#### Dax Toolkit:

# A Proposed Framework for Data Analysis and Visualization at Extreme Scale

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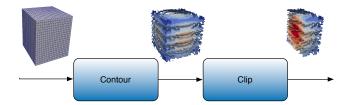
This work was supported in full by the DOE Office of Science, Advanced Scientific Computing Research, under award number 10-014707, program manager Lucy Nowell.

Motivation Dax Toolkit Results

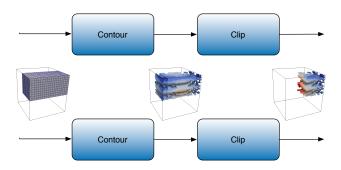
#### Dax Toolkit

A new visualization framework designed to exhibit the pervasive parallelism necessary for exascale machines.

## Visualization Pipeline



## Parallel Visualization Pipeline



#### Petascale To Exascale

	Jaguar – XT5	Exascale	Increase
Cores	224,256	100 million – 1 billion	~1,000×
Threads	224,256 way	1 – 10 billion way	~50,000×
Memory	300 Terabytes	10 – 128 Petabytes	~500×

Estimates consolidated from International Exascale Software Project Roadmap and the DOE Exascale Initiative Roadmap.

# MPI-Only Approach?

	Jaguar – XT5	Exascale	Increase
Cores	224,256	100 million – 1 billion	~1,000×
Threads	224,256 way	1 – 10 billion way	$\sim$ 50,000 $ imes$
Memory	300 Terabytes	10 – 128 Petabytes	$\sim$ 500 $\times$

Vis object code + state : 20 MB

On Jaguar : 20 MB x 200,000 processes = 4 TB

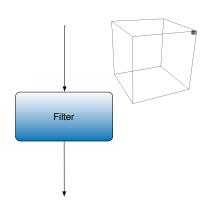
On Exascale:  $20 \text{ MB} \times 10,000,000,0000 \text{ processes} = 200 \text{ PB!}$ 

# Visualization Pipeline too heavyweight?

	Jaguar – XT5	Exascale	Increase
Cores	224,256	100 million – 1 billion	~1,000×
Threads	224,256 way	1 – 10 billion way	~50,000×
Memory	300 Terabytes	10 – 128 Petabytes	~500×

On Jaguar : 1 trillion cells  $\rightarrow$  5 million cells/thread On Exascale: 500 trillion cells  $\rightarrow$  50K cells/thread

#### Revisiting the Filter

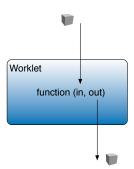


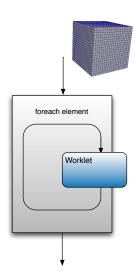
- ► Lightweight Object
- ► Serial Execution
- ▶ No explicit partitioning
- No access to larger structures
- ► No state

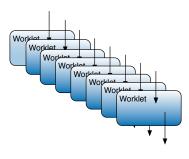
Motivation Dax Toolkit Results

Background Our Approach Related Work

function (in, out)







# Existing Approaches

Multicore extensions to VTK pipeline [Vo, et al. 2010]

- ▶ Pros: Can be applied to most existing VTK filters.
- Cons: High overhead for each execution thread; VTK algorithms optimized for sizeable chunks.

Functional field definitions (FEL/FM) [Bryson, et al. 1996]

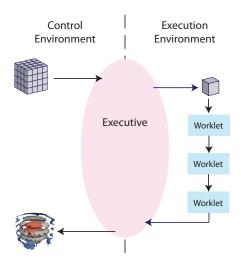
- Pros: Mesh flexibility; low memory overhead; lazy evaluation; straighforward to parallelize.
- Cons: Does not manage massive multi-threading; no mechanism for topology generation.

MapReduce [Dean and Ghemawat 2008] [Vo, et al. 2011]

- Pros: simple programming model for massive parallelism; custom systems specializing in large amounts of data.
- Cons: Difficult to cast visualization algorithms; global shuffling opeartion inefficient because it ignores known neighborhood or domain decompositions.

# Dax Toolkit

## Dax Programming Environment



#### Data Model

▶ dax::exec::Work\*

Corresponds to work performed by each Worklet.

```
dax::exec::WorkMapField
dax::exec::WorkMapCell
```

▶ dax::exec::Field

Provides access to data arrays.

```
dax::exec::FieldCell
dax::exec::FieldPoint
```

dax::exec::FieldCoordinates

#### **Execution Environment**

```
DAX_WORKLET void FieldWorklet(
    DAX_IN dax::exec::WorkMapField& work,
    DAX_IN dax::exec::Field& in_field,
    DAX_OUT dax::exec::Field& out_field)
{
    dax::Scalar in_value = in_field.GetScalar(work);
    dax::Scalar out_value = ...;
    out_field.Set(work, out_value);
}
```

#### Code Comparison

```
int vtkCellDerivatives::RequestData(...)
                                               DAX WORKLET void CellGradient(...)
ſ
  ...[allocate output arrays]...
  ...[validate inputs]...
  for (cellId=0: cellId < numCells: cellId++)
    ...[update progress]...
    input -> GetCell(cellId. cell):
                                                 dax::exec::Cell cell(work):
    subId = cell->GetParametricCenter(
                                                 dax::Vector3 parametric_cell_center
                               pcoords);
                                                   = dax::make_Vector3(0.5, 0.5, 0.5);
    inScalars -> GetTuples (
      cell->PointIds. cellScalars):
    scalars = cellScalars -> GetPointer(0):
                                                 dax::Vector3 value = cell.Derivative(
    cell->Derivatives(
                                                   parametric cell center.
     subId.
                                                   points.
     pcoords,
                                                   point_attribute,
     scalars,
                                                   0):
     1.
     derivs):
    outGradients -> SetTuple(cellId, derivs);
                                                 cell_attribute.Set(work, value);
  ...[cleanup]...
```

Motivation Dax Toolkit Results

# Results

### Implementation Assumptions

► GPU ≈ Exascale Node

► CUDA ≈ Development Environment on Exascale Node

# Performance Comparison

Mesh Size	VTK Time	Dax Time	Speedup		
$Elevation \to Gradient$					
144 <sup>3</sup>	2.75 s	0.013 (0.024) s	210 (114)		
256 <sup>3</sup>	15.52 s	0.074 (0.135) s	210 (115)		
512 <sup>3</sup>	125.75 s	0.589 (1.076) s	213 (117)		
$Elevation \to Sine \to Square \to Cosine$					
144 <sup>3</sup>	2.32 s	0.002 (0.006) s	1169 (386)		
256 <sup>3</sup>	12.99 s	0.013 (0.034) s	999 (382)		
512 <sup>3</sup>	103.88 s	0.110 (0.276) s	944 (376)		

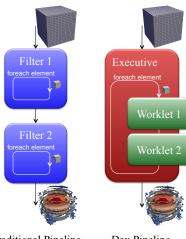
Performance comparison between Dax toolkit and VTK. Values in parentheses show the corresponding values with data transfer times included.

#### Challenges and Ongoing Work

 Topology modifying Worklets e.g. Marching Cubes/Streamlines

▶ I/O and Rendering

#### Conclusion



Traditional Pipeline

Dax Pipeline

#### Acknowledgements

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