

# Networking Simulation Clusters with Visualization Clusters for Real-Time Data Analysis

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## Motivation

### Basic Job Queue Paradigm

#### Computational Simulation

- Submit job
- Wait in queue
- Simulation runs to completion
- Results saved to disk

#### Post-processing of Data

- Read data from disk
- Visualize data
- Save resulting images / video to disk

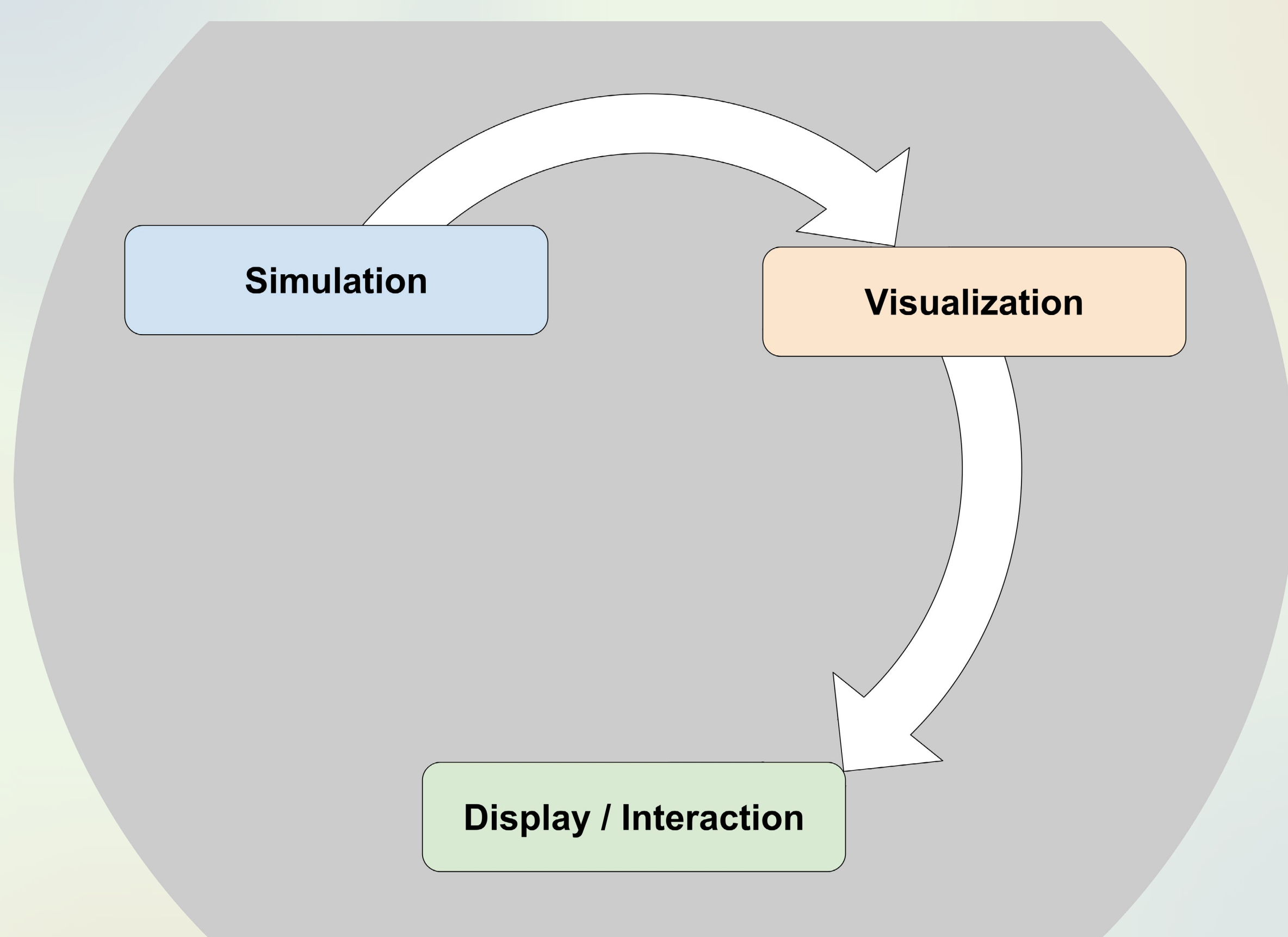
#### Viewing Results

- Read images / video from disk
- View and explore the results

### Real-Time Paradigm

- Concurrent visualization and display of a running simulation
- Data streamed between simulation and visualization resources in real-time

*Real-time streaming **eliminates costly reads to / writes from disk**, enables a **higher sampling rate** for analysis, and can **reduce the time-to-discovery!***



## Results

### Parallel Streaming

- Data distributed in  $M$  simulation ranks
- $N$  visualization ranks each connect to a simulation rank
- Data streamed in parallel ( $M$ -to- $N$ )

