

The 39th International Symposium on Lattice Field Theory (Lattice 2022)

Performance Optimization of Baryon-block Construction in the Stochastic LapH Method

Phuong Nguyen* (TU Munich/Intel)

Ben Hoerz (Intel)



Overview

1

Motivation

2

Baryon-block Construction

3

Reference Implementation

4

Single-core Optimization

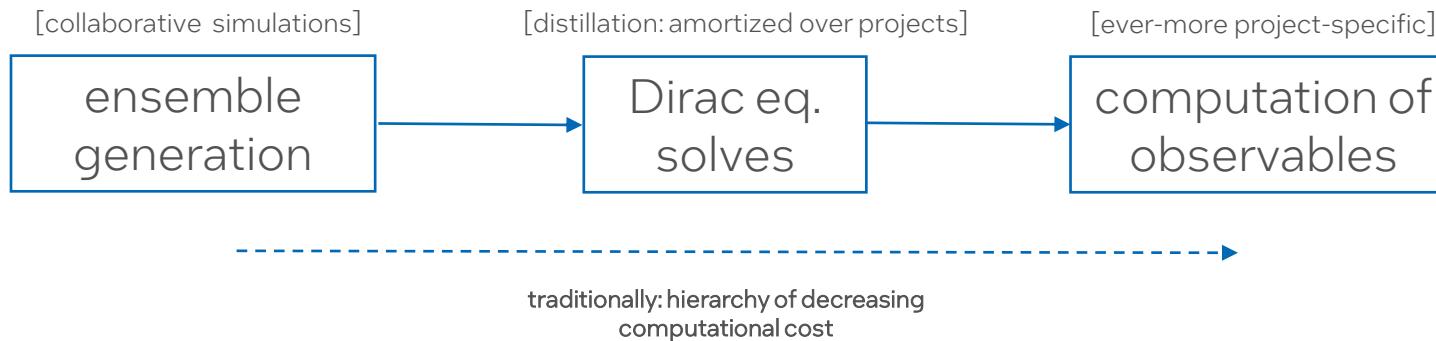
5

Node-level Optimization

6

Conclusion

1. Motivation



- Traditionally, comparatively moderate cost to compute observables
- Balance can shift for modern multi-hadron spectroscopy
 - e.g. Baryon systems with the stochastic LapH method
- **Challenge:** scaling to state-of-the-art ensembles
 - Design with CLS E250 in mind: $96^3 \times 192$, $a = 0.064$ fm
 - First multi-hadron results in mesonic sector
 - Projected cost of baryon blocks exceeds Dirac solves severalfold

[Morningstar et al. 1104.3870]

[talk by W. Soeldner Tue 15:40]

[talk by S. Paul Fri 15:10]

2. Baryon-block Construction

$$\mathcal{B}_{\mathbf{p}}^{d_1 d_2 d_3} = \sum_{\mathbf{x}} \sum_{a,b,c} e^{-i\mathbf{p}\mathbf{x}} \varepsilon_{abc} q_{\mathbf{x}a}^{d_1} q_{\mathbf{x}b}^{d_2} q_{\mathbf{x}c}^{d_3}$$

- q : set of (LatticeColorVector-valued) LapH quark fields.
- $d_1, d_2, d_3 = 1, \dots, n_{\text{Dil}}$
- Kernel called many times with same momentum set, different q (spin, noise combinations)
 - $\# \text{momenta} \ll V \Rightarrow$ precompute momentum phases for SFT
- Seek to exploit high arithmetic intensity
 - $O(n_{\text{Dil}}^3)$ compute with $O(n_{\text{Dil}})$ memory transfer.

3. Reference Implementation (1)

```
input: q1, q2, q3          // 3D array, [nD,nX,nColor]
       momBuf             // 2D array, [nMom, nX]
output: res                // 4D array, [nMom, nD1, nD2, nD3]
intermediate: diq          // 2D array, [nX, nColor]
           singlet         // 1D array, [nRows, nX]
for d1 = 1 to nD1; do      // dilution index 1
  for d2 = 1 to nD2; do    // dilution index 2
    for iX = 1 to nX; do
      calculate diq(iX,:) = q1(d1,iX,:)*q2(d2,iX,:)
    end for
  rowIndex = 0;
  for d3 = 1 to nD3; do   // dilution index 3
    for iX = 1 to nX; do
      calculate singlet(rowIndex, iX) = diq(iX,:)*q3(d3,iX,:)
    end for
    // blocked momentum projection
    rowIndex++;
    if rowIndex == nrows
      // With BLAS Level 3 - GEMM
      res(:, d1, d2, d3 - nrows : d3) = momBuf * singlet
      rowIndex = 0
    end if
  end for
end for
end for
```

$$\mathcal{B}_p^{d_1 d_2 d_3} = \sum_x e^{-ipx} \sum_{a,b,c} \varepsilon_{abc} q_{xa}^{d_1} q_{xb}^{d_2} q_{xc}^{d_3}$$

The diagram illustrates the components of the formula. It shows the summation over indices x and a, b, c . The term $q_{xa}^{d_1}$ is grouped under a red bracket labeled "momBuf". The term $q_{xb}^{d_2}$ is grouped under a red bracket labeled "singlet". The term $q_{xc}^{d_3}$ is grouped under a red bracket labeled "diq". The factor ε_{abc} is positioned between the "momBuf" and "singlet" groups.

