



DATA TRANSFER AND REUSE ANALYSIS TOOL FOR GPU-OFFLOADING USING OPENMP

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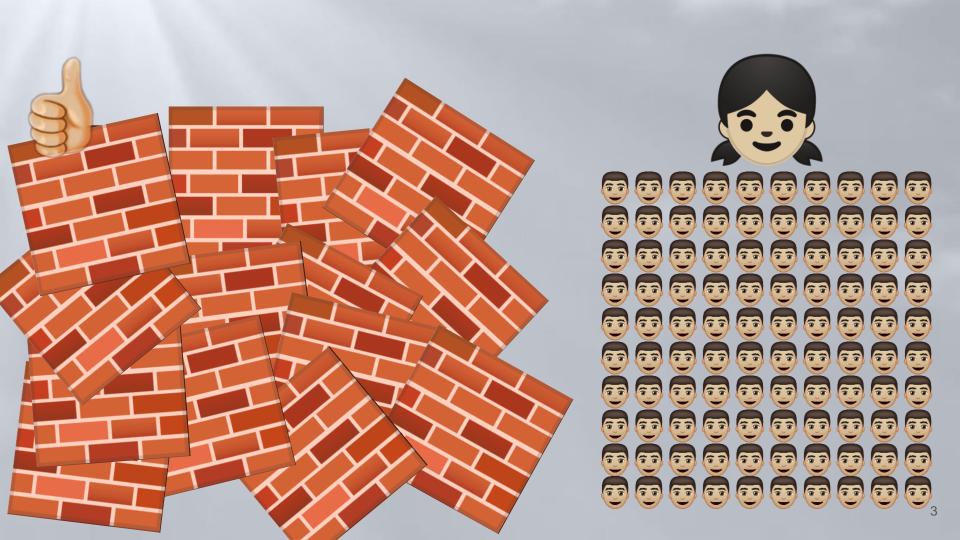
- HPC programs are parallel
- Long execution time
- Large size
- GPUs are increasingly important in HPC
 - ➤ Massive threading capability
 - > Energy efficient

MOTIVATION



A race car can travel faster, but a bus can carry more load





CHALLENGES OF PORTING To GPU





Portability

Highly dependent on underlying architecture and choice of programming model (CUDA)



Programmability

Different from existing programming languages. Extensive refactoring of code is required



Parallelism

What is the degree of parallelism?

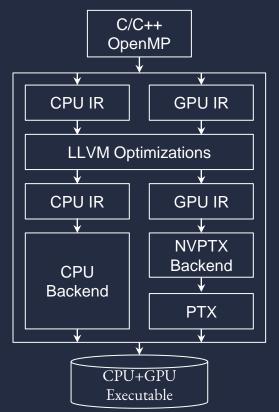


Data Handling

Requires explicit data transfers

- De-facto programming model for node-level parallelism
- OpenMP 4.X+ offers GPU programming ability
- OpenMP codes may also spend a significant portion of their execution time on data transfer
- Multiple GPU kernel calls may be reusing the same data

OpenMP



Data Transfer

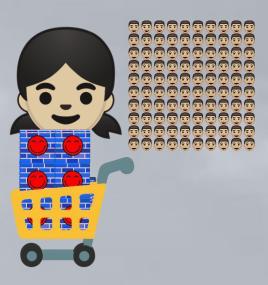
Explicit

Implicit

- target data map directive
- transfer all data *tofrom*
- compilers handle transfer



- 1. Color bricks in Blue
- 2. Draw smiley faces
- 3. Stack the bricks





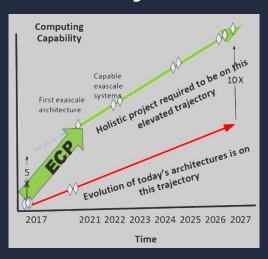
KEEP CALM

AND SEE THE

BIGGER PICTURE

Program Transformation for automatic GPUoffloading using **OpenMP**

Exascale Computing Project



- ❖ Major US DoE* project
- ❖ Deliver 2 capable Exascale system
- ♦ Exaflop/s rate is 10**18 floating point operations per second
- * Acceptable power
- Develop applications to utilize them
- Develop softwares to make them usable
- ❖ Grid Data Parallel Math library in C++ *Department of Energy 8

- Design & develop a compiler framework for C/C++ that can automatically
 - Recognize data reuse opportunities in an application
 - ➤ Insert pertinent "omp data map" directive in the source code accordingly.
 - ➤ Assumption target offloading code already inserted

