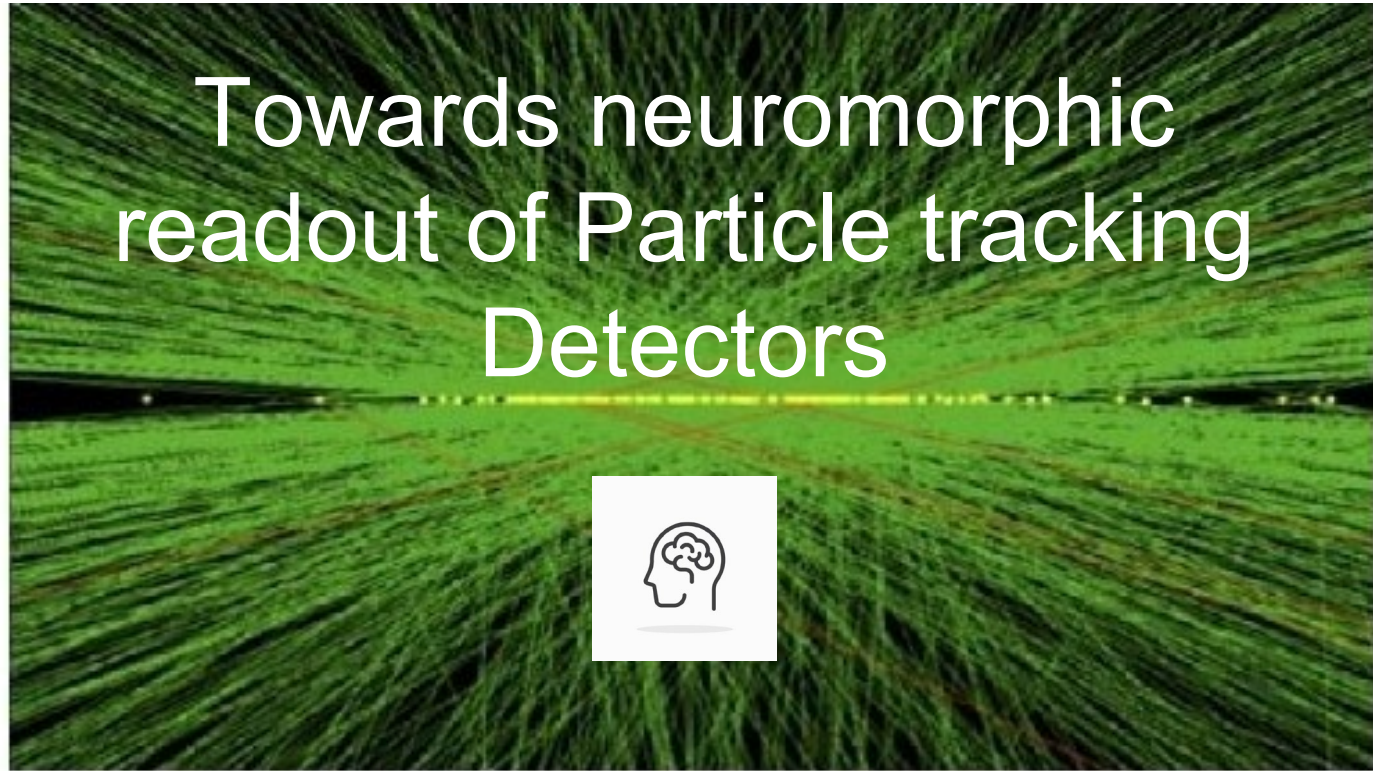




Alice Bean

University of Kansas

January 25, 2024



Courtesy Lindsey Gray, FNAL



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This is work in progress! I'm learning about all of this stuff. People who are helping:

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Fermi National Accelerator Laboratory

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University of Illinois, Chicago

Morris Swartz, Petar Maksimovic
John Hopkins University

Catherine Schuman
University of Tennessee, Knoxville

Derek Grove – University of Kansas,
Evelyn Silva - REU summer student, Vassar Univ.



References

Shruti R. Kulkarni et al., “On-sensor Data Filtering using Neuromorphic Computing for High Energy Physics Experiments”, ICONS 23 Proceedings, arXiv:2307.11242

Jieun Yoo et al., “Smart pixel sensors: towards on-sensor filtering of pixel clusters with deep learning,” arXiv: 2310.02474v (3 Oct 2023)

Manfred Krammer, Frank Hartmann, “Silicon Detectors”,
<https://indico.cern.ch/event/124392/contributions/1339904/attachments/74582/106976/IntroSilicon.pdf>

G. Giacomini, W. Chen, G. D’Amen, and A. Tricoli, *Fabrication and performance of ACcoupled LGADs*, Journal of Instrumentation 14 no. 09, P09004–P09004 (2019).

CMS Collaboration, A MIP Timing Detector for the CMS Phase-2 Upgrade, Tech. Rep. CERN-LHCC-2019-003. CMS-TDR-020, CERN, Geneva, Mar, 2019.



OUTLINE

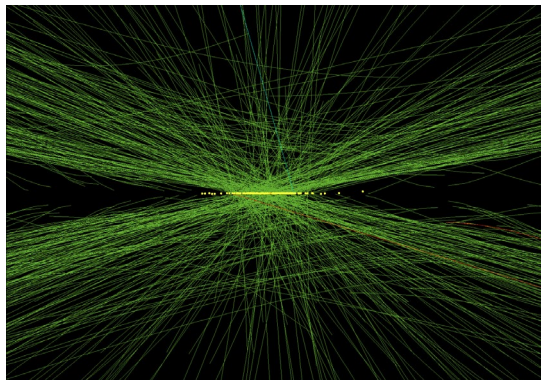
- CMS tracking detectors
- Smart Pixel Project with Neuromorphic algorithm
- Timing detectors
- Neutron detection application



CMS Experiment at CERN

- Physics Goals are many including studying the Higgs and searching for new physics beyond the Standard Model
- Plans are to continue to ramp up luminosity and take data until 2040
- High Luminosity- LHC (HL-LHC) detector being built to start taking data in 2029
- Presently # events/25ns ~50

HL-LHC up to 200!



Courtesy Lindsey Gray, FNAL



Need to have:

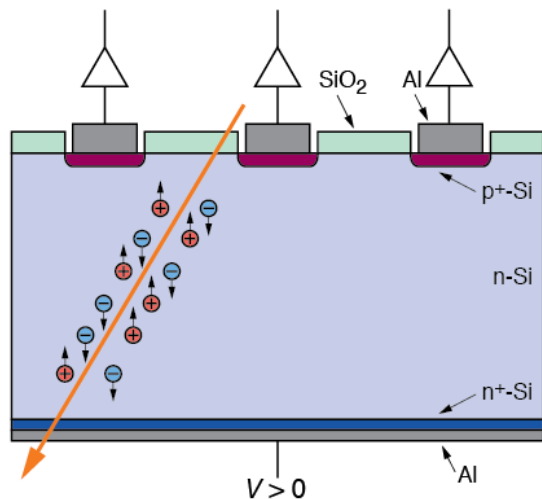
precision tracking –

measuring positions near collision
 $\leq 10 \mu\text{m}$

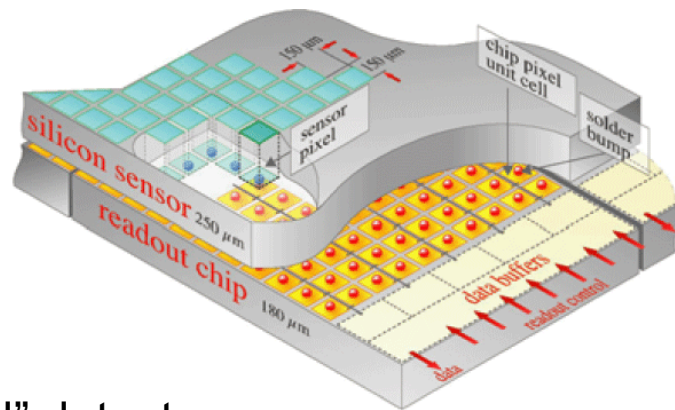
precision timing -

measuring time of arrival before calorimetry
 $\sim 25\text{-}40 \text{ ps}$

Silicon sensors used for tracking



Position resolution determined by pixel size, number of pixels hit and the ability for the cluster charge to be shared among pixels



Traditional CMS tracking “hybrid” detector

sensor collects charge from minimum ionizing particles, separate readout chip bump bonded to amplify/digitize signals

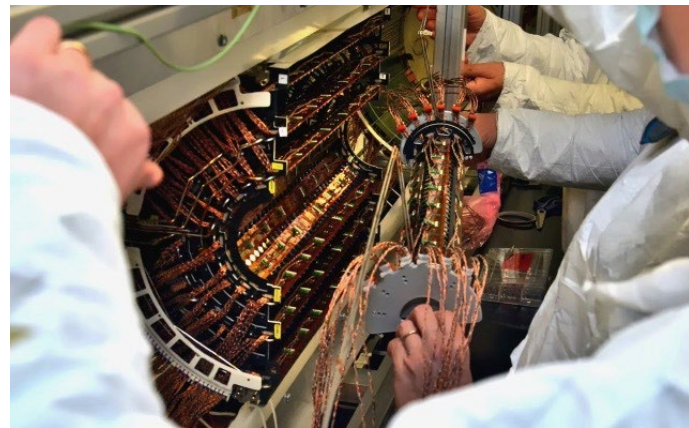


CMS silicon inner tracking detector

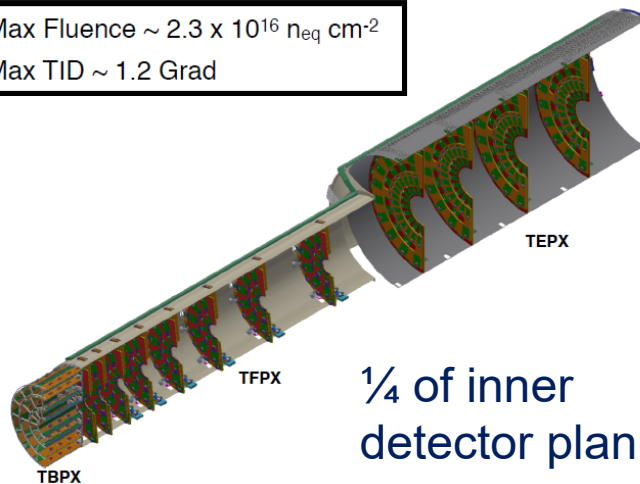
Current detector installed in 2017:
124 Million $100 \times 150 \mu\text{m}^2$ pixels
Measures positions of tracks to:
 $\sim 10 \mu\text{m}$ precision
starting at 3cm from beam

HL-LHC detector to be installed in 2028:
2 Billion $100 \times 25 \mu\text{m}^2$ pixels

Will be used to disentangle tracks
multiple interactions



Max Fluence $\sim 2.3 \times 10^{16} \text{ n}_{\text{eq}} \text{ cm}^{-2}$
Max TID $\sim 1.2 \text{ Grad}$



$\frac{1}{4}$ of inner
detector planned



Challenges

- POWER
- Data Rates
- Radiation
- Amount of Material
- Getting signals in and out
- CAN YOU BUILD IT?



Challenges

- **POWER**

50-60kW with 40-50kA@ 1.2V

- **Data Rate** collisions at 40MHz

operate readout chips @ 1.28 Gbps

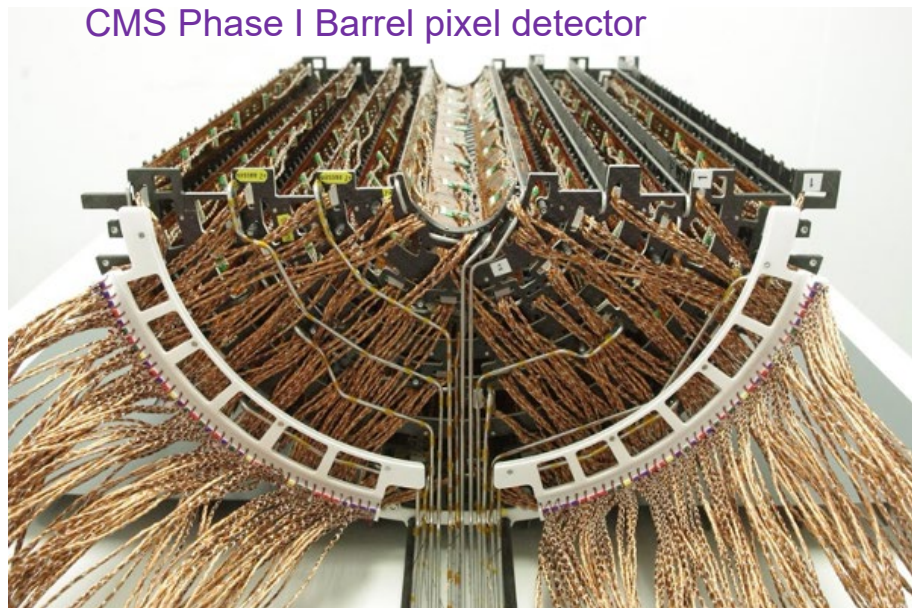
transmit off detector with 5-10 Gbps fibers

- **Radiation** particle rate 3 GHz/cm²

~ 1×10^{16} N/cm² by end of run

Challenges

- POWER
- Data Rates
- Radiation
- Amount of Material
- Getting signals in and out



The 36 AWG twisted pair e-link cables used to get signals out for HL-LHC detector at 1.28 Gbps are hard to build, have mass, and are hard to route!



