

用 Python 來玩粒子物理實驗吧

An Example of Using Convolutional
Neural Network to Correct Gamma-ray
Energy Measurement

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outreach @PyCon Taiwan 02, Jun, 2018





About myself and research

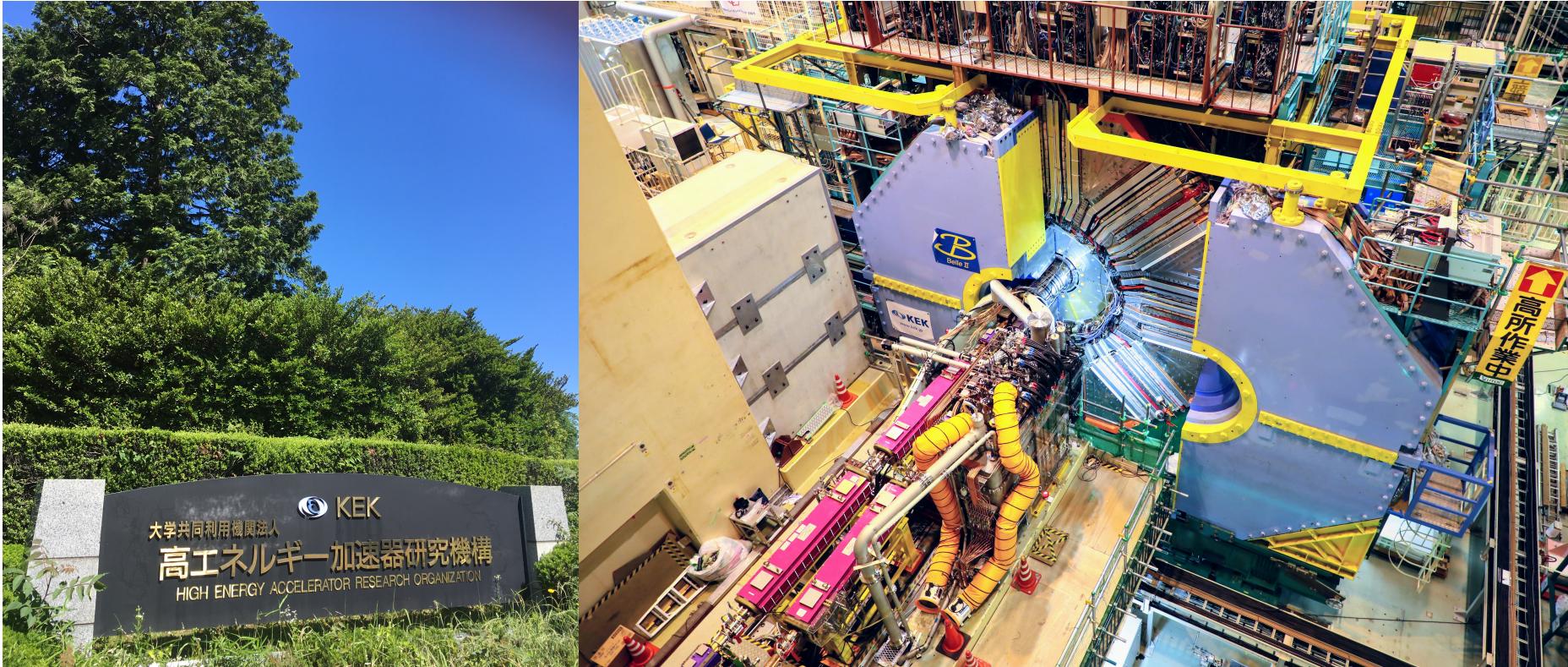


Name: 黃坤賢 (Kunxian Huang)
Position: Postdoctoral Research at NTU
Research field: Particle physics experiment
Also a PYTHON-lover



About my research

What I do at postdoc.....

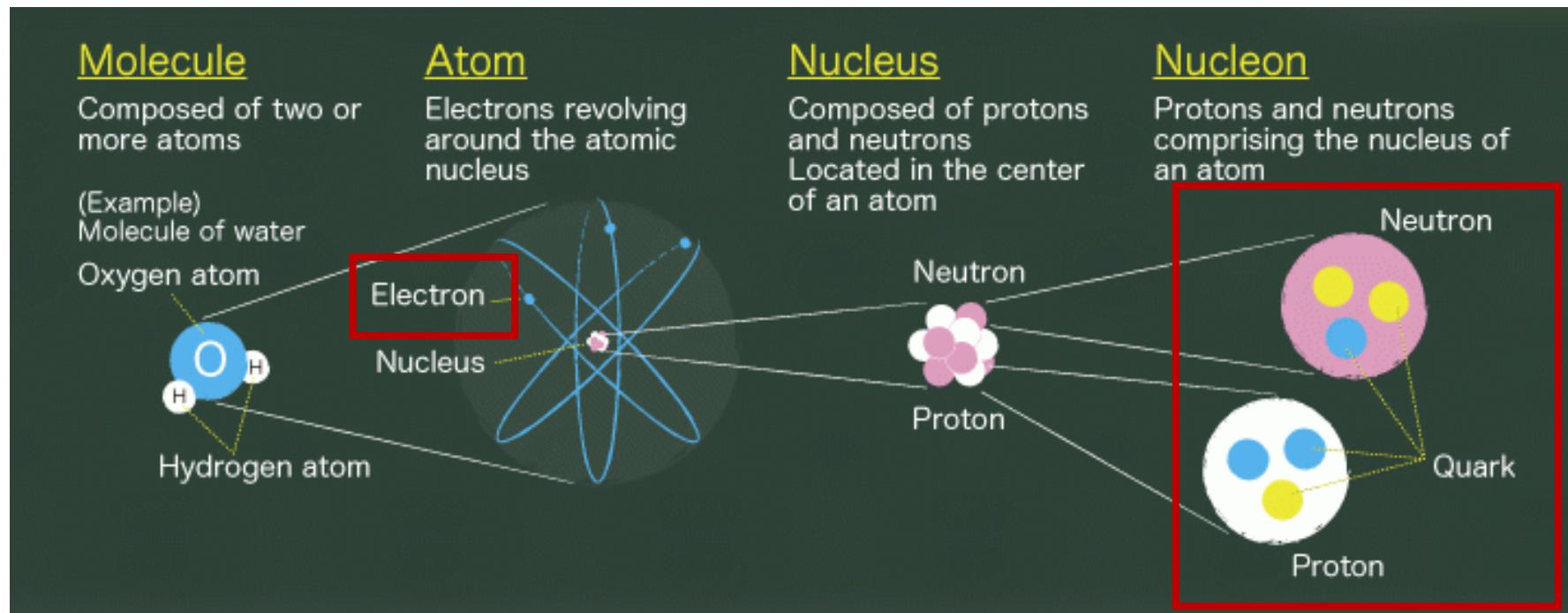


B meson experiment @Tsukuba, Japan: Belle II

Using Python to simulate, analyze, and storage for data, even distribute data worldwide groups

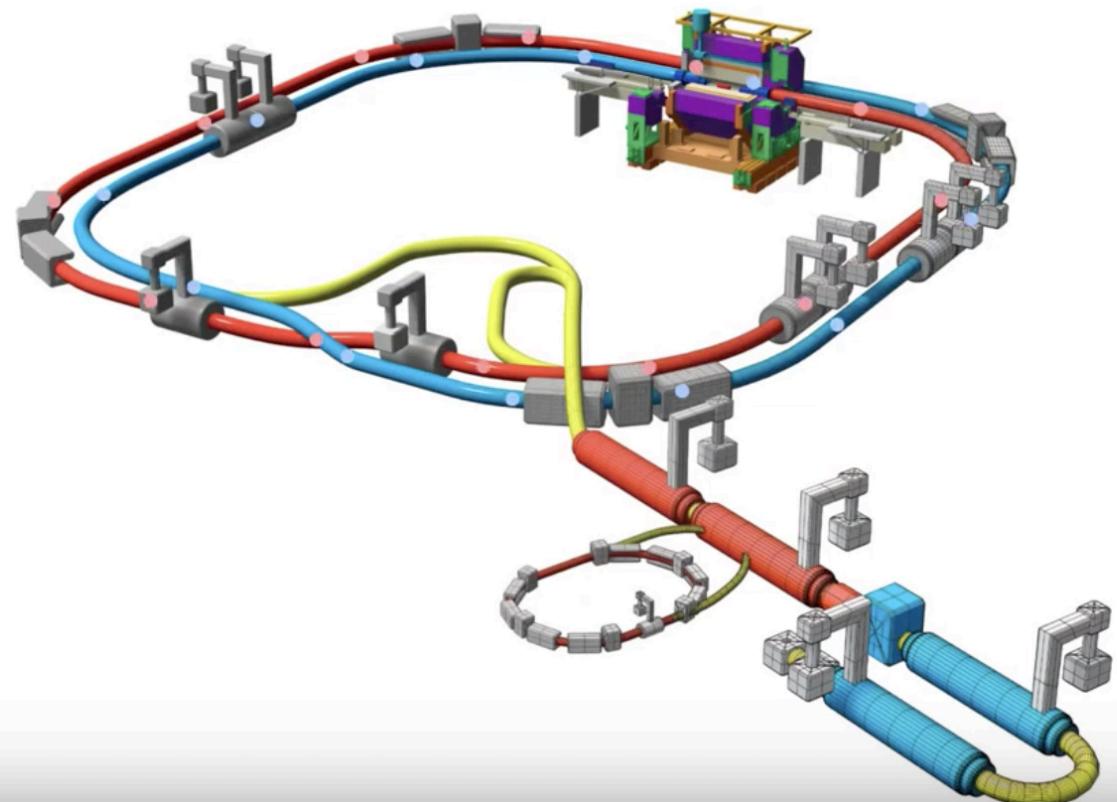
What is “particle” physics

- Particle physics studies the **tiniest** object of Nature.
→ “Elementary particles” are smaller than proton or neutron.





Particle physics experiment (or called High energy physics exp.)



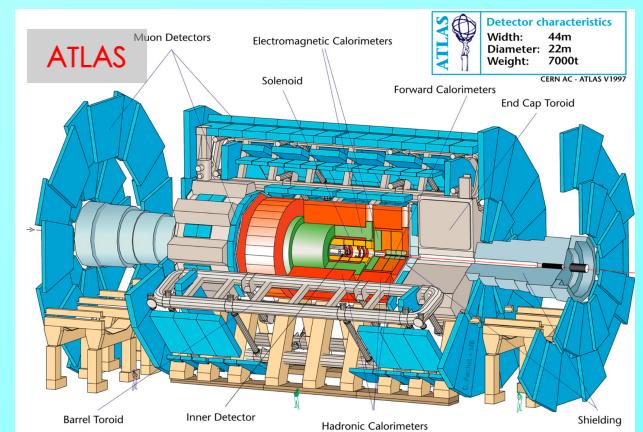
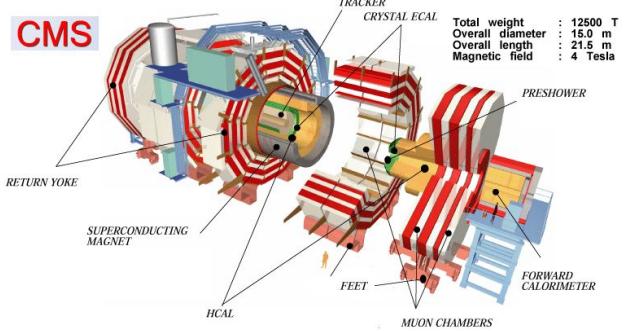
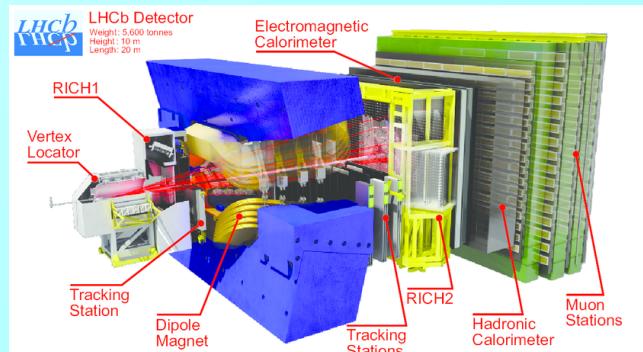
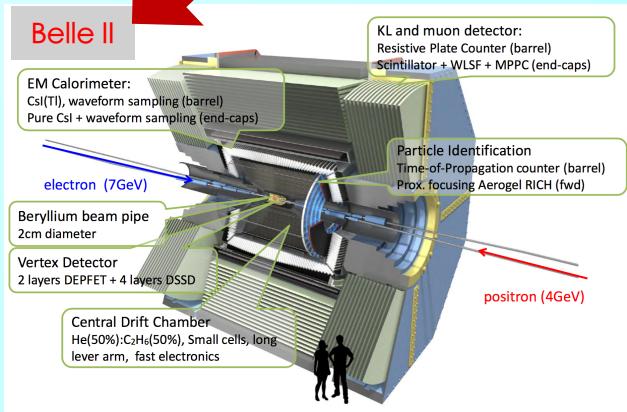
Accelerated particles are used to produce “unfamiliar” particles in our life.



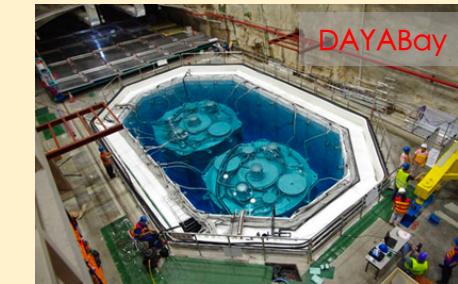
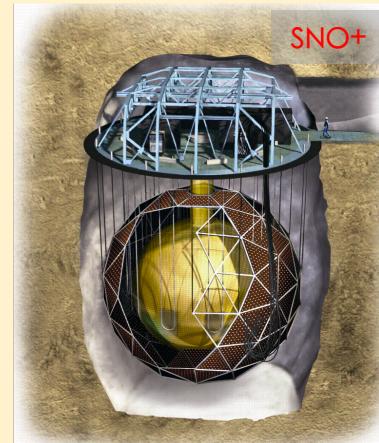
Large particle experiments worldwide

今天的主角

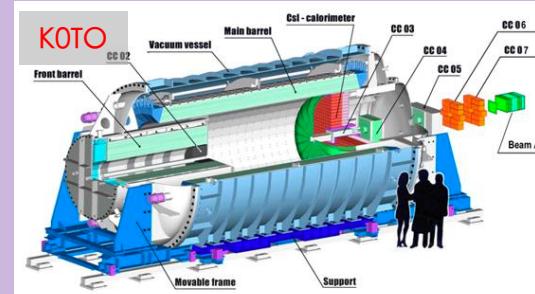
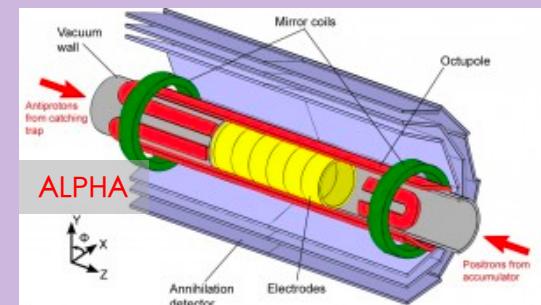
Collider exp.



Neutrino exp.



Other exp.





“Why” to do
Belle II experiment???



5 mins physics !!



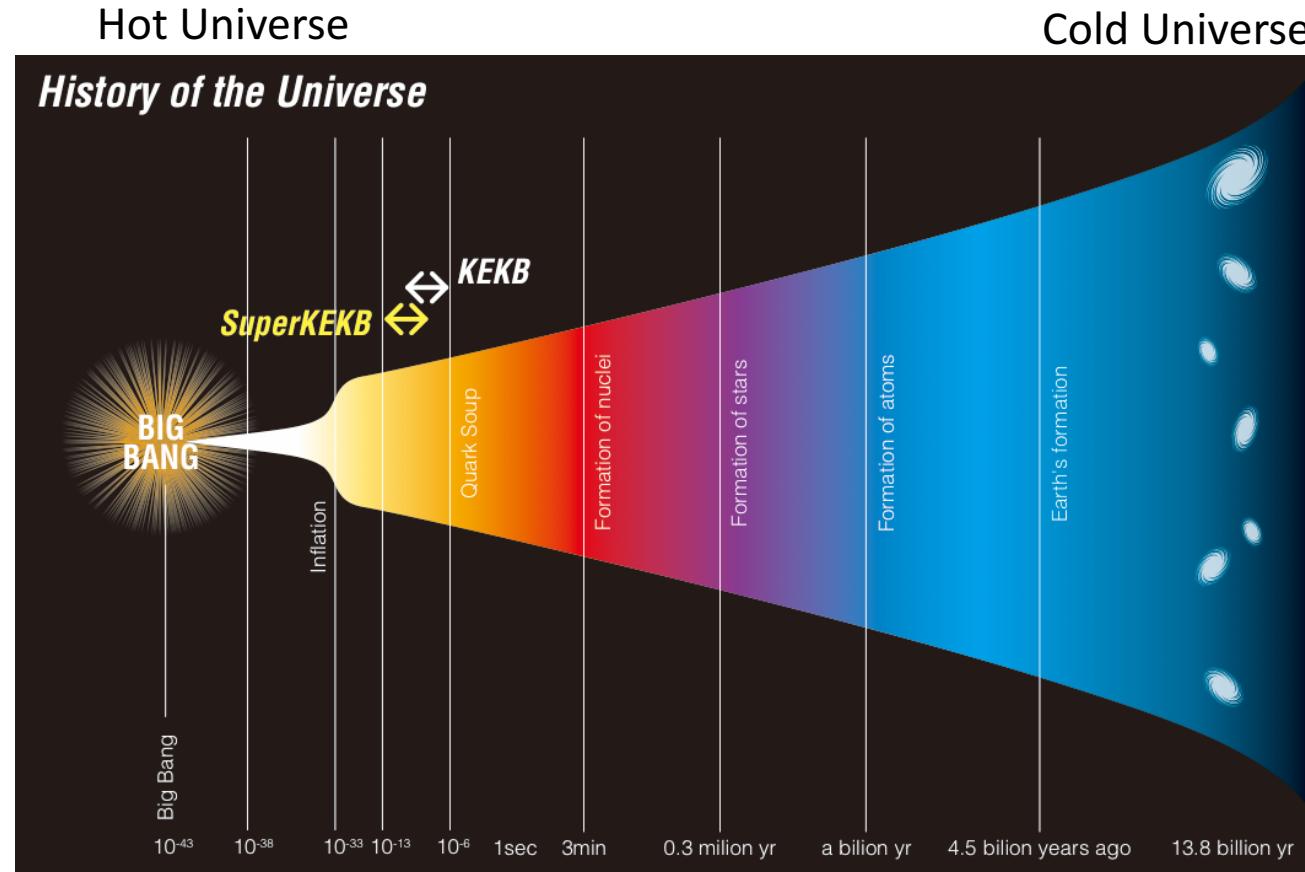


Introduction video

- Belle II detector observes B mesons produced by accelerator to measure CP violation and other new physics.
- Now, we can see matter in our Universe. Why?
CP violation is the key.
- <https://youtu.be/nGCrrgXSEOk?t=110>



Universe after Big Bang



SuperKEKB accelerator produces “un-familiar” particles that are common existing right after Big Bang.



