

How OPS (Optimizing Parallelizing System) May be Useful for Clang

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Optimizing parallelizing system (OPS)

OPS at a glance

- High-level internal representation: “Reprise”;
- Frontend (Clang based);
- Analysis:
 - Dependencies graph, Alias analysis, Computations graph, ...
- Transformations:
 - Recurrent loops parallelizing, Loop unrolling, Loop fusion, Loop nesting, Loop interchange

OPS at a glance (end)

- GUI for testing purposes;
- Data visualization:
 - Dependencies graph, ...
- Backends:
 - MPI, OpenMP, CUDA, VHDL, Clang, ...

Generating high-level code

- Automatic parallelization:
 - CUDA
 - OpenMP
 - MPI
 - VHDL
- Optimizing memory usage: tiling.

Clang + OPS integration



Figure 1: injecting OPS inside Clang

Low-level IR vs. High-level IR

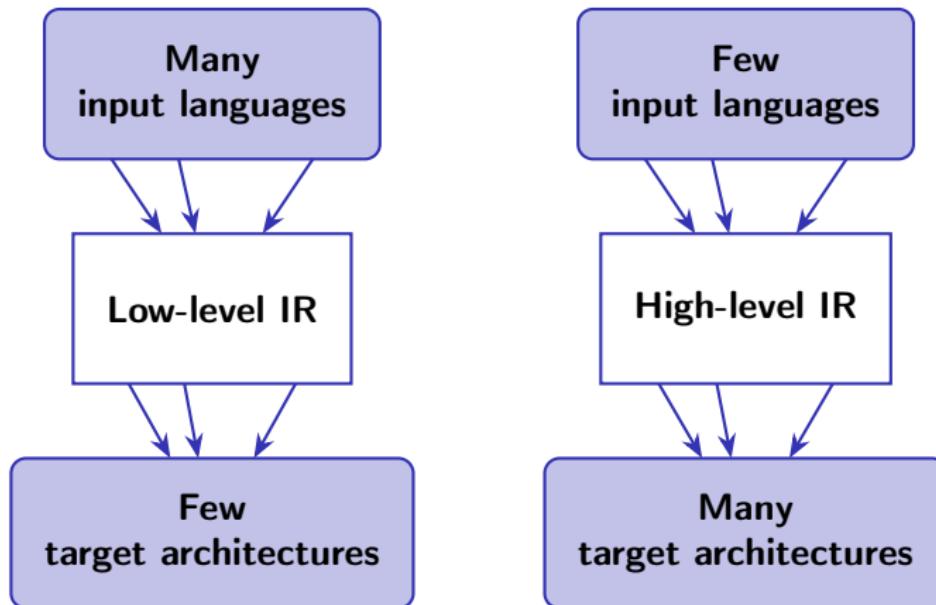


Figure 2: comparison of low-level and high-level internal representations

Advantages of using OPS

- Code generation for accelerators
 - GPU
 - FPGA
- Block data placement
 - shared memory
 - distributed memory
- Parallel programming visual aid: dependency visualization in terms of original source code.
- Dialog compilation