Challenges

CAN YOU BUILD IT?



CMS HL-LHC inner tracker

Yes! We think we can build it!

What about the next/better tracker?

Can you build it?

Need to reduce data at source for next generation detector

Information flow now

On Detector

Keep only pixels
with charge above
threshold, digitize

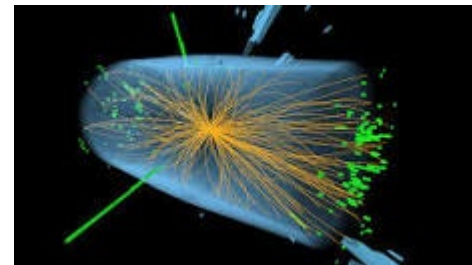
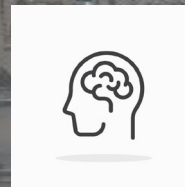


Pixel x, y, E

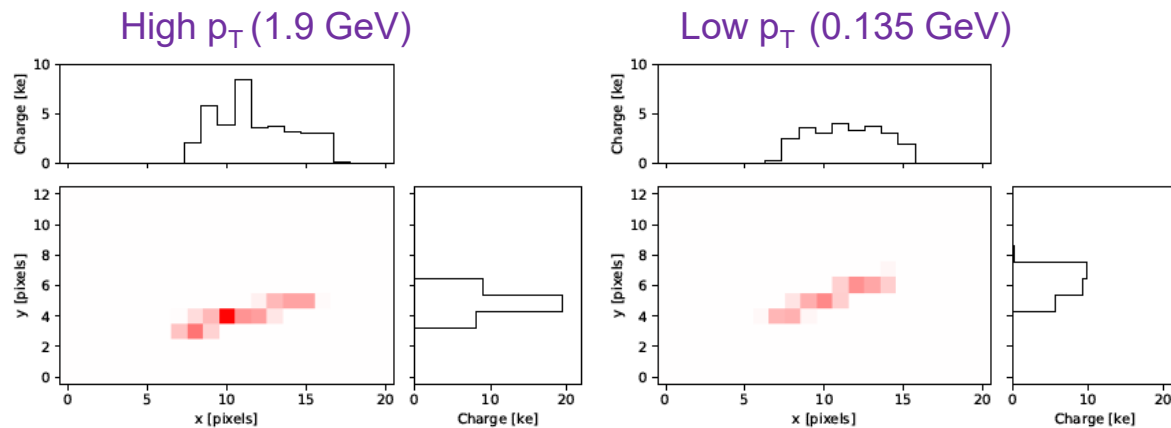
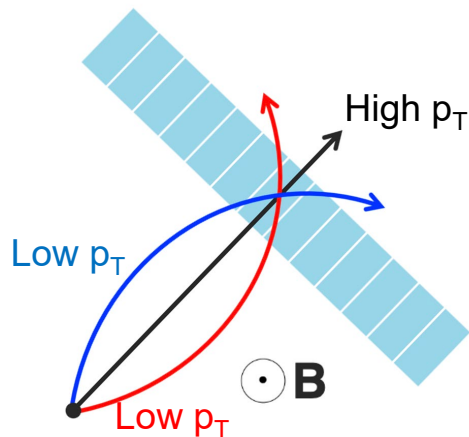
Off Detector

Find cluster position,
energy, tracking
Lots of AI

BRAIN



Smart Pixel Project -reduce data at source

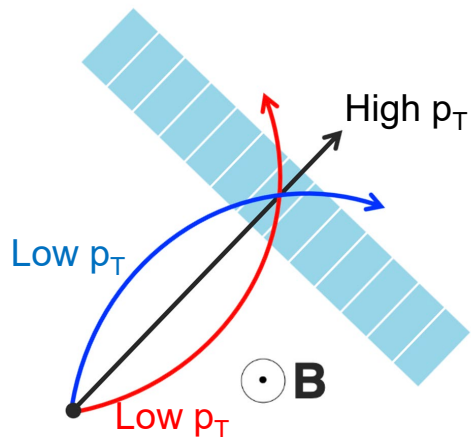


Application for **after** HL-LHC pixel detector

Use ASIC for filtering out pixel clusters with $p_T \leq 0.2$ GeV

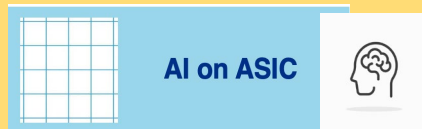
Smaller Pixels are $12.5 \times 50 \mu\text{m}$

Smart Pixel Project -reduce data at source



On Detector

Keep only pixels
in clusters with
 $p_T \geq 0.2 \text{ GeV}$

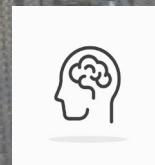


Pixel x, y ,
cluster η, ϕ, E

Off Detector

Find tracking
Lots of AI

BRAIN

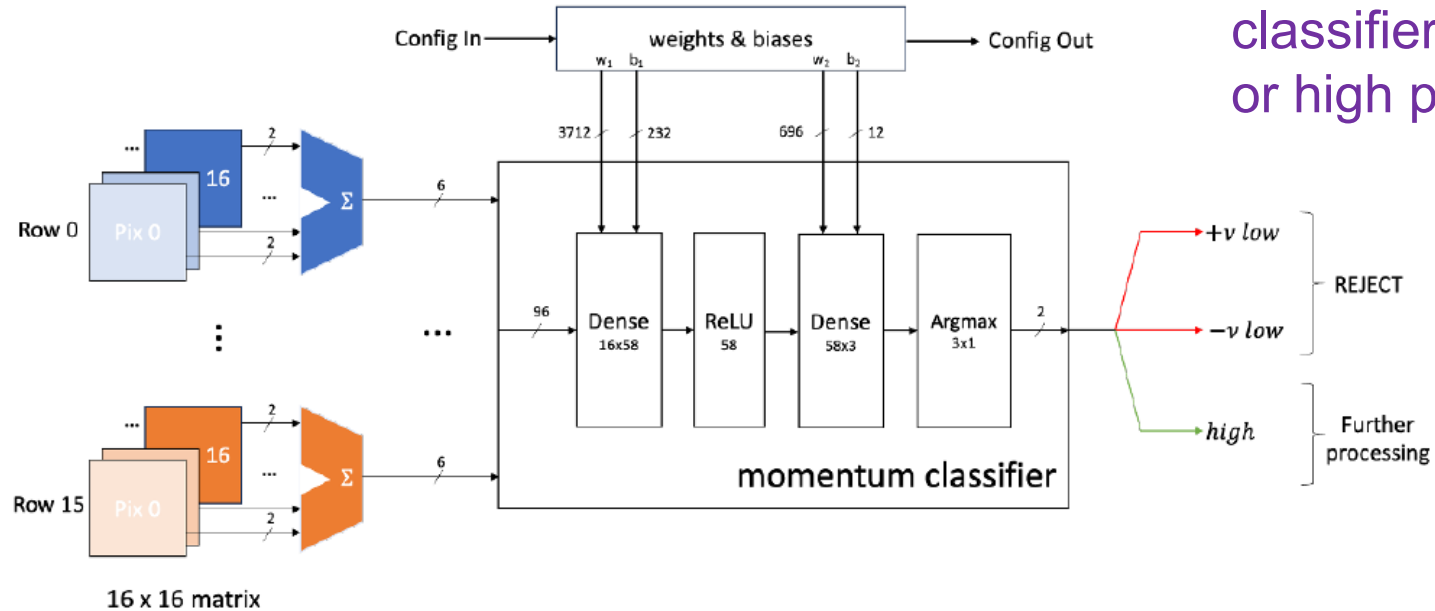


Application for after HL-LHC pixel detector

Use ASIC for filtering out pixel clusters with $p_T \leq 0.2 \text{ GeV}$

Pixels are $12.5 \times 50 \mu\text{m}$

Conventional Neural Network



Output is classifier of low or high p_T

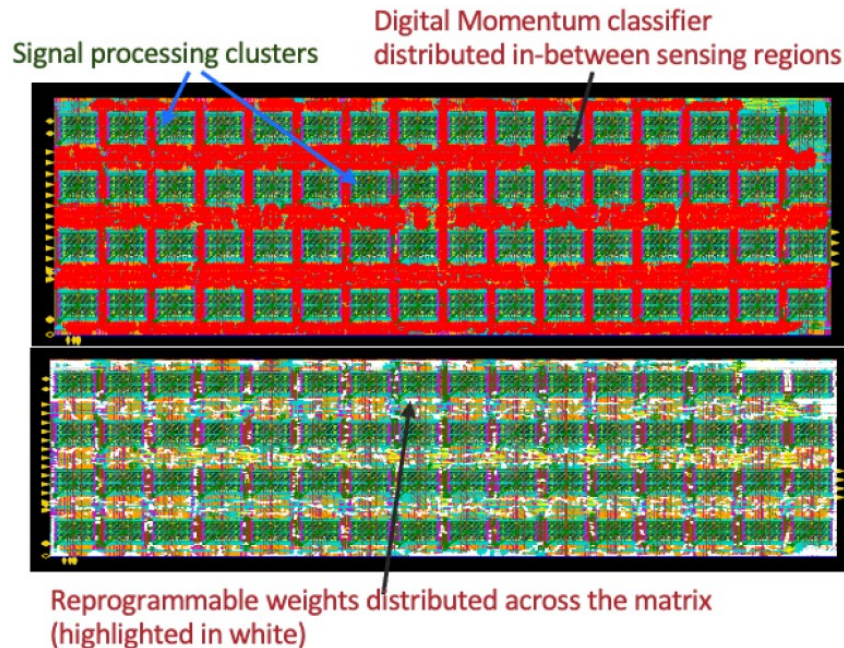
Smart pixel approach

First project: Implement
Classification neural network on chip:

- Use neural network with reprogrammable weights
- 28nm chip layout for 16x16 pixel array

Next → 3D sensor+chip together

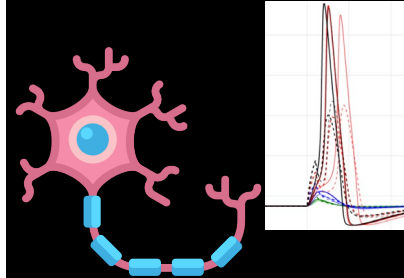
Also look at Spiking Neural
Network (SNN)
implementation



Neuromorphic Computing



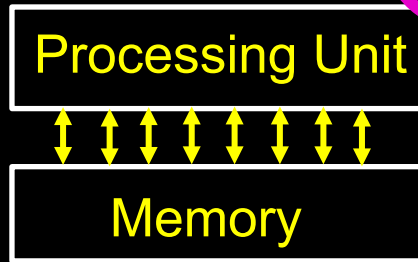
~20 watts



Neurons process
and store
information as
needed



Lots of watts



- Processing and
Memory Separated
- Clocked data uses
lots of power

Dynamic Vision Systems

Applications which use Neuromorphic Computing techniques today
These are event based or asynchronous images which only respond to changed pixels in images similar to how human eye works

They use less power can capture faster frame rates than conventional image processors

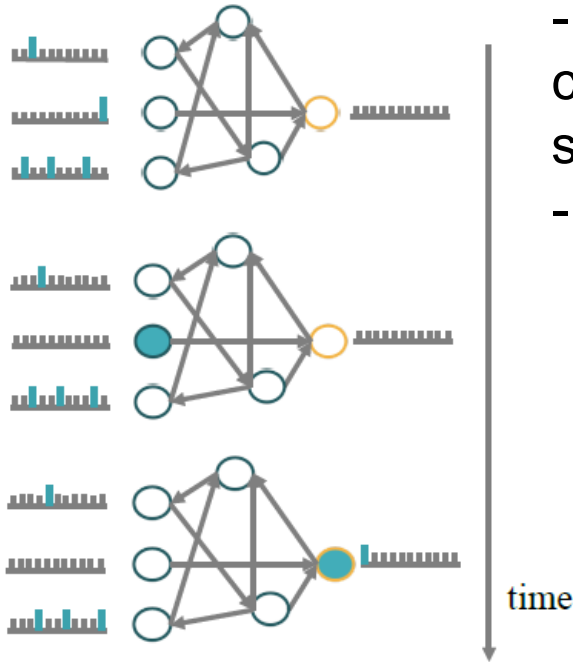
See: talk by Christoph Posch
at <https://indico.cern.ch/event/1245654>

Founded Prophesee systems can process 10K fps, with <10mW
(Prophesee.ai)

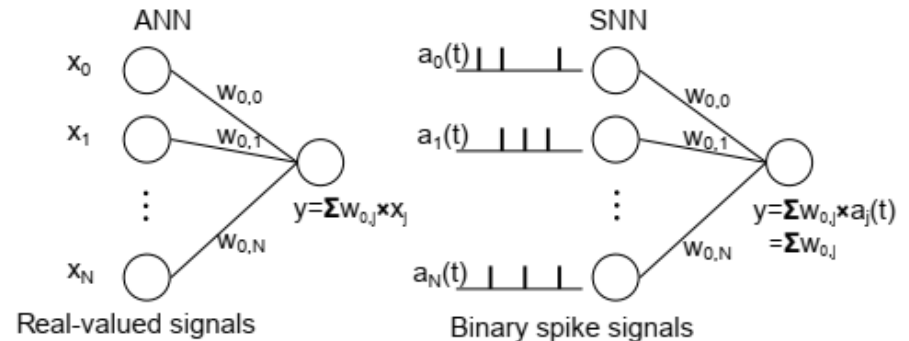


Spiking Neural Networks (SNN)

