

# ATLAS Analysis Model & the HL-LHC Conceptual Design Report

March 17 2020

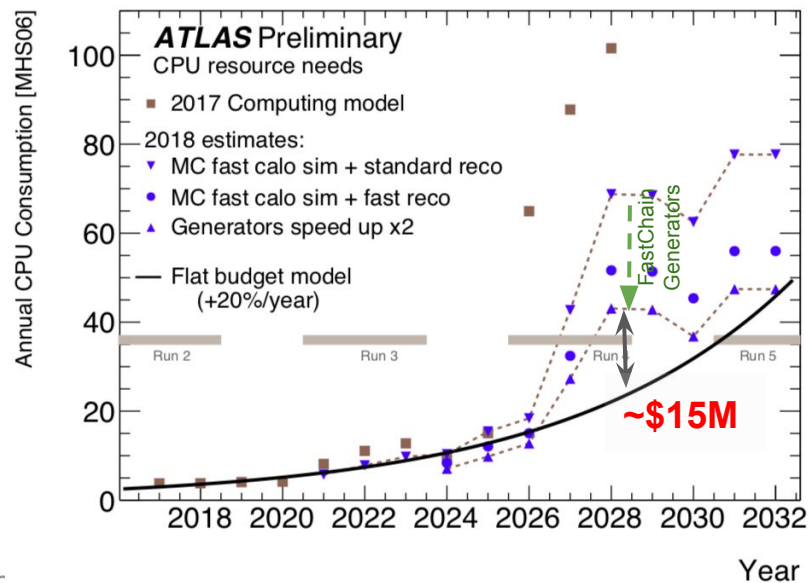
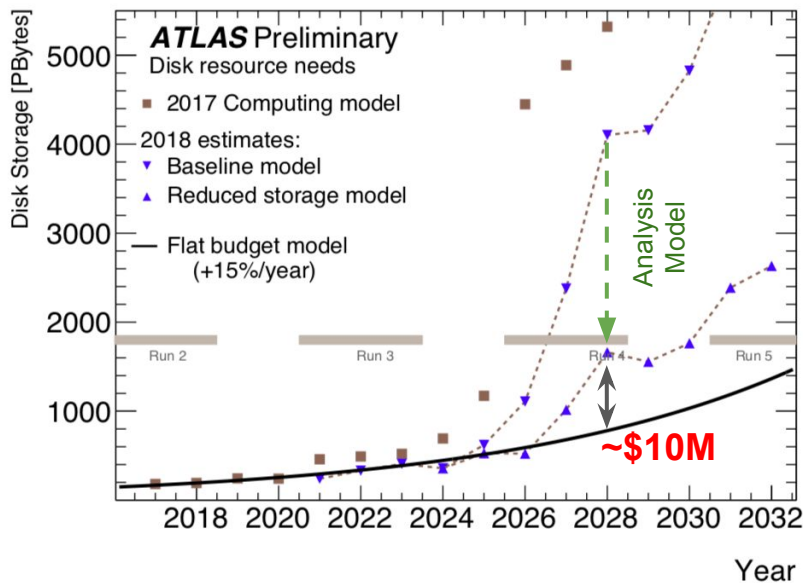
Paolo Calafiura (LBNL)

With lots of input from Davide Costanzo (Sheffield) and Wei Yang (SLAC)

1. ATLAS HL-LHC Computing Model and the Conceptual Design Report
2. US ATLAS Analysis Facilities (a.k.a. Shared T3s)
3. Personal musings on the evolution of Analysis Facilities