- pre-compute **diq** = $q1 \times q2$ once for all $d3$
- pack multiple **singlet** vectors into a matrix, then using BLAS-3 to calculate **momBuf** * **singlet**.



Problems ???

3. Reference Implementation (2)

Testing system

Hardware Info:

- 2 x Intel(R) Xeon(R) Platinum 8358 CPU @ 2.60GHz.
- 64 cores in total.

Theoretical Peak Performance:

- Single core: $2.6 \text{ GHz} * 2 \text{ FMA} * 2 \text{ Flops/FMA} * 8 \text{ Flops/AVX512 DP} = 83.2 \text{ GFLOPs}$
- Node performance: $64 \times 83.2 \text{ GFLOPs} = 5324 \text{ GFLOPs}$

3. Reference Implementation (3)

Profiling for single core

- Only 33% of Peak Performance
- ~27% Non-FP

```
$ vtune -collect hpc-performance ./kernel
...
CPU
    DP GFLOPS: 27.490
    ...
    Vectorization: 99.9% of Packed FP Operations
    Instruction Mix
        DP FLOPs: 73.5% of uOps
        Packed: 99.9% from DP FP
            128-bit: 26.1% from DP FP
            256-bit: 0.0% from DP FP
            512-bit: 73.9% from DP FP
        Scalar: 0.1% from DP FP
    x87 FLOPs: 0.0% of uOps
    Non-FP: 26.5% of uOps
    FP Arith/Mem Rd Instr. Ratio: 2.342
    FP Arith/Mem Wr Instr. Ratio: 12.229
    ...
...
```

Low Arithmetic Intensity

```
$ vtune -collect memory-access ./kernel
...
CPU Time: 10.515s
Memory Bound: 37.2% of Pipeline Slots
L1 Bound: 1.1% of Clockticks
L2 Bound: 10.3% of Clockticks
L3 Bound: 3.5% of Clockticks
DRAM Bound: 15.0% of Clockticks
Elapsed Time
    Store Bound: 1.5% of Clockticks
Time
    Loads: 19,793,093,775
    Stores: 4,173,125,190
    LLC Miss Count: 97,540,950
    Local DRAM Access Count: 39,002,730
    Remote DRAM Access Count: 45,503,185
    Remote Cache Access Count: 0
    Average Latency (cycles): 41
    ...
...
```

- DRAM bound & large number of stalled cycles for LLC Misses

4. Single-core Optimization (1)

Reference Implementation

```
input: q1, q2, q3      // 3D array, [nD,nX,nColor]
       momBuf        // 2D array, [nMom, nX]
output: res            // 4D array, [nMom, nD1, nD2, nD3]
intermediate: diq     // 2D array, [nX, nColor]
                  singlet // 1D array, [nRows, nX]
for d1 = 1 to nD1; do // dilution index 1
  for d2 = 1 to nD2; do // dilution index 2
    for iX = 1 to nX; do
      calculate diq(iX,:) = q1(d1,iX,:) * q2(d2,iX,:)
    end for
  rowIndex = 0;
  for d3 = 1 to nD3; do // dilution index 3
    for iX = 1 to nX; do
      calculate singlet(rowIndex, iX) = diq(iX,:) * q3(d3,iX,:)
    end for
    // blocked momentum projection
    rowIndex++;
    if rowIndex == nrows
      // With BLAS Level 3 - GEMM
      res(:, d1, d2, d3 - nrows : d3) = momBuf * singlet
      rowIndex = 0
    end if
  end for
end for
end for
```



