


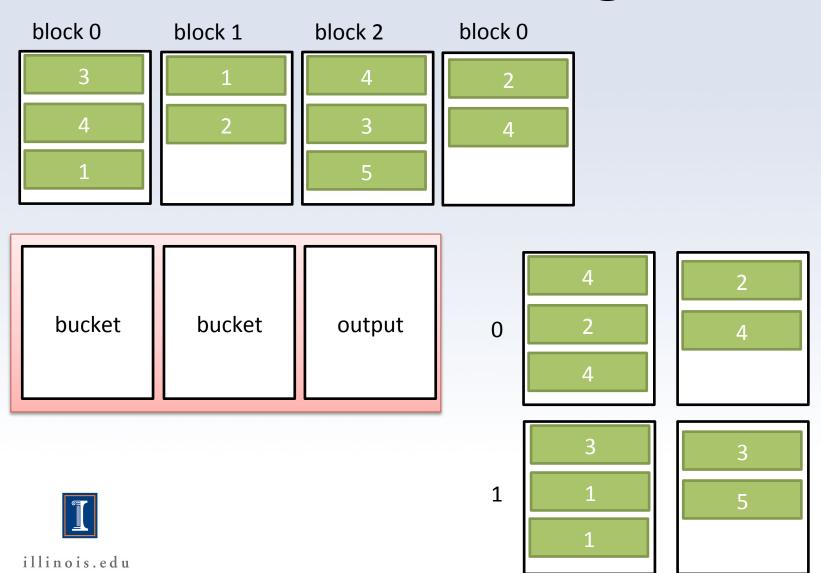


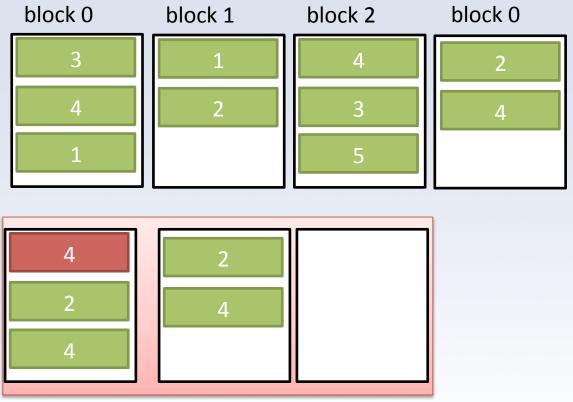




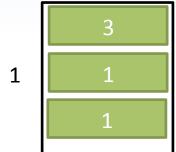


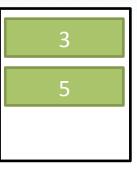


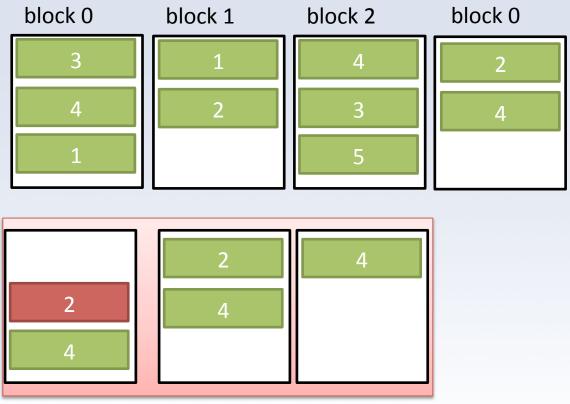




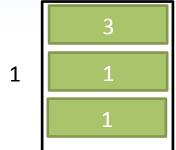


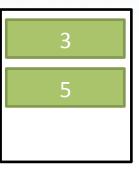


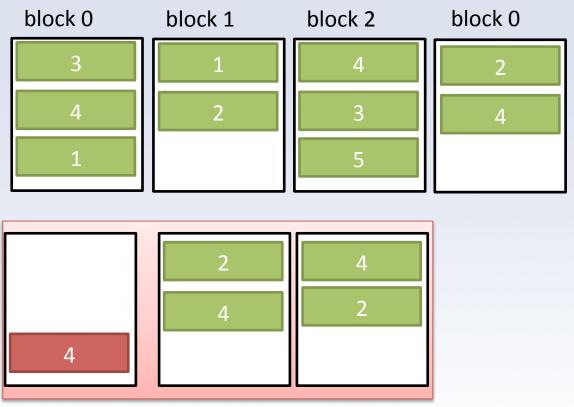






