

Research Review 2021

Spiral AI/ML: Co-optimization for High-Performance, Data-Intensive Computing in Resource-Constrained Environments

November 2021

Scott McMillan

Software Engineering Institute
Carnegie Mellon University
Pittsburgh, PA 15213

[DISTRIBUTION STATEMENT A] This material has been approved for public release and unlimited distribution.

Copyright 2021 Carnegie Mellon University.

This material is based upon work funded and supported by the Department of Defense under Contract No. FA8702-15-D-0002 with Carnegie Mellon University for the operation of the Software Engineering Institute, a federally funded research and development center.

The view, opinions, and/or findings contained in this material are those of the author(s) and should not be construed as an official Government position, policy, or decision, unless designated by other documentation.

NO WARRANTY. THIS CARNEGIE MELLON UNIVERSITY AND SOFTWARE ENGINEERING INSTITUTE MATERIAL IS FURNISHED ON AN "AS-IS" BASIS. CARNEGIE MELLON UNIVERSITY MAKES NO WARRANTIES OF ANY KIND, EITHER EXPRESSED OR IMPLIED, AS TO ANY MATTER INCLUDING, BUT NOT LIMITED TO, WARRANTY OF FITNESS FOR PURPOSE OR MERCHANTABILITY, EXCLUSIVITY, OR RESULTS OBTAINED FROM USE OF THE MATERIAL. CARNEGIE MELLON UNIVERSITY DOES NOT MAKE ANY WARRANTY OF ANY KIND WITH RESPECT TO FREEDOM FROM PATENT, TRADEMARK, OR COPYRIGHT INFRINGEMENT.

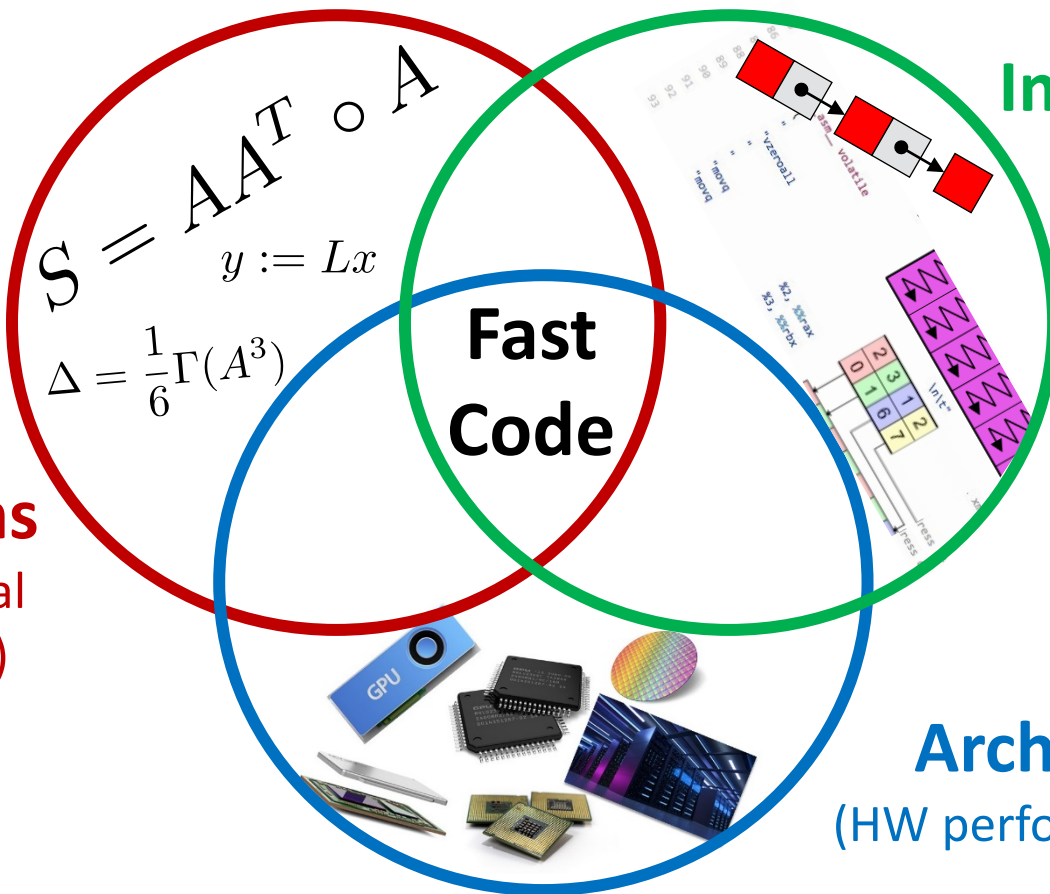
[DISTRIBUTION STATEMENT A] This material has been approved for public release and unlimited distribution. Please see Copyright notice for non-US Government use and distribution.

This material may be reproduced in its entirety, without modification, and freely distributed in written or electronic form without requesting formal permission. Permission is required for any other use. Requests for permission should be directed to the Software Engineering Institute at permission@sei.cmu.edu.

DM21-0887

Fast code depends on three interdependent ingredients

Algorithms
(mathematical
descriptions)



Spiral/AIML: Co-Optimization for High-Performance, Data-Intensive Computing in Resource-Constrained Environments



“Rapidly delivering artificial intelligence to a combat zone won’t be easy.” Col. Drew Cukor, USMC.

Problem(s)

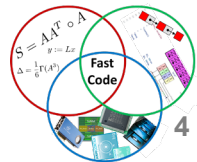
- Increasing complexity in computing architectures.
- Mission cost, size, weight, and power (CSWAP) constraints drive increasing use of FPGAs and ASICs (more complexity).
- Achieving performance from these platforms is *hard*.
- Achieving performance from **data-intensive applications** (graphs, ML, AI) is *hard*.

Solution

- Automatic code generation for data-intensive computations.
- Simultaneous, automatic co-optimization of hardware within CSWAP constraints.