CP violation in B meson

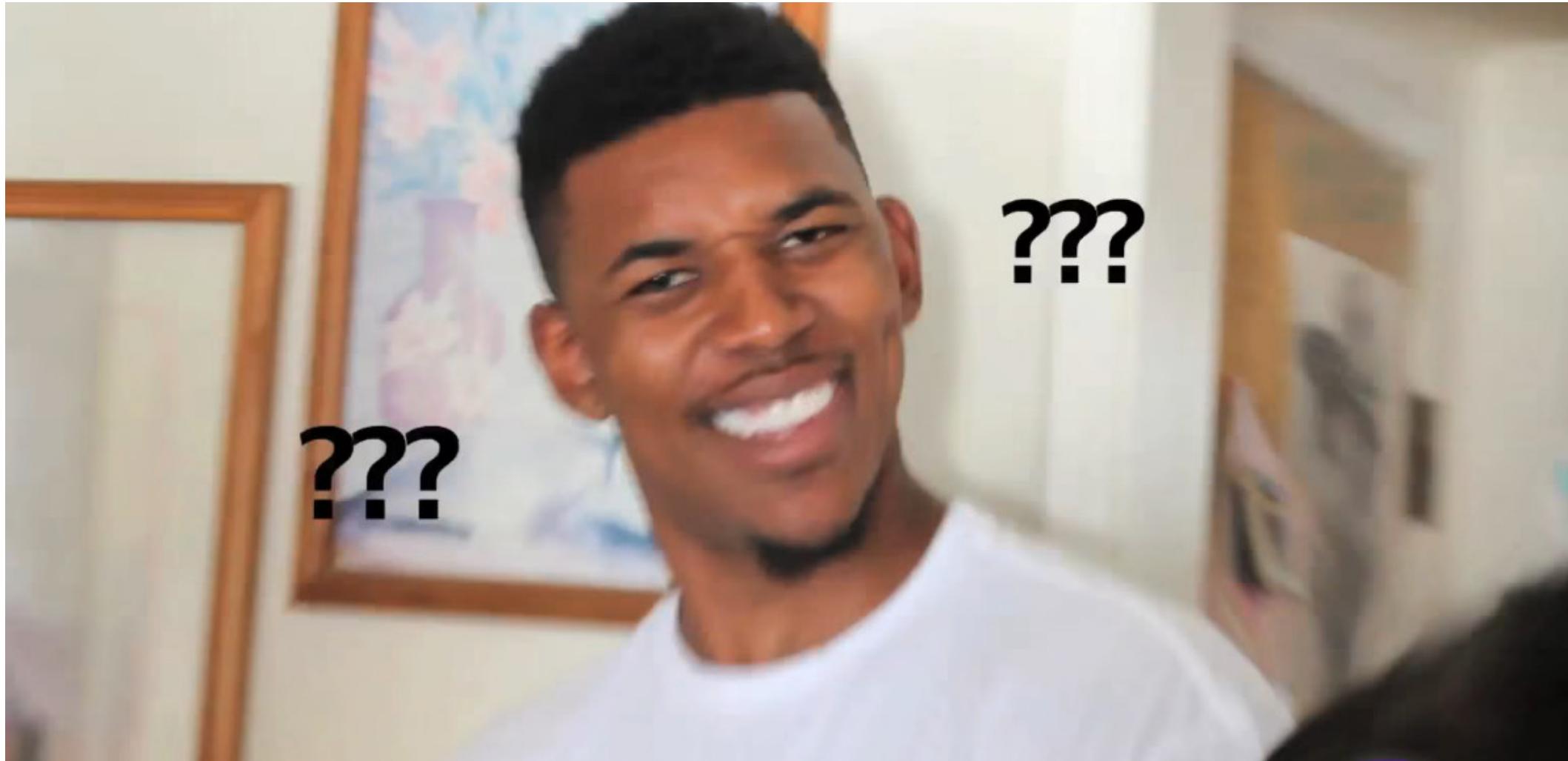
- CP violation (電荷宇稱不守恆) first discovered in kaon meson decay by James Cronin, Val Fitch in 1964.
- M. Kobayashi and T. Maskawa introduced the mix matrix with three generation in standard model, and it also implies CP violation in the third quark decay in 1973.
- Sanda and Carter proposed CP violation in B meson decays in 1980.



M. Kobayashi and T. Maskawa



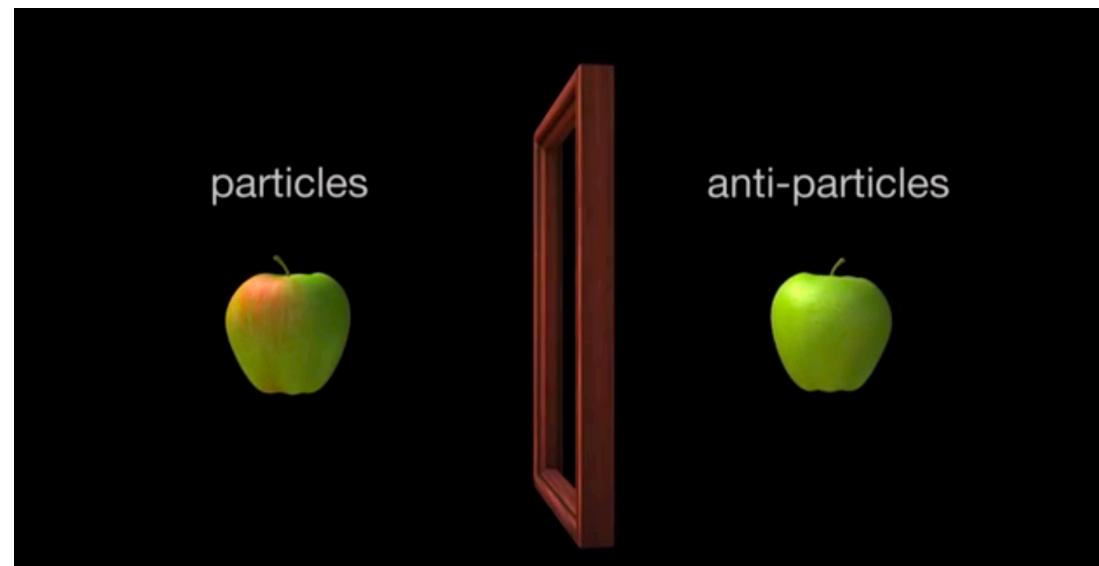
CP violation





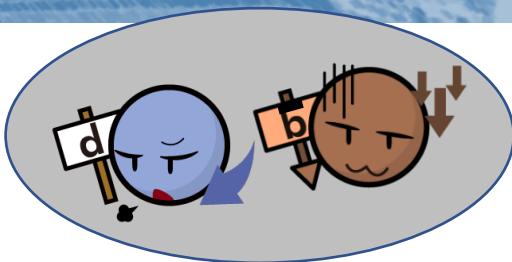
CP violation

- We believed that particle and anti-particle are the same (except **charge** and **parity**). But, why in our universe there are only matters (particles)??
- Because of charge-parity operator can be the particle-antiparticle conjugate operator, CP violation means **difference of particle-antiparticle**.
- Differences between matter-antimatter except C,P: life time, and decay branching ratio.





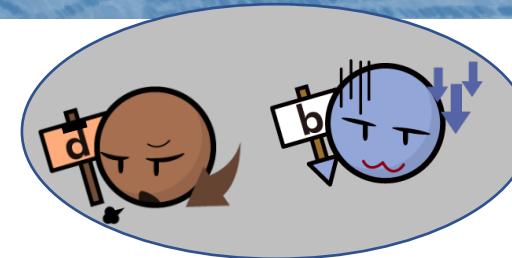
B mesons (B 介子)



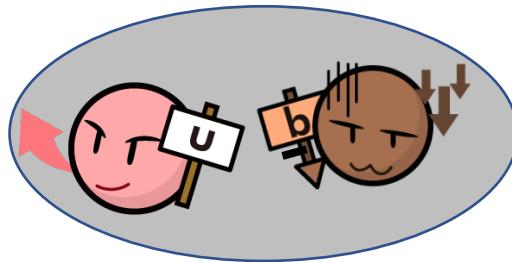
B^0 ($d\bar{b}$)



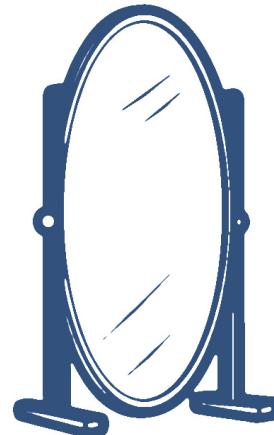
CP mirror



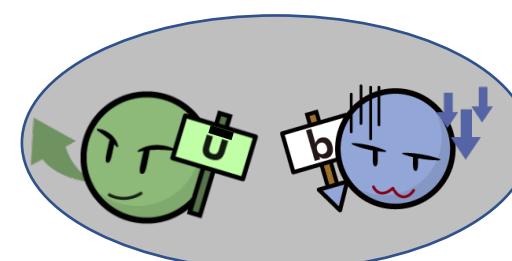
\bar{B}^0 ($\bar{d}b$)



B^+ ($u\bar{b}$)



CP mirror



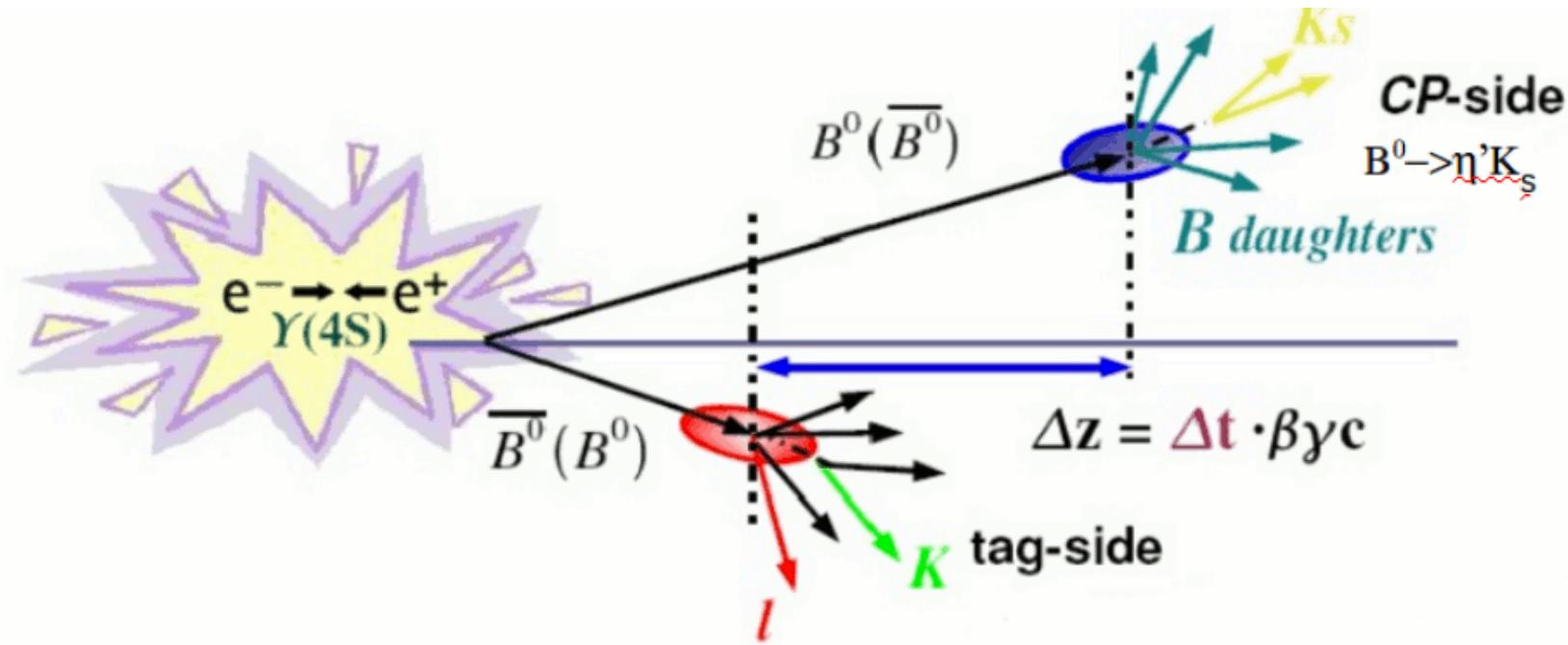
B^- ($\bar{u}b$)

Particle: More common in Universe

anti-Particle: rare in Universe

From Higgs-Tan
<http://higgstan.com/>

CP violation in B meson

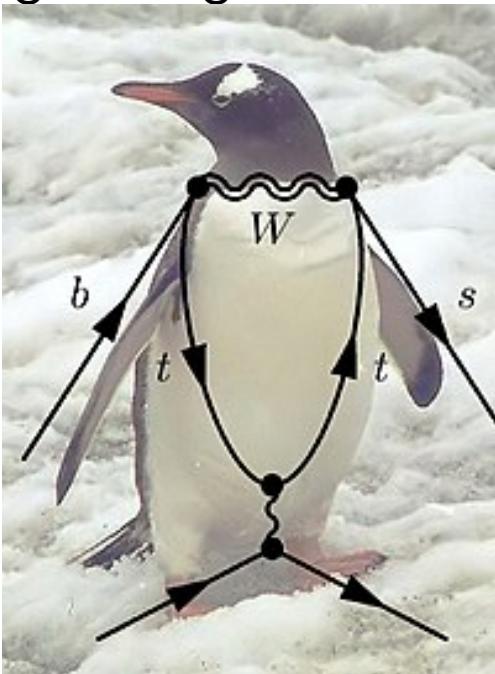


- The longer life time , the farer they goes.
- B meson and anti-B meson have different behavior in some decay modes.



Other new physics in Belle II

From wiki page of
“Penguin diagram”



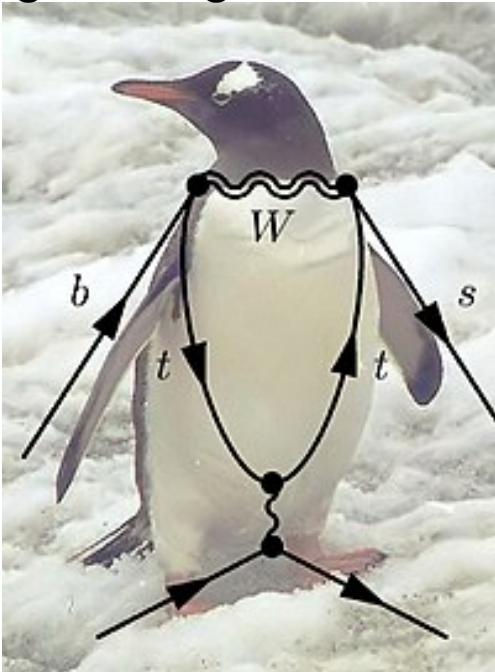
New intermedia particle
in penguin diagram decay



Dark matter search

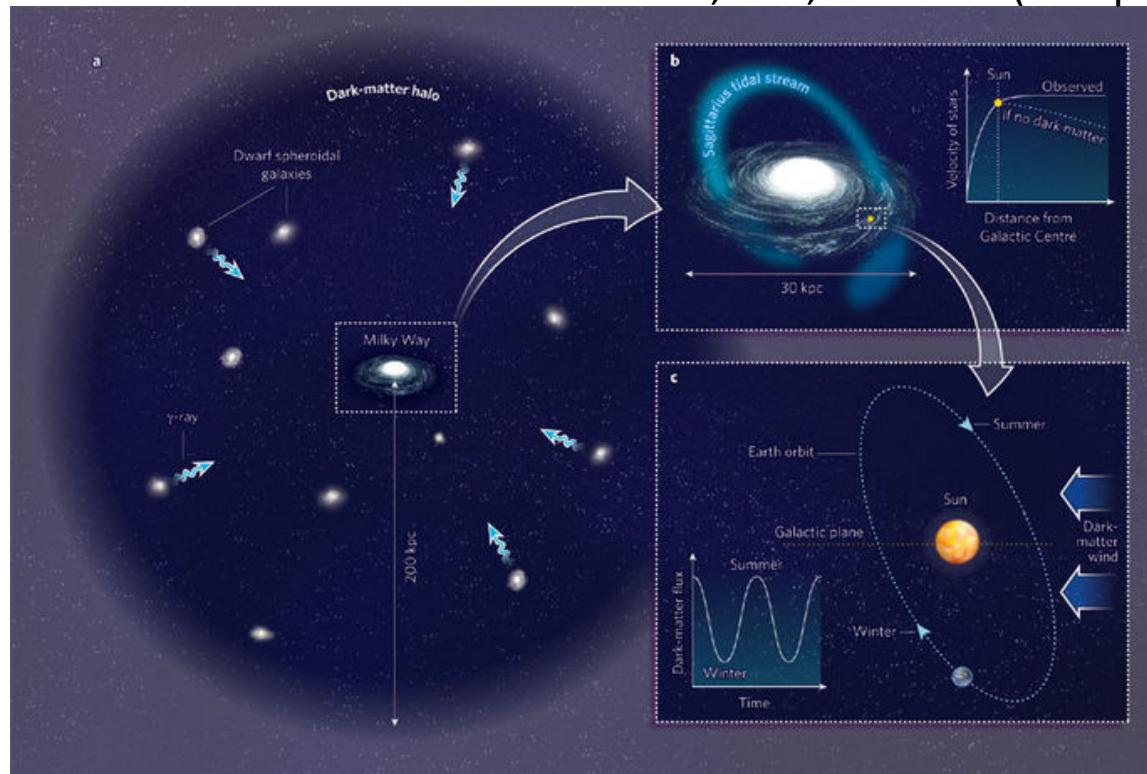
Other new physics in Belle II

From wiki page of
“Penguin diagram”



New intermedia particle
in penguin diagram decay

From Nature, 458, 587–589 (02 April 2009)



Dark matter search
~~(Gundam)~~



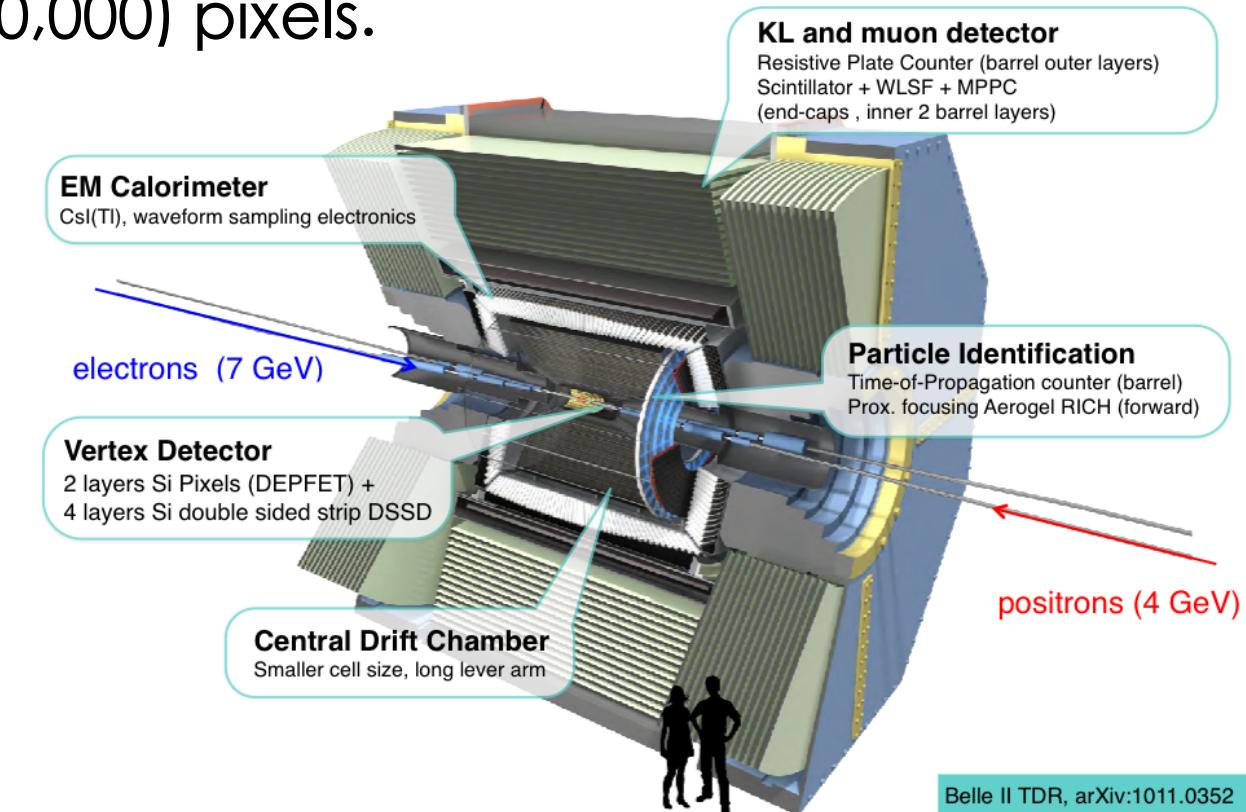
Physics explanation finished!!





