

### Chapel performance for some graph analytics





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### Overview

My brief history of supercomputing

Overview of the Chapel programming language

Computing graph hitting time moments

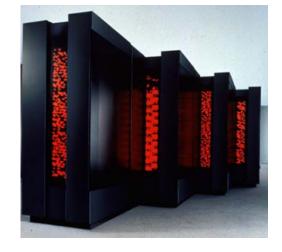
Triangle enumeration

Goal: understand Chapel's performance capabilities.

## My brief history of supercomputing

We used to have things called supercomputers.

Integrated computing environment designed for modeling and simulation.



Cray 1, serial #1@LASL 1976 1 64-bit processor

Cray 1, serial #1@LASL 1976 1 64-bit processor 160 MFLOPS, 8 MB 80 MHz, 115 kW 303 MB disk

Last was the TMC CM-5: OS, language/compiler (F90), runtime, tools (!)

TMC CM-5 @ LANL 1993 1024 x SuperSparc processors 597 GFLOPS, #1 Top500

## My brief history of supercomputing

DOE 1995 ASCI program defined COTS HPC machines.

- I was hired at LANL 6/94 essentially to help them transition from Cray vector machines.
- Lots of cool research trying to make them supercomputers: Cellular IRIX, file systems, fault tolerance, etc.

"Anyone can build a fast CPU.

The trick is to build a fast system." Seymour Cray

• Hardware, OS, compilers, languages, libraries, tools, ...

The Cray 1 delivered to LASL as a slab of hardware. LASL/LANL wrote OS, scheduler, compiler, debugger, etc.



1996 PVM UG@Santa Fe

## DARPA HPCS 2003 High Productivity Computing Systems

Productivity: time from idea to solution

Characteristics of a productive language

- Expressive
- Performance
- Portability
- Extensible

Importance of each as a function of your problem

## Chapel Programming Language

Cray High Performance Language <a href="https://chapel-lang.org/">https://chapel-lang.org/</a>
Informed by ZPL, HPF, MTA/XMT C and Fortran extensions, C++

Key construct : domain

- global view of a data structure and distribution
- first class index set

```
data parallel : forall i in Domain { }
task parallel : coforall { }, cobegin { }, begin { }
```

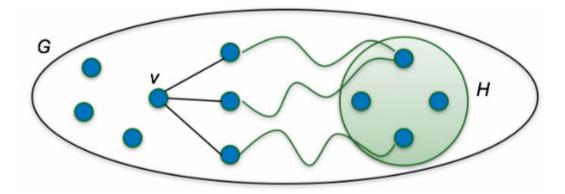
"The Rise and Fall of High Performance Fortran: an Historical Object Lesson", Kennedy, Koelbel, and Zima, 2007.

# Chapel domain

```
var MyDom = {1..n} : dmapped Block ( boundingBox={1..n} );
var x, y : [MyDom] real;
x += alpha * y;
resid = + reduce x^{**}2;
forall i in A.dom nnz { // Matrix-vector product
  y[A.rowidx[i]] += x[A.colidx[i]];
forall i in SomeDomain { // Stencil is also a domain
  y[i] = ( + reduce [ k in Stencil ] coeff ( i + k ) * x ( i + k );
```

# Graph hitting time

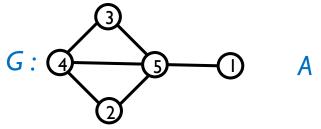
 A random variable for the number of (Markov chain) steps to reach a set of hitting set vertices H of a graph G



 Compute random variable distribution, i.e., the hitting time moments: mean, standard deviation, skew, and kurtosis.

"Advantages to modeling relational data using hypergraphs versus graphs", Wolf, Klinvexm, and Dunlavy, IEEE HPEC, 2016.

