



Flexible, Efficient Abstractions for High Performance Computation on Current and Emerging Architectures

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Motivation

- \cong Complex physics \Rightarrow complex software!
 - is this necessary, or just a result of looking at the problem in the wrong way?
- Changing from model "A" to model "B" ...
 - may require different transport equations
 - may introduce different nonlinear coupling
- Spatial discretization frequently permeates software design
 - Model developers typically must deal with "mesh loops"
 - often resort to "copy/paste/modify" tactics that are highly bug-prone
 - Future proofing:
 - What if you want to do OpenMP on these loops?
 - What happens when you learn that OpenMP is not the right tool?
 - pthreads, CUDA / OpenCL ... ?
- Questions: Can we write efficient software that...
 - ... naturally handles complexity and allows us to easily extend/replace existing models?
 - ... allows programmers to easily and robustly express intent while not worrying about "details?"
 - ... allows us to refactor for different hardware architectures without rewriting the code base?

Flexible...

Register all expressions

- Each "expression" calculates one or more field quantities.
- Each expression advertises its direct dependencies.

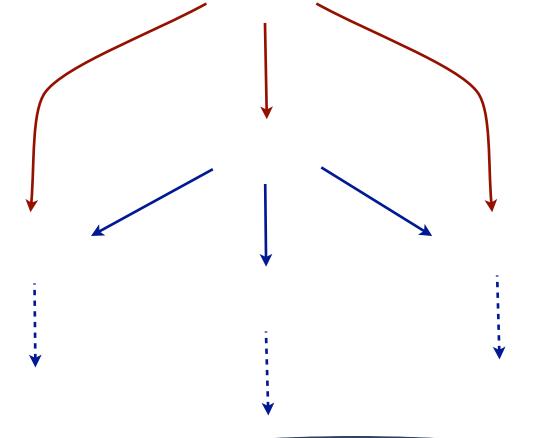
Set a "root" expression; construct a graph

- All dependencies are discovered/resolved automatically.
- Highly localized influence of changes in models.
- Not all expressions in the registry may be relevant/used.

From the graph:

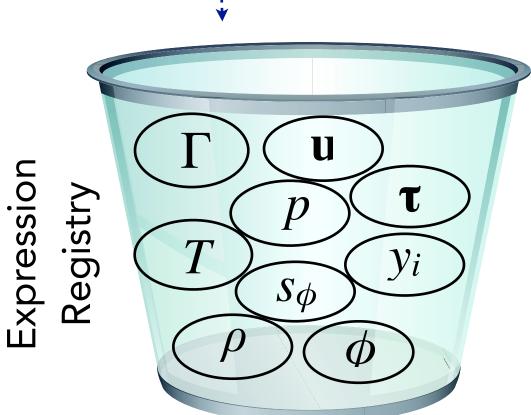
- Deduce storage requirements & allocate memory (externally to each expression).
- Automatically schedule evaluation, ensuring proper ordering.
- Asynchronous execution is critical! (overlap communication & computation)
- Robust scheduling algorithms are key.

