

# Polyhedral Optimizations of Explicitly Parallel Programs

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# Introduction

- Moving towards Extreme-Scale and Exa-Scale computing systems
  - Billions of billions operations per second
- Enabling applications to fully exploit the systems is **not easy** !
- How ??

# Introduction

- Two approaches from past work:
  - 1) Manually parallelize using explicitly-parallel programming models (E.g., CAF, Cilk, Habanero, MPI, OpenMP, UPC etc)
    - Optimizations performed by programmer not compiler !
    - Tedious ! But can deliver good performance, with sufficient effort
  - 2) Automatically parallelize sequential programs
    - Done by compilers not humans !
    - Easy ! But, limitations exist.

# Motivation and Our Approach

- **Motivation**

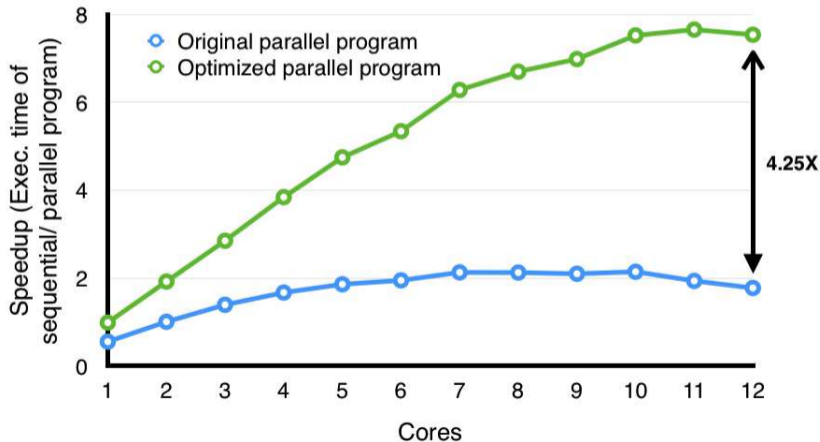
- Programmer expresses logical parallelism in the application and then let compiler perform optimizations accordingly

- **Our approach**

- *Automatically optimize explicitly-parallel programs*

# Glimpse of benefits

## Scalability of Jacobi benchmark [KASTORS] on Intel Westmere with 12 cores



- 1 Introduction and Motivation
- 2 Background**
- 3 Our framework
- 4 Evaluation
- 5 Related Work
- 6 Conclusions and Future work

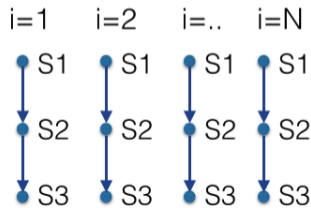
# Explicit Parallelism - Loop level parallelism

- Major difference between Sequential and Parallel programs
  - Sequential programs - total execution order
  - Parallel programs - partial execution order
- Loop-level parallelism (since OpenMP 1.0)
  - Loop is annotated with 'omp parallel for'
  - Iterations of the loop can be run in parallel

```

1 #pragma omp parallel for
2   for (i-loop) {
3       S1;
4       S2;
5       S3;
6   }

```



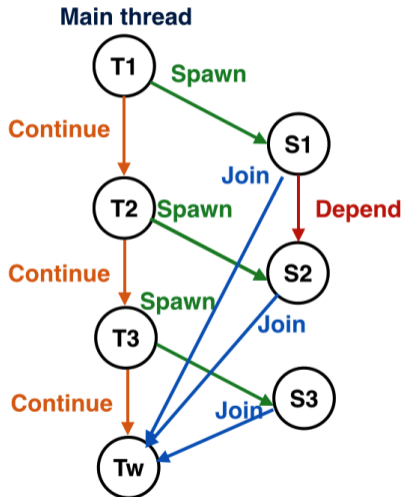
# Explicit Parallelism - Task level parallelism

- Task-level parallelism (OpenMP 3.0 & 4.0)
  - Region of code is annotated with 'omp task'
  - Synchronization
    - B/w parent and children - 'omp taskwait'
    - B/w siblings - 'depend' clause

```

1 #pragma omp task depend(out: A) // T1
2   {S1}
3 #pragma omp task depend(in: A) // T2
4   {S2}
5 #pragma omp task // T3
6   {S3}
7 #pragma omp taskwait // Tw

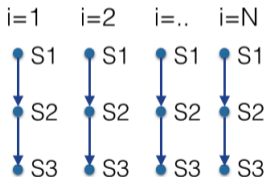
```



