

Overview: XENON needs and challenges

Chris Tunnell [Rice U.]

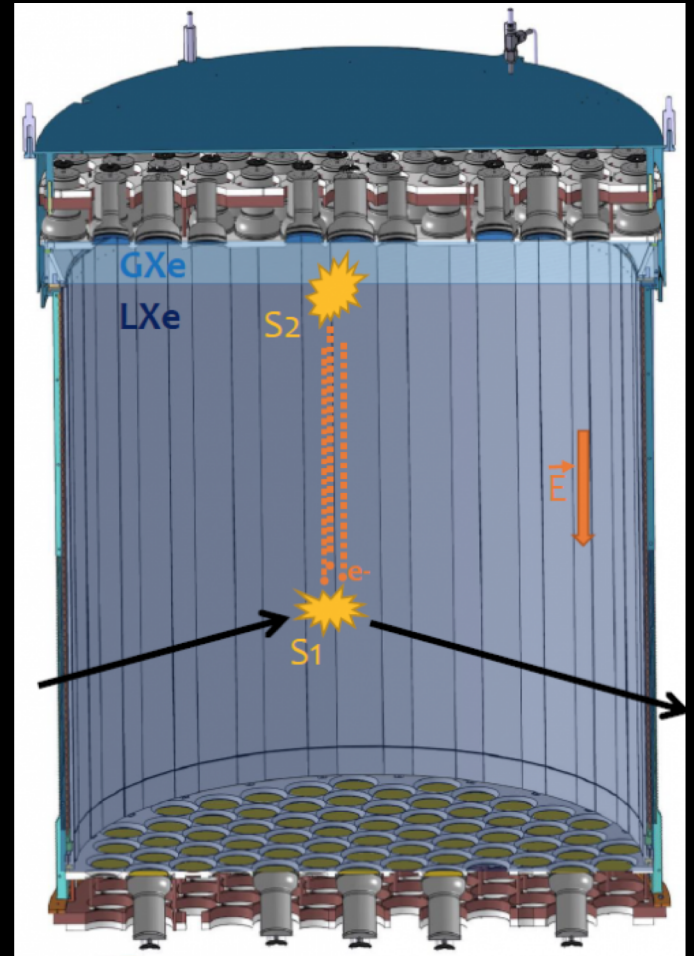
Oct 1, 2019

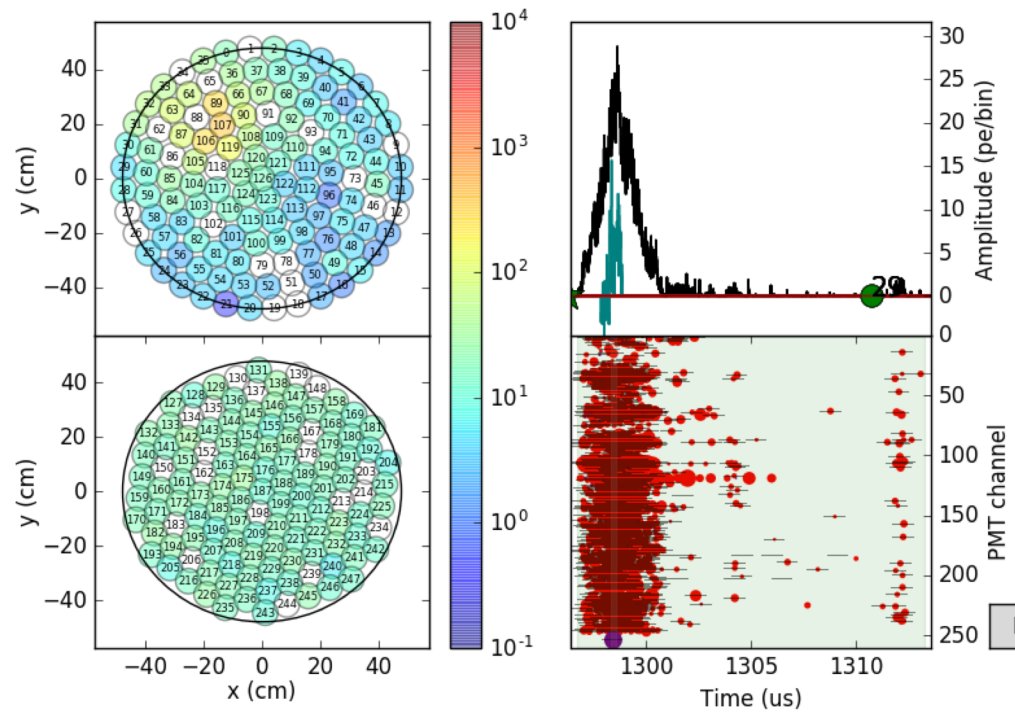
DANCE 2019



What is XENON?

