



Can We Have HPC Data Science?

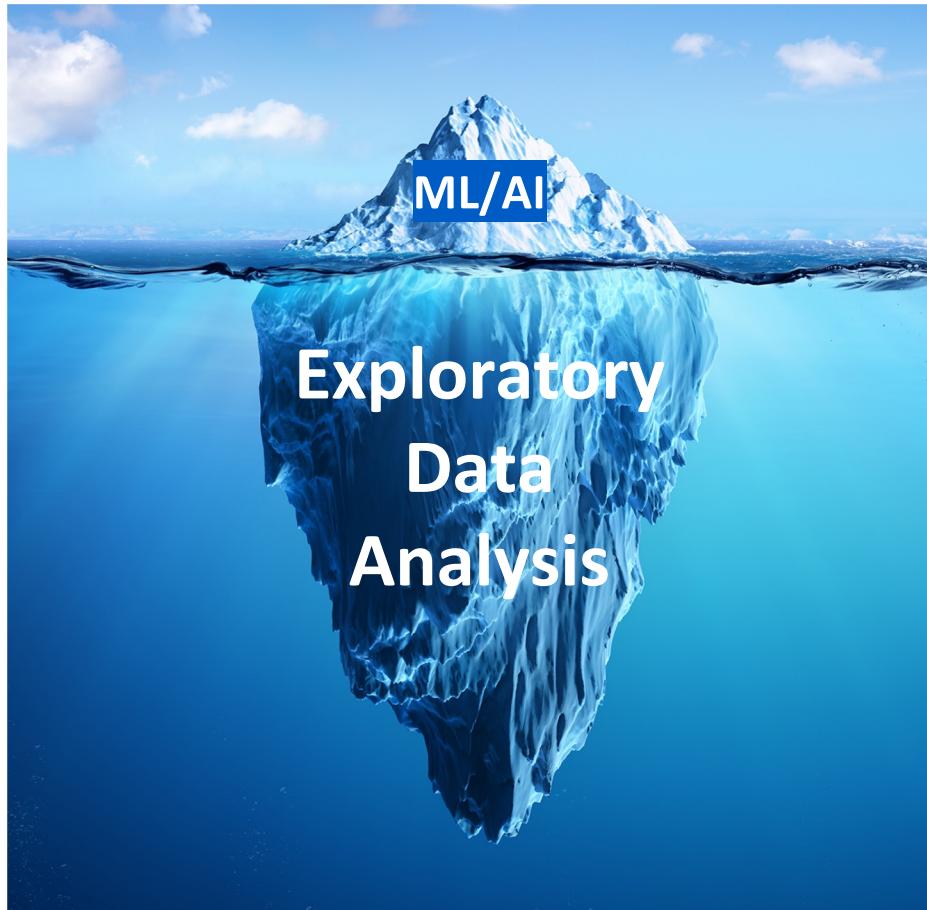
Dr. William Reus
US Department of Defense

Cryptanalytic Bombe at Bletchley Park, UK.
Image credit: Antoine Taveneaux, en.wikipedia.org/wiki/Bombe