$$\Gamma = \Gamma(T, p, y_i)$$



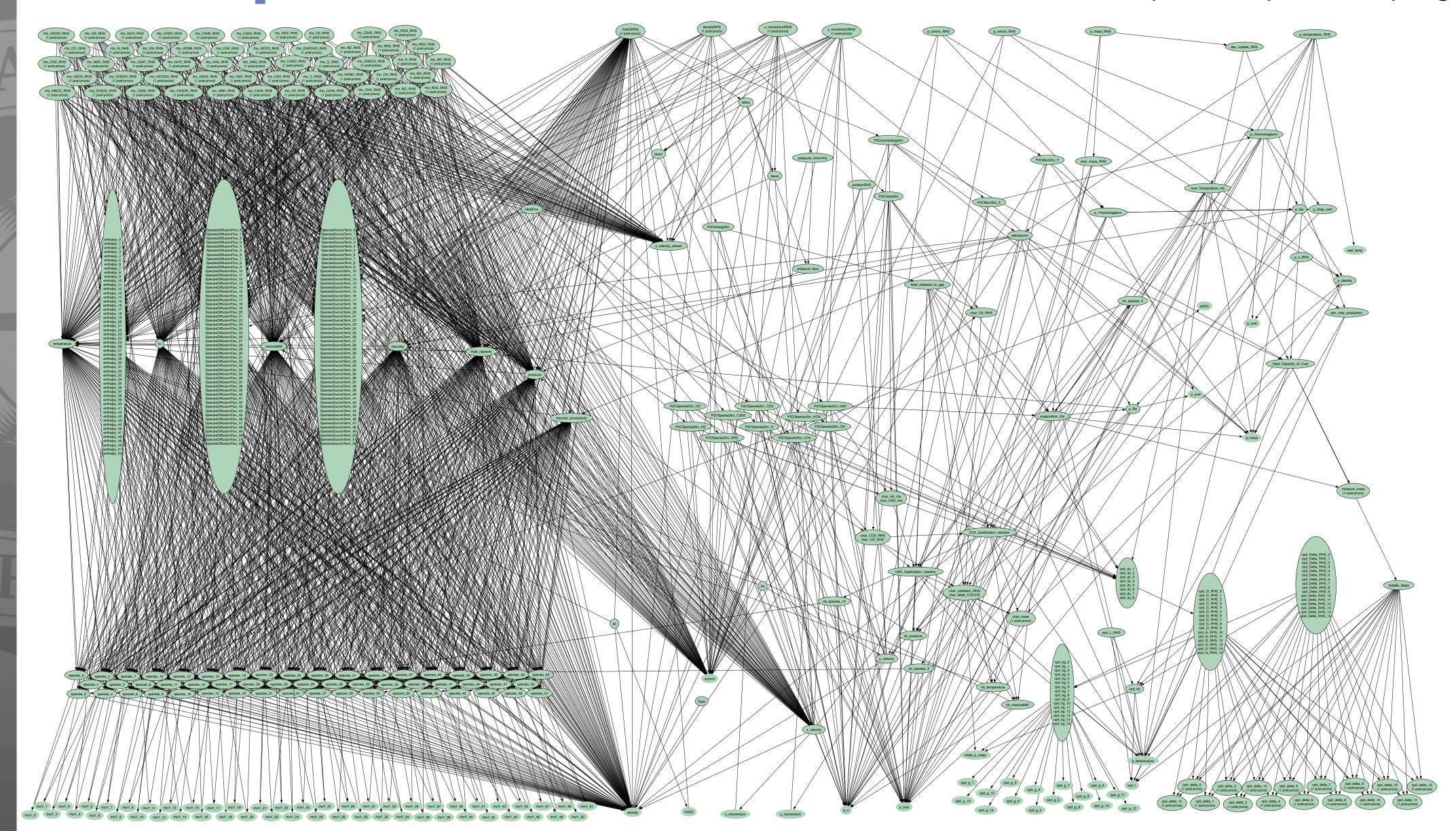
Direct (expressed) dependencies.

Indirect (discovered) dependencies.



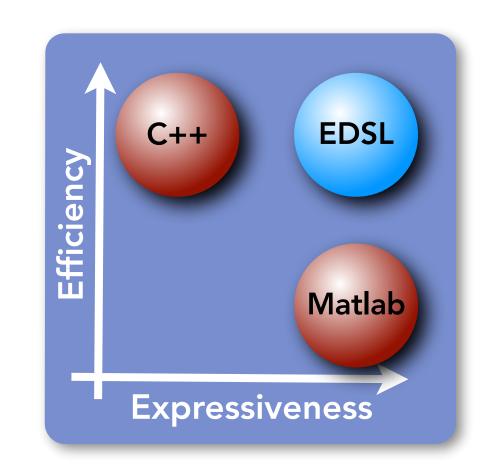
Example: coal combustion : ~35 ODEs per particle . Complex interphase coupling