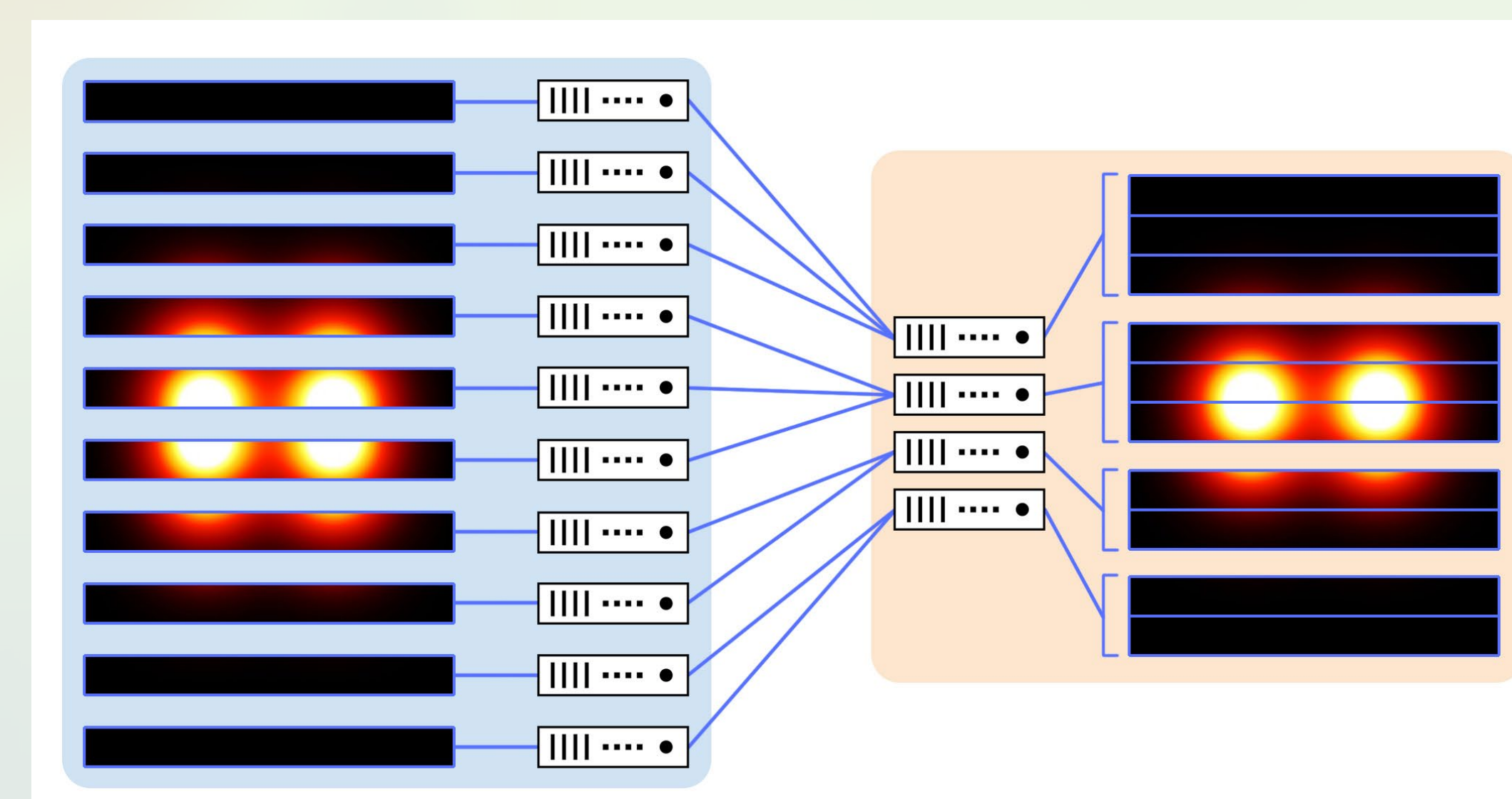
### Data Redistribution

- Visualization declares how it expects data
- Incoming data is redistributed to fit the expected layout
- Works regardless of layout in simulation