Liquid XENON dark matter detector instrumented with 248 photomultipliers and 10-ns flash ADCs. We make a world-leading new experiment every few years.

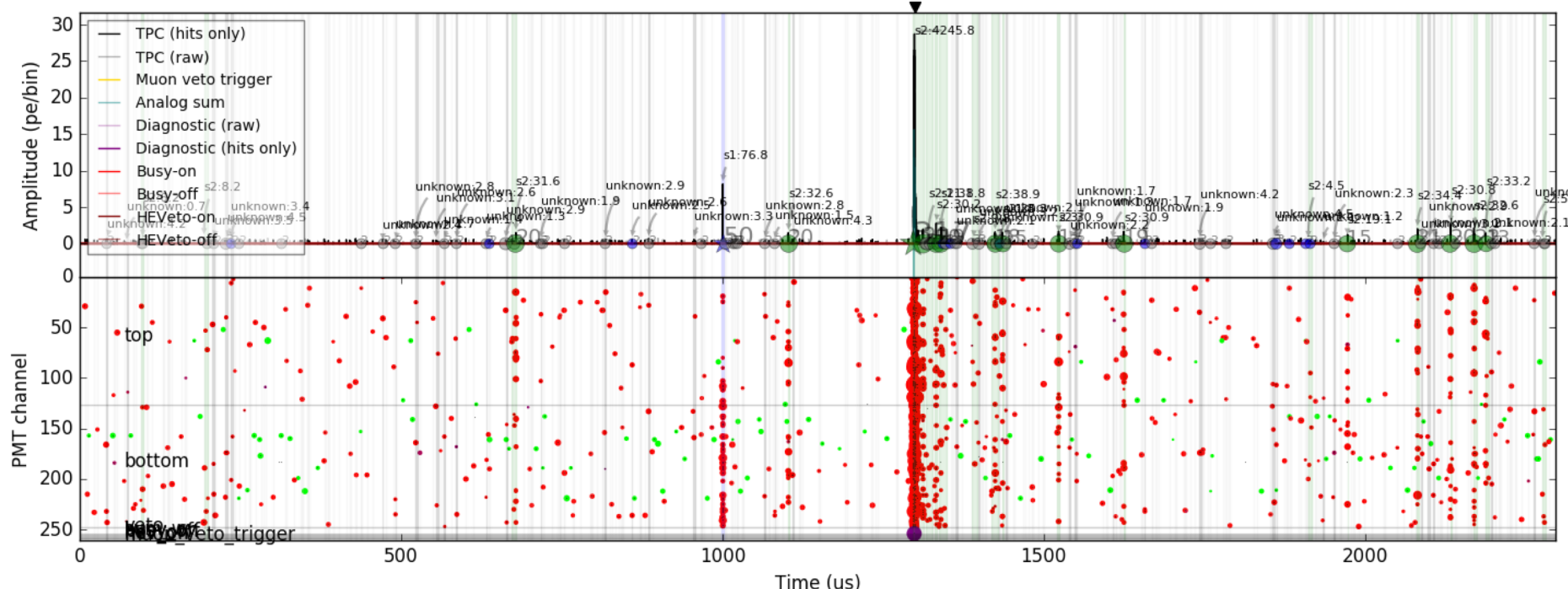




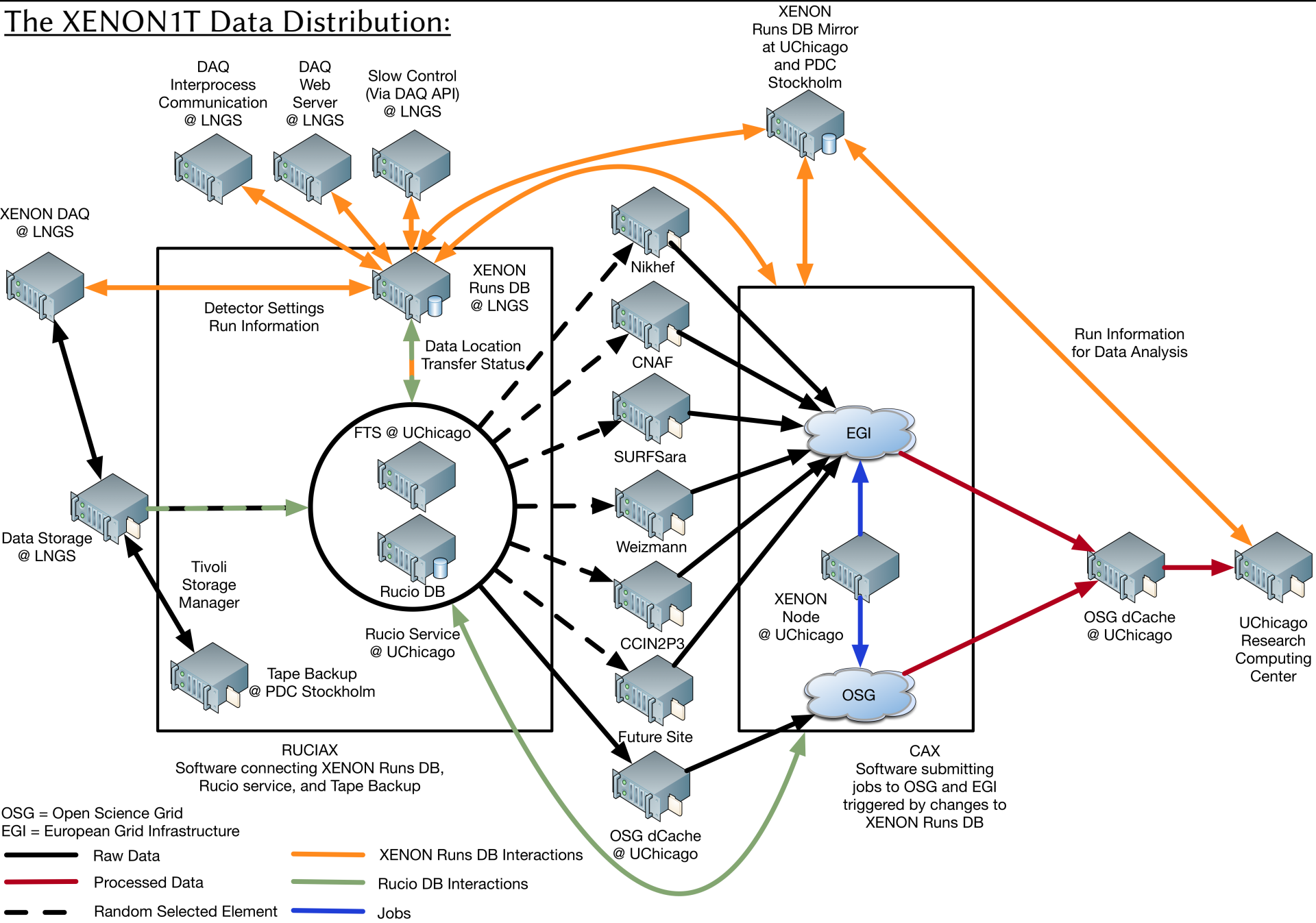
Event recorded at 2016/12/28, 17:47:14 UTC, 018090496 ns
 Suspicious channels (# hits rejected):
 52 (3), 63 (7), 84 (4), 122 (5), 126 (3), 138 (5), 141 (5), 143 (4),
 153 (4), 157 (8), 160 (3), 161 (9), 165 (3), 170 (3), 177 (4), 180
 (7), 189 (5), 199 (5), 211 (4), 212 (6), 219 (6)

Selected peak: s2 at 129675-131326, mean hit time 1299.01us
 Area: 4245.76 pe, contained in 2024 hits in 207 channels
 Fraction in top: 0.60
 Peak widths: hit time std = 1844ns,
 50% area range = 1110ns, 90% area range = 3128ns
 Top hitpattern reconstruction: (-17.58, 20.57), gof 366.6.
 Top spread: 22.6cm, Bottom spread: 32.1cm