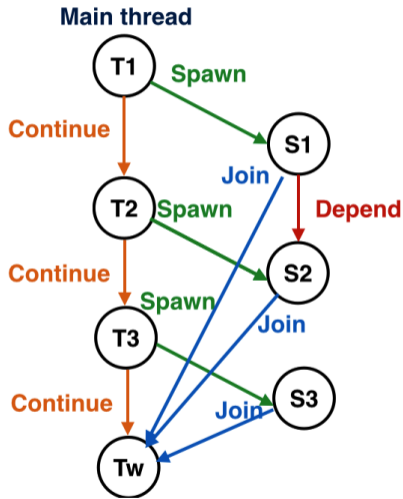


# Explicit Parallelism - Happens before relation

- Happens-Before relation
  - Specification of partial order among dynamic statement instances
  - $HB(S1, S2) = true \leftrightarrow S1$  must happen before  $S2$ , where  $S1$  and  $S2$  are statement instances.



$$HB(S1(i), S2(i)) = true$$



$$HB(S1, S2) = true, HB(S2, S3) = false$$

# Explicit Parallelism - Serial elision property

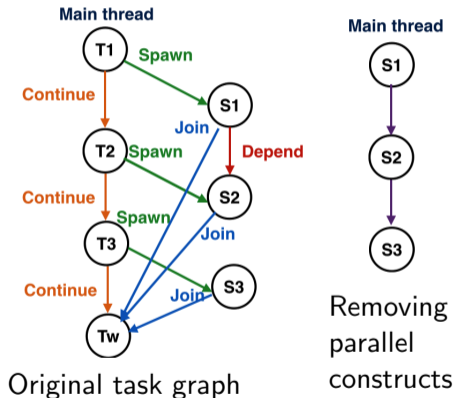
- Serial-Elision property
  - Removal of all parallel constructs results in a sequential program that is a valid (albeit inefficient) implementation of the parallel program semantics.

```

1 #pragma omp task depend(out: A) // T1
2   {S1}
3 #pragma omp task depend(in: A) // T2
4   {S2}
5 #pragma omp task // T3
6   {S3}
7 #pragma omp taskwait // Tw

```

Satisfies serial-elision



# Polyhedral Compilation Techniques

- Compiler techniques for analysis and transformation of codes with nested loops
- Algebraic framework for affine program optimizations
- Advantages over AST based frameworks
  - Reasoning at statement instance level
  - Unifies many complex loop transformations

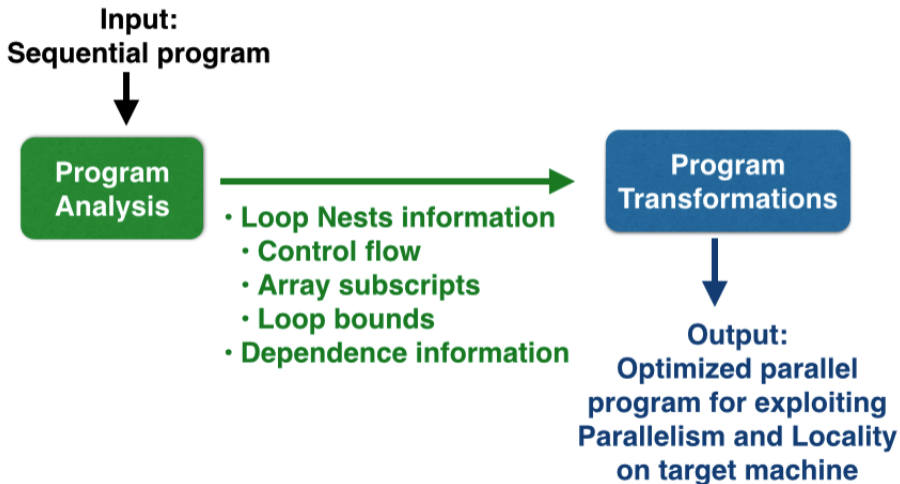
# Polyhedral Representation (SCoP)