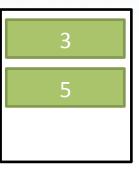


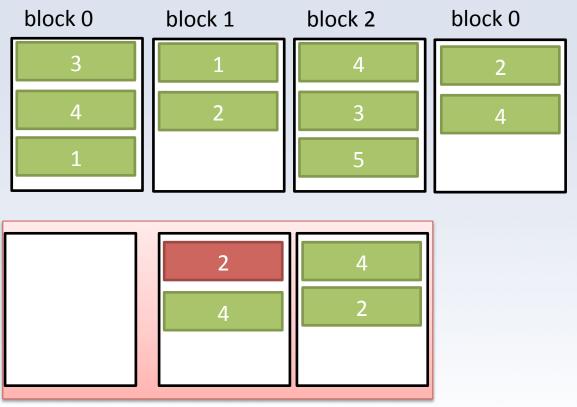






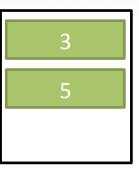


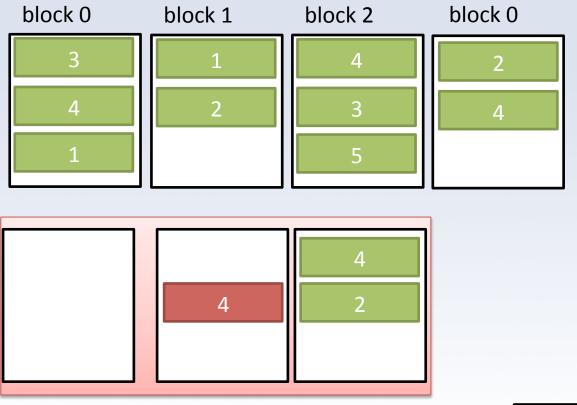






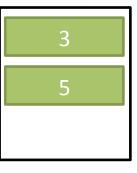


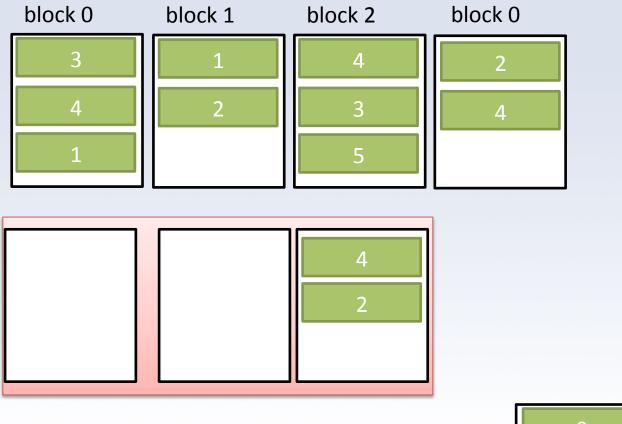






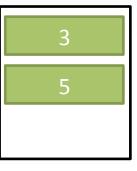


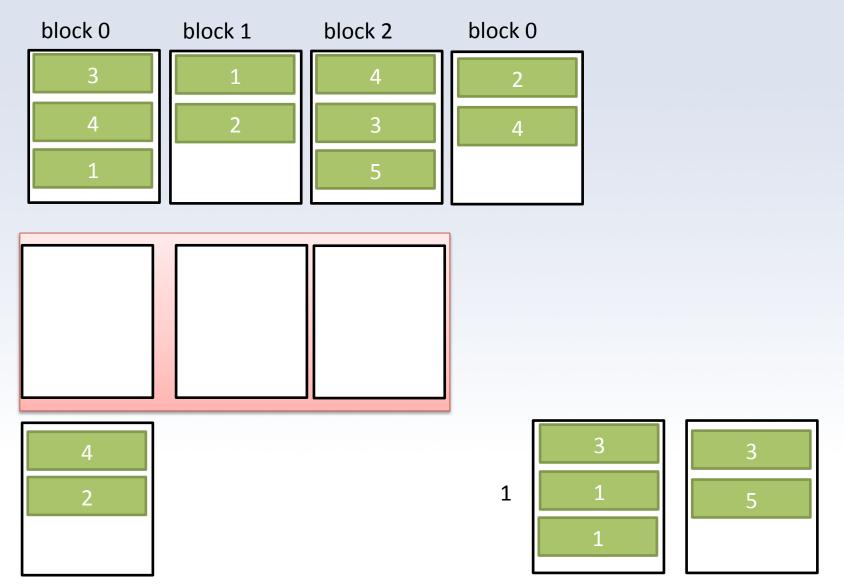


