



A Framework for Resource Aware Multithreading

Pacific Northwest
NATIONAL LABORATORY
Proudly Operated by Battelle Since 1965

Problem Formulation

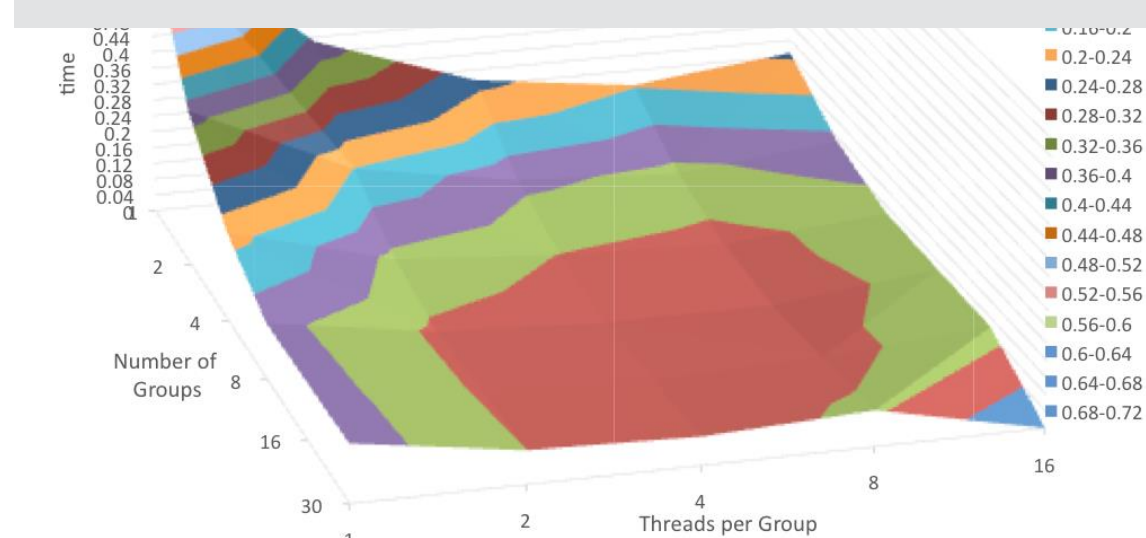
Current Data movement Strategy:

- Oblivious to the execution pattern of neighboring threads

Parallelism Vs Locality

- Accessing closed page is expensive both for performance and power
- Serialization of concurrent threads in memory banks

