

Computing challenges and opportunities for FAIR experiments





1. Distrust



2. Excitement



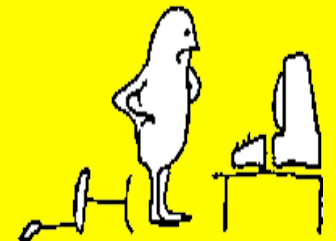
3. Astonishment



4. Enthusiasm



5. Love



6. Disillusionment



7. Fright



8. Horror



9. Fury



10. Frustration



11. The End



1. Distrust



2. Excitement



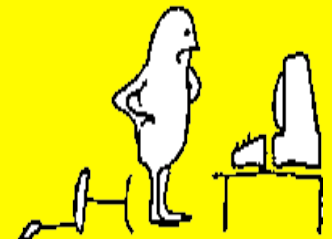
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4. Enthusiasm



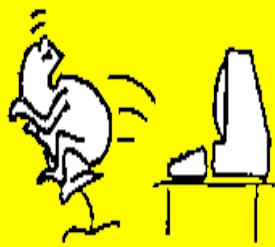
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10. Frustration

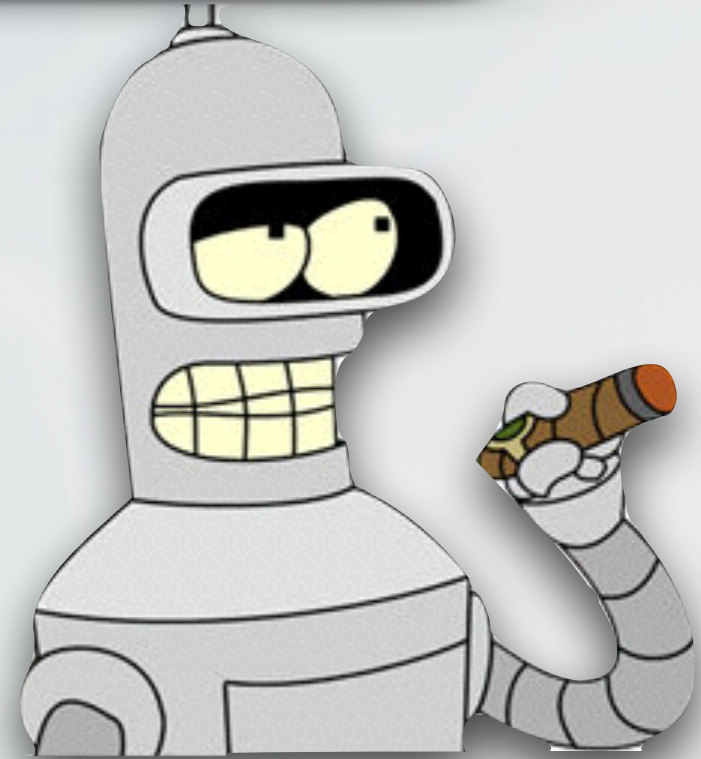


11. The End

Disclaimer:

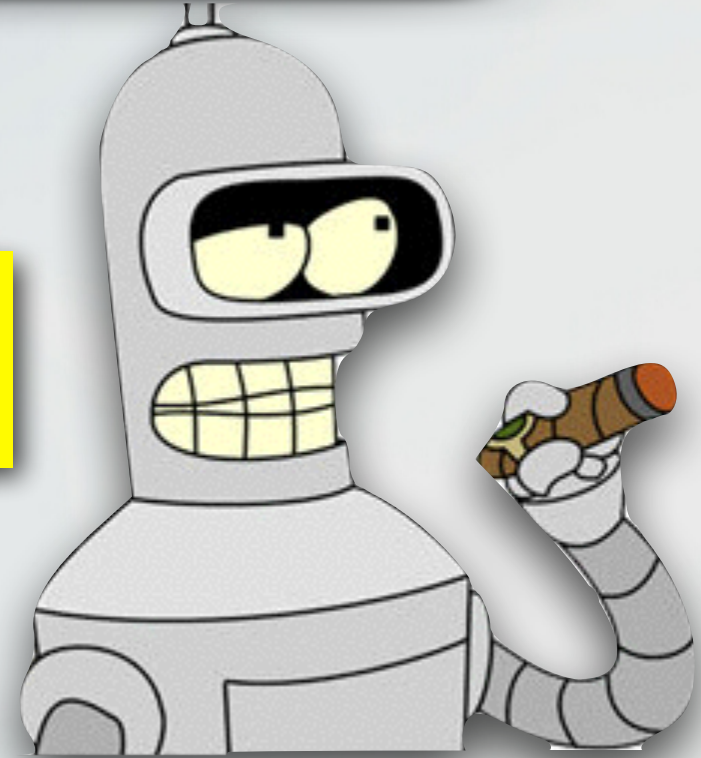
The information selected for this talk solely reflects my personal excitement, astonishment, enthusiasm, and love for (FAIR) computing.

Who are you? And why should I listen?



Who are you? And why should I listen?

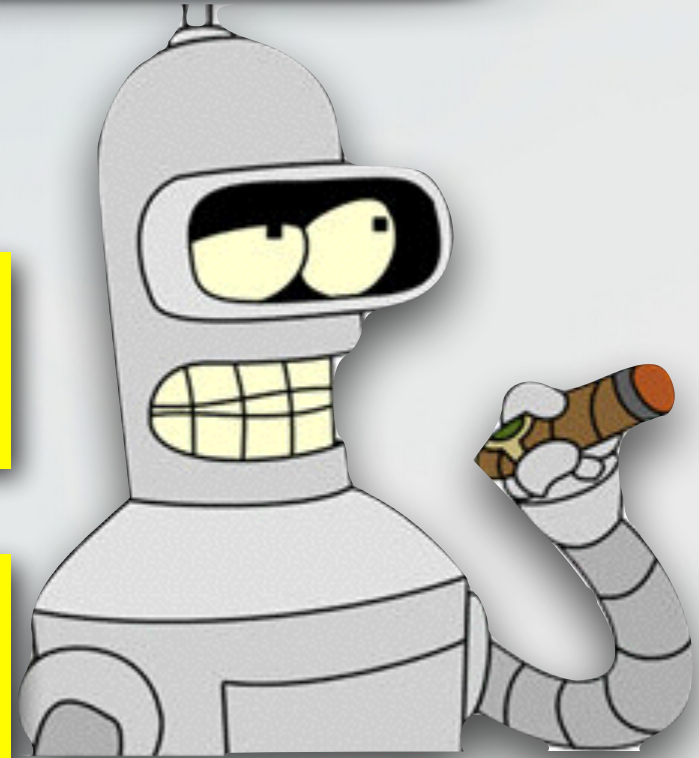
1. Everyone (dis)likes computing, good for the upcoming dinner discussion.



Who are you? And why should I listen?

1. Everyone (dis)likes computing, good for the upcoming dinner discussion.

2. Very large chance you have to deal with “FAIR computing” in some way.

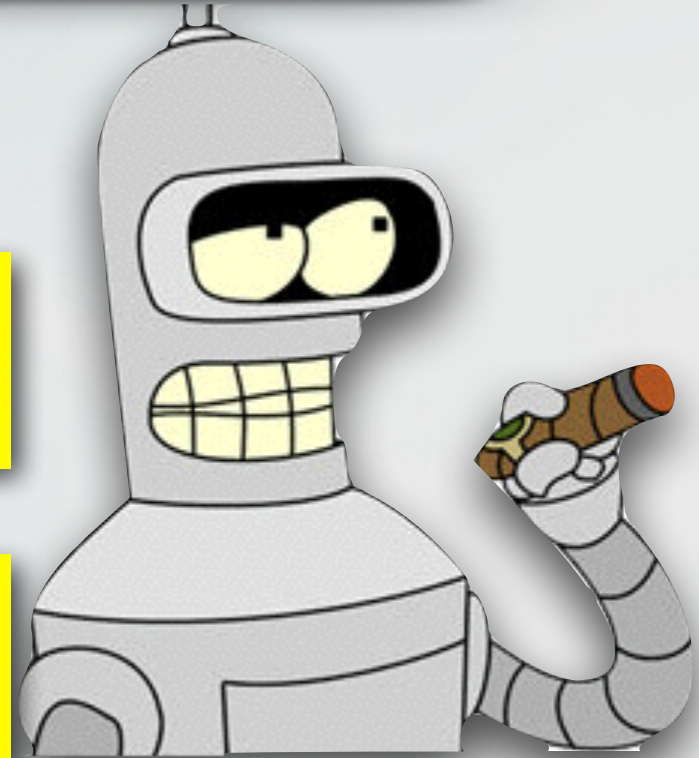


Who are you? And why should I listen?

1. Everyone (dis)likes computing, good for the upcoming dinner discussion.

2. Very large chance you have to deal with “FAIR computing” in some way.

3. Computing remains rapidly changing: new developments & opportunities!



BIG

DATA

USE
ONE SINCE
SOFTWARE

BUSINESS
REQUIRING
MOVING
ORGANIZATIONS

ANALYTICS

PARALLEL

UBIQUITOUS
RADIO-FREQUENCY
COMPLEXITY
SOLID
WIRELESS
TOLERABLE
SIZE
NEEDED

INTERNET

TECHNOLOGIE

USED
DISTRIBUTED
MAY
SOCIAL
MANAGEMENT
CAPTURE
STORE
DEFINING
CASE

STORAGE

COMPLEX
RESEARCH
MPP
DATABASES

SETS
EXAMPLES

SENSOR
ARCHIVES

ELAPSED
CURRENT
THOUGHT

APPLIED
DESCRIBING
AMOUNT

USING TYPES
WORKING
GARTNER
MASSIVELY
BIOLOGICAL
ALSO RELATED
HUNDREDS
CREATED

PERFORMANCE
DISK
RELATIONAL
TIME
SHARED
COMBAT
SIGNIFICANT
INCLUDE SYSTEMS
BIOGEOCHEMICAL
NETWORKS
INFORMATION

COMPUTING
TOOLS
SET
GENOMICS
ZETTABYTES
WITHIN
PROCESS
DEFINITION
SEARCH
OPPORTUNITIES

CONNECTOMICS
RECORDS
COST
CONTINUES
CITATION

DESKTOP
CURRENTLY
FC
TENS
CAPACITY
FORMS
PRESENTATIONS
NOW
PRACTITIONERS
WORLD'S

PROCESSING
LOGS
EVERY
LARGER

MANAGE
GROW

The “Big Data” challenges

Doug Lany, META Group Inc.

The “Big Data” challenges

Doug Lany, META Group Inc.

Volume: data volume grows faster than computing power.

Velocity: the rate at which data is processed for fast decision making.

Variety: heterogeneity of data types, representation, and interpretation.

Veracity: degree of certainty about data.

Complexity: data from complex network of many different sources.

FAIR and the “Big Data” challenge

Thorsten Kolleger, Stephane Pietri, Stefano Spataro, Volker Fries

FAIR and the “Big Data” challenge

Thorsten Kolleger, Stephane Pietri, Stefano Spataro, Volker Friesse

PANDA/CBM

- monolithic systems
- LHC-like rates & requirements
- up to 1 TB/s data rates
- online data reduction to ~1 GB/s
- total request 3×10^5 cores at Tier0
- distributed computing

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NUSTAR/APPA

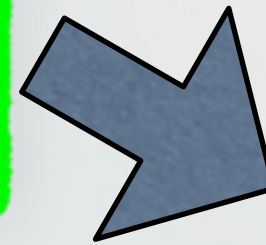
- many smaller experiments
- up to 1 GB/s data rates
- computing power $\sim 4 \times 10^3$ cores
- large variety in setups
- high flexibility in data processing
- complexity in database management