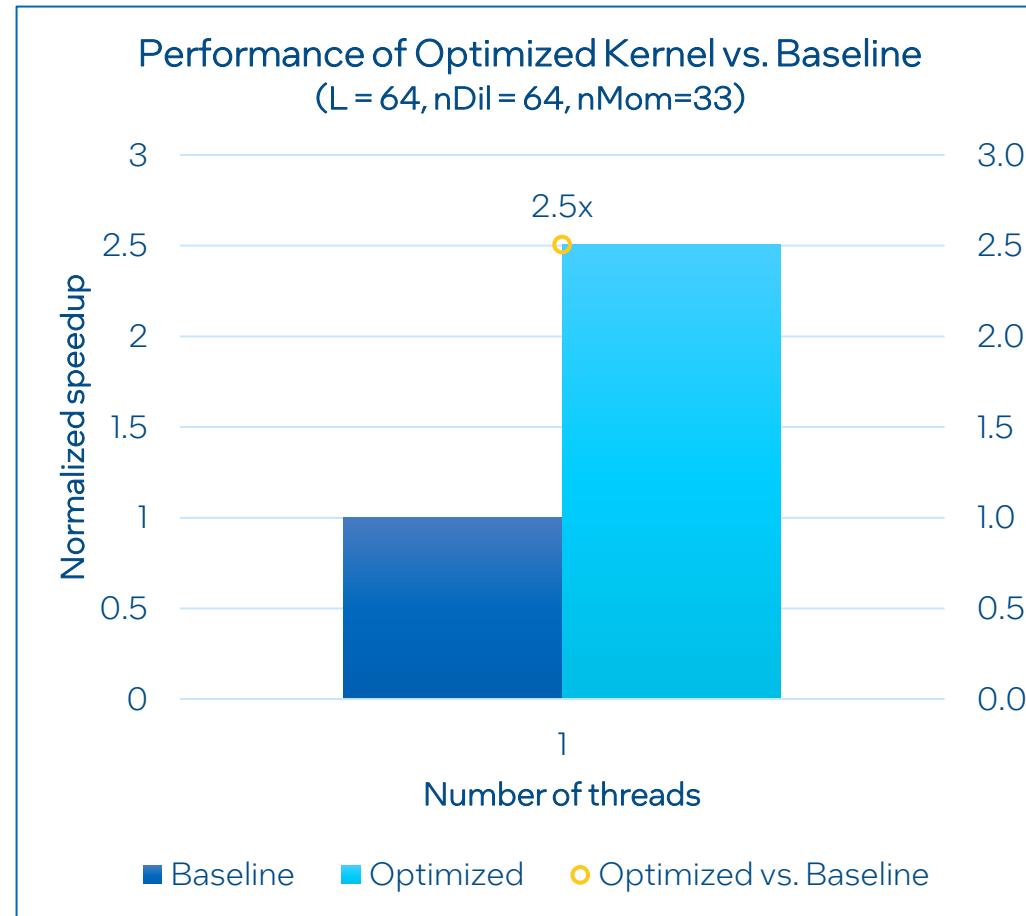
```
input: q1, q2, q3      // 3D array, [nD,nColor,nX]
       momBuf        // 2D array, [nMom, nX]
output: res            // 4D array, [nMom, nD1, nD2, nD3]
intermediate: diq     // 4D array, [bsizeD1, bsizeD2, nColor, bsizeX]
                  singlet // 4D array, [bsizeD3, bsizeD1, bsizeD2, bsizeX]
                  tmpBuf   // 4D array, [nD3, bsizeD1, bsizeD2, nMom]
for blockD1; do
  for blockD2; do
    init(tmpBuf, 0.)
    for blockX; do
      for d1 = 1 to bsizeD1; do
        for d2 = 1 to bsizeD2; do
          for iX = 1 to bsizeX; do
            calculate diq(:) = q1(~d1,:,iX) * q2(~d2,:,iX)
          end for
        end for
      end for
      for blockD3; do
        for d3 = 1 to bsizeD3; do
          for diq_elem in diq; do
            for iX = 1 to bsizeX; do
              calculate singlet(:) = diq_elem * q3(~d3,:,iX)
            end for
          end for
        end for
      end for
      // With JIT GEMM
      tmpBuf += singlet * momBuf
    end for // blockX
    storing res <- tmpBuf
  end for // blockD2
end for // blockD1
```

- Change data layout to have contiguous memory accesses.

- Cache Blocking in d1, d2, d3, nX.