Parameter sweep between threads aggregated by groups, size of groups and time

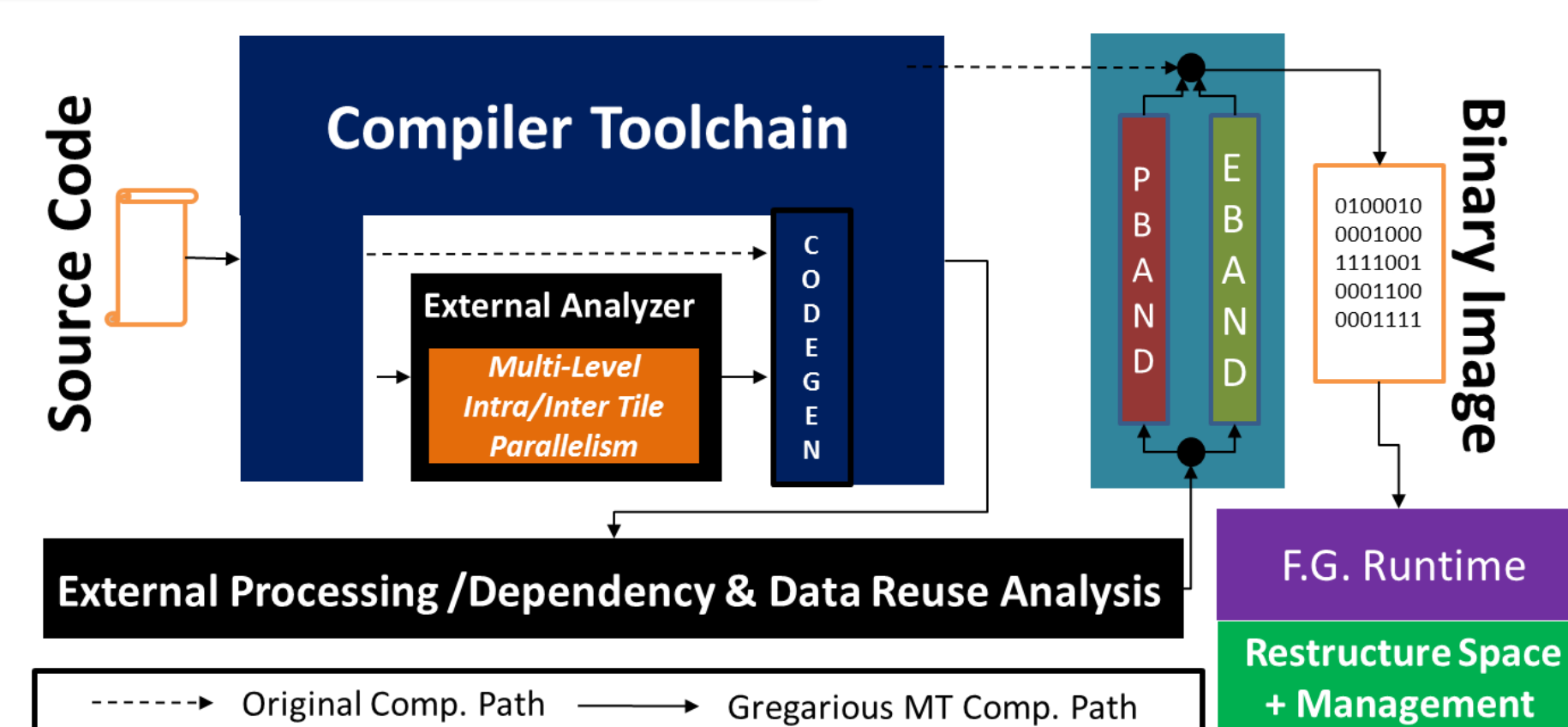


Aiming toward Exascale

- Stencil computation is at the heart of the simulation of several physical phenomena that is of scientific interest such as the Data Analysis Proxy App used in the Center for Exascale Simulation of Advanced Reactors.

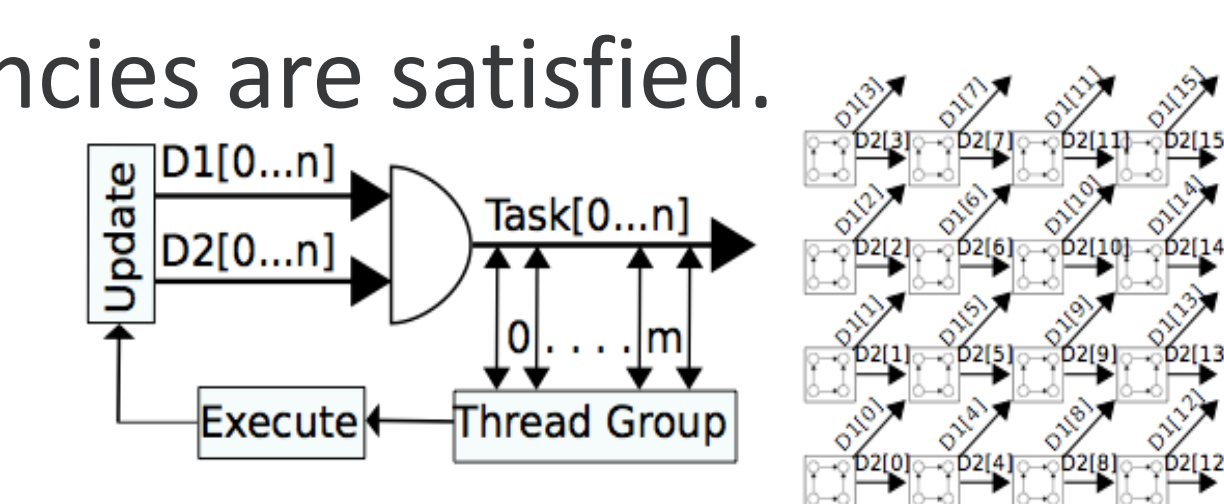
Group Locality

Threads collaborate to improve locality using static information gained from the compiler.