Prev peak Next peak Main S1 Main S2

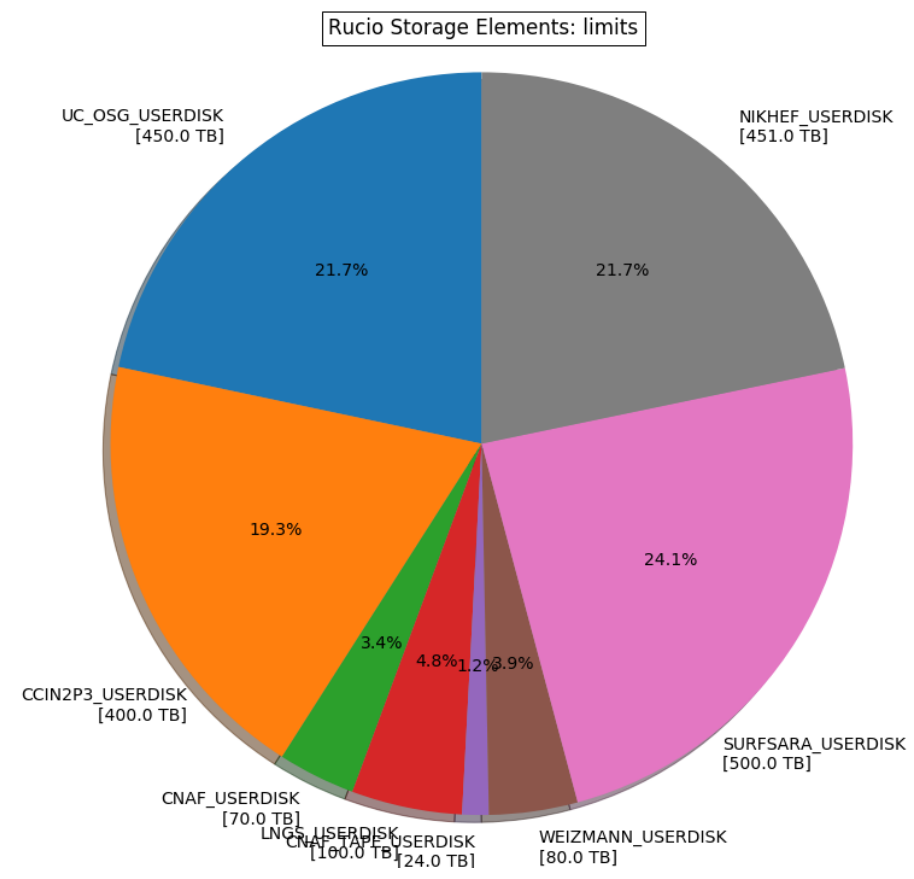
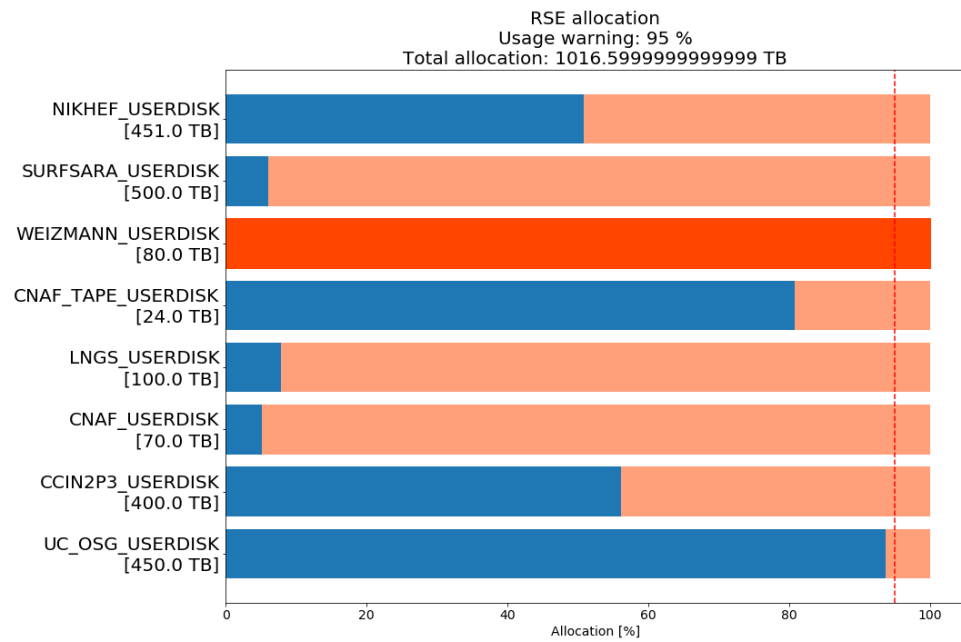