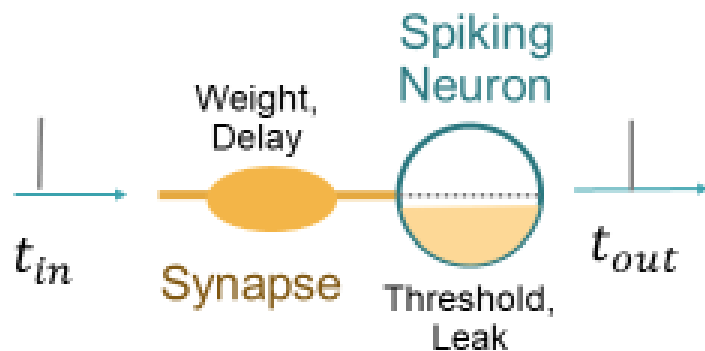
Spiking Neural Networks



- Data spikes excite “neurons” which communicate via synapses by transmitting spikes
- Use temporal information



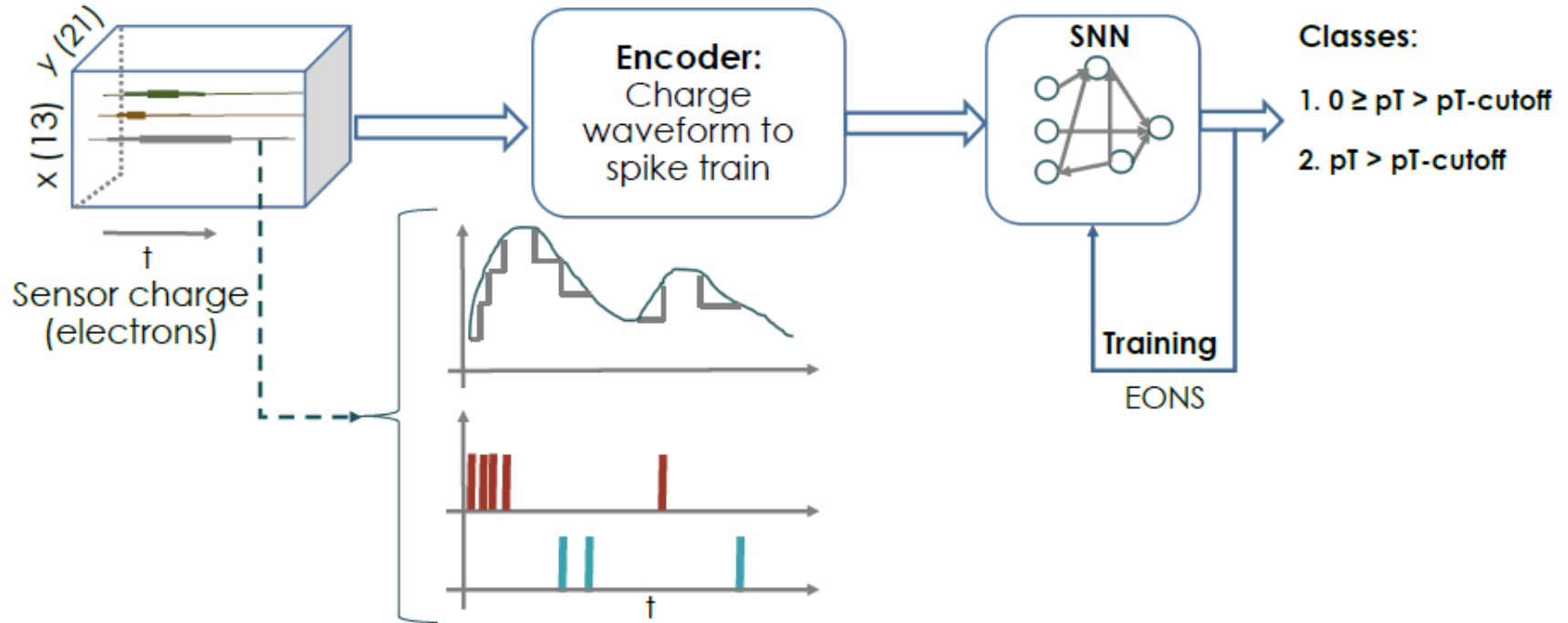
Neuromorphic Computing – Basic Model of Operation



- Inherent temporal dynamics
- Neuron fires with charge exceeding threshold – **Leaky Integrate and Fire**
- Multiple tunable parameters
 - Synaptic weights, delays
 - Neuronal thresholds, leaks
 - Network architecture

Oak Ridge National Lab has implemented models of these neurons and synapses in FPGA systems for radiation monitoring that use very little power

Smart pixel SNN implementation

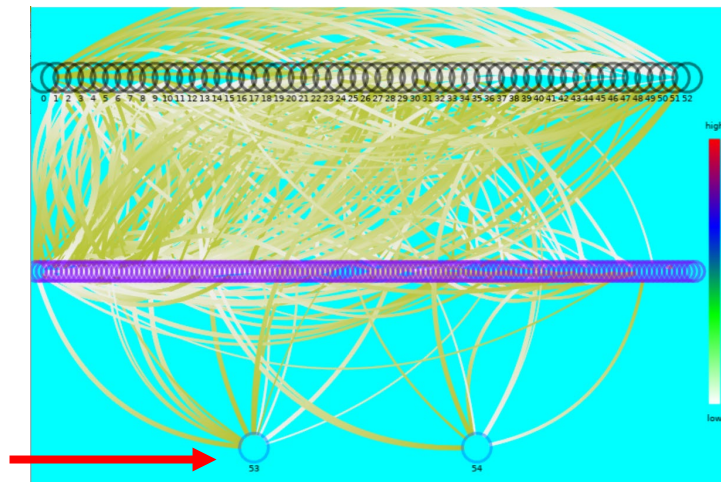




Results of SNN trained classifier for Smart Pixels

- Use Network with highest signal efficiency for $p_T > 2.0\text{GeV}$ clusters
 - Trained Network size: 84 neurons, 493 synapses

Models	DNN	DNN (quantized)	SNN (this work)
Signal Efficiency	94.8 %	91.7 %	91.89%
Data Reduction	24.02 %	25.71 %	25.47%
Neurons	128	128	84
Parameters	2049	2561	930

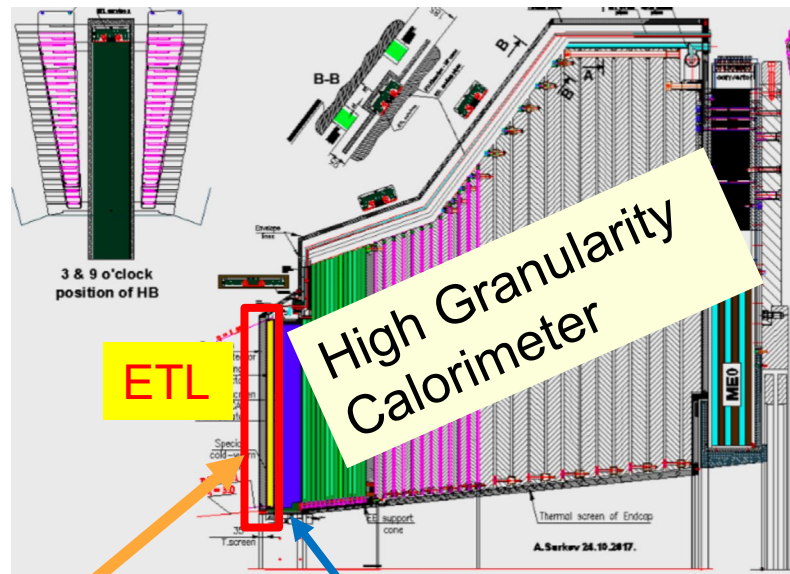


2 output neurons that fire depending on which of the two classes



Can we use SNNs for
data reduction for high
precision position and
timing detectors?

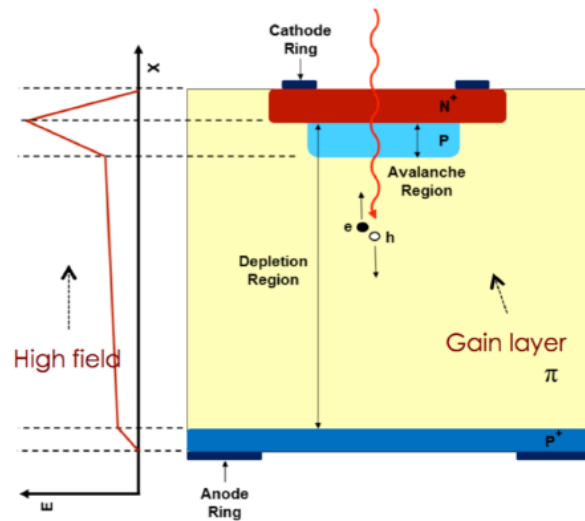
Precision Timing with Silicon Detectors



Interaction point
~1m away

Neutron
moderator

CMS HL-LHC Endcap Timing Layer



Low Gain Avalanche Diode (LGAD)
Measure time of track to ~25ps

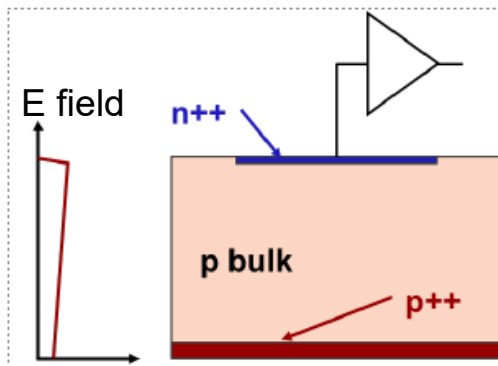


Challenges

- Electronics readout chip and readout system need to be carefully designed to obtain timing precision
 - currently 25 ps for CMS ETL
- Little charge sharing among pixels with current LGAD design
 - position resolution limited by pixel size
~1.3mm x 1.3mm

Timing information with Low Gain Avalanche Diode (LGAD) detectors

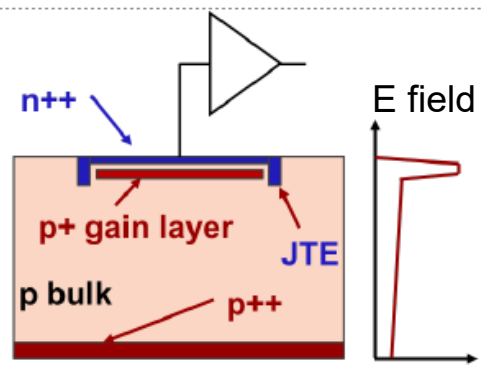
Traditional Silicon Detector



CMS Inner Tracker

need sufficient thickness

Ultra Fast LGAD

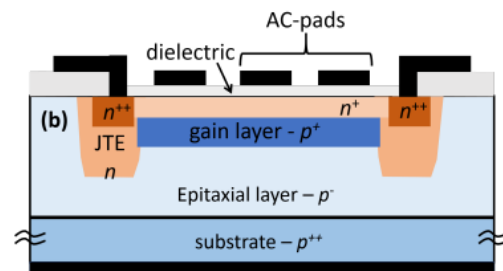


CMS ETL

thin but limited

acceptance due to segmentation

AC coupled LGAD



?

thin

no acceptance limit

Substrate

Charge Sharing
among pixels

yes exploited

limited charge sharing due
to electrode size

exploitable thanks to
small electrode

Position resolution

~3-10 μ m

~50-300 μ m

1-100 μ m?

Timing resolution

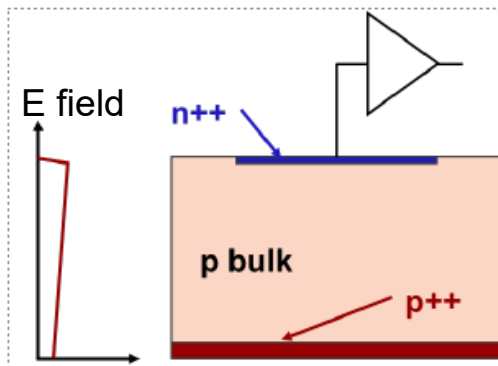
~ns

~20-40 ps

20-40ps?

Timing information with Low Gain Avalanche Diode (LGAD) detectors

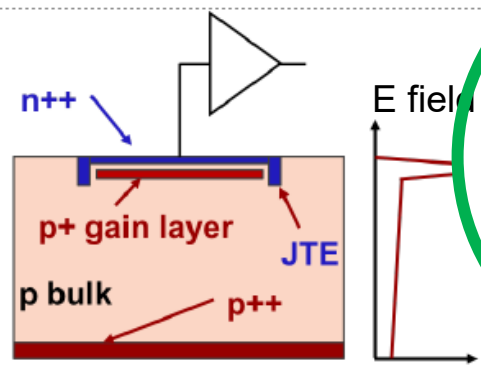
Traditional Silicon Detector



CMS Inner Tracker

need sufficient thickness

Ultra Fast LGAD

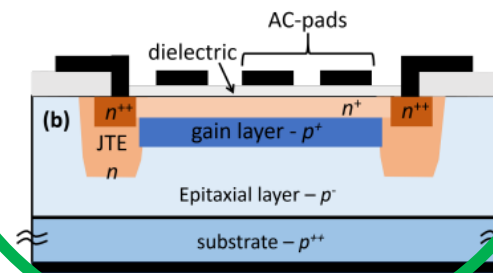


CMS ETL

thin but limited

acceptance due to segmentation

AC coupled LGAD



?

thin

no acceptance limit

Charge Sharing
among pixels

yes exploited

limited charge sharing due
to electrode size

exploitable thanks to
small electrode

Position resolution

~3-10 μ m

~50-300 μ m

1-100 μ m?