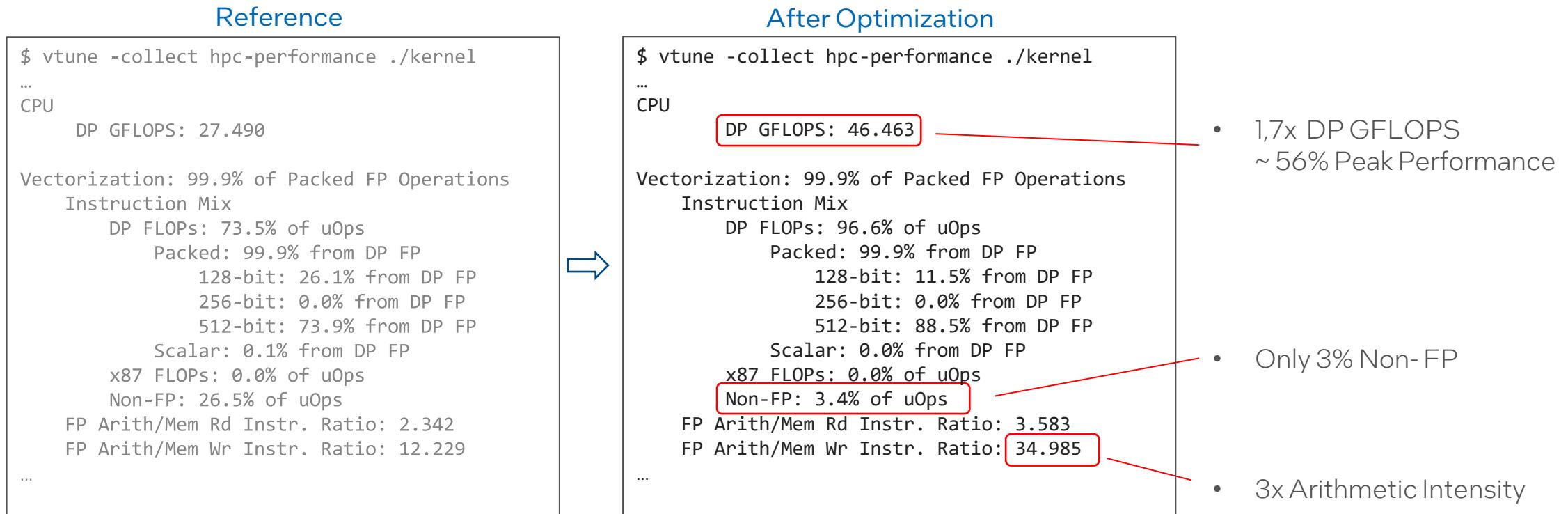
- Intel MKL Just-in-Time (JIT) Code Generation for Small Matrix Multiplication

4. Single-core Optimization (2)



4. Single-core Optimization (3)

Profiling with a small test case ($L = 32$, $nDil = 32$)



4. Single-core Optimization (4)

Profiling for memory-access

Reference

```
$ vtune -collect memory-access ./kernel
...
CPU Time: 10.515s
    Memory Bound: 37.2% of Pipeline Slots
        L1 Bound: 1.1% of Clockticks
        L2 Bound: 10.3% of Clockticks
        L3 Bound: 3.5% of Clockticks
    DRAM Bound: 15.0% of Clockticks
Elapsed Time
    Store Bound: 1.5% of Clockticks
Time
    Loads: 19,793,093,775
    Stores: 4,173,125,190
    LLC Miss Count: 97,540,950
        Local DRAM Access Count: 39,002,730
        Remote DRAM Access Count: 45,503,185
        Remote Cache Access Count: 0
    Average Latency (cycles): 41
...
```

After Optimization

```
$ vtune -collect memory-access ./kernel
...
CPU Time: 6.580s
    Memory Bound: 36.9% of Pipeline Slots
        L1 Bound: 5.7% of Clockticks
        L2 Bound: 2.4% of Clockticks
        L3 Bound: 14.7% of Clockticks
    DRAM Bound: 4.8% of Clockticks
Elapsed Time
    Store Bound: 0.0% of Clockticks
Time
    Loads: 12,038,361,140
    Stores: 1,235,037,050
    LLC Miss Count: 0
        Local DRAM Access Count: 0
        Remote DRAM Access Count: 0
        Remote Cache Access Count: 0
    Average Latency (cycles): 42
...
```



- L3 Bound instead of DRAM Bound
- 40% Less cycles for Loads
- 70% Less cycles for Stores
- 0 stalled cycles for LLC Misses

5. Node-level Optimization (1)

```
input: q1, q2, q3          // 3D array, [nD,nColor,nX]
       momBuf            // 2D array, [nMom, nX]
output: res               // 4D array, [nMom, nD1, nD2, nD3]
intermediate: diq         // 4D array, [bsizeD1, bsizeD2, nColor, bsizeX]
                     singlet // 4D array, [bsizeD3, bsizeD2, bsizeD1, bsizeX]
                     tmpBuf  // 2D array, [nD3, bsizeD2*bsizeD1,nMom]

for blockD1; do
  for blockD2; do
    init(tmpBuf, 0.)
    for blockX; do
      for d1 = 1 to bsizeD1; do
        for d2 = 1 to bsizeD2; do
          for iX = 1 to bsizeX; do
            calculate diq(:) = q1(~d1,:,iX) * q2(~d2,:,iX)
          end for
        end for
      end for
    end for
    for blockD3; do
      for d3 = 1 to bsizeD3; do
        for diq_elem in diq; do
          for iX = 1 to bsizeX; do
            calculate singlet(:) = diq_elem * q3(~d3,:,iX)
          end for
        end for
      end for
      tmpBuf += singlet * momBuf           // Using MKL JIT
    end for // blockX
    storing res <- tmpBuf
  end for // blockD2
end for // blockD1
```

- Which loop to parallelize?