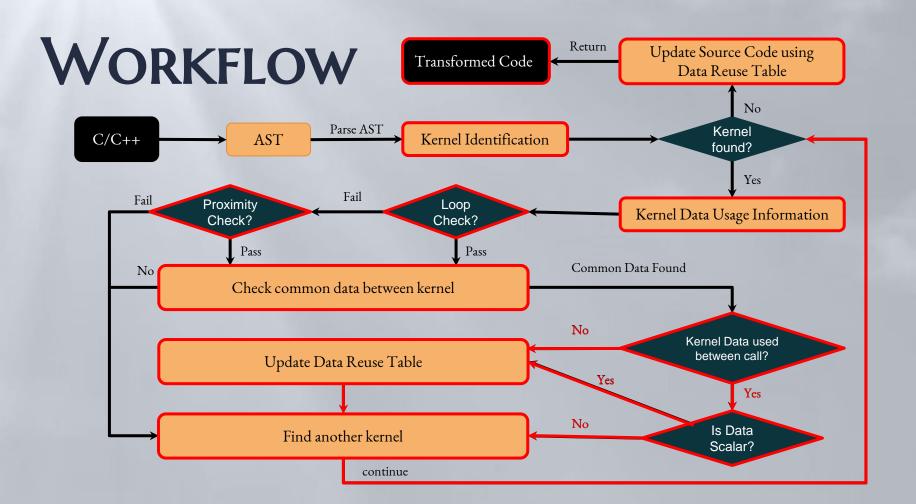
DATA REUSE ANALYSIS



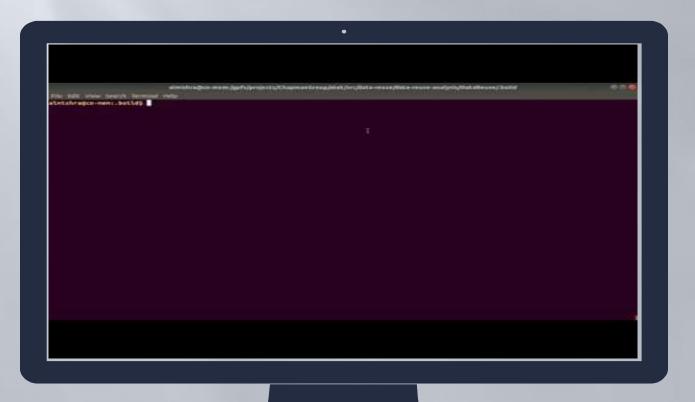
Example Code: Multiply 3 Matrices

```
#pragma omp target data map(to: A[0:N*N], B[0:N*N], C[0:N*N])
                        map(tofrom: D[0:N*N]) map(alloc: temp[0:N*N])
{ // data region starts
// Kernel 1
#pragma omp target teams distribute parallel for collapse(2)
  for (int i = 0; i < N; i++)
   for (int j = 0; j < N; j++) {
     temp[i * N + j] = 0;
      for (int k = 0; k < N; k++)
       temp[i * N + j] += A[i * N + k] * B[k * N + j];
// Kernel 2
#pragma omp target teams distribute parallel for collapse(2)
  for (int i = 0; i < N; i++)
   for (int j = 0; j < N; j++)
     for (int k = 0; k < N; k++)
       D[i * N + j] += temp[i * N + k] * C[k *
} // data region ends
```

Code inserted automatically



DEMO



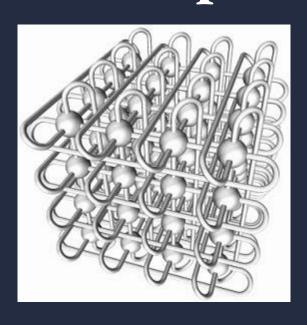
EXPECTATION

REALITY

```
#pragma omp target data map(to: A, B, C) \
                     map(tofrom: D) map(alloc: temp)
#pragma omp target teams distribute parallel for
 for (int i = 0; i < N; i++)
   for (int j = 0; j < N; j++) {
     for (int k = 0; k < N; k++)
       temp[i][j] += A[i][k] * B[k][j];
#pragma omp target teams distribute parallel for
 for (int i = 0; i < N; i++)
   for (int j = 0; j < N; j++)
     for (int k = 0; k < N; k++)
       D[i][j] += temp[i][k] * C[k][j];
```

```
#pragma omp target data map(to:temp)
#pragma omp target data map(to:A,B)
#pragma omp target teams distribute parallel for
  for (int i = 0; i < N; i++)
    for (int j = 0; j < N; j++) {
      for (int k = 0; k < N; k++)
        temp[i][j] += A[i][k] * B[k][j];
#pragma omp target data map(to:C) map(tofrom:D)
#pragma omp target teams distribute parallel for
  for (int i = 0; i < N; i++)
    for (int j = 0; j < N; j++)
      for (int k = 0; k < N; k++)
        D[i][j] += temp[i][k] * C[k][j];
```

