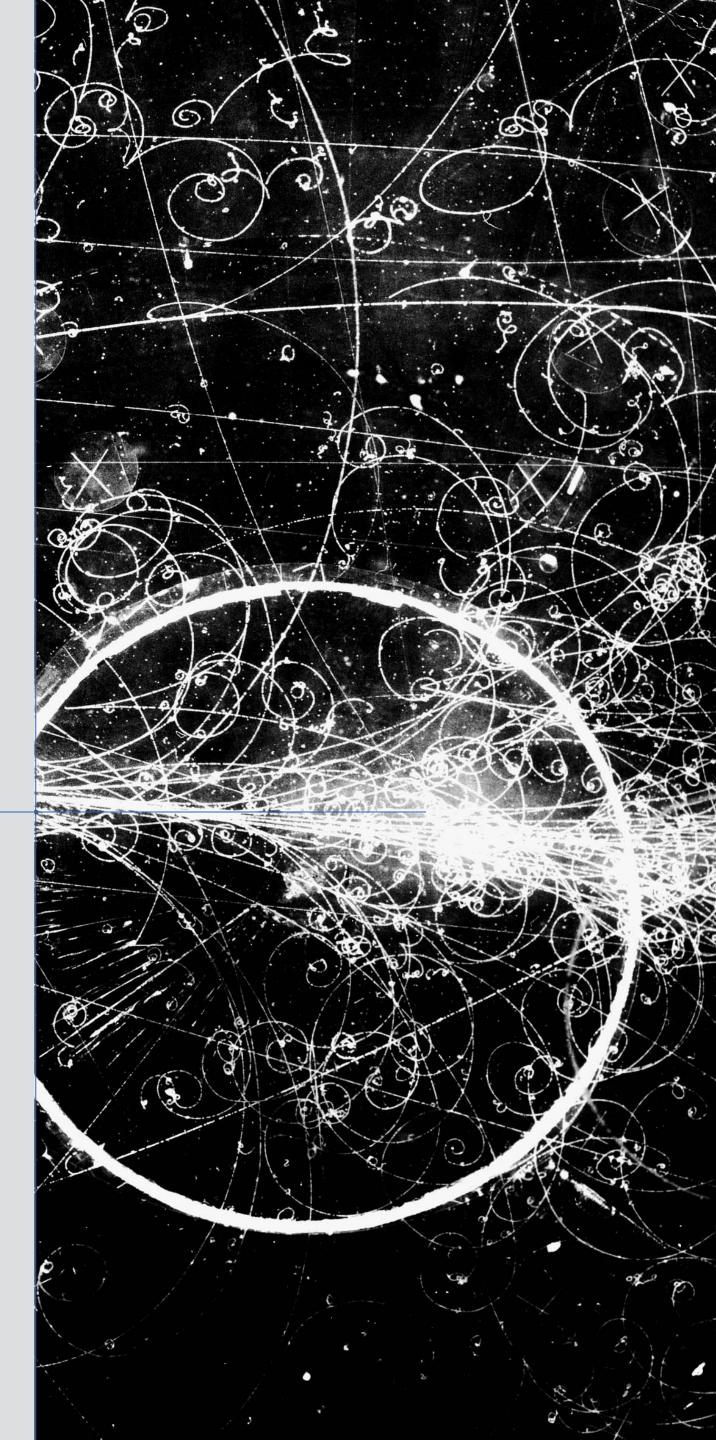
# (Introduction to) Data Acquisition

Advanced Graduate Lectures on practical Tools, Applications and Techniques in HEP June 6, 2024 Alessandro Thea

Rutherford Appleton Laboratory - PPD



Science and Technology Facilities Council



# Acknowledgements

#### Lecture inherited from Monika Wielers

colleagues!