5. Node-level Optimization (2)

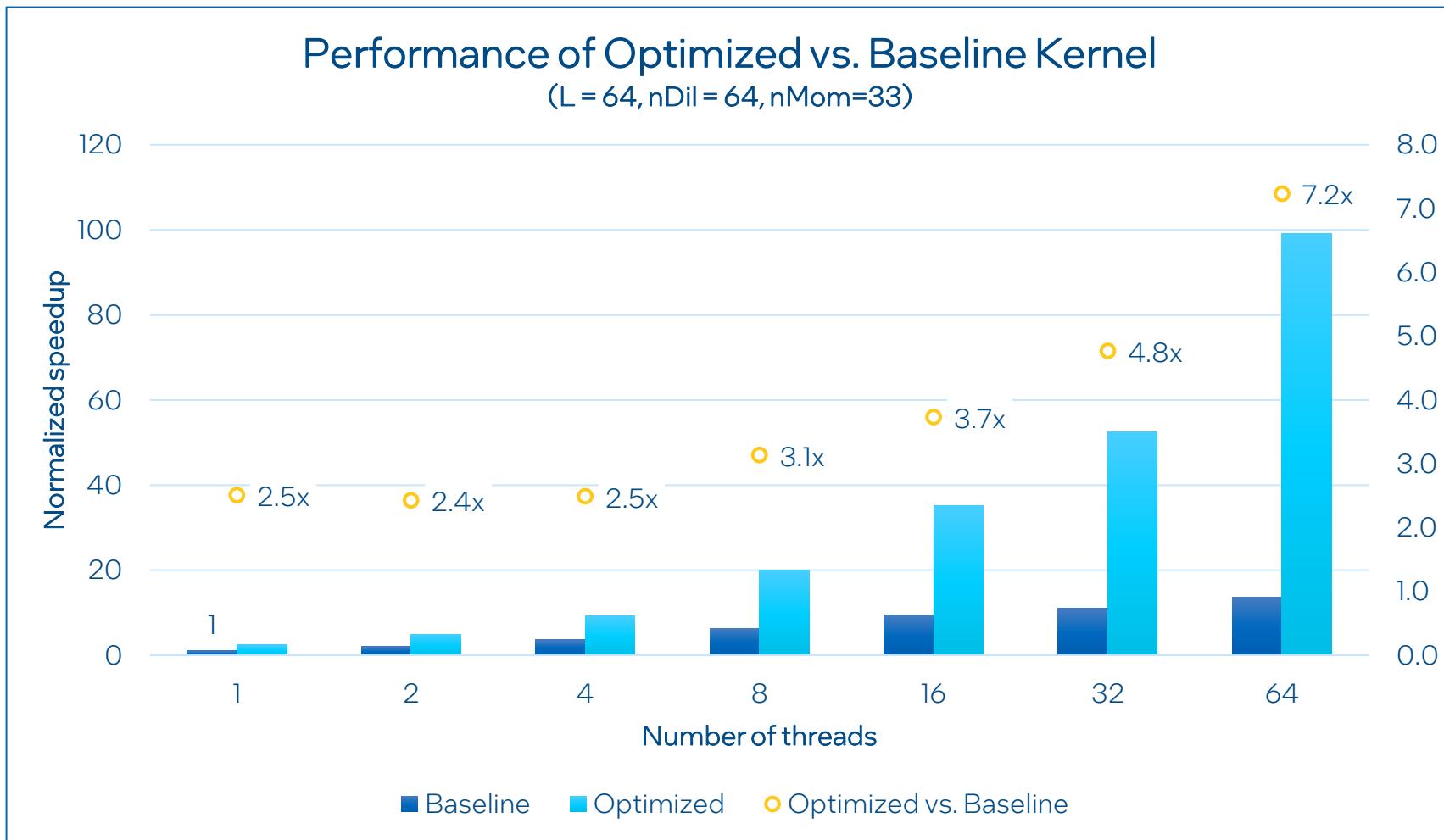
```
input: q1, q2, q3          // 3D array, [nD,nColor,nX]
       momBuf            // 2D array, [nMom, nX]
output: res               // 4D array, [nMom, nD1, nD2, nD3]
intermediate: diq         // 4D array, [bsizeD1, bsizeD2, nColor, bsizeX]
                     singlet // 4D array, [bsizeD3, bsizeD2, bsizeD1, bsizeX]
                     tmpBuf  // 2D array, [nD3, bsizeD2*bsizeD1,nMom]

#pragma omp parallel for collapse(2) firstprivate(q1, q2, q3)
for blockD1; do
  for blockD2; do
    init(tmpBuf, 0.)
    for blockX; do
      for d1 = 1 to bsizeD1; do
        for d2 = 1 to bsizeD2; do
          for ix = 1 to bsizeX; do
            calculate diq(:) = q1(~d1,:,ix) * q2(~d2,:,ix)
          end for
        end for
      end for
    end for
    for blockD3; do
      for d3 = 1 to bsizeD3; do
        for diq_elem in diq; do
          for ix = 1 to bsizeX; do
            calculate singlet(:) = diq_elem * q3(~d3,:,ix)
          end for
        end for
      end for
    end for
    tmpBuf += singlet * momBuf      // Using MKL JIT
  end for // blockX
  storing res <- tmpBuf
end for // blockD2
end for // blockD1
```