# Setting up linear system



Simple undirected graph

$$\mathbf{A} = \begin{pmatrix} 0 & 0 & 0 & 0 & 1 \\ 0 & 0 & 0 & 1 & 1 \\ 0 & 0 & 0 & 1 & 1 \\ 0 & 1 & 1 & 0 & 1 \\ 1 & 1 & 1 & 1 & 0 \end{pmatrix}$$

Adjacency matrix

Configured as linear system\*:  $(D-A)x_k = f(D, A, x_{k-1})$ 

for D = diagonal matrix of vertex degrees, x = moments where  $x_1$  mean,  $x_2$  standard deviation,  $x_3$  skew,  $x_4$  kurtosis

Solved using the Conjugate Gradient algorithm

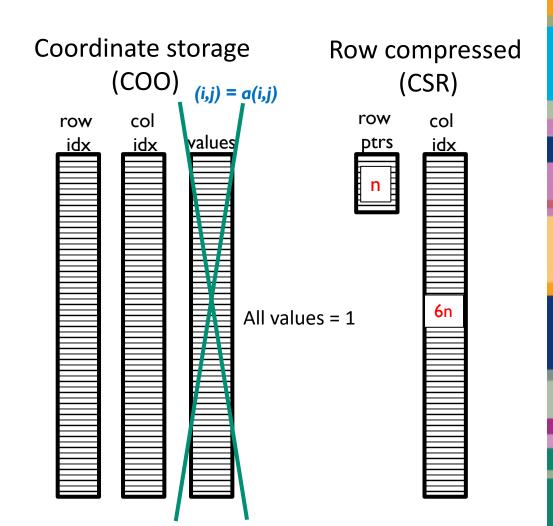
- Key kernel: matrix-vector product

<sup>\*</sup> Rich Lehoucq, report in progress.