Big Data in particle physics exp.

- Large detector use complex of sub-detectors to measure a single event. Parts of sub-detectors can detect specific particles.
- Recent years, each sub-detector are equipped with $O(1,000\sim 100,000)$ pixels. So, the complex detector have $O(1,000,000)$ pixels.
- **Big Data** techniques are widely used in large particle experiments.
- Example: Belle II detector
- Data size: 1MB
(after event builder by firmware)
- Trigger rate (event rate): max 30 kHz.
- Data rate: max 30 GB/sec



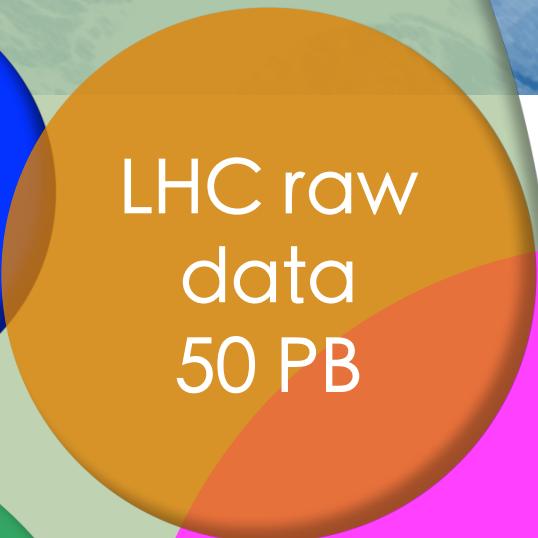


Big Data Science

Google
internet
archive
 $\sim 1\text{EB}$

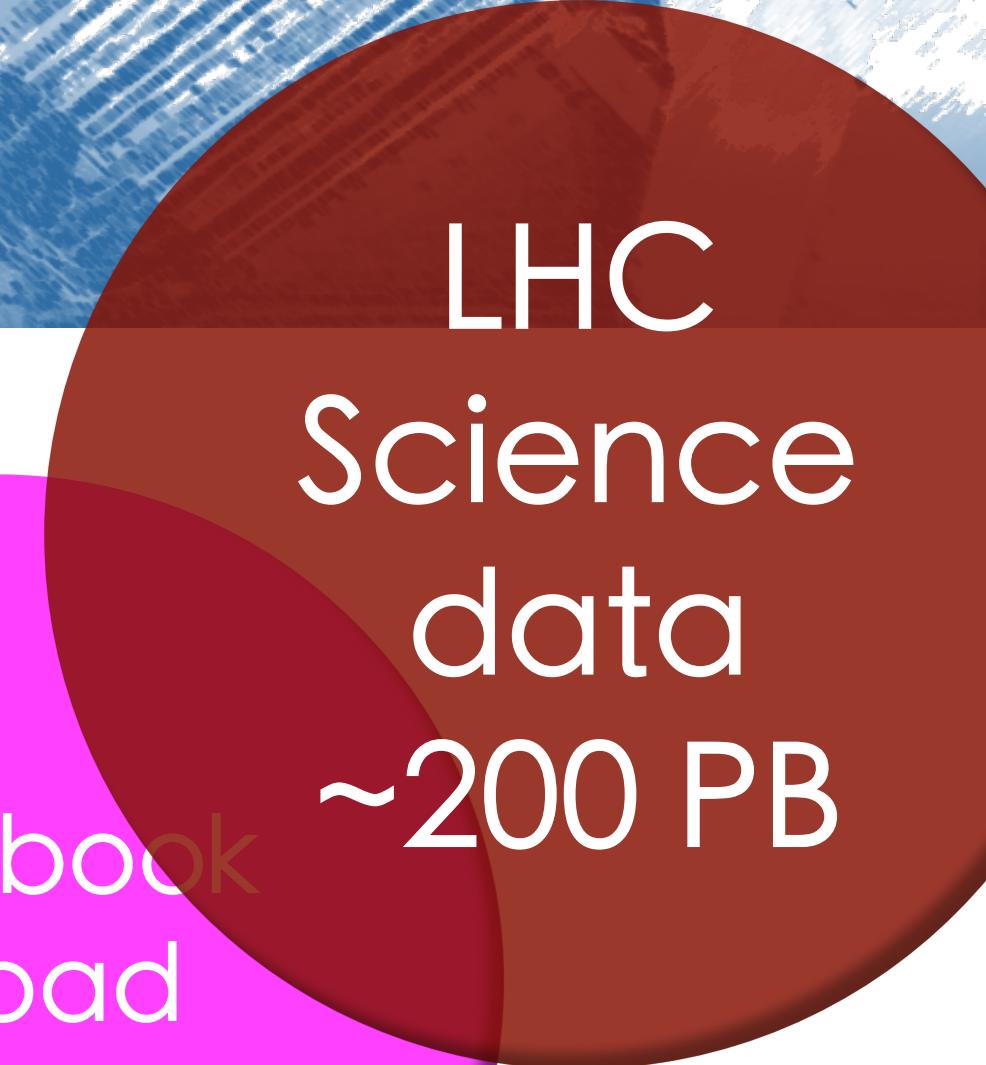


Belle II
data+
simulation
 $30\sim 50\text{ PB}$



50 PB

Facebook
upload
 $\sim 180\text{PB}$

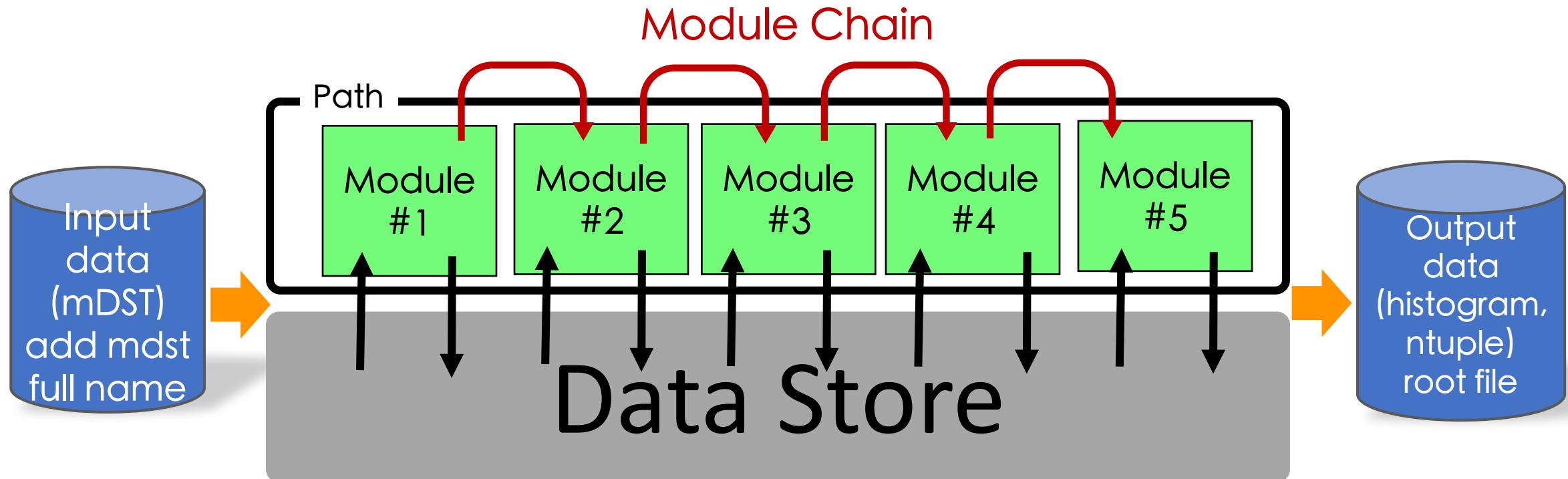


LHC
Science
data
 $\sim 200\text{ PB}$

Information from
Elizabeth Sexton-Kennedy (Fermi National Accelerator Lab. (US))
<https://indico.cern.ch/event/658060/contributions/2844782/>



Structure of Belle II data analysis



C++14:

- Module
- code to build particles from primitive objects (like tracks and calorimeter clusters)

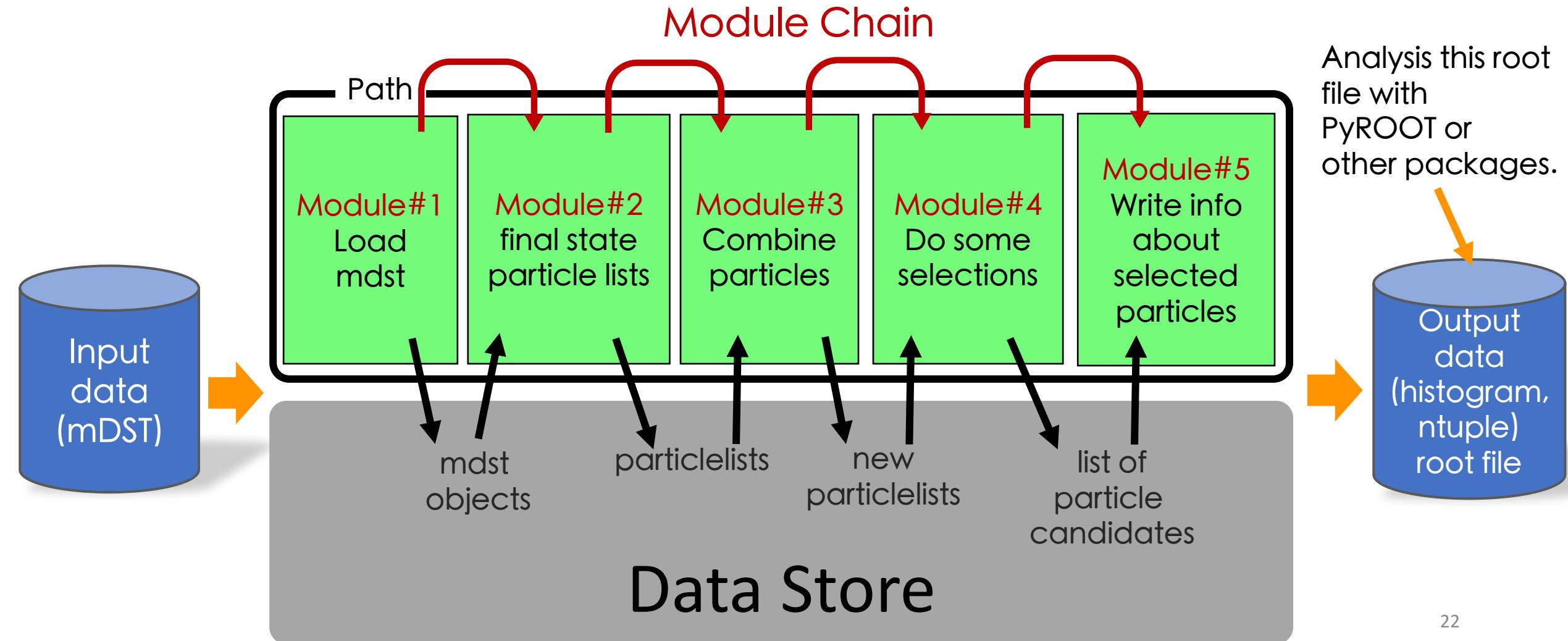


Python3.6:

- Load and configure C++ modules
- Other high-level analysis tasks.



Example of B meson candidates reconstruction



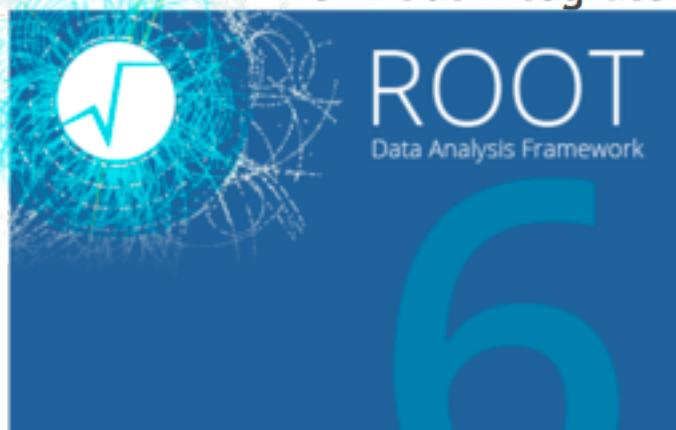


Numerical Analysis with Python at Belle II

- “ROOT” file format is usually for data storage and analysis format in HEP.
- Website: <https://root.cern.ch/>

ROOT is ...

A modular scientific software framework. It provides all the functionalities needed to deal with big data processing, statistical analysis, visualisation and storage. It is mainly written in C++ but integrated with other languages such as **Python** and R.



Installation instruction: (<https://root.cern.ch/building-root>)

- use of CMake cross-platform build-generator tool to build and install
- download source code: <https://root.cern.ch/get-root-sources>
- Don't forget to turn on cmake option “python” to install PyROOT



PyROOT
python binding
for ROOT



root_numpy
interface
between ROOT
and NumPy



root_pandas
loading/saving ROOT
files as pandas
DataFrames



ROOT
written in C++



Installing PyROOT, root_numpy, root_pandas

- Pre-requires: ROOT, Numpy, Pandas
- PyROOT
 - website: <https://root.cern.ch/pyroot>
 - enable option of “python” when installing ROOT using cmake.
- root_numpy
 - website: http://scikit-hep.org/root_numpy/index.html
 - installation:
 - git clone git://github.com/rootpy/root_numpy.git
 - python setup.py install --user
- root_pandas
 - website: https://github.com/scikit-hep/root_pandas
 - installation:
 - git clone https://github.com/scikit-hep/root_pandas.git
 - python setup.py install --user



Training course with Jupyter Notebook

- For new collaborators, Belle II hold a start kit workshop for 3 days per 4 month.
- Jupyter Notebooks on internal server.
- Self-sustained: combine documentation + code
- Aim for large tutor/student ratio.

jupyter B2T_Advanced_1_CS Last Checkpoint: 02/03/2018 (autosaved)

File Edit View Insert Cell Kernel Widgets Help

Cell Toolbar

In [10]:

```
%matplotlib inline
import root_pandas
from matplotlib.pyplot import *

# short sample, freshly made
#B = root_pandas.read_root('ana_PhiKs_ntuple_sample.root',key='b0phiKs')

# premade sample
B = root_pandas.read_root('~tamponi/public/CSTutorial/ana_PhiKs_ntuple_sample.root',key='b0phiKs')

B.describe()
```

Out[10]:

	exp_no	run_no	evt_no	B0_M	B0_ErrM	B0_SigM	B0_phi_M	B0_phi_ErrM	B0_phi_SigM	B0_K_S0_M	B0_hso20	B0_hso22	B0_hso24	E
count														
mean														
std														
min														

In [15]:

```
myrange = (-0.21,0.21)
figure(figsize=(10,7))
h = hist(B.B0_deltae[B.B0_isSignal==1],nbins,myrange,alpha=1.,color='red',label='signal')
h = hist(B.B0_deltae[B.B0_isSignal==0],nbins,myrange,alpha=0.5,color='green',label='background')
legend()
```

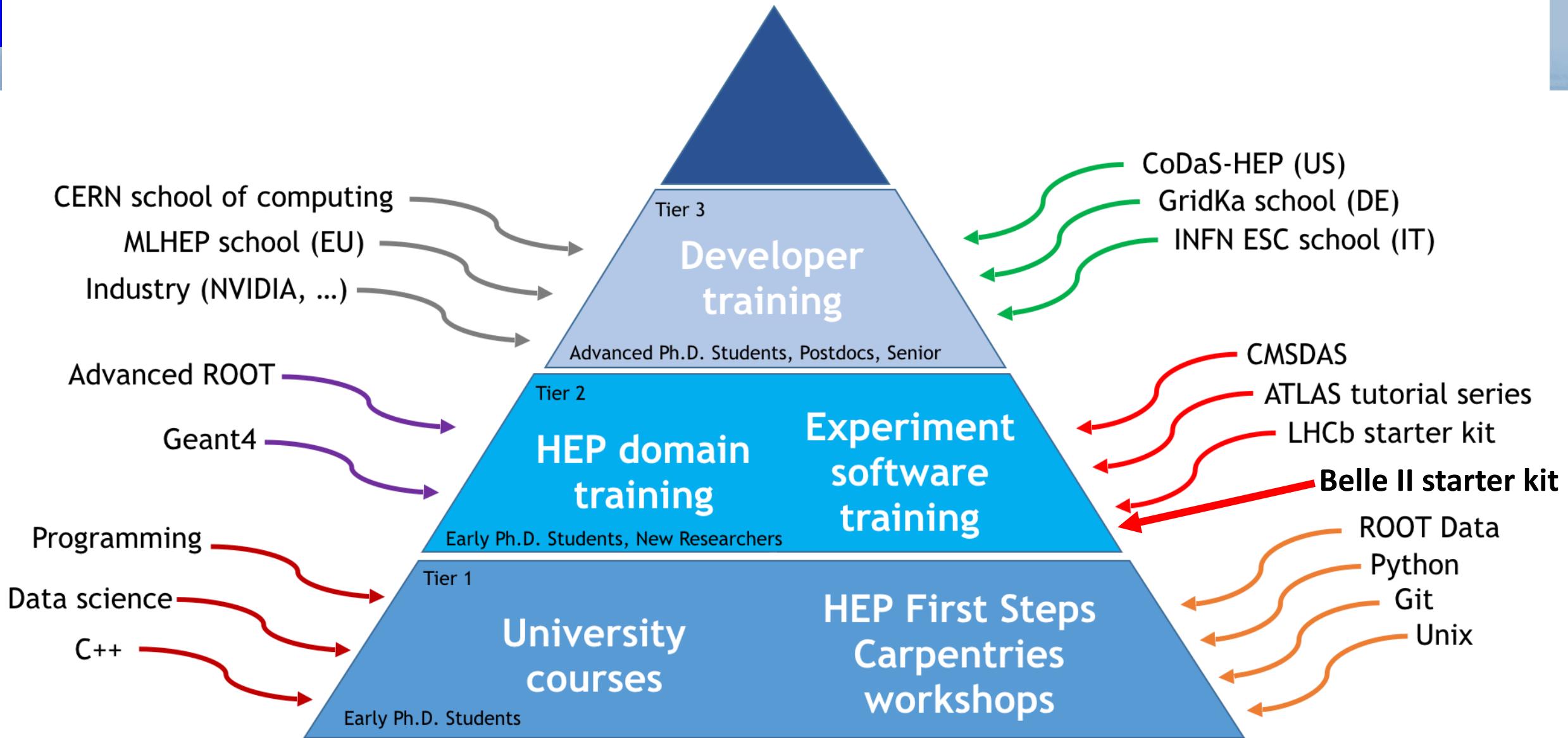
Out[15]:

```
<matplotlib.legend.Legend at 0x7ff10688af98>
```