- A statement (S) in the program is represented as follows in Static Control Part (SCoP):
  - 1) Iteration domain ( $\mathcal{D}^S$ )
    - Set of statement (S) instances
  - 2) Schedule ( $\Theta^S$ )
    - Assigns logical time stamp to the statement instances (S)
    - Gives ordering information b/w statement instances
    - **Captures sequential execution order of a program**
    - Statement instances are executed in increasing order of schedules
  - 3) Access function ( $\mathcal{A}^S$ )
    - Array subscripts in the statement (S)

# Polyhedral Compilation Techniques - Summary

- Advantages
  - Precise data dependency computation
  - Unified formulation of complex set of loop transformations
- Limitations
  - Affine array subscripts
    - But, conservative approaches exist !
  - Static affine control flow
    - Control dependences are modeled in same way as data dependences.
  - **Assumes input is sequential program**
    - **Unaware of happens-before relation in input parallel program**

# Automatic parallelization of sequential programs



- 1 Introduction and Motivation
- 2 Background
- 3 Our framework**
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# Polyhedral optimizations of Parallel Programs (PoPP)

Input: **Parallel program**  
(preferably with all possible  
logical parallelism)



Program  
Analysis

- 
- Loop Nests information
  - Control flow
  - Array subscripts
  - Loop bounds
  - Dependence information
  - **Happens-Before relation**

Program  
Transformations



Output:  
Optimized parallel  
program for exploiting  
Parallelism and Locality  
on target machine



# PoPP - Program Analysis

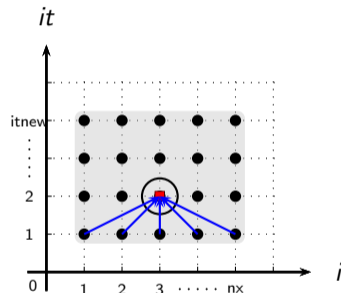
- Step1: Compute dependences based on the sequential order (use serial elision and ignore parallel constructs)

```

1 #pragma omp parallel
2 #pragma omp single
3 {
4     for (int it = itold + 1; it <= itnew; it++) {
5         for (int i = 0; i < nx; i++) {
6             #pragma omp task depend(out: u[i]) \
7             depend(in: unew[i])                // T1
8             for (int j = 0; j < ny; j++)
9 S1:                 u[i][j] = unew[i][j];
10        }
11        for (int i = 0; i < nx; i++) {
12 #pragma omp task depend(out: unew[i]) \
13 depend(in: f[i], u[i-1], u[i], u[i+1]) // T2
14     for (int j = 0; j < ny; j++)
15 S2:                 cpd(i, j, unew, u, f);
16        }
17    }
18 #pragma omp taskwait                // Tw
19 }

```

Conservative analysis, but may still capture vectorization possibility



(S2  $\rightarrow$  S1)  
dependences  
across it & i loops

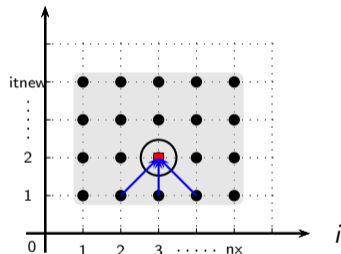
# PoPP - Program Analysis

- Step1: Compute dependences based on the sequential order (use serial elision and ignore parallel constructs)
- Step2: Compute happens-before relation (transitive closure)

```

1 #pragma omp parallel
2 #pragma omp single
3 {
4     for (int it = itold + 1; it <= itnew; it++) {
5         for (int i = 0; i < nx; i++) {
6 #pragma omp task depend(out: u[i]) \
7     depend(in: unew[i])           // T1
8         for (int j = 0; j < ny; j++)
9 S1:             u[i][j] = unew[i][j];
10        }
11        for (int i = 0; i < nx; i++) {
12 #pragma omp task depend(out: unew[i]) \
13     depend(in: f[i], u[i-1], u[i], u[i+1]) // T2
14        for (int j = 0; j < ny; j++)
15 S2:             cpd(i, j, unew, u, f);
16        }
17    }
18 #pragma omp taskwait           // Tw
19 }

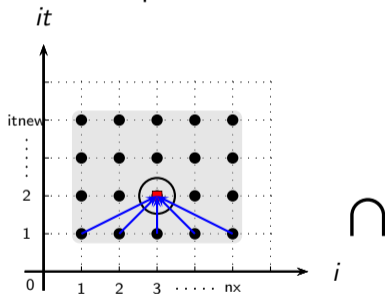
```