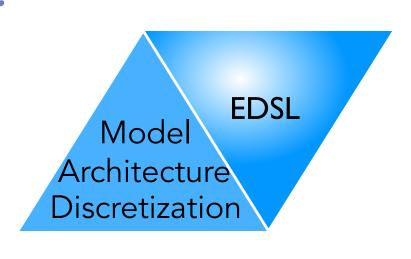
- 55 PDEs



Efficient...

- Expressive syntax (matlab-style array operations)
 - Programmer expresses intent (problem structure) not implementation.
- High performance
 - Should match hand-tuned code in performance.
- Extensible
 - Insulate programmer from architecture changes (e.g. multicore
 → GPU → ...).
 - EDSL "back-end" compiles into code for target architecture.
- "Plays well with others"
 - Allow programmer to write in C++ and inter-operate with EDSL.
 - Not an "all-or-none" approach: enable incremental adoption.
 - Allows concurrent development of EDSL and application codes.





Field Expressions

$$\vec{c} = \vec{a} + sin(\vec{b})$$



Manual C++

Thread 1

_

Thread n

```
Field::const_iterator ia1 = a1.begin();
Field::const_iterator ib1 = b1.begin();
for(Field::iterator ic1 = c1.begin();
    ic1 != c1.end();
    ++ic1, ++ia1, ++ib1) {
    *ic1 = *ia1 + sin(*ib1);
};
...
Field::const_iterator ian = an.begin();
Field::const_iterator ibn = bn.begin();
for(Field::iterator icn = cn.begin();
    icn != cn.end();
    ++icn, ++ian, ++ibn) {
    *icn = *ian + sin(*ibn);
};
```

Nebo EDSL

- Data parallel handled internally.
- Thread deployment (resizable threadpool).
- GPU deployment.
- Compile-time guarantee of field compatibility for given operations.

Chained Stencil Operations

```
\phi = -\nabla \cdot \mathbf{q}= \nabla \cdot (\lambda \nabla T)
```

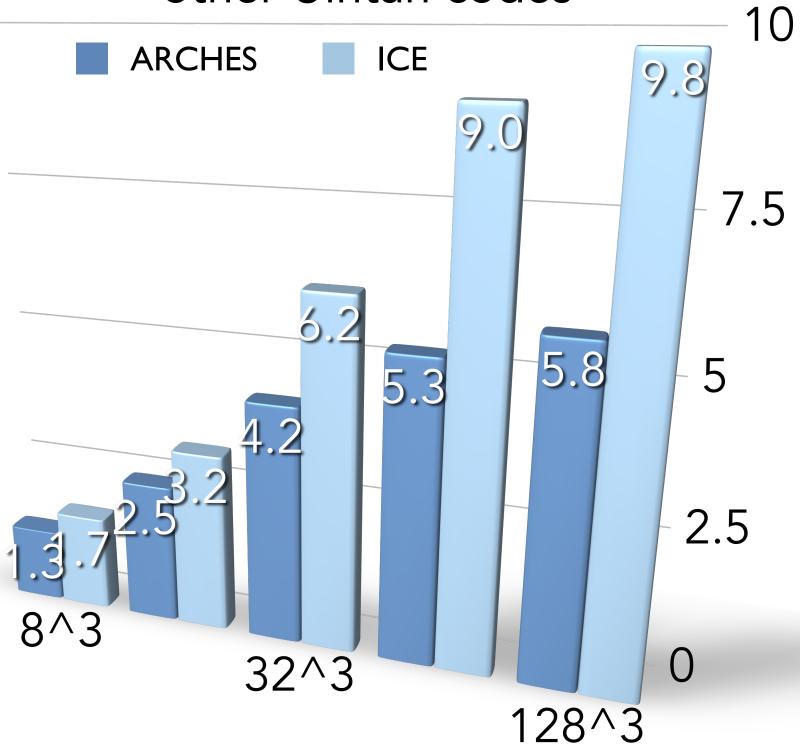
```
// field type inference:
typedef FaceTypes<FieldT>::XFace XFluxT;
typedef FaceTypes<FieldT>::YFace YFluxT;
typedef FaceTypes<FieldT>::ZFace ZFluxT;

// operator type inference:
typedef OpTypes<FieldT>::DivX DivX;
typedef OpTypes<FieldT>::DivY DivY;
typedef OpTypes<FieldT>::DivZ DivZ;
```