Approach

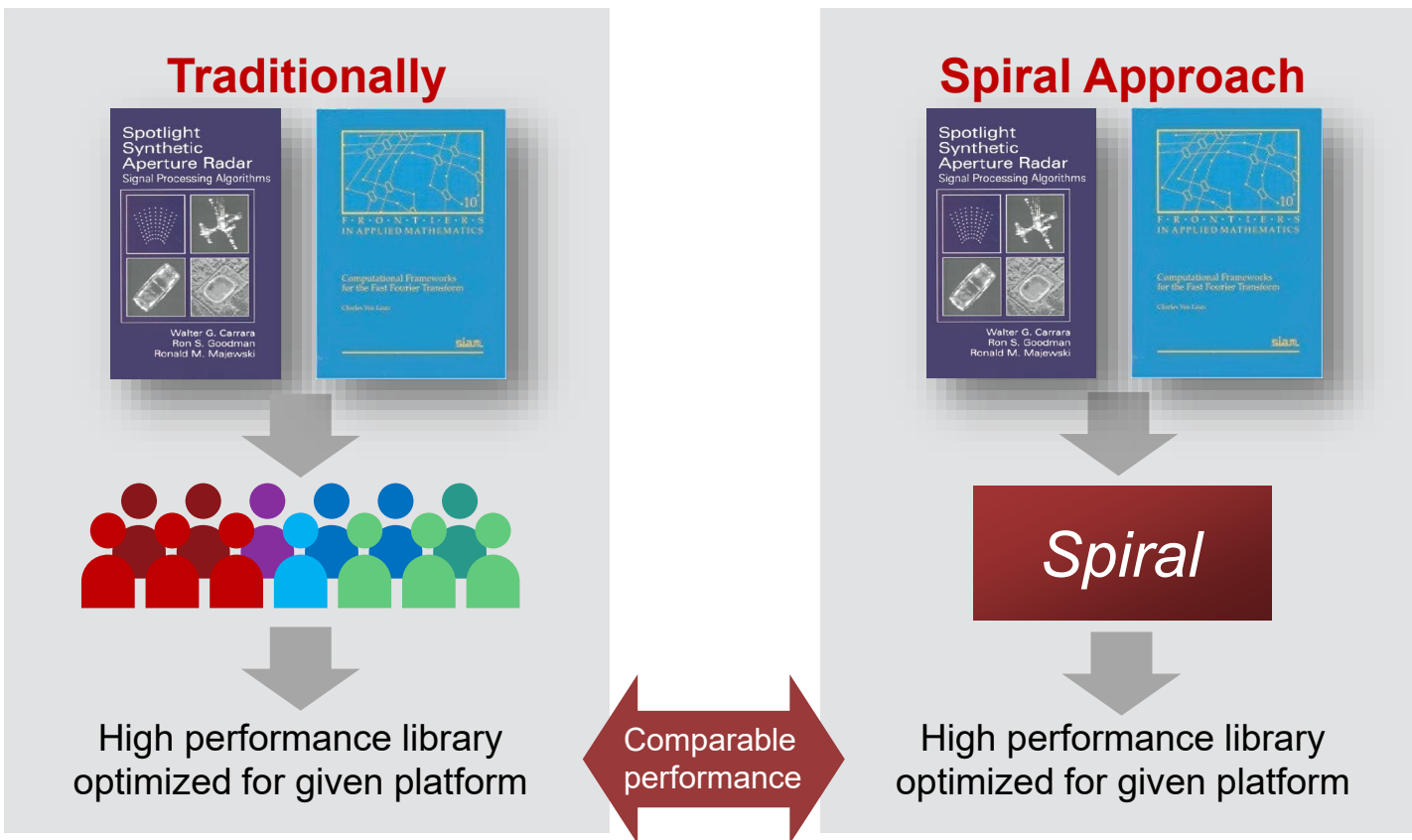
- Identify common AI/ML/Graph computational primitives and their **mathematical descriptions**.
- Develop **hardware performance models** allowing Spiral to choose between components satisfying CSWAP requirements.
- Encode knowledge about **efficient implementations** of graph, ML, and AI computational primitives into **Spiral code-generation technology**.



What is Spiral?

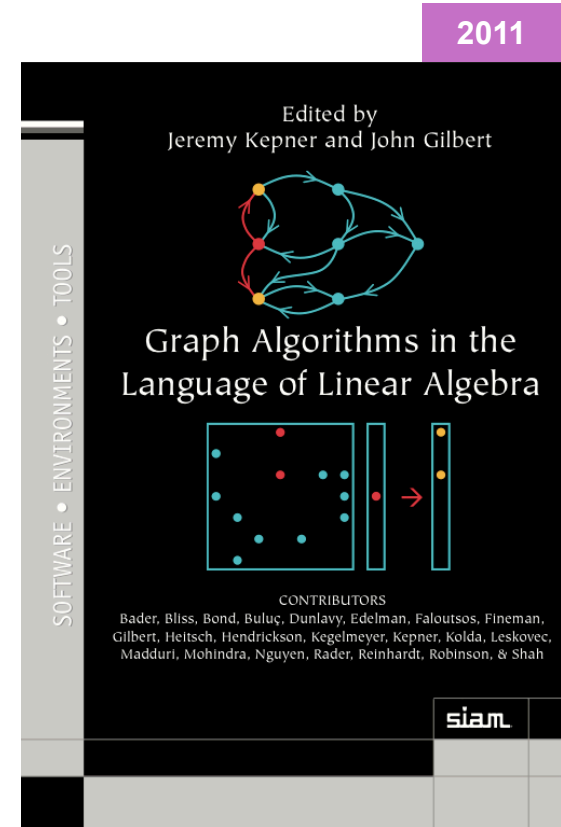


What is Spiral?



Previous Work Addressed Spiral for Graphs

- Spiral understands the rules of linear algebra.
- Much research in graph algorithms using linear algebra
- However, it involves *sparse* and *irregular* data.
- FY2016-2018: working to “teach” the data structures, operations, and optimizations used in graph algorithms
 - Needed a mathematical specification of common operations
 - Led to involvement in the **GraphBLAS** community



Foundational GraphBLAS References

Mathematical Foundations of the GraphBLAS

IEEE HPEC 2016

Jeremy Kepner (MIT Lincoln Laboratory Supercomputing Center), Peter Aaltonen (Indiana University), David Bader (Georgia Institute of Technology), Aydın Buluç (Lawrence Berkeley National Laboratory), Franz Franchetti (Carnegie Mellon University), John Gilbert (University of California, Santa Barbara), Dylan Hutchison (University of Washington), Manoj Kumar (IBM), Andrew Lumsdaine (Indiana University), Henning Meyerhenke (Karlsruhe Institute of Technology), [Scott McMillan](#) (CMU Software Engineering Institute), Jose Moreira (IBM), John D. Owens (University of California, Davis), Carl Yang (University of California, Davis), Marcin Zalewski (Indiana University), Timothy Mattson (Intel)

Design of the GraphBLAS API for C

IEEE HPEC 2017

Aydın Buluç[†], Tim Mattson[‡], [Scott McMillan](#)[§], José Moreira[¶], Carl Yang^{*,†}