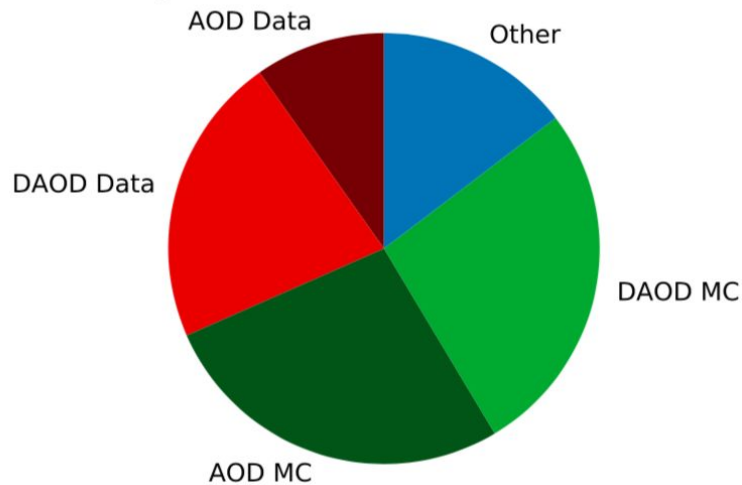
# The Infamous Resource Plots

- LHC Phase-2 computing was advertised as a difficult problem to solve
  - The CPU/disk needed is above what we expect ([link to plots](#))
  - Current model presented on 4 Dec 2018 at ATLAS weekly ([link](#))
- Update of the model (and plots) will come this Spring.
  - **Several ideas for the CPU problem, fewer ideas for the disk problem (harder!)**
  - **Tradeoffs: memory-cpu, disk-network, cpu-disk.**

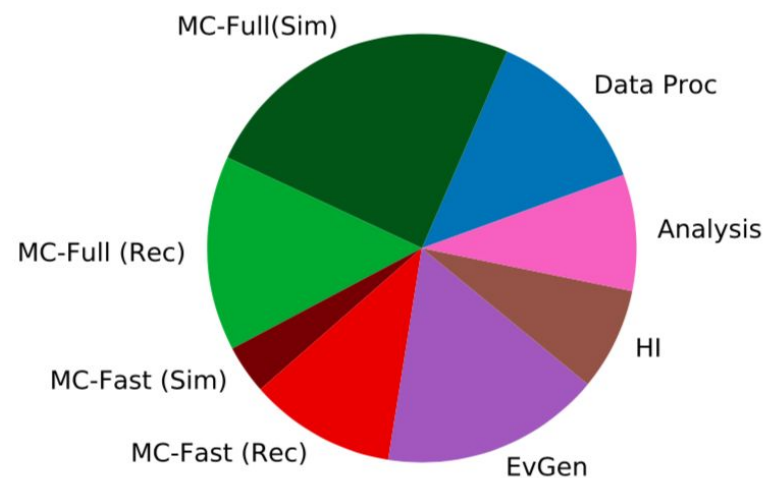


- Almost 90% of projected Run 4 disk usage
  - Over 50% of disk dedicated to MC
- Non-negligible CPU impact