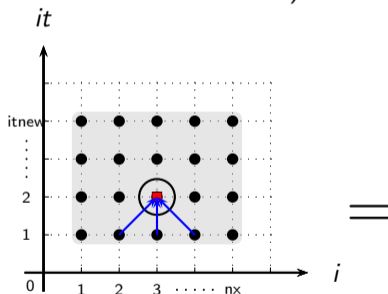
(S2→S1) HB edges  
across it & i loops

# PoPP - Program Analysis

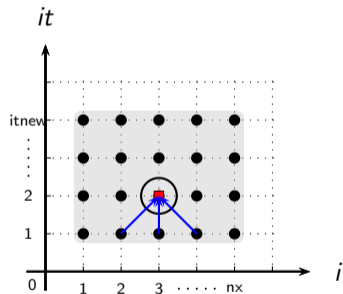
- Step1: Compute dependences
- Step2: Compute Happens-before relation (transitive closure)
- Step3: Intersect 1 & 2 (Gives best of both worlds)



Conservative  
dependences  $\mathcal{P}_1^{S2 \rightarrow S1}$



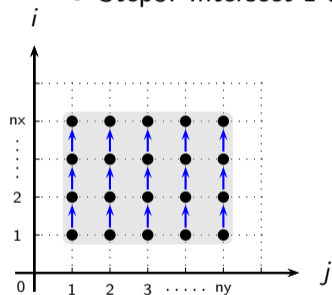
HB relation  
 $\mathcal{HB}_1^{S2 \rightarrow S1}$



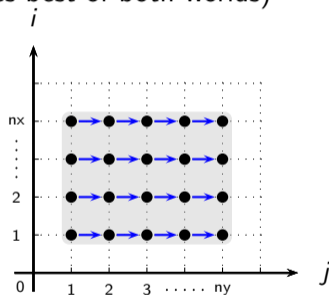
Refined dependences  
 $\mathcal{P}'_1^{S2 \rightarrow S1}$

# PoPP - Program Analysis

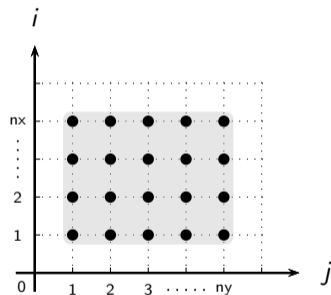
- Step1: Compute dependences
- Step2: Compute Happens-before relation (transitive closure)
- Step3: Intersect 1 & 2 (Gives best of both worlds)



Conservative  
dependences  $\mathcal{P}_1^{S1 \rightarrow S1}$   
(j-loop is parallel for S1)



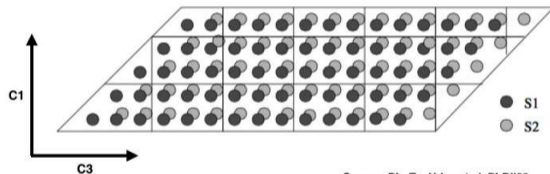
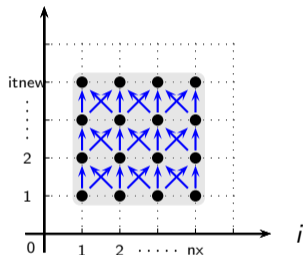
HB relation  
 $\mathcal{HB}_1^{S1 \rightarrow S1}$  (i-loop is  
parallel for S1)



Refined dependences  
 $\mathcal{P}'_1^{S1 \rightarrow S1}$  (No  
dependences for S1)