FAIR and the “Big Data” challenge

Thorsten Kolleger, Stephane Pietri, Stefano Spataro, Volker Fries

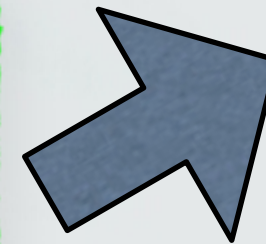
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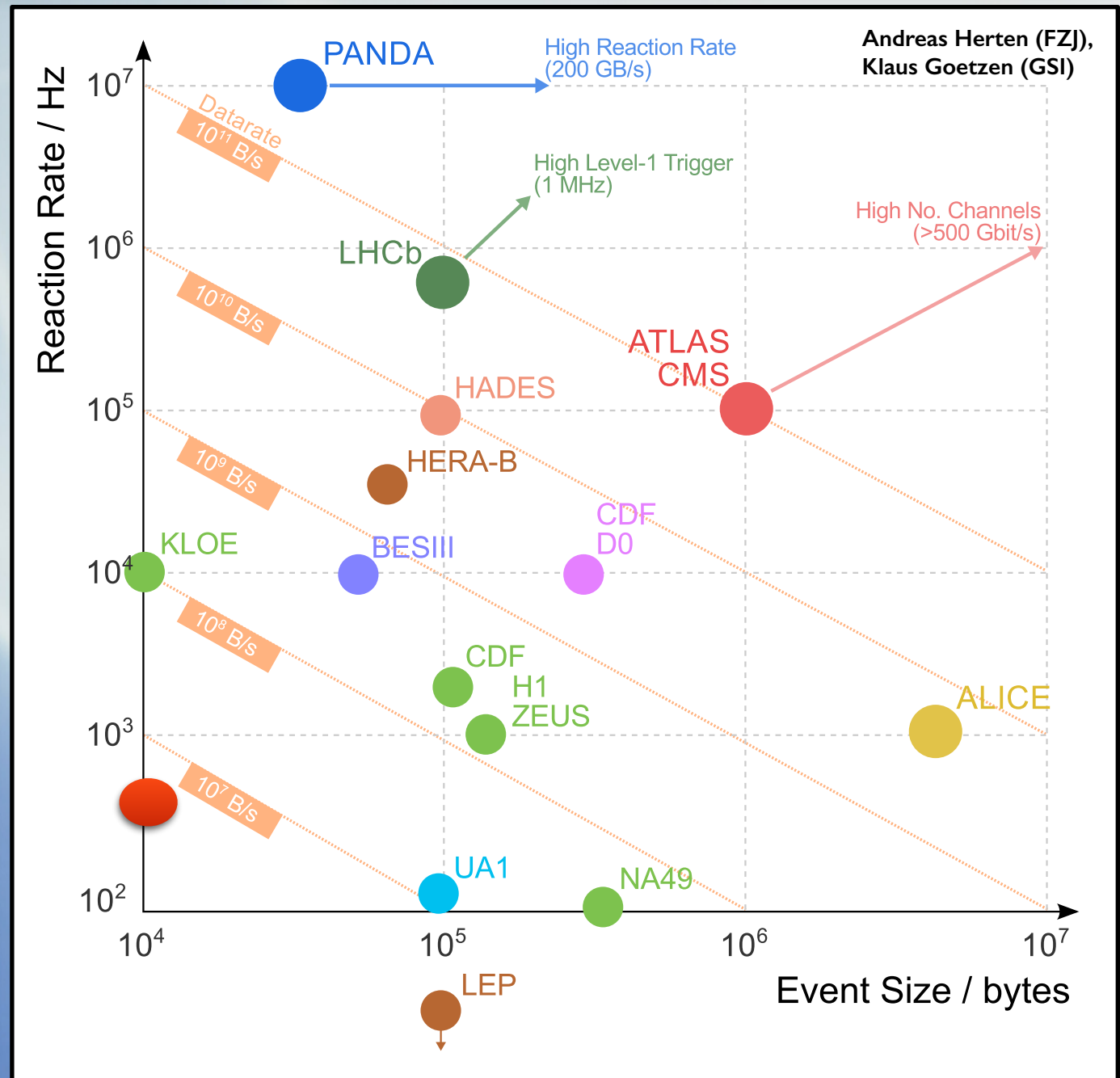
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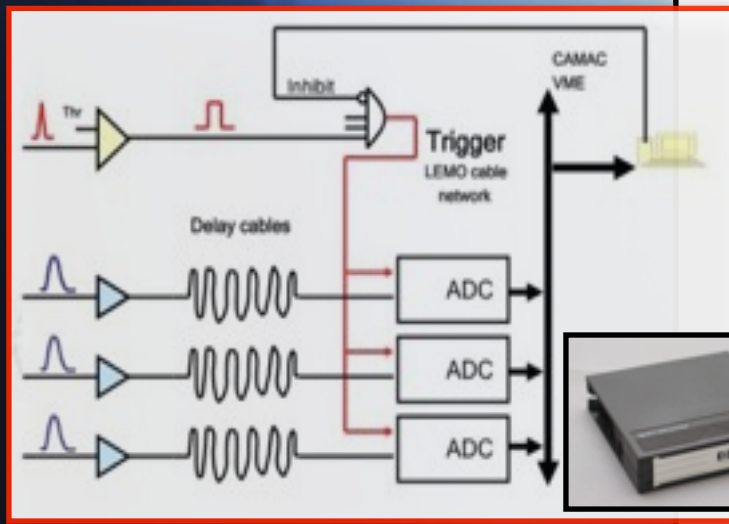
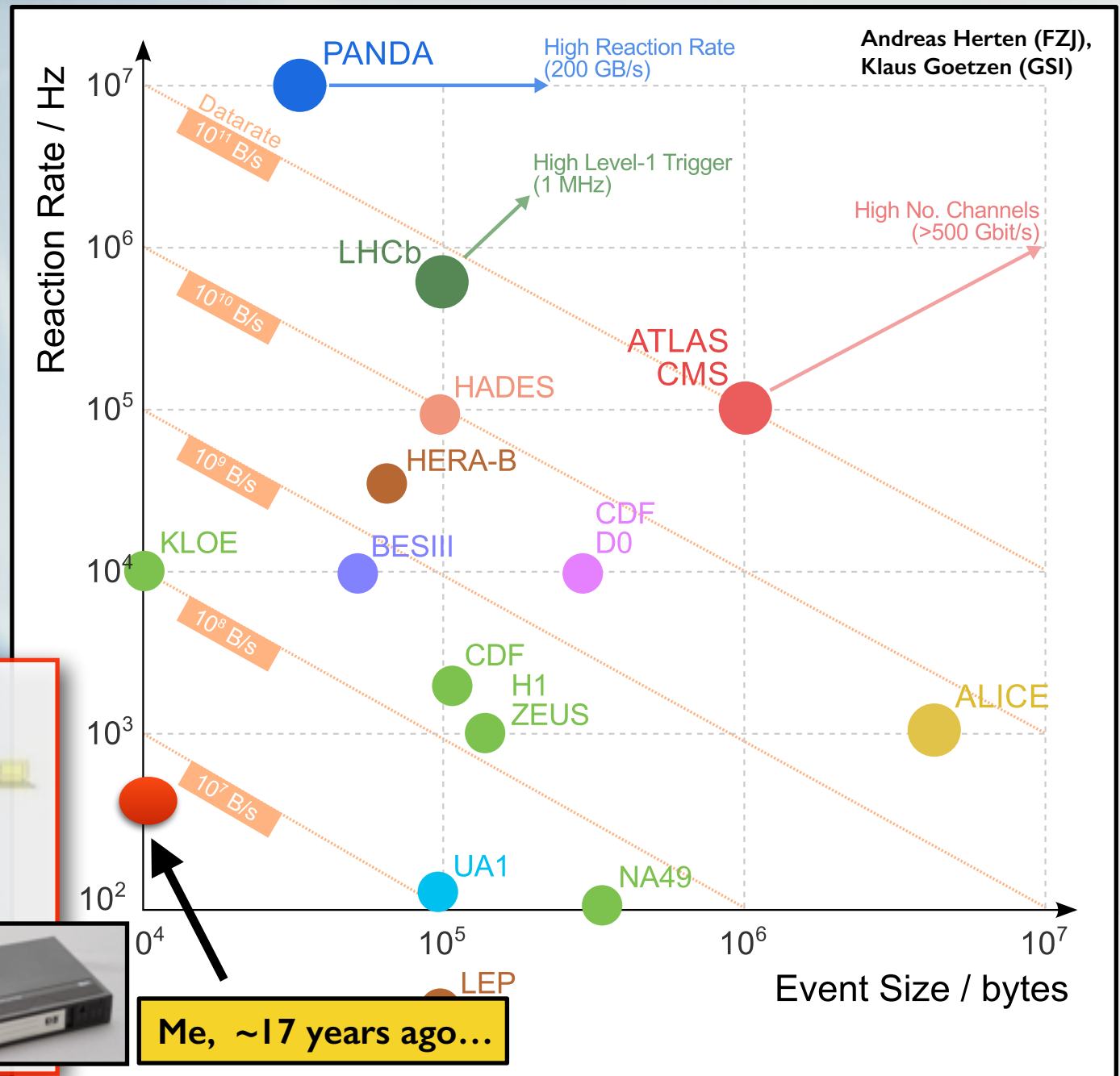


Volume
Velocity
Veracity
Variety
Complexity

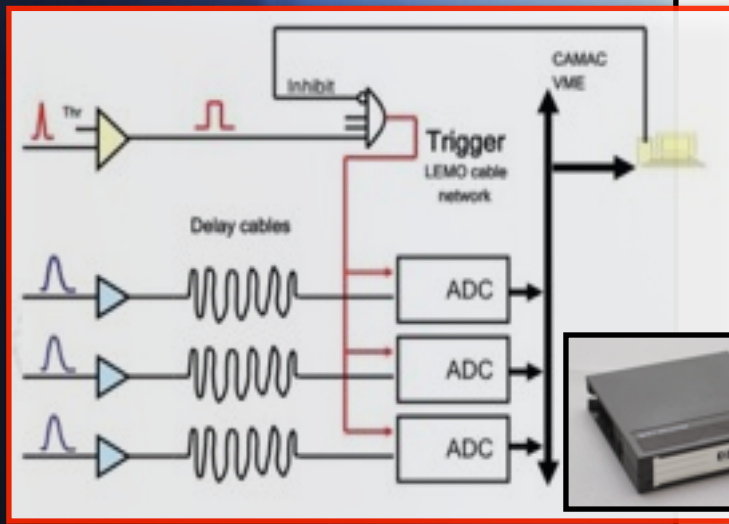
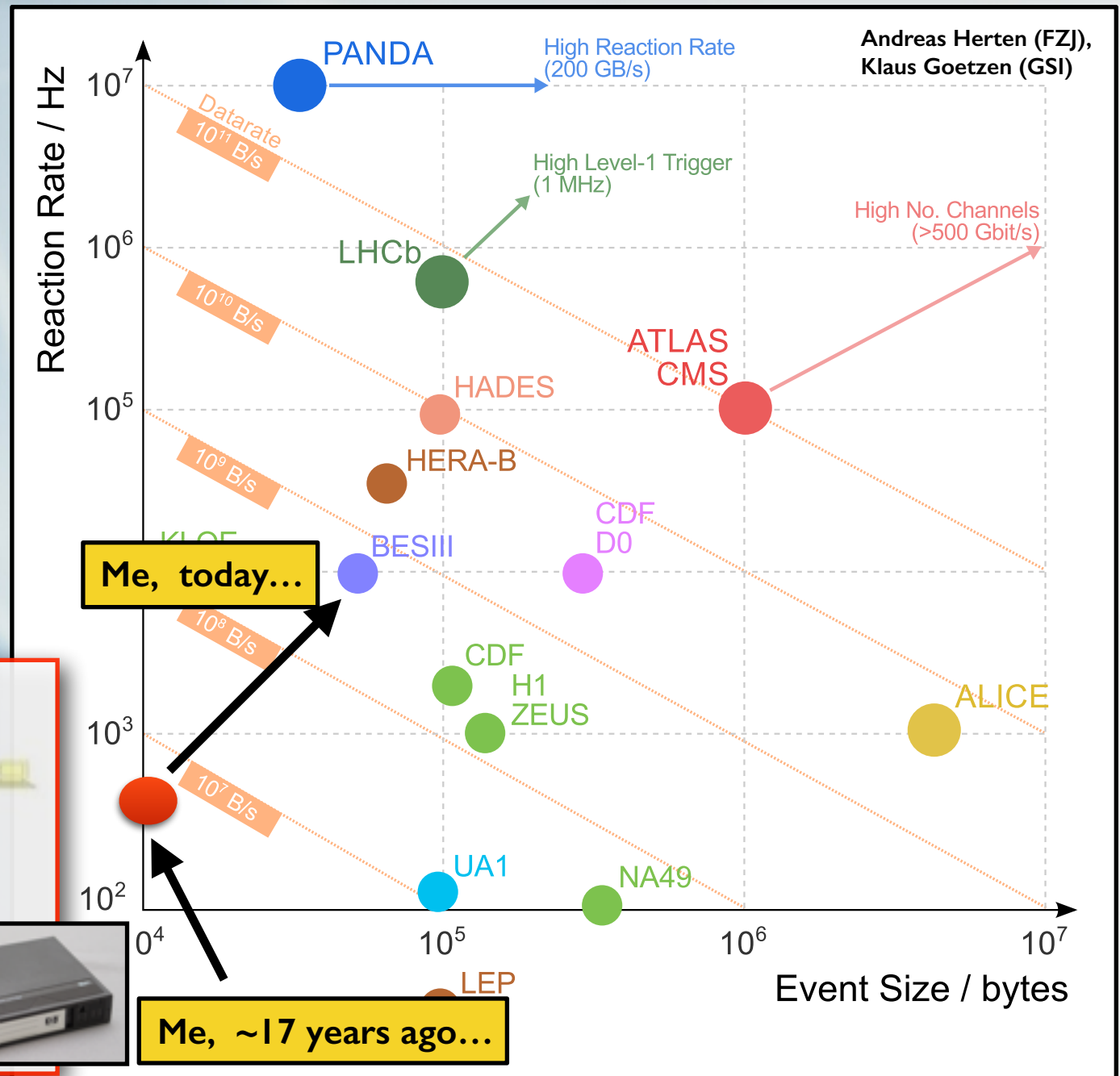
The data-processing challenge



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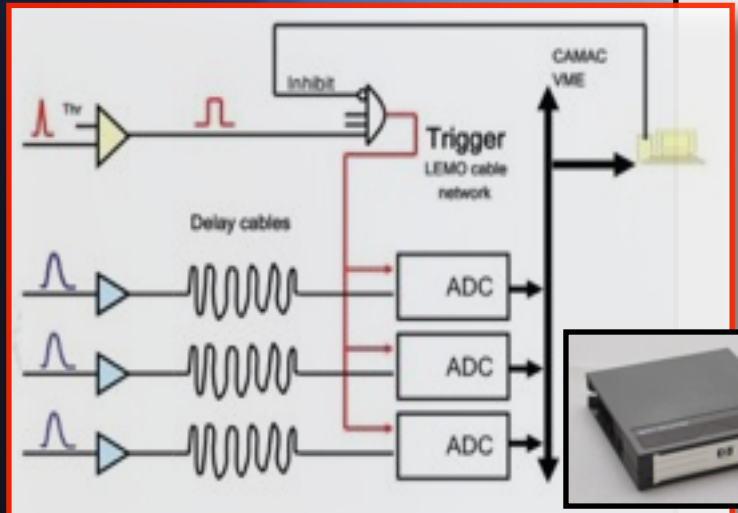
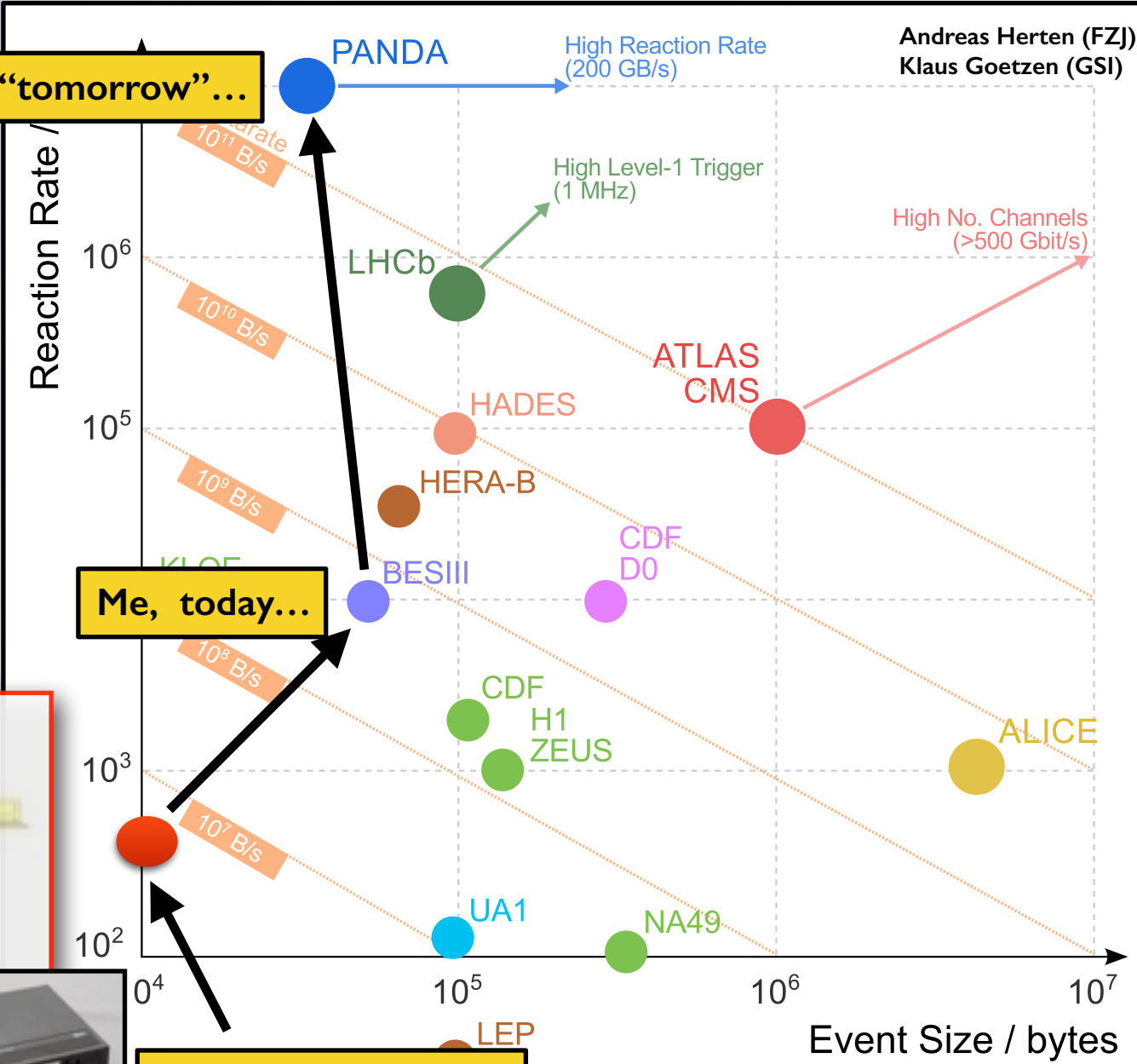
The data-processing challenge

Me and some of you, "tomorrow"...