**ATLAS** Preliminary. 2028 Disk resource needs  
Reduced storage model

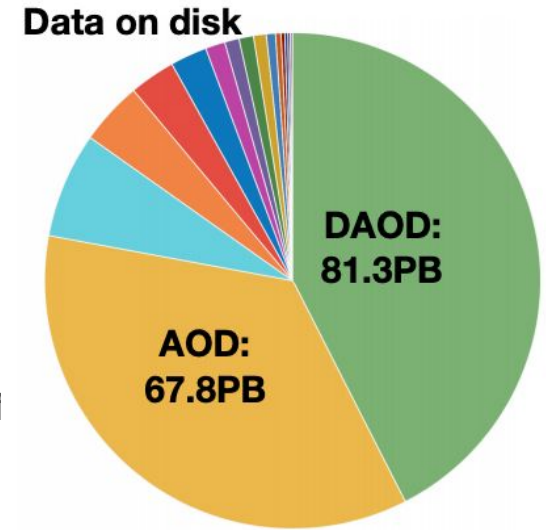


**ATLAS** Preliminary. 2028 CPU resource needs  
MC fast calo sim + fast reco, generators speed up x2

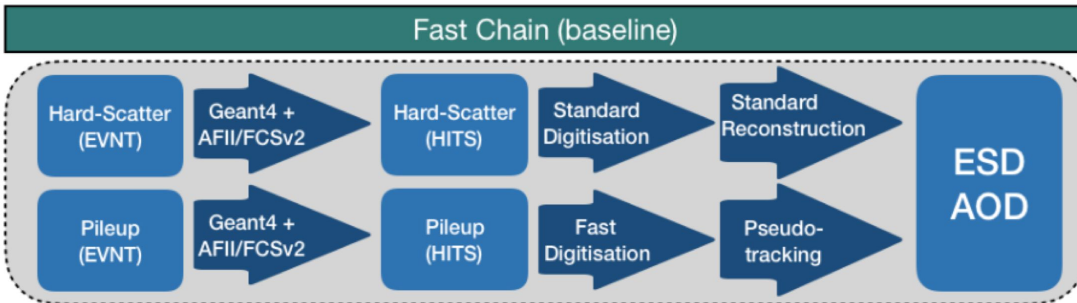


# Analysis Model Evolution

- Most of the disk space is used by analysis formats (AOD and DAOD)
  - Analysis model has a direct impact on disk resources
  - Run-2 analysis model was too “expensive”
  - Analysis Model Study Group for Run-3 (aka AMSG-R3) provided recommendations for Run-3 ([CHEP 2019](#))
  - Most analysis to move to a new DAOD\_Phys format
- Further evolution for Run-4
  - Use a smaller pre-calibrated format - DAOD\_PHYSLITE
  - Large use of FastChain. Are AODs needed as persistent MC format 1



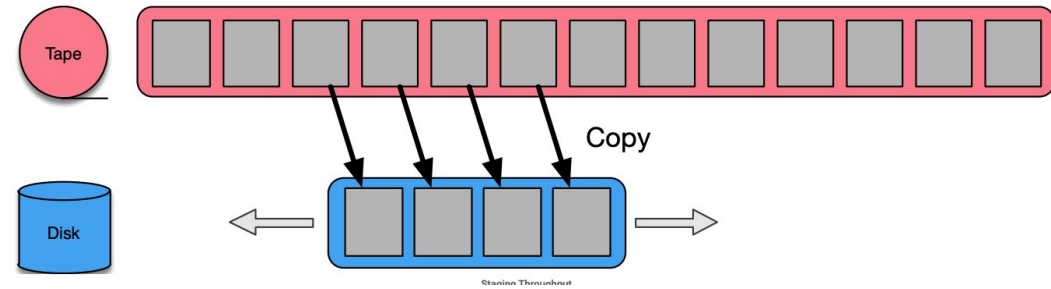
Run 2



# Idea: Data carousel. Store AOD on tape

## 'data carousel'

- Use tape as storage medium for AODs
- Recall AODs from tape in a rolling buffer
  - Retrieve only data used by workflow
- Process data (or MC) from disk once available there
- Reduce number of AOD copies needed on disk



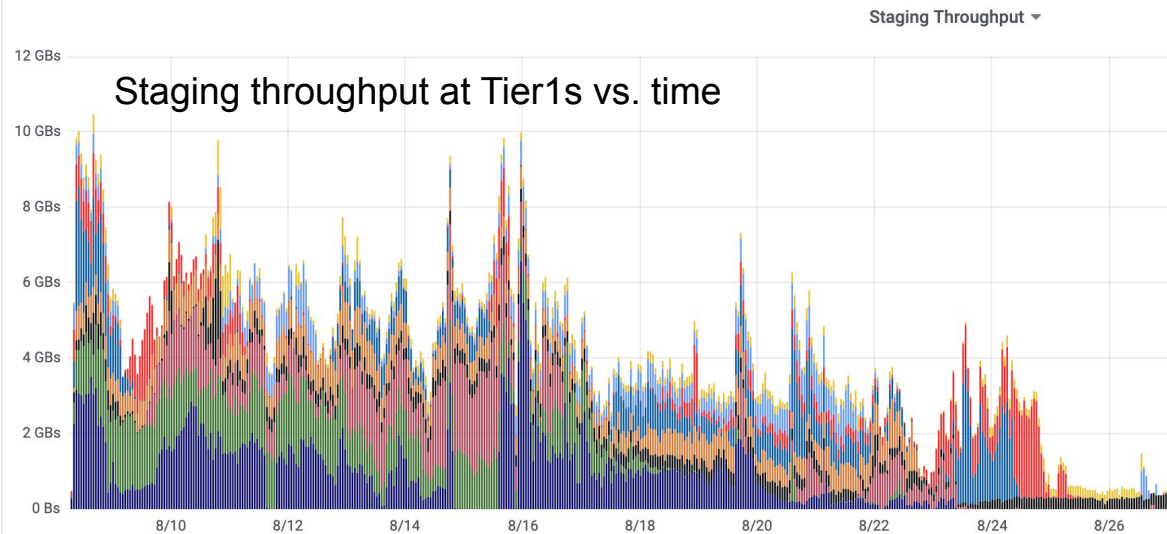
## Tested tape performance in production For AOD Derivations