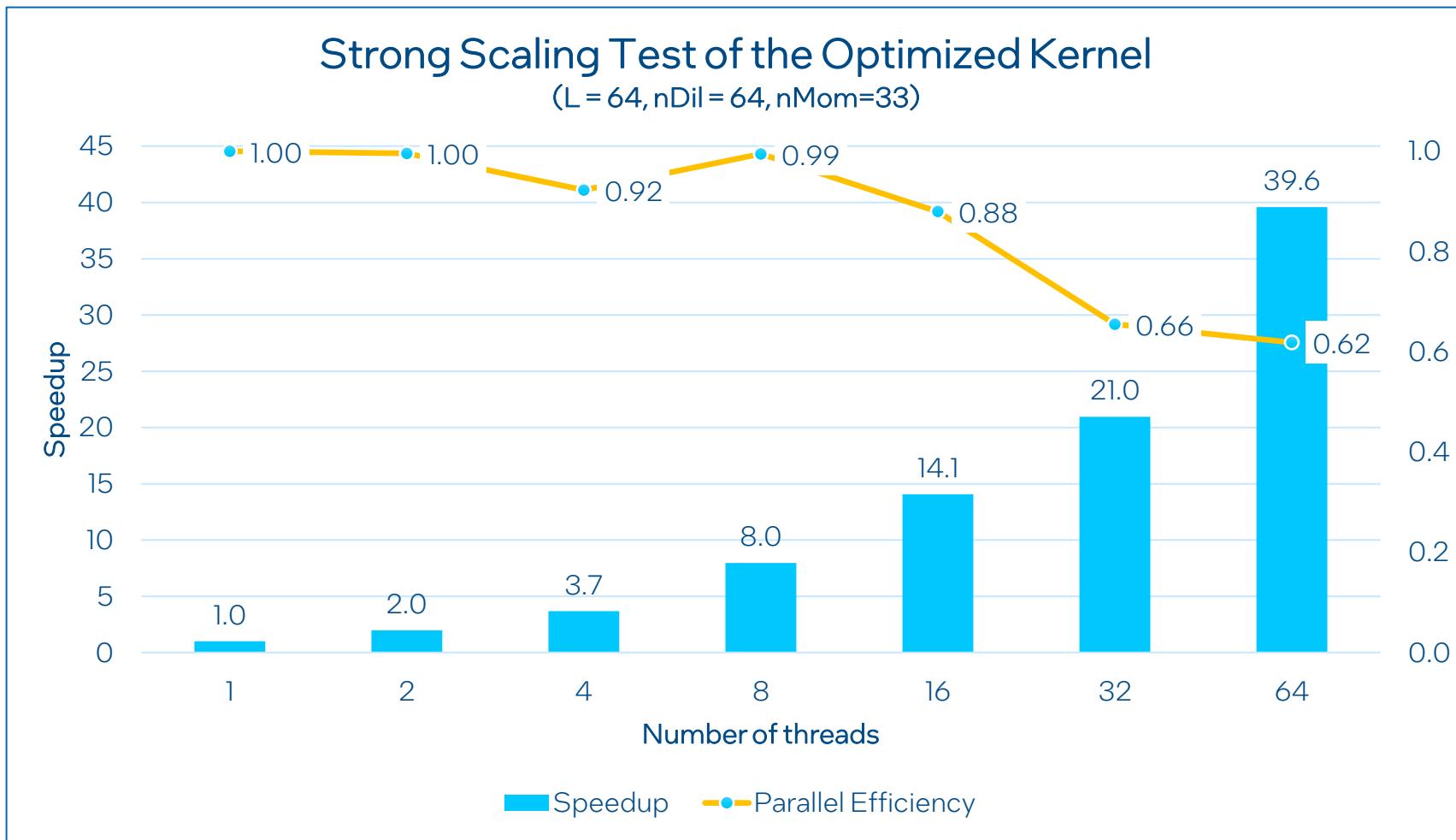
Using:

- **Collapse**: parallelize multiple loops
- **Firstprivate**: declare “read-only” for q1, q2, q3.
- Binding set:
\$OMP_PLACES=cores
\$OMP_PROC_BIND=close