@Belle II starter kit
training for new comers

HEP Software Training

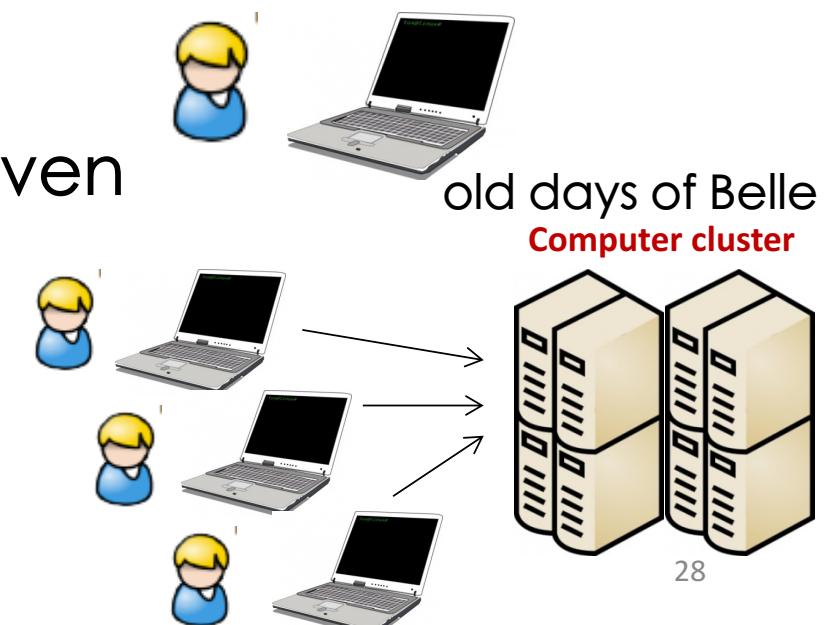
From Peter Elmer
@WLCG and HSF workshop 2018





Needs of Grid Computing

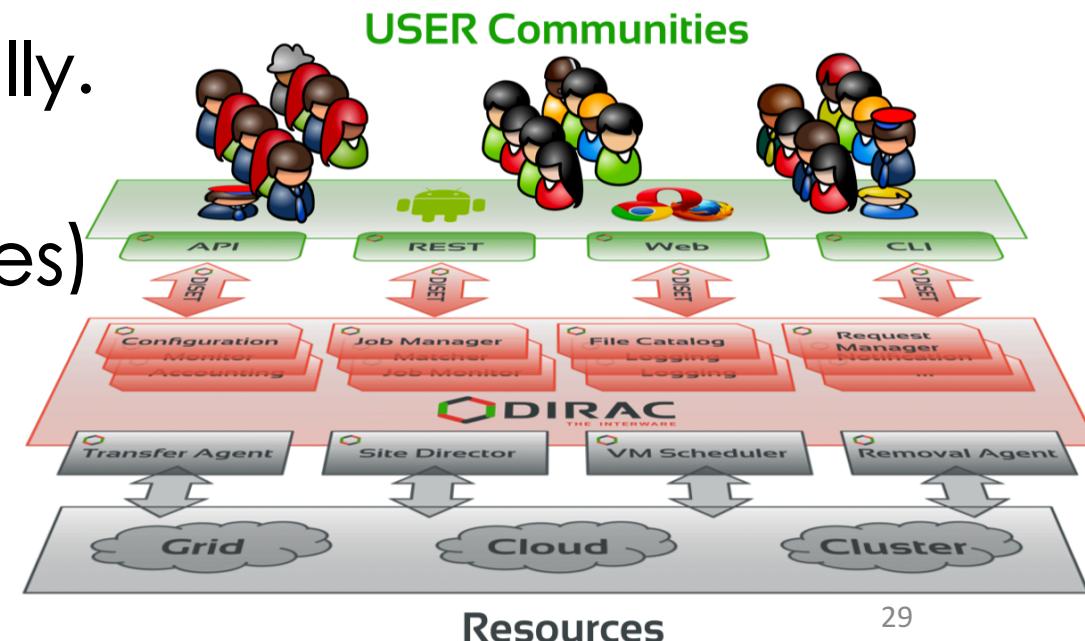
- In the age of Belle, KEK constructed central computing cluster for the computing resource.
- Belle II: 50 times luminosity, higher background rate, 1 MB size per event.
- Bigger data, need more CPUs & memories.
→dramatically larger computing resources are needed.
- Belle II collaboration: 25 countries,
~100 institutes world wide.
- Each institute contribute their cluster, grid, or even buy cloud service from company.
- Concept of distributed resource usage:
 - Send Job (to central service)
 - “Heterogeneous” resource distributed over the world.
 - Data distributed over the world.





Distributed GRID computing

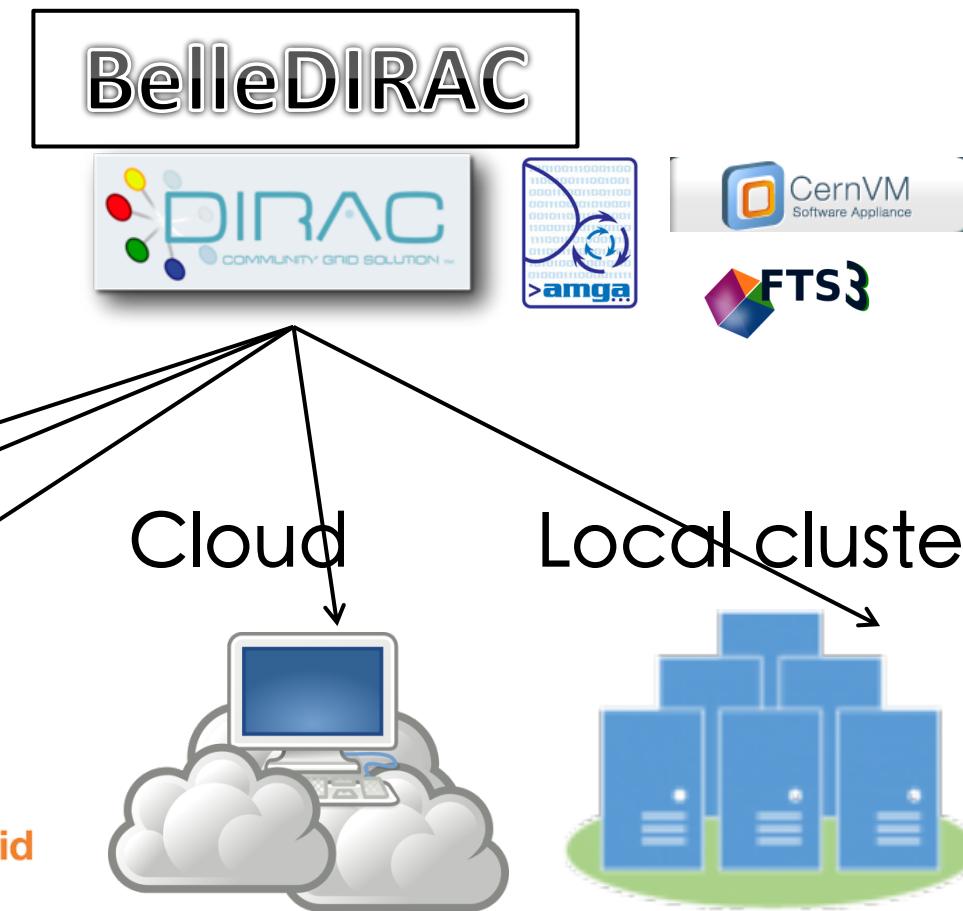
- “DIRAC” (Distributed Infrastructure with Remote Agent Control)
- A software framework for distributed computing providing a complete solution to one (or more) user community requiring access to distributed resources.
- Developed by LHCb experiment initially.
Then, CMS, ATLAS, and Belle II joined.
- Source code @GITHUB: (PYTHON codes)
<https://github.com/DIRACGrid/DIRAC>





Belle II DIRAC system

BelleDIRAC: system developed by Belle II computing team as an extension of DIRAC.

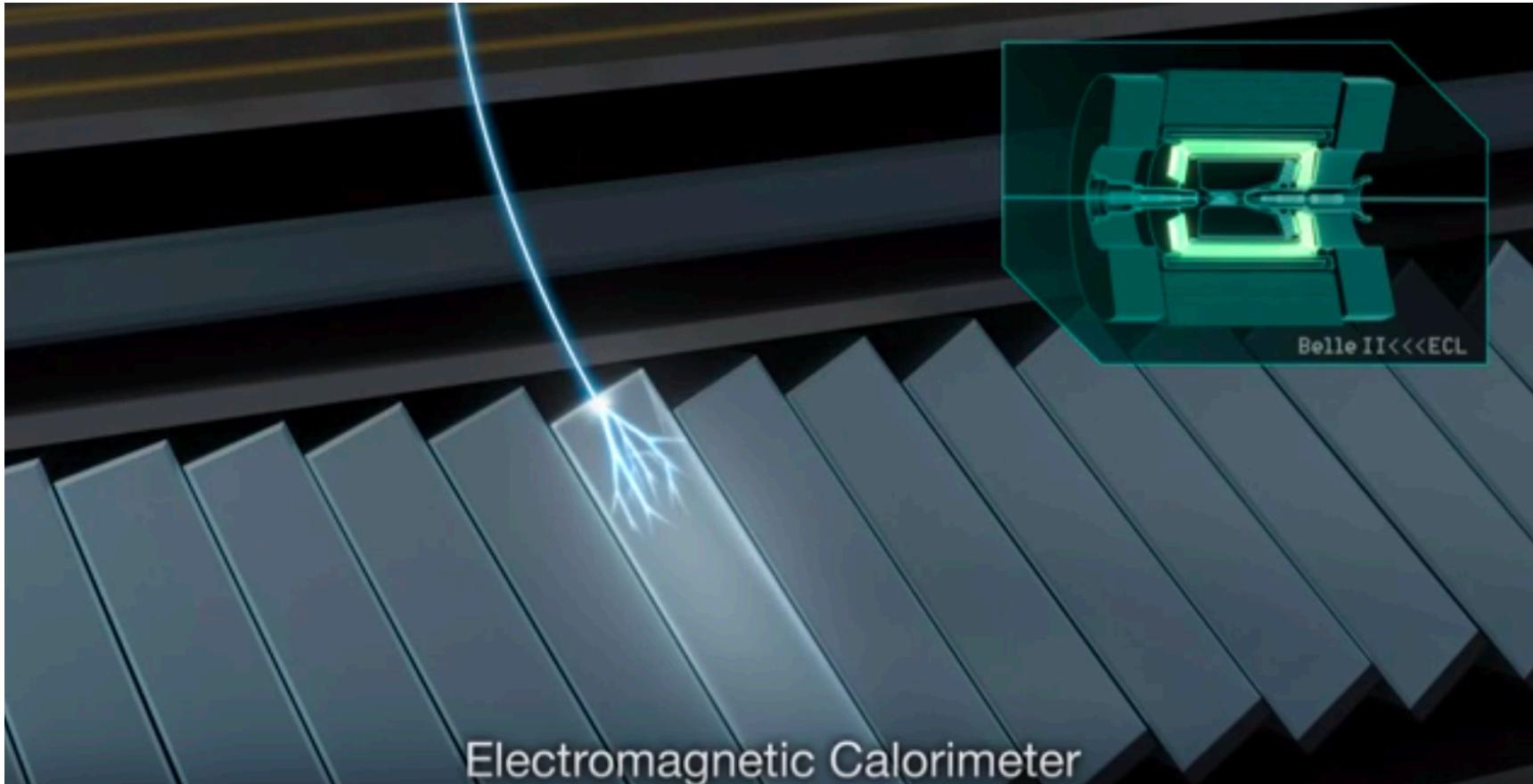




An Example of Using Convolutional Neural Network to Correct Gamma-ray Energy

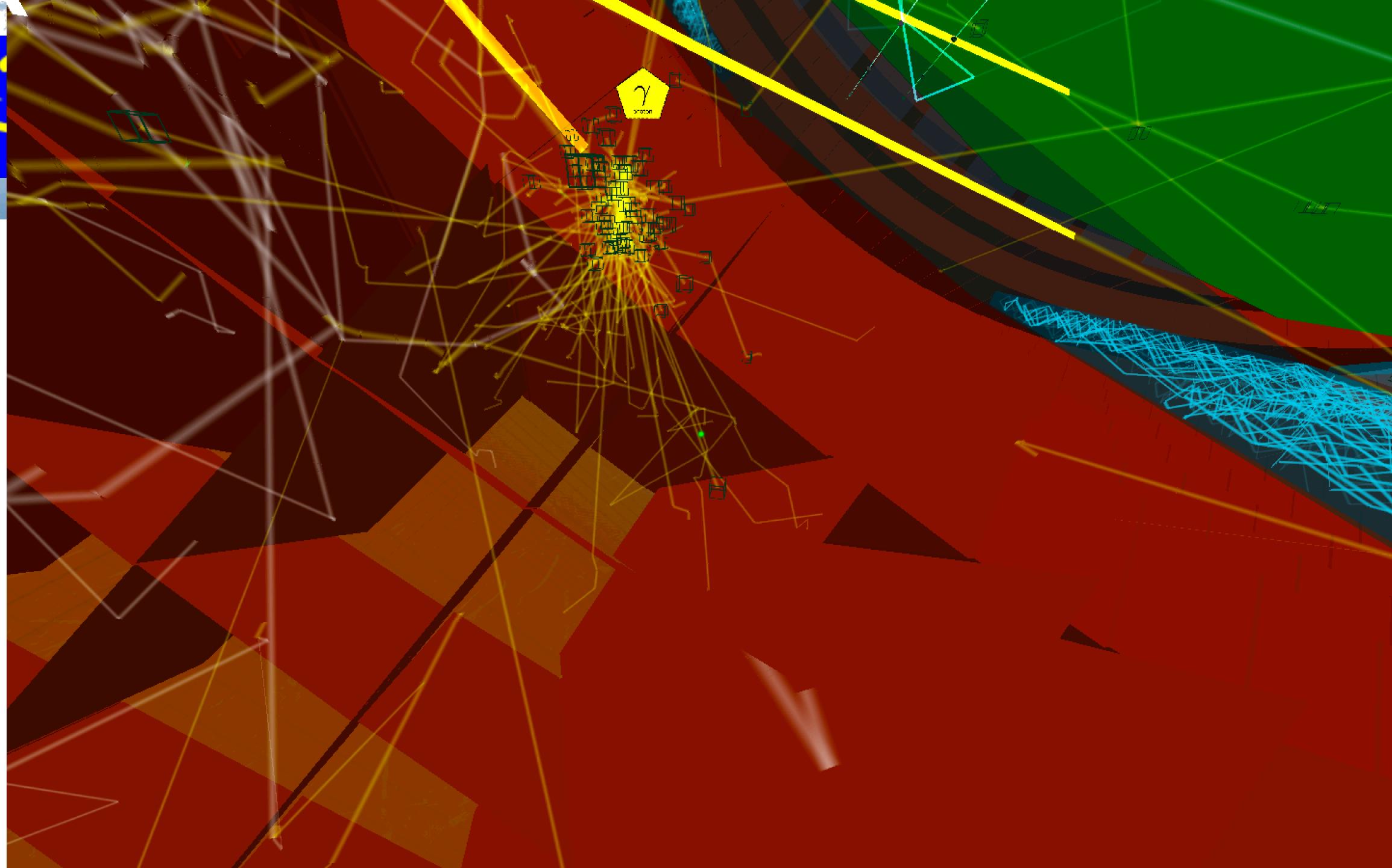


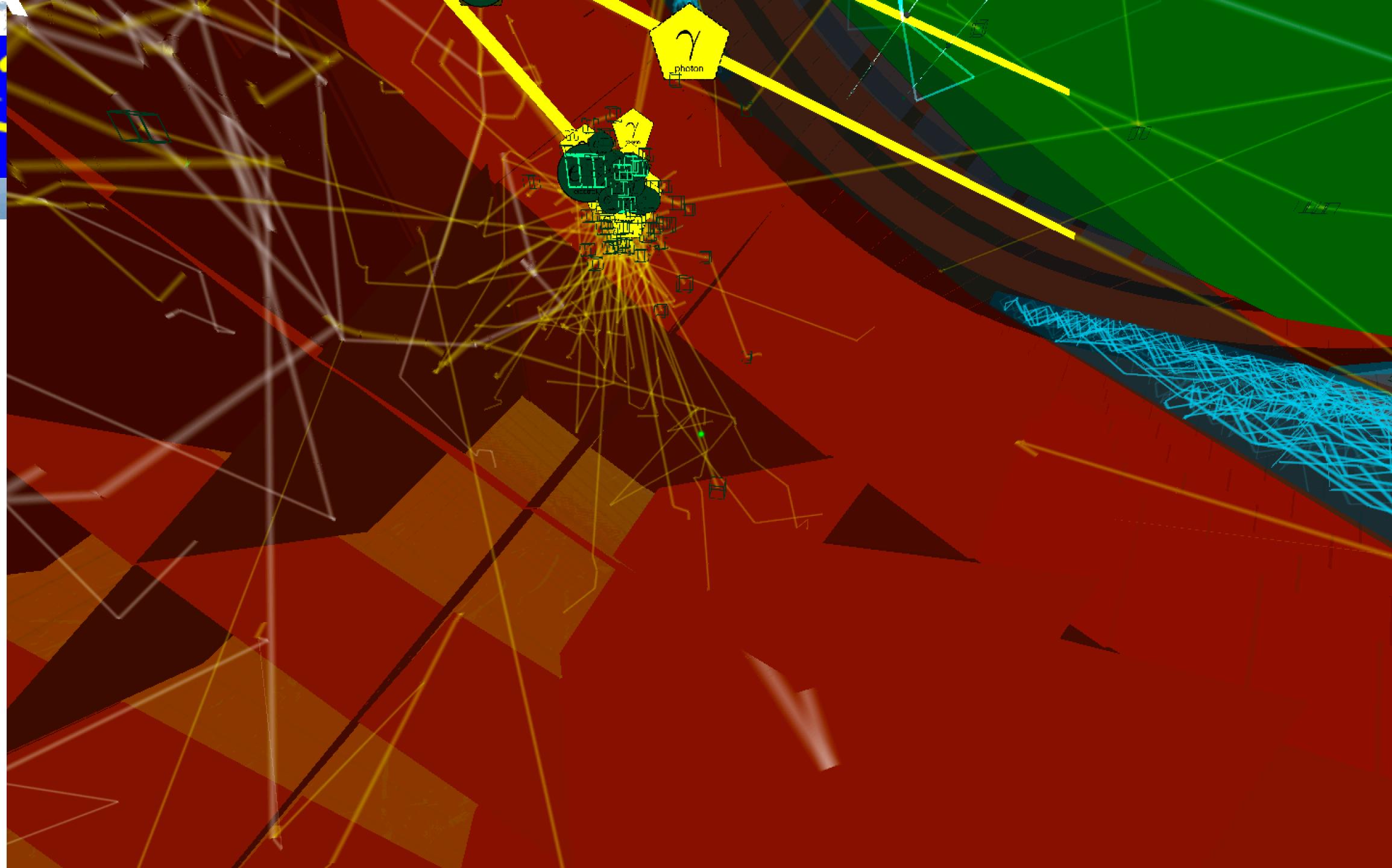
Gamma-ray energy measurement in Crystal array detector