The Fine Grain Execution Runtime

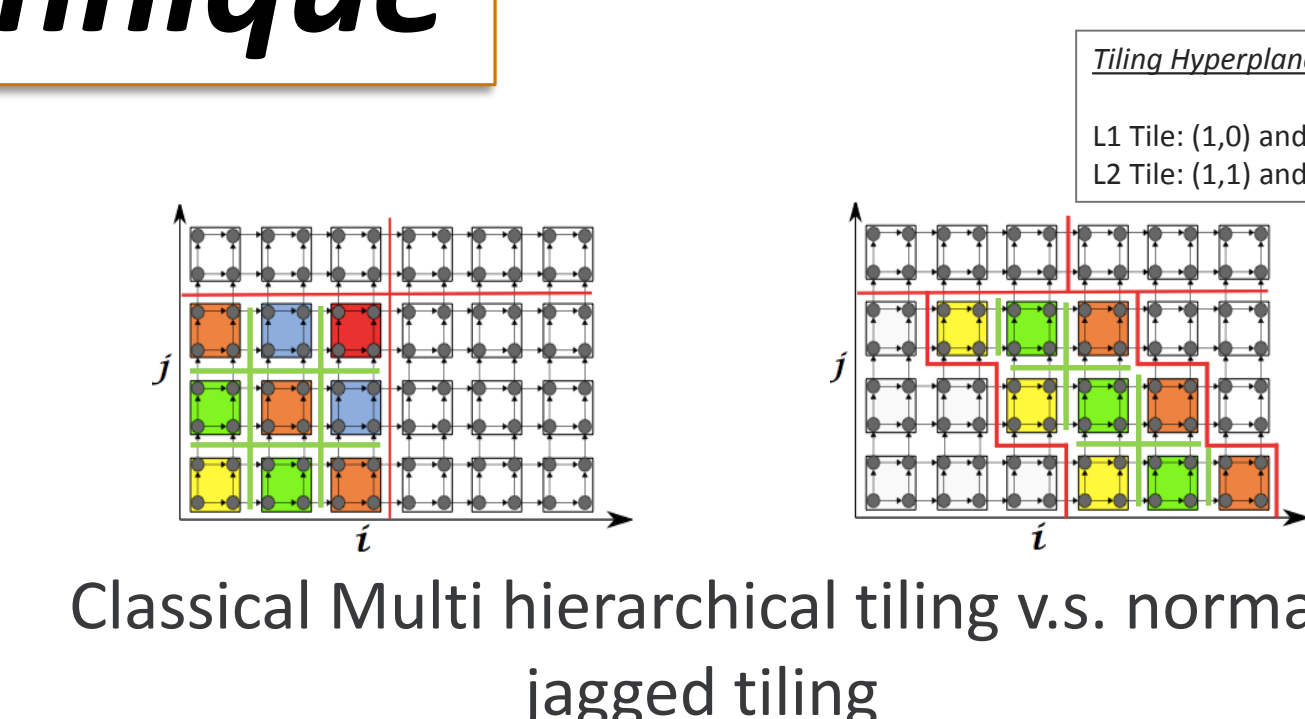
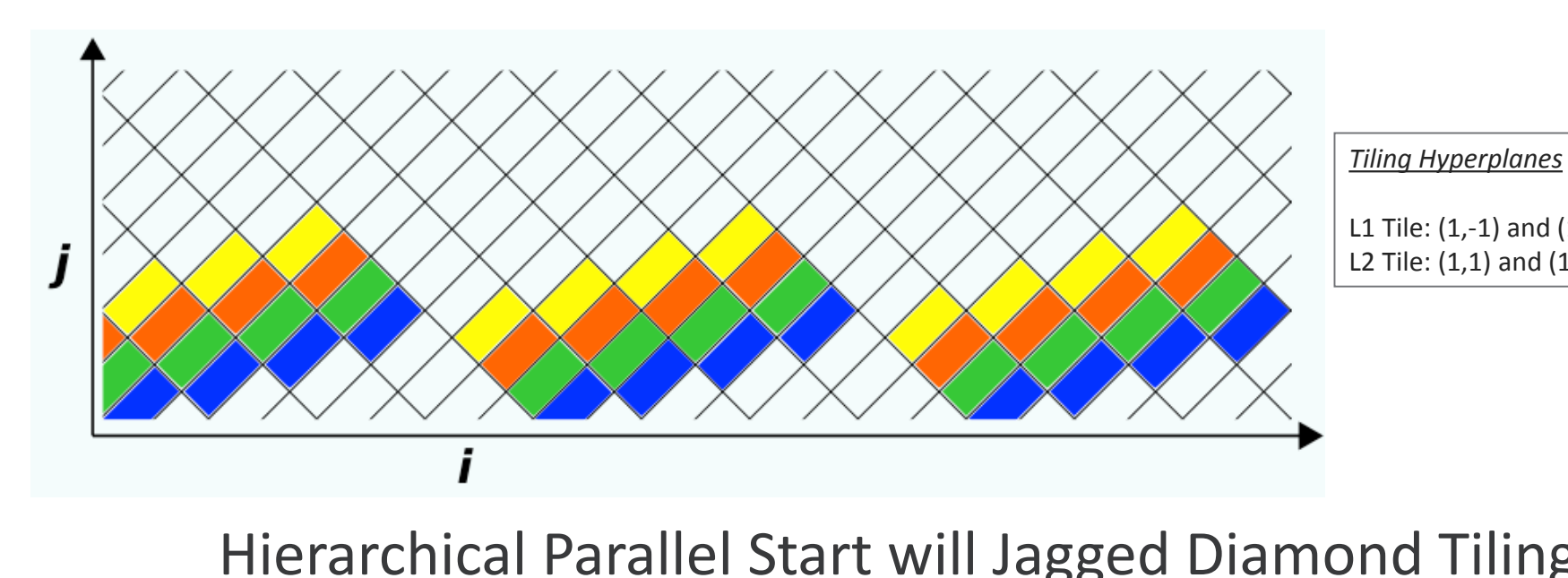
- Dependencies and tasks (i.e. tiles) are represented by bit masks.
- Dependency bit mask are AND'ed to get task mask.
- Inner tiles run in parallel as dependencies are satisfied.
- Take advantage of outer tile locality.
- Hierarchical