The XENON1T Data Distribution:



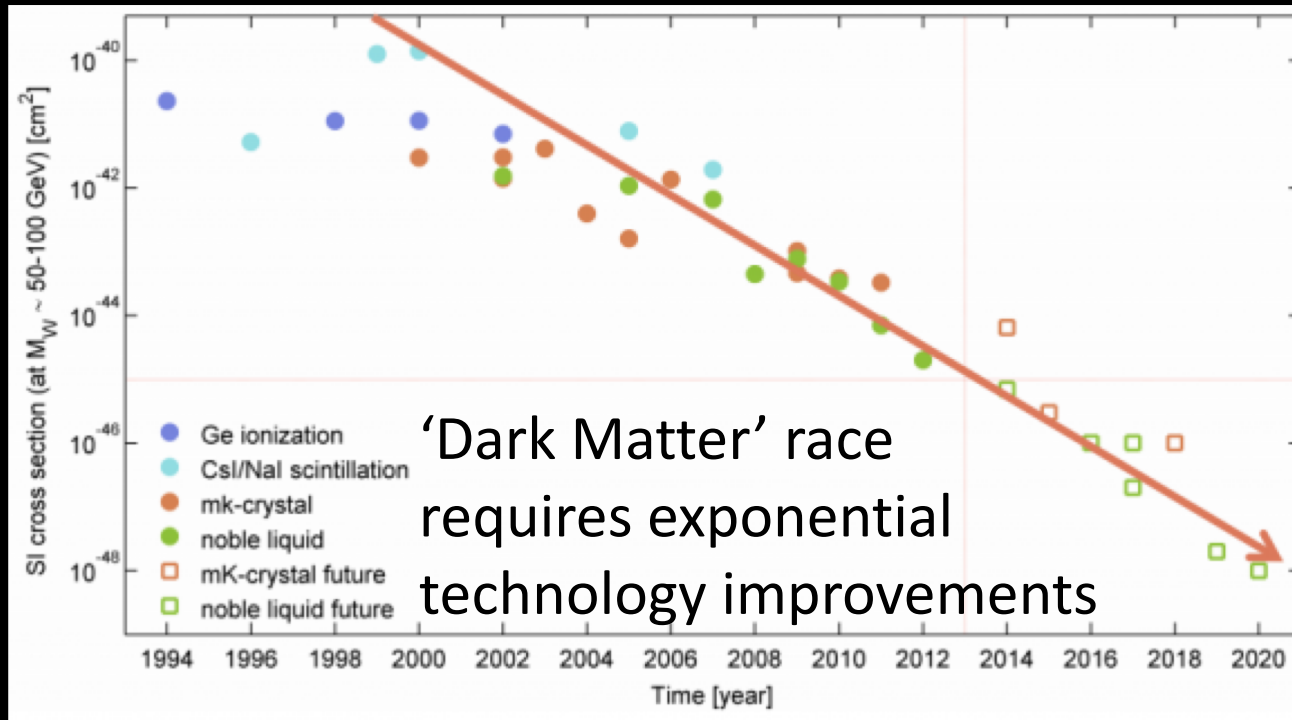
The XENON1T Disk Allocation and Requirement

- Data have two copies:
 - US: OSG dCache at UChicago (hold only relevant data)
 - Europa: One of several computing centers
- Tape copy in Stockholm Independent from Rucio