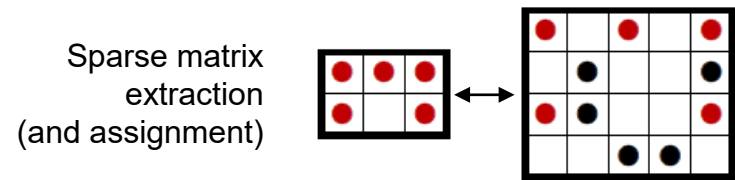
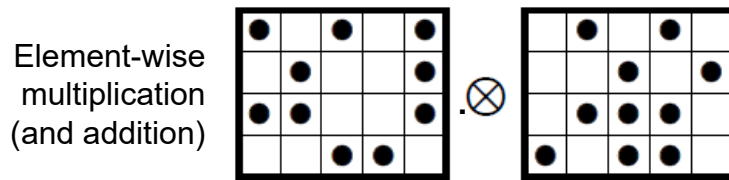
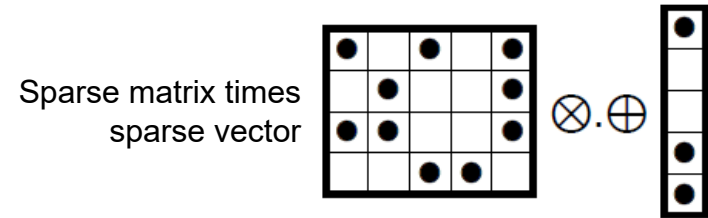
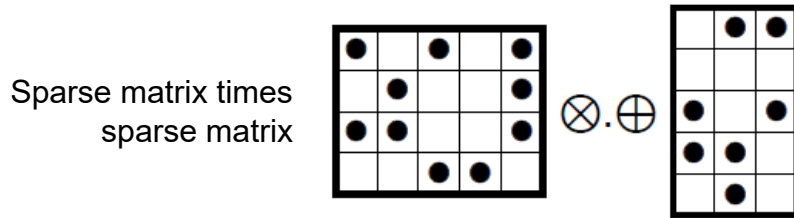
[†]*Computational Research Division, Lawrence Berkeley National Laboratory*

[‡]*Intel Corporation* [§]*Software Engineering Institute, Carnegie Mellon University* [¶]*IBM Corporation*

^{*}*Electrical and Computer Engineering Department, University of California, Davis, USA*

GraphBLAS Primitives

- Basic objects (opaque types)
 - Matrices (sparse or dense), vectors (sparse or dense), algebraic operators (semirings)
- Fundamental operations over these objects



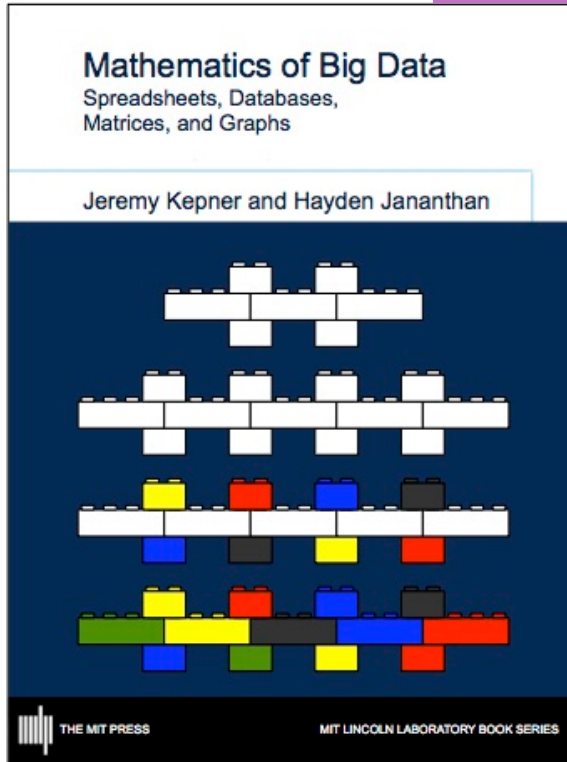
...plus reduction, transpose, Kronecker product, filtering, transform, etc.

Buluc, A.; Mattson, T.; McMillan, S.; Moreira, J.; & Yang, C. "The GraphBLAS C API Specification," v1.0 May 2017, v1.1 May 2018, v1.3 Sep 2019. **v2.0 Nov 2021 (at SC21)**. <http://graphblas.org>

GraphBLAS Is Not Just for Graphs

- It is a general linear algebra specification with a rich set of operations on any type of matrix and vector data.
- Recent work has shown applicability of GraphBLAS for operations on data in tabular form.
- **AI, specifically ML, workloads use both sparse and dense data...**

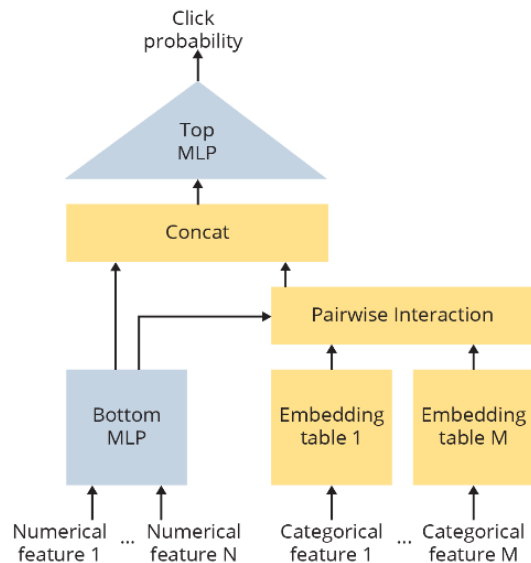
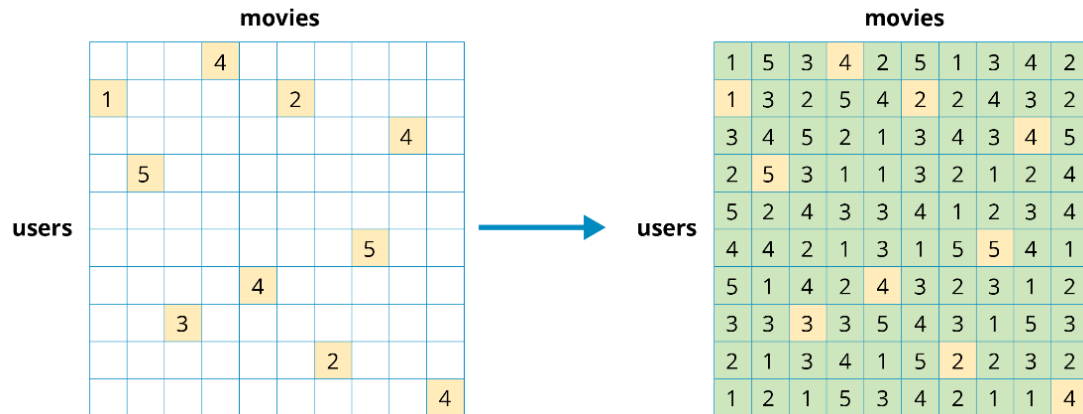
2018