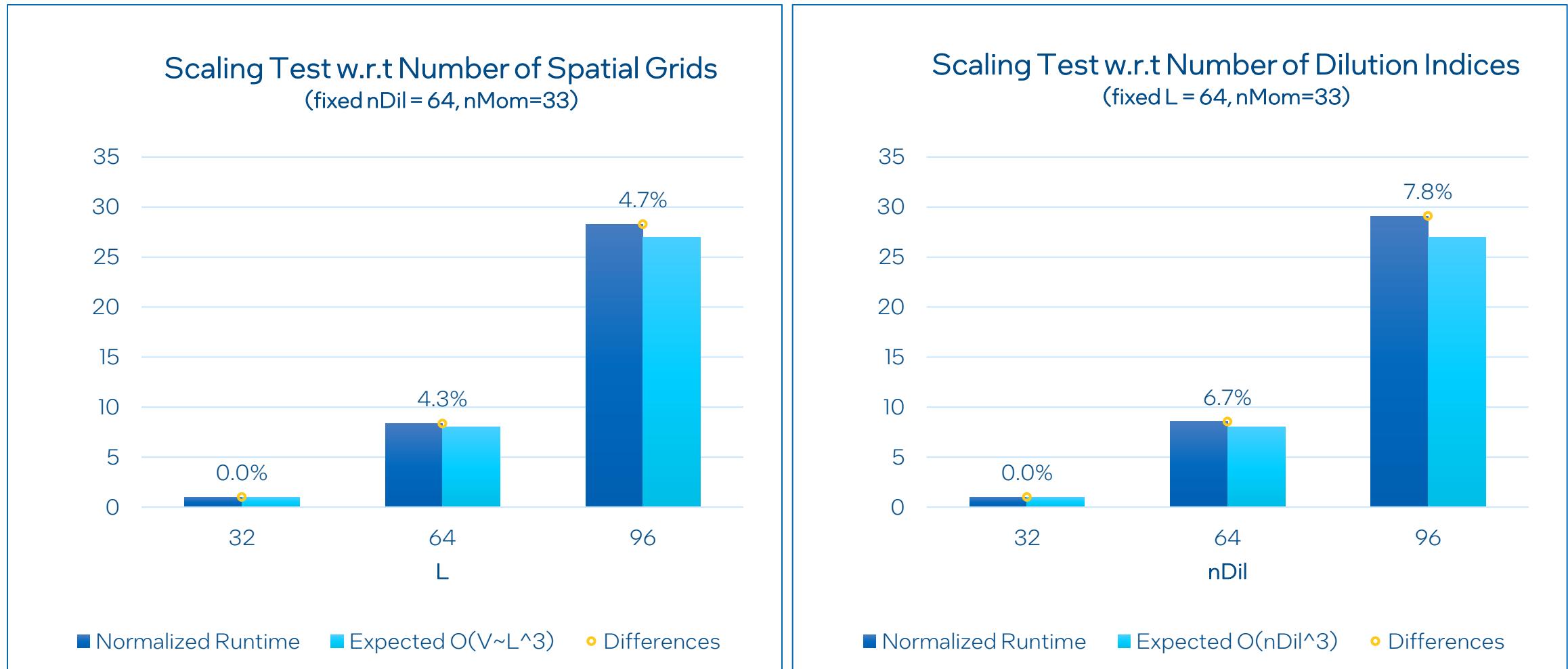
5. Node-level Optimization (3)



5. Node-level Optimization (4)



5. Node-level Optimization (5)



6. Conclusion

- Modern spectroscopy methods entail large measurement cost.
- Challenge: scale computation of observables to ever-larger problem sizes for simulations near the physical point.
- Optimized computation of stochastic LapH baryon blocks, achieving
 - up to **7.2x speedup** over previous QDP + gemm-based implementation
 - favorable scaling with core count, good parallel efficiency
 - good scaling with problem size (N_{dil} , V), no nasty surprises.
- Get ready for more ambitious production runs.

The Intel logo is displayed in white against a solid blue background. The word "intel" is written in a lowercase, sans-serif font. A small, solid blue square is positioned above the letter "i". The letter "i" has a vertical stroke extending upwards from its top loop. The letter "t" has a vertical stroke extending downwards from its top loop. The letters "n", "e", and "l" are standard lowercase forms.