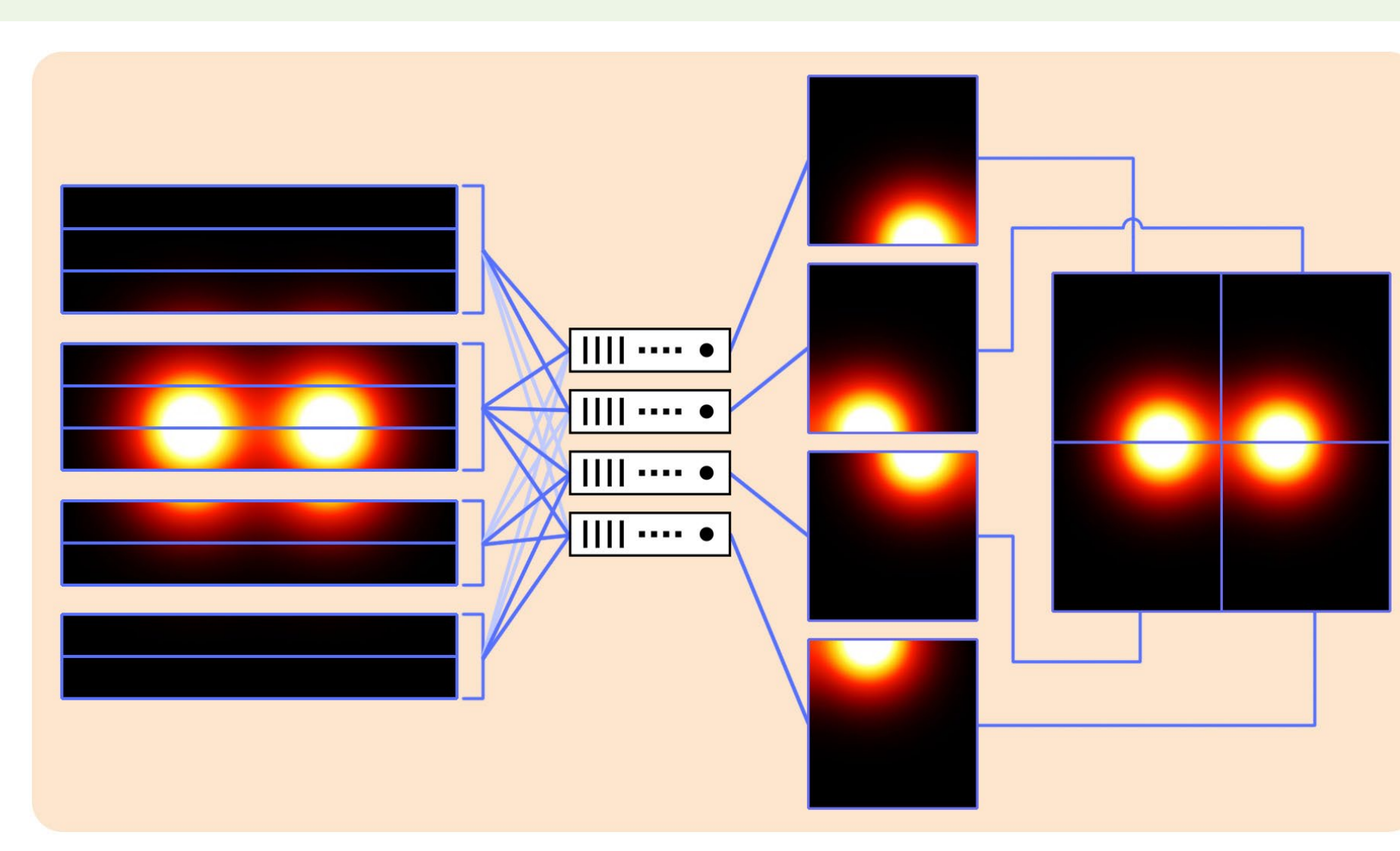
### Use Case: 2D Heat Diffusion

Grid Dimensions	Raw Data Size	Processed Data Size
2048x2048	3.2 GB	19.9 MB
4096x4096	12.8 GB	61.0 MB
8192x8192	51.2 GB	217.8 MB
16384x16384	204.8 GB	830.9 MB

Raw data saved directly from 4-byte float array. Processed data saved as JPEG images rendered from visualization application.



Parallel data streaming of 2D heat diffusion data. This illustration shows 10 simulations ranks streaming data to 4 analysis ranks.



Data redistribution by the visualization application. Regardless of incoming data layout, data is automatically reorganized to fit the layout specified by the visualization application.