GraphBLAS Operations Can Also Perform ML

- Recommender Systems predict broad information from sparse data
- Deep Learning Recommendation Models (DLRM) has a mixed sparse-dense workloads

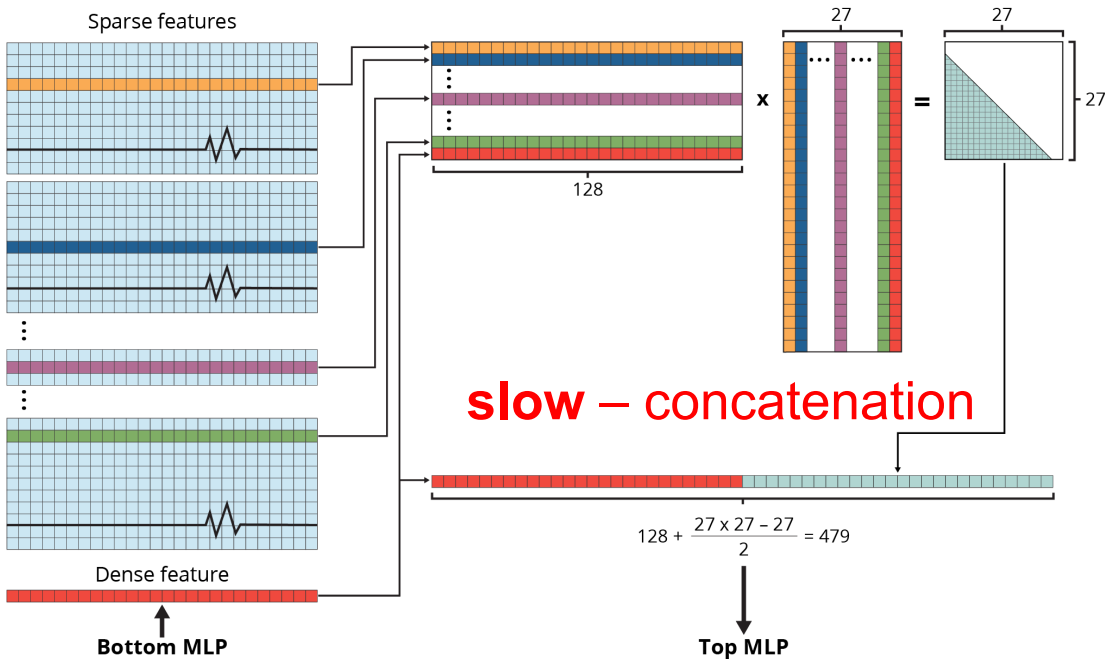
RS model fills in the empty spots



Deep Learning Recommendation Model (DLRM)

Co-Funded by DARPA SDH

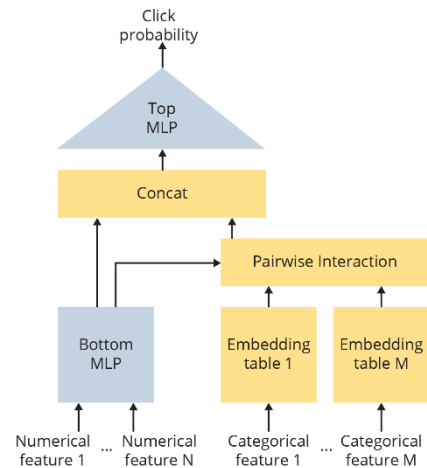
slow –
DRAM
Access



fast –
Vector Math

fast –
Vector Math

fast –
Vector Math



FY19-21: Adding More GraphBLAS for ML to Spiral

Three primary lines of R&D

- **Algorithms**

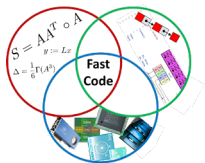
- Analyzing graph, ML and AI workloads
- Identifying computational kernels
- Developing data structures

- **Implementations**

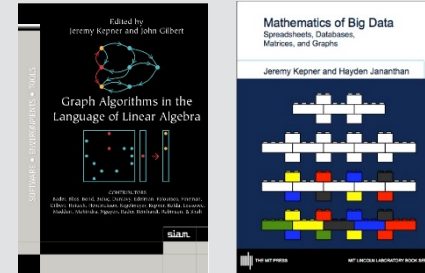
- Expanding the GraphBLAS API
- Optimizing GraphBLAS Template Library
- Developing the Graph Kernel Collection

- **Architectures**

- Developing performance models for GPU and multi-core CPU



Spiral Approach



Spiral

High performance library
optimized for given platform

SPIRAL is an Advanced Source-to-Source Compiler

Compiler Analogies

AST

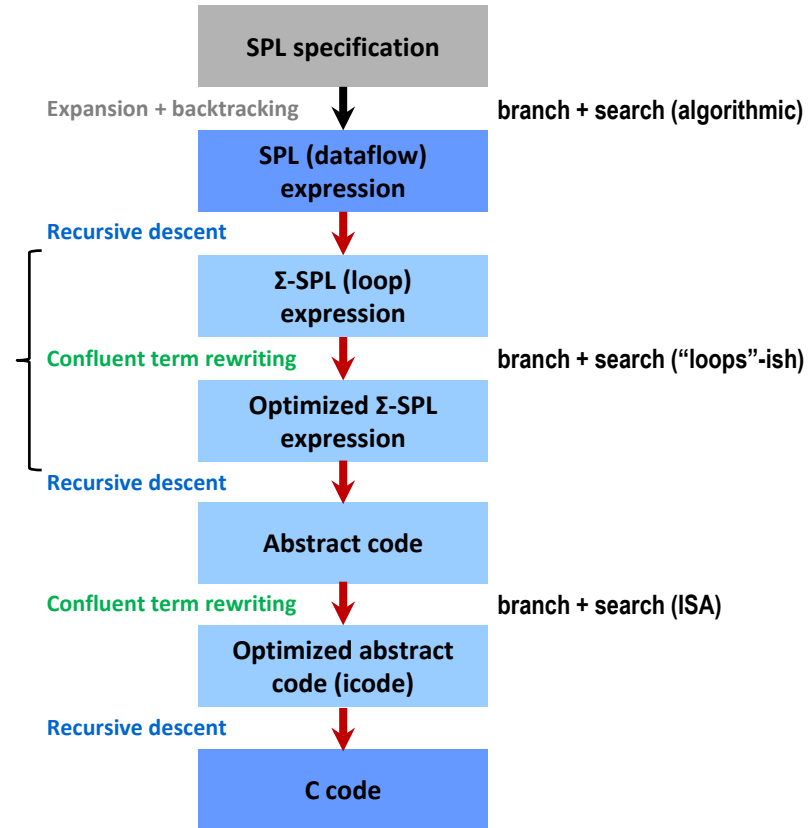
Finest grained AST

algorithm breakdown, and some architecture-specific features; e.g., SIMD vs. warp, number of threads, etc.