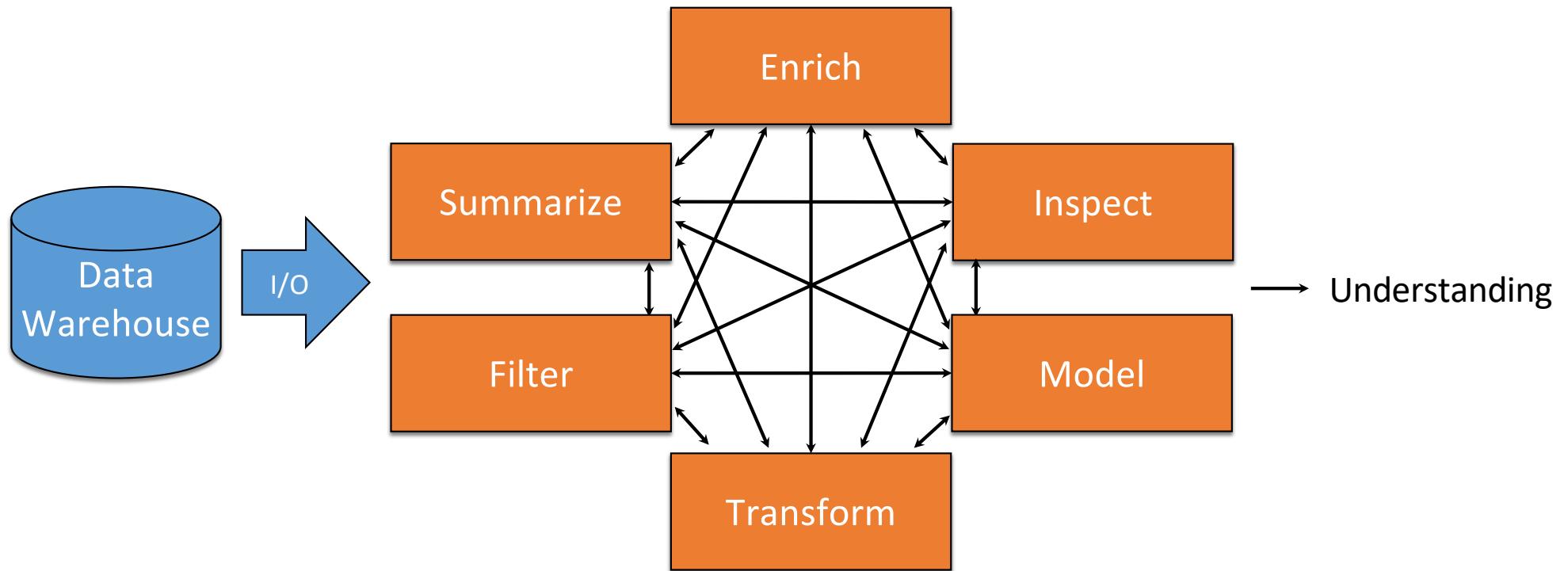
HPC → Data Science

1. HPC data science requires
 - a. Interactivity (EDA)
 - b. Data warehousing (ETL)
2. Challenges
 - a. EDA is bursty
 - b. HPC software is siloed
 - c. Data breaks everything
3. A culture shift
 - a. Think outside the slurm job
 - b. Adopt an ecosystem mentality
 - c. Honor data engineering

