# Advancing Streaming Readout and Processing Architectures at JLAB

**EICUG/EPIC Collaboration Meeting** 

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### Breaking the Limits: The Problems We Must Solve

#### Scientific Data Expansion



#### Get more out of experiments

- Integration of all detectors into the event identification process.
- Increase science/\$ ratio
- Preserve filtered, widely open trigger data for future physicist.



CLAS12 DAQ dead-time



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## Vertical and horizontal scaling and it's limitations

#### Vertical scaling

- Some portion of the code remains serial
- Memory contention, leading to a plateau in performance.
- Concurrent accesses to CPU caches, increasing cache misses. If threads rely on disk or network I/O, and context switching.

#### Horizontal (X) scaling

- Batch processing
  - Issues:
    - Limited local resources
    - Require data migration and temporal persistency (IO latency)

#### CLAS12 scaling curve Amdahl fit





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# Stream Data Workloads Require Scalable Processing

#### Social media

Science





IOT vs. IOST





Precise genetic correlations between IOSTs

The Art of Scalability. by Martin L. Abbott and Michael T. Fisher. ISBN-13: 978-0134032801



### End-to-End Streaming Data Acquisition





Jefferson Lab

homas Jefferson National Accelerator Facility



### CODA/ERSAP Role in Streaming DAQ





### DAQ stream source: VTP







# Y Scaling: Event Reactive Microservices Framework



Agile framework that makes easy software evolution over time!

Programming V. Gyuriyan, D. Abbott, N. Brei, M. Goodrich, G. Heyes, SM : Shared Memory E. Jastrzembski, D. Lawrence, B. Raydo, C. Timmer Published at: IEEE Xplore **DPS**: Data Processing Station DOI: https://doi.org/10.1109/TNS.2023.3242548



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ERSAP provides event level parallelization. Supports arbitrary fragmented data (Z scaling).

UE : User Engine

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heterogeneity.



# Efficient X and Z scaling for streaming experiments

- We need network solutions that recognize experiment unique markers
- Shape, sort, and forward data based on experiment-specific markers.
- Dynamic load balancing to ensure optimal and efficient use of computational resources.



*EJ-FAT Joint ESnet JLab FPGA Accelerated Transport Load Balancer* Stacey Sheldon, Yatish Kumar, Michael Goodrich, Graham Heyes <u>arXiv:2303.16351</u> [cs.NI]





# 4x8 CH Scintillator SRO pipeline at JLAB Test Lab





Photos courtesy of Hanjie Liu

Triggered data are waveforms read out over the VME bus.

Stream data are integrated sums and times of all hits over a threshold in the calorimeter regardless of the trigger status.



### SRO and data processing in action











### Conclusion

- We have successfully carried out several streaming readout campaigns, including beam tests that demonstrated the feasibility of the streaming readout approach. The Streaming CODA and ERSAP frameworks are at the forefront of the SRO effort at the JLAB
- The first time, a fully remote, distributed data stream processing workflow has been successfully demonstrated using production-level physics data across DOE computing facilities using EJFAT and ERSAP. This marks a significant milestone in our efforts to enable real-time, cross-site data curation and analysis at scale.



# Thank you