Optimization Phase

IR
of a compiler

IR
mapping to HW specific instruction sets



SPIRAL is an Advanced Source-to-Source Compiler

Mathematical Descriptions

Compiler Analogies

AST

Finest grained AST

algorithm breakdown, and some architecture-specific features; e.g., SIMD vs. warp, number of threads, etc

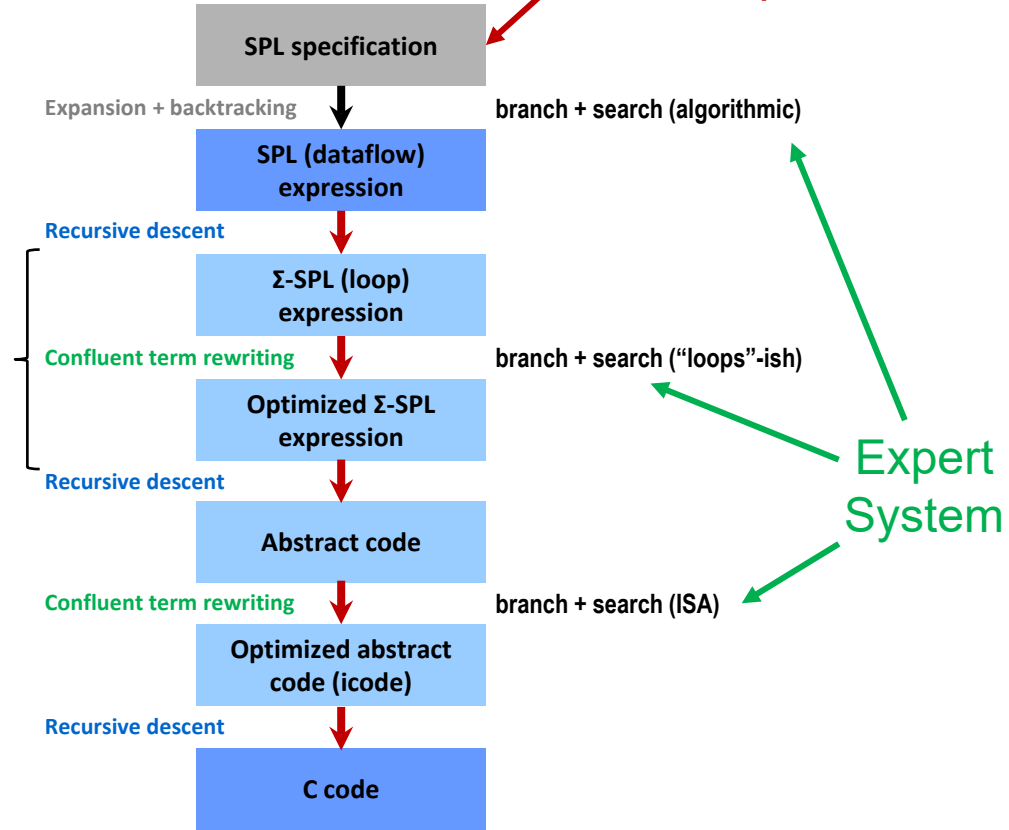
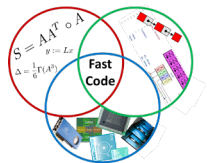
Optimization Phase

Performance Models

IR of a compiler

IR

mapping to HW specific instruction sets



SPIRAL: Code Generation Stages/Additions

AST

Additions

SPL specification

Parsing GraphBLAS syntax. Support for new sparse data formats, semiring operators, and sizes specified at run-time.

Expansion + backtracking

branch + search (algorithmic)

SPL (dataflow) expression

Computation graphs (DAGs) of operations to perform optimizations (GBTLX, Oct20)

Recursive descent

Σ -SPL (loop) expression

Confluent term rewriting

branch + search (loops)

Optimized Σ -SPL expression

New representation for bilinear (binary, even n-ary) operators allowing for mxm, mxv, ewise, apply, etc.

Recursive descent

Abstract code

Confluent term rewriting

branch + search (ISA)

Optimized abstract code (icode)

Generated kernels for sparse linear algebra applications

Recursive descent

C code

Finest grained AST algorithm breakdown, and some architecture-specific features; e.g., SIMD vs. warp, number of threads, etc.

Optimization Phase

IR of a compiler

IR mapping to HW-specific instruction sets

Artifact Highlights: Algorithms

Publications and Implementations

1. **LAGraph: a community effort to curate algorithms built on top of GraphBLAS (IPDPSW'19)**
2. Linear algebraic formulation of edge-centric k-truss algorithms with adjacency matrices (HEPC'18)
3. Graph signal processing and deep learning (Signals'20)
4. Delta-stepping SSSP: from vertices and edges to GraphBLAS implementations (IPDPSW'19)
5. Linear algebraic depth-first search (ARRAY'19)
6. Linear algebraic Louvain clustering algorithm (IPDPSW'20)
7. Edge entropy as an indicator of effectiveness of Graph Neural Networks for node classification (SSC'20)
8. Bipartite graph generation using Kronecker product (IPDPSW'20)
9. **The design of LAGraph (SIAM CSE'21)**
10. **LAGraph: Linear Algebra, Network Analysis Libraries, and the Study of Graph Algorithms (IPDPSW'21)**
11. GraphSAINT, graph neural networks
12. Deep Learning Recommender Models
13. GAP Benchmark algorithms
14. Triangle Centrality analysis

LAGraph: Linear Algebra, Network Analysis Libraries, and the Study of Graph Algorithms

Gábor Szárnyas*, David A. Bader[†], Timothy A. Davis[‡], James Kitchen[§],
Timothy G. Mattson[¶], Scott McMillan^{||}, Erik Welch[§]