Me, today...

Me, ~17 years ago...

Andreas Herten (FZJ),
Klaus Goetzen (GSI)



“Tomorrows” experiment

Physics

"Tomorrows" experiment

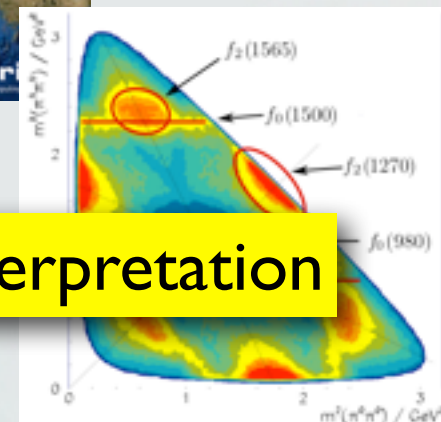
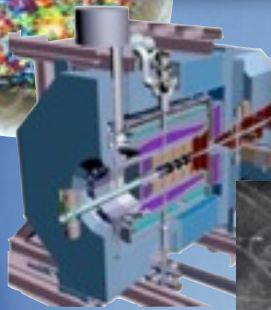
Detector

Feature extraction

Online computing

Distributed computing

Data interpretation



"Tomorrows" experiment

Physics

Detector

Feature extraction

Online computing

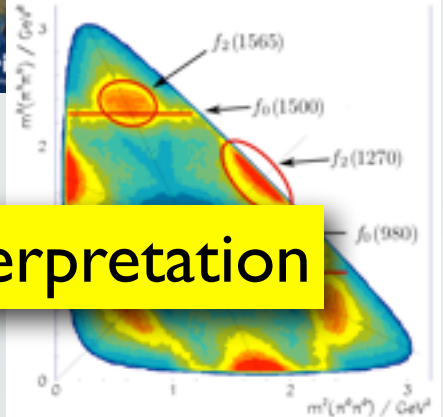
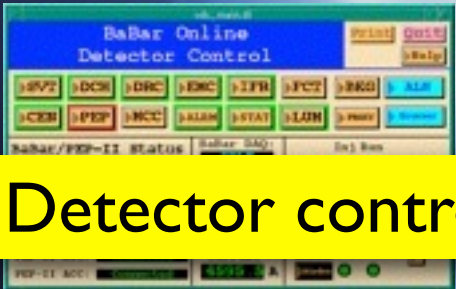
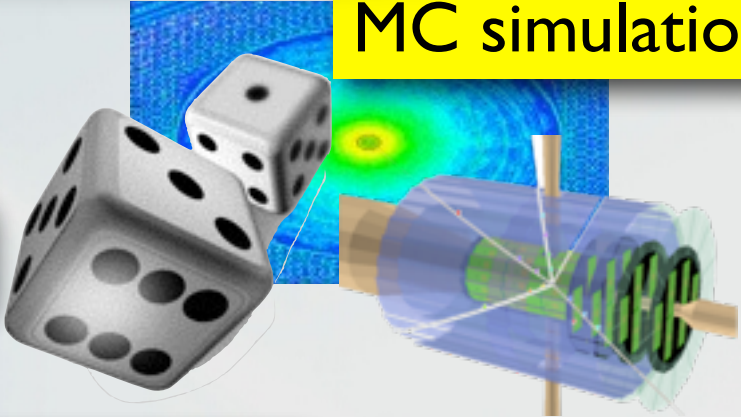
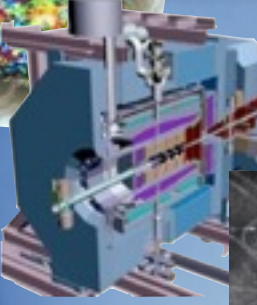
MC simulations

Distributed computing

Detector control

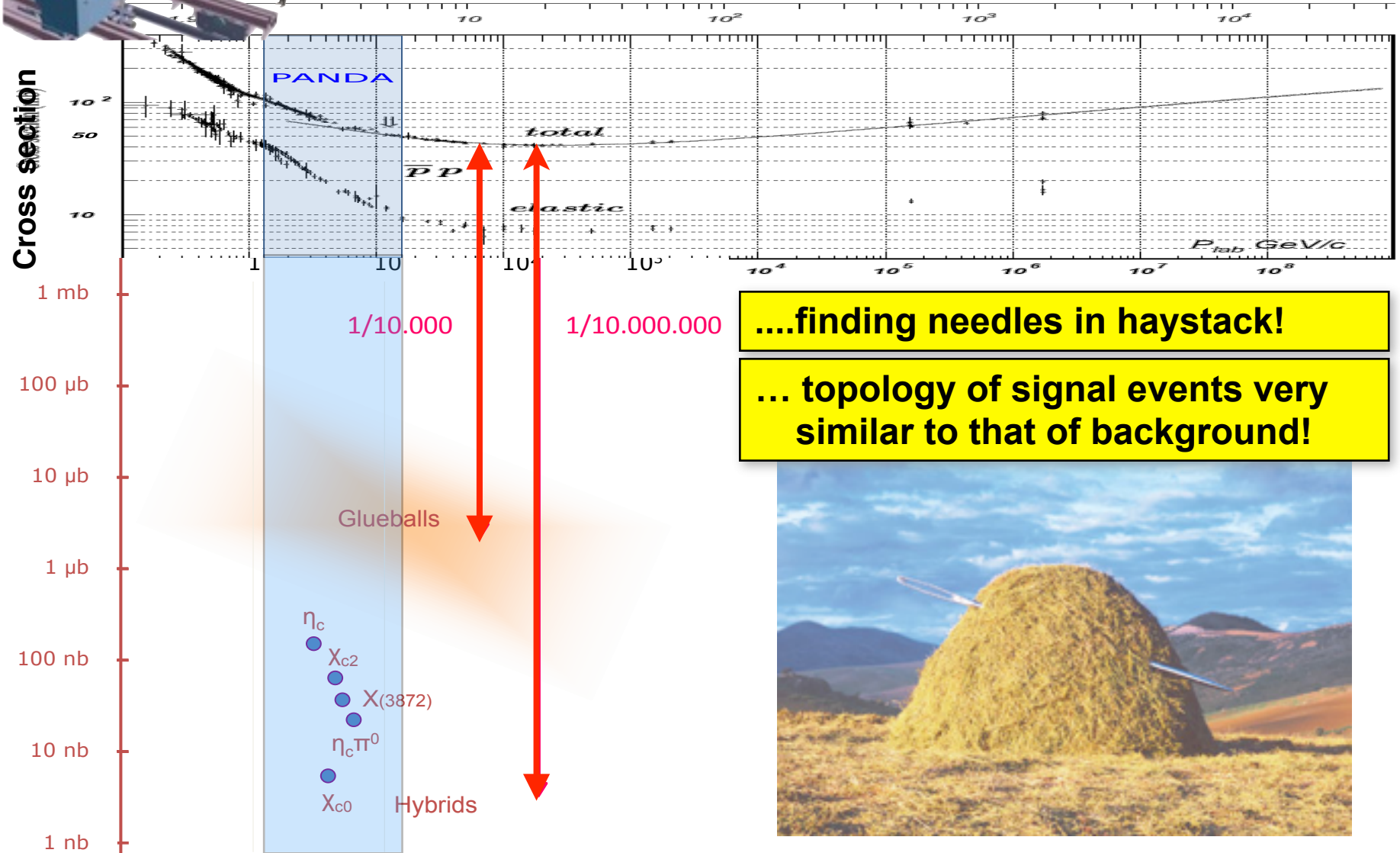
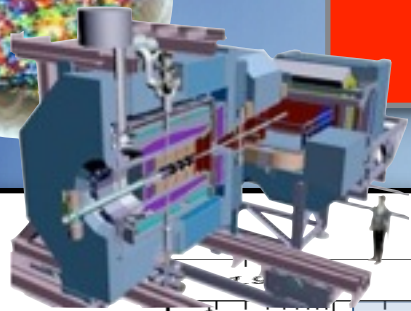
Database management

Data interpretation



The “physics” challenge! (PANDA as an example)

Klaus Goetzen (GSI)



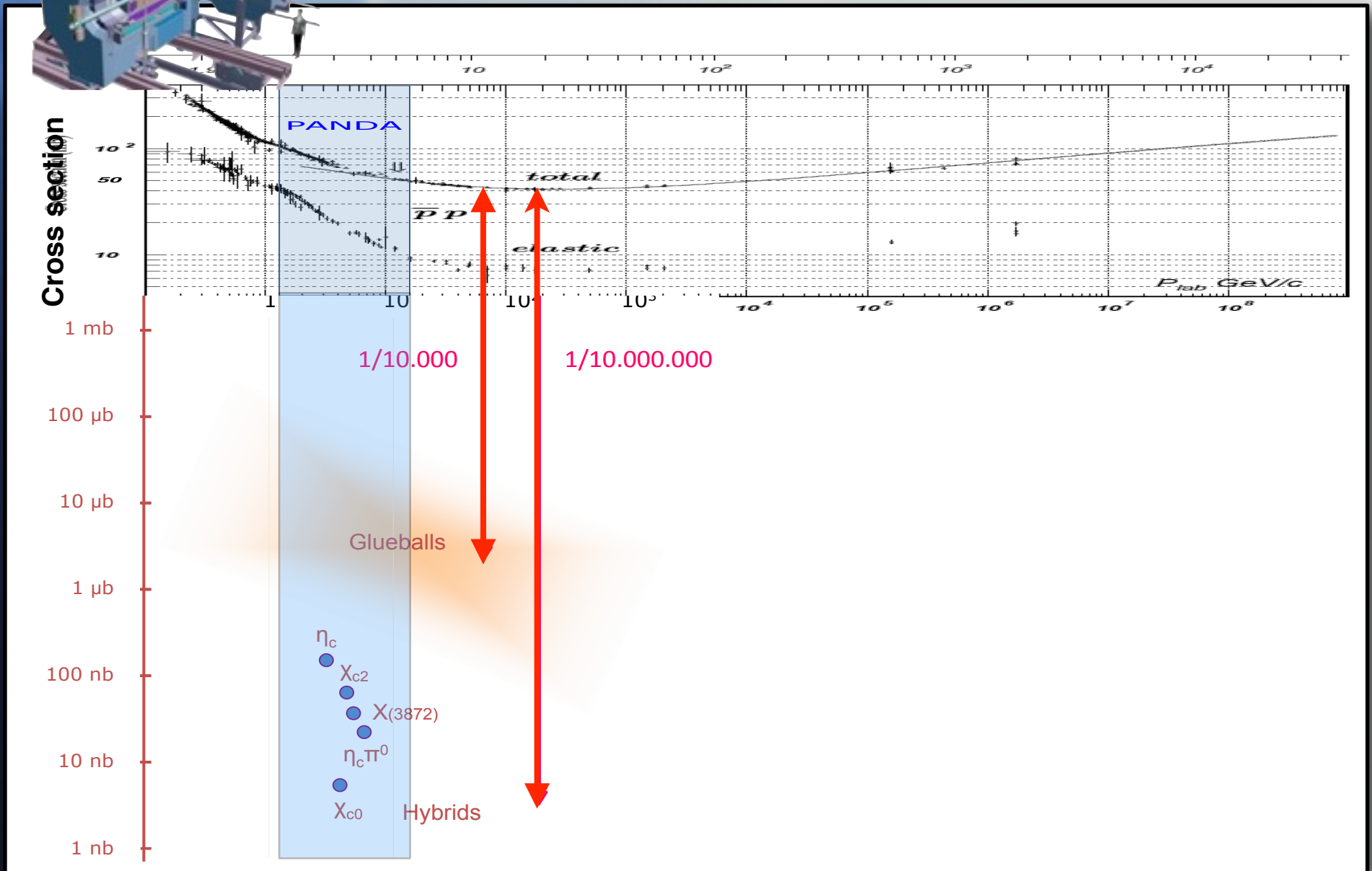
...finding needles in haystack!

... topology of signal events very similar to that of background!