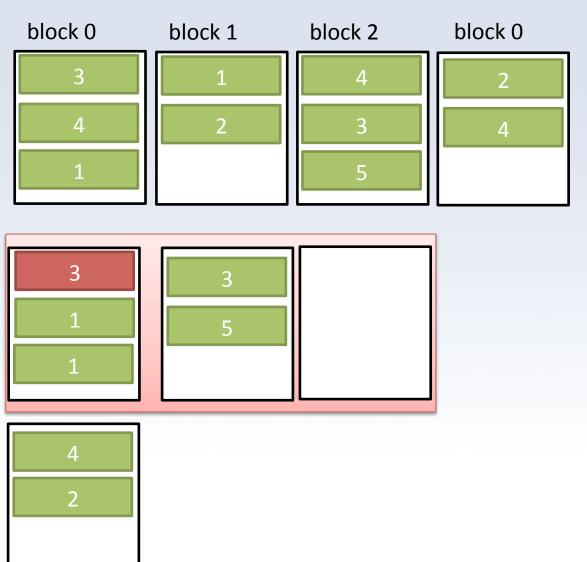


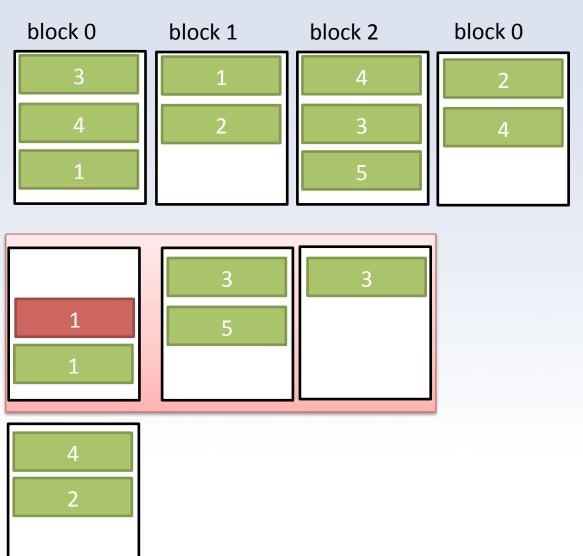


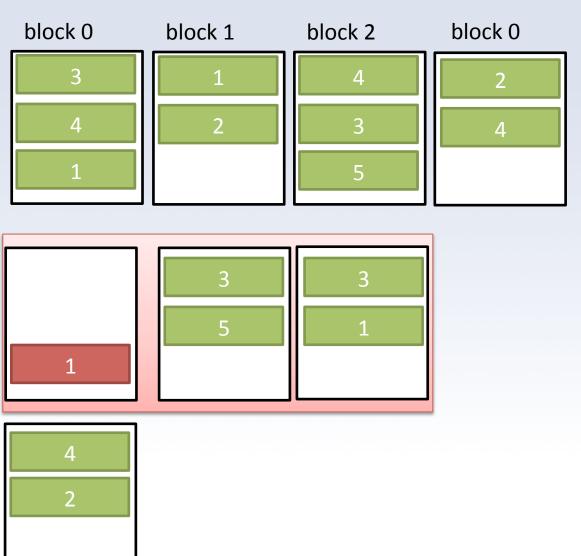


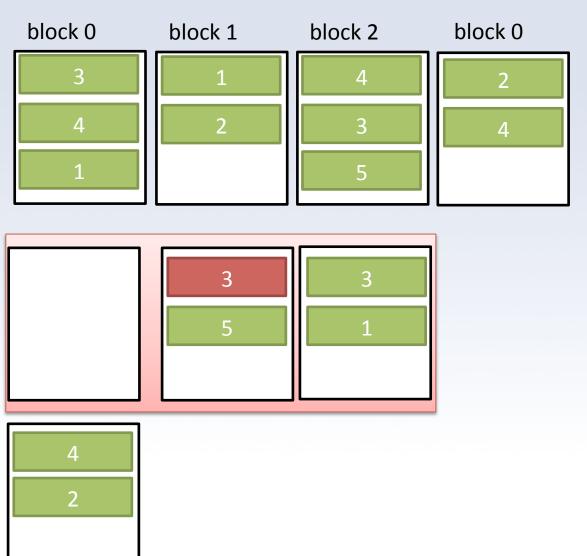


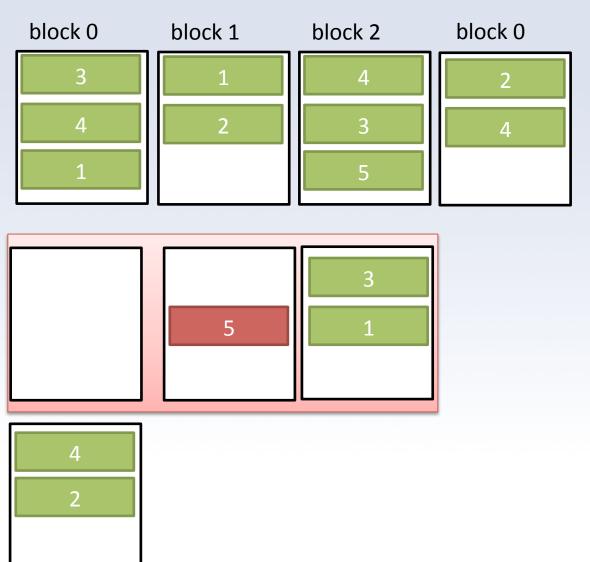


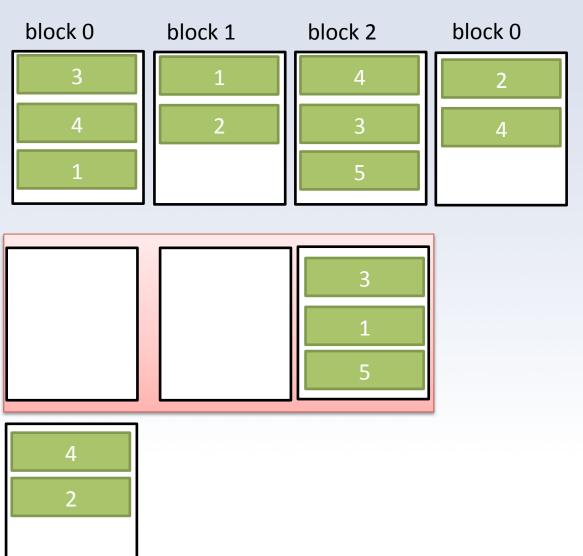


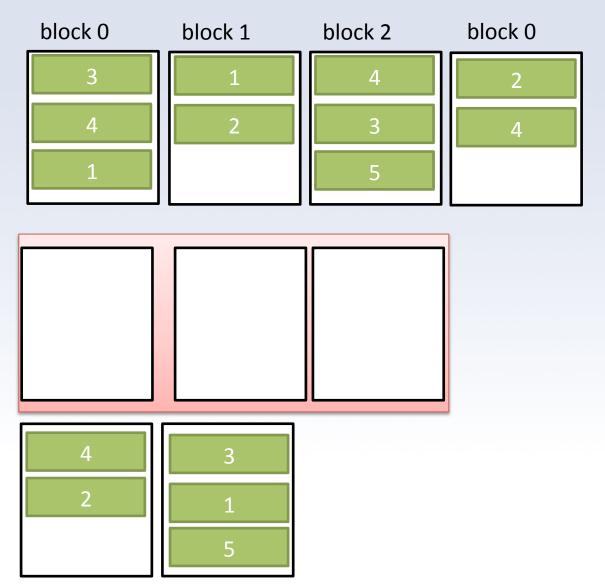












Analysis

- Requires that each bucket fits in memory
 - $-\mathrm{B(R)/(M-1)} \leq \mathrm{M}$
 - Approximately $B(R) \leq M^2$
- Hashing all the data takes B(R) reads and B(R) writes
- Reading in the buckets takes B(R) reads
- Total cost: 3B(R)



- We can do γ in a similar way
 - Hash each record into buckets using grouping attributes
 - Bring bucket into memory
 - Compute aggregation functions on each bucket