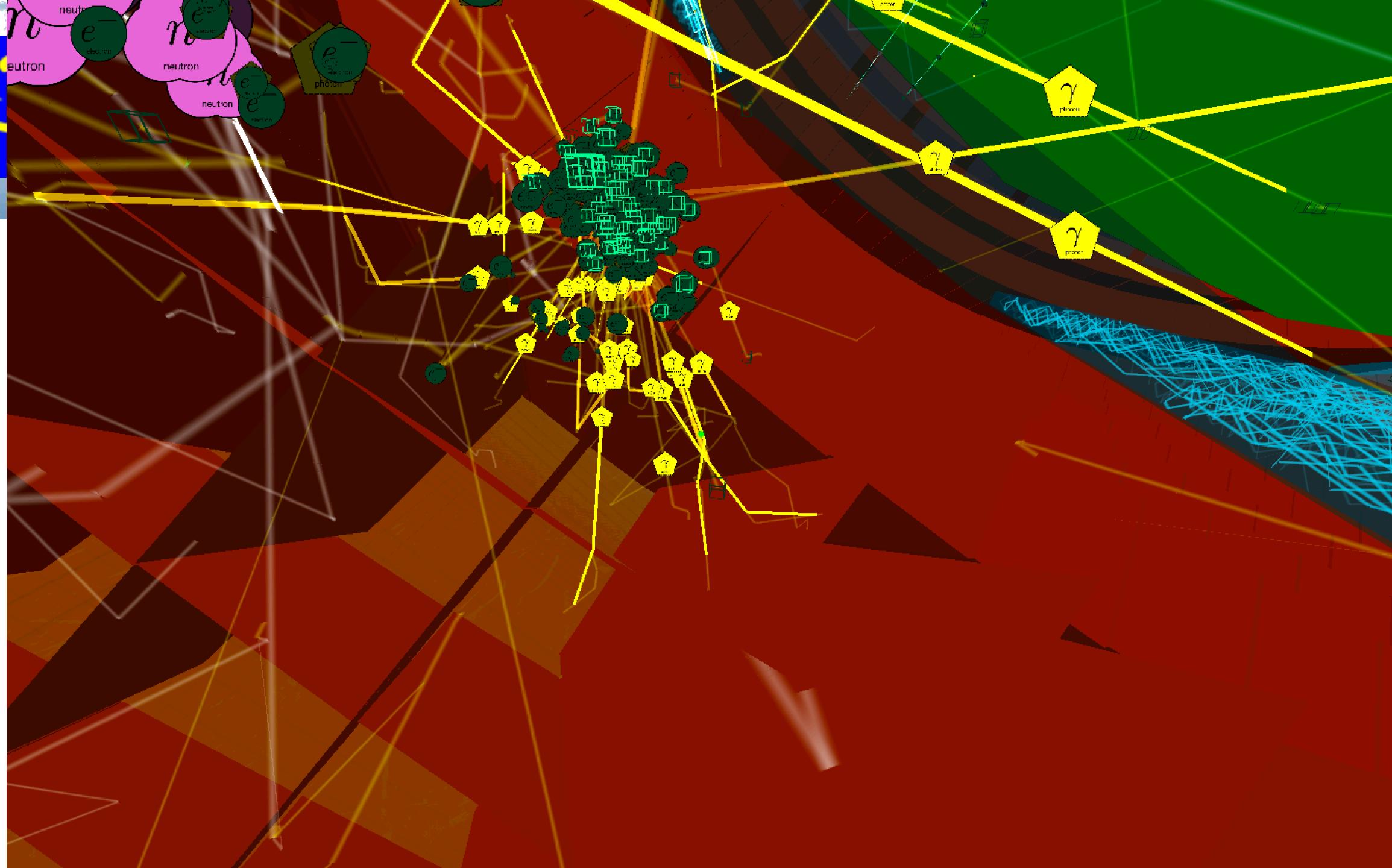


From KEK youtube channel

<https://www.youtube.com/watch?v=nGCrgXSEOlk&t=1s>





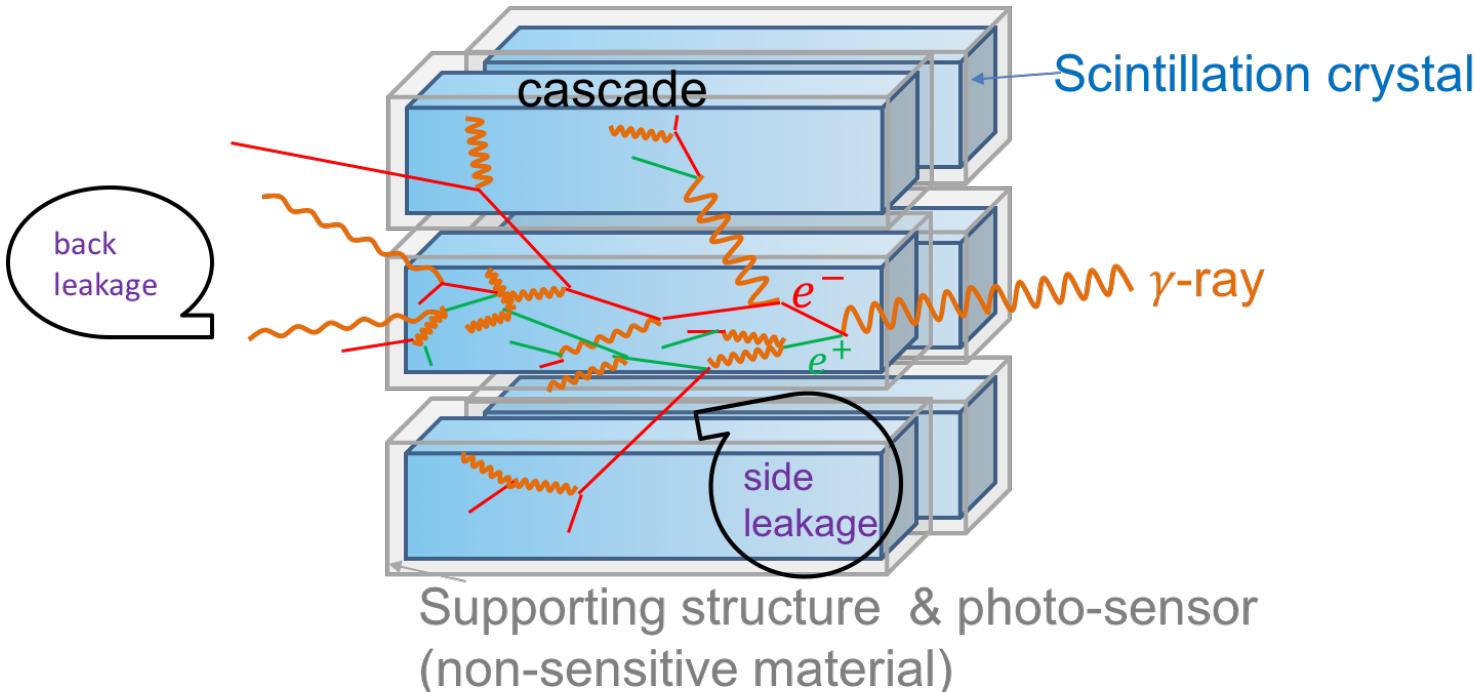




From Belle II VR

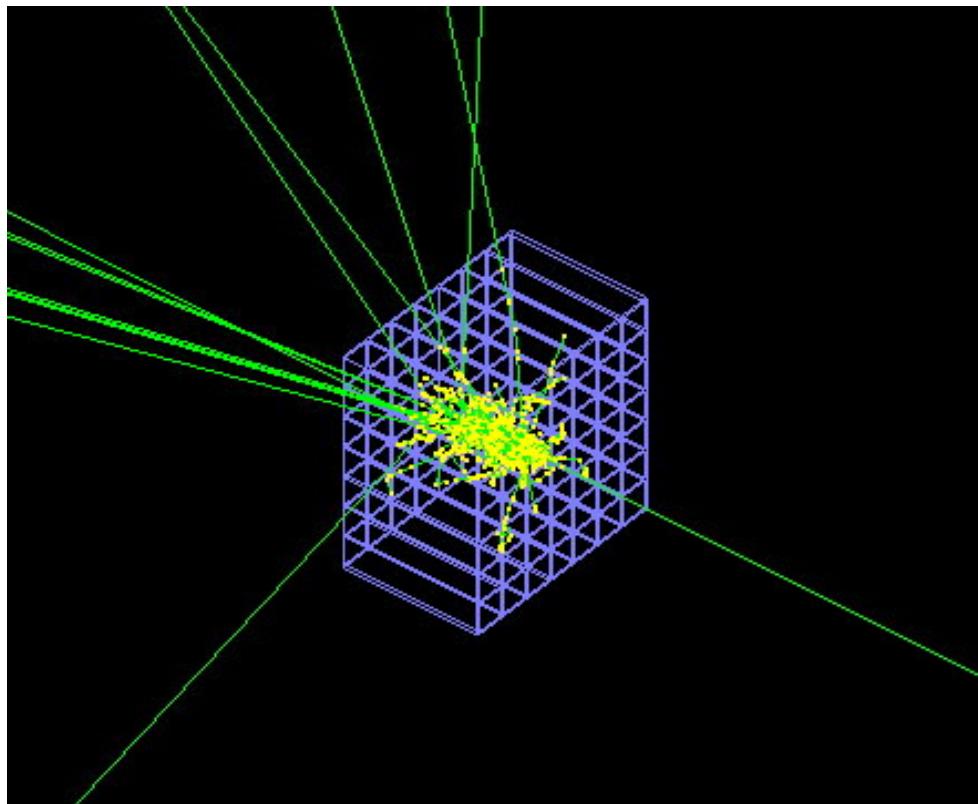
https://store.steampowered.com/app/810020/Belle_ll_in_Virtual_Reality/

Gamma-ray energy leakage @crystal array



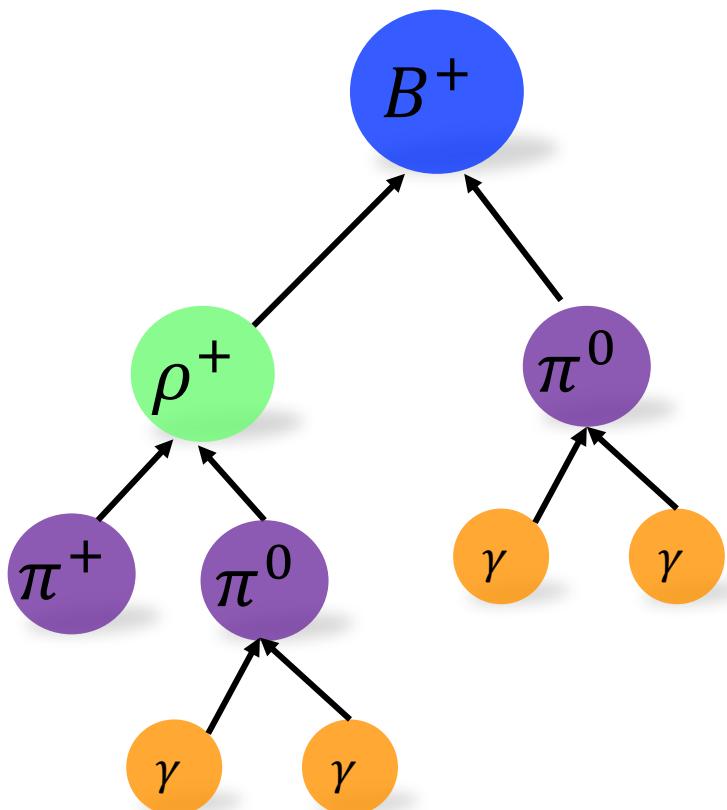
- Gamma-ray measurement at scintillation crystal:
 γ -ray with energy **0** (>0.1 GeV) transforms to charged electrons (positions). The charged particles moving inside crystal will cause light emission. Then, we use photon-sensor to measure light
→ obtain energy of gamma-ray.

Gamma-ray energy leakage @crystal array

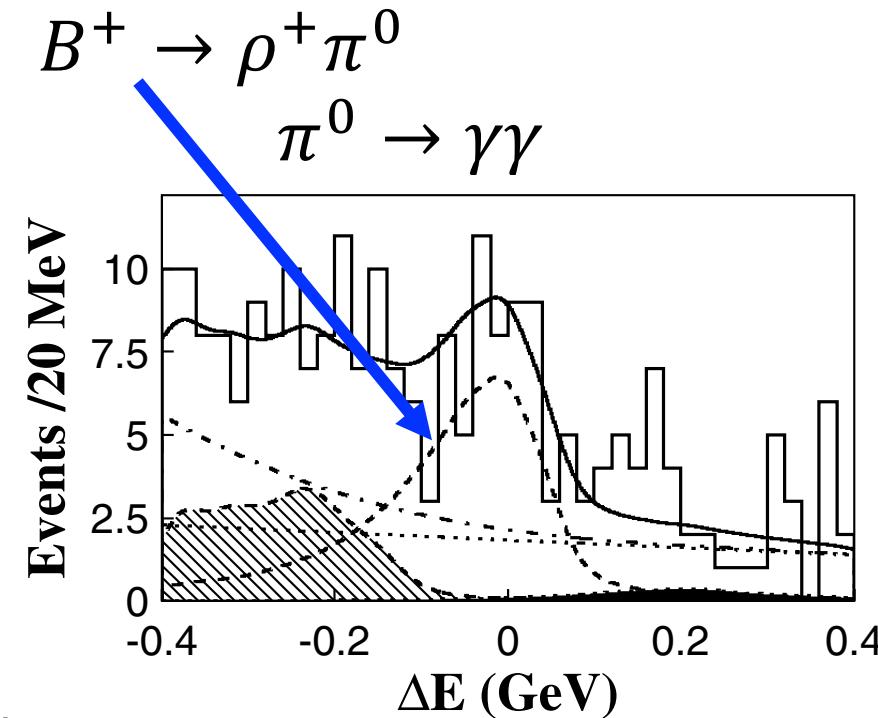


A simulated γ -ray event:
side leakage and back leakage can be seen.

Gamma-ray energy leakage



$$E_B = E_{\dots} + E_\gamma + E_\gamma + E_\gamma + E_\gamma$$



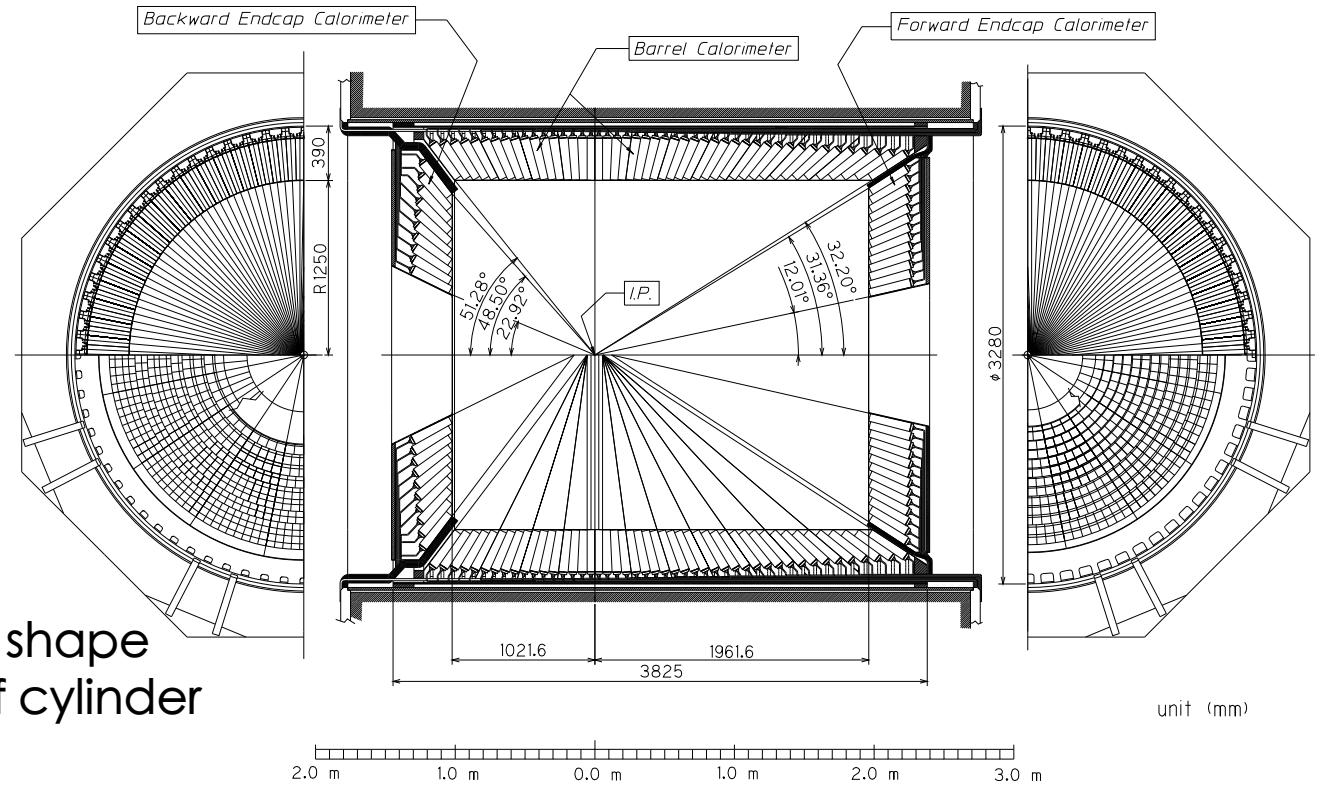
From PRL 94, 031801 (2005)

Problem:
 Signal (dash line) tail overlap with other backgrounds
 →Backgrounds also have uncertainty
 →Affect final measurement uncertainty



ECL detector structure

BELLE CsI ELECTROMAGNETIC CALORIMETER



- ~8,000 CsI crystals
- ~100 kinds of crystal shape
- Equipped outside of cylinder

Backward Endcap

Barrel

Forward Endcap

CsI crystal wiki:
https://en.wikipedia.org/wiki/Caesium_iodide

From Ikeda, Hitomi
PhD disserlation,
“Development of the CsI(Tl)
Calorimeter for the
measurement of CP Violation
at KEK B-factory “



CsI Crystal 30X30 Array model

30 by 30 CsI crystal array geometry in Geant4 simulation

- Simulate^[1] γ -ray incident into CsI crystal array.
- Each crystal has below geometry:

55 mm X 55 mm (front face)

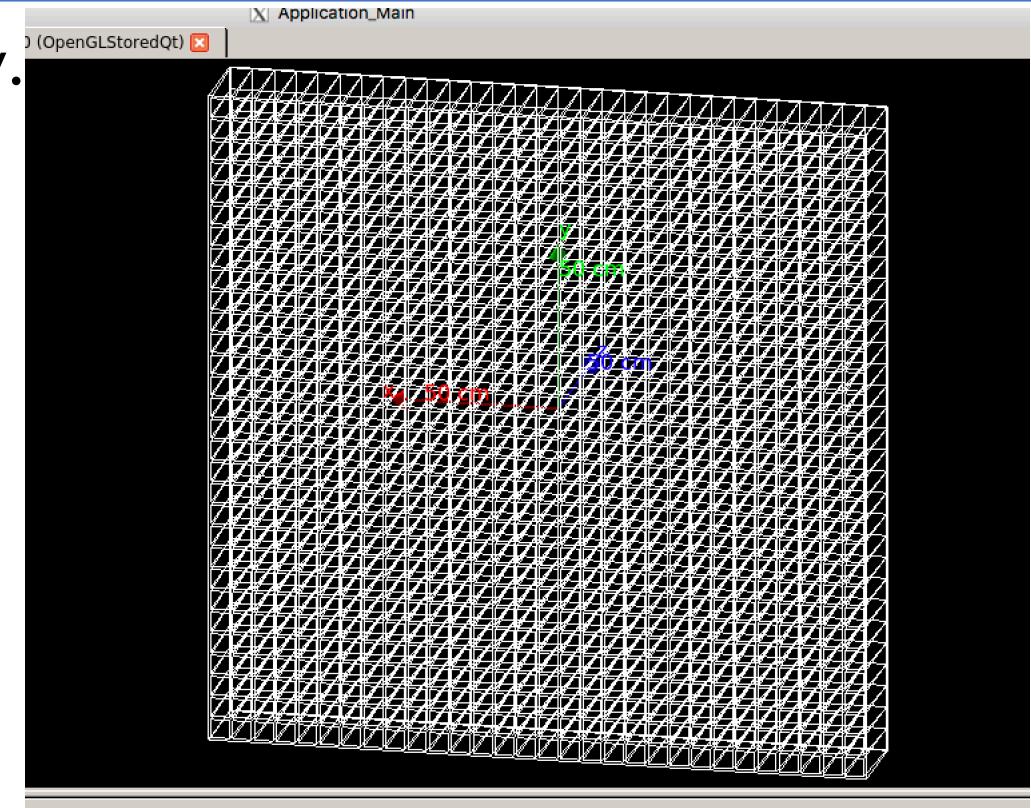
65 mm X 65 mm (rear face)

length: 30 cm

The 0.5 cm gap is filled with aluminum.