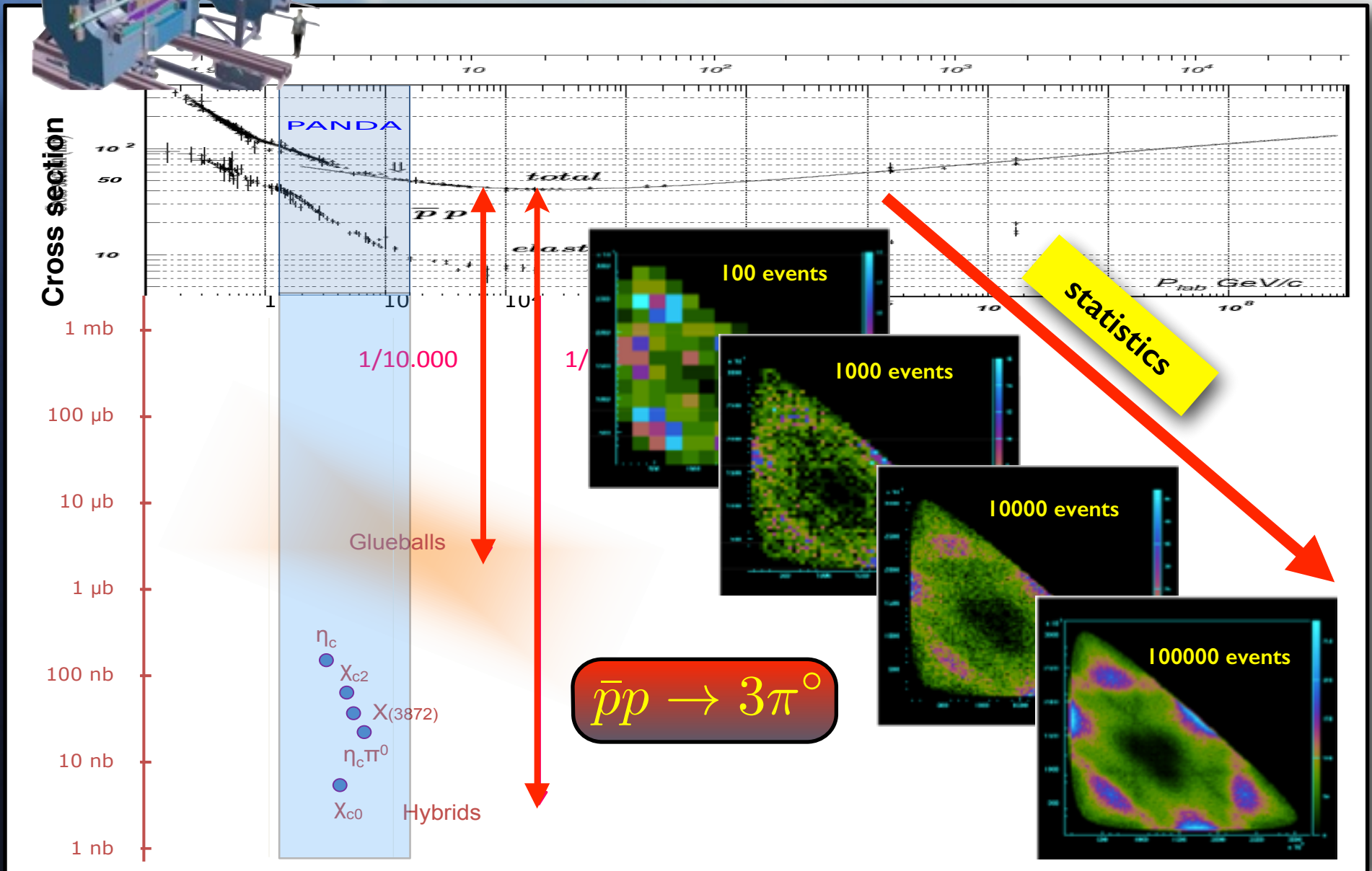
The “physics” challenge! (PANDA as an example)

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The "physics" challenge! (PANDA as an example)

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Versatile experiments

Era of large-scale facilities and huge collaborative efforts

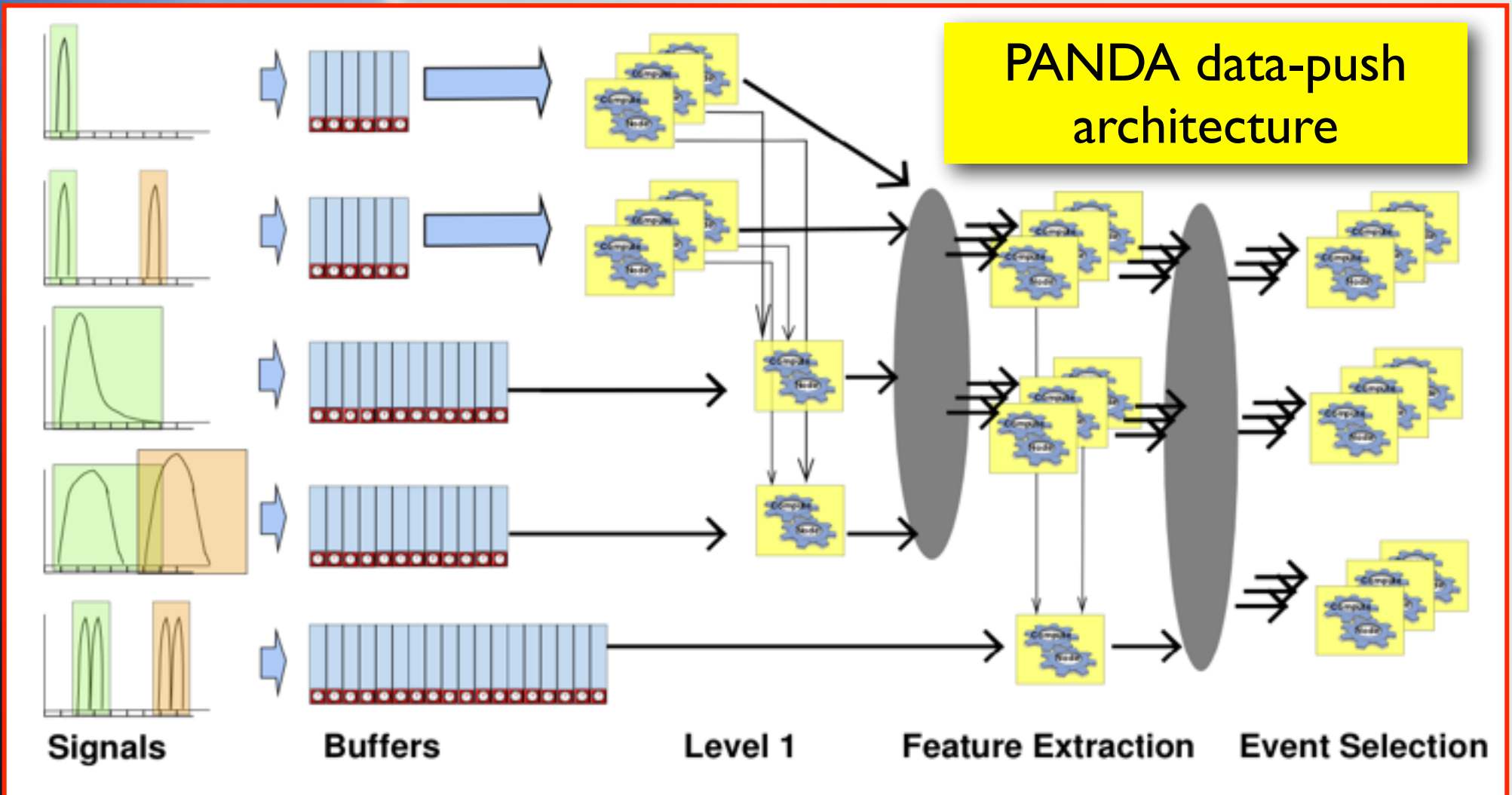
Experiments aim to cover a broad physics program

Not one “golden” channel

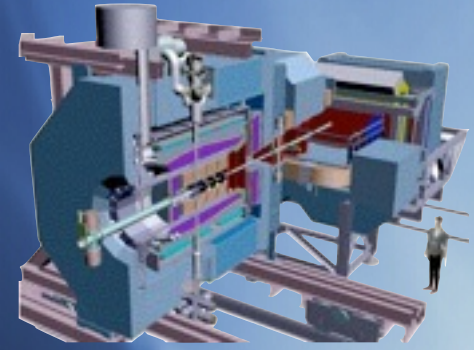
Experiments need to become capable and flexible to accommodate versatility



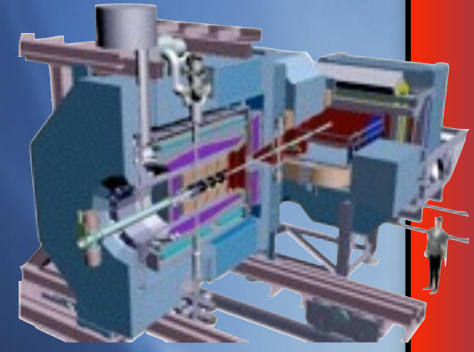
Trigger-less Data Acquisition



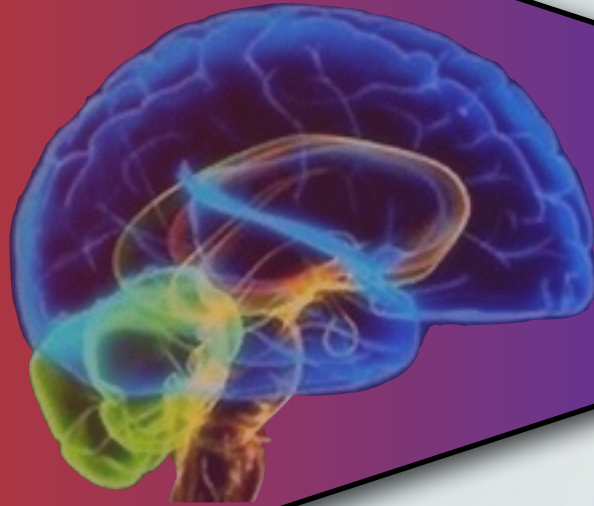
continuous data sampling with self-triggered detector front-end ---> flexible event selection



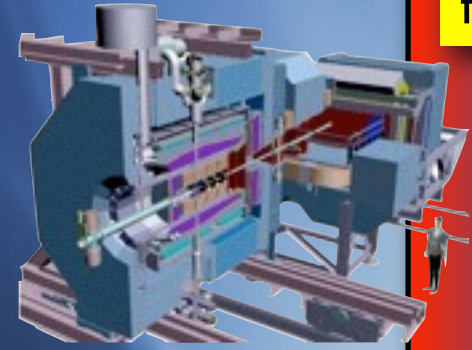
10^7 /sec.



10^7 /sec.



10^4 events/sec.



10^7 /sec.

feature extraction

track finding

event ordering

vertex fitting

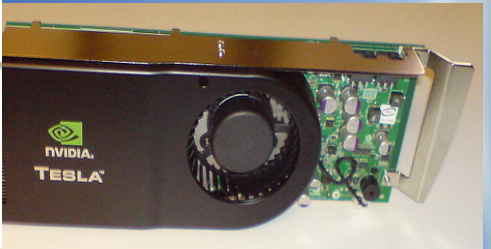
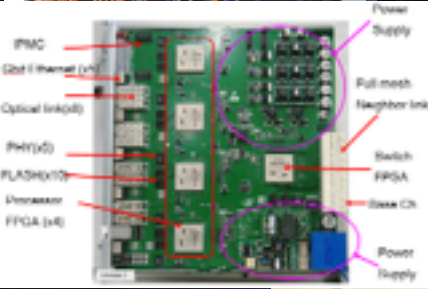
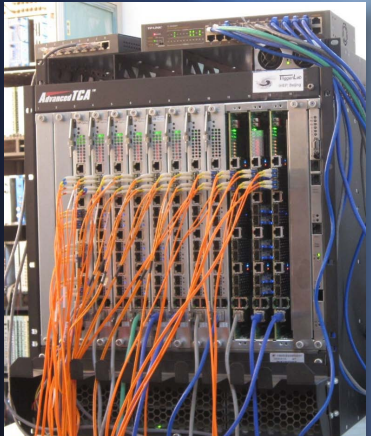
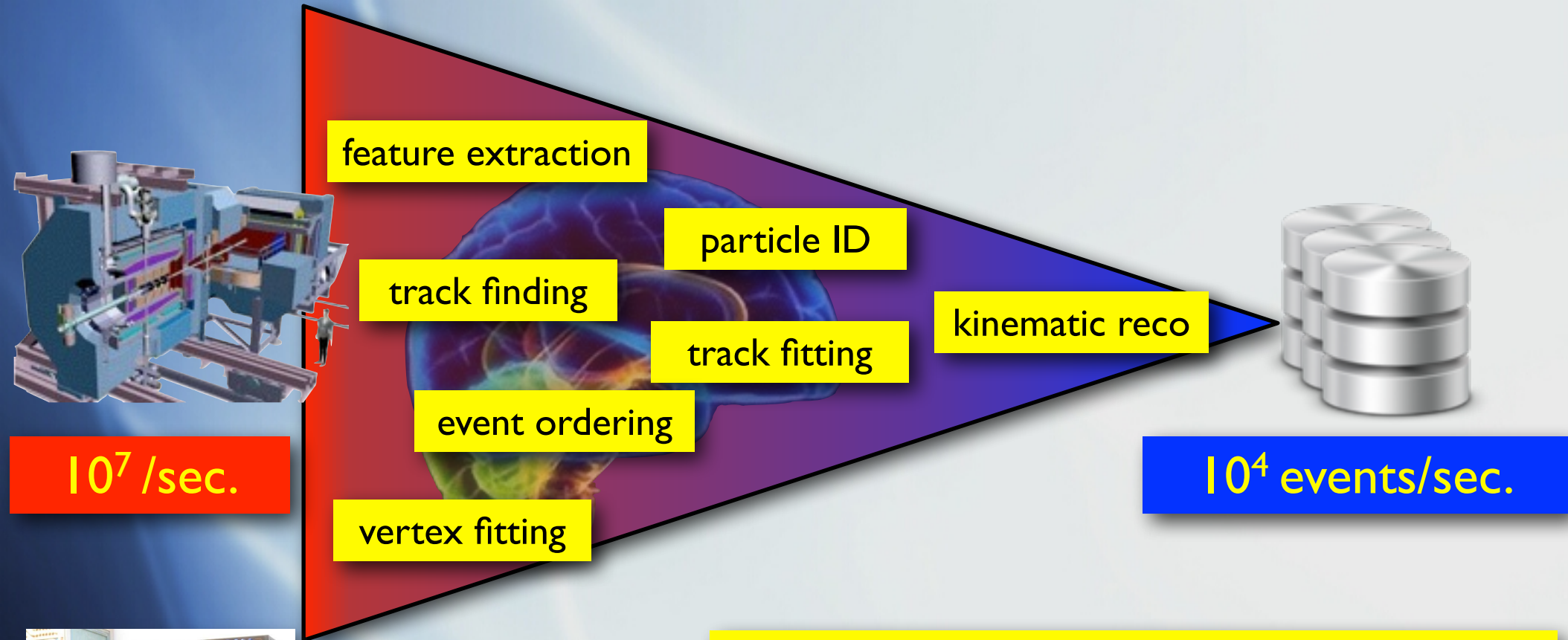
particle ID

track fitting

kinematic reco



10^4 events/sec.



