# Beyond the Instrument: How to Manage and Master Your Core Facility Data

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# Overview of Seattle Children's Research Institute

#### **Core Services**

## **Shared Research Resources (SRR)**

- ABSL-3 Core Facility
- Behavioral Phenotyping Core
- Flow Cytometry Core
- Genomics and Spatial Biology CoLab
- Microscopy and Histopathology CoLab
- Preclinical Imaging Core



Center Support Services

- Research IT
  - Research Storage Services (RSS)
- Research Informatics
  - Research Scientific Computing
    - High Performance Computing Core
    - Bioinformatics Collaboration and Consultation



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# **Data Workflow in SRR**

# **Core Services**

#### **Shared Research Resources (SRR)**

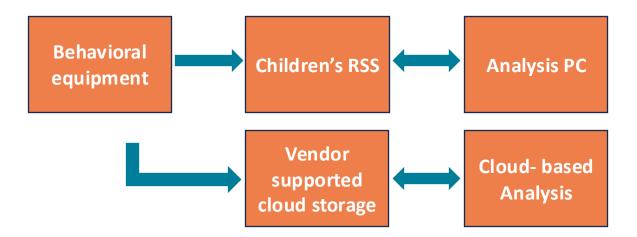
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## **Data Workflow in SRR**

#### **Behavioral Phenotyping Core (BPC)**

- Data type: Behavioral videos (few hours to > 24 hours)
- Data size: ~2 TB/ day
- Data storage: AWS cloud (integrated with Allentown Inc.) → Children's RSS
- Data Analysis: Machine-Learning; High-end PC with a dedicated graphics card

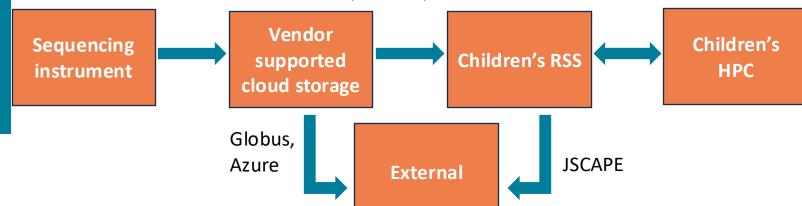




## **Data Workflow in SRR**

#### Genomics and Spatial Biology CoLab (GSB)

- Data type: Sequencing data from Illumina/ PacBio instruments
- Data size: ~10 GB- 1 TB per file
- Data storage: Instrument cloud → Children's RSS
- Data Analysis: Children's HPC
- External Transfer: Globus Connect, JSCAPE, Azure





**Data Generation > Data Processing > Data Analysis .....Data Storage + Transfer** 



## **Data Generation** > Data Processing > Data Analysis .....Data Storage + Transfer

Instrument	Avg. Raw data	Usage
Zeiss LSM 900 (Airyscan) Confocal	1- 80 GB	3-4 users per day; 8- 10 hours per day
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Zeiss Axioscan 7 slide scanner	2- 20 GB	20-100 slides per week
LifeCanvas SmartSPIM light-sheet	1 TB	moderate- heavy usage
ImageXpress Nano High-Content Imager		Low-moderate usage
CSU W1 SoRa Spinning disk Confocal		moderate- heavy usage



## Why use HPC for microscopy data?

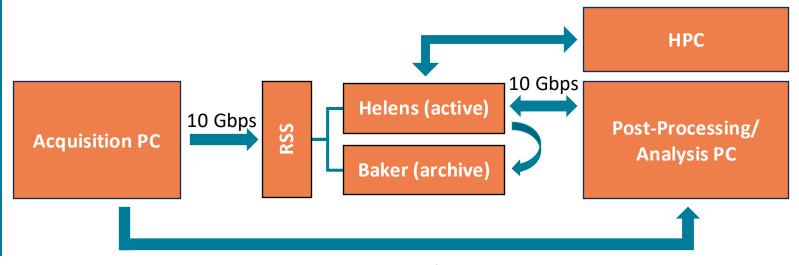
- Explosion in data volume and complexity (increasing > 5x in recent years)
- Growing demand for rapid analysis and sharing
- Need for scalable, reliable infrastructure