Timing resolution

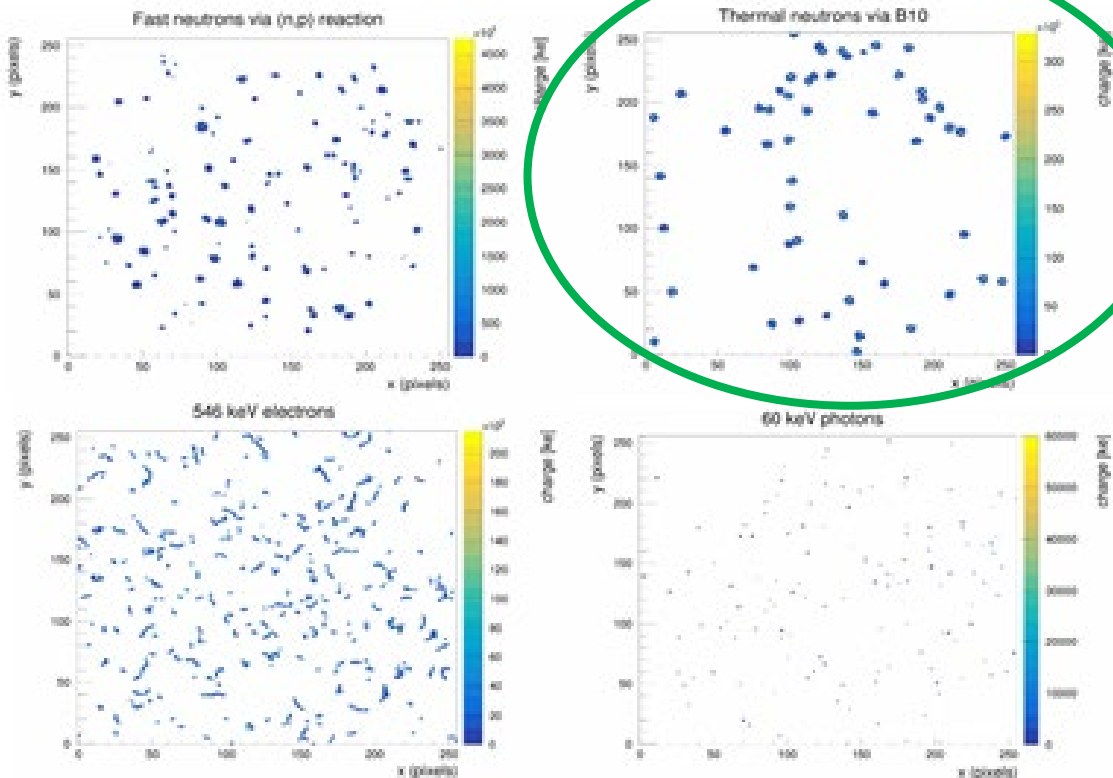
~ns

~20-40 ps

20-40ps?

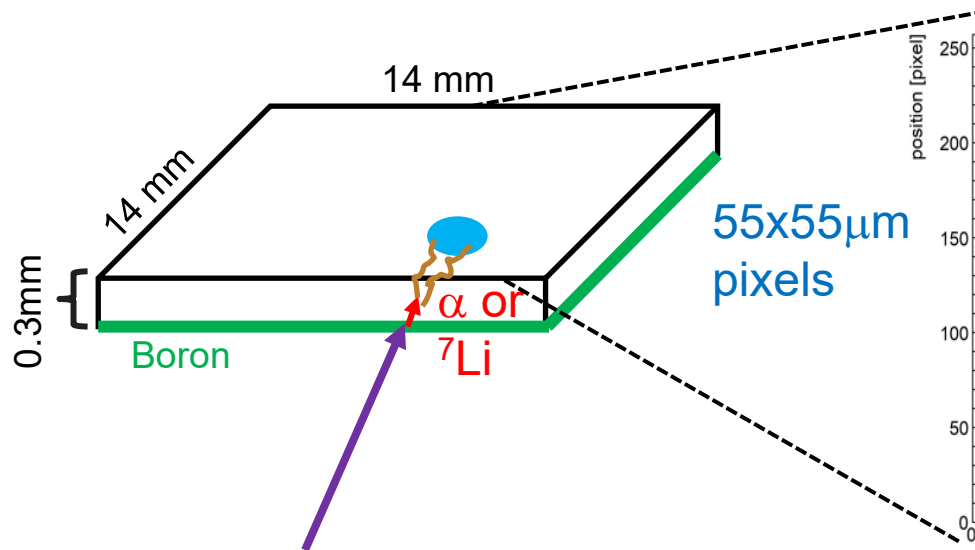
Neutron Beam monitor application

- Thermal neutrons interact with ^{10}B . Large clusters form from plasma charge effect
- Needs precision timing and tracking
- Eventually will use AC coupled LGAD detectors

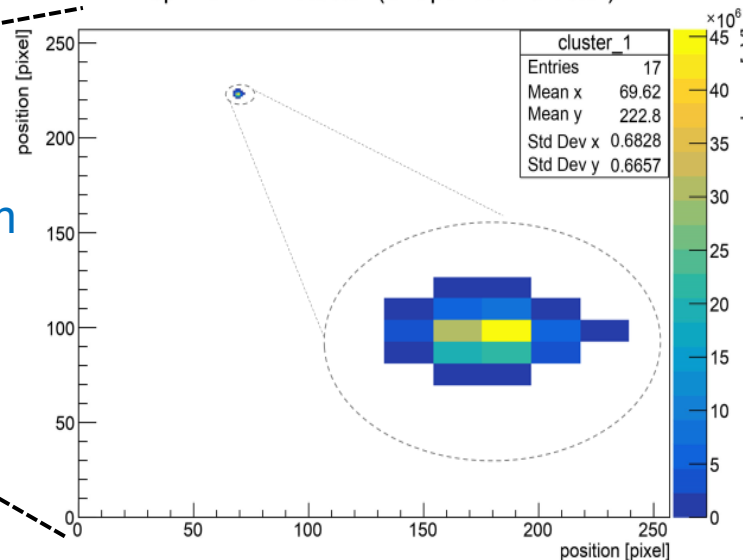


Neutron Beam Monitor Application

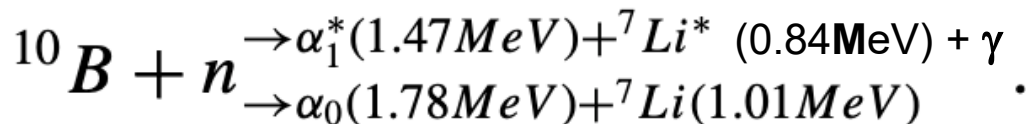
First start with traditional silicon simulation



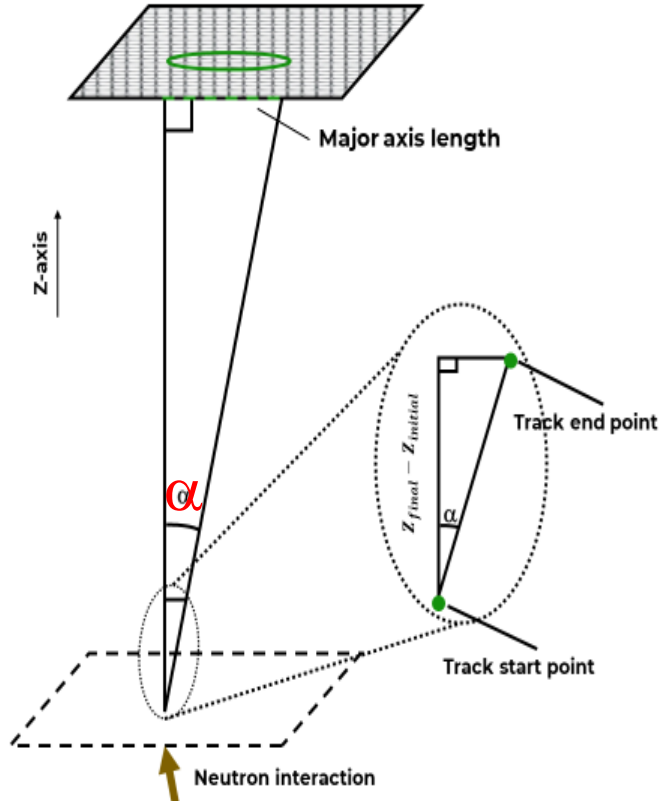
Top View of Detector (Snapshot of Cluster)



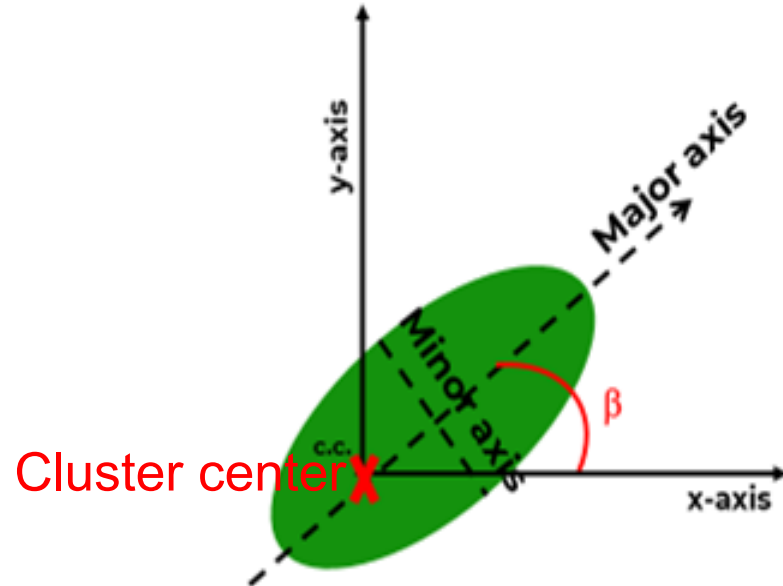
Thermal neutron - n



Size of cluster ~35 pixels



2D Top View of Detector



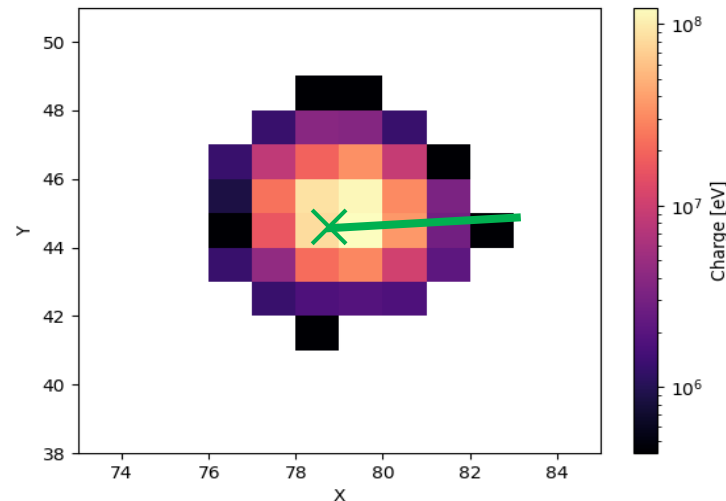
First steps

- Use traditional silicon detector (not LGAD) to start
55×55μm pixels in 0.3mm thick detector
- Use Timepix4 ASIC for readout
can use time-over-threshold,
time of Amplitude (<200ps) with
16 bit counter, ~800e threshold,
3.6MHz/mm²/s hit rate capability
- Simulate events with Allpix2 software

(<https://allpix-squared.docs.cern.ch/>)