Data Science

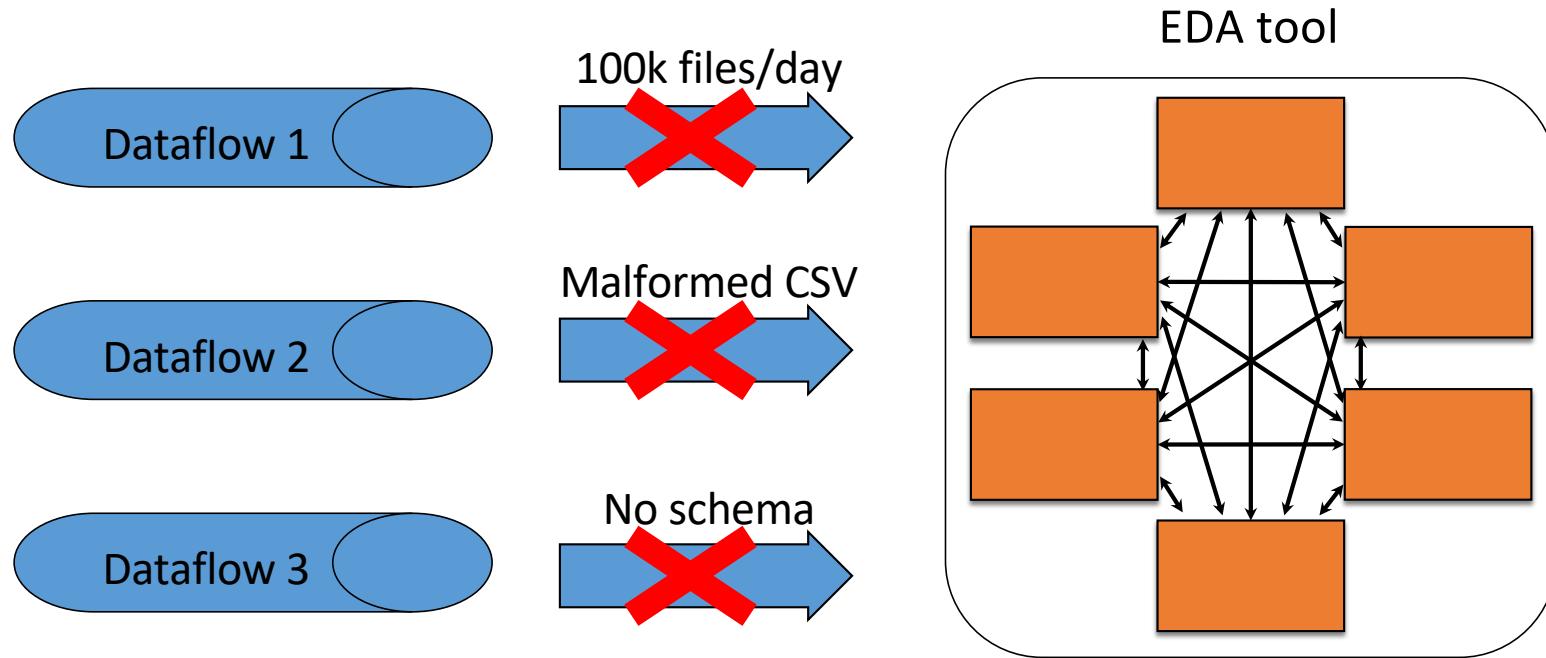


EDA: Learning the Language of a Dataset

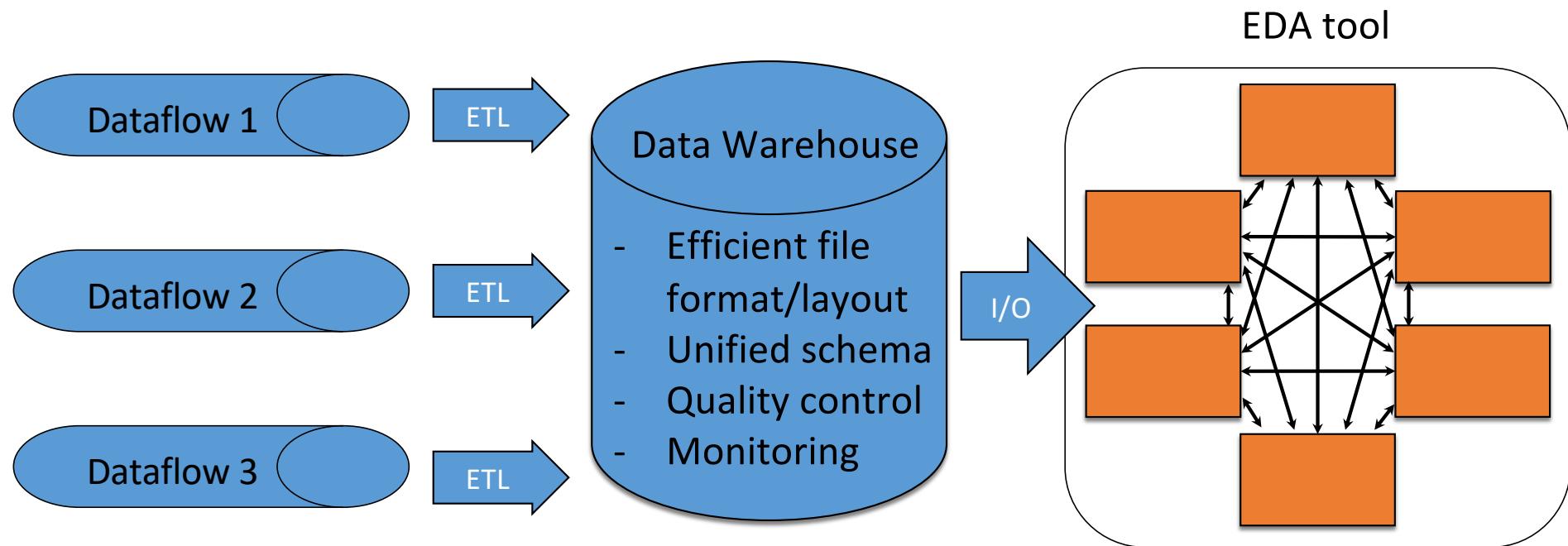


Synonyms: Data Cleaning, Data Wrangling, Data Modeling, Data Mining, etc.

Why Data Warehousing?

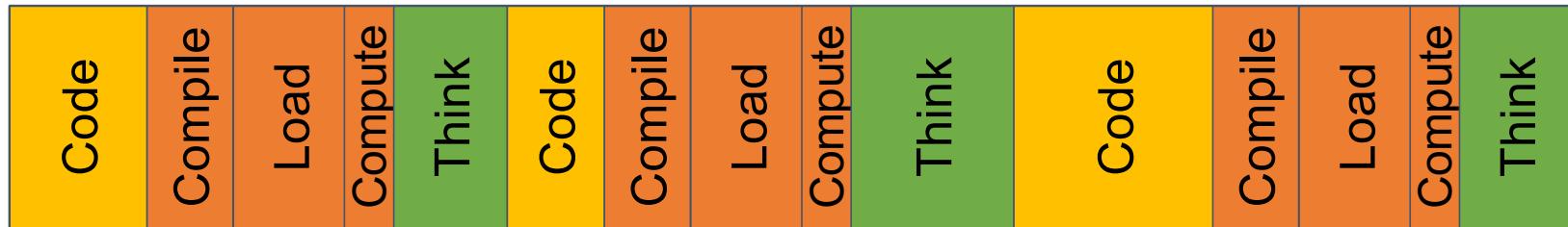


Why Data Warehousing?

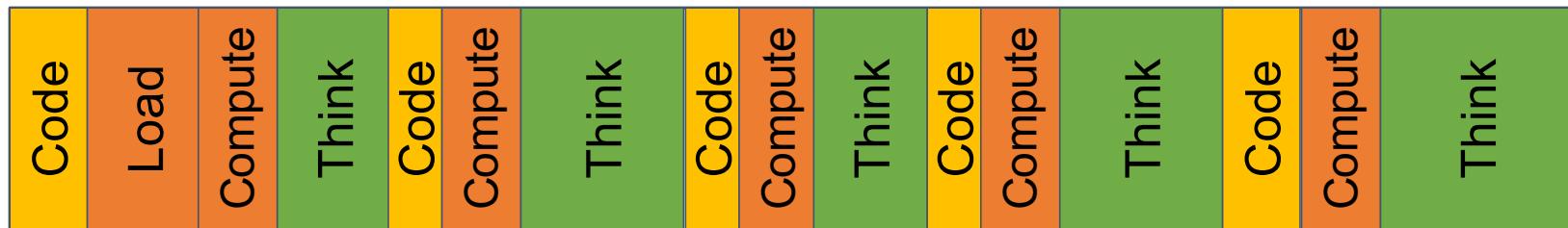


Why Interactive EDA?