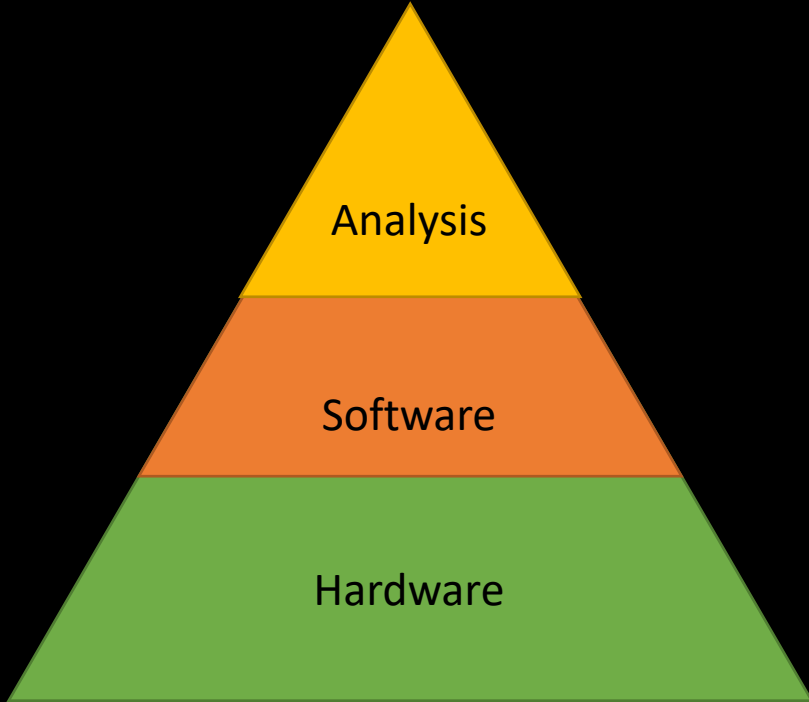


- In total: ~2 PB available
- Distributed worldwide
- Connected to computing centers

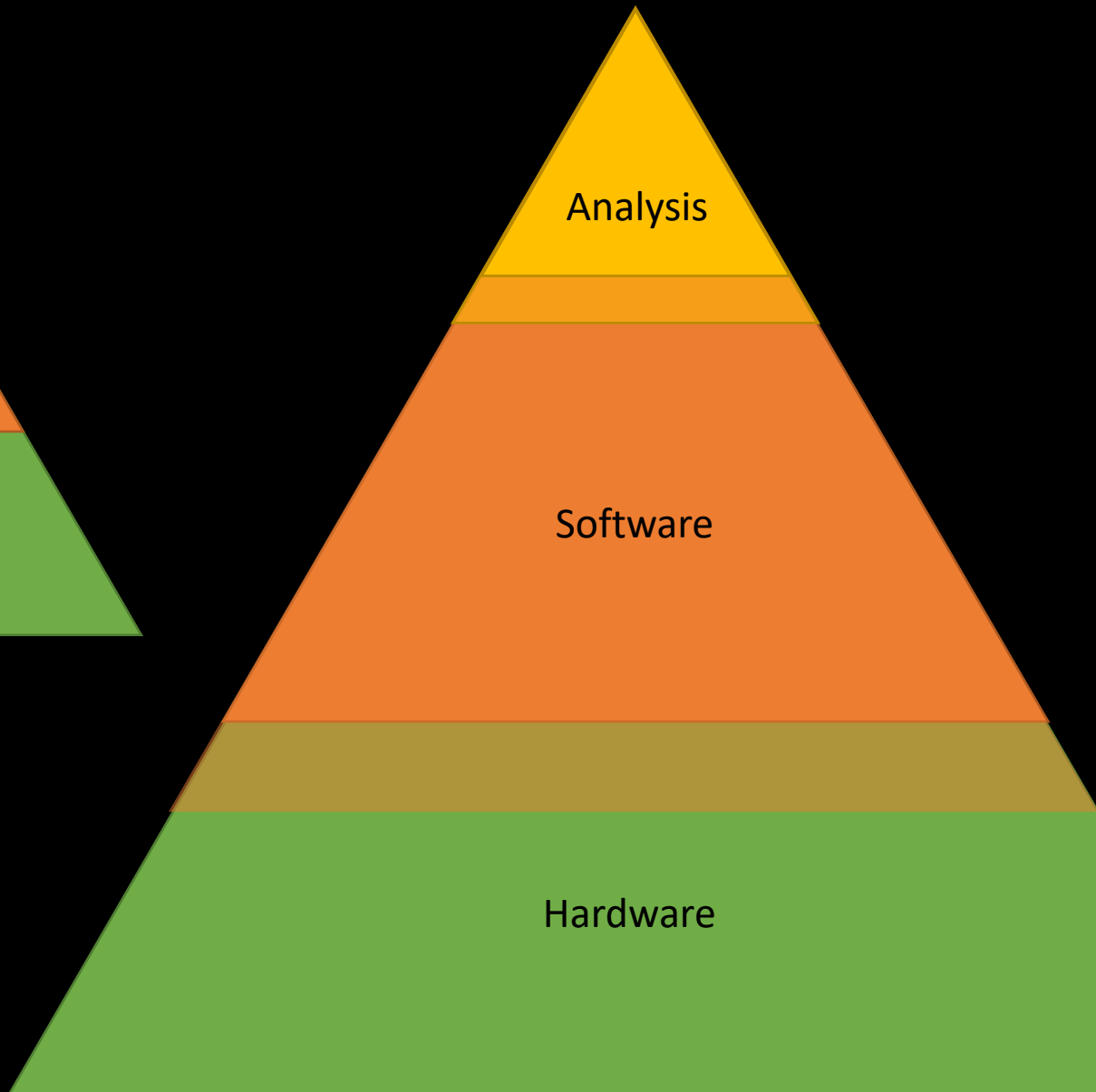
Claim 1: Exponential increase in experimental requirements (incl. computational)