Measured γ -ray energy

$$E_{Total} = \sum_{i=1}^{900} E_{i,deposited}$$



[1]GEANT 4

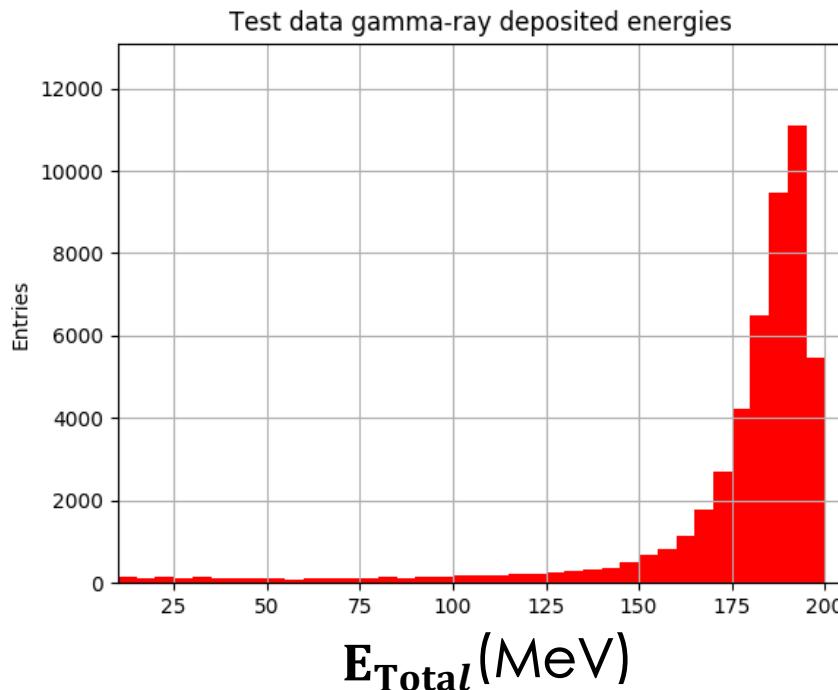
C++ API for particle and radioactive source simulation

<https://geant4.web.cern.ch/>

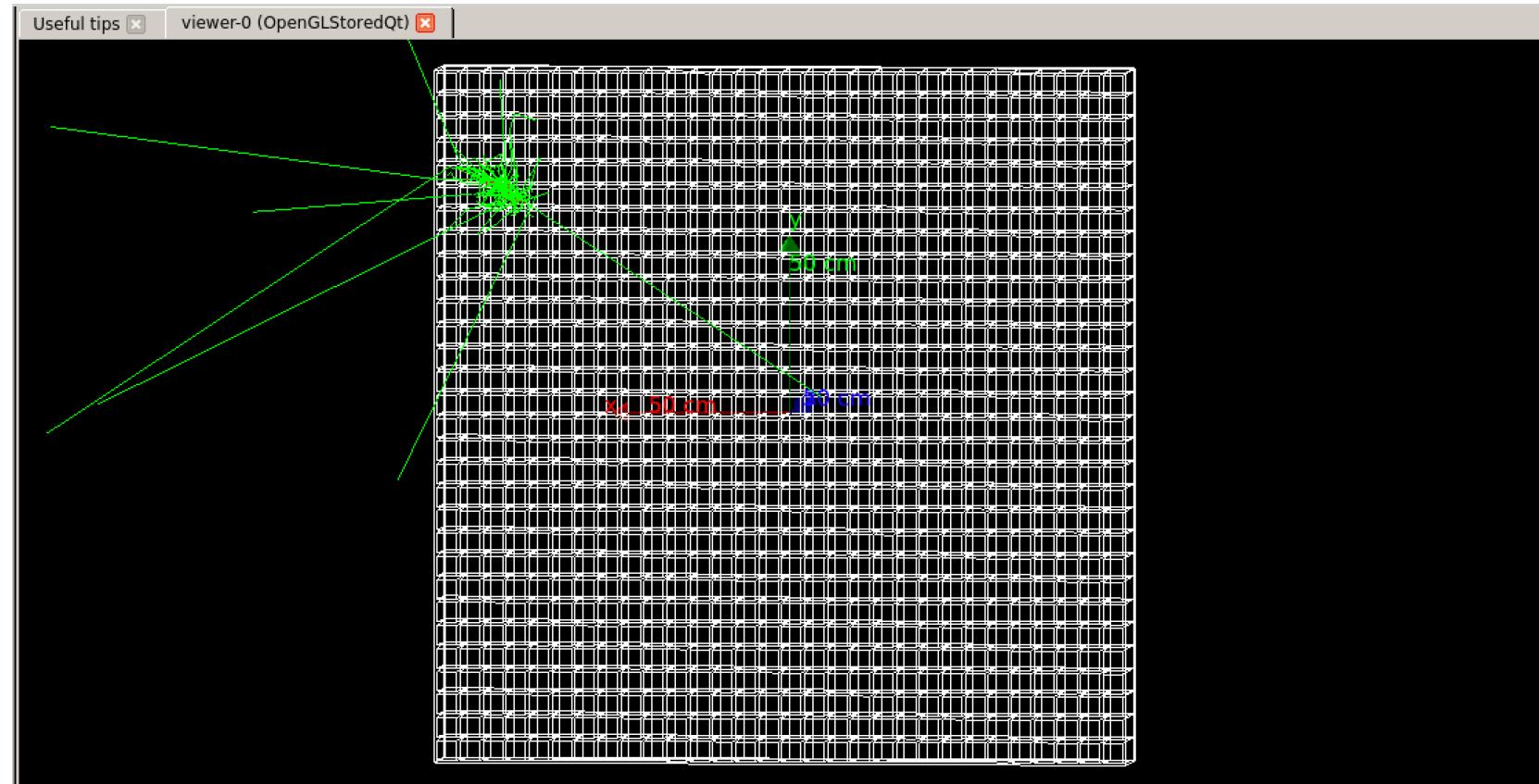


CsI Crystal 30X30 Array model

Incident of 200 MeV, 10,000 times



Measured γ -ray energies (E_{Total}) may not be close to 200 MeV.

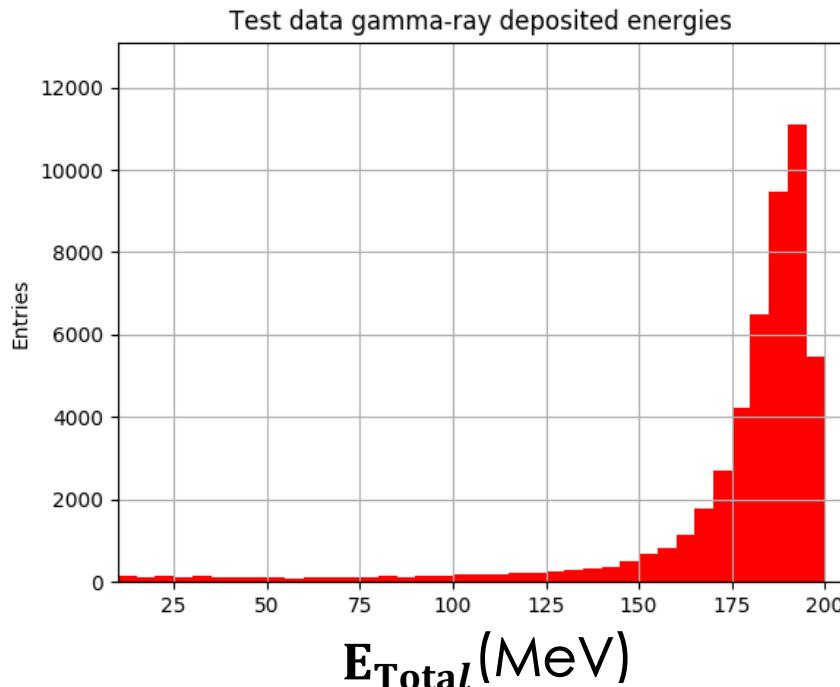


A simulated γ -ray event of 30X30 CsI array:
energy leakage can be seen.



Regression problem

Incident of 200 MeV, 10,000 times

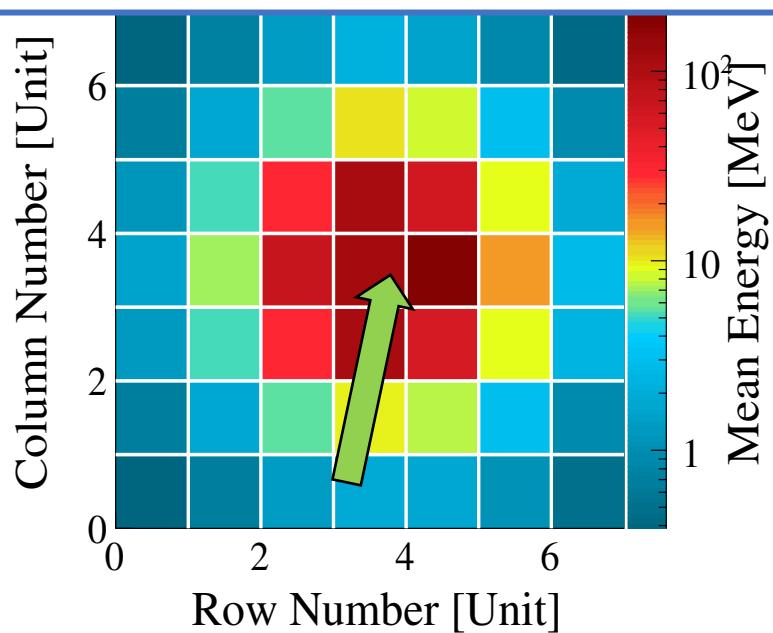


Problem:
How to make prediction of energy (E_{pre})
that close to “true” gamma-ray
energy (E_{True}).
→ Regression problem

Measured γ -ray energies (E_{Total})
may not be close to 200 MeV.

Energy pattern on CsI crystal array

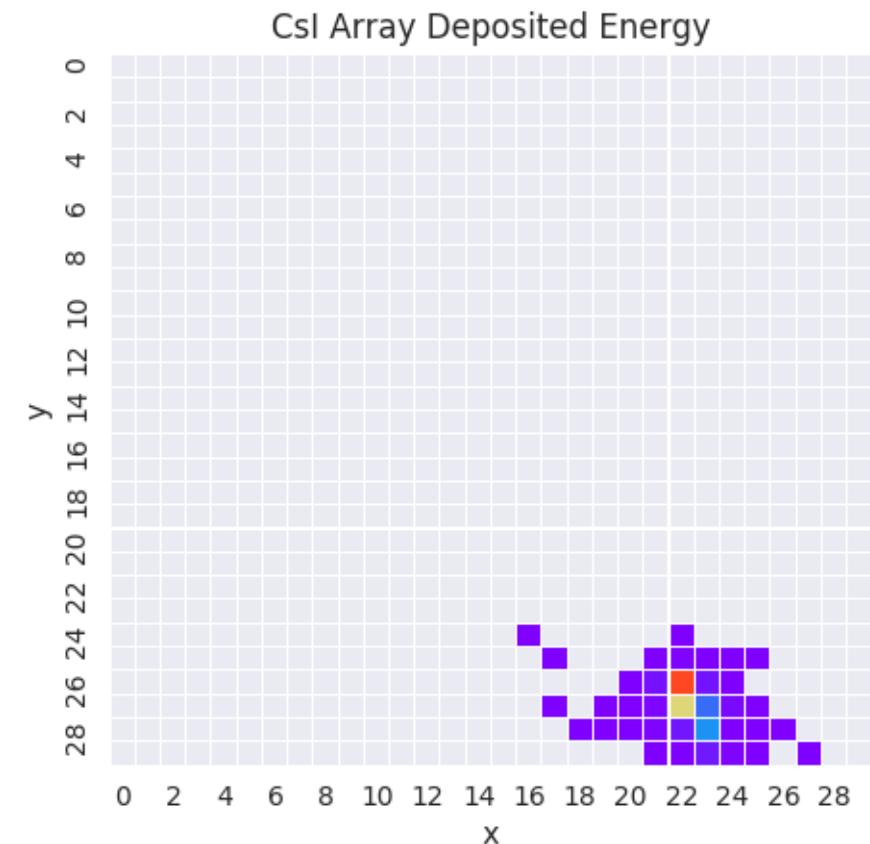
Photon of energy 1 GeV hits
2/3 towards the right
of the central crystal. 1000
simulated events



- The deposited energy pattern is supposed to be related to incident angle and position. Also, the energy leakages relate to energy pattern.
- As the crystals are homogeneous, energy pattern can be the same over the crystal array.
- Regression problem.
- Solution: **Convolutional Neural Network (CNN)**
- Introduction of CNN webpage:
<http://cs231n.github.io/convolutional-networks/>
(page of Cs231n class Standford Univ.)



Input Array, Predict Energy, Target



Input Data:
Deposited Energy of CsI Array

Target: (y_{True})

Ratio of Total deposited energy/True energy:

$$y_{True} = (E_{Total}/E_{True})$$

Prediction: ($y_{Predict}$)

Predict energy by CNN:

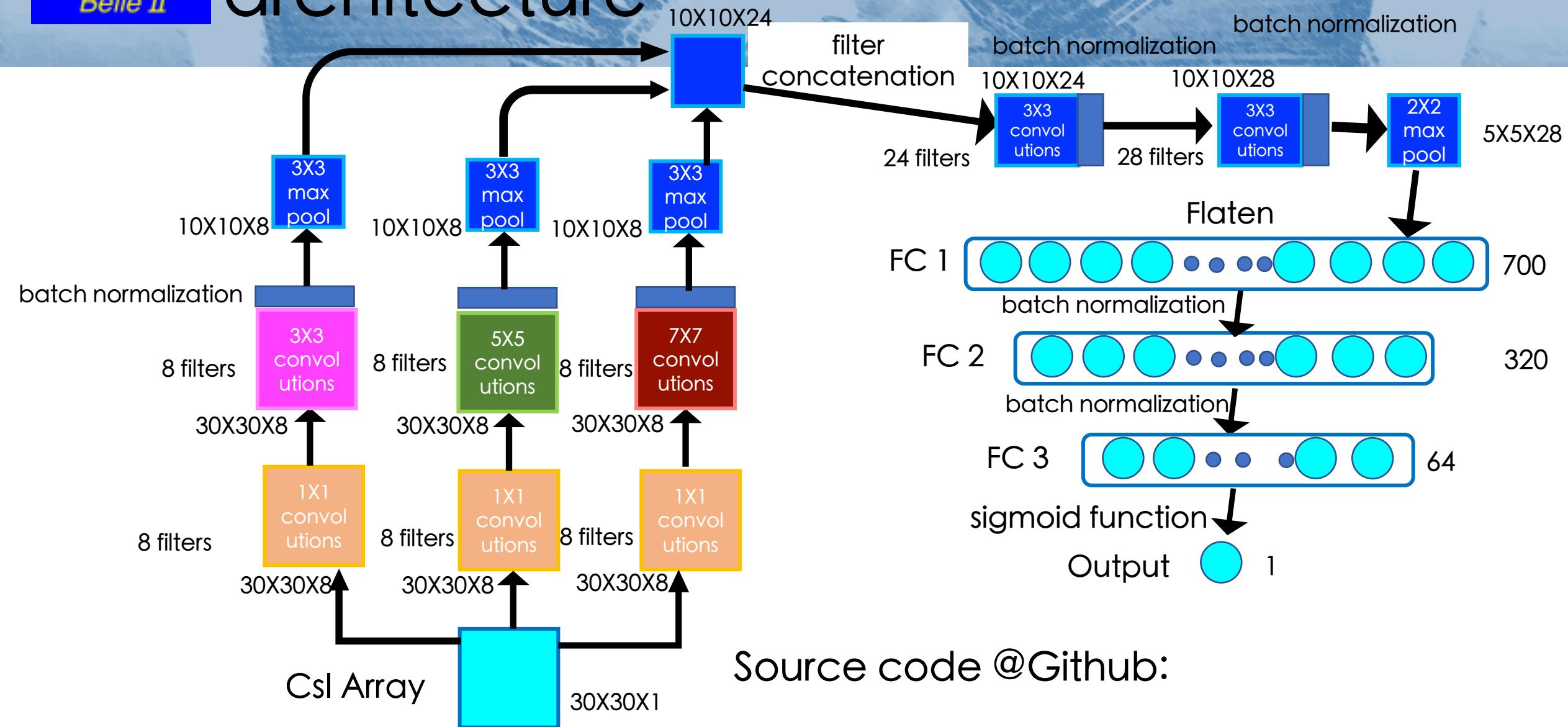
$$E_{CNN} = E_{Total}/y_{Predict}$$

Loss function:

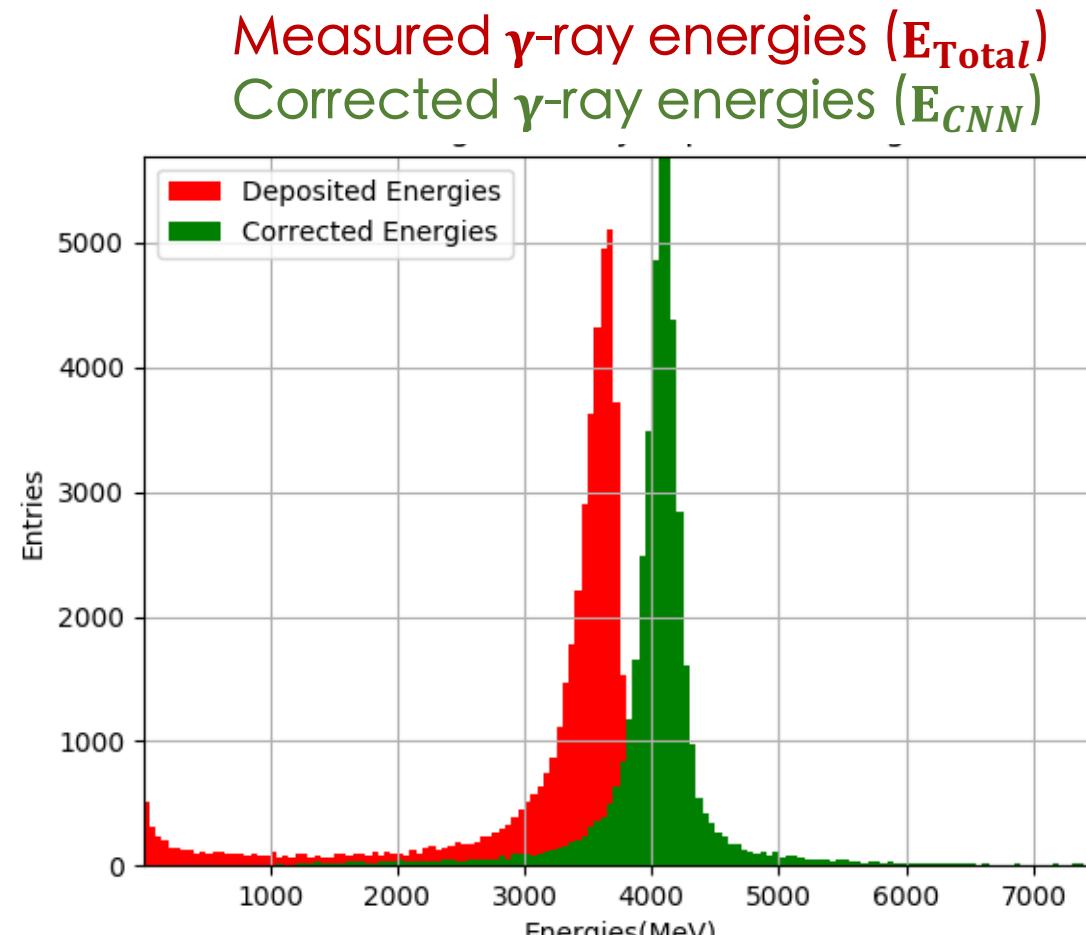
$$\text{MSE of } (y_{Predict} - y_{True})$$