Compiled workflow

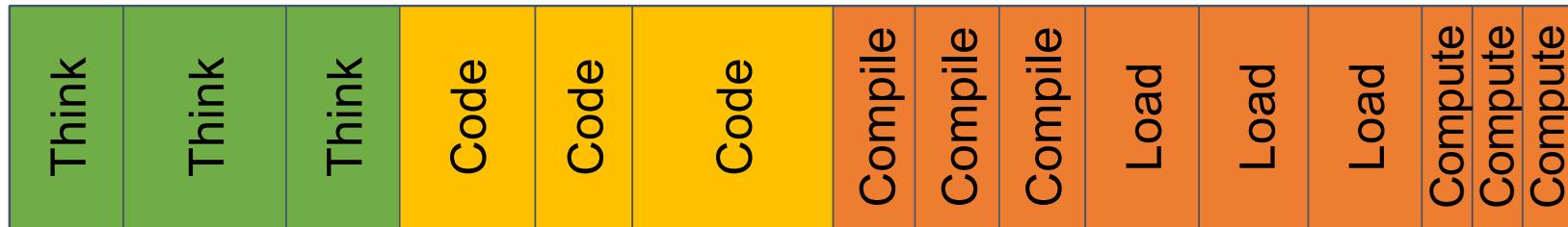


Interactive workflow

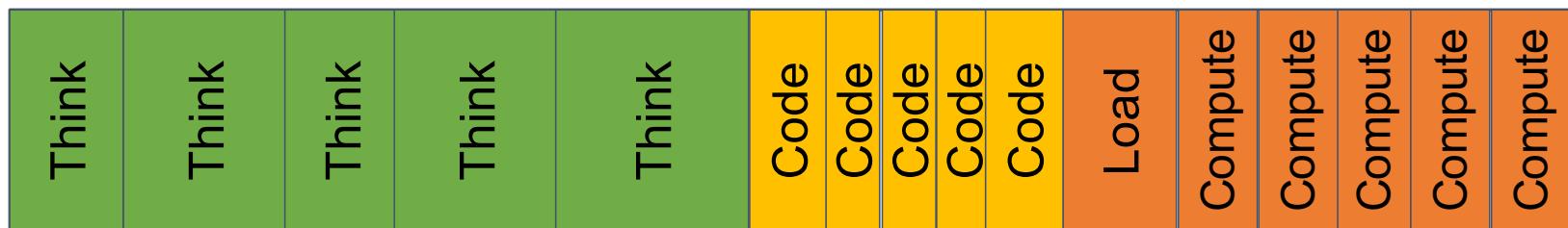


Difference 1: Overhead

Compiled workflow

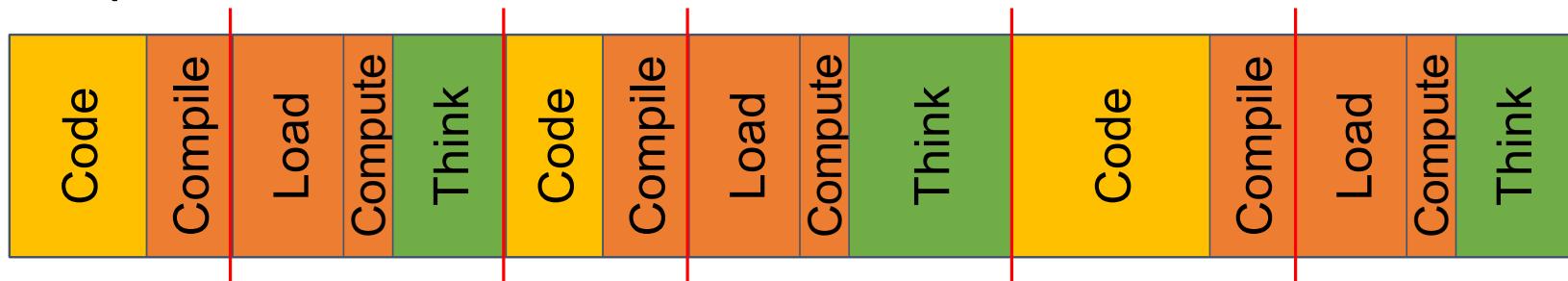


Interactive workflow

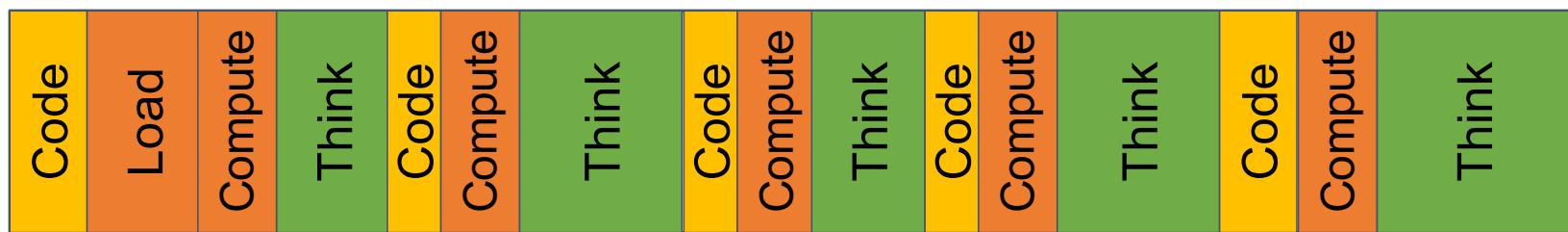


Difference 2: Context Switching

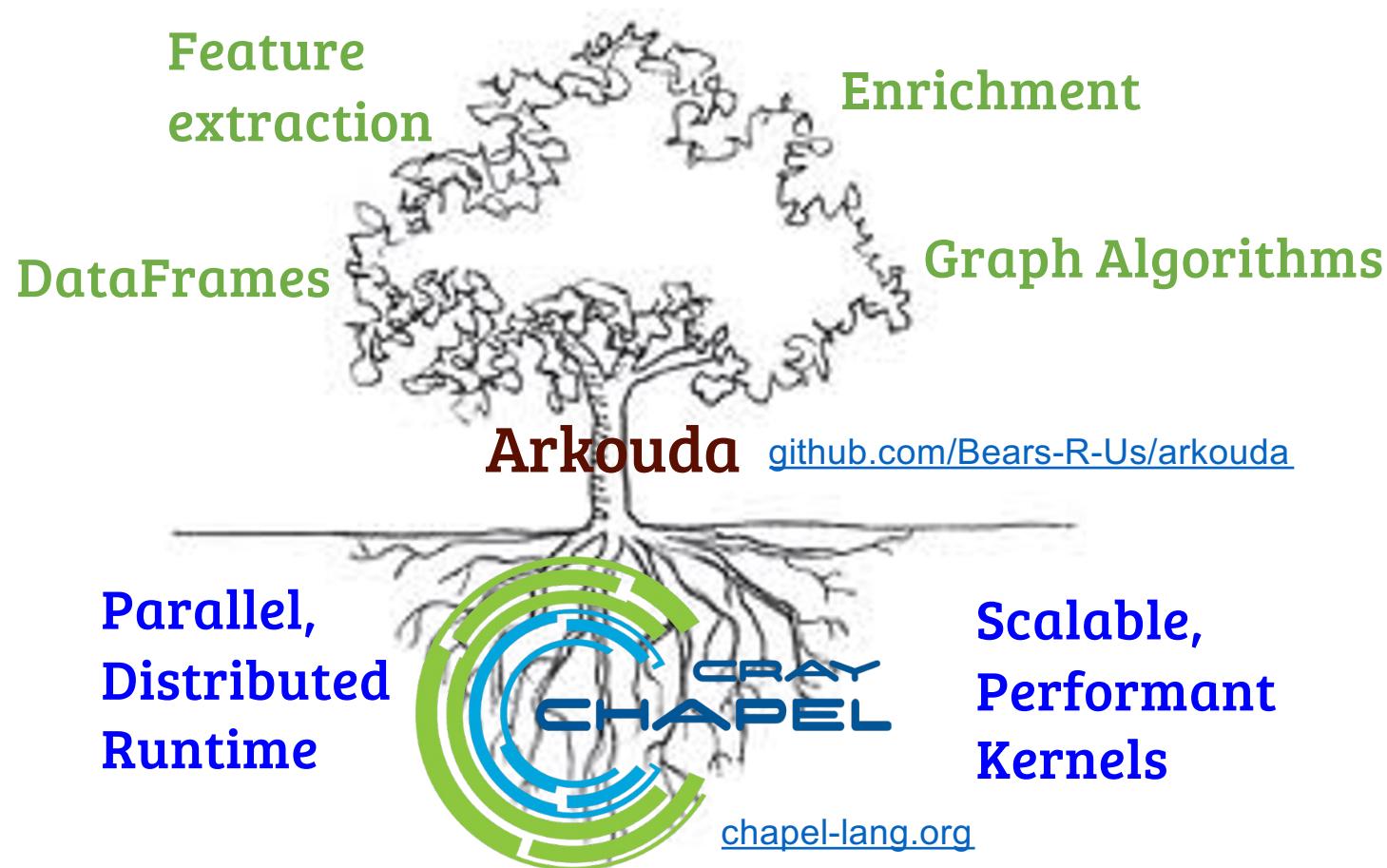
Compiled workflow



Interactive workflow



Arkouda: NumPy for HPC



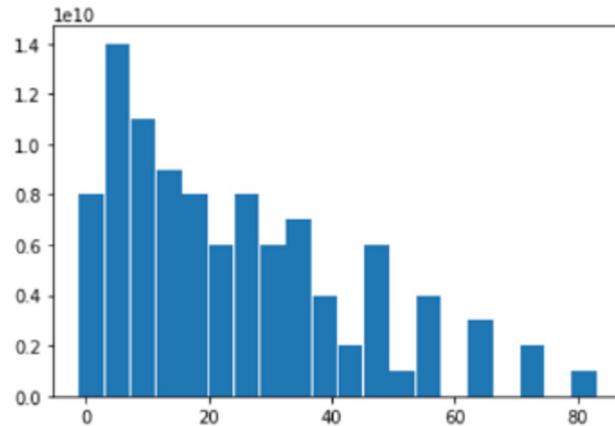
Interactive EDA

```
In [9]: A = ak.randint(0, 10, 10**11)