*CWI Amsterdam [†]New Jersey Institute of Technology [‡]Texas A&M University [§]Anaconda, Inc.
[¶]Intel Corp. ^{||}Software Engineering Institute, Carnegie Mellon University

Download and/or contribute code at:

<https://github.com/GraphBLAS/LAGraph>

Artifact Highlights: Implementations

APIs:

15. GraphBLAS C API Spec v1.3 (9/19)
16. GraphBLAS C API Spec v2.0 for multithreading (IPDPSW'21)
17. GraphBLAS C++ API Spec (draft, IPDPSW'20)
18. Distributed GraphBLAS API (IPDPSW'20)
19. Spiral/GBTLX source-to-source translator, (HPEC'20)

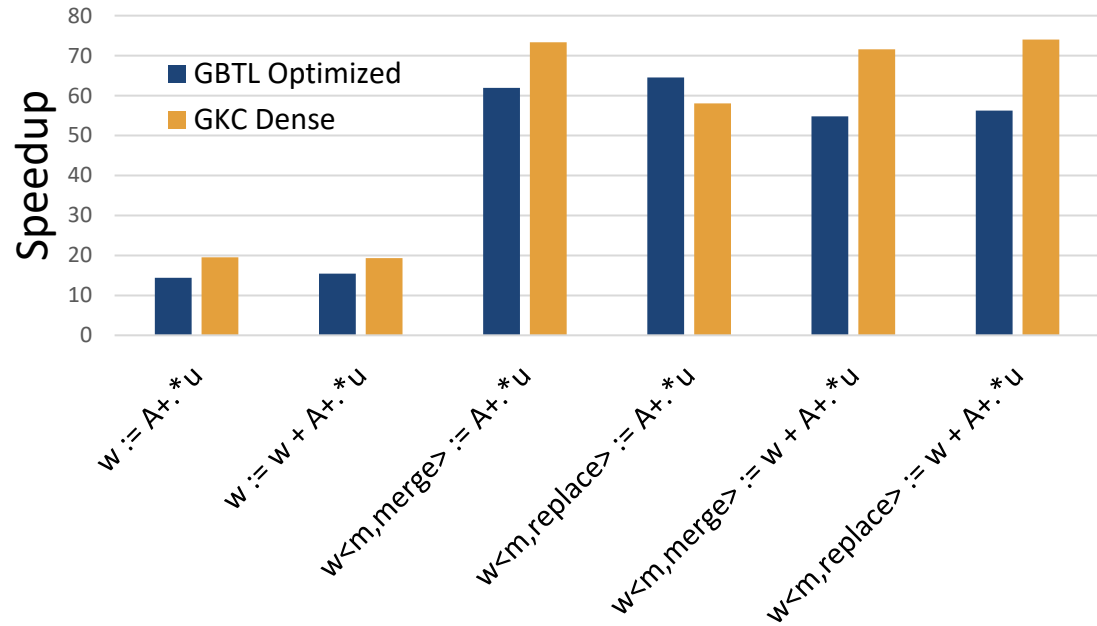
Open-source code:

20. [GraphBLAS Template Library \(GBTL\) v3.0](#)
21. [Graph Kernel Collection](#)
22. LAGraph Algorithms Repository, v1.0: a community effort to curate algorithms built on top of GraphBLAS
23. NWGraph: A Library of Generic Graph Algorithms and Data Structure in C++20 (with PNNL/UW)

Tutorials:

- “A Hands-on Introduction to GraphBLAS”
 24. Tutorial in C (HPEC'18, '19)
 25. Tutorial in python (SIAM CSE'21, ICS'21, HPEC'21)

Speedup of GBTL++ and GKC SpMV over GBTL v3.0

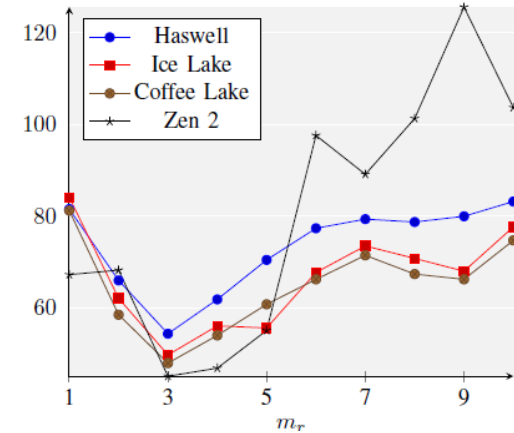


Artifact Highlights: Architectures

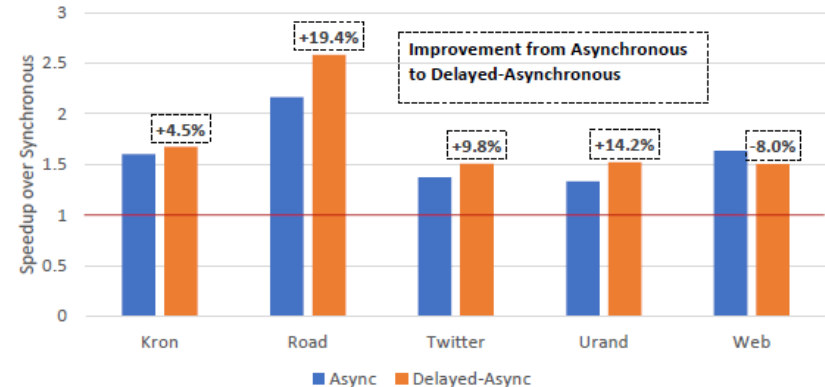
Analyses and Publications:

26. PageRank acceleration for large graphs with scalable hardware (FPGA) and 2-step SpMV (HPEC'18)
27. Preliminary exploration of large-scale triangle counting on shared-memory, multicore systems (HPEC'18)
28. Exploration of fine-grained parallelism for load balancing eager k-truss on GPU and CPU (HPEC'19)
29. Efficient SpMV operation for large and highly sparse matrices using scalable multi-way merge parallelization (Micro'19)
30. FESIA: Fast, efficient set intersection approach on modern CPUs (ICDE'20)
31. Toward an objective metric for exact triangle count (HPEC'20)
32. Triangle counting with cyclic distribution (HPEC'20)
33. Evaluation of graph analytic frameworks using the GAP benchmark suite (IISWC'20)
- 34. Analytical models for portable Deep Learning Recommender Model (DLRM) implementations (manuscript)**
35. A method for fusing non-element-wise layers in DNNs (HPEC'21)
36. Modeling matrix engines for portability and performance (manuscript)
- 37. Delayed asynchronous iterative graph algorithms (HPEC'21)**

Effects of Different Tile Size (m_r) for Batch Syrk Execution time (ms)



Page Rank: Async and Delayed-Async vs Synchronous, 112 Thread Cascade Lake



External Collaborators

Andrew Lumsdaine/PNNL/UW

Jesun Firoz/PNNL

Tony Liu/UW

Kevin Deweese/UW

Tim Mattson/Intel Research

Michel Pelletier/Gragegon

Ben Brock/UCB

Aydin Buluc/LBNL/UCB

Oguz Selvitopi/LBNL

James Demmel/UCB

Roger Pearce/LLNL

Geoff Sanders/LLNL

Ben Priest/LLNL

Natalia Vassilieva/Cerebras

Trevor Steil/UMinn.

