

# Streamlining Computation and Communication for Distributed Science Workflows

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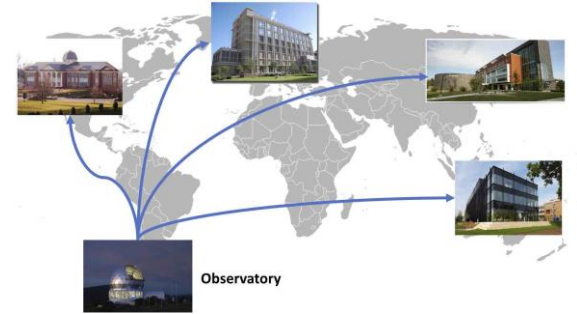
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# Distributed Science

- Remote data processing
  - Data is moved from data collection facilities to HPC clusters to analyze
- Collaborative projects
  - Data sharing to enable collaboration
- Reproducible research
  - Central repositories to store data





# How to process large-scale remote data?

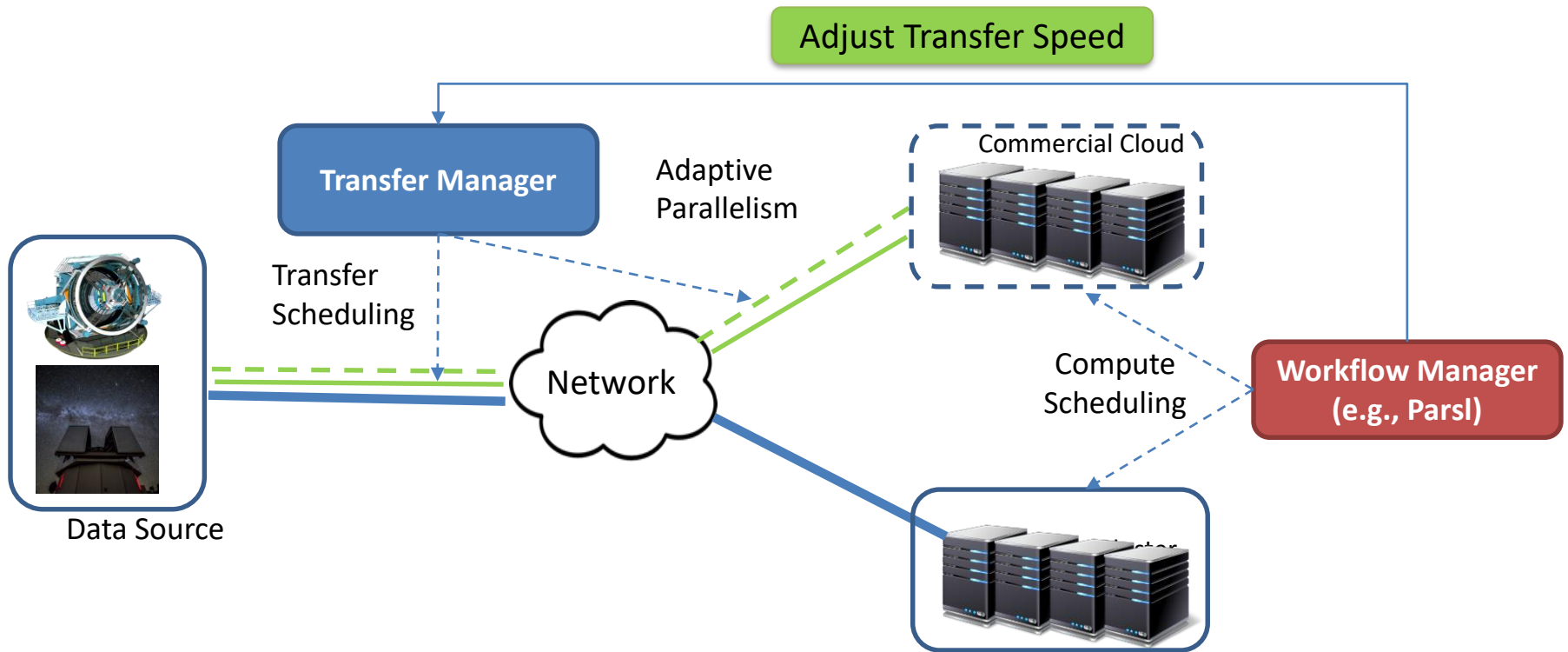
## 1. Download first, process later

-  Long execution time → Transfer time + Processing time
-  Need for large staging space → Potentially increased storage cost