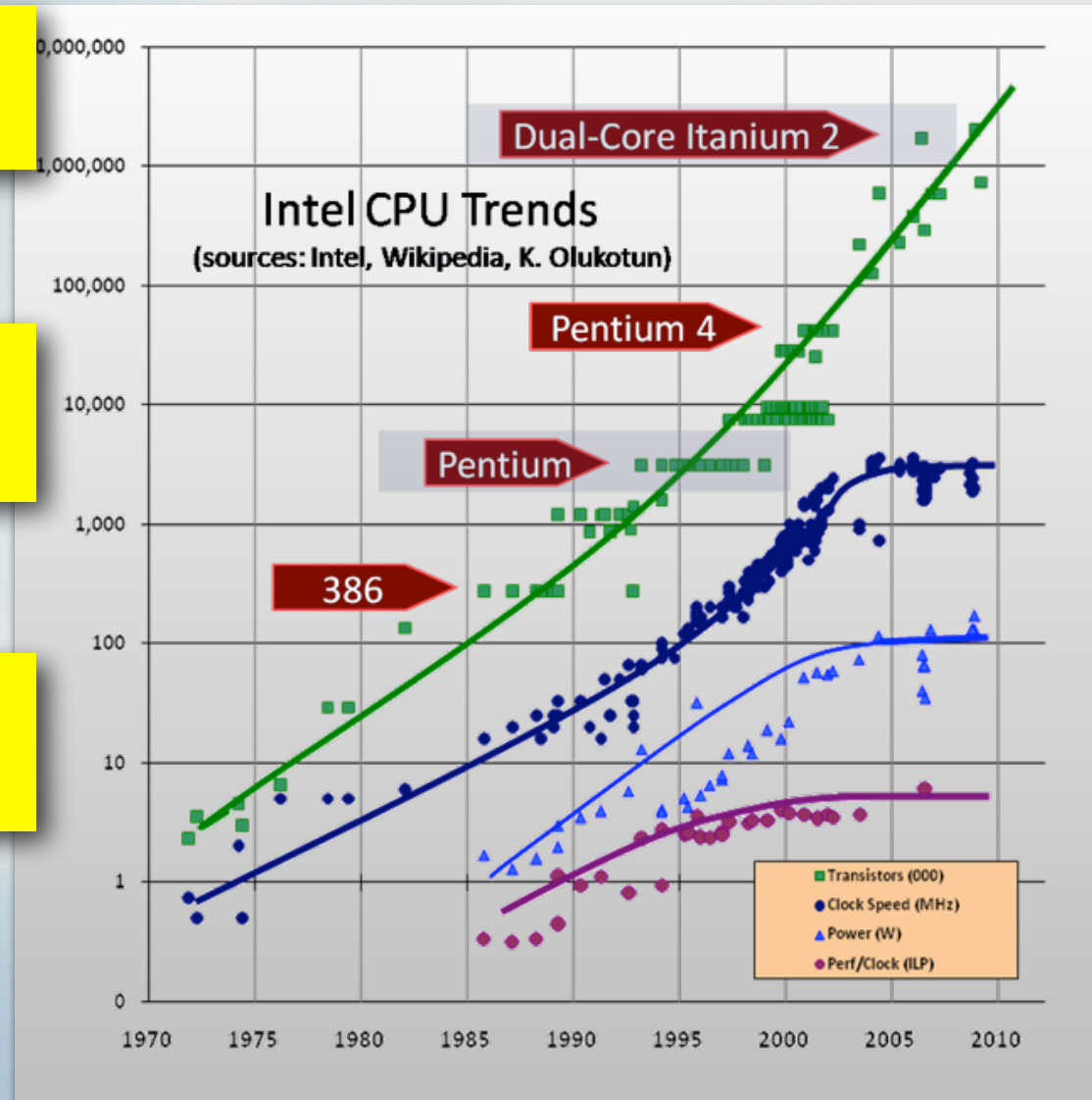
In-situ event reconstruction?
 Yes, no other choice! But ...
 How to implement on hardware?
 Which hardware to use?
 Streaming data, pileup?
 How to adapt existing code?
 How to simulate online processing?
 How to link components together?

“The free lunch is over”

Lets admit! We were spoiled by the frequency-scaling era

Cores don't get faster, but more parallel! Requires new approaches!

Even then, not all jobs are easy parallelizable (Amdahl's law)



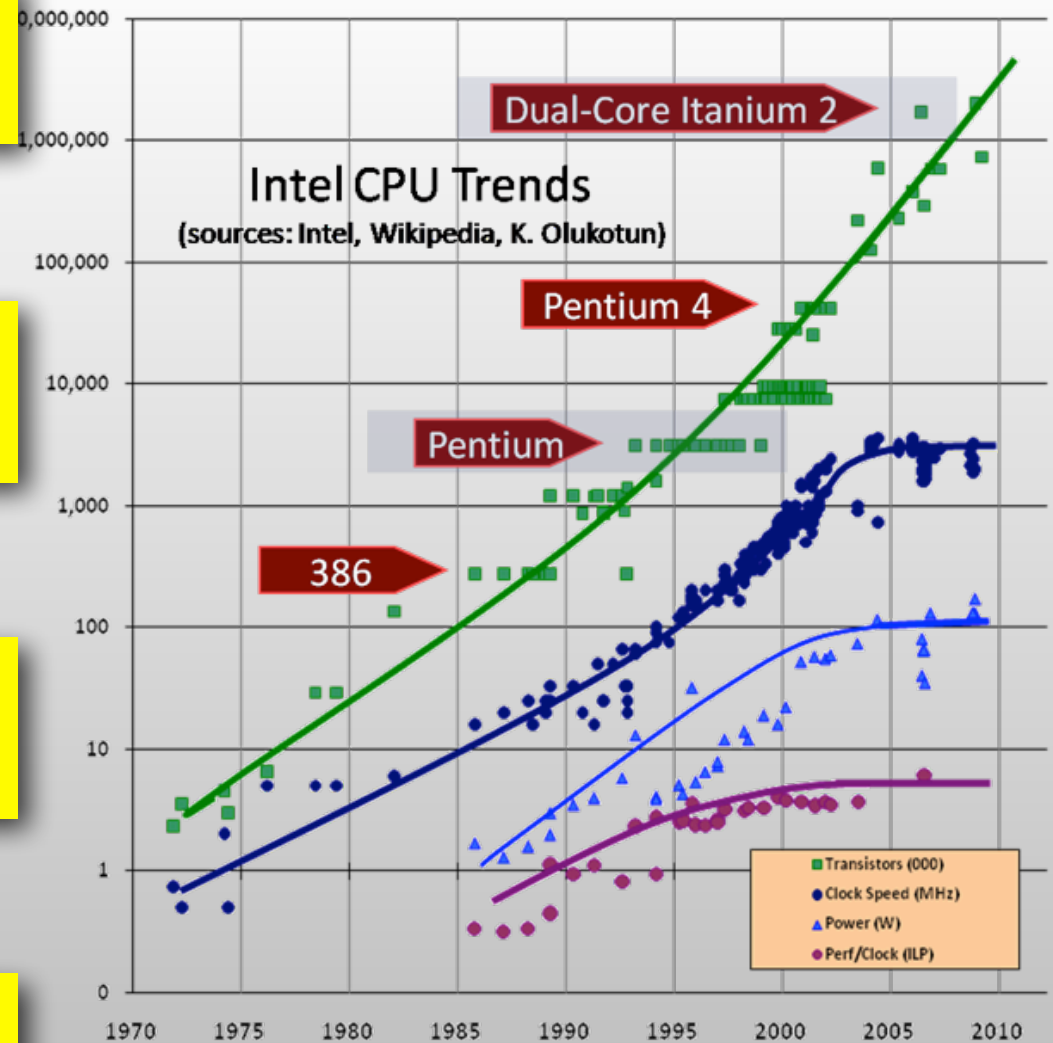
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Innovative ideas needed!



“Intelligence” in practise: GPUPWA

Niklaus Berger

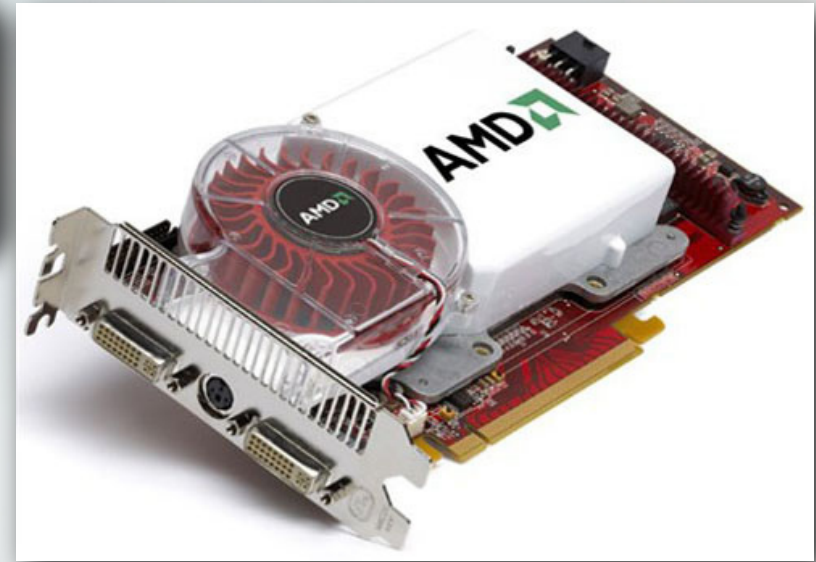
Likelihood, given n data points at Ω_i

$$\mathcal{L} \propto \prod_{i=1}^n \frac{I(\Omega_i)}{\int \eta(\Omega) I(\Omega) d\Omega}$$

Product over data events

Detection efficiency

Normalisation integral over phase space

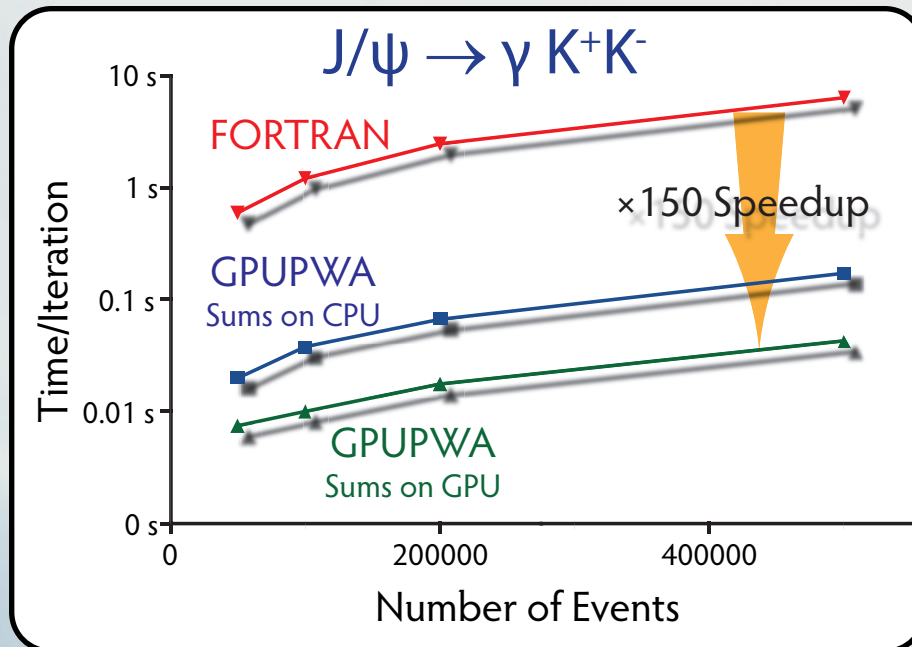


Resonance searches with BESIII data

Partial-wave analyses for a huge data samples are extremely time consuming

Innovative idea! Use SIMT capabilities on cheap graphic cards

Massive parallel floating point operations and lookup tables ideally suited for the fitting problem



Towards real-time applications

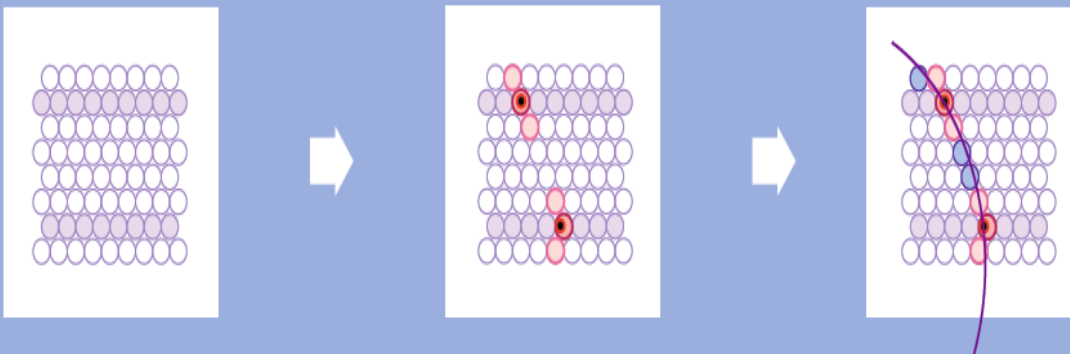
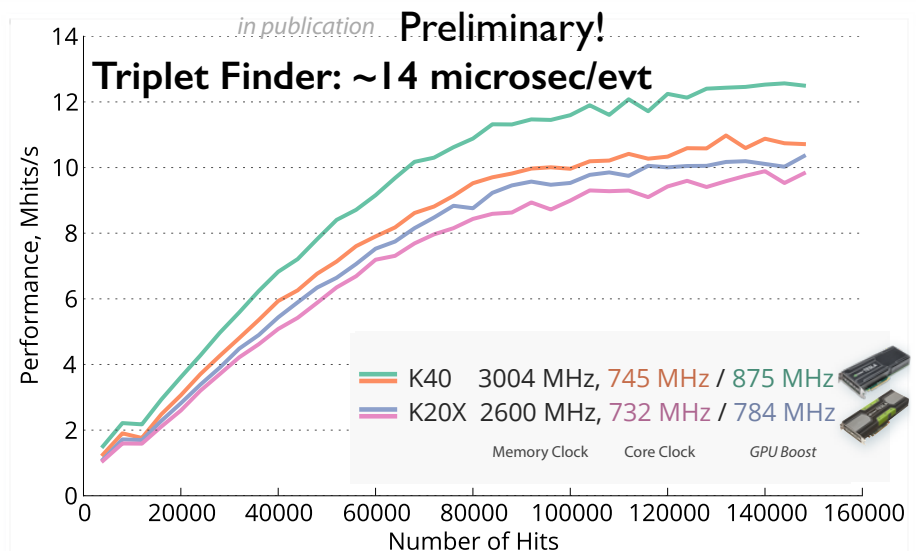
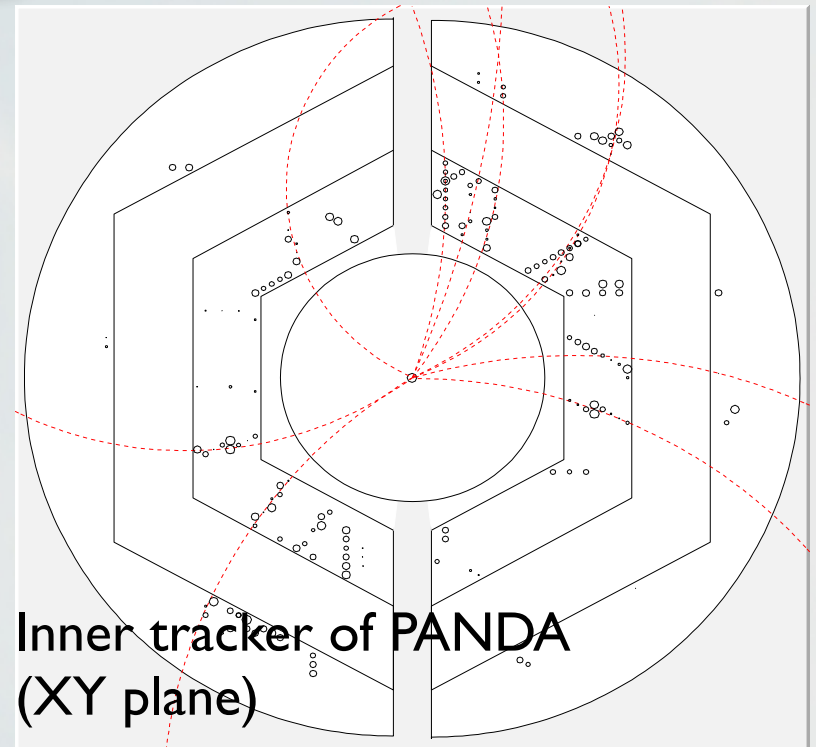
Track finding & fitting using **GPUs**

