

# Vectorized Bloom Filters for Advanced SIMD Processors

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### Bloom filters

#### Introduction

- Original version [Bloom 1970]
  - Represents a "set of items"
  - Answers: "Does item X belong to the set?"
- Supports 2 operations
  - Insert an item in the set
  - Check if an item exists in the set
- Probabilistic data structure
  - Allows false positives

### Bloom filters

#### Description

- The data structure
  - A bitmap (an array of bits) of m bits
  - \* A number of hash functions
- Insert an item in the set
  - \* Compute hash functions h(x,m), g(x,m), ...
  - \* Set bits h(x,m), g(x,m), ...
- Search an item in the set
  - \* Test bits h(x,m), g(x,m), ...

### Bloom filters

#### Errors

- False negatives are not possible
  - If item x in set: h(x,m), g(x,m), ... are all set
- False positives are possible
  - h(x,m), g(x,m), ... may be set by other items
  - ❖ 1 bit not set: 1-1/m
  - \* k bits not set: (1-1/m) ^ k
  - \* k bits not set with n items in the filter: (1-1/m) ^ kn
  - \* 1 target bit is set: 1 (1-1/m) ^ kn
  - \* k target bits are set: [1 (1-1/m) ^ kn] ^ k

### Bloom filters in Databases

#### Semi-Joins

- Evaluate selections
  - Select tuples from table R if R.y > 5
  - Select tuples from table S if S.y < 3</li>
- Truncate join inputs using Bloom filters
  - Discard R tuples if R.x not in the S.x set
  - Discard S tuples if S.x not in the R.x set
- Join remaining tuples
  - Filter tuples that the Bloom filters missed

#### The query:

```
select *
from R, S
where R.x = S.x
and R.y > 5
and S.y < 3
```

### Bloom filters in Databases

#### In parallel/distributed databases

- Filter data to reduce network traffic
  - Network << RAM</li>
  - Probing the Bloom filter > send over the network
  - Broadcast the filters —> small cost

#### In main-memory database execution

- Filter data as early as possible to reduce the working set
  - Filter before partitioning
  - If after: Bloom filter probing > hash table probing
  - Bloom filter fits in the cache often

#### Scalar implementation

- Iterate over the hash functions / bit-tests
  - 1 access & bit-test / time
  - 1 hash function / time
- Good performance —> short-circuit
  - Bit-test fail —> stop inner loop
  - Most keys fail early
- Bad performance —> short-circuit
  - Branching logic —> branch mis-predictions & pipeline bubbles

#### Scalar implementation

```
for (o = i = 0 ; i != tuples ; ++i) {
 key = keys[i];
                                         // read the key
 for (f = 0; f != functions; ++f) { // iterate over functions
   h = hash[f](key);
                                         // compute the hash function
   if (bit test(bitmap, h) == 0)
                                         // perform bit-test (x86 instruction)
     goto failure;
                                         // early abort if bit-test fails
                                         // copy the payload to output
 rids out[o] = rids[i];
 keys out[o++] = key;
                                         // write the key to output
failure:;
                                         // jump here if not qualified
```

- Use multiplicative hashing
  - 1 multiplication
  - Universal family
  - Pair-wise independent functions easy

#### Scalar implementation

- How much can be done?
  - Unroll hash functions
  - Separate branches (prediction states) per function
  - Better branch prediction (hopefully)

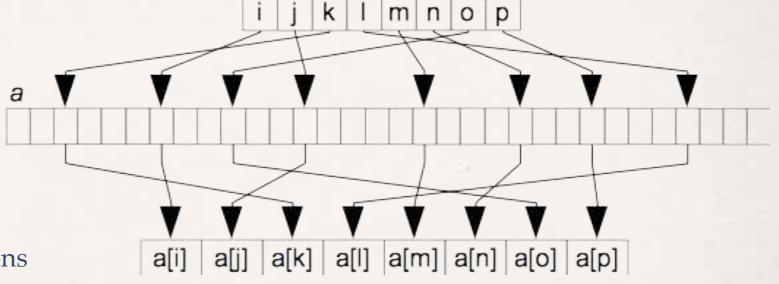
### SIMD in Databases

#### SIMD on query execution

- General usage
  - Scan, aggregation, index search [Zhou et.al. 2002]
- For sorting / compressing
  - Comb-sort [Inoue et al. 2007]
  - Merge-sort using bitonic merging [Chhugani et al. 2008]
  - \* Range partitioning [Polychroniou et al. 2014]
  - Dictionary (de-)compression [Willhalm et al. 2009]
- For indexing
  - Tree index search [Kim et al. 2010]
  - Hash table probing using multi-key buckets [Ross 2006]