## Impact

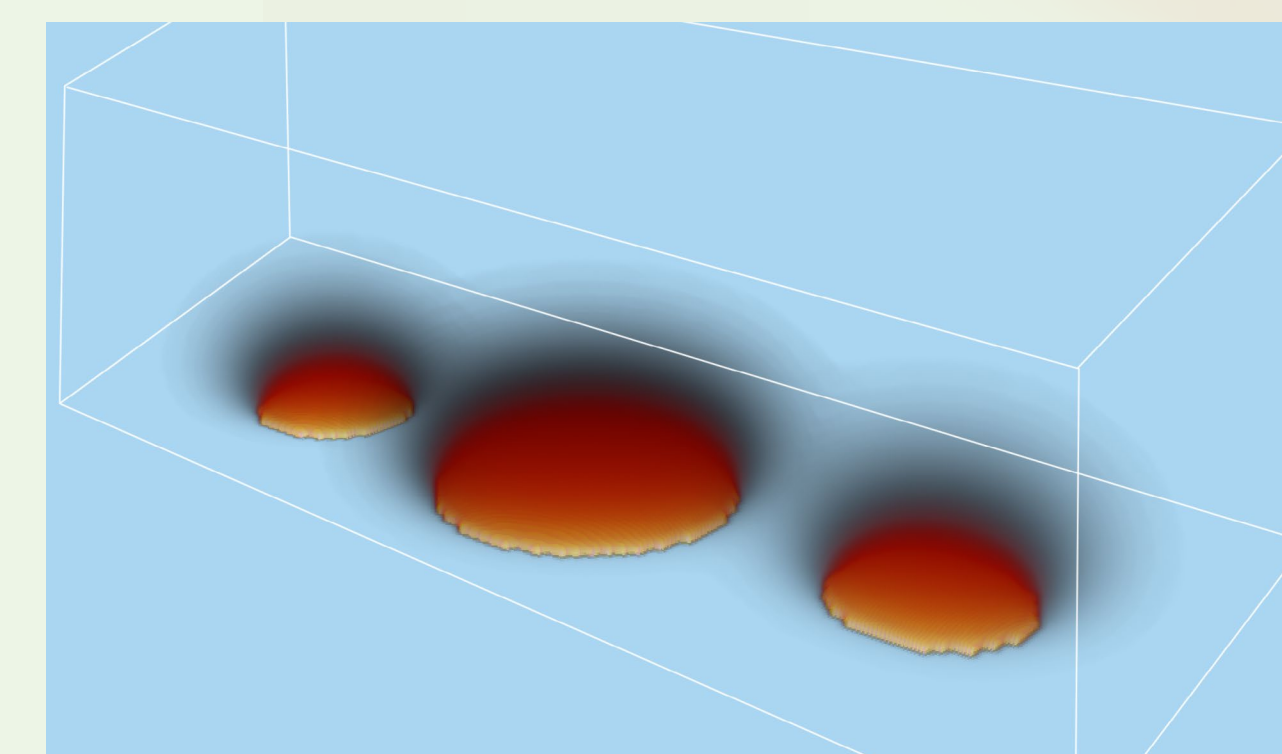
### Wide-Ranging Applicability

#### Computational Simulation and Visualization

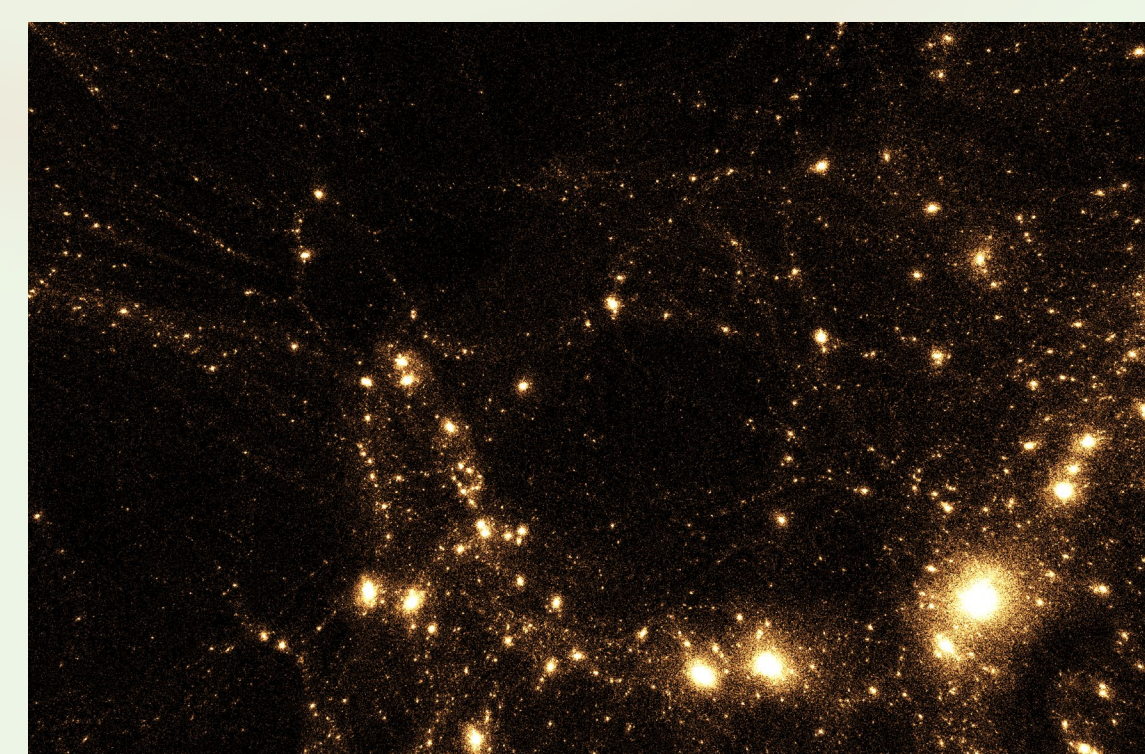
- Improves sampling rate without increasing storage needs
- More fully utilizes compute resources (leverages high-bandwidth networks and GPUs)
- Visualization can be run and viewed on demand
- Necessary for exascale computing