Inner tracker of PANDA:

- Hough transformations (2D)
- Riemann track finder (3D)
- Triplet finders
-

Work in progress...

Andreas Herten, Tobias Stockmanns (FZJ)



Towards real-time applications

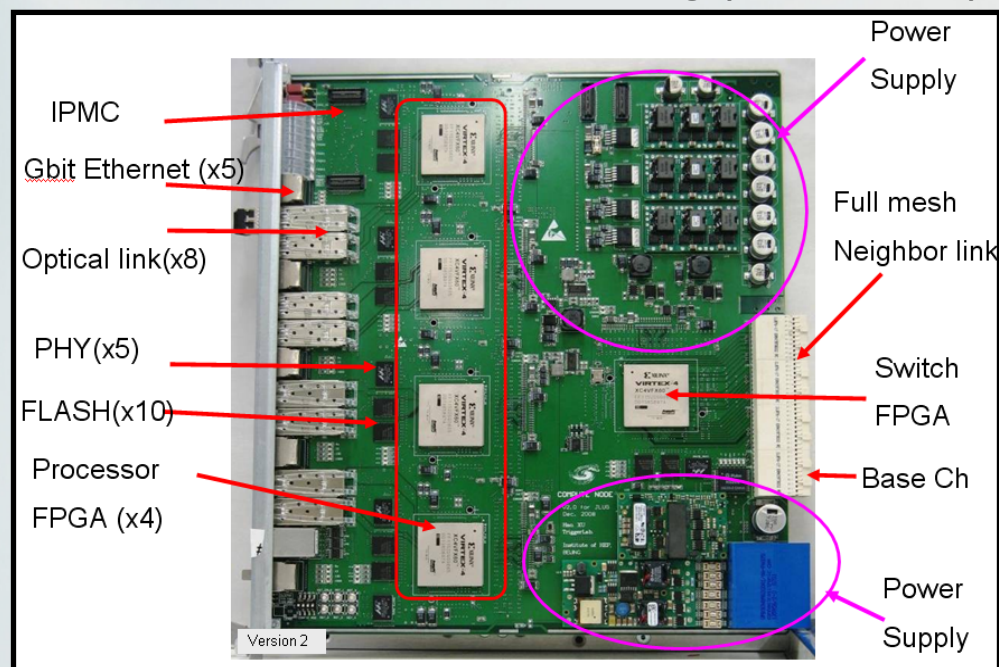
Track finding & fitting using **FPGAs**

VHDL: Very-high-speed integrated circuits **Hardware Description Language** (dataflow language)

Helix fitter for tracker of PANDA,
1st iteration only :
260 microseconds/event

Device Utilization Summary				
Logic Utilization	Used	Available	Utilization	Note(s)
Number of Slice Flip Flops	17,187	50,560	33%	
DCM autocalibration logic	14	17,187	1%	
Number of 4 input LUTs	27,438	50,560	54%	
DCM autocalibration logic	8	27,438	1%	
Number of occupied Slices	18,159	25,280	71%	

Yutie Liang (Uni Giessen)



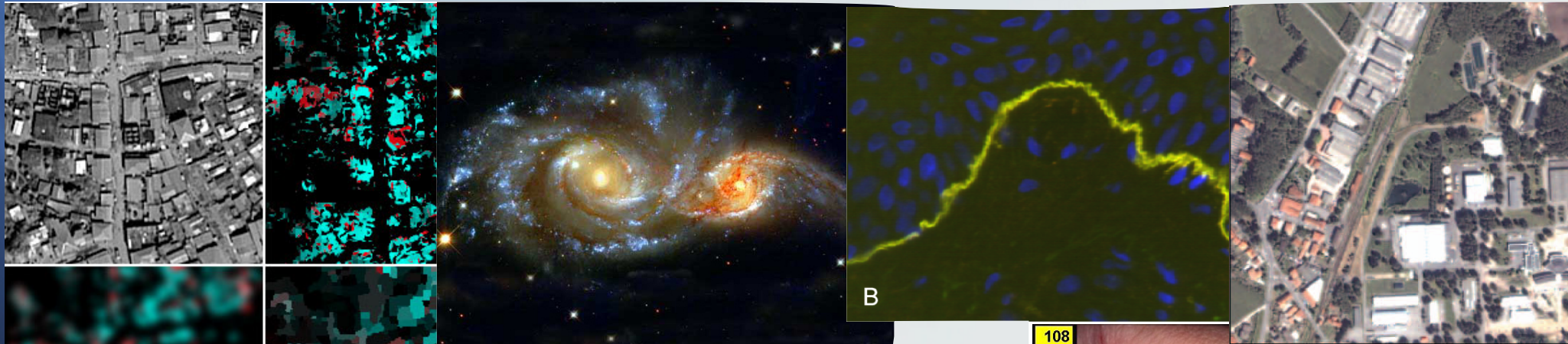
High Performance and scalable **Compute Node**

5 FPGAs: Virtex-4 FX60
10Gb DDR2 RAM
~32 Gbps bandwidth
2 PowerPCs
ATCA compliant

see talks by Yutie Liang (PANDA) and Jan Michel (CBM) on Thursday afternoon in Tier 6

Consolidating beyond our communities!

Mohammad Babai, Michael Wilkinson, JM (University of Groningen)

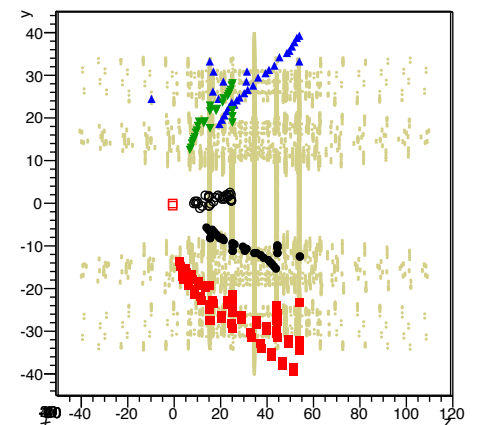
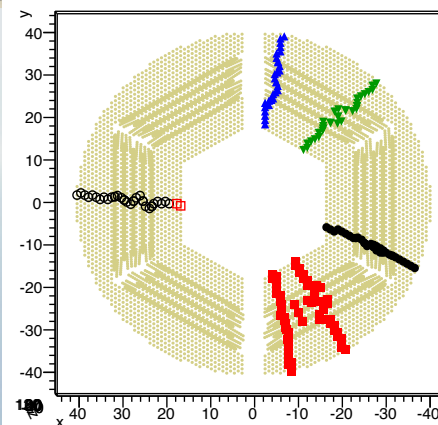
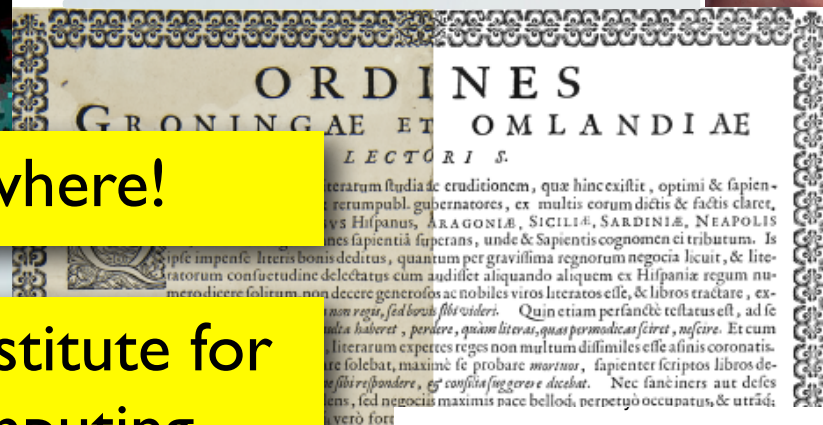


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Image data are everywhere!

Collaboration with Institute for Mathematics and Computing Science at University Groningen

Vision-based algorithm for track finding for fast track filtering



**Modern computing is all about
FRAMEWORKS**

Providing access to technologies

Details hidden for the “user”

Modular, virtual, and versatile

Geant 4



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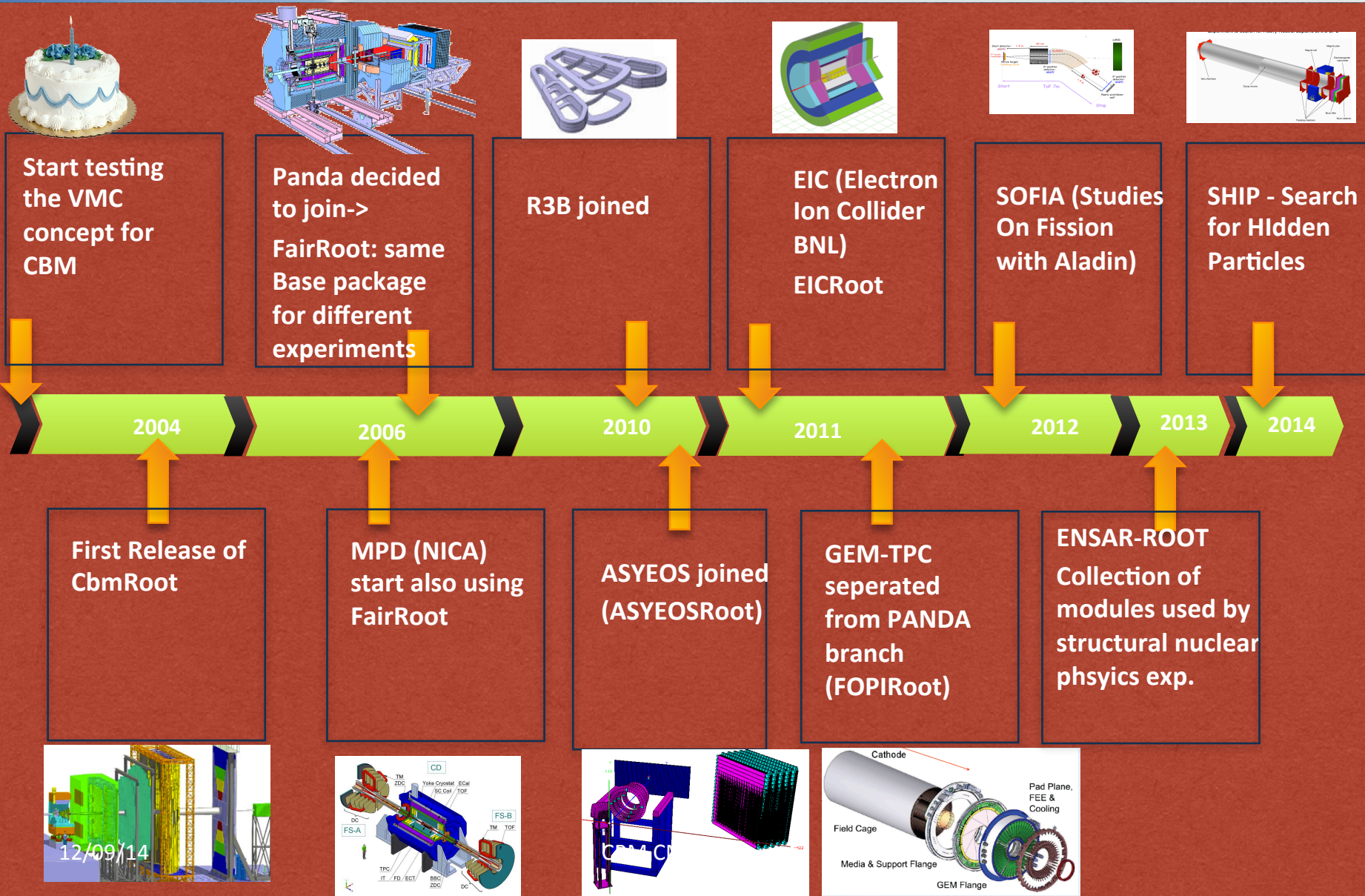
Collaborative developments