#### SIMD loads

- Sequential
  - 128/256/512 sequential bits
  - Align —> better performance
  - Mask reads
- Fragmented
  - \* 32/64 bits from multiple locations
  - Indexes in another SIMD register
  - Loaded values packed in SIMD
  - Since Intel Haswell (2009)



#### SIMD without gathers

- Scalar accesses
  - 256-bit load = 32-bit load
  - Pack in less space
  - Tree node accesses [Kim et.al. 2009]
  - Multi-key hash buckets [Ross 2006]
- Fragmented accesses
  - Extract index from SIMD to scalar
  - Load each item individually
  - Pack values in SIMD

```
// extract indexes
i1 = mm256 cvtsi128 si64(index);
i2 = mm256 cvtsi128 si64(
      mm256 permute4x64 epi64(index, 1));
i3 = _mm256_cvtsi128_si64(
      _mm256_permute4x64_epi64(index, 2));
i4 = mm256 cvtsi128 si64(
      _mm256_permute4x64_epi64(index, 3));
// load values one at a time
v1 = mm load epi64(&data[i1]);
v2 = mm load epi64(\&data[i2]);
v3 = mm load epi64(\&data[i3]);
v4 = mm load epi64(\&data[i4]);
// pack values
v12 = mm256 unpacklo epi64(v1, v2);
v34 = mm256 unpacklo epi64(v3, v4);
value = mm256 permute2x128 si256(v12,
                             v34, 64);
```

#### Using SIMD for Bloom filters

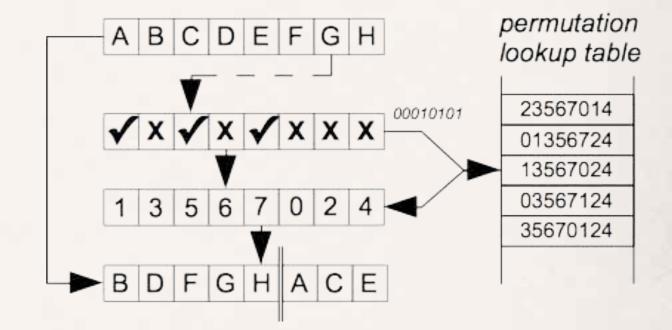
- Vectorizing hashing / access / bit-test
  - Multiplicative hash in SIMD
  - 32-bit gather to access the bitmap on hash div 32
  - Mask with 1 bit shifted using hash mod 32
- "How" to vectorize >1 functions?
  - ★ k=1 —> similar to selection scan
  - Maintain short-circuit
  - Avoid branching
  - Minimize loads/stores

```
// multiplicative hashing
hash = _mm256_mullo_epi32(key, factor);
hash = _mm256_srli_epi32(hash, shift);

// bit-test
index = _mm256_srli_epi32(hash, 5);
bit = _mm256_and_si256(hash, mask_31);
data = _mm256_i32gather_epi32(bitmap, index, 4);
bit = _mm256_sllv_epi32(mask_1, bit);
data = _mm256_and_epi32(data, bit);
aborts = _mm256_cmpeq_epi32(data, mask_0);
```

#### SIMD 2-way partitioning

- Using SIMD permutations
  - Register to register "gather"
  - "Pull"-based shuffling
- Using boolean result bitmap as an index
  - Get boolean results —> extract bitmap
  - Load permutation mask
  - Permute vector to "true" and "false"
  - ♦ W SIMD lanes = 2^W permutation mask



```
// load 8-way permutation mask
bitmap = _mm256_movemask_ps(aborts);
mask = _mm_load_epi64(&perm_table[bitmap]);
mask = _mm256_cvtepi8_epi32(mask);

// permute keys & rids
key = _mm256_permutevar8x32_epi32(key, mask);
rid = _mm256_permutevar8x32_epi32(rid, mask);
```

Best stored in W \* 2^W bytes —> L1 for 8-way SIMD

#### Conditional control flow transformation

- Maintain short-circuit logic
  - Never do multiple bit-tests for the same key
  - First bit-test fails —> second bit-test wasted
  - Process a different input key per lane
- Arbitrary hash function per lane
  - Maintain function indexes (per lane)
  - Any hash function (per lane)
  - Function index = k —> tuple qualifies!