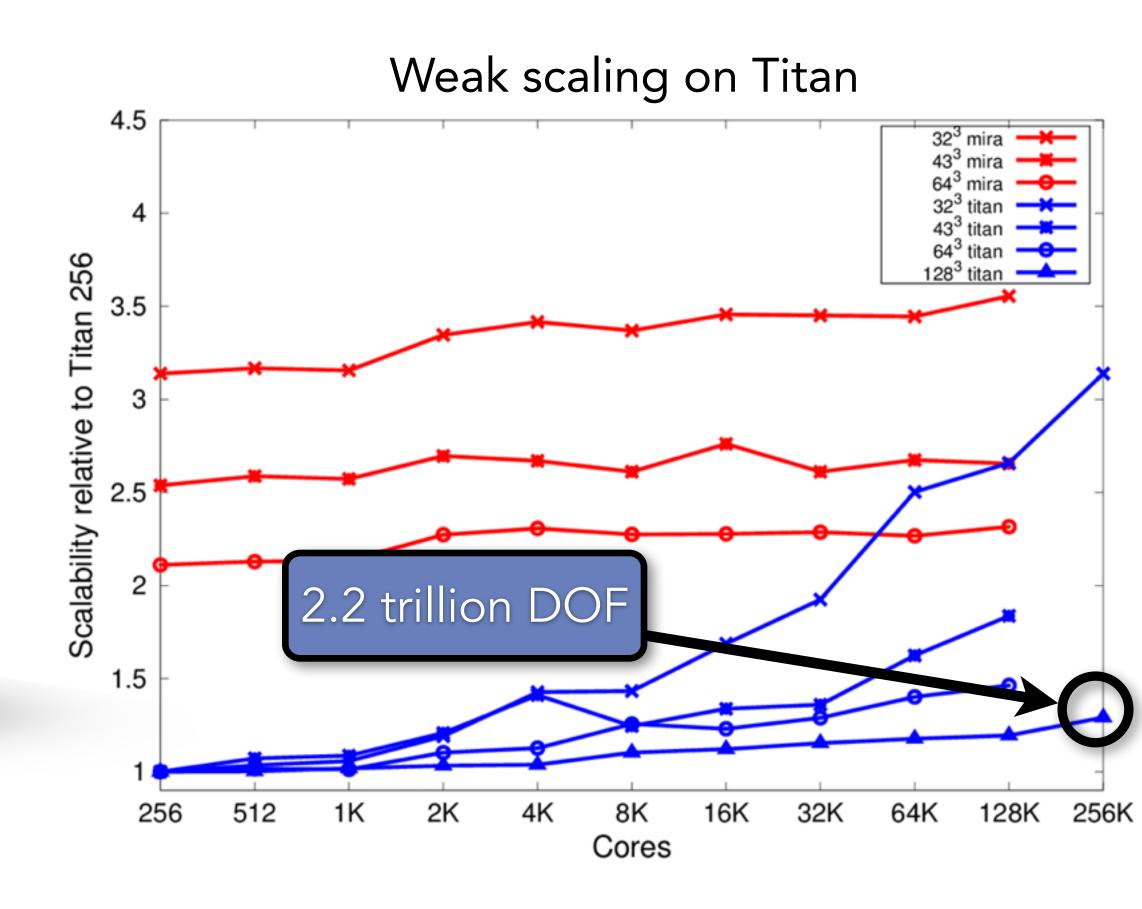
- One inlined grid loop, no temporaries.
- Better performance & scalability than without chaining.
- Compile-time consistency checking (field-operator and field-field compatibility).
- Runtime consistency checks for ghost cell validity.

