

Near real-time streaming analysis of big fusion data

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2021-12-03

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Goal: Tackle fusion data analytics at ITER scale to expand operational capabilities of fusion facilities



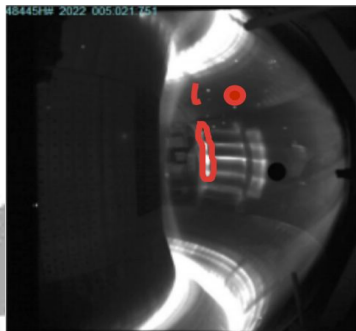
- **Driver: Compute environments are changing - ITER will have only minimal compute resources on-site.**
 - Increase in compute power driven by accelerators (GPU/TPU/ASIC/WSE)
 - Storage volume and speed don't keep up with increase in compute power
- **Driver: Long-pulse devices need fast turnaround time for data analysis**
 - Measurements need to be analyzed as plasma pulses are ongoing - overlap experiments and data analysis to save time.
- **Driver: ML algorithms perform best when trained on big data sets**
 - Manual training will become infeasible. ML models will need to be trained online as new data becomes available.

Solution: Move to near real-time streaming data analysis.

For this we are developing the [DELTA framework](#)



Connect experiments, scientists, and compute resources to optimize scientific discovery

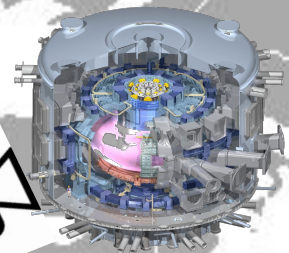


Hot spots detected!!

Move the plasma gap 2cm inward



Remote scientist sees analyzed data - while experiment is running



Experiment:
ITER
KSTAR
NSTX-U...

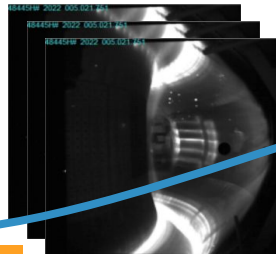
In the future: run simulations



Data analysis at remote HPC center - while experiment is running

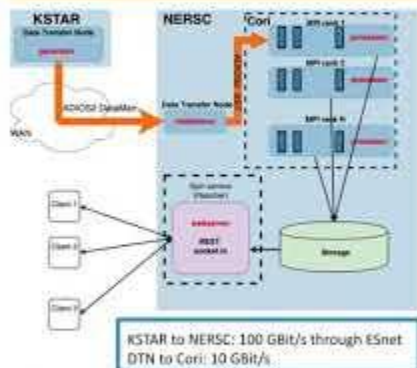


Stream data while experiment is running





Delta framework is a distributed system that facilitates federated data analysis



- generator streams data into HPC facility
- Data is **streamed** using ADIOS2 library
- **middleman** serves as a relay
- **processor** receives data stream and performs analysis on HPC resource
- Analyzed data is **stored in a database** where it is accessible to externally facing services
- **Webserver** running on rancher serves visualization requests from web-clients

In December 2020 we used NERSC to analyze KSTAR ECEi measurements in between shots.

Streaming the measurements to NERSC, analyzing and visualizing them took about 5 minutes.