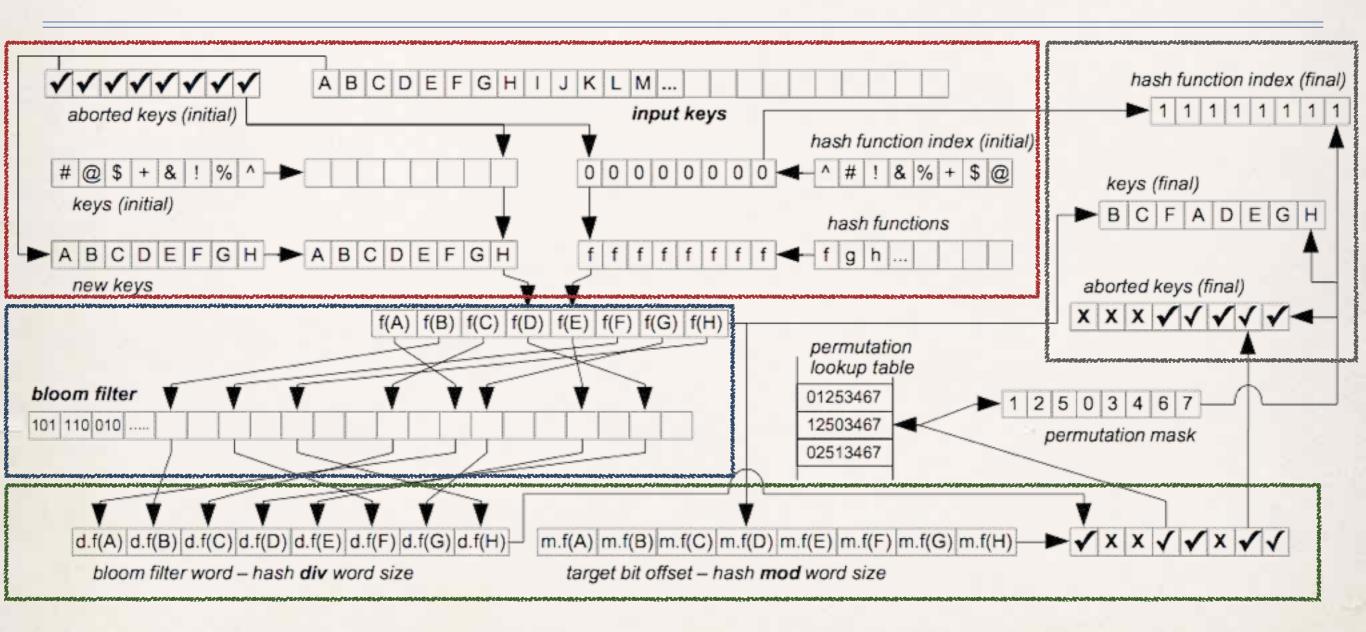
"Gather" hash functions from register (not L1)

#### Conditional control flow transformation

- Dynamic input reading
  - Recycle lanes that failed a bit-test
  - Permute SIMD vector in two parts
  - Refill aborted part of the vector
  - Advance input pointer
  - Word-aligned access
- Dynamic output writing
  - SIMD permute —> write qualifiers
  - Advance output pointer