Putting it Together: Performance & Scalability

Speedup using DSL* relative to other Uintah codes



*Comparison to ICE and ARCHES, sister codes in Uintah, on a 3D Taylor-Green vortex problem. Run on a single processor.



"1" indicates perfect weak scaling

Multicore & GPU Performance

Test: mockup of a diffusion-reaction problem.

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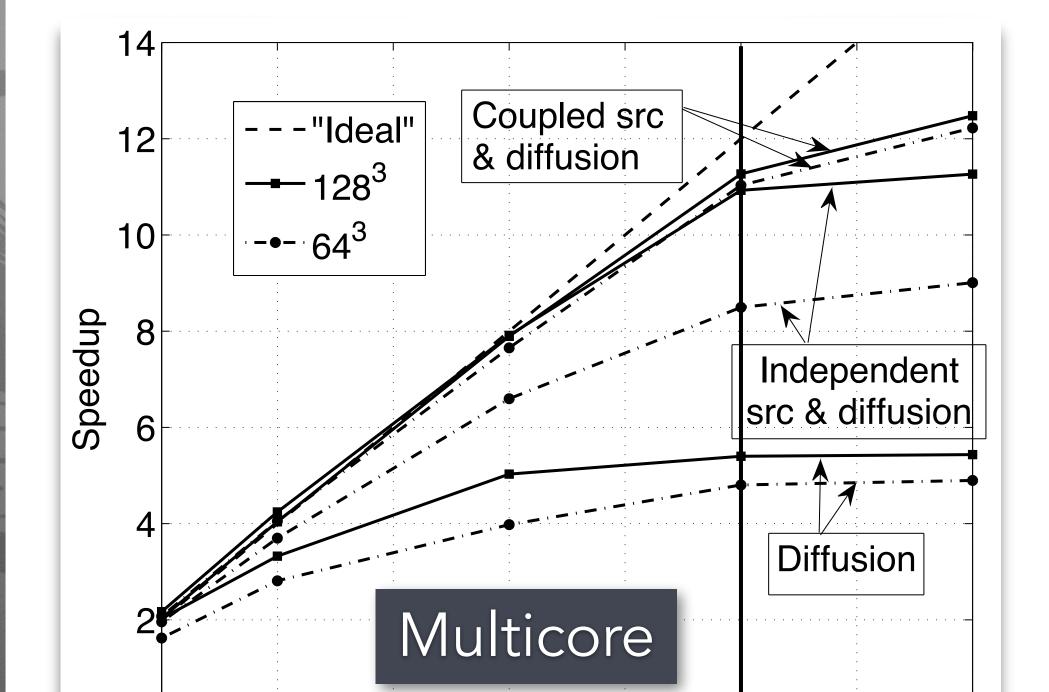
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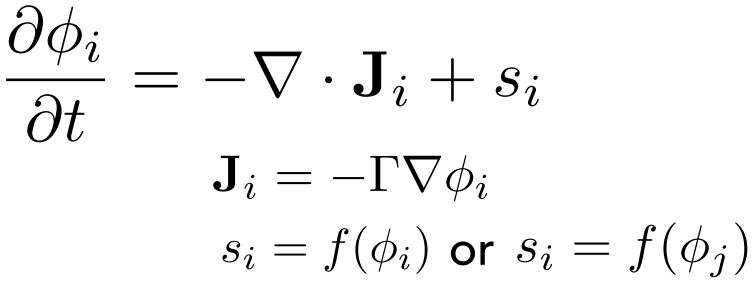
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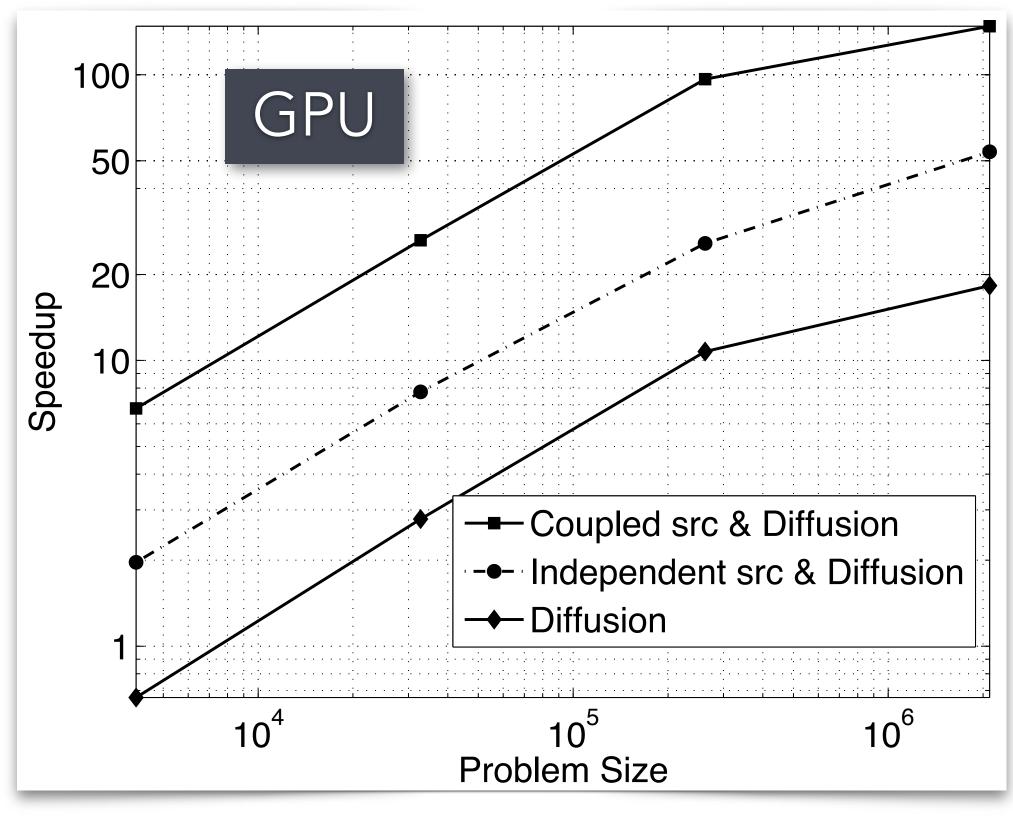
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Thread Count

- Easily dial in the number of equations (30 here).
- Diffusion is an inexpensive stencil calculation.
- Reaction is an expensive point-wise calculation.







Parting Thoughts

Hierarchical parallelization allows for flexible usage of available resources:

- Domain decomposition (SIMD)
 - Should allow a process to do computation on "interior" while waiting on communication from neighbors.
- Task decomposition (MIMD)
 - Decompose the solution into a DAG that can be scheduled asynchronously.
- Vectorized parallel (SIMD)
 - Break grid operations across multicore, GPU, etc.

DAG representation is a scalable abstraction that:

- Handles problem complexity gracefully.
- Provides convenient separation of the problem's structure from the data.
- Allows sophisticated scheduling algorithms to optimize scalability & performance.

(E)DSLs are very useful

- Future-proofing: separate intent from implementation.
- EDSLs allow seamless transition of a code base and leverage existing compilers.
- Template metaprogramming pushes work from run-time to compile-time for more efficiency.