#### • Many ideas, material also borrowed from Andrea Venturi, Francesca Pastore and other TDAQ





## Outline

- 1. Introduction
  - 1.1. What is DAQ?
  - 1.2. System architecture
- 2. Basic DAQ concepts
  - 2.1. Digitization, Latency
  - 2.2. Deadtime, Busy, Backpressure
  - 2.3. De-randomization
- 3. Scaling up
  - 3.1. Readout and Event Building
  - 3.2. Buses vs Network
- 4. DAQ challenges for large-scale experiments





## What is DAQ?

#### **D**ata **A**c**Q**uisition (**D**A**Q**) is

- the process of **sampling signals**
- that **measure** real world physical conditions
- and **converting** the resulting samples **into digital** numeric values that can be manipulated by a computer



### [Wikipedia]





## What is DAQ?

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- the process of **sampling signals**
- that **measure** real world physical conditions
- and **converting** the resulting samples **into digital** numeric values that can be manipulated by a computer Ingredients:
- **Sensors**: convert physical quantities to electrical signals
- Analog-to-digital converters: convert conditioned sensor signals to digital values
- Processing and storage elements

### [Wikipedia]





## What is DAQ?

#### DAQ is an **heterogeneous** field (a.k.a. dark arts)

• with boundaries not well defined

#### An **alchemy** of

- physics
- electronics
- computer science
- hacking
- networking
- experience

Where money and manpower matter as well



#### [Real life]







## DAQ duties

Gather data produced by detectors

• Readout

Form complete events

• Data Collection and Event Building

Possibly feed extra processing levels Store event data

• Data Logging

Manage operations

Control, Configuration, Monitoring









## Interlude: data vs *interesting* data

Interesting physics data typically a small fraction of sampled signals

#### really, Really, REALLY SMALL

Logging all recorded data is unpractical (and costly)

• sometimes technically unfeasible

Online data reduction before logging becomes imperative

#### That's the job of the **Trigger**!

- DAQ and Trigger deeply entwined
  - often referred as TDAQ

#### All about the Trigger systems in the next lecture Dr. Will Panduro

• All you wanted to know about trigger and never dared to ask!





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# Trigger in a nutshell

Selects interesting events **AND** rejects boring ones, *in real time* 

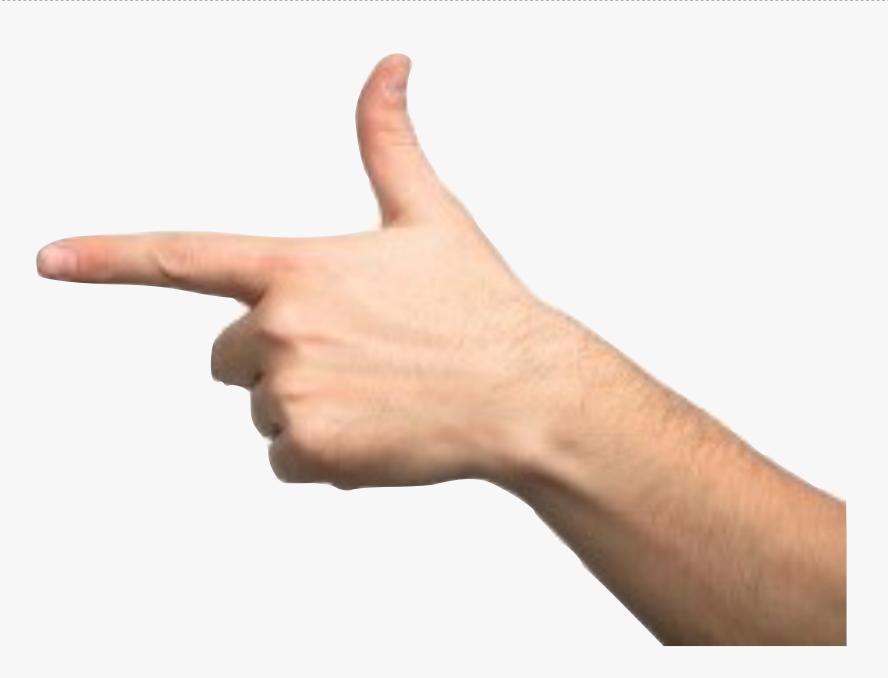
- Selective: efficient for "signal" and resistant to "background"
- Simple and robust: Must be predictable at all times!
- **Fast:** Late is no better than never