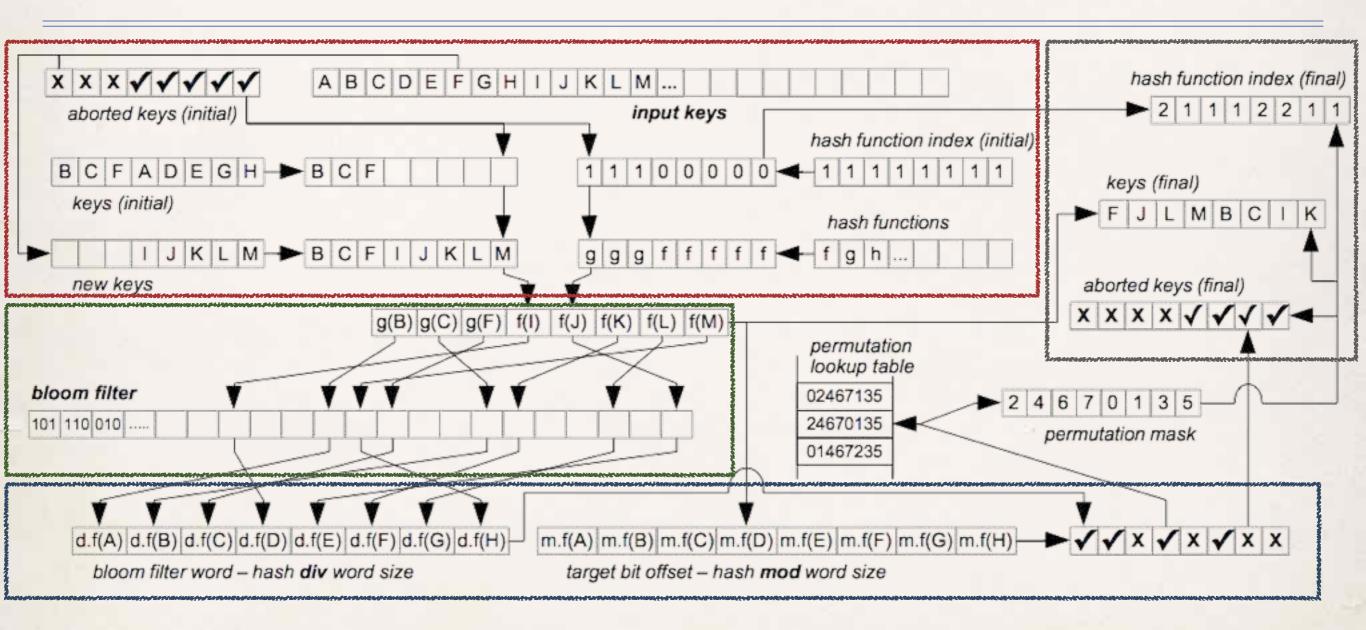
```
// read new keys & payloads
new key = mm256 maskload epi32(keys, aborts);
new val = _mm256_maskload_epi32(vals, aborts);
// clear aborted data
key = mm256 andnot si256(aborts, key);
rid = mm256 andnot si256(aborts, rid);
fun = mm256 andnot si256(aborts, fun);
// mix old with new items
key = mm256 \text{ or } si256(key, new key);
rid = mm256 \text{ or } si256(rid, new rid);
// perform bit-tests and permute data
bitmap = [...]
// advance input pointers by counting bits
keys += mm popcnt u64(bitmap);
rids += mm_popcnt_u64(bitmap);
```

## Example



- First loop
  - 32-bit keys, no payloads, no output code
- 1) Input & hashing 2) Bitmap access
- 3) Bit-testing
  4) Permutations

## Example



- Second loop
  - 32-bit keys, no payloads, no output code
- 1) Input & hashing 2) Bitmap access
- 3) Bit-testing
  4) Permutations

#### Writing the output

- Use branching
  - Low selectivity —> rarely taken
  - Skipped otherwise
- Filter data
  - SIMD permute
  - Store sequentially
  - Qualifiers "aborted"
  - Advance output pointers
  - Same as selection filtering

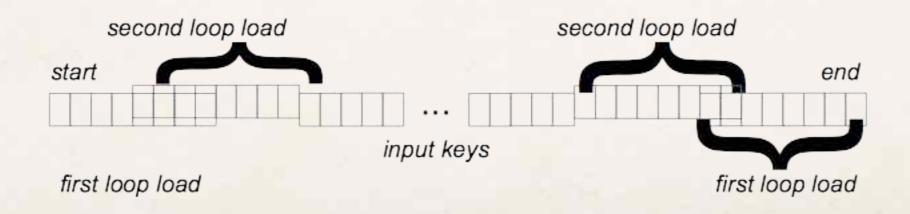
```
// any qualifiers ?
done = mm256 cmpeq epi32(fun, functions);
done = mm256 andnot si256(aborts, done);
if (! mm256 testz si256(done, done)) {
  // load permutation mask
  bitmap = mm256 movemask ps(done);
 mask = mm256 loadl epi64(&perm table[bitmap]);
 mask = mm256 cvtepi8 epi32(mask);
  // permute data and mask
  key out = mm256 permutevar8x32_epi32(key, mask);
  rid out = mm256 permutevar8x32 epi32(key, mask);
  done = mm256 permutevar8x32 epi32(done, mask);
  // write qualifiers to output
  mm256 maskstore epi32(keys out, done);
  _mm256_maskstore_epi32(rids_out, done);
  // update output pointer by counting bits
  keys out += mm popcnt u64(done);
  rids_out += _mm_popcnt_u64(done);
```