Has model for Timepix output already included

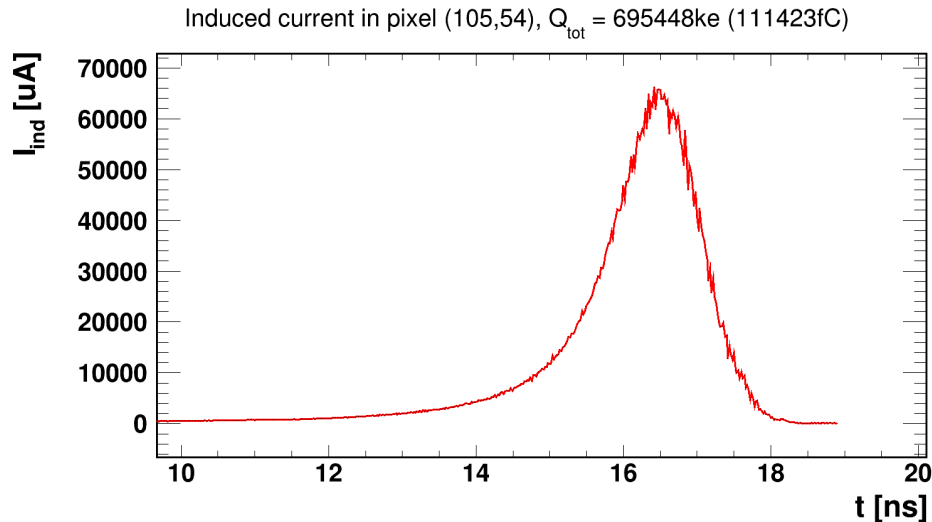
- Calculate Offline x_c , y_c , E , α , β and compare with
truth information to measure resolution



res (x_c , y_c) ~1μm

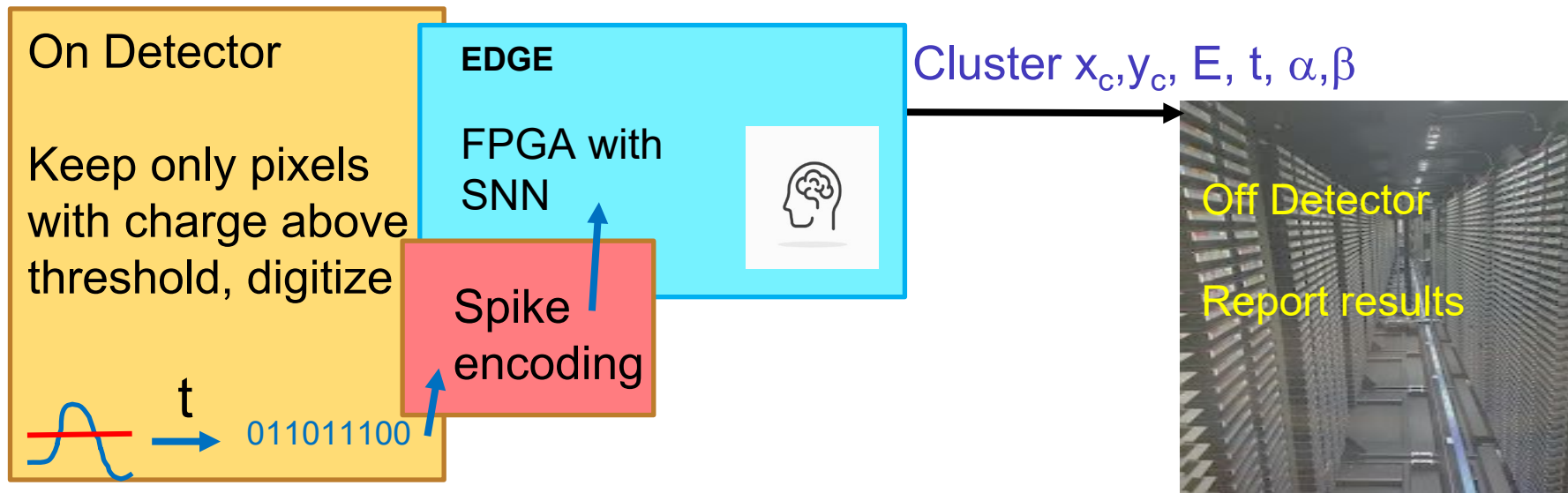


Next Steps



- Put time characteristics into simulation
 - Implement detailed model of charge multiplication, electrostatic repulsion (plasma effect), electric field (using TCAD) in MC Charge transport code (Allpix²) and output
- Study encoding for SNN
- Train SNN to output regression variables
- Implement SNN in FPGA

Start with SNN implementation on FPGA



SNN with Digital stream pipeline maps well to FPGA
 Has low latency for sparse data
 DNNs are harder to implement on FPGAs

Later SNN implementation ASIC

On Detector



Dream for full
SNN analysis of
cluster information
with ASIC

Cluster $x_c, y_c, E, t, \alpha, \beta$

Off Detector

Report results



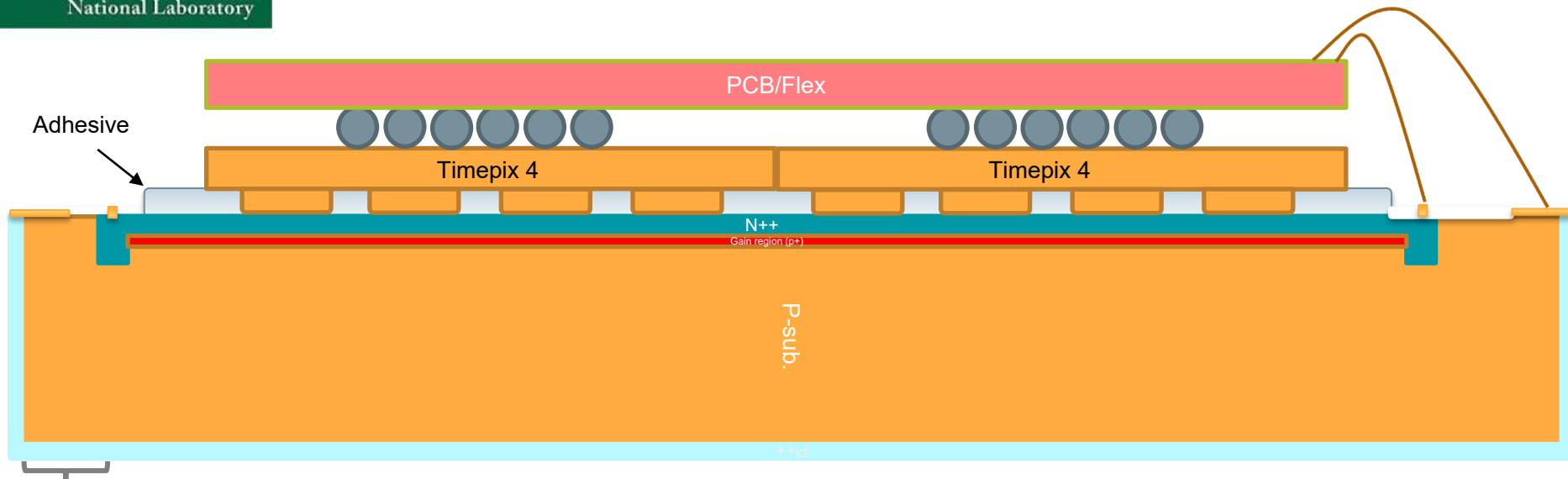
Big issue: Spike Encoding implementation at speed



Challenges

CAN YOU BUILD IT?

M. Benoit Proposal



Guard ring/Active edge region

Can achieve capacitive coupling and transmit signal with good efficiency via a thin adhesive layer

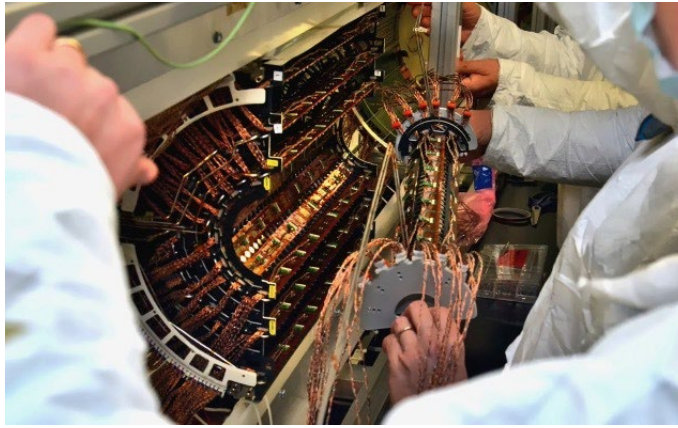
- **No electrode required** on the AC-LGAD side, making it a simpler structure to manufacture.
- AC Coupling can be achieved via a combination of **thin oxide-adhesive layer**, **ENIG pads from ACF experience**
- **No complex interconnect** method necessary, no lithography
- Multiple ASICs can be tiled on a **large area AC-LGAD**
- **Large charge deposition and multiplication provide spatial amplification**

Summary

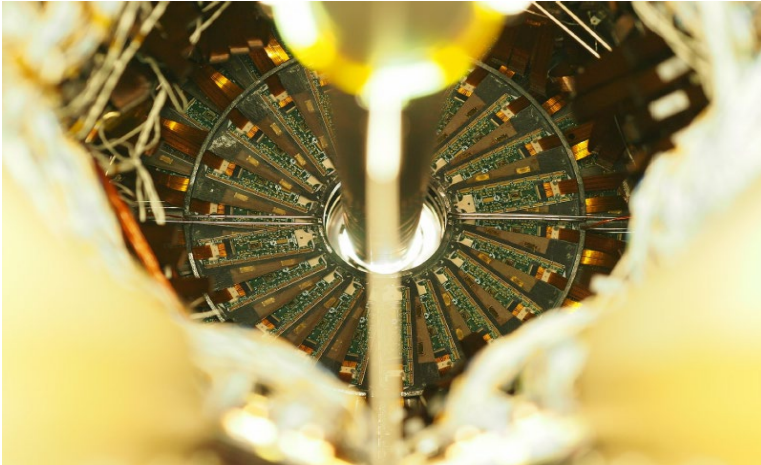
- Dreams of particle detectors with precise position and time resolution have to confront how to reduce the data output from the detector
- Lots of work now in implementing AI on ASICs for particle tracking applications
- Many technical hurdles are being overcome
- There is a pathway to implementing SNNs on the detector

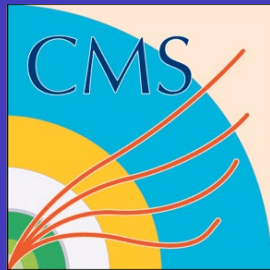


BACKUP



CMS inner detector installed in 2016 built partially at KU

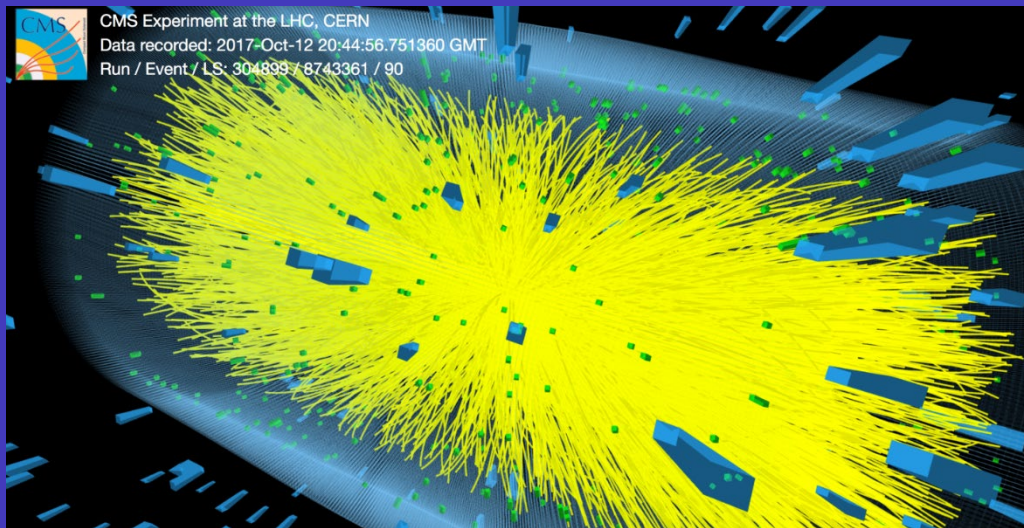




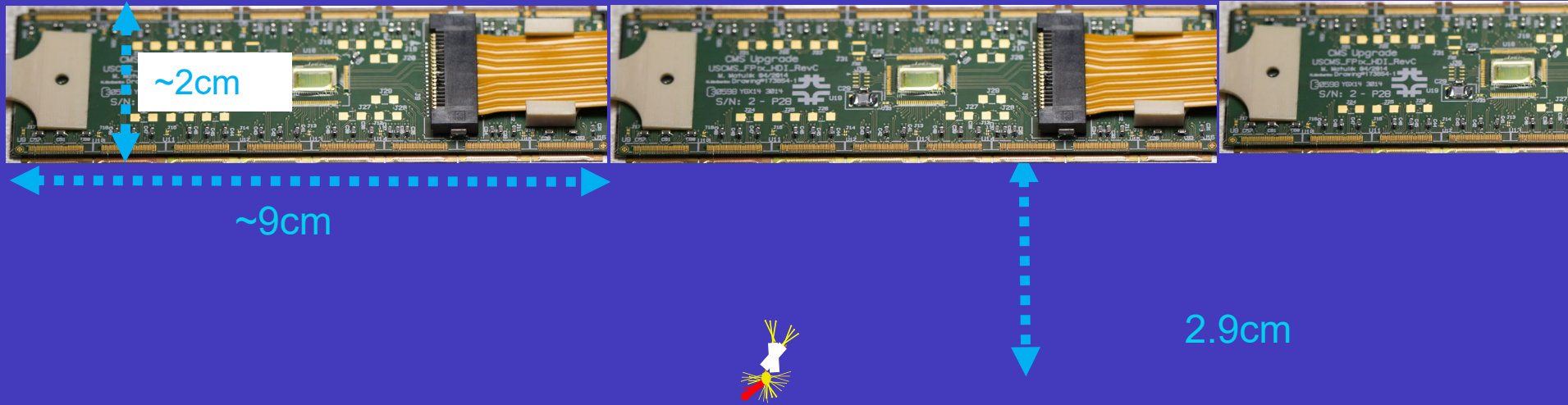
Build detectors that track particle trajectories

Current CMS inner tracking detector

- made from silicon
- includes 124 million pixels
- measures positions with resolution $\sim 10\mu\text{m}$



They then travel to detector



Zoom out some more - this shows part of the first layer of detectors

Standard Model of Particle Physics

www.quarked.org

FUNDAMENTAL PARTICLES

Fundamental Particles are the smallest things scientists have discovered. So far, no-one has been able to split them into smaller pieces. Everything else is made of combinations of these particles.