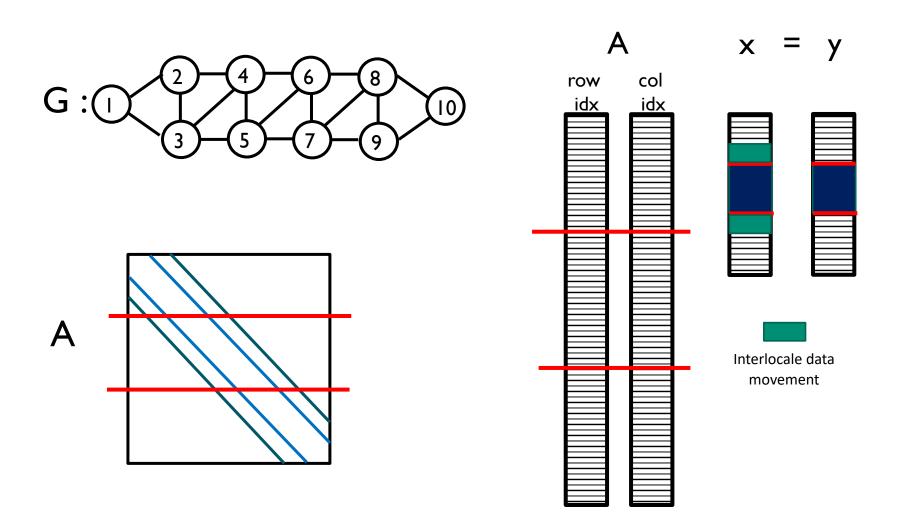
## Storing the sparse matrix

### Chapel sparse domain

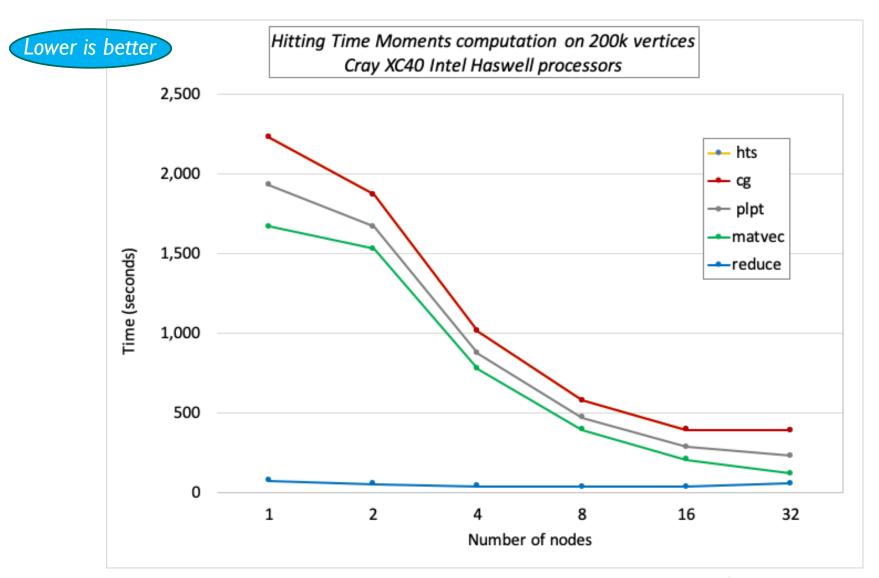
- Define dense domain
- Define subset of it: sparse domain
- Not (yet) performant
- Using for triangle enumeration in unique way



### Example: banded matrix A, in COO format

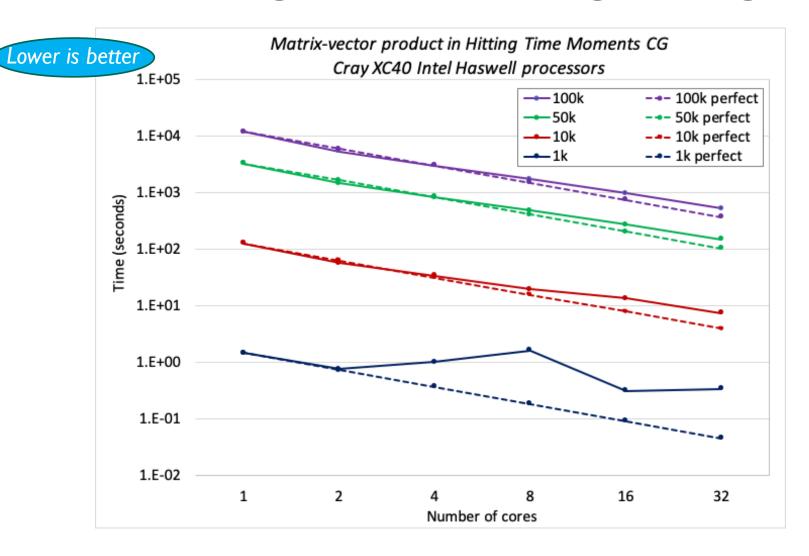


### Strong scaling\*



\* qthreads + ugni

### MatVec: single node\* strong scaling



<sup>\*</sup> qthreads runtime

### Arkouda

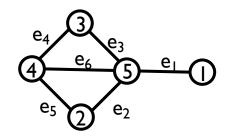
Jupyter-like notebook, launching jobs onto HPC using Chapel.

Based on numpy-like arrays

Implemented a gather/scatter capability, called "aggregation"

https://github.com/mhmerrill/arkouda

### Triangle enumeration Key computation: sparse MatMat



#### Adjacency matrix Edge Incidence matrix

"A Task-Based Linear Algebra Building Blocks Approach for Scalable Graph Analytics,", Wolf, Stark, and Berry, IEEE HPEC 2015.

# Performance Tools

#### CrayPat

■ No longer supports Chapel.

#### Tau

■ Single node support

#### **HPCToolKit**

Returns profile with missing function names, even when compiling with -g

#### **LDMS**

- PAPI sampler runs with Chapel code, but gives '0' for all data collected.
- Network samplers should work to show communication (TBD).

#### ChplBlamer

Academic tool from University of Maryland (Jeff Hollingsworth); supported?

# Summary

Scaling performance currently pretty good. Arkouda "aggregation" better?

We're assuming no known graph structure.

Exploring various matrix storage formats:

COO, CSR, Chapel sparse domain

User supplied Chapel operator capability.

Need better tools!

#### Future work

- Matrix "in place" implementation, to support full application.
- Additional processors, eg ARM, GPU and interconnects.