An Efficient Tiling Technique

The Jagged Tiling technique

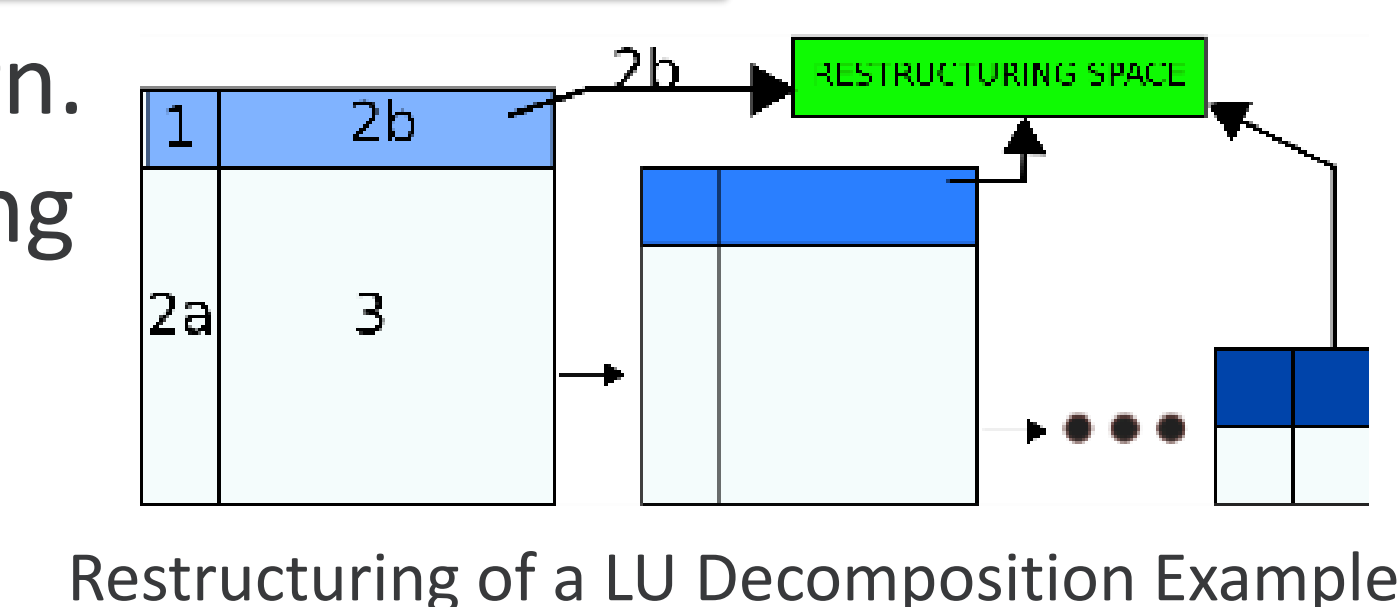
- Designed to improve intratile parallelism.
- Uses Polyhedral formulation to create the different tiling levels



- Two flavors:
 - **Jagged Tiling** (pipeline start codes) & **Jagged Diamond Tiling** (parallel start code)

The Data Restructuring Framework

- Mem. access latency dependent on access pattern.
- Access pattern = Access Function + Mem. Mapping
- Based on Reuse of Data
- Mapping of one address space to the other using invertible transformation matrix.