Luke D'Alessandro/Indiana

Jeremy Kepner/MIT/LL

Jose Moreira/IBM

Manoj Kumar/IBM

Eric Hein/EMU/Lucata

David Bader/NJIT

Paul Burkhardt/DOD

Kaushik Velusamy/UMBC

DARPA: HIVE, SDH, DSSOC,
PAPPA, DPRIVE

Phil Ratzloff/SAS

Matt Galati/SAS

Jeffrey Vetter/ORNL

Albert-Jan Yzelman

Danielle Spampinato

Now at Huawei, Zurich

Tim Davis/TAMU

Scott Kolodziej/TAMU

Jinhao Chen/TAMU

Et al.

Roi Lipman

RedisLabs, Israel

James Kitchen/Anaconda

Eric Welch/Anaconda

Intel MKL/Graph Working Group (Russia)

Andrey Nikolaev, Alexey Kukanov, Kirill Voronin

Intel MKL/Graph Working Group

Henry Gabb, Ramesh Peri, et al.

Gabor Szarnyas/

Hungarian Academy of Sciences

External Collaborators

Andrew Lumsdaine/PNNL/UW

Jesun Firoz/PNNL

Tony Liu/UW

Kevin Deweese/UW

Tim Mattson/Intel Research

Michel Pelletier/Grapphagon

Ben Brock/UCB

Aydin Buluc/LBNL/UCB

Oguz Selvitopi/LBNL

James Demmel/UCB

Roger Pearce/LLNL

Geoff Sanders/LLNL

Ben Priest/LLNL

Natalia Vassilieva/Cerebras

Intel MKL/Graph Working Group (Russia)

Andrey Nikolaev, Alexey Kukanov, Kirill Voronin

Spiral AI/ML: Co-optimization for High-Performance, Data-Intensive Computing
© 2021 Carnegie Mellon University

Trevor Steil/UMinn.

Luke D'Alessandro/Indiana

SuiteSparse GraphBLAS Library

Available as Ubuntu package

Used for Matlab sparse

Tim Davis/TAMU

Scott Kolodziej/TAMU

Jinhao Chen/TAMU

Et al.

James Kitchen/Anaconda

Eric Welch/Anaconda

Intel MKL/Graph Working Group

Henry Gabb, Ramesh Peri, et al.

[DISTRIBUTION STATEMENT A] This material has been approved for public release and unlimited distribution.

Jeremy Kepner/MIT/LL

Jose Moreira/IBM

Manoj Kumar/IBM

Eric Hein/EMU/Lucata

David Bader/NJIT

Paul Burkhardt/DOD

Kaushik Velusamy/UMBC

DARPA: HIVE, SDH, DSSOC, PAPP, DPRIVE

Phil Ratzloff/SAS

Matt Galati/SAS

Jeffrey Vetter/ORNL

Albert-Jan Yzelman

Danielle Spampinato

Now at Huawei, Zurich

Roi Lipman

RedisLabs, Israel

Gabor Szarnyas/

Hungarian Academy of Sciences

External Collaborators

Andrew Lumsdaine/PNNL/UW

Jesun Firoz/PNNL

Tony Liu/UW

Kevin Deweese/UW

Tim Mattson/Intel Research

Michel Pelletier/Graphagon

Ben Brock/UCB

Aydin Buluc/LBNL/UCB

Oguz Selvitopi/LBNL

James Demmel/UCB

Roger Pearce/LLNL

Geoff Sanders/LLNL

Ben Priest/LLNL

Natalia Vassilieva/Cerebras

Intel MKL/Graph Working Group (Russia)

Andrey Nikolaev, Alexey Kukanov, Kirill Voronin

Trevor Steil/UMinn.

Luke D'Alessandro/Indiana

Python packages:
pygraphblas, grblas
(available with pip)

James Kitchen/Anaconda
Eric Welch/Anaconda

Intel MKL/Graph Working Group
Henry Gabb, Ramesh Peri, et al.

Albert-Jan Yzelman
Danielle Spampinato
Now at Huawei, Zurich

Roi Lipman
RedisLabs, Israel

Gabor Szarnyas/
Hungarian Academy of Sciences

External Collaborators

Andrew Lumsdaine/PNNL/UW

Jesun Firoz/PNNL

Tony Liu/UW

Kevin Deweese/UW

Tim Mattson/Intel Research

Michel Pelletier/Grapphagon

Ben Brock/UCB

Aydin Buluc/LBNL/UCB

Oguz Selvitopi/LBNL

James Demmel/UCB

Roger Pearce/LLNL

Geoff Sanders/LLNL

Ben Priest/LLNL

Natalia Vassilieva/Cerebras

Intel MKL/Graph Working Group (Russia)

Andrey Nikolaev, Alexey Kukanov, Kirill Voronin

Trevor Steil/UMinn.

Luke D'Alessandro/Indiana

RedisGraph Database:
(available in Google Cloud)

Jeremy Kepner/MIT/LL

Jose Moreira/IBM

Manoj Kumar/IBM

Eric Hein/EMU/Lucata

David Bader/NJIT

Paul Burkhardt/DOD

Kaushik Velusamy/UMBC

DARPA: HIVE, SDH, DSSOC,
PAPPA, DPRIVE

Phil Ratzloff/SAS

Matt Galati/SAS

Jeffrey Vetter/ORNL

Albert-Jan Yzelman

Danielle Spampinato

Now at Huawei, Zurich

Tim Davis/TAMU

Scott Kolodziej/TAMU

Jinhao Chen/TAMU

Et al.

James Kitchen/Anaconda

Eric Welch/Anaconda

Intel MKL/Graph Working Group

Henry Gabb, Ramesh Peri, et al.