Convolutional Neural Network architecture

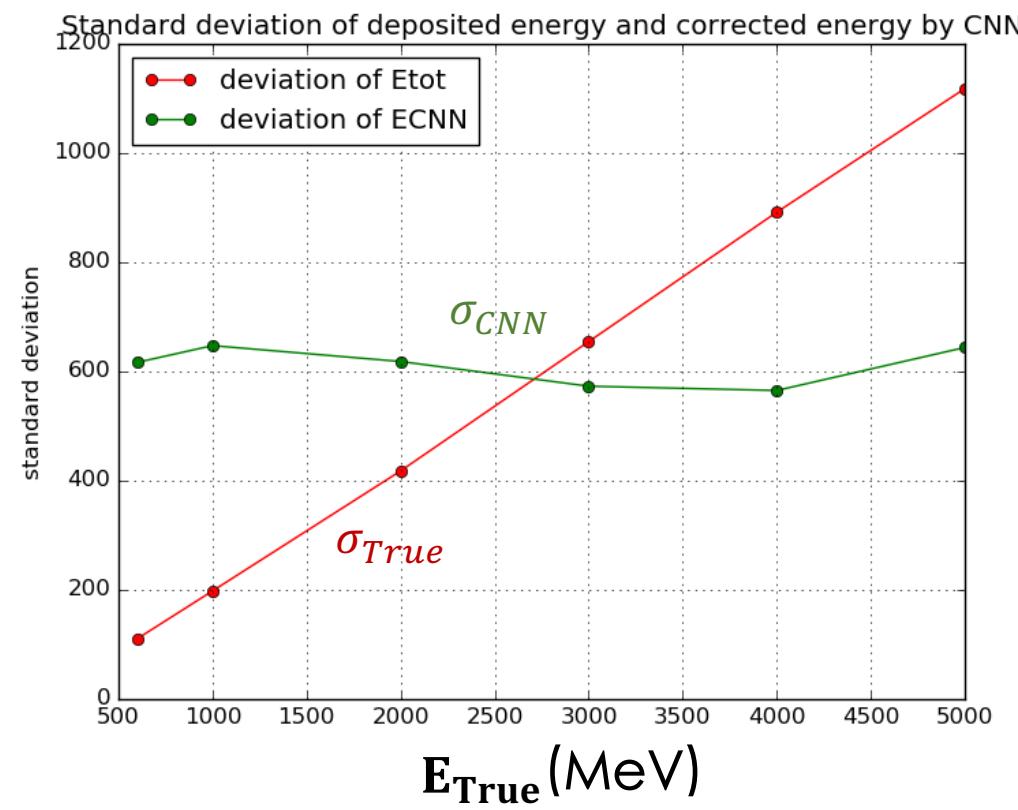
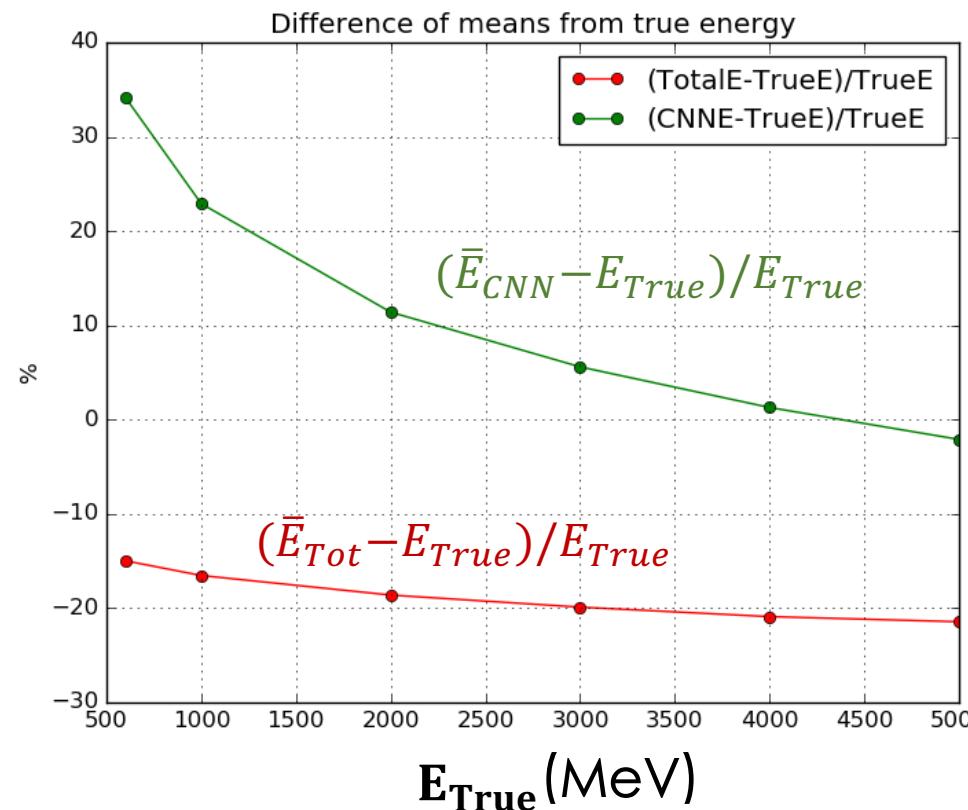


Test data result of mono-energy events



Incident of 4000 MeV γ -ray, 10,000 times

Test data result of mono-energy events



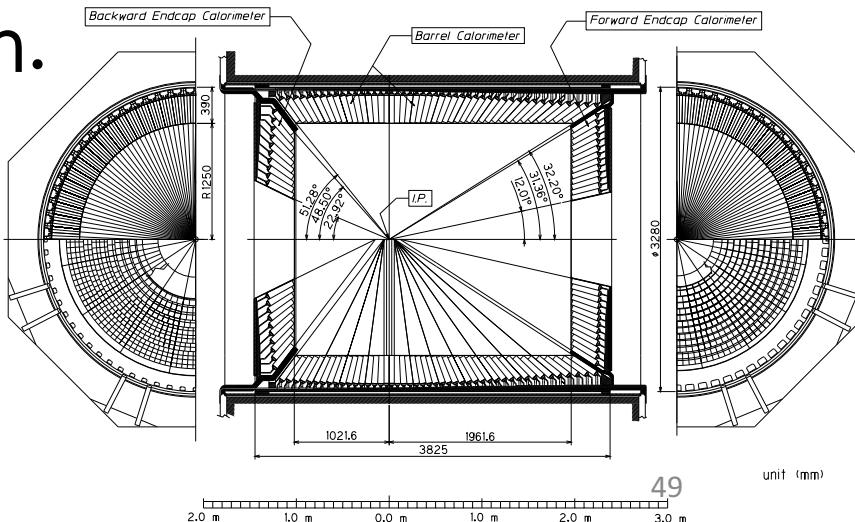
Correction of CNN method performance at high energy (>2000MeV) is good.
 However, many gamma-ray having energy less than 2GeV in experiment
 →Further improvement is needed.



Simple CNN may be not good enough.....

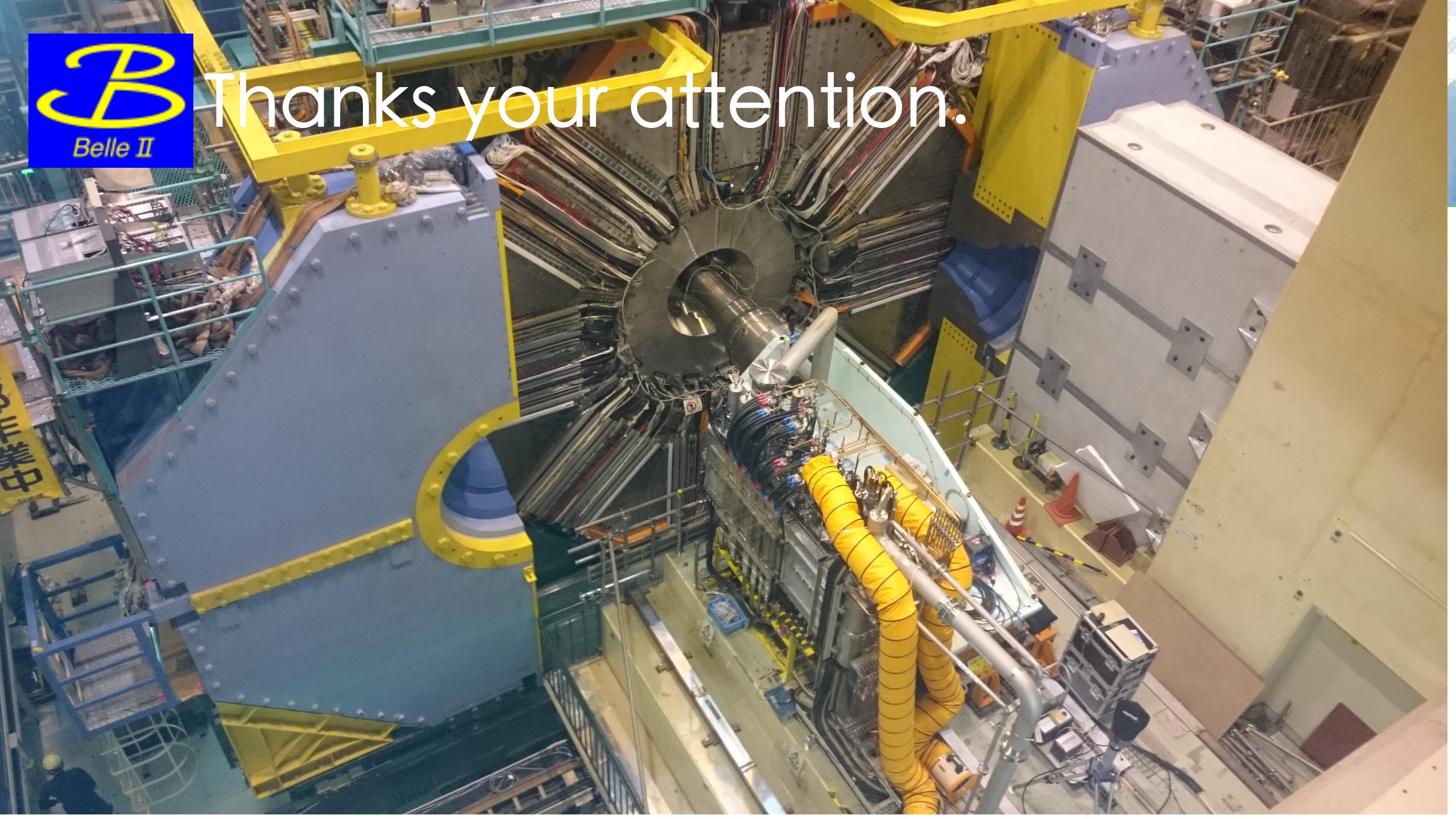
- Proposed improvement:
 - group events by their energy pattern
-->train multiple CNN models
 - target function: $(E_{\text{Total}}/E_{\text{True}}) \rightarrow (E_{\text{total}}/E_{\text{peak}})$
 - Others...
- Using CNN model for real “Belle II” problem.
 - Heterogeneous CsI crystal
 - Crystal arrangement: cylinder geometry
 - Background gamma-ray and electronics noise

BELLE CsI ELECTROMAGNETIC CALORIMETER





Thanks your attention.





Thanks your attention.

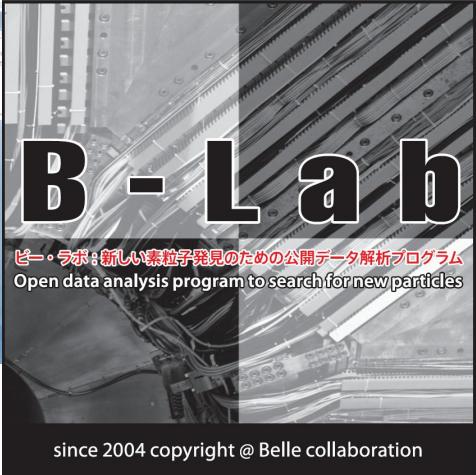


Thanks your attention.

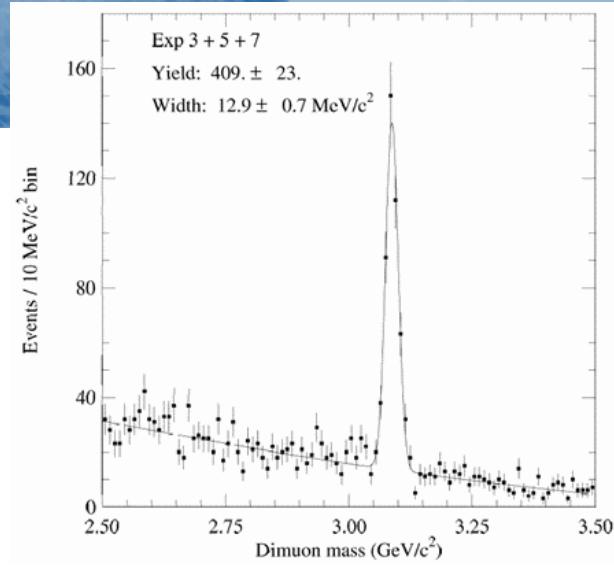


References

- Belle II experiment official website: <https://www.belle2.org/>
- Belle II facebook page: <https://zh-tw.facebook.com/belle2collab/>
- KEK Youtube channel:
<https://www.youtube.com/channel/UCMvGyDHrm820nYLUhwtm2DQ>
- niconico生放送 Belle II first collision:
<http://live.nicovideo.jp/watch/lv312661713#49:24:30>



B-Lab: open data project of Belle



- Open data of “Belle” experiment.
(only 0.1% of full data, $\sim 1\text{fB}^{-1}$)
- B-Lab program is a project to open part of the Belle experimental data mainly to high school science clubs via the Internet and to give them a chance to "search for new particles."
- website: <http://belle.kek.jp/b-lab/b-lab-english/>
- Data in format of ROOT files.
Can be processed by PyROOT, root_numpy, and root_pandas.

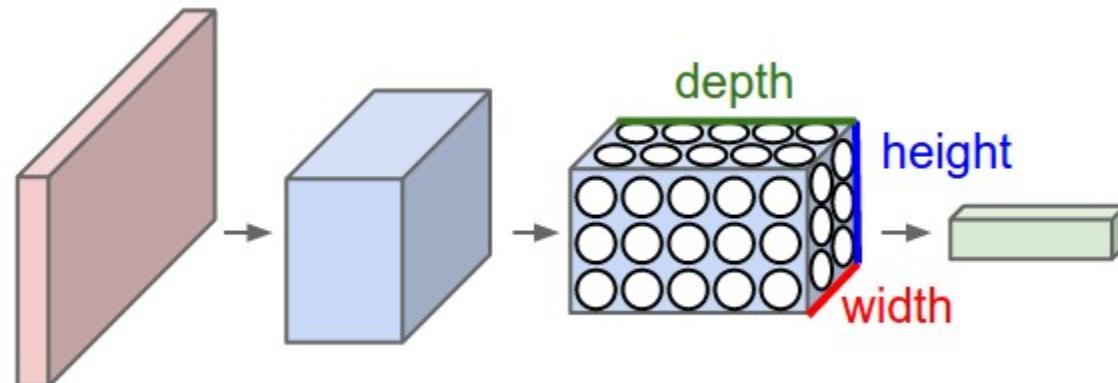


Backup



Convolutional Neural Network

- “convolution” layer



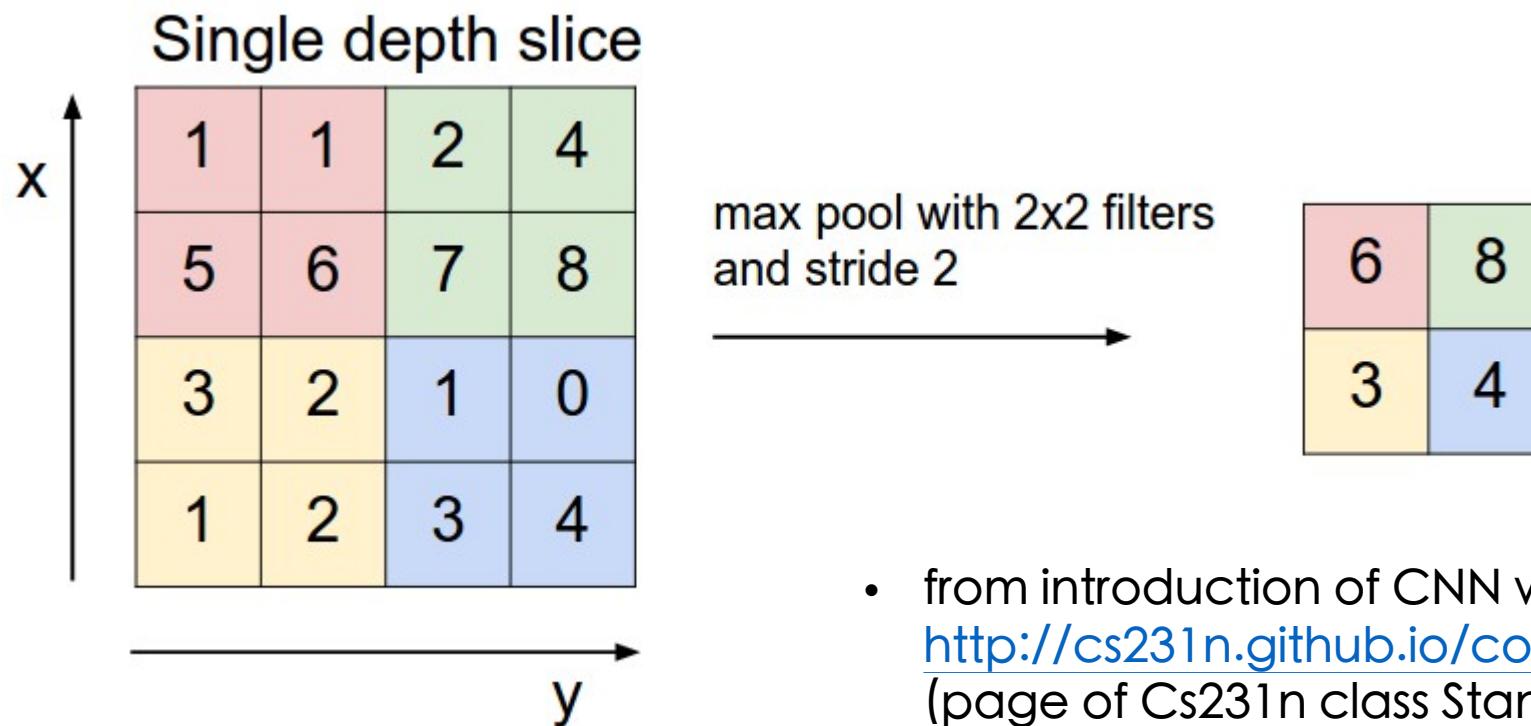
Input Volume (+pad 1) (7x7x3)	Filter W_0 (3x3x3)	Filter W_1 (3x3x3)	Output Volume (3x3x2)
$x[:, :, 0]$	$w_0[:, :, 0]$	$w_1[:, :, 0]$	$o[:, :, 0]$
0 0 0 0 0 0 0 0 0 0 2 1 2 0 0 1 2 0 2 2 0 0 0 2 1 2 0 0 0 1 0 0 1 0 0 0 2 0 0 2 1 0 0 0 0 0 0 0 0	1 0 1 -1 -1 -1 1 -1 0 0 1 1 0 0 -1 1 0 0 -1 -1 0	1 -1 0 -1 1 1 -1 0 0 1 0 0 1 1 0 0 0 0 -1 0 -1	-2 -2 0 -1 -2 6 -2 -2 -3 4 4 2 3 9 -1 2 4 5
$x[:, :, 1]$	$w_0[:, :, 1]$	$w_1[:, :, 1]$	$o[:, :, 1]$
0 0 0 0 0 0 0 0 1 0 2 2 0 0 0 0 1 2 0 2 0 0 0 1 2 0 1 0 0 2 0 0 2 1 0 0 0 0 0 0 0 0	0 1 1 0 0 -1 1 0 0 -1 -1 0 1 -1 -1 0 1 0	1 0 0 1 1 0 0 0 0 -1 0 -1 0 1 1 -1 0 1	4 4 2 3 9 -1 2 4 5
$x[:, :, 2]$	$w_0[:, :, 2]$	$w_1[:, :, 2]$	$o[:, :, 2]$
0 0 0 0 0 0 0 0 2 0 1 1 1 0 0 0 1 1 0 2 0 0 1 1 1 2 2 0 0 1 1 0 1 2 0 0 0 0 0 2 2 0 0 0 0 0 0 0 0	1 0 1 0 1 0 -1 0 -1 0 1 1 -1 0 1 0 0 0	0 1 1 -1 0 -1 <br;-1 0="" 1<br=""></br;-1> 0 0 0 0 1 1 -1 0 1	-1 0 -1 0 1 1 -1 0 1 0 0 0 0 1 1 -1 0 1
	Bias b_0 (1x1x1)	Bias b_1 (1x1x1)	
	$b_0[:, :, 0]$	$b_1[:, :, 0]$	
	1	0	

toggle movement

- from introduction of CNN webpage:
<http://cs231n.github.io/convolutional-networks/>
(page of Cs231n class Standford Univ.)