# PoPP - Program Transformations

- Step4: Use refined dependences in existing optimizations



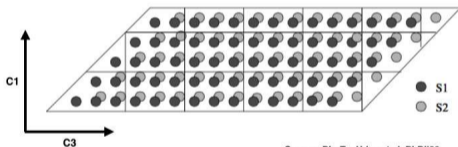
Source: PLuTo, Uday et.al, PLDI'08

Refined dependences,  $\mathcal{P}'_1^{S2 \rightarrow S1}$

- Skewing and tiling the iteration space

# PoPP - Code generation

- Step5: Generate optimized code using fine grained synchronization



Source: PLuTo, Uday et.al, PLDI'08



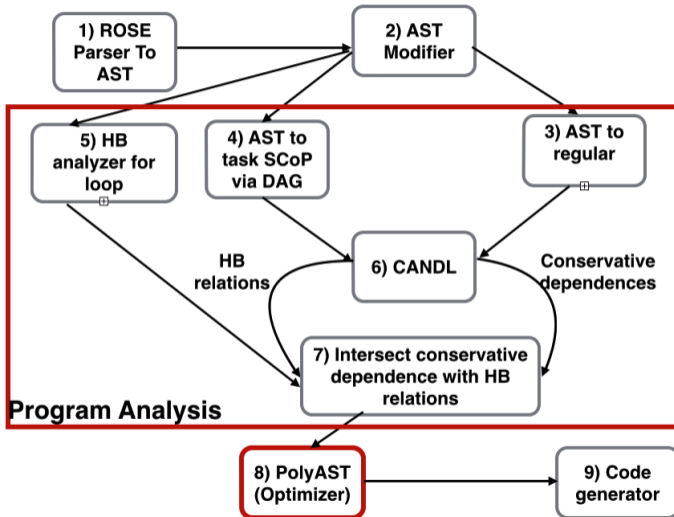
```

2 #pragma omp parallel for \
3   private(c3,c5) ordered(2)
4 for (c1 = itold + 1; c1 <= itnew; c1++) {
5   for (c3 = 2 * c1; c3 <= 2 * c1 + nx; c3++) {
6     #pragma omp ordered \
7       depend(sink: c1-1, c3) depend(sink: c1, c3-1)
8       if (c3 <= 2 * c1 + nx + -1) {
9         for (c5 = 0; c5 < ny; c5++)
10 S1:      u[-2*c1+c3][c5] = unew[-2*c1+c3][c5];
11       }
12
13       if (c3 >= 2 * c1 + 1) {
14         for (c5 = 0; c5 < ny; c5++)
15 S2:      cpd(-2*c1+c3-1, c5, unew, u, f);
16       }
17     #pragma omp ordered depend(source)
18   }}

```

- Doacross loop synchronization - OpenMP 4.1

# PoPP - Workflow (in ROSE Compiler)



# PoPP - Transformations & Code Generation

- Transformations - PolyAST framework [Shirako et.al SC'2014]
  - To perform loop optimizations
  - Hybrid approach of polyhedral and AST-based transformations
  - Detects reduction, doacross and doall parallelism from dependences
- Code Generation
  - Doall parallelism - `omp parallel for`
  - Doacross parallelism - `omp ordered depend`
    - Allows fine grained synchronization in multidimensional loop nests



# Extensions to Polyhedral frameworks

- Correctness of Intersection approach
  - Serial-elision property makes it correct !
- Computing conservative dependences
  - Non-affine subscripts, Unknown function calls, Non-affine conditionals etc
  - Extended access functions to support
- Extracting and Encoding task-related constructs in polyhedral representation (SCoP)
  - Constructed task SCoP to compute HB relation

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# Evaluation: Benchmarks and Platforms

	<b>Intel Xeon 5660 (Westmere)</b>	<b>IBM Power 8E (Power 8)</b>
<b>Microarch</b>	Westmere	Power PC
<b>Clock speed</b>	2.80GHz	3.02GHz
<b>Cores/socket</b>	6	12
<b>Total cores</b>	12	24
<b>Compiler</b>	gcc -4.9.2	gcc -4.9.2
<b>Compiler flags</b>	-O3 -fast(icc)	-O3

- KASTORS -Task parallel (3)
  - Jacobi, Jacobi-blocked, Sparse LU
- RODINIA -Loop parallel (8)
  - Back propagation, CFD solver, Hotspot, Kmeans, LUD, Needle-Wunch, Particle filter, Path finder
- Unanalyzable data access patterns - 7 benchmarks