#### Example Applications

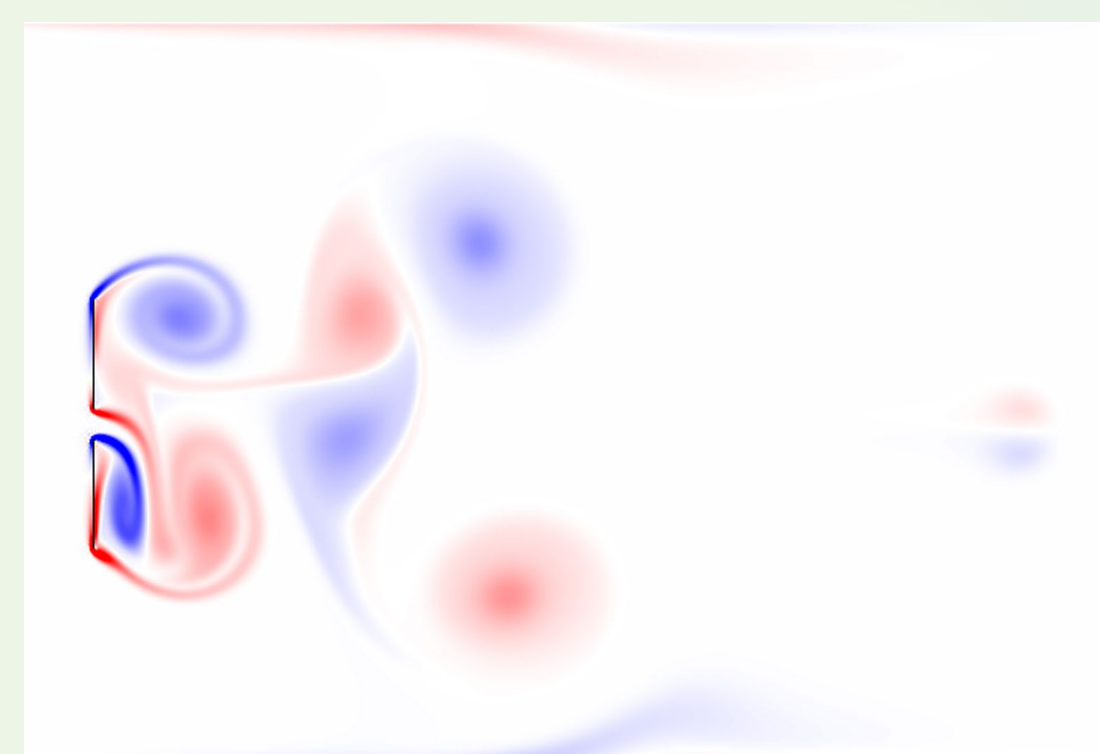
- 2D regular grid simulations → texture mapped rendering
- 3D regular grid simulations → direct volume rendering
- Particle-based simulations → particle rendering



3D heat diffusion.



Dark matter particle physics.



2D computational fluid dynamics.

### Benefits

#### Reusable Visualization Applications

- Generalizable (examples on left)
- Handle data from a variety of different simulations

#### Little Overhead for Simulations

- Data redistribution done in the visualization application
- Limits run-time costs to the simulation

