

# Discovering Optimal Execution Policies in KRIPKE using RAJA

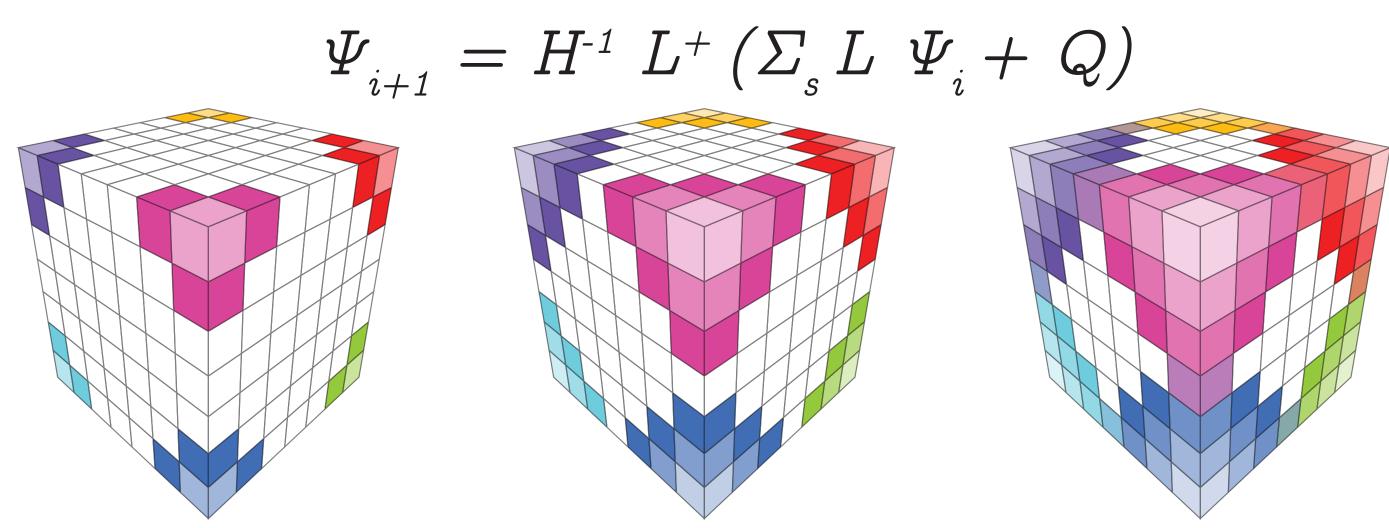
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# Motivation

- Legacy physics applications need updating to run well on newer architectures but are not always designed for architecture flexibility
- With architectures changing frequently (multicore, many-core, GPU), applications need to be adaptable to many different architectures.
- Adaptive, flexible programming layers are necessary to intelligently search large optimization spaces

# KRIPKE

- KRIPKE is a proxy application for Sn particle transport developed at LLNL
- Highly dimensional: composed of directions, groups, zones, and moments
- Many possible nestings of data and execution. Difficult to find the best
- Solves the linear Bolzmann equation using sweeps over a 3D domain space
- Goal: find optimal execution policies for common configurations of KRIPKE



Sweep (t=1)

Sweep (t=2)

Time sequence of the sweep kernel  $(H^{-1})$  moving through the mesh. Multiple sweeps can occur at the same time. Grid contention occurs when a location has equal manhattan distance from two or more sources (corners).

# **RAJA** Performance Portability Layer

- Provides C++ abstractions to enable architecture portability
- Predefined execution policies exist for SIMD, OpenMP, and CUDA
- Nested and advanced loop transformations (tiling, reordering) are available
- Goal: use RAJA to drive optimization search space exploration for KRIPKE

#### Example RAJA Execution Policy to apply NestedPolicy< ExecList< seq\_exec, seq\_exec, omp\_for\_nowait\_exec, simd\_exec>, OMP\_Parallel< Tile< TileList< tile\_none, tile\_none, tile\_none, tile\_fixed<512>>,

Permute<PERM\_JIKL>

Basic loop implementation for d in range(0,dom<IDirection>(id)): for nm in range(0,dom<IMoment>(id)): for g in range(0,dom<IGroup>(id)): for z in range(0,dom<IZone>(id)):