FERMIONS

Fermions are particles that cannot exist together if they have identical properties.

BOSONS

Bosons are particles that can group together even if they have exactly identical properties.

PARTICLES

Quarks there are 6 Flavors of Quarks:

Positive Electric Charge LESS MASS	>>	MORE MASS
		
Up (u)		Charm (c)
Negative Electric Charge LESS MASS	>>	MORE MASS
		
Down (d)		Strange (s)
	>>	MORE MASS
		
		Bottom (b)

Leptons there are 6 types of Leptons:

No Electric Charge		
		
Electron-Neutrino (ν_e)	Muon-Neutrino (ν_μ)	Tau-Neutrino (ν_τ)
Negative Electric Charge LESS MASS	>>	MORE MASS
		
Electron (e)	Muon (μ)	Tau (τ)

ANTI-PARTICLES

Anti-quarks

Negative Electric Charge LESS MASS	>>	MORE MASS
		
Anti-up ($\bar{u}$)		Anti-charm ($\bar{c}$)
Positive Electric Charge LESS MASS	>>	MORE MASS
		
Anti-down ($\bar{d}$)		Anti-strange ($\bar{s}$)
	>>	MORE MASS
		
		Anti-bottom ($\bar{b}$)

Anti-leptons

No Electric Charge		
		
Anti-electron-Neutrino ($\bar{\nu}_e$)	Anti-muon-Neutrino ($\bar{\nu}_\mu$)	Anti-tau-Neutrino ($\bar{\nu}_\tau$)
Negative Electric Charge LESS MASS	>>	MORE MASS
		
Positron ($\bar{e}$)	Anti-muon ($\bar{\mu}$)	Anti-tau ($\bar{\tau}$)

Force Carrier Particles:

Electromagnetic Force: Photons



Weak Force:



Strong Force: Gluons





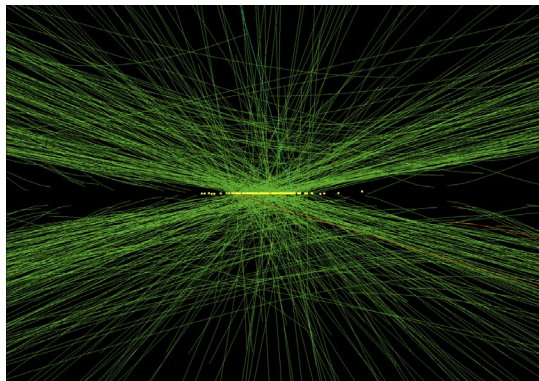
Calculate Offline x_c , y_c , E , α , β



Getting more data implies improving tracking detector

- New detector being built to start taking data in 2029
- Electronics supports taking 40 Million snapshots/second with up to 200 collisions expected per snapshot

Now



Simulation of
expected
snapshot with
new detector



Courtesy Lindsey Gray, FNAL

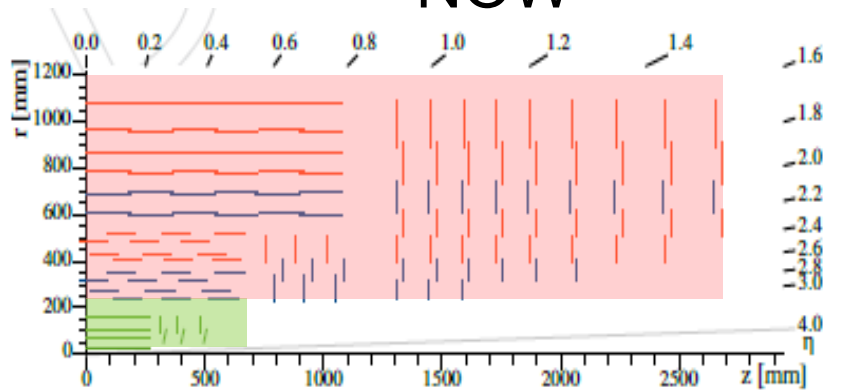
Need: Radiation tolerance, Increased Granularity, Robust Pattern Recognition



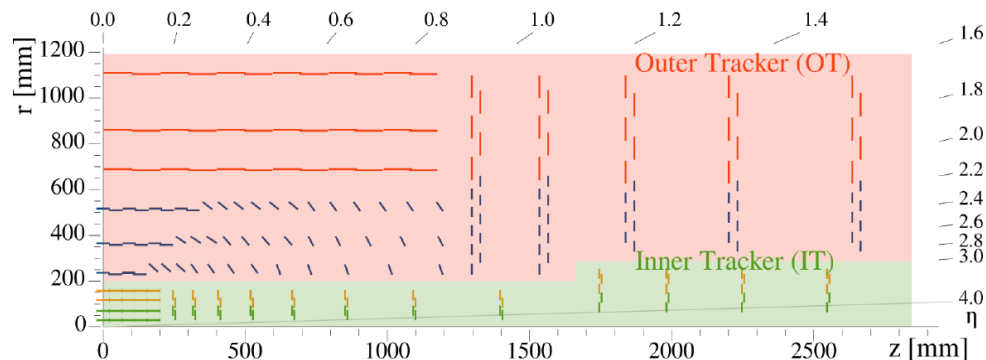
Design of HL-LHC CMS tracking detector

Tracking Detectors

NOW



Proposed HL-LHC



1/4th of total tracker

Both have Inner pixel detector with Outer silicon strips

Bigger differences include:

Inner pixel detector has 124M channels

No trigger information



Inner pixel tracker has 2B channels!

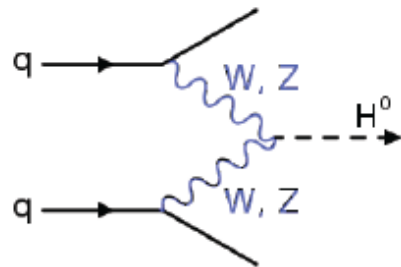
Outer tracker has L1 trigger modules

Some Physics Goals

1. Study the SM Higgs

Vector Boson Fusion production important
need forward jets and leptons

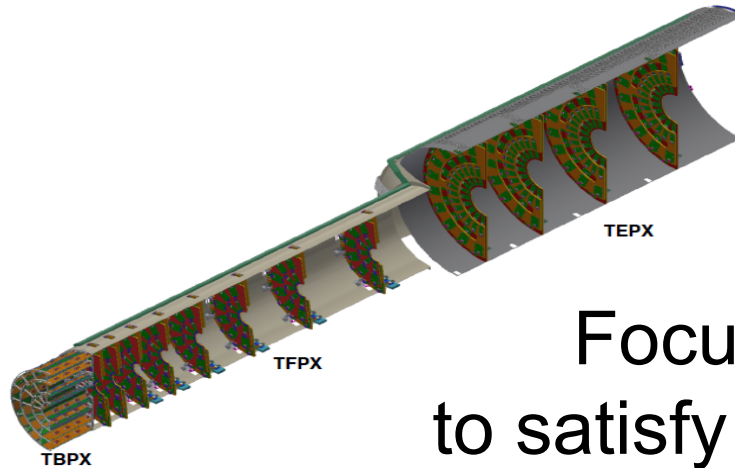
Higgs Production
via Vector Boson
Fusion



2. Look for Beyond Standard Model physics

Need good mass resolution:

Improved 2-track separation, Reduced
material in tracking volume



Focus here: **Inner Tracker**
to satisfy most of the requirements

Requirements,

Contribution to the level-1 trigger – from outer tracker

Increased granularity

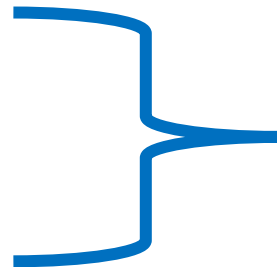
Improved two-track separation

Robust pattern recognition

Extended tracking acceptance in η

Radiation Tolerance

Reduced material in tracking volume



- Large channel count

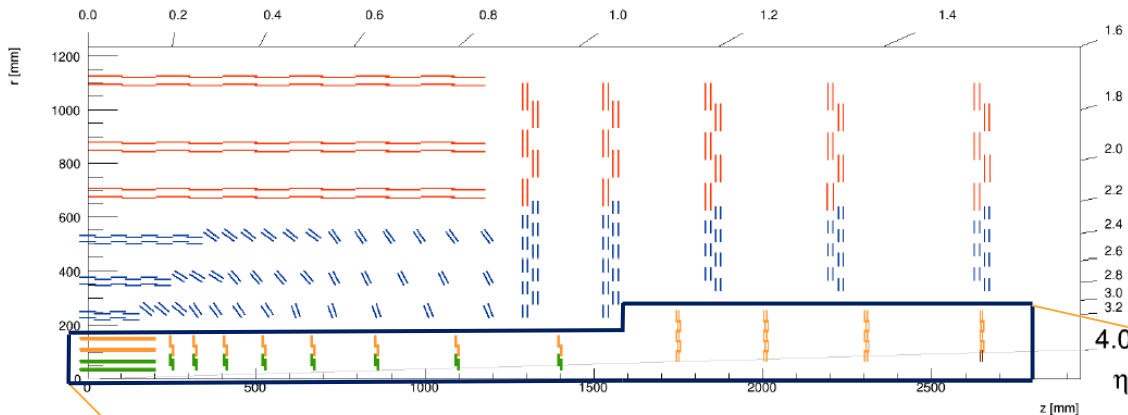
- Smaller pixel size
(by factor of 6)

100 x 25 μm^2

- More forward detectors

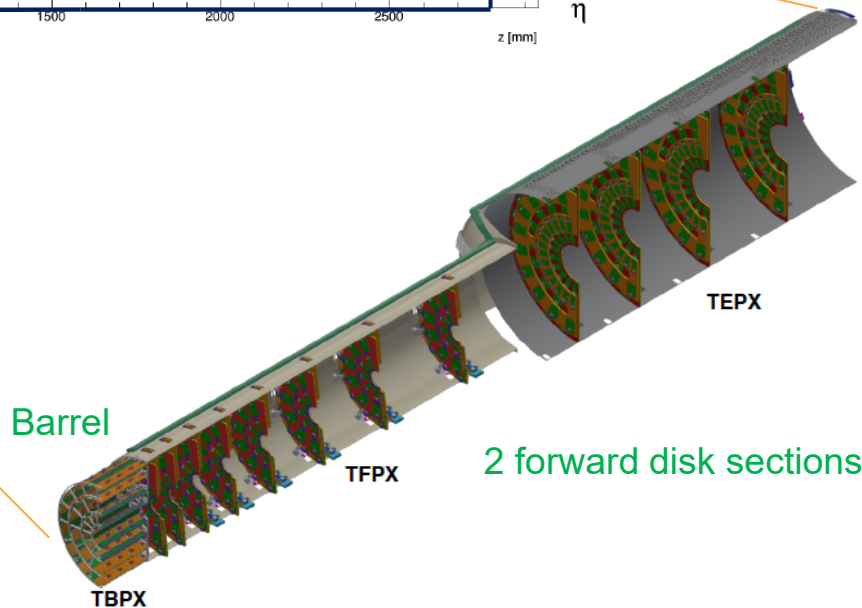


The CMS Tracker for HL-LHC

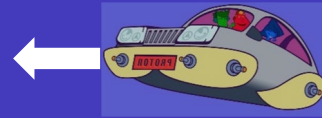
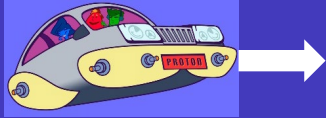


Inner Pixel Tracker

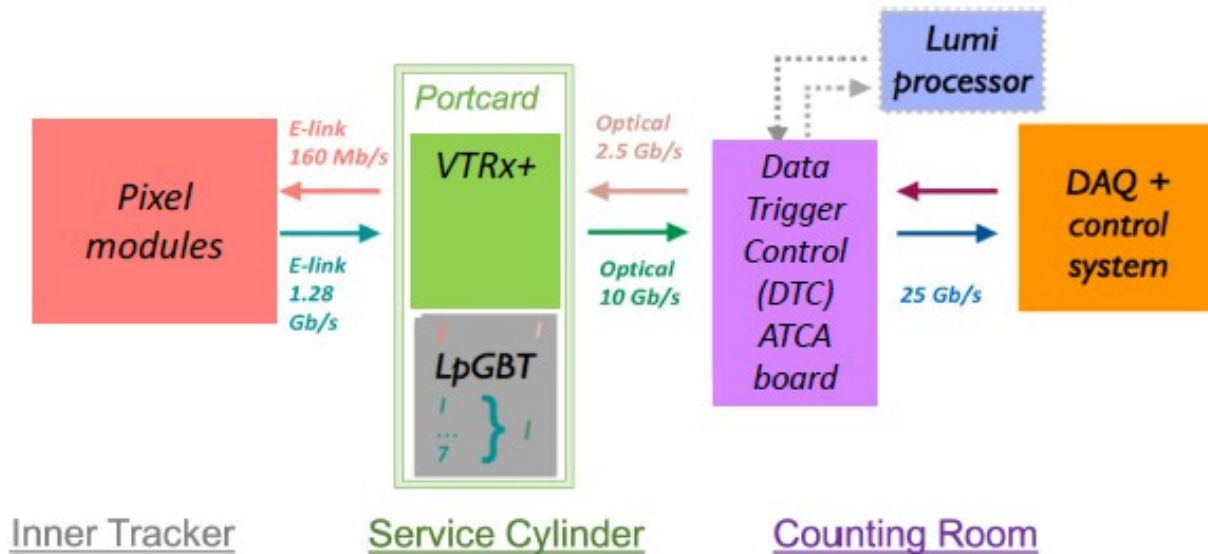
- 4352 modules, 4.9 m², 2×10^9 pixels
- Coverage up to $|\eta|$ of 4.0
- Occupancy < 0.1%
- Hybrid modules with 2 or 4 readout chips
- Mechanics made so it can be removed for maintenance
- Maximum Fluence $\sim 2.3 \times 10^{16} n_{eq} \text{ cm}^{-2}$