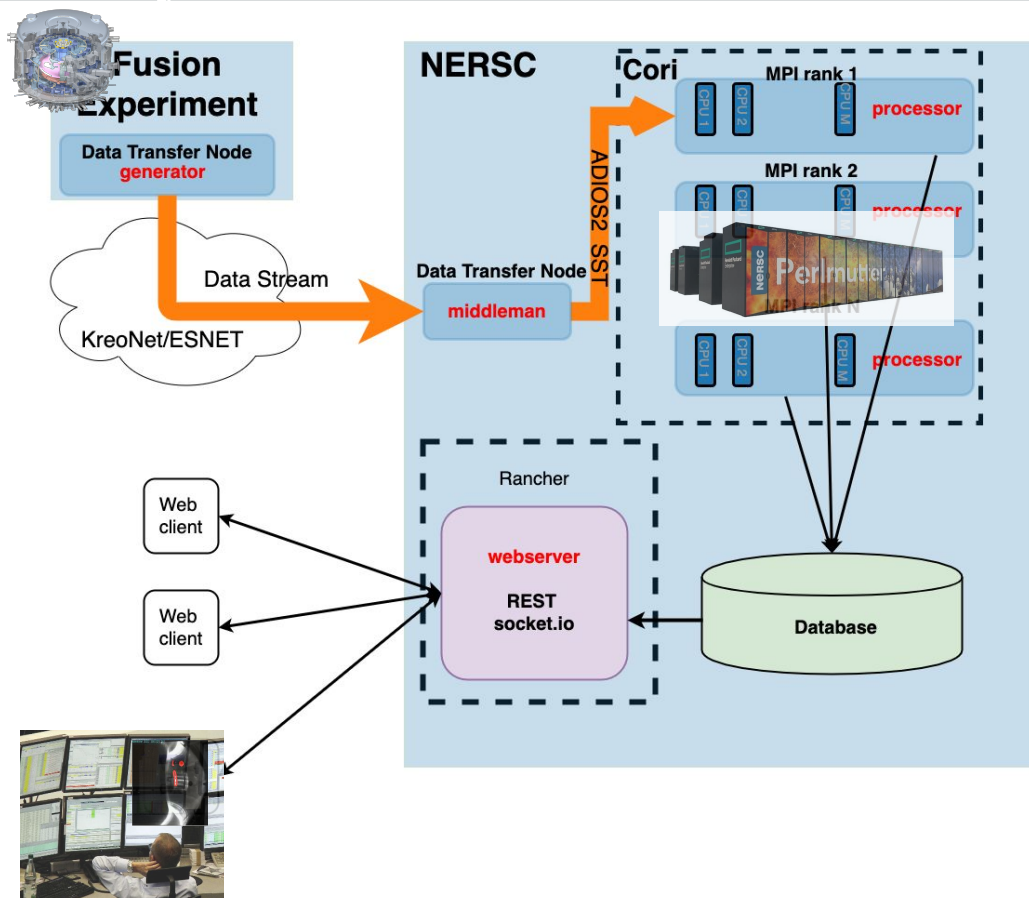
Shot cadence at KSTAR is ~10 minutes.

Screen layout:

| Network bandwidth | Network bandwidth | CPU utilization |
|--------------------------------|-----------------------|-------------------------------------|
| KSTAR Data generator | NERSC relay | Cori(NERSC) Data analysis |

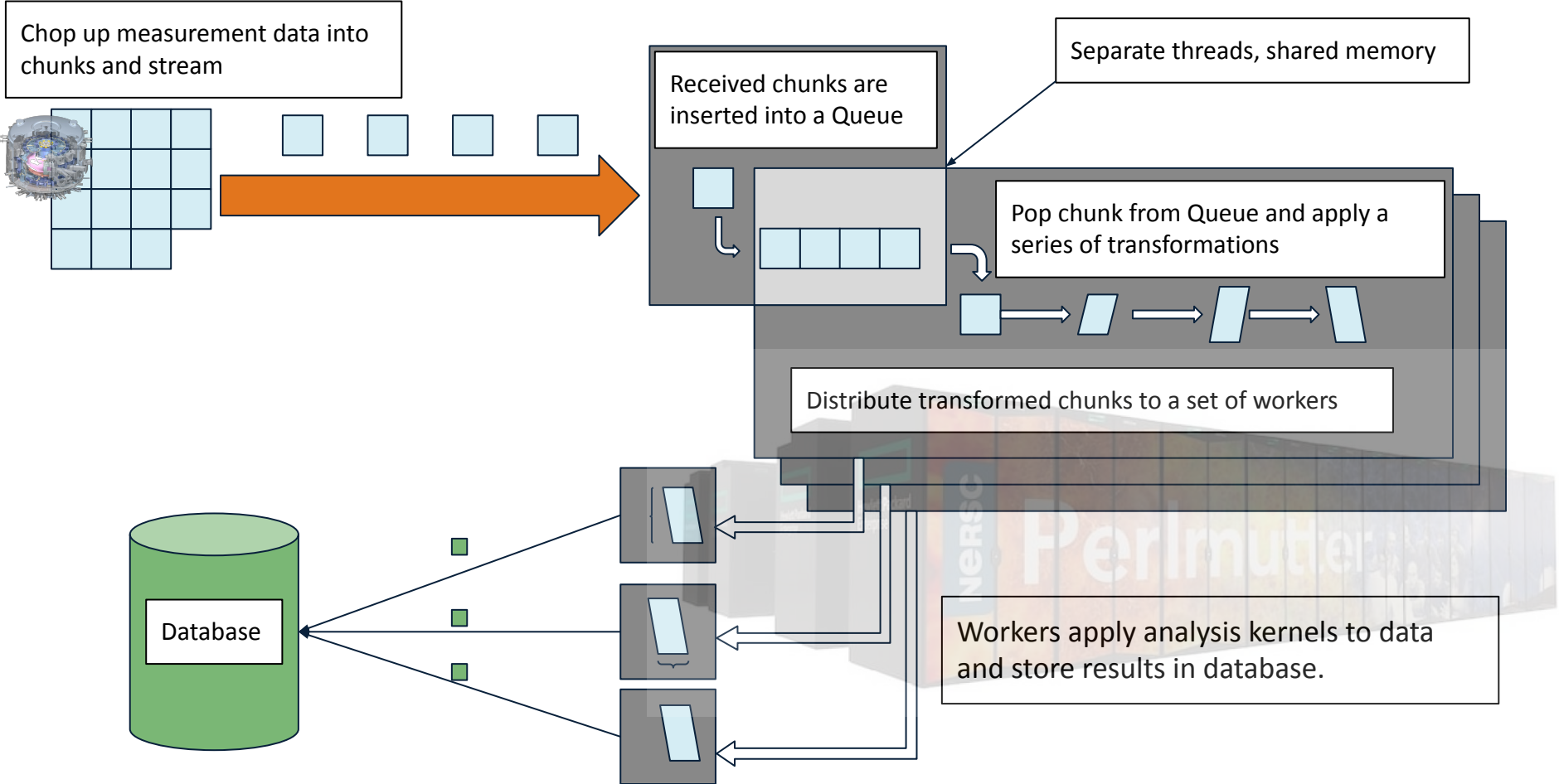
[Video link](#)