B = ak.randint(0, 10, 10**11)
C = A * B
hist = ak.histogram(C, 20)
Cmax = C.max()
Cmin = C.min()
executed in 3.96s, finished 13:45:28 2019-09-12
```

800 GB arrays that persist in distributed memory on compute nodes

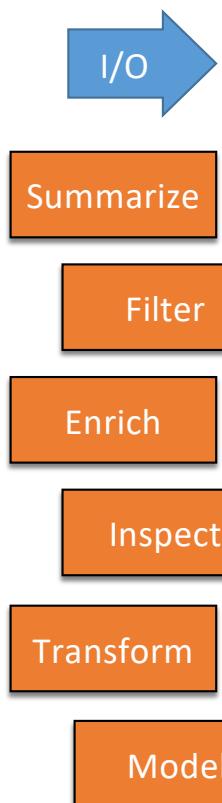
```
In [10]: bins = np.linspace(Cmin, Cmax, 20)
_ = plt.bar(bins, hist.to_ndarray(), width=(Cmax-Cmin)/20)
executed in 193ms, finished 13:45:28 2019-09-12
```

Parallel computations



Inline visualization

Interactive EDA



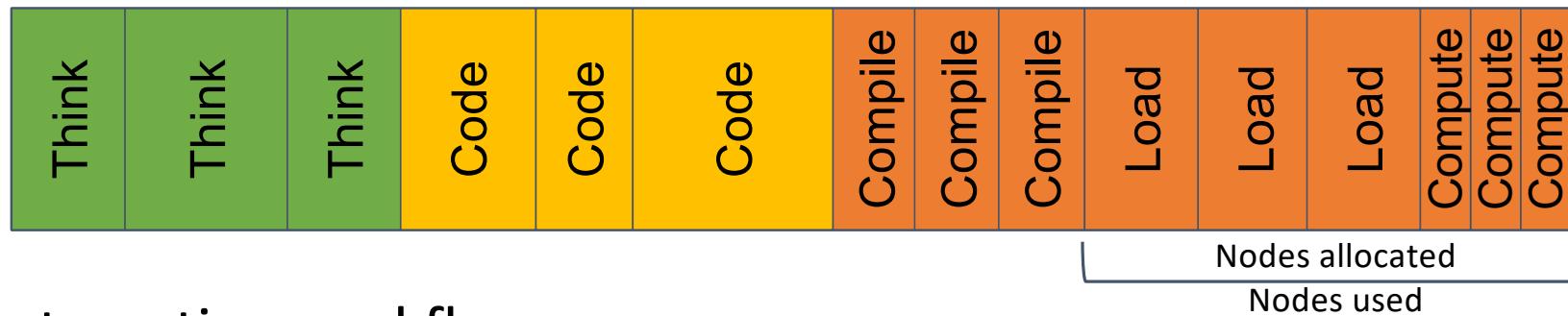
Operation	Example	Approx. Time (seconds)
Read from disk	<code>A = ak.read_hdf()</code>	30-60
Scalar Reduction	<code>A.sum()</code>	< 1
Histogram	<code>ak.histogram(A)</code>	< 1
Vector Ops	<code>A + B, A == B, A & B</code>	< 1
Logical Indexing	<code>A[B == val]</code>	1 - 10
Set Membership	<code>ak.in1d(A, set)</code>	1
Gather	<code>B = Table[A]</code>	4 - 120
Get Item	<code>print(A[42])</code>	< 1
Sort Indices by Value	<code>I = ak.argsort(A)</code>	15
Group by Key	<code>G = ak.GroupBy(A)</code>	30
Aggregate per Key	<code>G.aggregate(B, 'sum')</code>	10