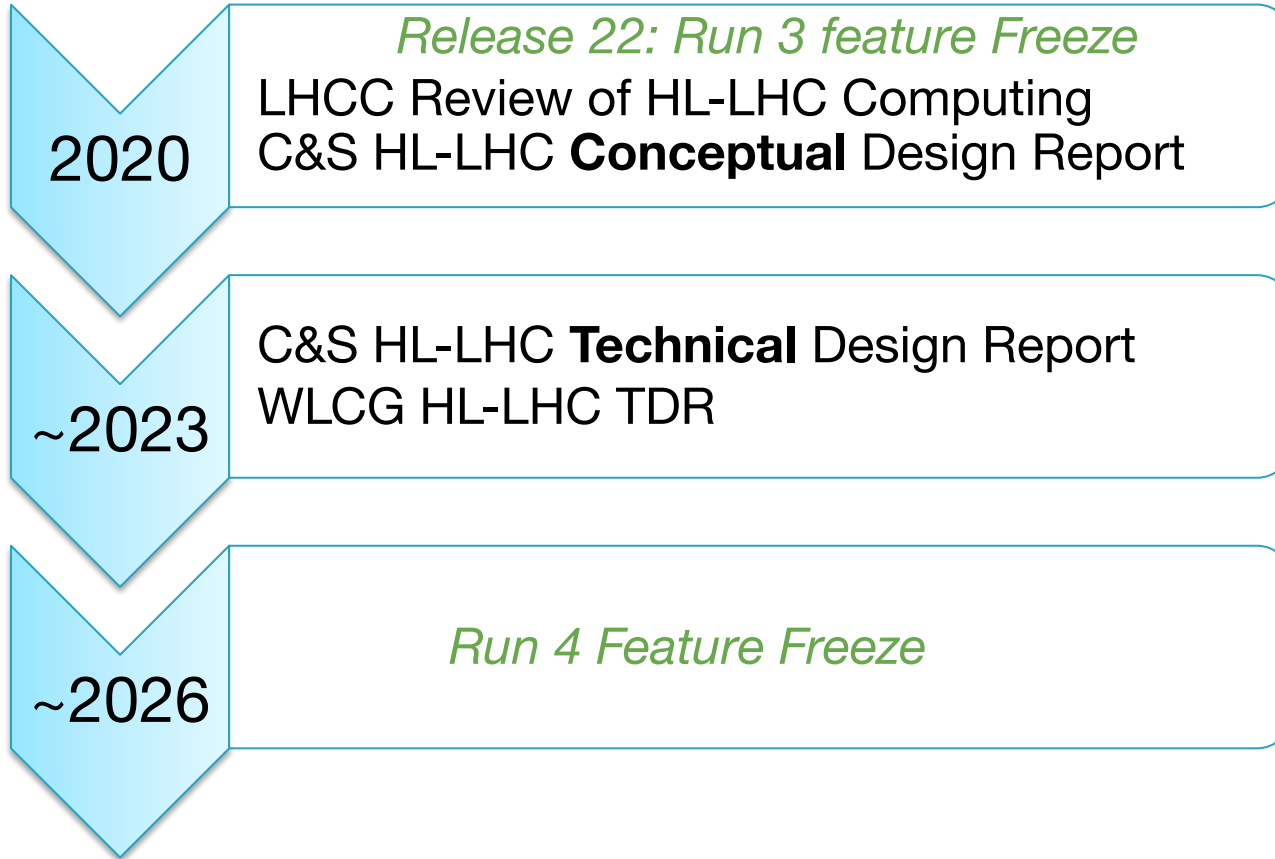
- Encouraging results: it worked

## AOD access from disk is 100 PB/month

- For tape access we have to schedule access orderly
- Ongoing R&D and technical work to automate this on the data and job management sides

## Data Carousel "commended" at last CRSG





# US ATLAS Analysis Facilities

## In Production:

- Two Analysis Facilities (a.k.a Shared Tier 3s) in operation @ SLAC and BNL for several years
- So far mostly use retired hardware from BNL T1 and former WT2 (SLAC Tier 2)
  - Direct FTE funding from US ATLAS
- The two T3s are part of US ATLAS Computing Facility and provide Physics Analysis Support
- No clear division between these two types of tasks and personnels.
  - All work together. Coordination meetings across WBSs

## Pre-production Facilities:

- Cal State Fresno Virtual T3
  - An all-AWS Analysis Facility
- Univ. of Chicago ML(+Analytics) Platform
  - GPU resources for all ATLAS users <https://www.atlas-ml.org/>



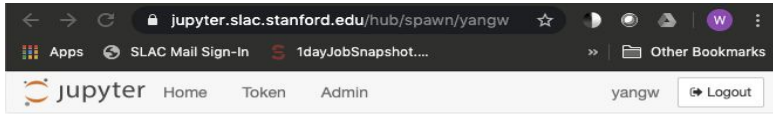
# Role of the Analysis Facilities

Directly serve US ATLAS physicists.

- Login, local batch, have home directory and data directory at AFs.