Roi Lipman
RedisLabs, Israel

Gabor Szarnyas/
Hungarian Academy of Sciences

External Collaborators

Andrew Lumsdaine/PNNL/UW

Jesun Firoz/PNNL

Tony Liu/UW

Kevin Deweese/UW

Tim Mattson/Intel Research

Michel Pelletier/Grapphagon

Ben Brock/UCB

Aydin Buluc/LBNL/UCB

Oguz Selvitopi/LBNL

James Demmel/UCB

Roger Pearce/LLNL

Geoff Sanders/LLNL

Ben Priest/LLNL

Natalia Vassilieva/Cerebras

Trevor Steil/UMinn.

Luke D'Alessandro/Indiana

Jeremy Kepner/MIT/LL

Jose Moreira/IBM

Manoj Kumar/IBM

Eric Hein/EMU/Lucata

David Bader/NJIT

Paul Burkhardt/DOD

Kaushik Velusamy/UMBC

DARPA: HIVE, SDH, DSSOC,
PAPPA, DPRIVE

Phil Ratzloff/SAS

Matt Galati/SAS

Jeffrey Vetter/ORNL

Albert-Jan Yzelman

Danielle Spampinato

Now at Huawei, Zurich

Tim Davis/TAMU

Scott Kolodziej/TAMU

Jinhao Chen/TAMU

Et al.

James Kitchen/Anaconda

Eric Welch/Anaconda

Roi Lipman

RedisLabs, Israel

Intel MKL/Graph Working Group (Russia)

Andrey Nikolaev, Alexey Kukanov, Kirill Voronin

Intel MKL/Graph Working Group

Henry Gabb, Ramesh Peri, et al.

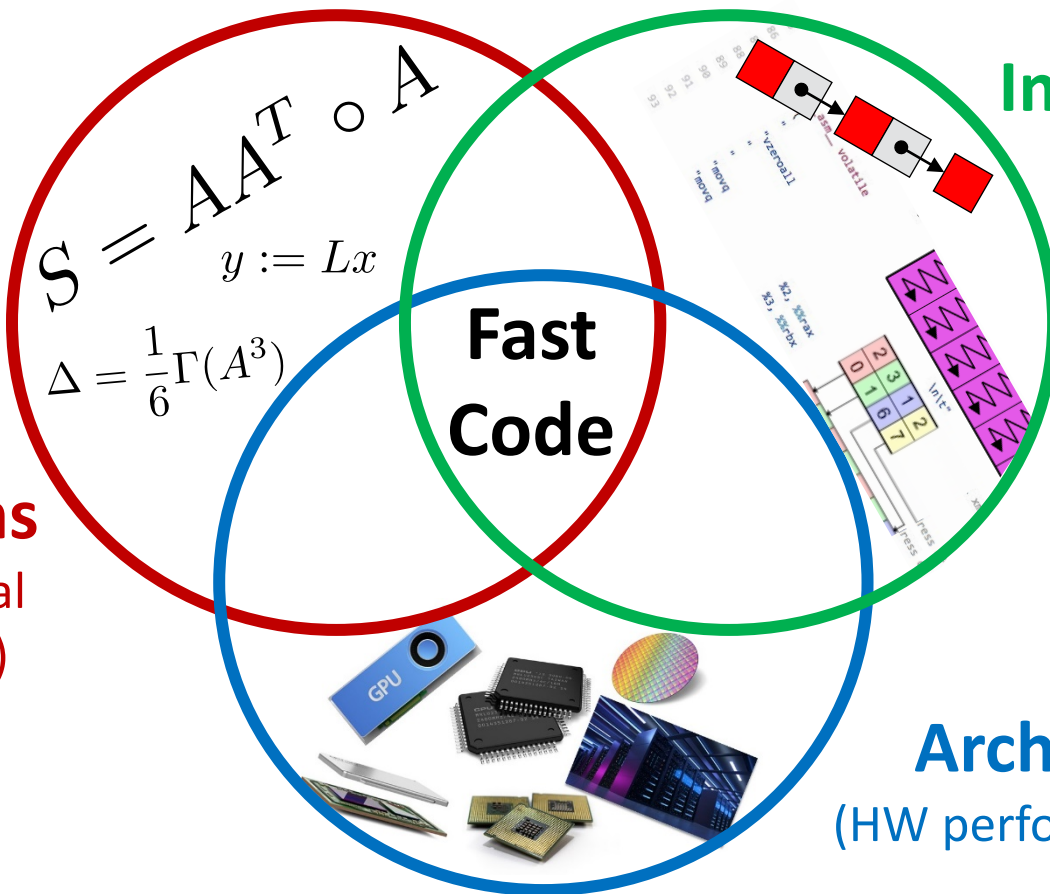
Gabor Szarnyas/

Hungarian Academy of Sciences

**A Hands-On Introduction
to the GraphBLAS
(Tutorials in C or Python)**

Fast code depends on three interdependent ingredients

Algorithms
(mathematical descriptions)



Implementations
(primitive ops and data structures)

Architectures
(HW performance models)

Team Members

SEI



Scott McMillan

Principal Engineer
The AI Division
Software Engineering Institute



John Wohlbier

Senior Research Scientist
The AI Division
Software Engineering Institute



Oren Wright

Senior Researcher
The AI Division
Software Engineering Institute



Jason Larkin

Senior Researcher
The AI Division
Software Engineering Institute

CMU ECE



Franz Franchetti

Kavčič-Moura Professor, ECE
Associate Dean for Research
College of Engineering



Jose Moura

Associate Department Head
for Research and Strategy



Tze Meng Low

Assistant Research Professor



James C. Hoe

Professor of Electrical and
Computer Engineering

Team Members

CMU Staff



Dr. Het Mankad



Dr. Daniele Spampinato
(now at Huawei)

SEI Interns



Courtney Rankin



Sandra Sanjeev
(now at Microsoft)

CMU Students



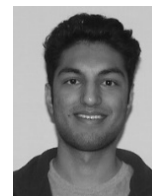
Elliott Binder



Mark Blanco



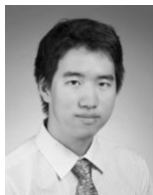
Peter Oostema



Sanil Rao



Upasana Sridhar



Mark Cheung



Guanglin Xu



George Ralph



Arvind Srinivasan



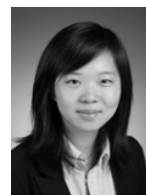
John Shi



Dr. Fazle Sadi
(Intel)



Dr. Anuva Kulkarni
(Motional)



Dr. Jiyuan Zhang
(Facebook)



Paul Brouwer



Dr. Vadim Zaliva
(Cambridge Univ.
Digamma)