Two Protons collide



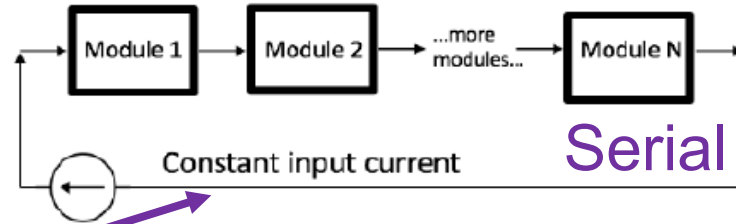
More Challenges



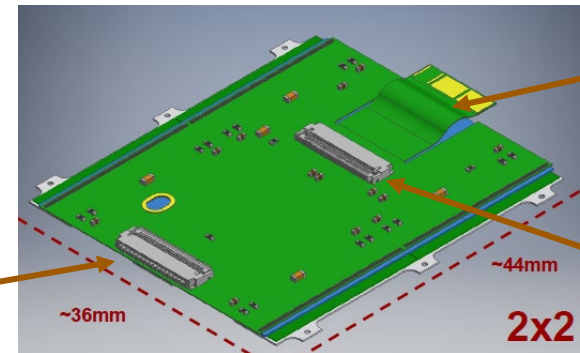
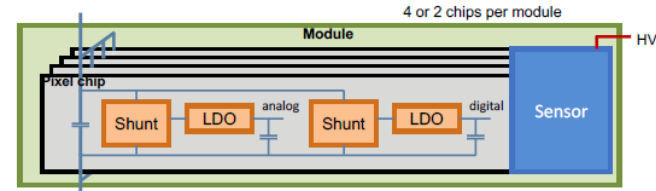
- POWER
- Data Rates
- Materials and getting signals in and out

KU Modules and Powering

- **Enormous power budget**
50-60kW with 40-50kA @ ~1.2V
- Problems with direct powering
too much material required
- Problems with local conversion
 - DCDC converter radiation hardness
 - space limitations
- Pixel Readout chip is only active electronics on Module High Density Interconnect (HDI)



Serial Powering



Power output connector

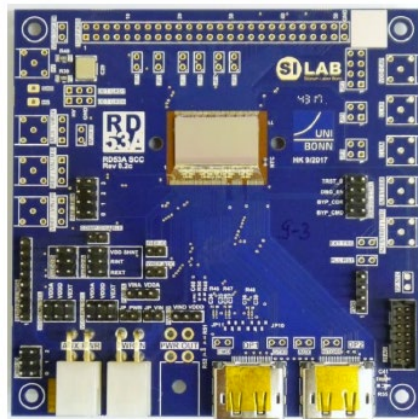
Power input pigtail

I/O connector for e-link

Module HDI design

Readout Chip Development by RD53

Collaboration for ATLAS and CMS



NOW:

Chip communicates at
400 Mb/s

- 65 nm CMOS technology
- $50 \times 50 \mu\text{m}^2$ cell
- Low threshold ($\leq 1000 e^-$)
- High Hit and Trigger rate
 - up to $4 \times 1.28 \text{ Gb/s}$
- Radiation Resistant
- Serial Powering capabilities
- CML Signal Driver
- CMS chip size:
 - $16.8 \times 21.6 \text{ mm}^2$
 - 336×432 cells

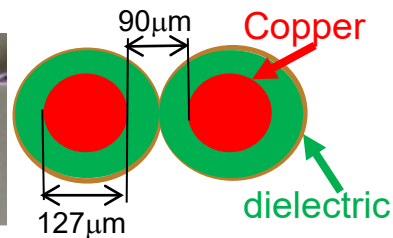
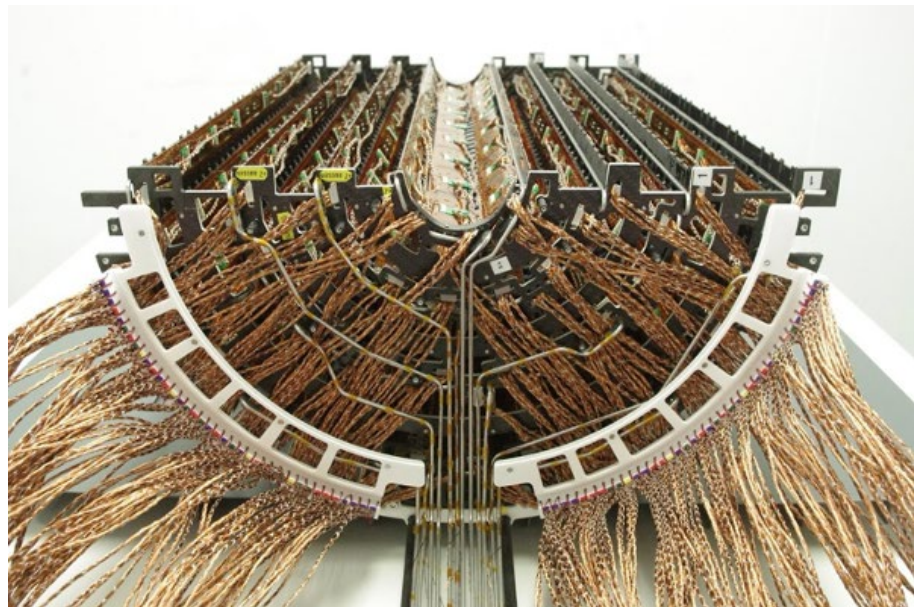


An example – electronic links

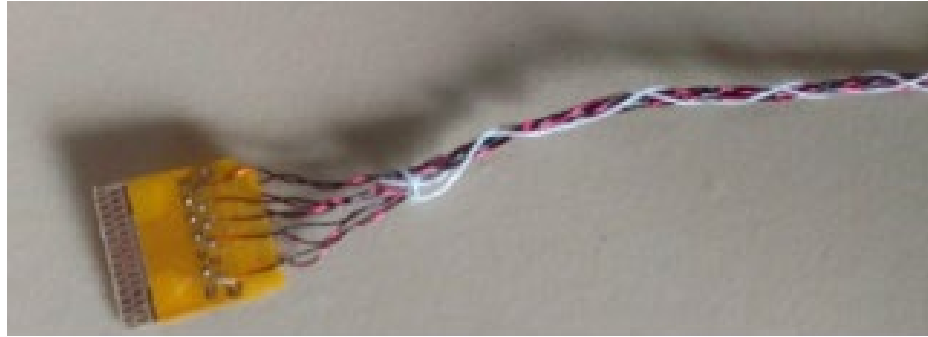
Use differential micro-twisted pairs

Pixel Barrel detector NOW
uses 36 AWG twisted pairs

So what's to worry about: **1.28 Gbps**



- Need to match 100Ω impedance
- Small wire has big skin losses
- Transmitting signals up to 2m
- Cross talk
- manufacturing
- Fragile construction ...



E-links can be built to transmit 1.28 Gbps signals up to ~2m

Building and testing these is non-trivial

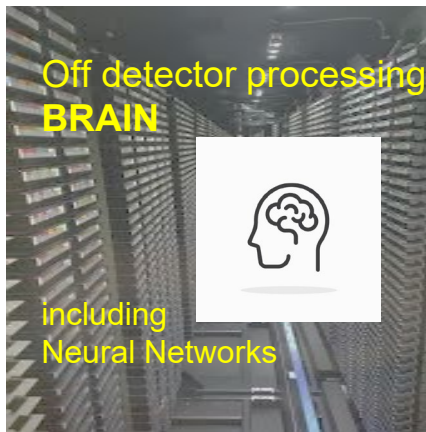


What about the future?

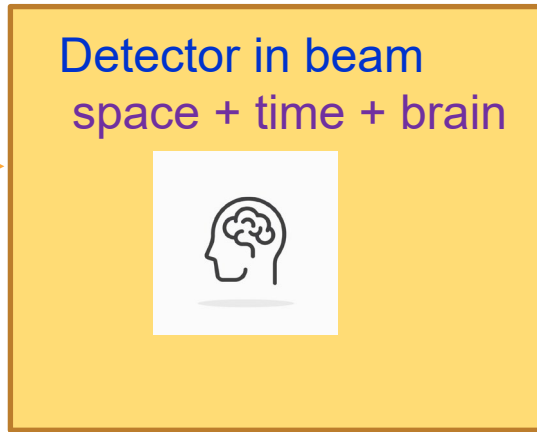


Need to move off detector processing onto the detector so not as much data is transported off detector

NOW



FUTURE



Use neuromorphic (brain inspired) computing

- Timing/Event driven data
Dynamic Vision Sensing
- Spiking Neural networks (SNN)
- Detector needs new sensor electronics/readout chip with SNN embedded

Can improve:

- time and space resolution – incorporating timing information directly
- Energy consumption



Conclusion

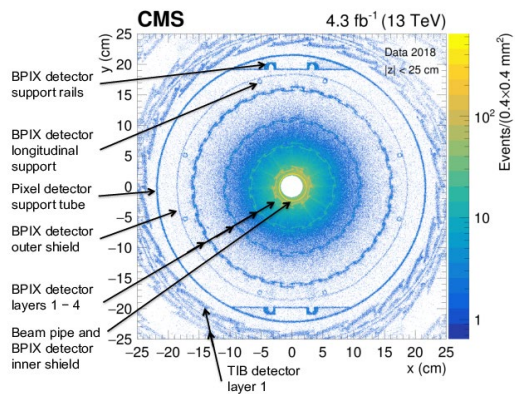
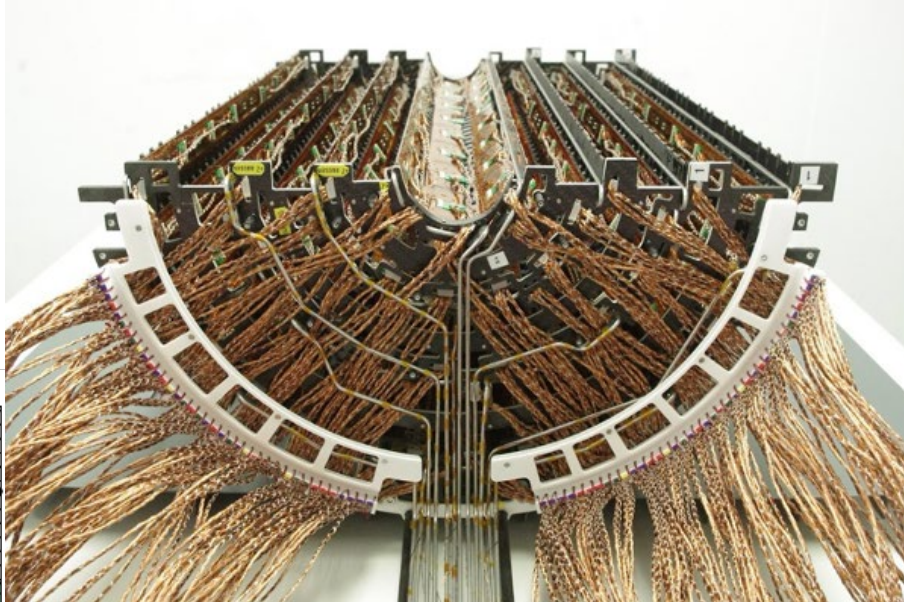
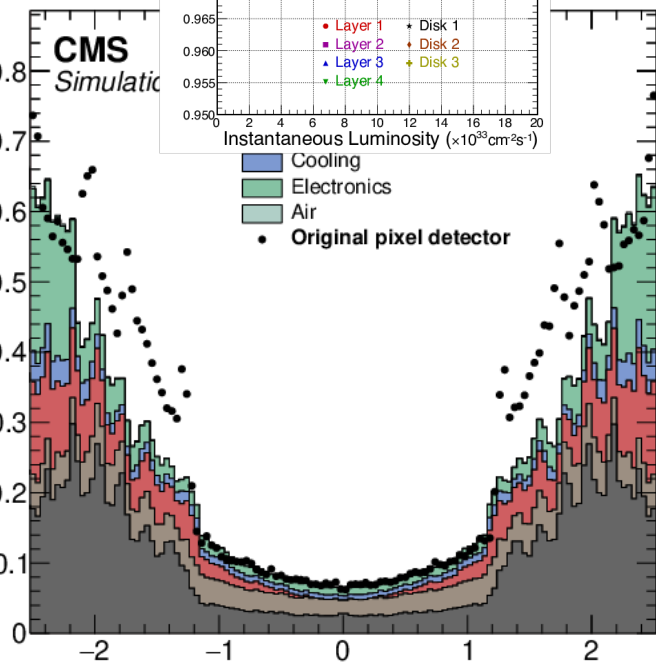
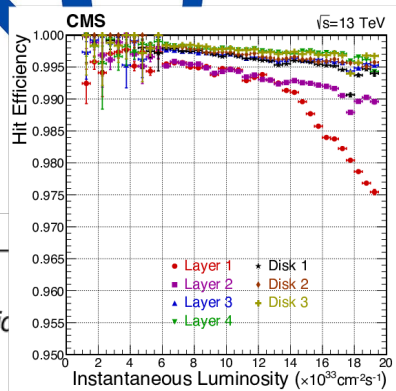
The CMS collaboration