  - Similar to familiar lxplus environment
- **Users need support**
  - from how to use unix, to how to use batch, to how to run rucio and Athena

Complement grid based analysis capability.

- **Aim at easy**-to-use, easy-to-understand
- Optimized for **late stage data analysis**, batch and/or interactive
- **Quick turnaround** time
- Support both traditional (root,prooflite) and modern (scipy, jupyter) data analysis platform
- Support rucio and PanDA to scale out big tasks to grid



## Spawner Options

### LSST Isstsqre/sciplat-lab Images

updated at Thu Oct 31 16:30:45 2019 UTC

- Recommended
- Latest Weekly
- Latest Release
- Weekly 2019\_43
- Weekly 2019\_42
- Weekly 2019\_41
- Release 18\_1\_0
- Release 18\_0\_0

### SuperCDMS Images

- CDMS JupyterLab Stable - v 1.8
- CDMS JupyterLab Unstable - Testing only

### ATLAS Images

- ATLAS Jupyterlab Image - v01
- ATLAS Jupyterlab Image - v01 (GPU)

### SLAC Machine Learning Images

- SLAC JupyterLab Image (GPU) v20200211.0
- SLAC JupyterLab Image (GPU) v20191101.0
- SLAC JupyterLab Image (GPU) v20190712.2
- SLAC RAPIDS JupyterLab Image (GPU) v20190719.1
- SLAC RAPIDS JupyterLab Image (GPU) v20200212.0