DELTA is a framework that enables near real-time analysis of data streams from fusion experiments



- **generator** streams data into HPC facility.
 - Input are data files. Future: stream directly from diagnostic
- **middleman** serves as a relay.
 - Data never touches the file system
- **processor** receives data stream and performs analysis on HPC resource.
 - Distributes work using a master-worker architecture
- Analyzed data is stored in a **database** where it is accessible to externally facing services.
 - Modern databases facilitate flexible real-time access by downstream applications
- **Webserver** running on Spin (container orchestration service) serves visualization requests from web-clients.
 - Make analyzed data instantaneously accessible from around the world. Allow to use modern web-based collaboration tools.

Delta processor simultaneously receives a data stream and runs analysis.

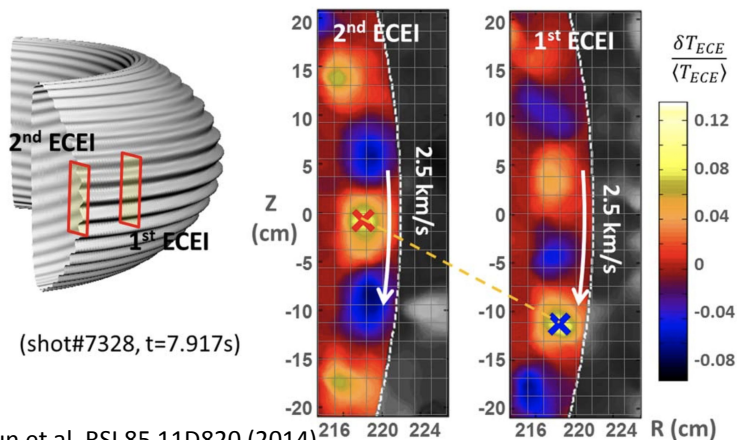


Define a compute intensive benchmark workload to test scalability of the DELTA processor



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ECEI visualizes large scale 2.5d plasma structures



Yun et al. RSI 85 11D820 (2014)

- KSTAR ECE diagnostic: samples 24x8=192 channels with ~MHz sampling rate
- Diagnostics produces image time-series with about 1GB/sec

ECE benchmark workflow:

Estimate the power spectrum of each channel. Then calculate:

- Cross-power: $S_{XY}(\omega) = E[X(\omega)Y^\dagger(\omega)]$
- Coherence: $C_{XY}(\omega) = |S_{XY}(\omega)| / S_{XX}(\omega)^{1/2} S_{YY}(\omega)^{1/2}$
- Cross-phase: $P_{XY}(\omega) = \arctan(\text{Im}(S_{XY}(\omega))/\text{Re}(S_{XY}(\omega)))$
- Cross-correlation: $R_{XY}(t) = \text{IFFT}(S_{XY}(\omega))$

for all $\binom{192}{2} = 18336$ channel pair combinations (X,Y).