- is taking more data and looking for Beyond Standard Model physics
- is building a new tracking detector for the HL-LHC

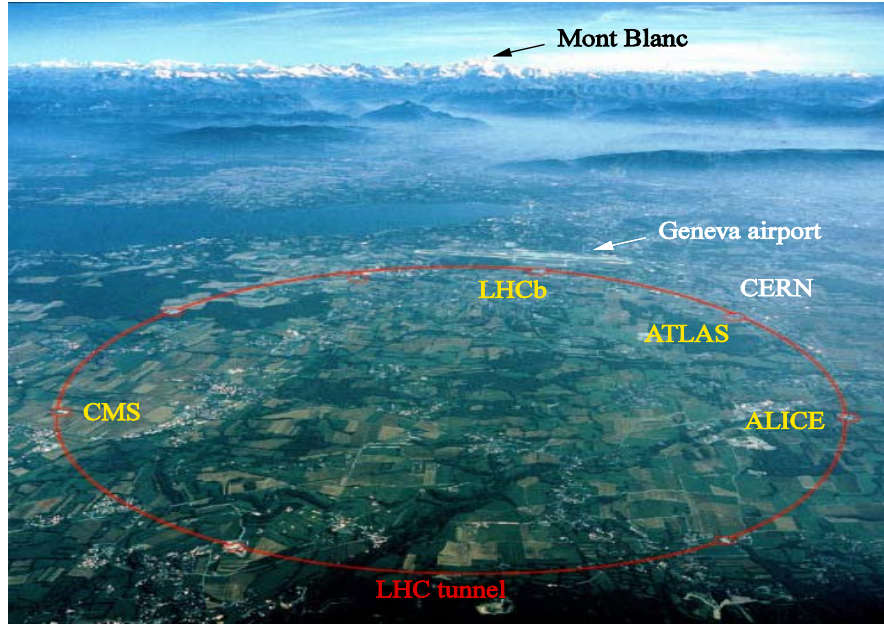
Challenges for building the detector include:

- Power
- Data Rates
- Materials and getting signals in and out

Explored challenges of e-link cables to carry signals



I am part of the CMS collaboration ~3000 physicists
 Operate a detector located at the Large Hadron Collider near
 Geneva, Switzerland



CMS DETECTOR

Total weight : 14,000 tonnes
 Overall length : 28.7 m
 Magnetic field : 3.8 T

STEEL RETURN YOKE
 12,500 tonnes

SILICON TRACKERS
 Microstrips ($80 \times 180 \mu\text{m}$) $\sim 200\text{m}^2 \sim 9.6\text{M}$ channels

SUPERCONDUCTING SOLENOID
 Niobium titanium coil carrying $\sim 18,000\text{A}$

MUON CHAMBERS
 Barrel: 250 Drift Tube, 480 R
 End-arms: 540 Drift Tube, 250 R

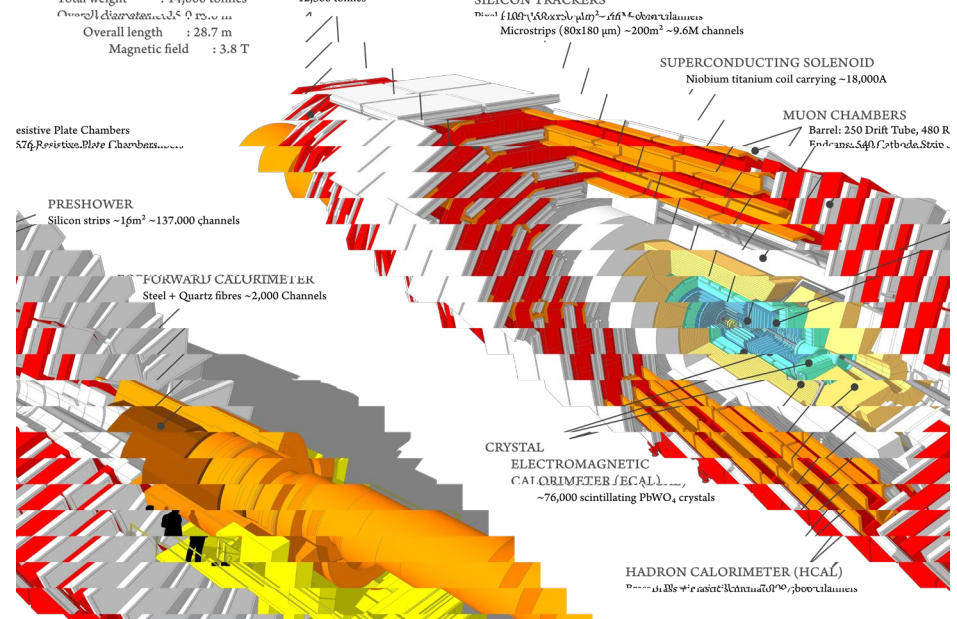
EXISTENTIVE PLATE CHAMBERS
 576 Resistive Plate Chambers

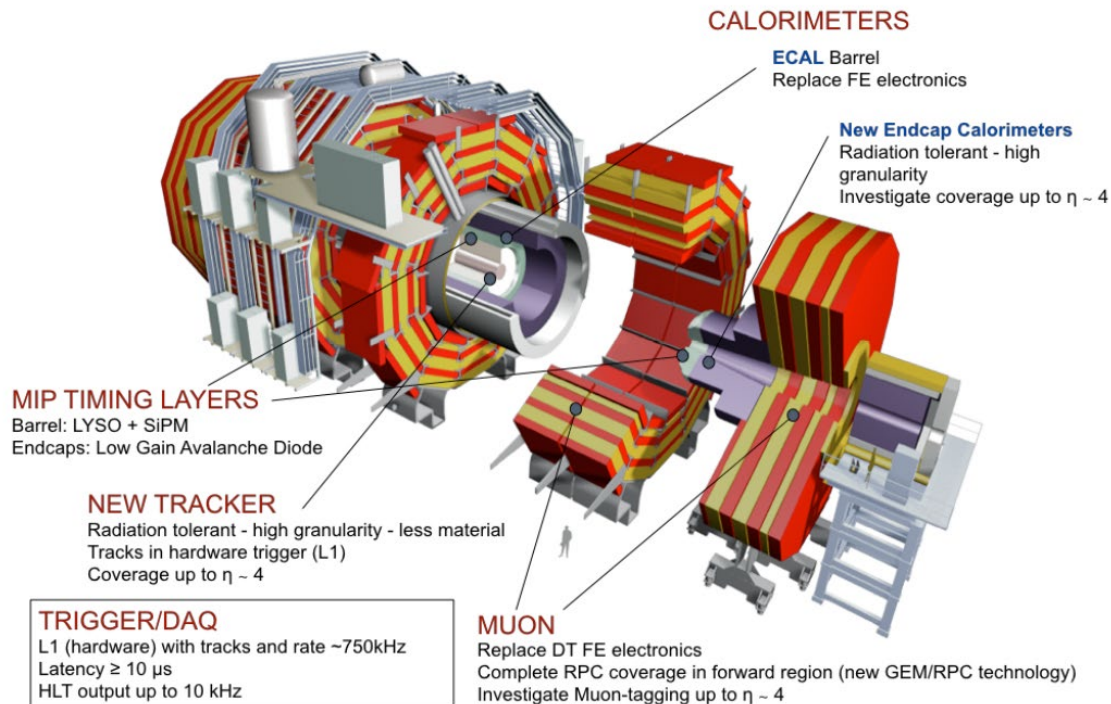
PRESHOWER
 Silicon strips $\sim 16\text{m}^2 \sim 137,000$ channels

FORWARD CALORIMETER
 Steel + Quartz fibres $\sim 2,000$ Channels

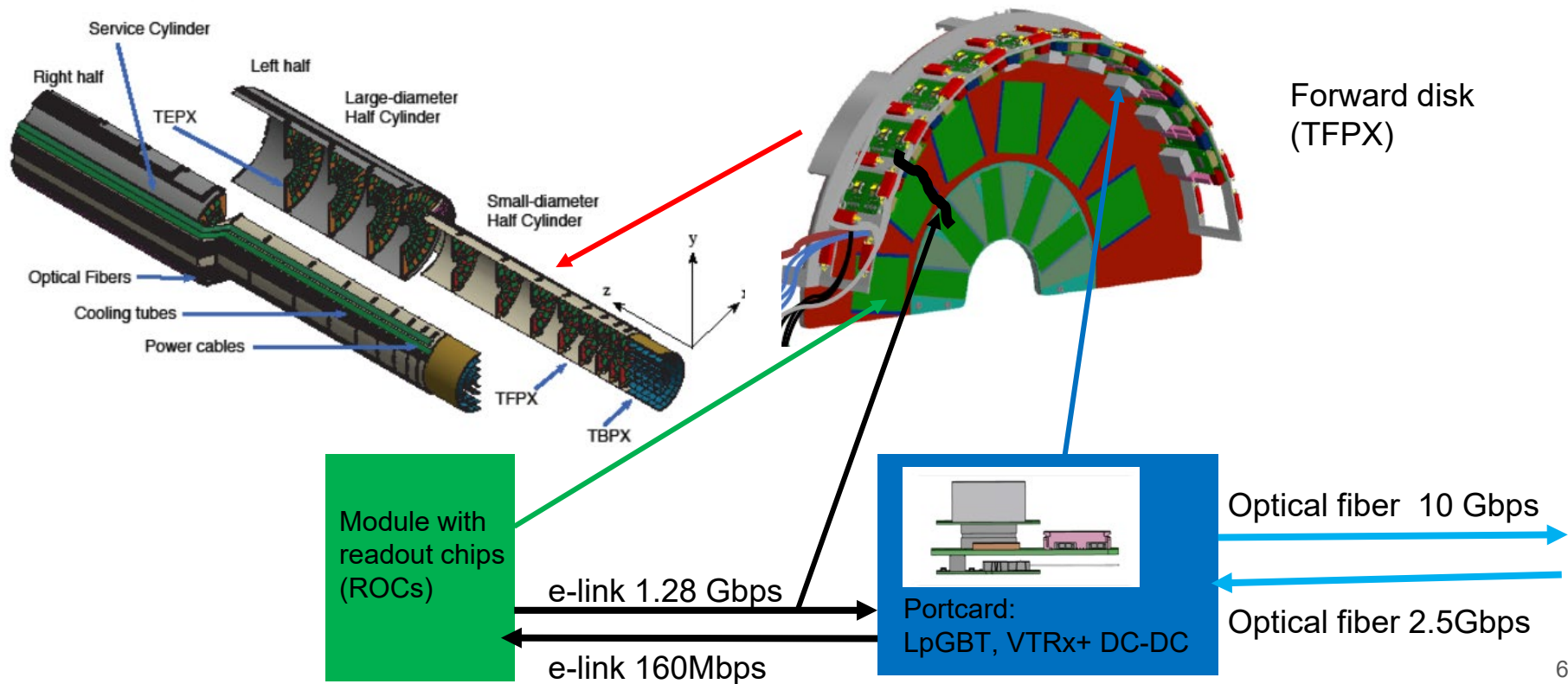
CRYSTAL
 ELECTROMAGNETIC
 CALORIMETER (ECAL)
 $\sim 76,000$ scintillating PbWO_4 crystals

HADRON CALORIMETER (HCAL)
 $\sim 1,136$ scintillating PbWO_4 crystals





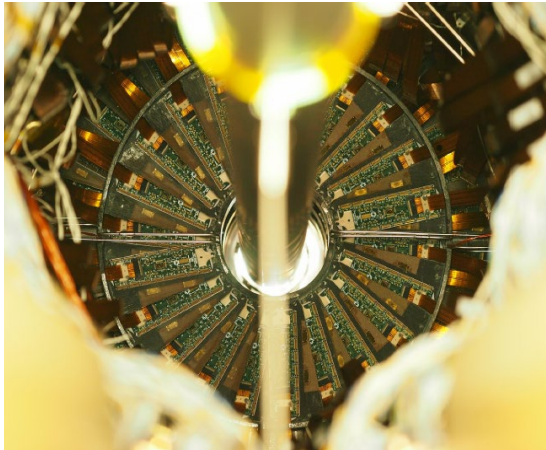
On Detector Readout Chain





Experimental Particle Physics and Detectors

Alice Bean
University of Kansas
Oct 20, 2022



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