Two Pass Hashing for \cup , \cap , -

- Hash both relations using same hash function
- Perform one pass algorithms on each bucket
- Requires 3(B(R)+B(S)) I/O operations
- Requires bucket of smaller table to fit in memory: min(B(R),B(S))<M²



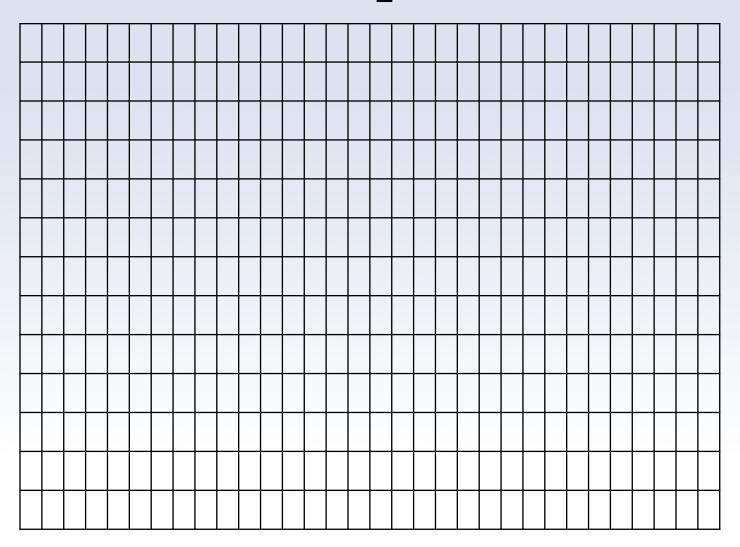
- Hash both relations *on join attributes* using same hash function
- Perform one pass join algorithm on each bucket
- Requires 3(B(R)+B(S)) I/O operations
- Requires bucket of smaller table to fit in memory: min(B(R),B(S))<M²



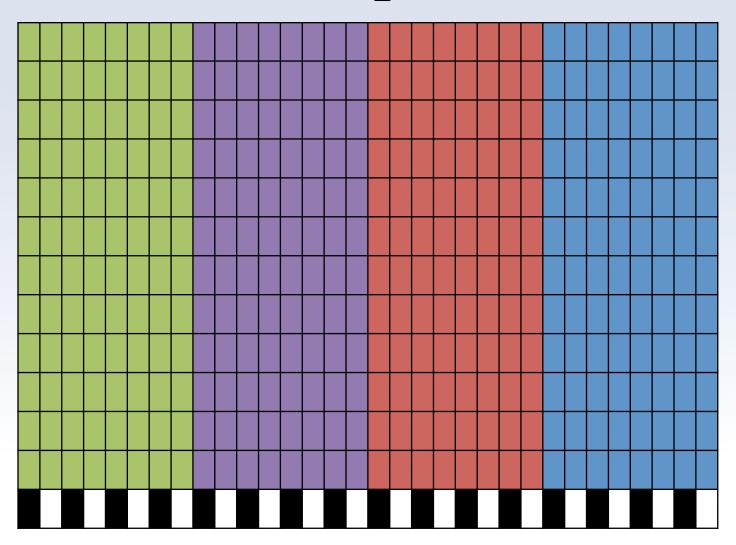
Hybrid Hash Join

- We can do better:
 - keep m of the k buckets in memory
 - never write these buckets to the disk
 - make these buckets larger than one buffer
 - perform joins immediately
 - keep one buffer for each of the remaining k-m buckets, use normal hash join for them











Hybrid Hash Join

- Requires that $mB(S)/k+k-m \le M$
 - B(S)/k expected size of "big" buckets
 - there are m "big" buckets in memory
 - k-m remaining buckets we write to disk
- Making the "big" buffers as big as possible, k≈B(S)/M
 - Requires B(S)≤M



Hybrid Hash Join

- Compared to hash join, saves 2 I/O operations per block of S in memory
- m/k buckets remain in memory
- Saves 2(m/k) reads and writes
- If $k \approx B(S)/M$, we save 2M/B(S) I/Os
- Cost: (3-(2M/B(S)))(B(R)+B(S))



2 Pass Hash-Based Summary

Operator	M required	I/O Cost
δ, γ	٧B	3B
U,∩,-	V(B(S))	3(B(R)+B(S))
\bowtie	√(B(S))	3(B(R)+B(S))
\bowtie	√(B(S))	(3-2M/B(S))x (B(R)+B(S))



Practical Example

- On a system with 8GB=2³³ of RAM
 - we can sort/join/deduplicate 2⁶⁶
 - Bigger than an exabyte of data!
- On a system with 256GB=2³⁸
 - we can sort/join/deduplicate 2⁷⁶ bytes of data
 - almost a yottabyte!