## 2. Streamline compute and transfer

-  Mismatch of compute and transfer speeds
-  Performance variability due to interference

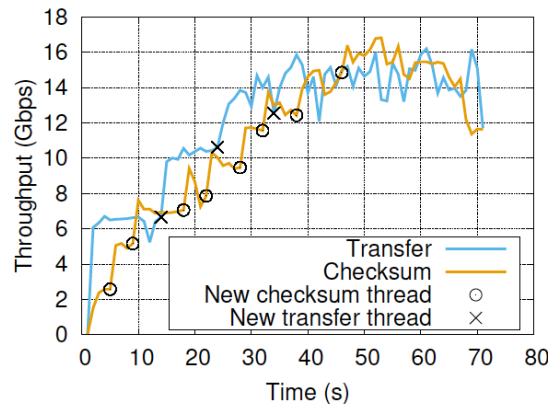
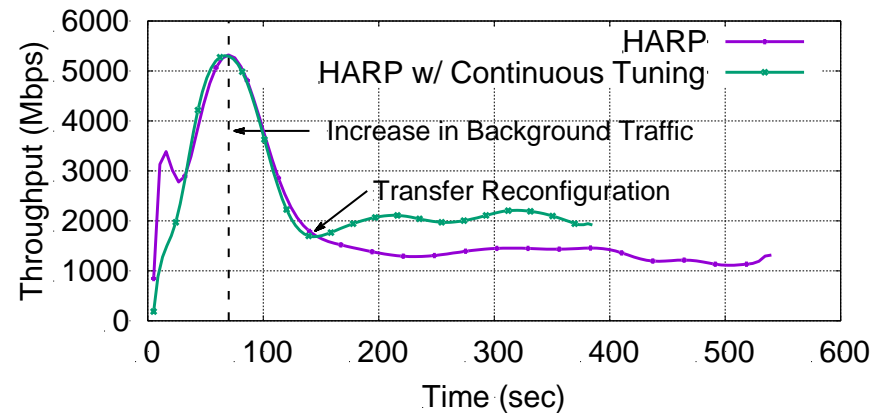
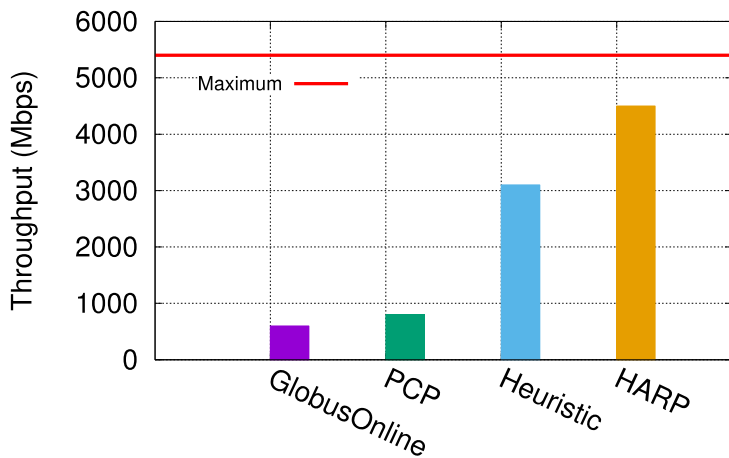
# End-to-End Workflow Parallelism