Performance charts (updated nightly): <https://chapel-lang.org/perf/arkouda/>

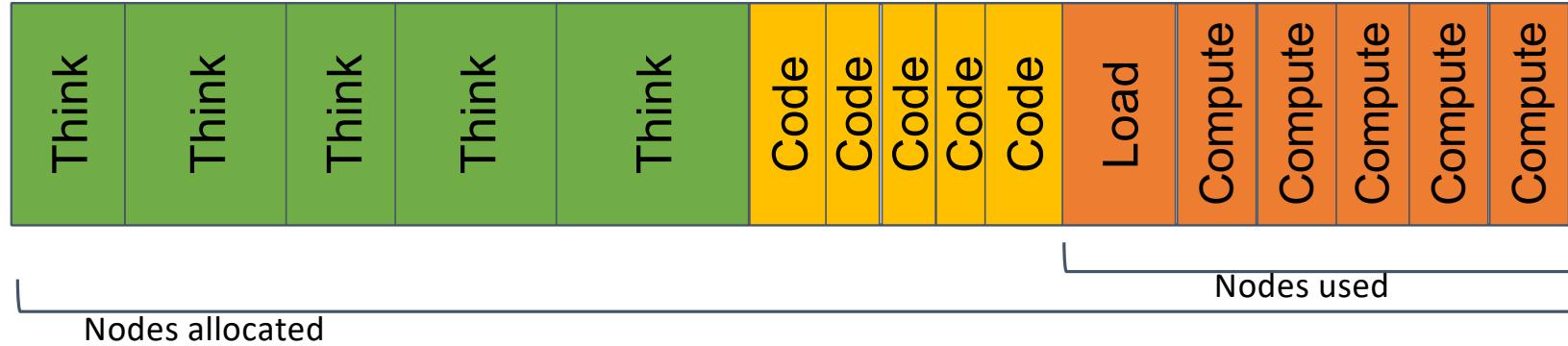
- A, B are 50 billion-element arrays of 32-bit values
- Timings measured on real data
- Hardware: Cray XC40
 - 96 nodes
 - 3072 cores
 - 24 TB
 - Lustre filesystem