Nested Policy applied to loop #pragma omp parallel for z2 in range(0,dom<IZone>(id),512): for d in range(0,dom<IDirection>(id)): for nm in range(0,dom<IMoment>(id)): #pragma omp for nowait for g in range(0,dom<IGroup>(id)): for z in range(z2, z2+512):

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Sweep (t=3)

# **Policy Description and Generation**

#### Policy Search Space

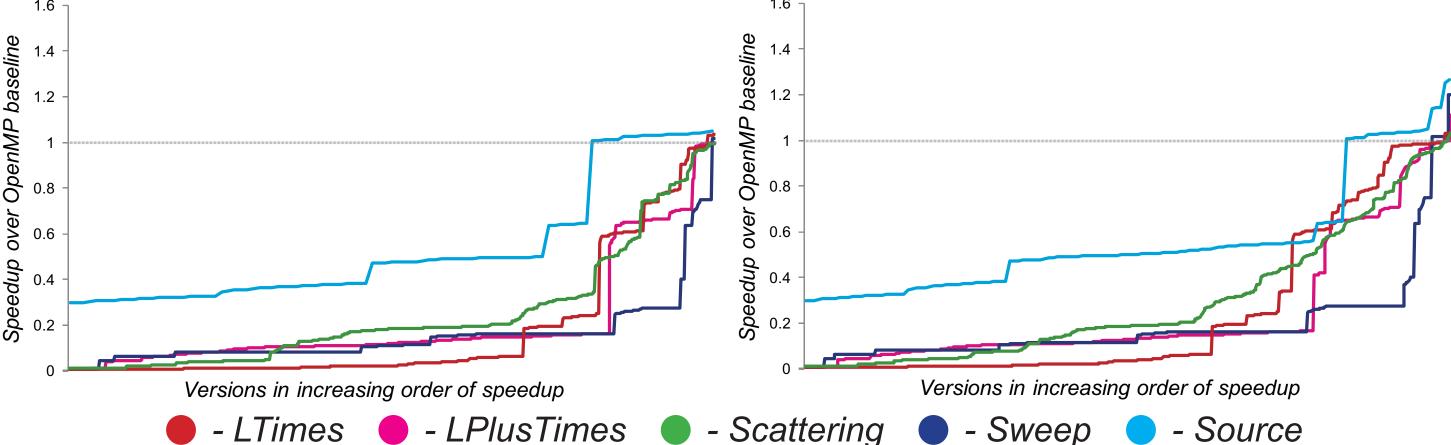
- Four execution policies: sequential, SIMD, OpenMP, collapsed OpenMP
- Five tiling policies: no tiling and fixed tiles of sizes 8, 32, 128, and 512
- Considered only loop valid nests, tiles must fit in L3 cache, no nested thread parallelism, OpenMP clauses only with OpenMP loop nests
- Policies are generated for each independent loop nest
- Five different loop nests:
  - 1. LTimes  $[L] \rightarrow 4$ -nested loop with 850K versions
  - 2. LPIusTimes [ $L^+$ ] -- 4-nested loop with 850K versions
  - 3. Scattering [ $\Sigma_{c}$ ] -- 4-nested loop with 850K versions
  - 4. Sweep  $[H^{-1}]$  -- 3-nested loop with 2.9K versions
  - Source [Q] -- 2-nested loop with 0.45K versions

# **Optimization Space Exploration**

- Assume kernel executions are independent of one another
- Too costly to run each execution policy for a larger Sn transport code.
- We propose two different strategies to explore the optimization space
- Goal: find optimal execution policies of kernels without exhaustive execution

### Hill-climbing Strategy

 $V \leftarrow \text{all versions of KRIPKE}$  $F \leftarrow \text{all features of a loop nest}$  $count \leftarrow 0$ do while count < threshold $p \leftarrow \mathsf{rand}(V)$  $best \leftarrow p$ foreach  $i, f \in \mathsf{shuffle}(\mathsf{enumerate}(F)) \mathsf{do}$ foreach  $option \in F_i$  do  $p_i \leftarrow option$  $count \leftarrow count + 1$ if time  $(p) < time (best) best \leftarrow p$  end • Limited to 10% of total search space • Limited to 20% of total search space • Speedup up to 3.1% over baseline.