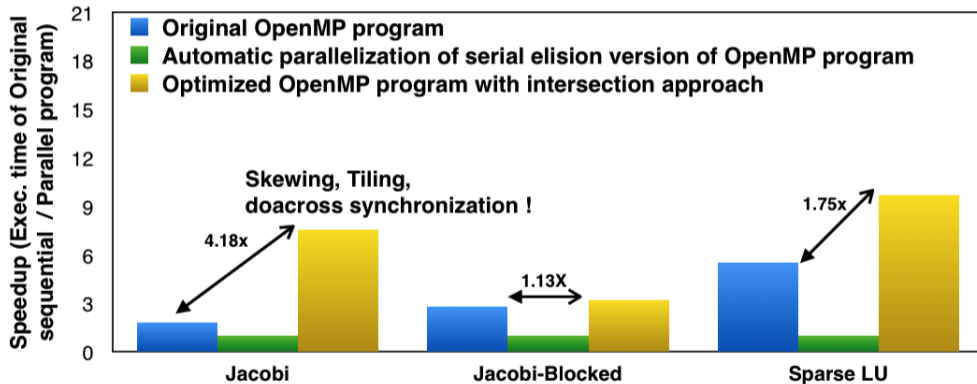
# Variants

Variants in the experiments

- **Original OpenMP program (Blue bars)**
  - Written by programmer
- **Automatic parallelization and optimization of serial elision version of OpenMP program (Green bars)**
  - Automatic optimizers
- **Optimized OpenMP program with intersection approach (Yellow bars)**
  - Our framework (PoPP)

# KASTORS suite + Intel Westmere (12 cores)

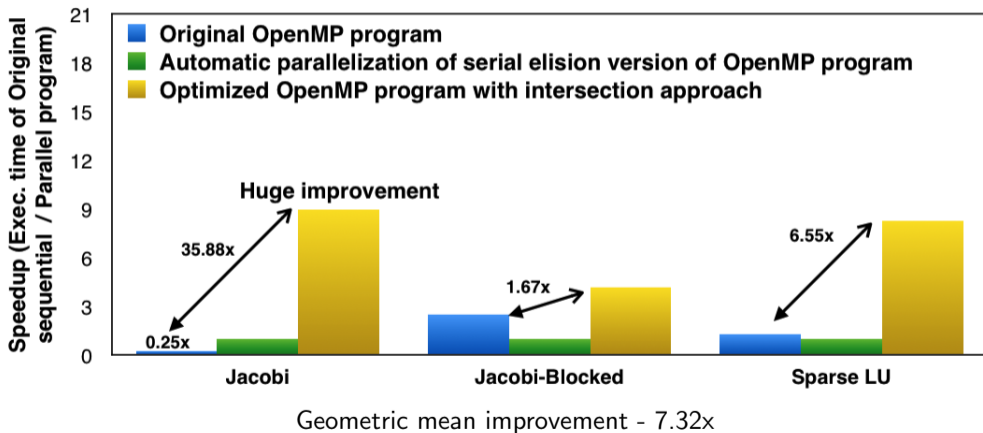
## Task-Parallel benchmarks (KASTORS) on Intel westmere (12 cores)



Geometric mean improvement - 2.02x

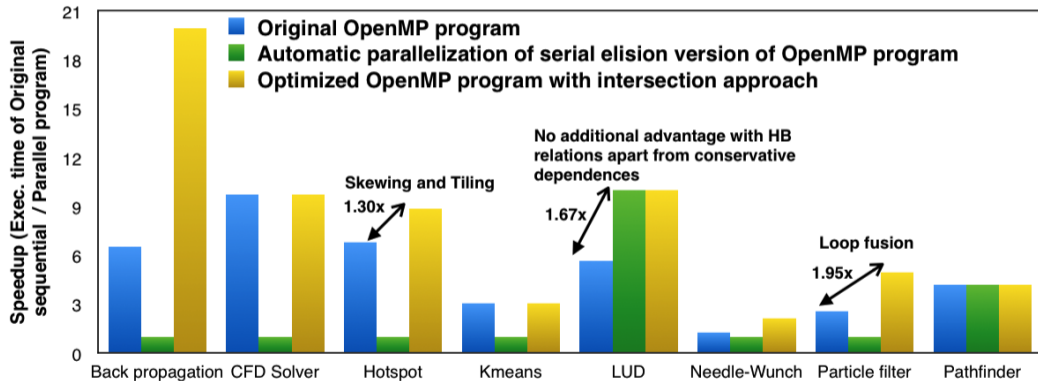
## KASTORS suite + IBM Power8 (24 cores)