# Real-time Data Transfer Tuning with HARP

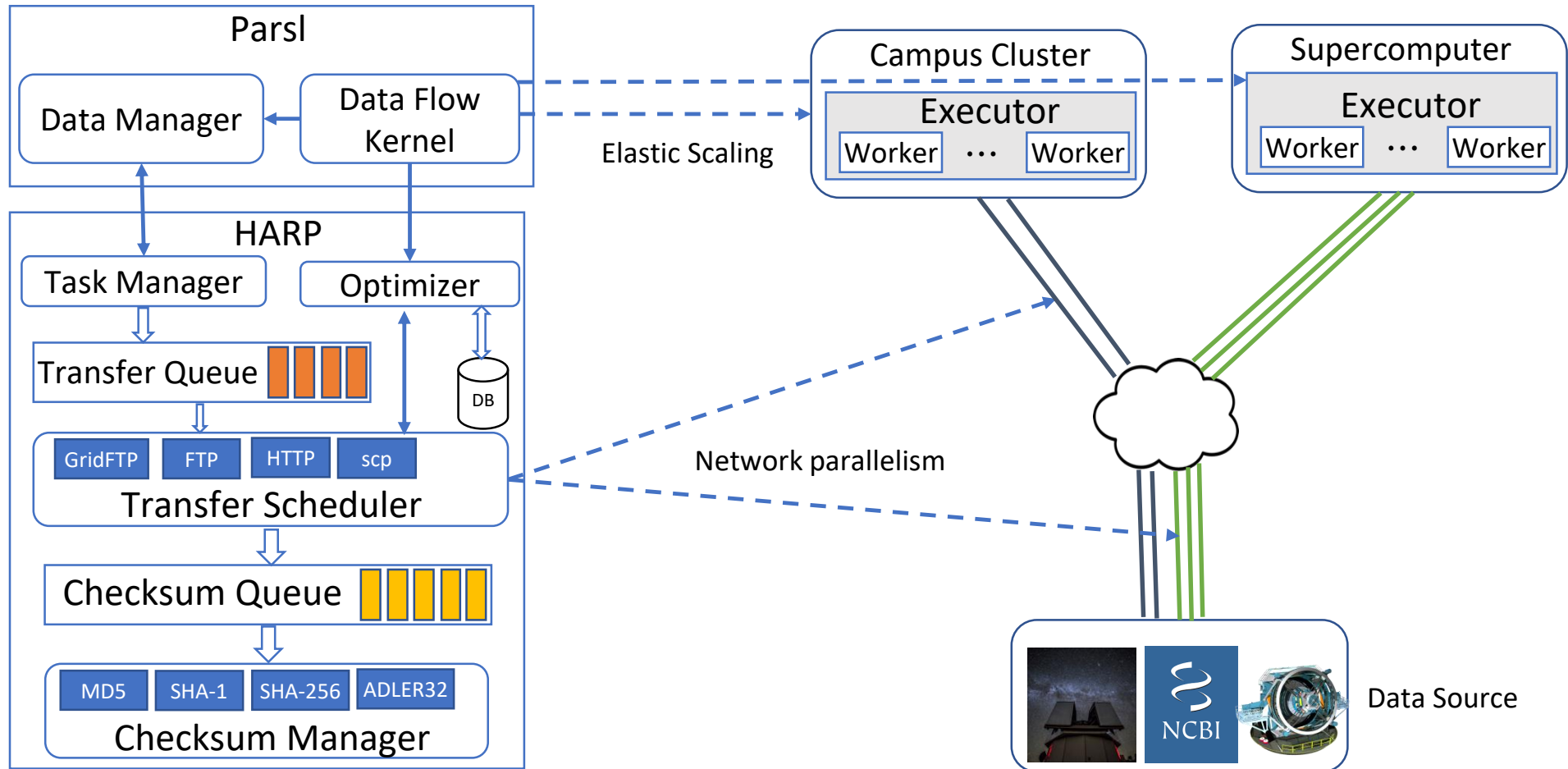
High-performance data transfers

Responsive to changing conditions



Scalable Integrity Verification

# Parsl Integration



## Why Parsl?

- Supports computational parallelism
- Easy to integrate new data transfer module
- Responsive and helpful development team

Thanks!



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