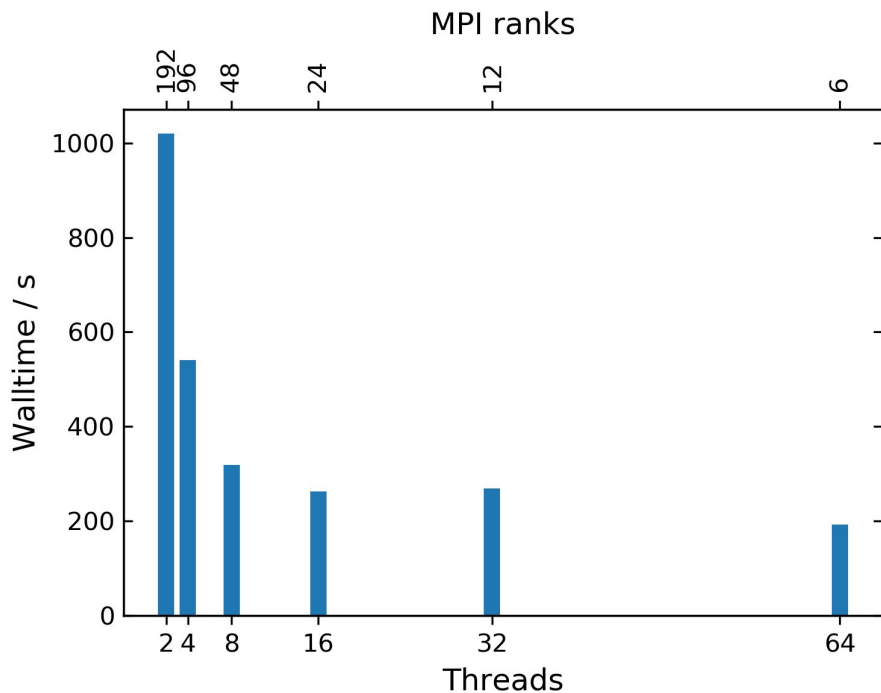
- Each calculation above is implemented as a separate kernel.
- For a single shot, we partition the ECEI data into 500 chunks of size 24x8x10,000.
- Code adapted from [fluctana](#), M. Choi. Other codes possible.

Use cases for C, S, P, R:

Estimation of local dispersion relation (flow velocity),
2d characterization of T_e turbulence, identification of
avalanche-like T_e transport events:

Choi et al. NF 57 126058 (2017); Choi et al. NF 59 086027 (2019);

The benchmark scales well over multiple CPU nodes until we hit architectural bottlenecks



Run benchmark on 6 Cori nodes (Xeon Haswell CPU with 32 cores, 128GB RAM).

- 192 MPI ranks / 2 hyperthreads: Too much communication, CPU cores are not effectively utilized.
- 6 MPI ranks / 64 hyperthreads: Shortest walltime, about 190s.

KSTAR shot cadence: approx. 10 minutes

Fastest execution time: 190 seconds.

Caveats:

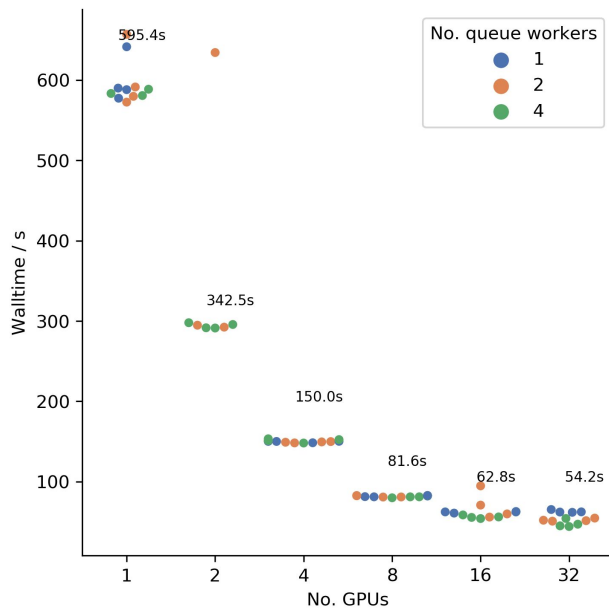
- Data is read from filesystem, no streaming.
- Data analysis results are discarded

Walltime does not change much when streaming + storage is added.