Spawn

Login via SLAC unix account

A list of ATLAS images are added at here

- Notebook with PyROOT (C++ or Python)
- uproot (pure python)
- root\_numpy

←

← If you want GPUs

Start your Jupyter container

# Recent US AFs survey and other data

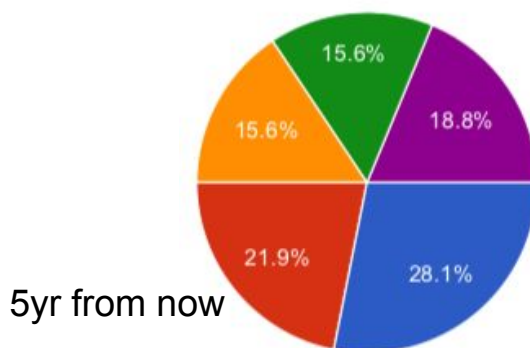
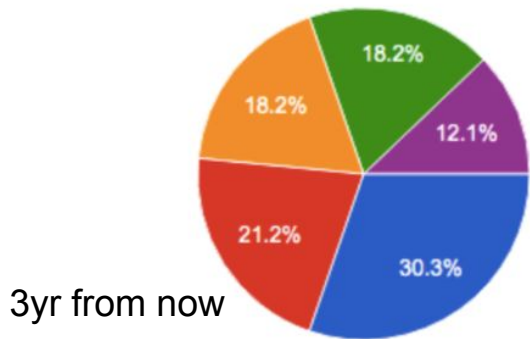
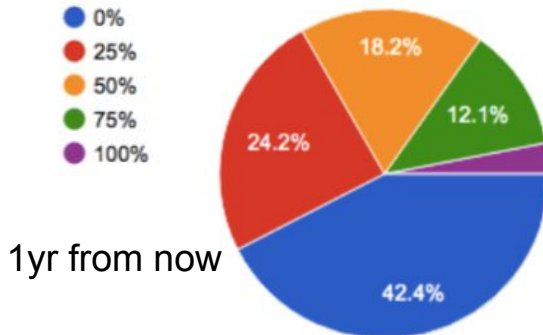
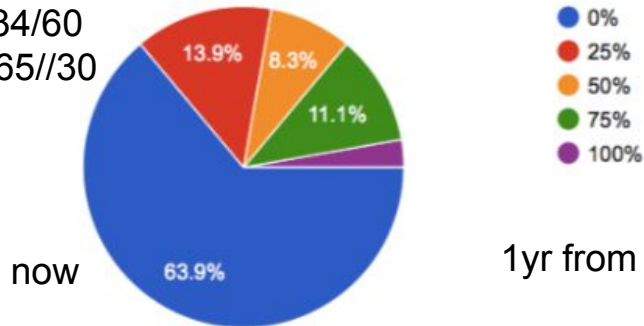
Survey Question: (excluding desktop and Grid usage)  
**What percentage of your group's computing do you expect to do at the US shared T3s now, 1yr, 3yr, 5yr from now**