#### Loop unrolling

- In general
  - Interleave instructions to increase IPC
  - Crucial for in-order CPUs
  - (Should be) irrelevant in out-of-order CPUs
  - Can still improve performance
  - Limited by number of registers to hold "state"
- For Bloom filters
  - Dynamic input reading —> naive loop unrolling does not work
  - Read the input from non-overlapping locations

#### Loop unrolling

- In Bloom filter probing
  - Read the input from non-overlapping locations
  - Simplest way: low —> high & high —> low
  - Allows for 2-way loop unrolling
  - Stop when the two pointers meet



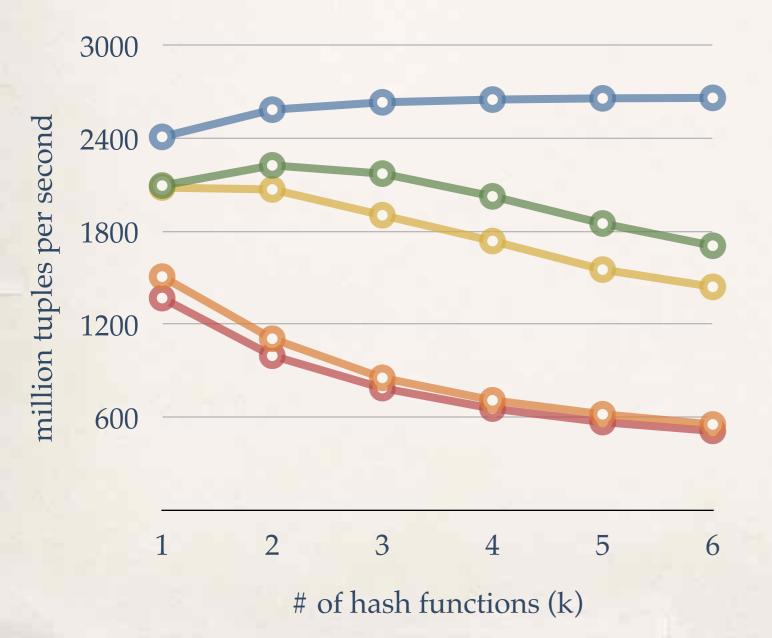
#### Experimental setup

- Hardware platform & software setting
  - \* 1 Intel Xeon E3-1675 v3 CPU @ 3.5 GHz with 4-cores & 2-way SMT
  - 32 GB of DDR3 RAM @ 1600 MHz
  - Running 8 threads & shared Bloom filter
  - Using 32-bit keys & 32-bit payloads

#### Figures

- Scalar soft: standard scalar implementation
- Scalar hard: scalar implementation with unrolled branches
- SIMD single: standard SIMD implementation
- SIMD double: SIMD implementation with unrolled loop