Multipass Algorithms

- Two pass algorithms can be extended into multipass algorithms through recursion
- Can be used to sort *extremely* large datasets
- In reality, we'd need distributed systems to even *store* this much data!
- Map Reduce framework/Hadoop

Index-Based Algorithms

- With an index, we can perform some operations faster
- Need to talk about two different kinds of indexes
 - clustering index search keys all appear together in blocks in the index file
 - unclustered index search keys are scattered through the index file

Clustered Index

block x		block x+1	block x+2
	kkkkkkkkk	kkkkkkkkkkkkkkkkkkkkkkkk	kkkkkkkk



Index Based $\sigma_{a=v}(R)$

- Read the index to obtain pointers to records satisfying a=v
- If the index is clustered, only need to read B(R)/V(R,a)
 - proportion of blocks in R with a a=v
- If index is unclustered, read T(R)/V(R,a)
 - values are scattered through index
 - might have to read a block for every tuple!



Index Based R\sim S

- Assume S is indexed on join attribute
- Table scan R and use index to look up corresponding join attributes in S
- Cost:
 - B(R) blocks for table scan
 - B(S)/V(S,Y) for each lookup (clustered)
 - T(S)/V(S,Y) for each lookup (unclustered)



Index Based R\sim S

- Assume S is indexed on join attribute
- Table scan R and use index to look up corresponding join attributes in S
- Cost:
 - -B(R)B(S)/V(S,Y) (clustered)
 - -B(R)T(S)/V(S,Y) (unclustered)



Zig-zag Join

- Requires a sorted index (e.g. B-tree)
- Start at smallest key in each sorted index
- If keys are equal, start joining
- Otherwise, skip in index with smaller key
- Cost: B(R)+B(S)





S: 2 2 4 4 6 7







S: 2 2 4 4 6 7



START JOINING!



START JOINING!



R: 1 3 4 4 4 5 6



S: 2 2 4 4 6 7



R: 1 3 4 4 4 5 6



S: 2 2 4 4 6 7

START JOINING!



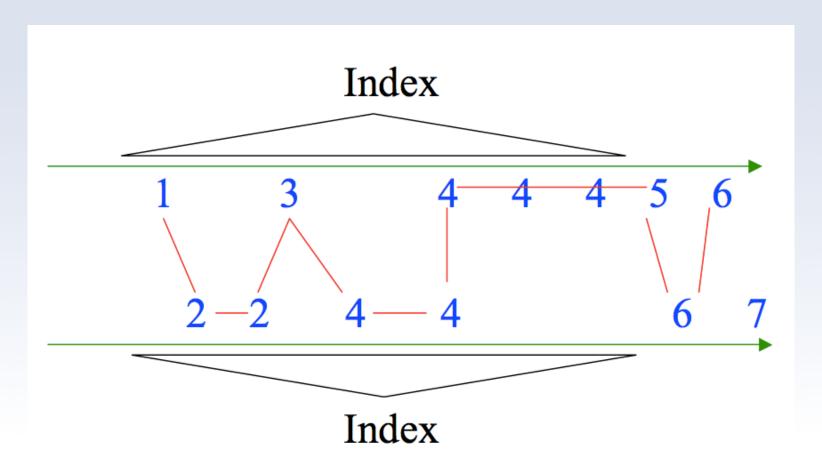
R: 1 3 4 4 4 5 6



S: 2 2 4 4 6 7



Why "zig-zag"?





Zig-zag Join

- With B-tree, might speed this up
 - If joinable entries are sparse, don't scan through both sorted tables
 - Instead, look up bigger value in the other Btree
 - Skips reading most values that don't join!



S: 2 2 4 4 6 7



R: 1 3 4 4 4 5 6



S: 2 2 4 4 6 7





Next week

- We'll start putting all this together:
 - Compile SQL queries into operators
 - Estimate costs of query plans
 - Figure out an optimal query plan