## Task-Parallel benchmarks (KASTORS) on IBM Power8 (24 cores)



# RODINIA suite + Intel Westmere (12 cores)

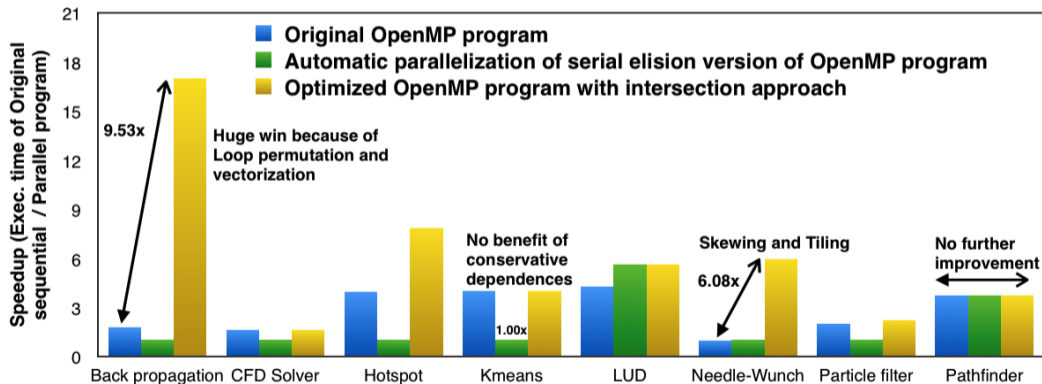
## Loop-Parallel benchmarks (Rodinia) on Intel Westmere (12 cores)



Geometric mean improvement - 1.48x

# RODINIA suite + IBM Power8 (24 cores)

## Loop-Parallel benchmarks (Rodinia) on IBM Power8 (24 cores)



Geometric mean improvement - 1.89x



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# Related work

- Dataflow analysis of explicitly parallel programs
  - Extensions to data-parallel/ task-parallel languages [J.F.Collard et.al Europar'96]
  - Extensions to X10 programs with async-finish languages [T. Yuki et.al PPOPP'13]
  - Above work is limited to analysis but we also focus on transformations.
- PENCIL - Platform Neutral Compute Intermediate Language [Baghdadi et.al. PACT'15]
  - Prunes data-dependence relation on parallel loops
  - No support for task parallel constructs as yet
  - Enforces certain coding restrictions related to aliasing, recursion etc.

# Related work (contd)

- Polyhedral optimization framework for DFGL [Sbirlea et.al LCPC'15]
  - Dataflow programming model - Implicitly parallel
  - Optimizations via polyhedral & AST-based framework
- Preliminary approach to optimize parallel programs [Pop and Cohen CPC'10]
  - Extract parallel semantics into compiler IR and perform polyhedral optimizations
  - Envisaged on considering OpenMP streaming extensions

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# PoPP - Conclusions and Future work

- Conclusions: Our approach
  - Reduced spurious dependences from conservative analysis by intersecting with HB relation
  - Broadened the range of legal transformations for parallel programs
  - Integrated HB relation from task-parallel constructs into Polyhedral frameworks
  - Geometric mean performance improvement of 1.62X on Intel Westmere and 2.75X on IBM Power8 - Larger improvements !!
- Future work:
  - Parallel constructs that don't satisfy serial-elision property
  - Extend to distributed-memory programming models (Eg: MPI)
  - Happens-Before relation for debugging
  - Beyond polyhedral

# Finally,

- *Optimizing explicitly parallel programs is a new direction for Parallel Architectures and Compilation Techniques (PACT)!*
- Acknowledgments
  - Support in part by DOE Office of Science Advanced Scientific Computing Research program through collaborative agreement DE-SC0008882
  - Rice Habanero Extreme Scale Software Research Group
  - PACT 2015 program committee
- Thank you!