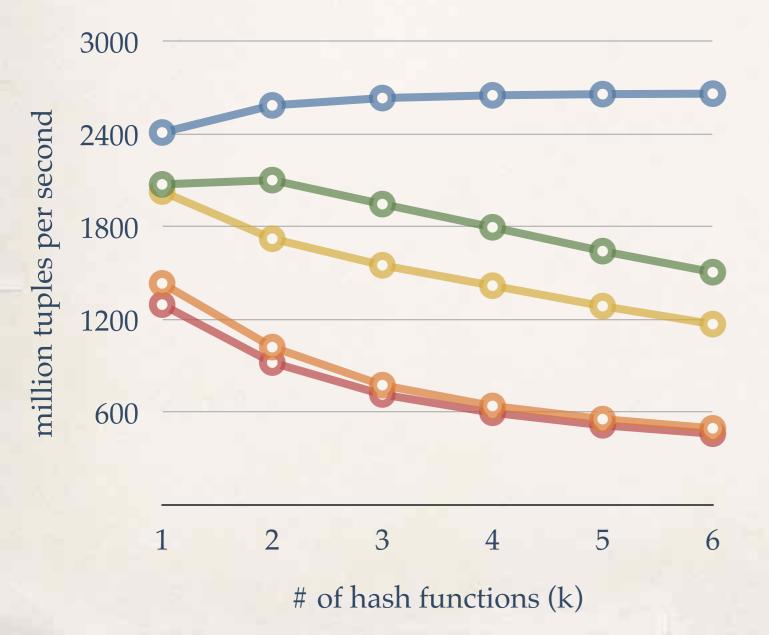
#### \* L1 cache resident Bloom filter



- Bandwidth
- SIMD (double)
- SIMD (single)
- Scalar(soft)
- Scalar (hard)

- 16 KB Bloom filter
- 10 bits / item
- 5% qualify
- more than 3X improvement!

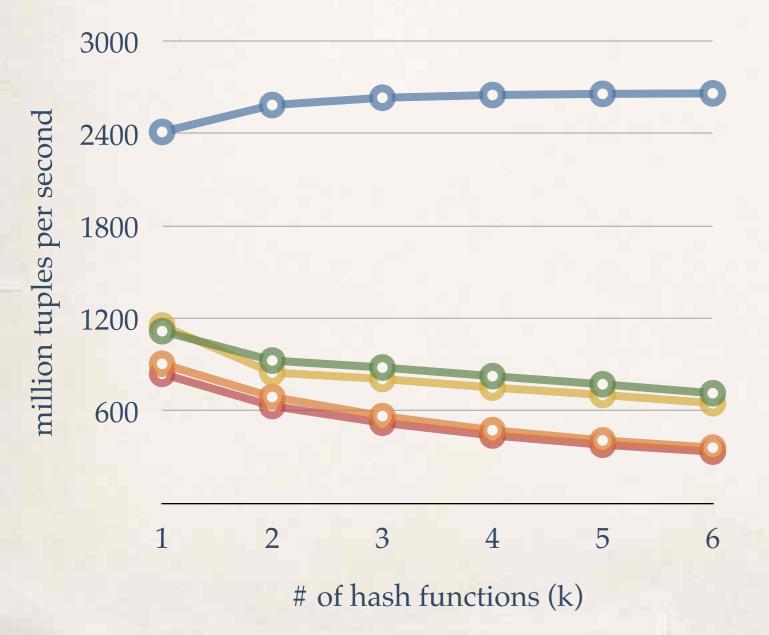
#### L2 cache resident Bloom filter



- Bandwidth
- SIMD (double)
- SIMD (single)
- Scalar(soft)
- Scalar (hard)