Benchmark analysis scales well on multiple CoriGPU nodes until we hit architectural bottlenecks



- Kernels C,S,P perform similar computation -> Fuse them into a single kernel.
- Implement kernel using numba to enable just-in-time-compilation and GPU execution
- Strong scaling up to 16 GPU nodes.
- After that, pre-processing throughput becomes the bottleneck.



Caveats:

- Time shown is for processing 100 chunks.
- Data is read from filesystem, no streaming.
- Data analysis results are discarded
- Data chunks are pre-processed using a short-time Fourier transform and bandpass filter

This is about 5x slower than the CPU case.

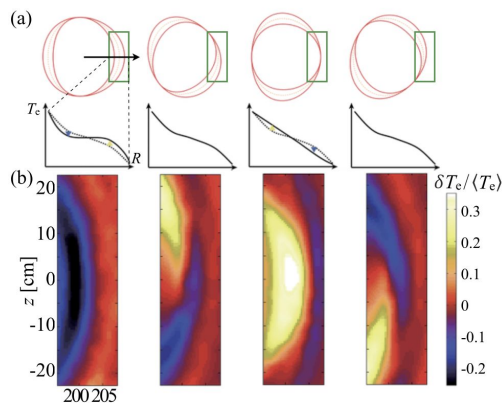
- Memory access pattern in the kernel is challenging
- Overhead due to memory transfer from/to GPU

Machine learning can be leveraged at multiple stages in the DELTA framework to aid scientific discovery



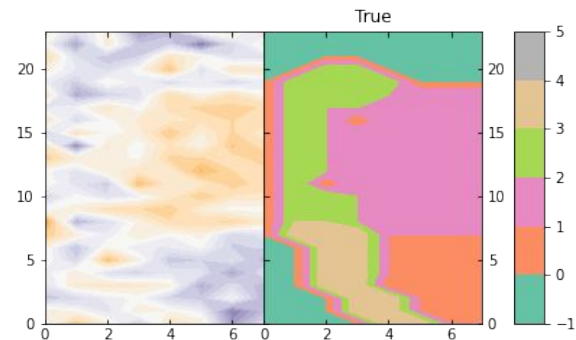
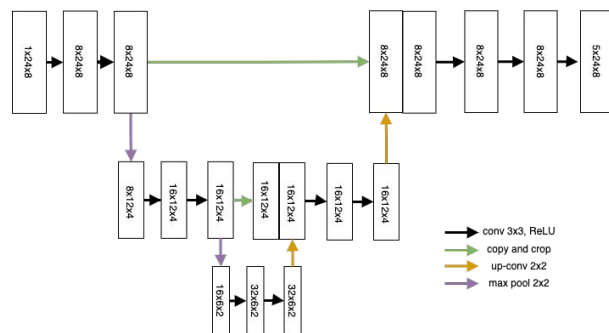
ML models can be used

- in pre-processing to run selectively run data analysis routines based on results from a ML detector
- to store detected features in a database
- in online visualization



- Magnetic island perturb the $\langle T_e \rangle$ profile as it rotates through ECEI view.
- Sampling $\delta T_e / \langle T_e \rangle$, ECEI detects radial phase inversion structures caused by magnetic islands.

- U-Net is a CNN encoder-decoder architecture
- Small model with $\lesssim 10k$ parameters.



Input is a 24x8 array of ECEI data

Output is a 24x8 array where each pixel is assigned a value that encodes which part of the image it belongs to.



<http://lb.test123.development.svc.spin.nersc.org/dashboard/newvue>

Goal: Connect experiments, scientists, compute using advanced networks to optimize scientific discovery for fusion energy sciences.

For this, are developing the Delta framework which reliable facilitates near real-time streaming analysis of big fusion data using distributed HPC resources and web-based interactive visualization tools.



- **If you would like to use DELTA for your analysis we want to hear from you.**

Future work will explore

- **Operationalize streaming data analytics for fusion: Develop a persistent service to be used in daily ops.**
- Develop a power-exhaust monitoring system
 - Add support for more data sources (f.ex. IR, Mirnov Coils, bolometer)
 - Add support for more advanced workflows: simulations that use measurements as input
- Explore use of industry streaming analysis software (f.ex. ray.io and apache foundation suite) at NERSC
- Coupling to IDA systems



Churchill et al. [2021](#)

Kube et al. [2021](#)

Kube et al. [2020](#)

Kube et al. [2020](#)

Choi et al. [2020](#)

Churchill et al. [2020](#)

Choi et al. [2016](#)

Choi et al. [2018](#)