Then



Now



Corollary 2:
We became “data
intensive” in last decade
with requirements per
developer also increasing

Transition from informally run unorganized collectives (e.g. XENON10/100) at LNGS to collaboration structure with more division of labor (XENON1T) spreading data around.

How keep up?

Claim 3:

Few inter-collaboration
computational R&D efforts,
despite need and funding
availability

Claim 4:

Our requirements
mismatch with other
communities, hard to
benefit from their prior
work

XENON: Using LHC data management tools was
(worthwhile) struggle for a while to adapt these tools to us

LZ: NERSC is for high performance instead of high
throughput computing, so refactoring overhead

DarkSide: User management

Corollary 4: LHC experiments have support timescales and legacy requirements irrelevant for us

- We build experiments every few years instead of decades
 - Fewer developers, fewer users
 - Smaller code base

Claim 5:

Around LHC commissioning, technology companies surpassed HEP in Big Data

- Can we follow the “LSST” model of organically using new things to lower manpower requirements and increase science output?
- Can we forget ROOT and join another community that is better supported and transferable?

us



Root



Py Data



Claim 6:

All general frameworks s*ck,
so better to modularize to
limit scope where useful

e.g. matplotlib plotting, uproot/HDF5 I/O, nestpy
NEST microphysics, g4py

What does this mean for software development?