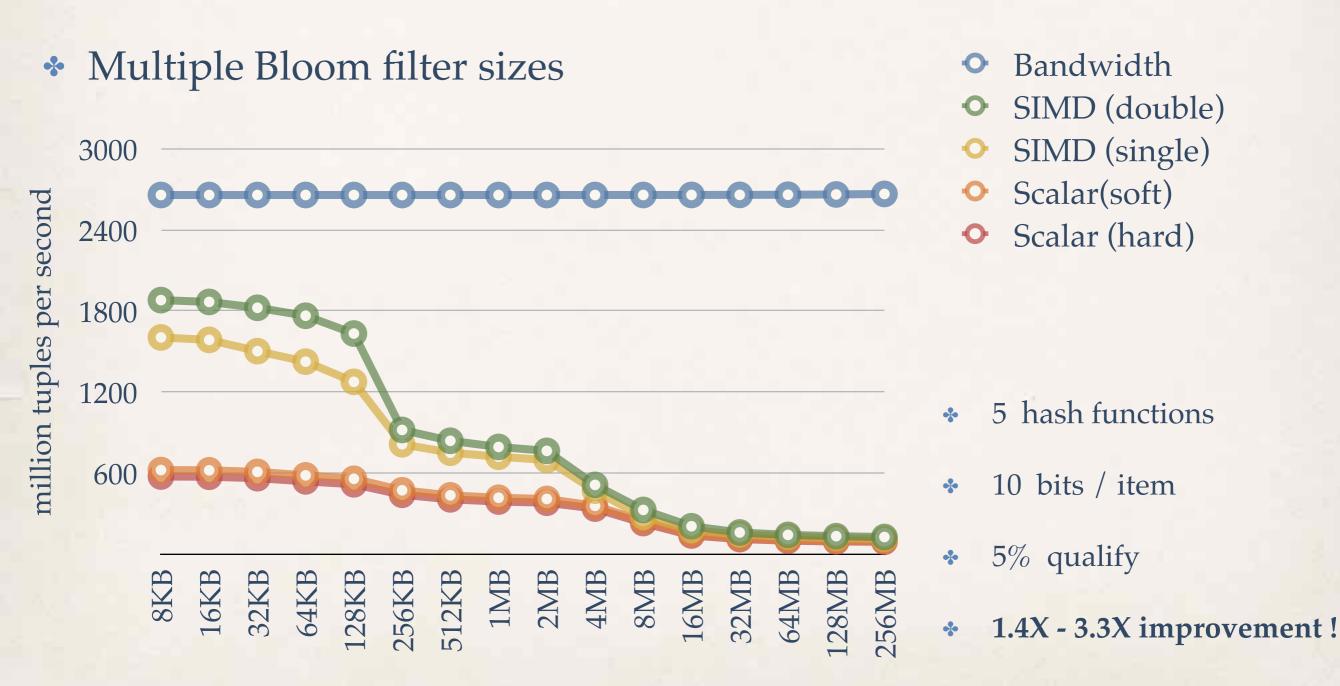
- 128 KB Bloom filter
- 10 bits / item
- 5% qualify
- more than 3X improvement!

#### L3 cache resident Bloom filter



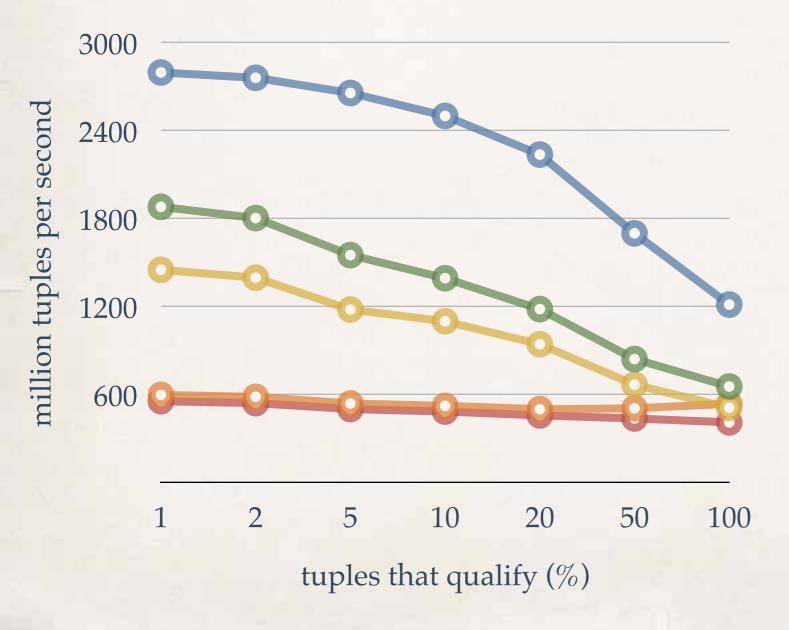
- Bandwidth
- SIMD (double)
- SIMD (single)
- Scalar(soft)
- Scalar (hard)

- 2 MB Bloom filter
- 10 bits / item
- 5% qualify
- more than 2X improvement!



Bloom filter size

#### Multiple selectivities



- Bandwidth
- SIMD (double)
- SIMD (single)
- Scalar(soft)
- Scalar (hard)
- 128 KB Bloom filter (L2)
- 5 hash functions
- 10 bits / item
- still faster on 100% selectivity

### Conclusions

#### Vectorized Bloom filters

- Implementation
  - Access data non-sequentially in SIMD
  - \* Eliminate conditional control flow
  - Maintain short-circuit
  - Non-trivial loop unrolling
  - \* Re-usable techniques for other structures
- Performance
  - More than 3X faster when cache-resident
  - Still faster when operating off the cache or all tuples qualify

# Questions