With minimal *controlled* **latency** 

• time it takes to form and distribute its decision

The implementation of "Trigger" has significantly evolved in the past decades

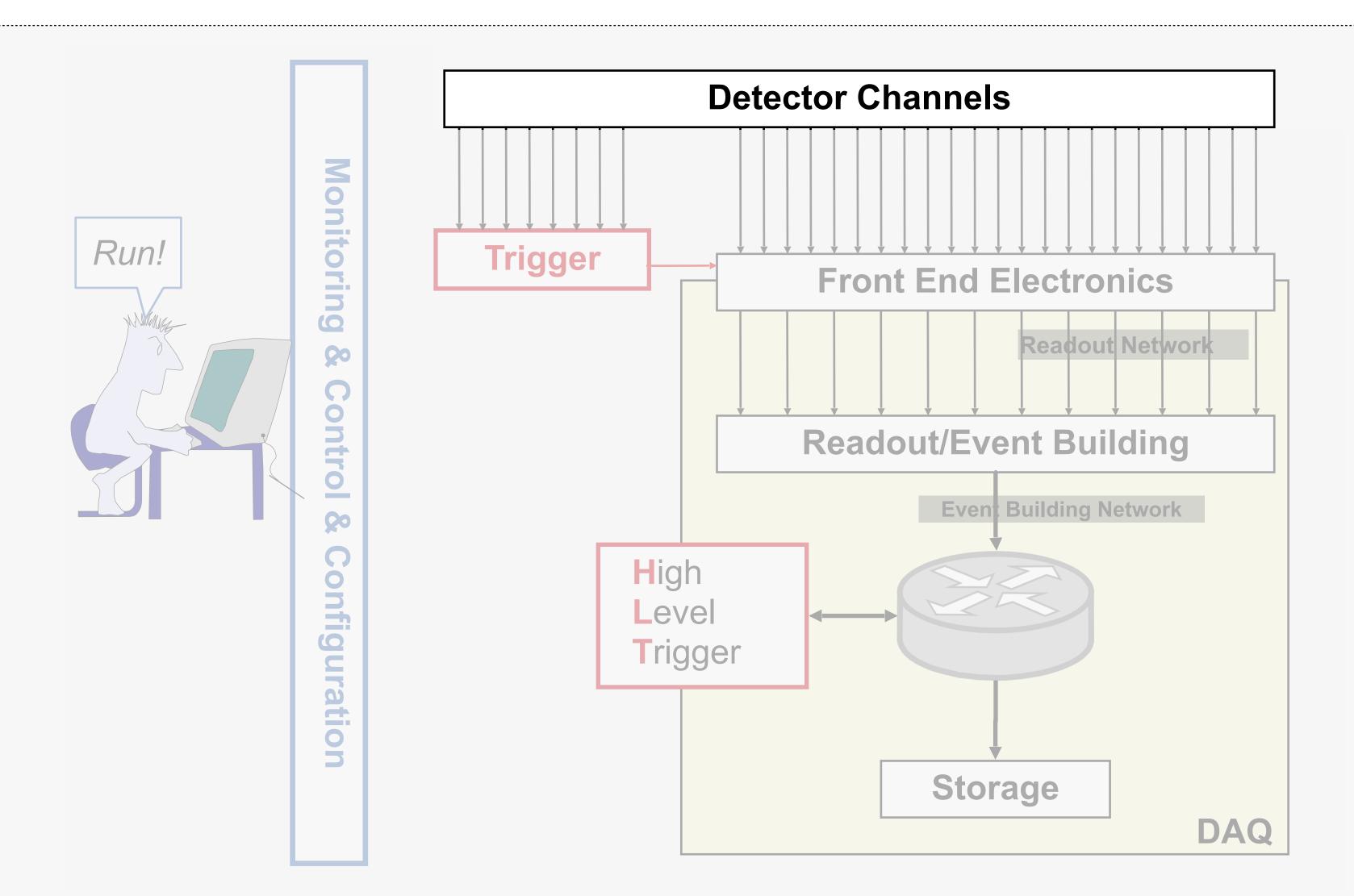
- **LEP**: trigger the sampling of (slow) detector
- (*High Level Trigger HLT*)