DAQ

Processing

Analysis

Triggerless Readout

500 MB/s
MongoDB

Software Trigger

Pickle files

1 PB/year

Distributed Processing
OSG Chicago

Distributed storage
Nikhef Chicago
Weizmann Bologna OSG
Lyon

Tape

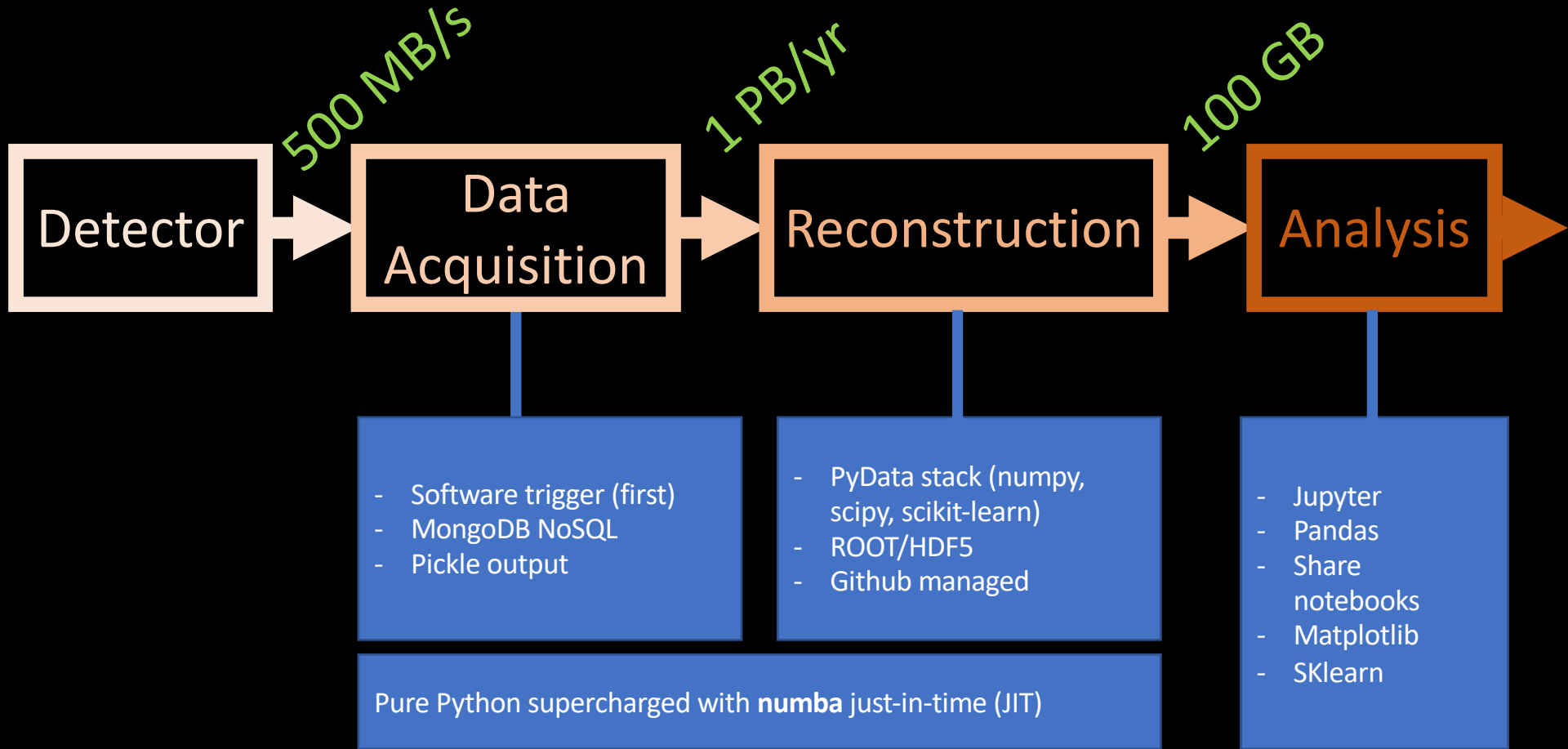
50 TB
Common Storage
Chicago

Reduce

1 GB
Table/HDF5

IPython Notebooks

Data Flow Essentials



Challenges and Needs

- Challenge: “Crank up the gain” necessary for dark matter program
 - Effect: 1) 500 MB/s streaming rate to 2) PB/yr storage rate
 - New challenge 1): How to build fast upgradable streaming system?
 - Subchallenge: Physics education fails to teach basic skills in programming, Data Science, or transferable skills... so we have to teach this as best we can
 - Subchallenge: Sit between HEP and High Tech, where they don't play together
 - Subsubchallenge: Building advanced system with few people requires being frantically technologically aggressive
 - New challenge 2): How to handle PB of data with minimal manpower?
 - Subchallenge: NSF does not have national labs, so highly distributed
 - Subchallenge: Efforts to use prior investment generally has burdensome overhead for our scale experiment
 - Subchallenge: Retaining people with these skills difficult
 - Subsubchallenge: No clear training path for undergraduates, graduates, postdocs, staff
 - Subsubchallenge: Also no clear career trajectory
 - Subsubchallenge: Lack of steady operational funds
 - Subchallenge: How do analysts interact with these data without giant software stacks and in an interactive manner (e.g. JupyterHub)?

Numba: Python just-in-time compiler

- Few 'array-oriented' compilers though common use case and hardware optimizations exist.
- Wasn't possible few years ago, **Python faster than your C++ code.**

```
@vectorize
def sinc(x):
    if x==0.0:
        return 1.0
    else:
        return sin(x*pi)/(pi*x)
```

```
1 ; ModuleID = 'sinc_mod_7b29370'
2
3 define double @sinc(double %x) {
4 Entry:
5   %0 = fcmp oeq double %x, 0.000000e+00
6   br il %0, label %BLOCK_12, label %BLOCK_16
7
8 BLOCK_12:                                ; preds = %Entry
9   ret double 1.000000e+00
10
11 BLOCK_16:                                ; preds = %Entry
12   %1 = fmul double %x, 0x400921FB54442D18
13   %2 = call double @llvm.sin.f64(double %1)
14   %3 = fmul double %x, 0x400921FB54442D18
15   %4 = fdiv double %2, %3
16   ret double %4
17
18 BLOCK_47:                                ; No predecessors!
19   ret double 0.000000e+00
20 }
21
22 declare double @llvm.sin.f64(double) nounwind readonly
```

Backup

- Details of computing scheme in arXiv: 1903.11441
- Code under XENON1T, XENONnT, AxFoundation Github organizations