Explored versions are shown by increasing speedup over OpenMP baseline. Subspace search does better than hill-climbing because the strategy was more likely to cover more tiling policies and consider non-local search spaces.

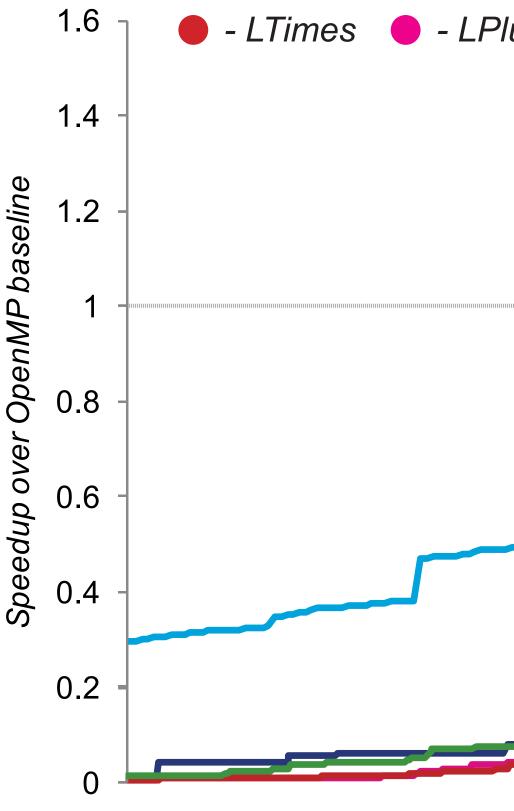
## Subspace Search Strategy

 $V \leftarrow \text{all versions of KRIPKE}$  $F \leftarrow \text{all features of a loop nest}$  $count \leftarrow 0$ do while count < threshold $V' \leftarrow \{ \mathsf{rand} (V) \}$ foreach  $i, f \in \mathsf{shuffle}(\mathsf{enumerate}(F)) \mathsf{do}$ foreach  $option \in F_i$  do  $V_{option} \leftarrow \{v_i \leftarrow option \ \forall \ v \in V'\}$  $count \leftarrow count + |V'|$  $V' \leftarrow V' \cup V_{option}$ end remove all but top k from V'

• Speedup up to 25.3% over baseline

# Performance Analysis

### Exhaustive Execution



- **Compiler:** Clang 3.8.0 with OpenMP support (-03 -march=native)

## **Comparison to Exhaustive Execution**

- entire KRIPKE proxy application by 19.5%.

# **Conclusion and Future Work**

## Future Work

# Acknowledgments and Resources



lusTimes	- Scattering	- Sweep	- Source

Versions in increasing order of speedup • Architecture: dual-socket Intel Xeon E5-2670, 32GB DDR3 RAM

• To evaluate our search strategies, we run all generated versions of KRIPKE.

• The best discovered policies improves over the basline performance of the

• Hill-climbing achieves up to 95.6% of optimal performance while subspace search achieves up to 98.8% of optimal performance.

• Used the RAJA performance portability layer to explore a large optimization space efficiently within the KRIPKE Sn transport proxy application

• Two different search space strategies can yield results up to 98.8% of optimal while only exploring 20% of the total search space.

• The best known execution time of KRIPKE improves by 19.5%.

• Expand results to include GPU execution policies (NVIDIA Kepler/Pascal) and nested parallelism with many-core (Intel Knight's Landing) architectures • Augment tiling policies to include multi-level tiling. This will be useful when targeting future architectures with complex memory hierarchies.

• Construct an accurate control-flow graph-based performance prediction model. The predictor replaces exaustive execution with only compilation.

[1] A. J. Kunen, T. S. Bailey, P. N. Brown, *KRIPKE - A Massively Parallel Transport Mini-App*, American Nuclear Society M&C, 2015 [https://codesign.llnl.gov/kripke.php] [2] R. D. Hornung and J. A. Keasler, *The RAJA Portability Layer: Overview and Status*, Tech Report, LLNL-TR-661403, Sep. 2014. [https://github.com/llnl/RAJA]