Human resource friendly

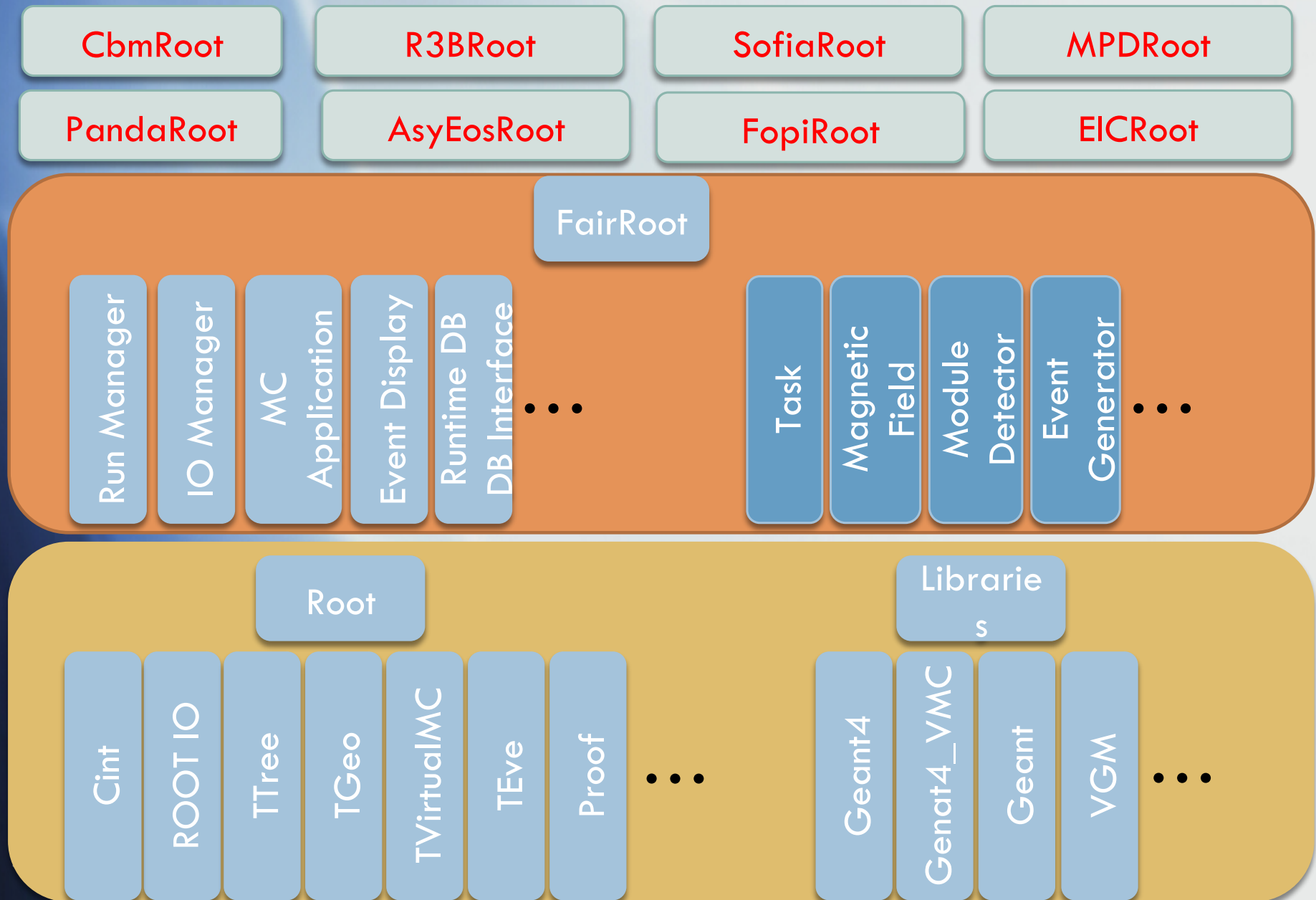
Geant 4



Example of a success story: FAIRROOT



Example of a success story: FAIRROOT



2013: FAIRROOT meets ALICE O²

2013: FAIRROOT meets ALICE O²



ALICE

ALICE O²:

- DAQ, online & offline with one framework

2013: FAIRROOT meets ALICE O²



ALICE



ALICE O²:

- DAQ, online & offline with one framework

FAIRROOT:

- Concurrency, merging online and offline

2013: FAIRROOT meets ALICE O²



ALICE



ALFA

ALICE O²:

- DAQ, online & offline with one framework

+

FAIRROOT:

- Concurrency, merging online and offline

=

ALFA:

- Join forces in a combined framework!

2013: FAIRROOT meets ALICE O²



Why join efforts?

FAIR: testing with real data and existing detectors

ALICE: interested in all the features of FAIRROOT

ALFA

2013: FAIRROOT meets ALICE O²



ALFA

Why join efforts?

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ALICE: interested in all the features of FAIRROOT

Mission of the framework

Ready for massive data reduction

Tighter coupling between online and offline reconstruction software

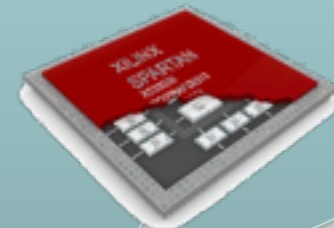
Dataflow model based on message queues

- BSD sockets API
- Bindings for 30+ languages
- Lockless and Fast
- Automatic re-connection
- Multiplexed I/O

iMatix Corporation
huge open source community



ØMQ



ALFA

FAIRMQ:

- o Asynchronous messaging toolkit
- o Broad scala of messaging pattern
- o Easy and scalable networking
- o Communication layer: ØMQ, nanoMSG

Data centers and distributed computing

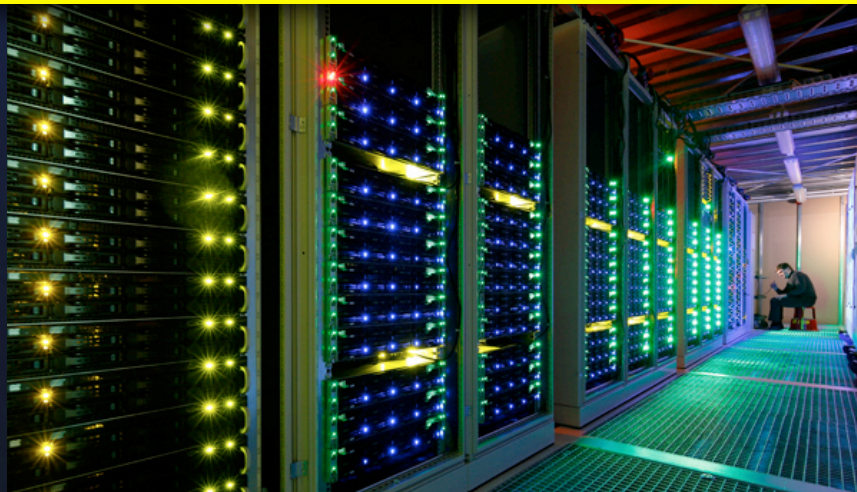
Volker Lindenstruth, Jan Trautmann

A common data center
for FAIR (**GreenITCube**)

19 M€ funding from
Helmholtz AI program

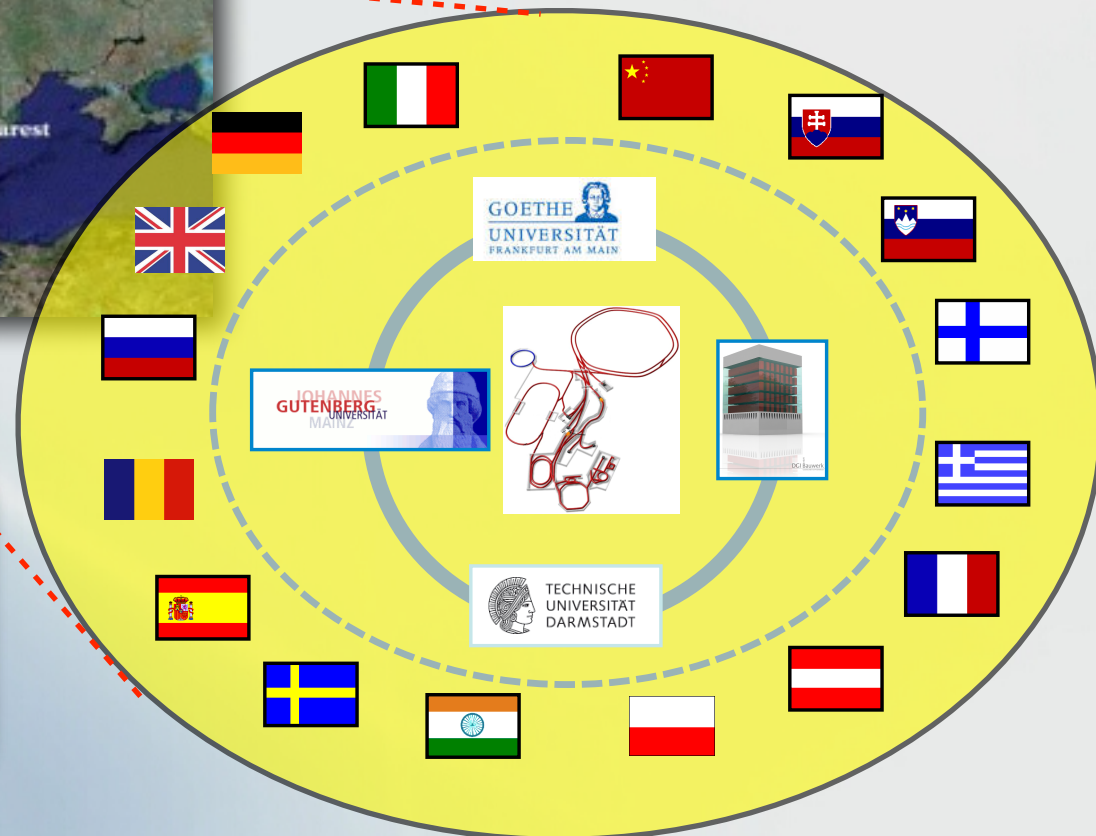
12 M€ building cost
7 M€ initial installation

Completion in Q3/2015



768 19" racks
4 MW cooling (baseline)
Max. cooling power 12 MW
>2300 fibers to exp^s

Data centers and distributed computing



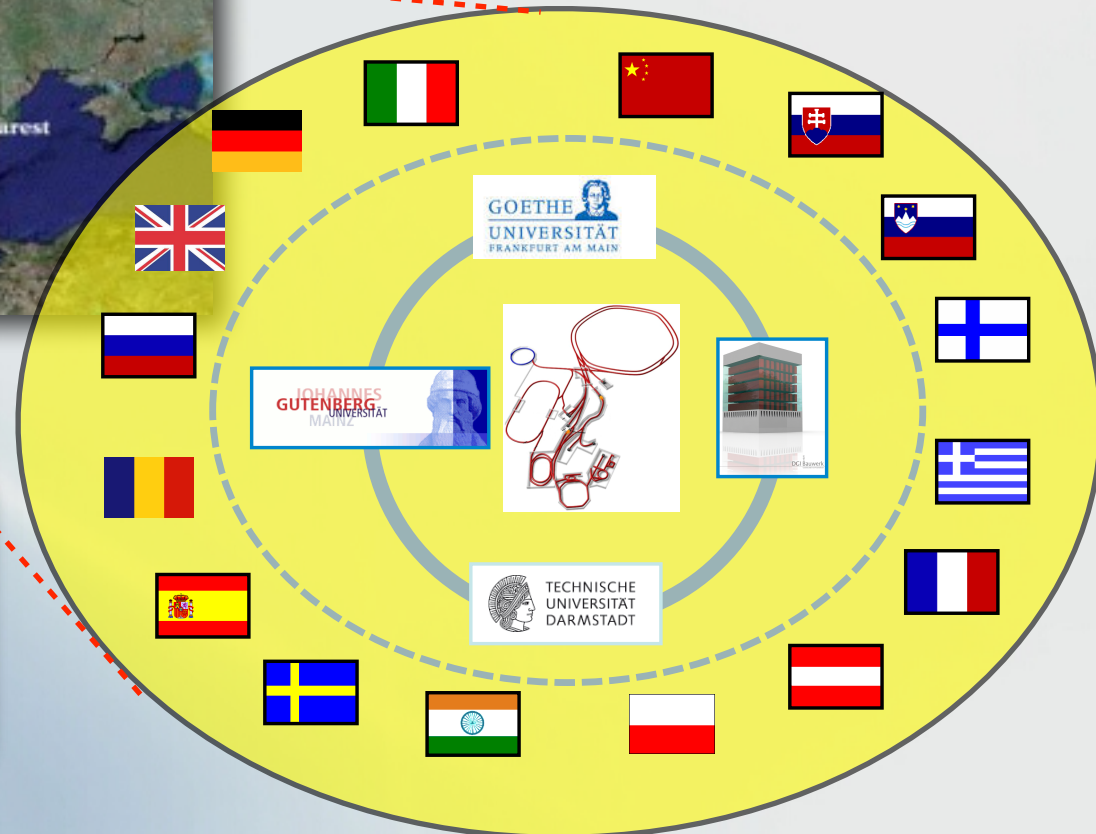
Close integration of HPC computing centers into FAIR
T0/T I: Teralink network

Data centers and distributed computing

Distributed computing

- o K.I.S.S. principle
- o small number of large centers
- o possibly heterogeneous
- o usage of standard protocols

Close integration of HPC computing centers into FAIR
T0/T1: Teralink network



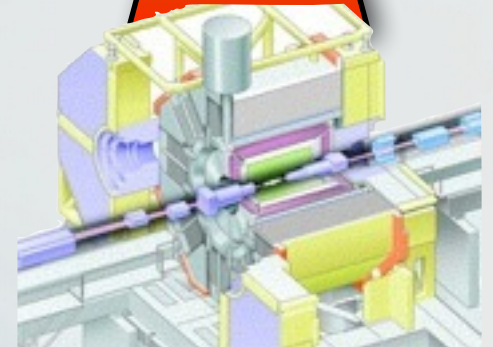
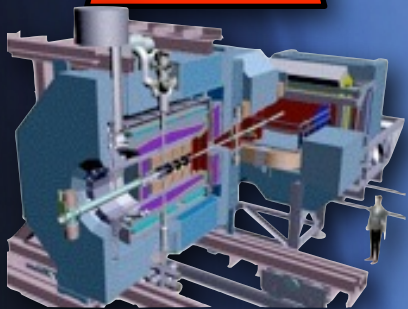
Data, data, and even more data



Data, data, and even more data



“Waterfall”