36 responses

# users / active users

BNL: 134/60

SLAC: 65//30



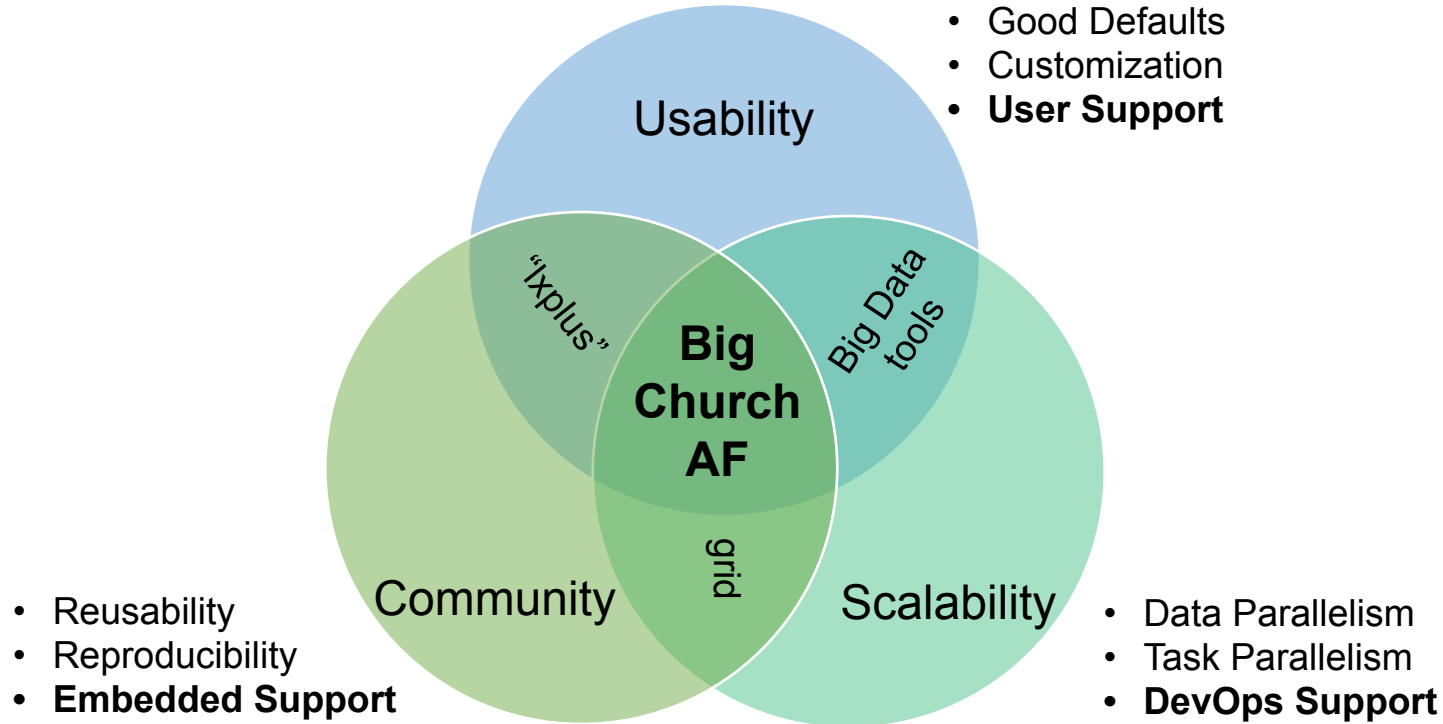
BNL and SLAC active user list show that almost no user actively use both AFs.

- Plenty of resources available
- Migration overhead?

Geographic locations seems to play a role - most of the SLAC users are on the West Coast.

- Looking for personalized support?

- Why aren't US ATLAS AFs as crowded as Ixplus?
- Why isn't everyone using the (clearly superior) scientific python stack?
  - *My guess:* Users don't have time to learn how to be efficient: they **stick to defaults**
    - Migrating a community (e.g. an analysis group) is even harder: **shared knowledge**
- If we want AFs to succeed we need to overwhelm users with support:
  - User-level support:
    - From beginner defaults, to cluster configurability
      - We are doing this, need more effort, particularly in entry-level tutorials
  - Embedded support:
    - Analysis groups are high-value customers, embed one “**account executive**”
      - Ideally a **50/50 physics/tools** person with knowledge & contacts on both sides
        - These people are precious but do exist, we need a **career path** for them
  - DevOps support:
    - Cutting edge analysis tools are hard to deploy at scale
      - Fledgling **DevOps community**
        - Highly marketable skills, need to create a **pipeline** and a rewarding career path for those who chose to remain
- Challenge: can we provide this level of support across a Distributed Facility?



# Backup

We asked many questions to the ATLAS Upgrade Physics community:

(My) highlights from six groups replies:

- Precision measurements will likely require **full sim** (egamma)
- Systematics often **limited by MC accuracy** not statistics (egamma, taus)
- HH(bbtt) currently **limited by MC statistics**
- **ML** classifiers expensive to train. **Full sim** needed (taus, HDBS)
- **MC needs** expected to **scale linearly** with lumi. (Exotics)
- **NNLO and NNNLO** need will increase (Exotics)
- **TLA** for photons and b-jets (Exotics)
- **DAOD\_PHYS\_LITE** good enough for tau **systematics**