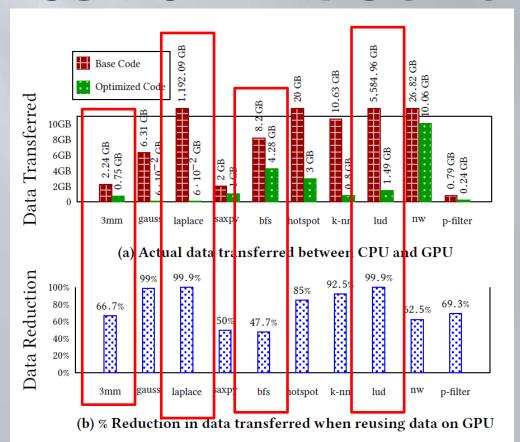
Experimental Setup



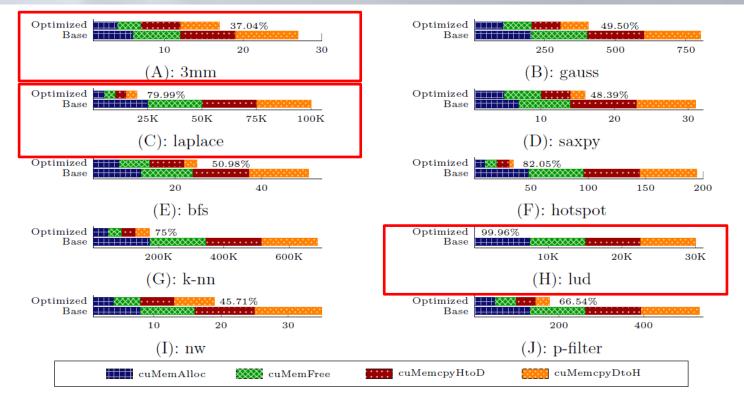
- SeaWulf cluster at Stony Brook University
- ♦ NVIDIA Tesla V100
- * SOLLVE
 - > LLVM version 8.0
- 4 microbenchmark and 6
 application from Rodinia

 Benchmark Suite
 - > Base Code
 - > Optimized Code

RESULTS - REDUCTION



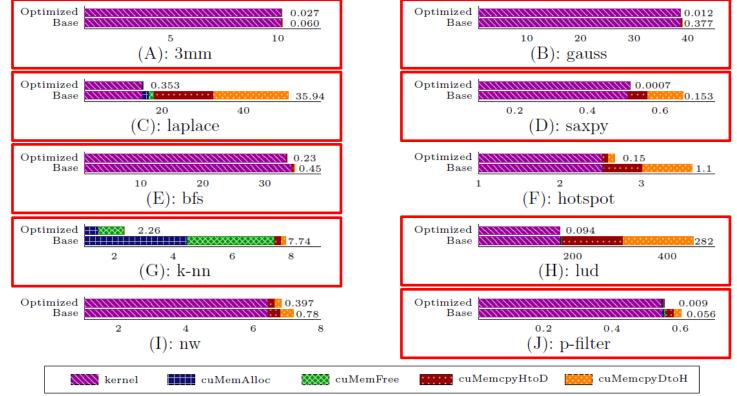
RESULTS - CUDA APIS



Number of calls to data transfer CUDA APIs.

The % at the tip of optimized code represents reduction in total number of calls when compared to base code.

RESULTS - EXECUTION TIME



Time taken (in sec) for different data management APIs and kernel computation time on V100 GPU. The numbers at the tip of each graph represents the time taken for data transfer only (in sec).

Related Work



Similar experiment ran using GCC with similar result

Other works

- ➤ An asymmetric distributed shared memory model for heterogeneous parallel systems. I Gelado, J E Stone, J Cabezas, S Patel, N Navarro, W M W Hwu
- ➤ Automatic cpu-gpu communication management and optimization. T B Jablin, P Prabhu, J A Jablin, N P Johnson, S R Beard, D I August.
- ➤ OMPSAN: Static verification of OpenMP's data mapping constructs. P Barua, J Shirako, W Tsang, J Paudel, W Chen, V Sarkar.

- If data is not reused on GPU
 - The performance of some application reduces significantly
- No loss of performance in any other cases
- User can accept or reject the transformation
- Future
 - Extend proximity check
 - Unified Memory
 - OpenMP 5.0
- Help other research aimed at Automatic GPU offloading of code

Conclusion & Future Work





Any questions?





You can find our work at:

• https://github.com/almishra/data-reuse-analysis

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