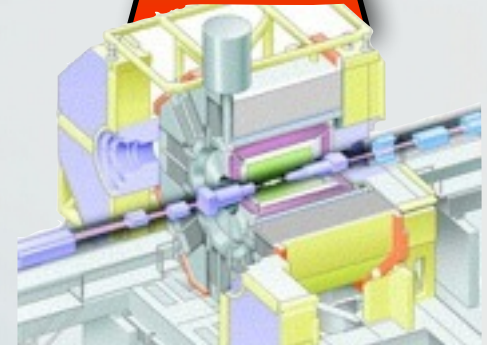
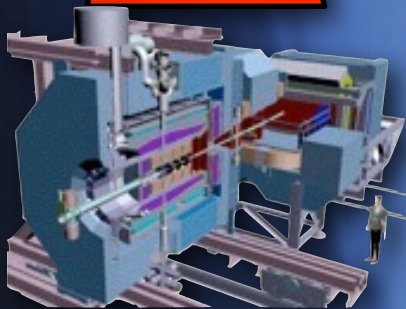
Data, data, and even more data



“Target”



“Waterfall”



Data, data, and even more data

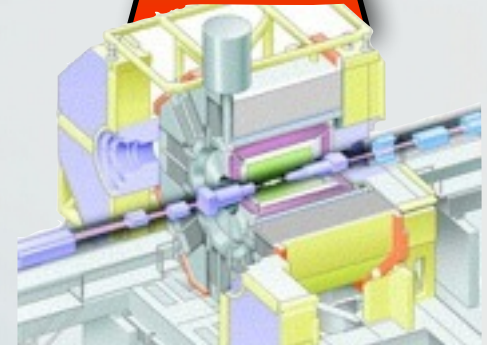
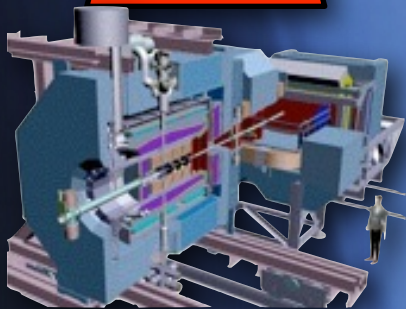


“Target”



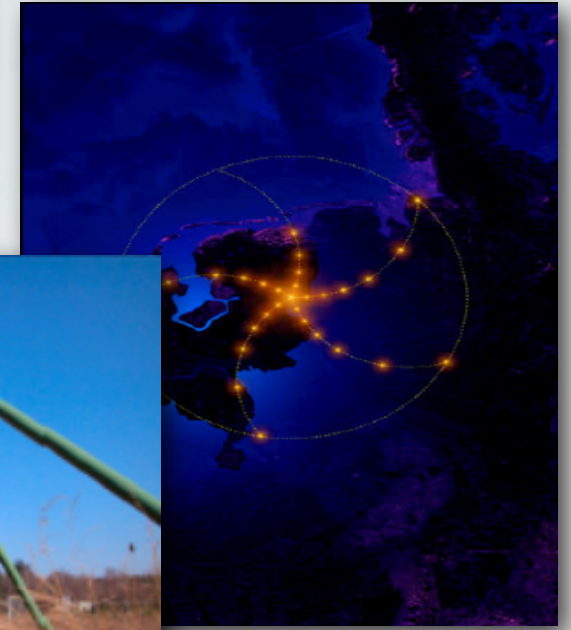
How do we manage, preserve and extract knowledge from the overwhelming flood of data efficiently, reliably and sustainably?

“Waterfall”



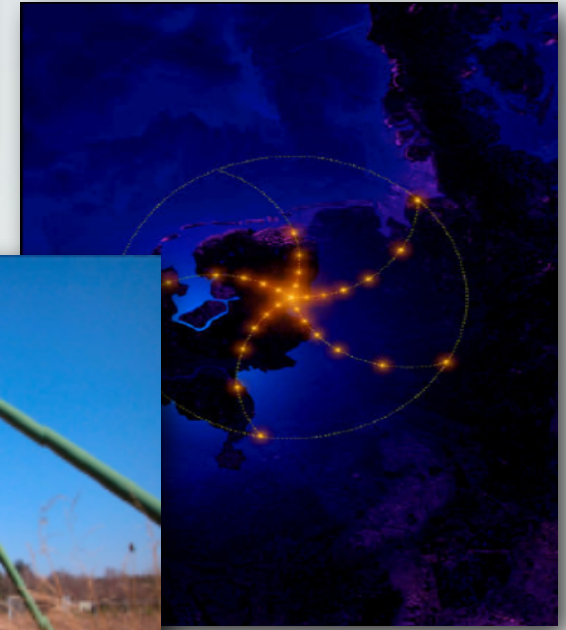
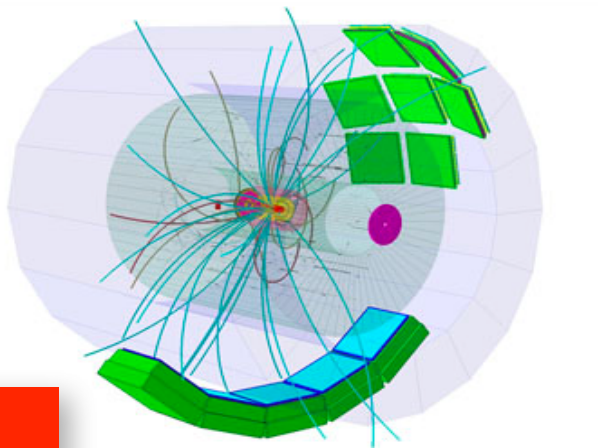
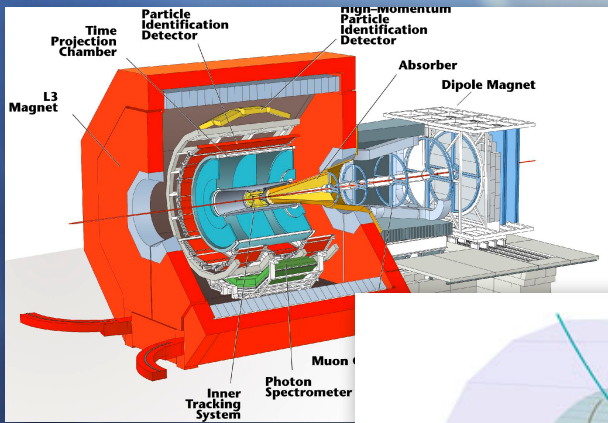
Data Processing, Publishing, and Mining

Data Processing, Publishing, and Mining



Astronomy

Data Processing, Publishing, and Mining



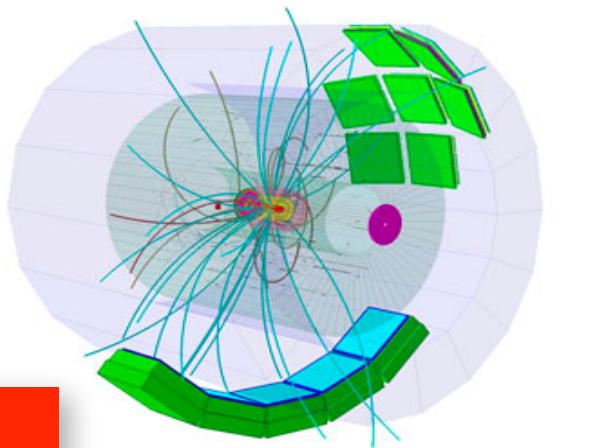
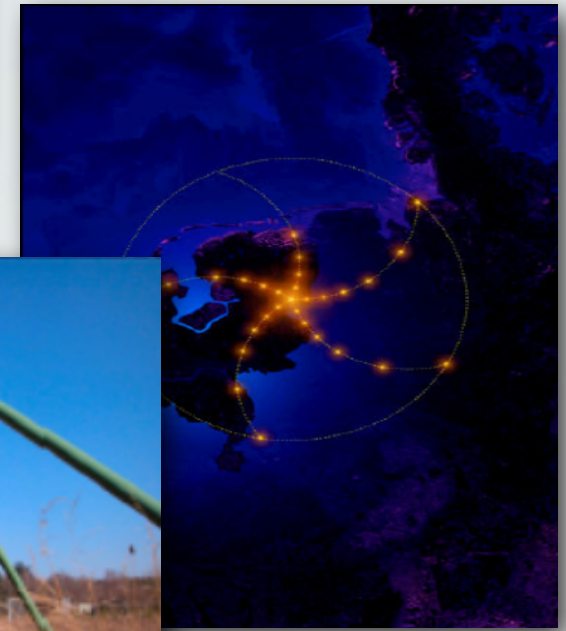
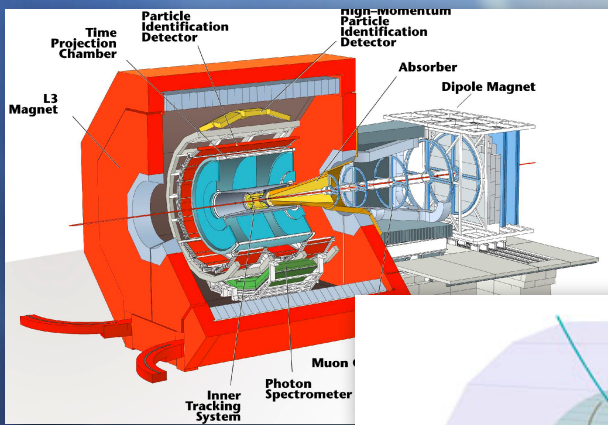
HEP
(old style)

Astronomy

“Waterfall”

forward chaining
“tier” architecture
driven by raw data
process in pipeline
operators push data
results in releases
static archive
raw data is obsolete

Data Processing, Publishing, and Mining



**HEP
(old style)**

Astronomy

“Waterfall”

forward chaining
“tier” architecture
driven by raw data
process in pipeline
operators push data
results in releases
static archive
raw data is obsolete

“Target”

backward chaining
“target” architecture
driven by user query
process on-the-fly
users pull data
information system
dynamic archive
raw data is sacred

The 21st International Conference on Computing in High Energy and Nuclear Physics

C H E P 2 0 1 5

@Okinawa Institute of Science and Technology Graduate University (OIST), Okinawa, Japan

April 13-17, 2015

Keywords for abstract submission

APPLICATION AREAS

- Computing facilities and infrastructures • Computing model
- DAQ • Data analysis • Data preservation • Data stores
- Experiment frameworks for WAN distributed computing
- Middleware and services for production infrastructures
- Monitoring • Multi-discipline • Outreach • Reconstruction
- Simulation • Trigger

TECHNOLOGIES

- Cloud computing • Collaborative tools • Continuous integration systems
- Control systems • CPU architecture, GPU, FPGA • Data algorithms
- Data handling/access • Databases • Event processing frameworks
- High performance computing • Networking • Parallel programming
- Performance and validation tools • Software design
- Software development process • Storage systems • Virtualization

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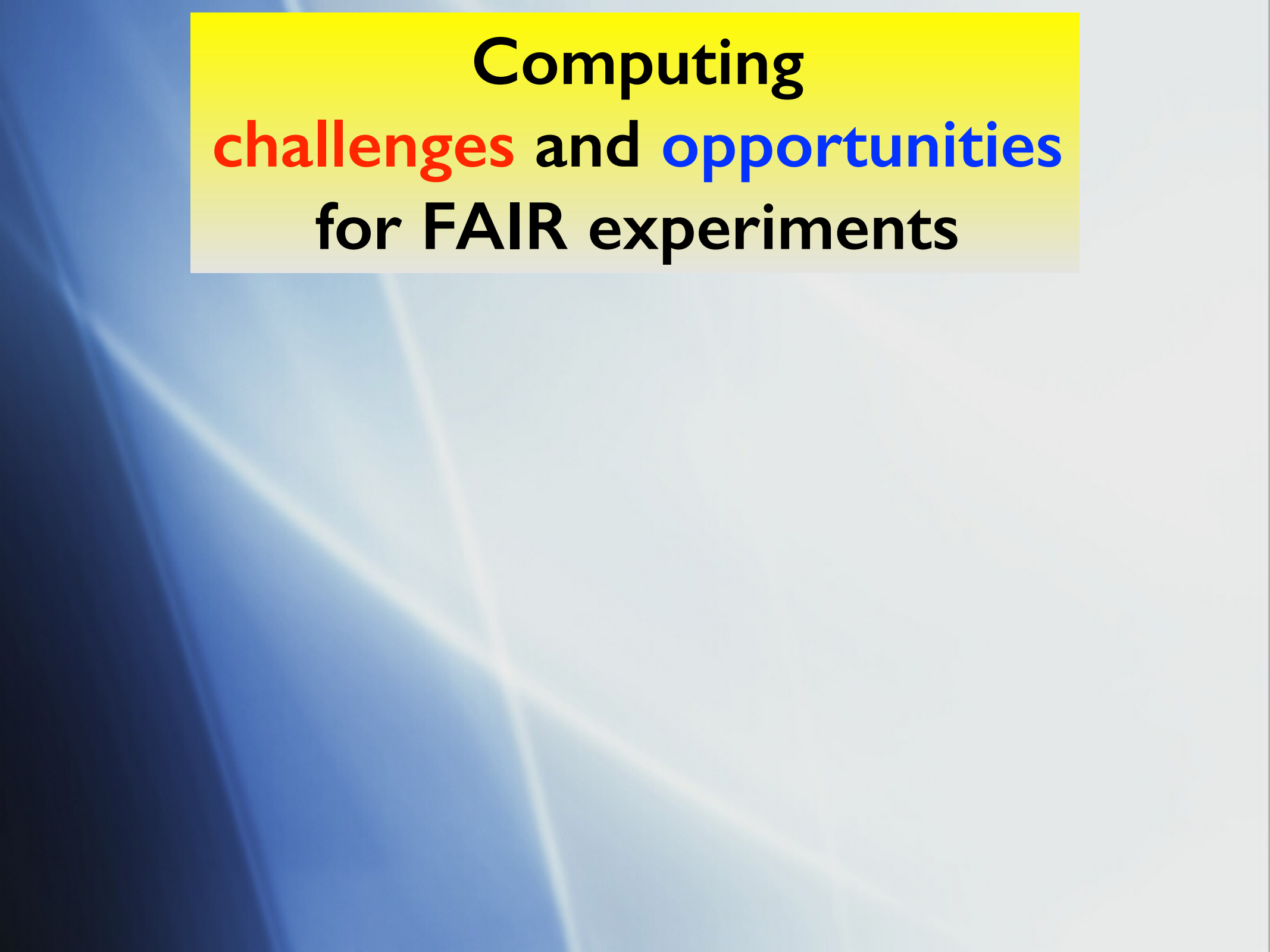
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News



2014/10/9: Abstract submission deadline has been extended until October 25.

2014/8/19: 1st bulletin is released.



Computing
challenges and **opportunities**
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The ultimate aim: place physics at the
center and physicist in the driver seat