Challenge: Trapped Capacity

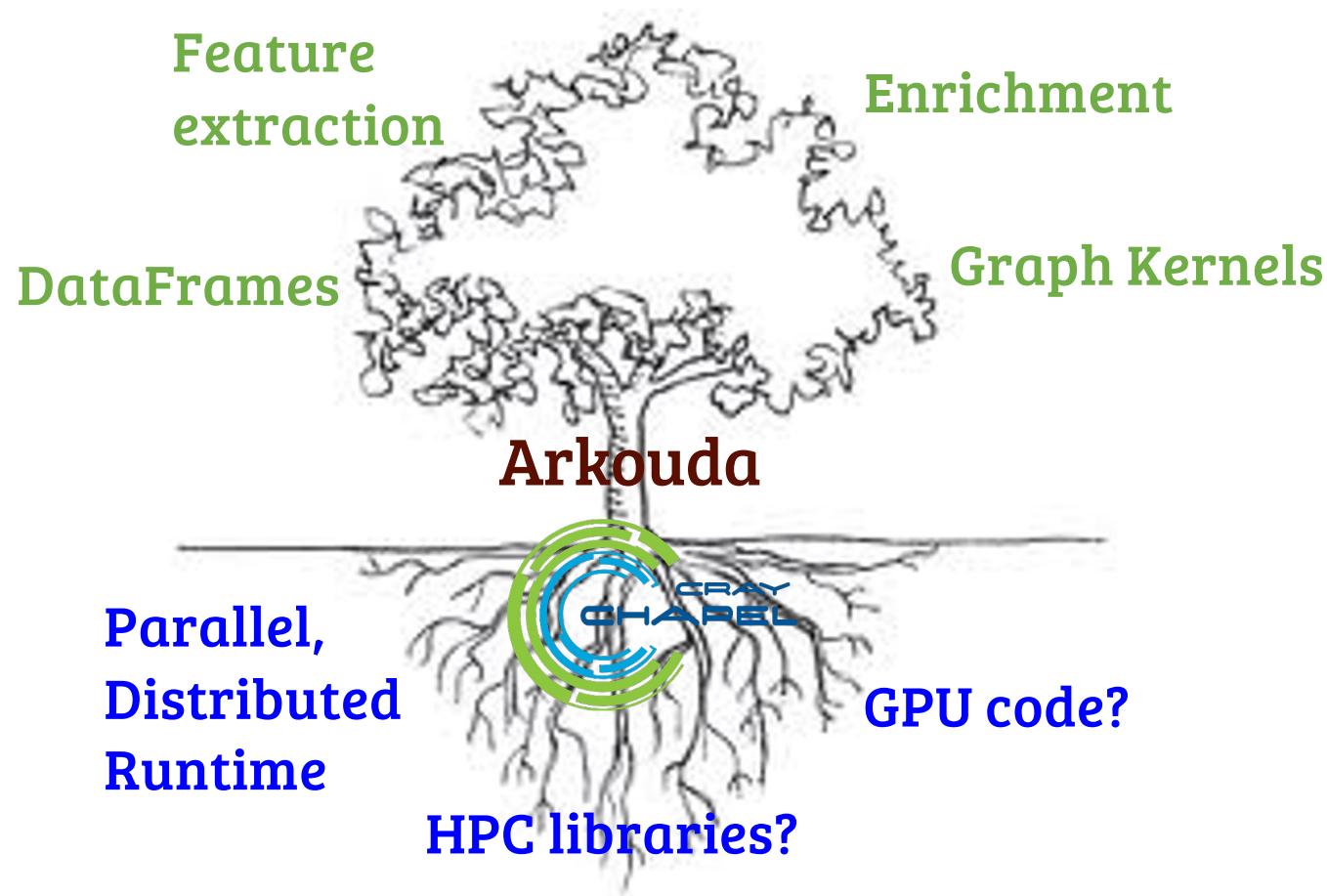
Compiled workflow



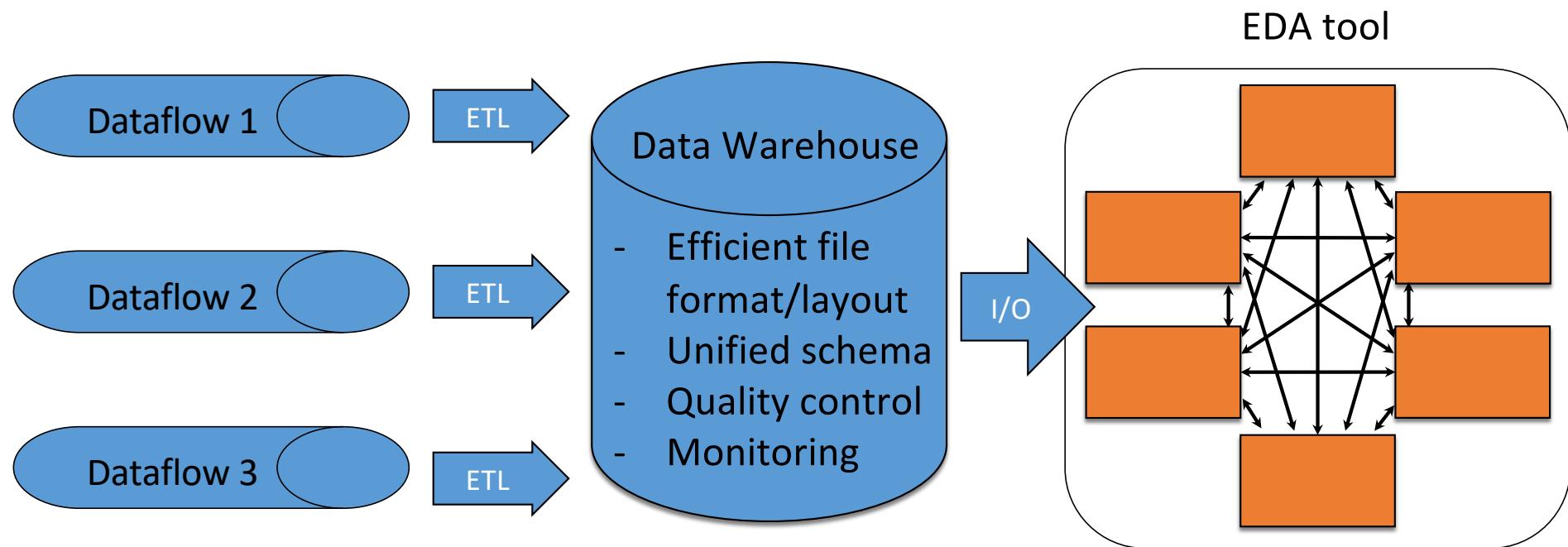
Interactive workflow



Challenge: Tool Integration



Challenge: HPC Data Engineering



HPC → Data Science

1. Think outside the slurm job
 - a. Vision: persist variables to fast FAM/FAS and release nodes between computations
 - b. Vision: interact with data already loaded by teammate
 - c. Example: Hierarchical object storage
2. Develop for the ecosystem
 - a. Vision: load and clean a graph in Arkouda, then call out to [HPC graph library], all within Jupyter notebook (no context switching!)
 - b. Example: require standard data interfaces/layouts
3. Honor HPC data engineering
 - a. Vision: data warehouse with cloud-like reliability and HPC scalability
 - b. Example: professionalize and incentivize HPC data engineering like we do HPC software and hardware engineering

Acknowledgements

- Arkouda developers and users
- Chapel team at HPE
- Partners in academia and national labs



chapel-lang.org



github.com/Bears-R-Us/arkouda