## Future Directions

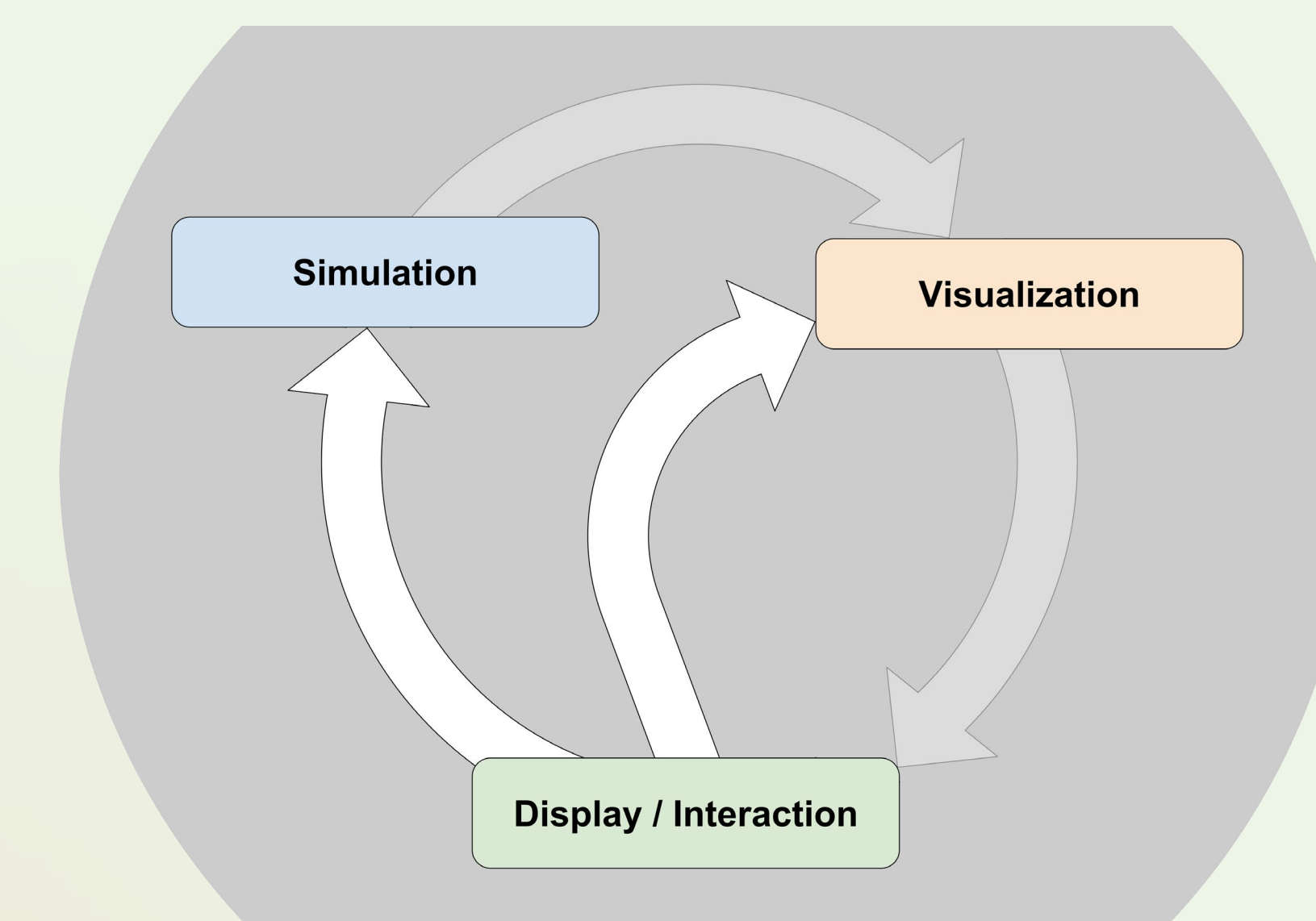
### Computational Steering

#### Feedback Loop for Simulation and Visualization

- User interaction to modify simulation or visualization
- Checkpointing system (allow for 'undo')
- Make HPC Applications look and feel like a desktop app

#### Example Uses

- Focus on area of interest
- Tweak parameters of simulation or visualization
- Explore 'what if' scenarios



### Acknowledgements

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