

High-Performance X-Ray Scattering Data Analysis

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Background

HipGISAXS:

a massively-parallel high-performance grazing incidence small angle X-ray scattering data analysis software.

- Written in C++ with
 MPI + OpenMP [+ CUDA.]
- **Double-precision complex** number computations.



An Example Kernel

```
//O(10^{6})
for(int z = 0; z < nqz; ++ z) {</pre>
    int y = z % nqy;
    vector3c_t mq = rotate(qx[y], qy[y], qz[z], rot);
    complex t qpar = sqrt(mq[0] * mq[0] + mq[1] * mq[1]);
    ... more computations ...
    complex t temp ff(0.0, 0.0);
    for(int i r = 0; i_r < rsize; ++ i_r) {</pre>
                                                              // O(1) - O(10)
        for(int i h = 0; i h < hsize; ++ i h) {</pre>
                                                           // O(1) - O(10)
            ... more computations ...
            complex t expo val = exp(0.5 * mq[3] * h[i h]);
            complex_t sinc val = sinc(0.5 * mq[3] * h[i_h]);
            complex t bess val = cbessj(qpar * r[i r], 1) / (qpar * r[i r]);
            temp ff += sinc val * expo val * bess val;
        }
    }
    ... more computations ...
    complex_t temp2 = exp(temp1);
    ff[z] = temp ff * temp2;
}
```

Optimizing for Intel Processors: Platforms for Analysis

• Edison (Cray XC30) @ NERSC:

- Intel Ivy Bridge (Xeon E5-2695). 12 cores.
- **Babbage** @ NERSC:

> Intel Xeon Phi (KNC). 60 cores.

Optimizing for Intel Processors: Threading

- Mostly embarrassingly-parallel computations.
- Primary performance analysis tools used:
 - Intel VTune, TAU, PAPI.
- Effective threading using **OpenMP**:



- **Compiler-based auto-vectorization.** (Intel compiler 15.0.)
- Requirements for auto-vectorization:
 - Loop should be single-block, typically without branches/jumps.
 - Loop must be countable.
 - No backward loop-carried dependencies.
 - No special functions or subroutine calls (unless inlined.)
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Pragmas and explicit directives failed.

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- VML mode LA (Low Accuracy, 4 ulp.)
 - IvyBridge: Time = 2.68x base [speedup = 0.37], # instructions = 3.49x base.
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- VML mode EP (Enhanced Performance, 50% bits accurate.)
 - *IvyBridge:* Time = 0.50x base [speedup = 1.98], # instructions = 0.61x base.
 - MIC: Time = 0.086x base [speedup = 11.64].

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```
inline avx_m256c_t avx_mul_ccp(avx_m256c_t a, avx_m256c_t b) {
    avx_m256c_t v;
    avx_m256_t temp1 = _mm256_mul_pd(a.real, b.real);
    avx_m256_t temp2 = _mm256_mul_pd(a.imag, b.imag);
    avx_m256_t temp3 = _mm256_mul_pd(a.real, b.imag);
    avx_m256_t temp4 = _mm256_mul_pd(a.imag, b.real);
    v.real = _mm256_sub_pd(temp1, temp2);
    v.imag = _mm256_add_pd(temp3, temp4);
    return v;
}
```



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    return v;
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• Performance:

typedef struct {

} avx m256c t;

mm256d real;

mm256d imag;

Time = 0.35x base [speedup = 2.86], # instructions = 0.28x base.

Conclusions and Insights

- Auto-vectorization does not always work.
- Intrinsics are best for DP/complex computations.
 - Provide most flexibility in achieving higher performance for *non-typical* computes.

• **Biggest surprises: Intel MKL performance, e.g.** v?Exp()

- DP complex, average MKL time = **0.93x** base [speedup = **1.08**]
- DP real, average MKL time = **0.53x** base [speedup = **1.87**]
- SP complex, average MKL time = **0.34x** base [speedup = **2.97**]
- SP real, average MKL time = **0.25x** base [speedup = **4.07**]
- Would be great if efficient implementations of special functions like Bessel, Sinc were available.
- Already taking Intel's help.

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