#### HPC can:

- Handle large-scale processing (TB-scale datasets)
- Accelerate image analysis workflows (segmentation, registration, AI)
- Faster turnaround for users
- Enable automated, reproducible workflows
- Better sharing and collaborative access



**Data Generation > Data Processing > Data Analysis .....Data Storage + Transfer** 





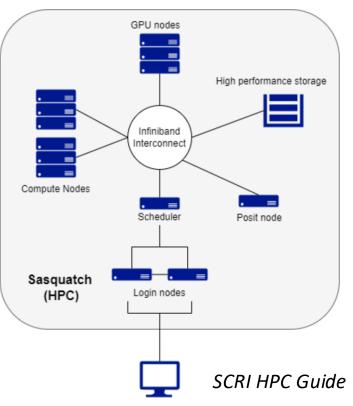
Direct Processing during acquisition via 10 Gig connection

**Data Generation > Data Processing > Data Analysis .....Data Storage + Transfer** Helens (active) HPC RSS Baker (archive) Light-sheet **Acquisition PC** For long-term **Direct Processing** via 10 Gig 25 Gbps **Short-term Data Post-Processing PC Cloud Storage** (De-striping, Stitching) Storage 'MOSAIC' For long-term 25 Gbps **Analysis PC** 



# **HPC Workflow**

- Onboarding to request an account
- Orientation and Training
- Connect to HPC (login nodes)
  - Home directory (100 GB)
  - Association directory (1 TB increments)
- Transfer data from RSS to HPC (not queued)
- Use Slurm for scheduling (jobs are queued)
  - Determine the resources required
  - CPU, GPU, memory, time
- Job dispatched to one or more of the nodes
- Job completed- output written to the specified subfolder in association directory
- Transfer analyzed data to RSS





# **Using HPC for Microscopy Data**

## Use Case 1: 3D/ 4D/ 5D Confocal analysis

- Large screen of >800 mutants
- High-content imaging of several 384-well plates
  - 10+ ROIs per well
  - Multichannel z-stacks; time series
- Repetitive pipeline (masking, filter, segmentation, quantification of 5+ parameters)
- Ideal for containerized batch analysis on HPC
- 1.3 million cell instances segmented per day on HPC
- 30 experiments run parallel for analysis



# **Using HPC for Microscopy Data**

# **Use Case 2: Light-sheet Imaging**

- 1 TB per sample
- Destriping, stitching, tile correction, atlas registration
- Too slow and resource heavy on local machines

## Pilot project

- Identify workflows common to all users (post-processing of light-sheet data)
  - Destriping using 'Pystripe'
  - Stitching using 'TeraStitcher'
- Deploy shared storage + processing containers
- Evaluate usage and impact → scale as needed



# **Integration Challenges**

#### **Data Transfer Bottlenecks**

 Slow data transfer speeds from acquisition/ RSS to HPC storage creates workflow delays.

#### **Workflow Compatibility**

- Many microscopy tools are Windows-only or GUI-based, while HPC environments are typically Linux and CLI-based.
- Software requiring **proprietary licenses** may not be easily portable to HPC.

### **Usability Gap**

- Most users are more comfortable with GUIs, but HPC access often requires command-line skills.
- Limited coding experience makes it hard for users to optimize or build pipelines on their own.



# **Progress & Lessons Learned**

#### Progress made:

- Direct Processing Pipelines via 10 Gbps
- Established **25 Gbps** dedicated link for light-sheet data transfer
- Launched specialized **image analysis workshops** (Python scripts) for researchers
- Identified pilot projects for HPC integration

#### **Lessons learned:**

- Core's Responsibility: Bridge the Gap. We must translate researcher needs to IT requirements and vice versa.
- The Partnership Model: Make IT your best friend. Ongoing discussions with Research Support Services (RSS) and HPC are crucial for policy and infrastructure.
- Actionable Start: Create and deploy basic, common pipelines (e.g., deconvolution, batch processing) to immediately lower the barrier to entry.



"People. Ideas. Technology – in that order." -attributed to John Boyd





Hope. Care. Cure.