Block array placement

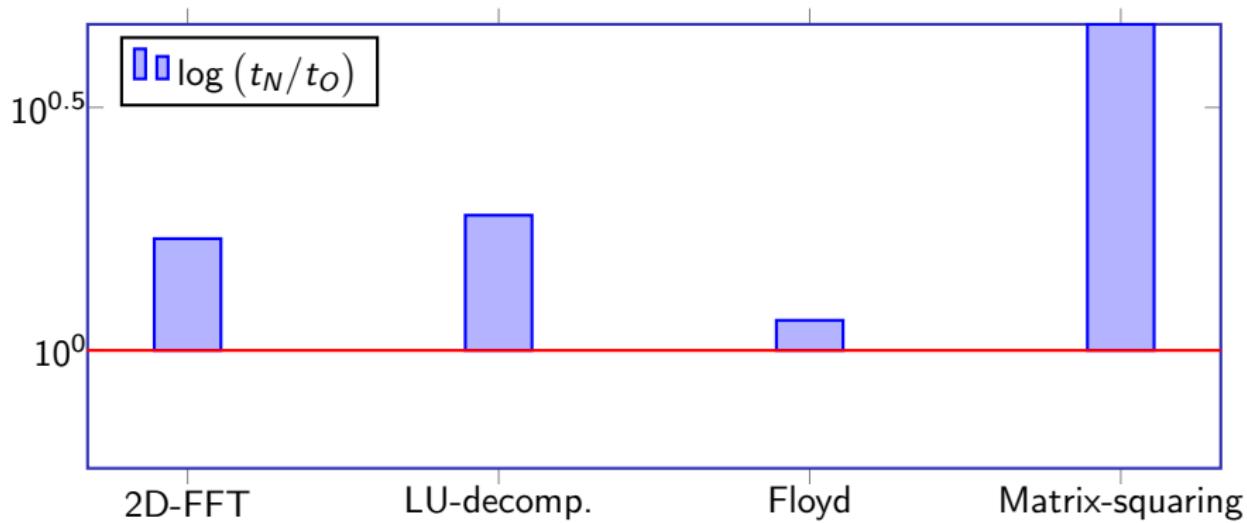


Figure 3: running times for algorithms with and without block placement in shared memory

Block affine array placement in distributed memory

Result

Automatically generated MPI code + block affine data placement.

Table 1: solving 3D Dirichlet problem for Poisson equation with iterative Jacobi algorithm

Size	Running time, s			
	1 node	2 nodes	4 nodes	8 nodes
128 × 128 × 128	38.25	19.74	10.9	5.97
256 × 256 × 256	310.19	165.64	87.11	48.64
384 × 384 × 384	1078	697.95	356.9	190.12
512 × 512 × 512	2786.47	1432.2	776.14	418.38