Convolutional Neural Network

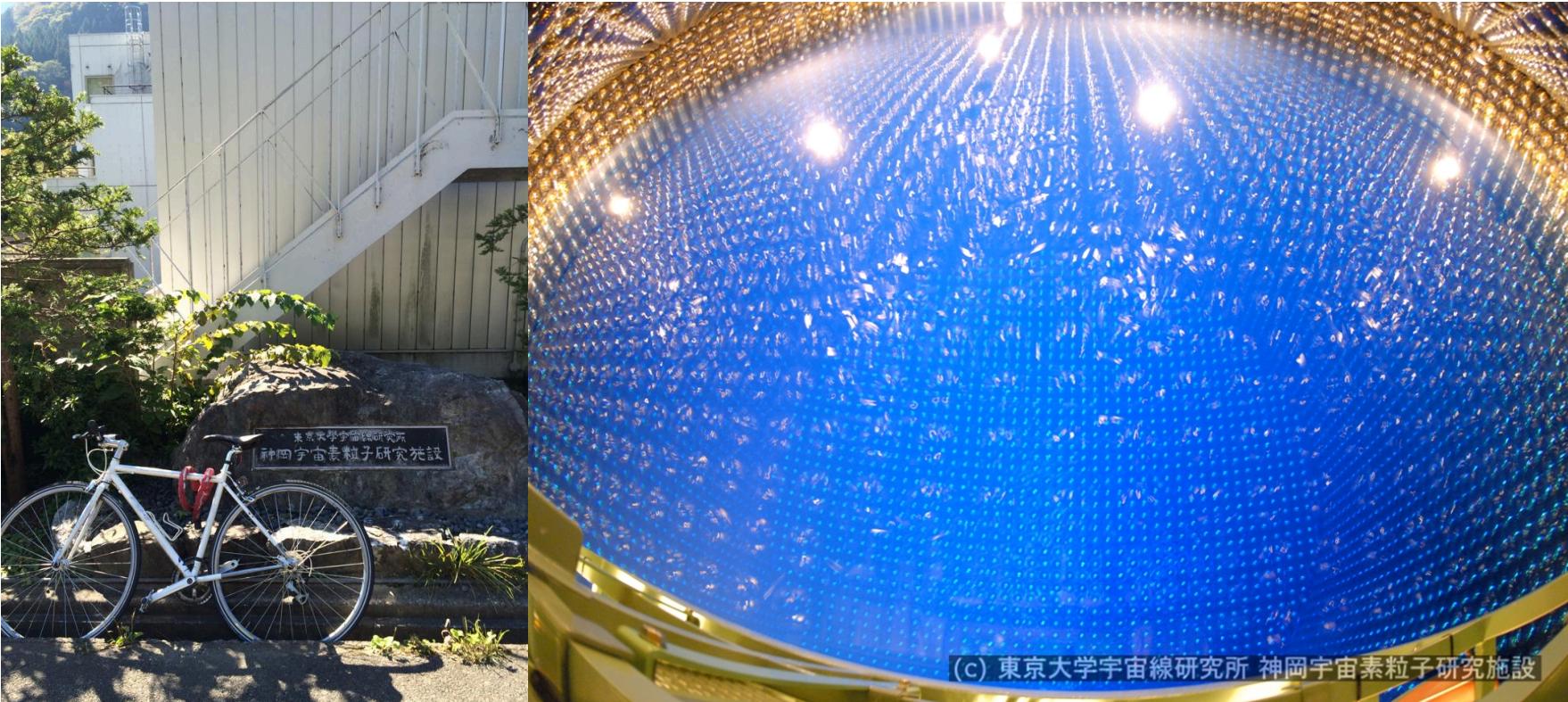
- “pooling” layer





About my research

What I did for my PhD.....

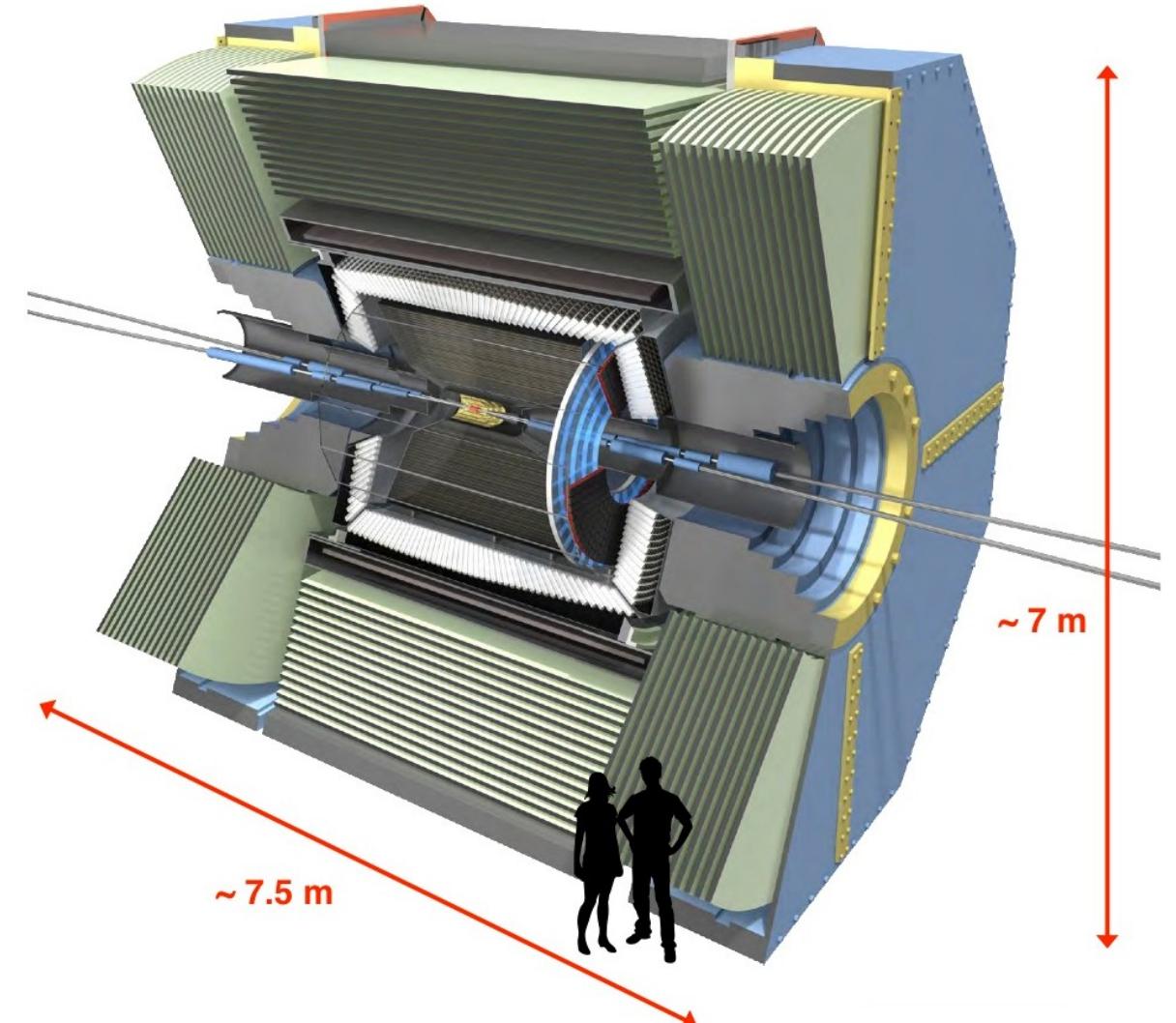


Neutrino experiment@Kamioka, Japan: T2K
Using Python to analyze Super-Kamiokande data



Belle II detector size

• “”





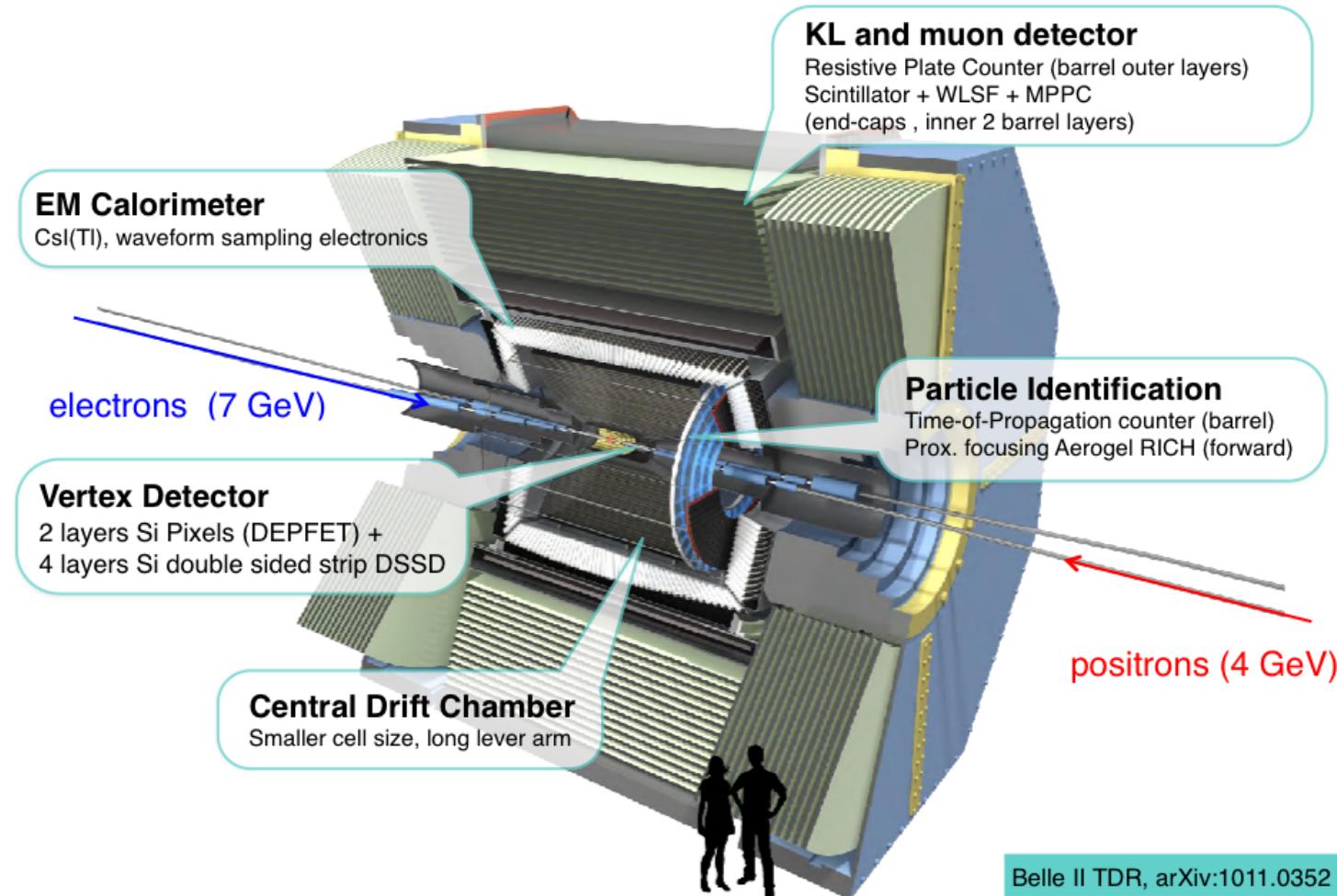
Big Data in particle physics exp.

- Large detector use complex of sub-detectors to measure a single event. Parts of sub-detectors can detect specific particles.
- Recent years, each sub-detector has $O(1,000\sim 10,000)$ pixels. So, the complex detector usually have totally $O(1,000,000)$ pixels.
→ one event has 1 million of “features” raw data.
- “Big Data” techniques are widely used in large particle experiments.



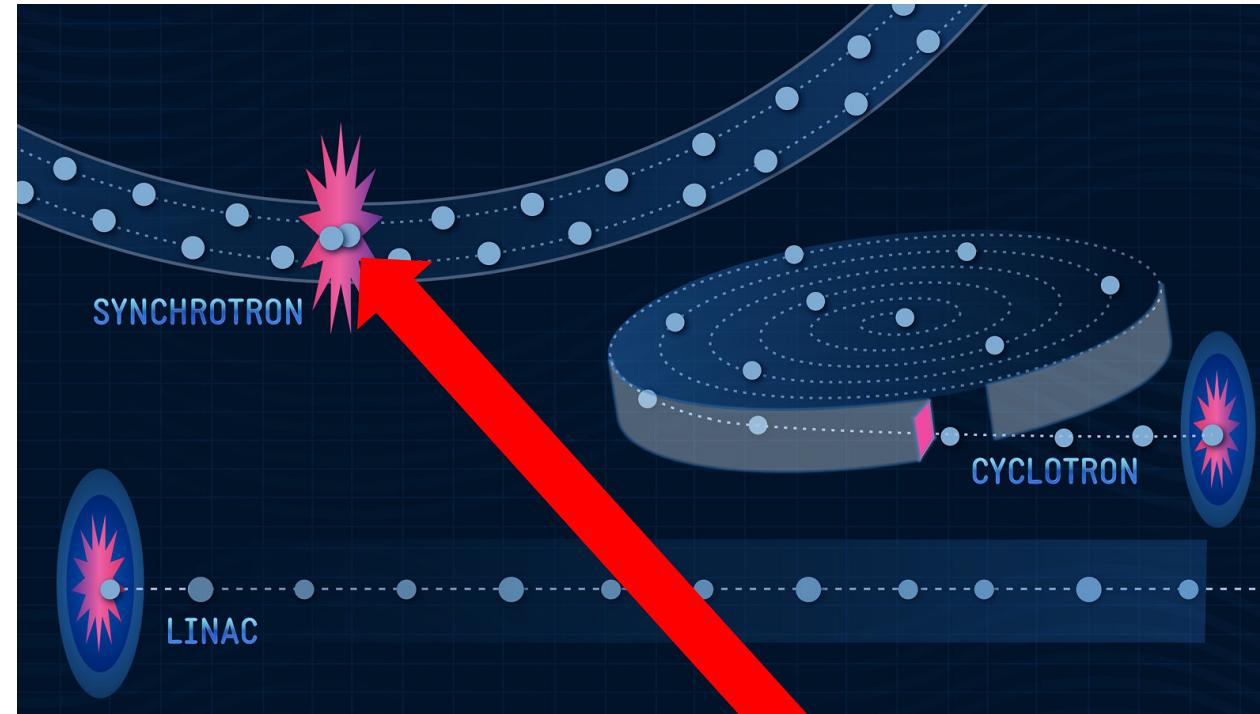
Big Data in particle experiments

- Example: Belle II detector
 - PXD: 1600*250*2 pixels
 - SVD: 187 sensors and $O(1,200\sim 1,500)$ strips/per sensor
 - CDC: 56,576 wires
 - TOP: ~8,000 electronics channels
 - ARICH: 131,760 pixels
 - ECL: 8,736 crystals
 - KLM: 17,000 photo-sensors
- Data size: 1MB
(after event builder by firmware)
- Trigger rate: max 30 kHz.
- Data rate: max 30 GB/sec





Particle physics experiment (or called High energy physics exp.)

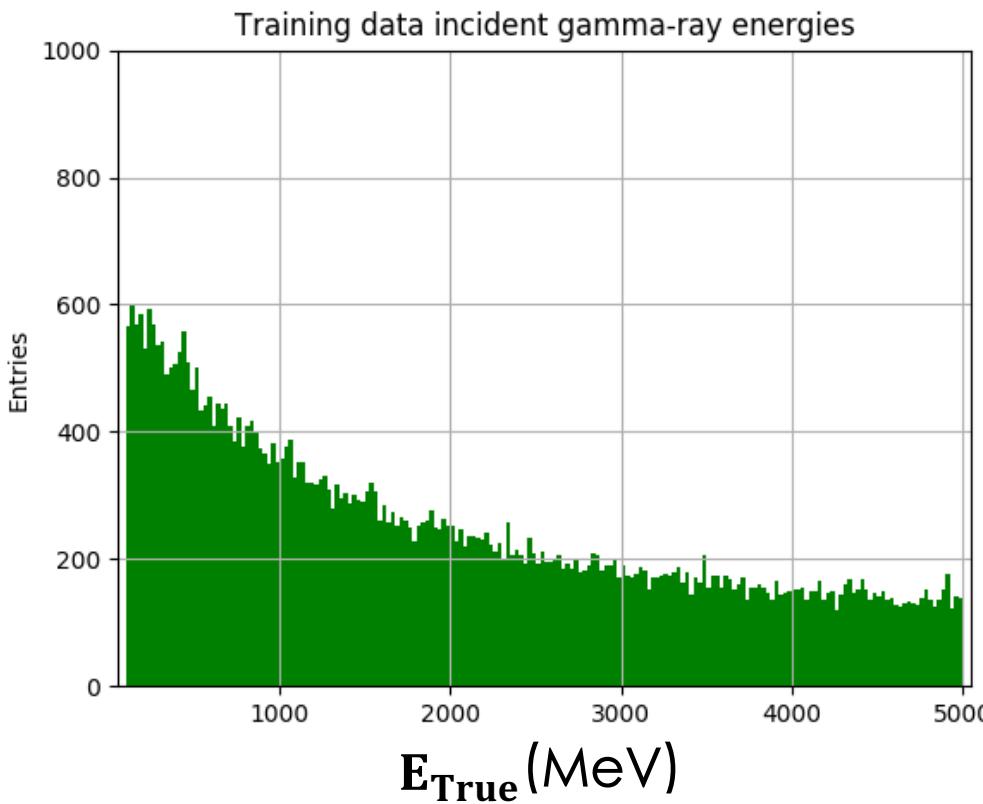


From Symmetry magazine
“A primer on particle accelerators”
07/12/16

Accelerated particles are used to produce “unfamiliar” particles in our life.



Training samples & Target



Target: (y_{True})

Ratio of Total deposited energy/True energy:

$$(E_{\text{Total}}/E_{\text{True}})$$

Another target option: ($E_{\text{total}}/E_{\text{peak}}$)

E_{peak} : peak value of total energy histogram of specific true energies



Test data result of mono-energy events

E_{real} (MeV)	600	1000	2000	3000	4000	5000
mean of E_{total} (MeV)	510.2	834.7	1627.9	2403.4	3164.4	3928.2
mean of E_{CNN} (MeV)	805.0	1229.1	2228.1	3167.9	4051.6	4895.1

E_{real} (MeV)	600	1000	2000	3000	4000	5000
RMS of E_{total} (MeV)	110.6	198.1	417.9	654.7	891.73	1118.1
RMS of E_{CNN} (MeV)	616.73	647.4	618.3	573.1	565.22	643.99

施工中



Backup

E_{real} (MeV)	500	1000	2000	2500	3000	3500	4000
RMS of E_{total} (MeV)	67.3	149.3	334.0	432.1	530.7	629.5	742.4
RMS of E_{NN} (MeV)	62.5	146.2	330.0	428.2	527.1	620.3	736.3