Initial Results

Experimental Platform:

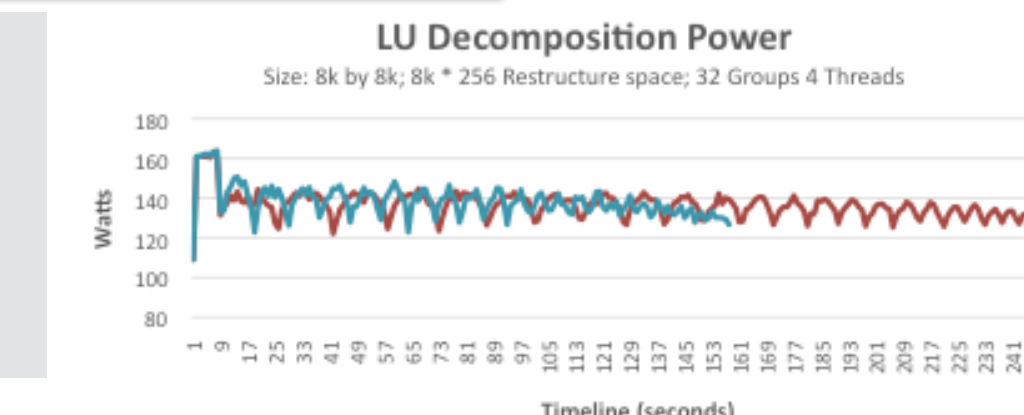
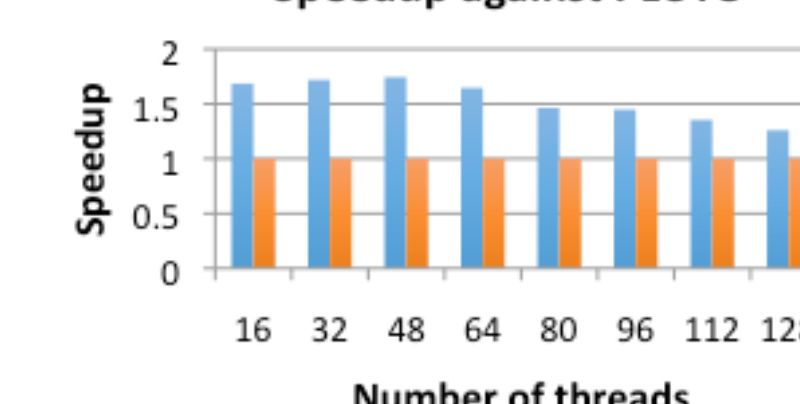
Intel Xeon Phi 7110P
61 Core at 1.1 GHz
L1 cache 32KB, L2 cache 512 KB

Threads per Group: 4

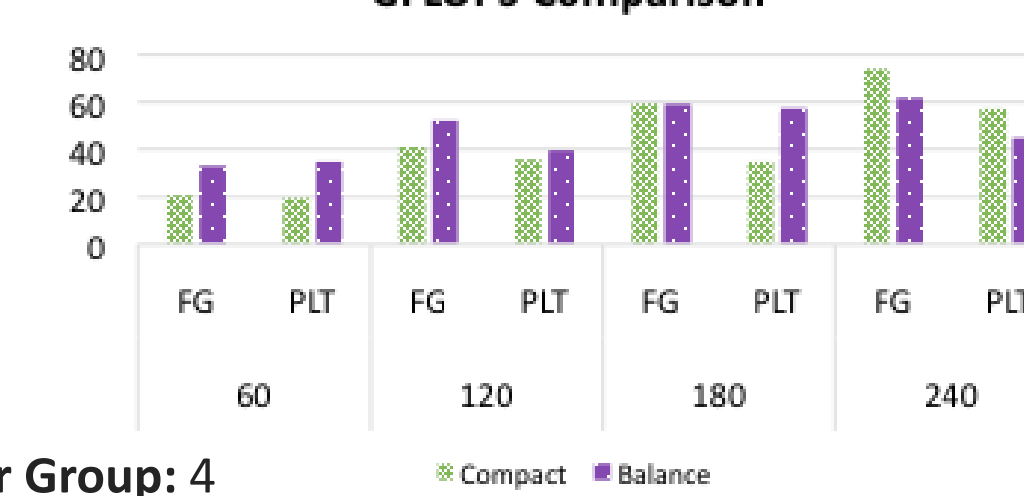
FG : Fine-grained execution

FG+RES: Fine-grained execution combined with restructuring. Restructuring is based on reuse of data.

Seidel-2D 10Kx10K 6K Times
Speedup against PLUTO



Heat 3D:
GFLOPS Comparison



FG : Fine-grained execution of tile inside L2 using Jagged Tiling
PLT: PLUTO generated tiled OpenMP code.

Conclusions

- Increase performance by better utilization of resources
- Overall better energy profiles due to reduction in time and a negligible changes in power consumption
- Future work includes more in-depth characterization & memory structure mapping