Graphics accelerator

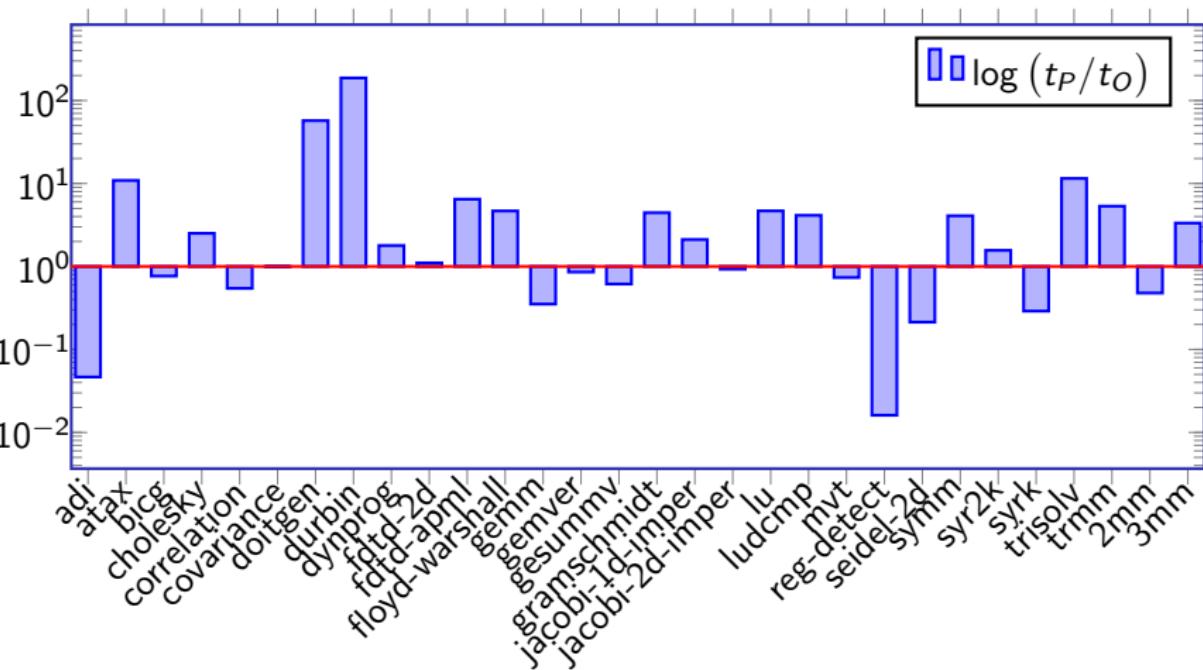
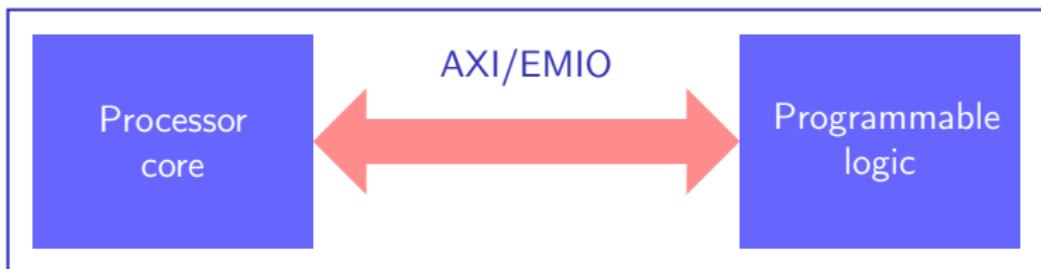
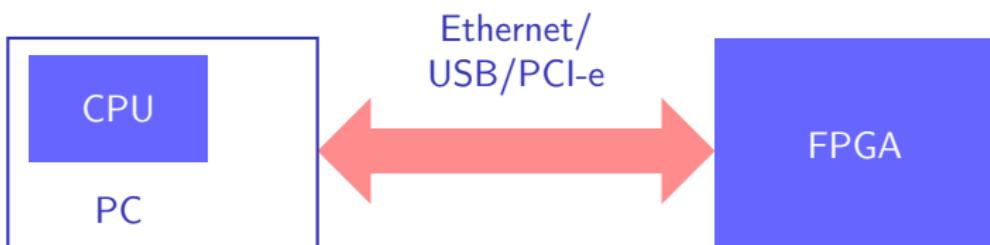


Figure 4: comparison of OPS-based solution with PPCG (<http://ppcg.gforge.inria.fr/>)

FPGA accelerator



a) on one core



b) on separate cores

Figure 5: the hybrid compute systems with FPGA/CPU

Dependency visualization

```
{  
    for(i=1 ; (i<10) ; i=(i+1))  
    {  
        for(j=1 ; (j<10) ; j=(j+1))  
        {  
            a[i][j] = (b[i][j] + a[i][(j-1)]) ;  
            c[i][j] = (a[i][j] + d[(i+1)][j]) ;  
            d[i][j] = 10 ;  
        }  
    }  
}
```

Анти зависимость
Носители (уровни): 0

Figure 6: parallel programmer's training simulator output

Dependency refinement in dialog

Example (Floyd algorithm)

```
for (k = 0; k < n; ++ k)
    for (i = 0; i < n; ++ i)
        for (j = 0; j < n; ++ j)
            if (a[i][j] > a[i][k] + a[k][j])
                a[i][j] = a[i][k] + a[k][j];
```

- Neither loop can be automatically parallelized due to data dependency.
- Actually, the dependency is not realized if $a[i][i] \geq 0$.
- The dialog compiler may ask the question to the programmer.

OpenOPS



- OpenOPS source code <https://github.com/0psGroup/open-ops>
- OPS website <http://www.ops.rsu.ru/>
- Web auto-parallelizer <http://ops.opsgroup.ru/en/>