Backup (1)

$$\mathcal{B}_{\mathbf{p}}^{d_1 d_2 d_3} = \sum_{\mathbf{x}} \sum_{a,b,c} e^{-i\mathbf{p}\mathbf{x}} \varepsilon_{abc} q_{\mathbf{x}a}^{d_1} q_{\mathbf{x}b}^{d_2} q_{\mathbf{x}c}^{d_3}$$

$$\varepsilon_{abc} = \begin{cases} +1 & (abc) \in \{(123), (231), (312)\}, \\ -1 & (abc) \in \{(321), (132), (213)\}, \\ 0 & \text{otherwise.} \end{cases}$$

Backup (2) – Different approaches for baryon constructions

$$q_{a\mathbf{x}}^d = \sum_{l=1}^{N_{\text{ev}}} Q_{dl} \phi_{a\mathbf{x}}^l$$

$$\mathcal{B}_{\mathbf{p}}^{d_1 d_2 d_3} = \sum_{\mathbf{x}} \sum_{a,b,c} e^{-i\mathbf{p}\mathbf{x}} \varepsilon_{abc} q_{\mathbf{x}a}^{d_1} q_{\mathbf{x}b}^{d_2} q_{\mathbf{x}c}^{d_3}$$

$$\mathcal{T}_{\mathbf{p}}^{lmn} = \sum_{\mathbf{x}} e^{-i\mathbf{p}\mathbf{x}} \sum_{a,b,c} \varepsilon_{abc} \phi_{\mathbf{x}a}^l \phi_{\mathbf{x}b}^m \phi_{\mathbf{x}c}^n$$

$$\mathcal{B}_{\mathbf{p}}^{d_1 d_2 d_3} = \sum_{l,m,n=1}^{N_{\text{ev}}} Q_{d_1 l} Q'_{d_2 m} Q''_{d_3 n} \mathcal{T}_{\mathbf{p}}^{lmn}$$

Backup (3)- Reference Implementation with OpenMP

```
input: q1, q2, q3          // 3D array, [nD,nX,nColor]
       momBuf            // 2D array, [nMom, nX]
output: res                // 4D array, [nMom, nD1, nD2, nD3]
intermediate: diq          // 2D array, [nX, nColor]
                  singlet   // 1D array, [nRows, nX]
for d1 = 1 to nD1; do      // dilution index 1
  for d2 = 1 to nD2; do    // dilution index 2
#pragma omp parallel for
  for ix = 1 to nX; do
    calculate diq(ix,:) = q1(d1,ix,:)* q2(d2,ix,:)
  end for
  rowIndex = 0;
  for d3 = 1 to nD3; do  // dilution index 3
#pragma omp parallel for
  for ix = 1 to nX; do
    calculate singlet(rowIndex, ix) = diq(ix,:)* q3(d3,ix,:)
  end for
  // blocked momentum projection
  rowIndex++;
  if rowIndex == nrows
    // With BLAS Level 3 - GEMM
    res(:, d1, d2, d3 - nrows : d3) = momBuf * singlet
    rowIndex = 0
  end if
end for
end for
end for
```

Backup (4) - Profiling for OpenMP

CPU

SP GFLOPS: 0.000
DP GFLOPS: 2,462.079
x87 GFLOPS: 0.000
CPI Rate: 0.456
Average CPU Frequency: 2.329 GHz
Total Thread Count: 64
Vectorization: 100.0% of Packed FP Operations
Instruction Mix
 DP FLOPs: 100.0% of uOps
 Packed: 100.0% from DP FP
 128-bit: 11.0% from DP FP
 256-bit: 0.0% from DP FP
 512-bit: 89.0% from DP FP
 Scalar: 0.0% from DP FP
 x87 FLOPs: 0.0% of uOps
 Non-FP: 0.0% of uOps
FP Arith/Mem Rd Instr. Ratio: 4.066
FP Arith/Mem Wr Instr. Ratio: 69.245

Effective Physical Core Utilization: 82.8% (52.970 out of 64)
Effective Logical Core Utilization: 41.9% (53.635 out of 128)
Serial Time (outside parallel regions): 0.029s (0.4%)
Parallel Region Time: 8.066s (99.6%)
Estimated Ideal Time: 6.862s (84.8%)
OpenMP Potential Gain: 1.204s (14.9%)

CPU Time: 446.248s

Memory Bound: 22.5% of Pipeline Slots
L1 Bound: 4.7% of Clockticks
L2 Bound: 1.5% of Clockticks
L3 Bound: 9.0% of Clockticks
DRAM Bound: 0.7% of Clockticks
Store Bound: 0.0% of Clockticks
NUMA: % of Remote Accesses: 75.0%
UPI Utilization Bound: 0.0% of Elapsed Time

Time

Loads: 667,973,038,590
Stores: 39,202,676,045
LLC Miss Count: 0
 Local DRAM Access Count: 0
 Remote DRAM Access Count: 0
 Remote Cache Access Count: 0
Average Latency (cycles): 23
Total Thread Count: 64
Paused Time: 2.456s

Bandwidth Utilization

Bandwidth Domain	Platform Maximum	Observed Maximum	Average
DRAM, GB/sec	354	46.900	13.052
DRAM Single-Package, GB/sec	177	28.200	10.753
UPI Utilization Single-link, (%)	100	34.500	11.405