• LHC: trigger the readout of on-detector buffers (*Level-1*) or trigger logging to permanent storage

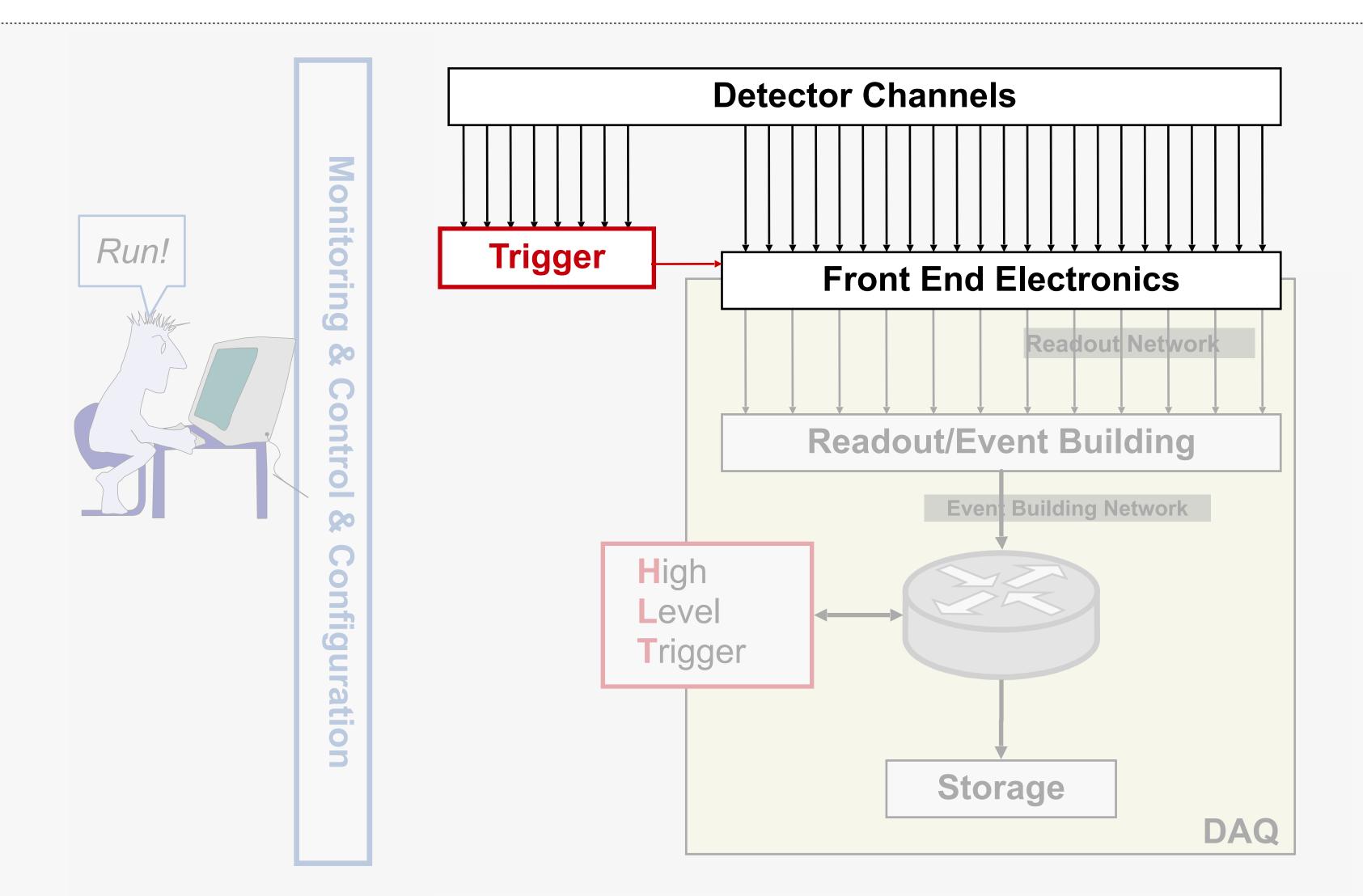








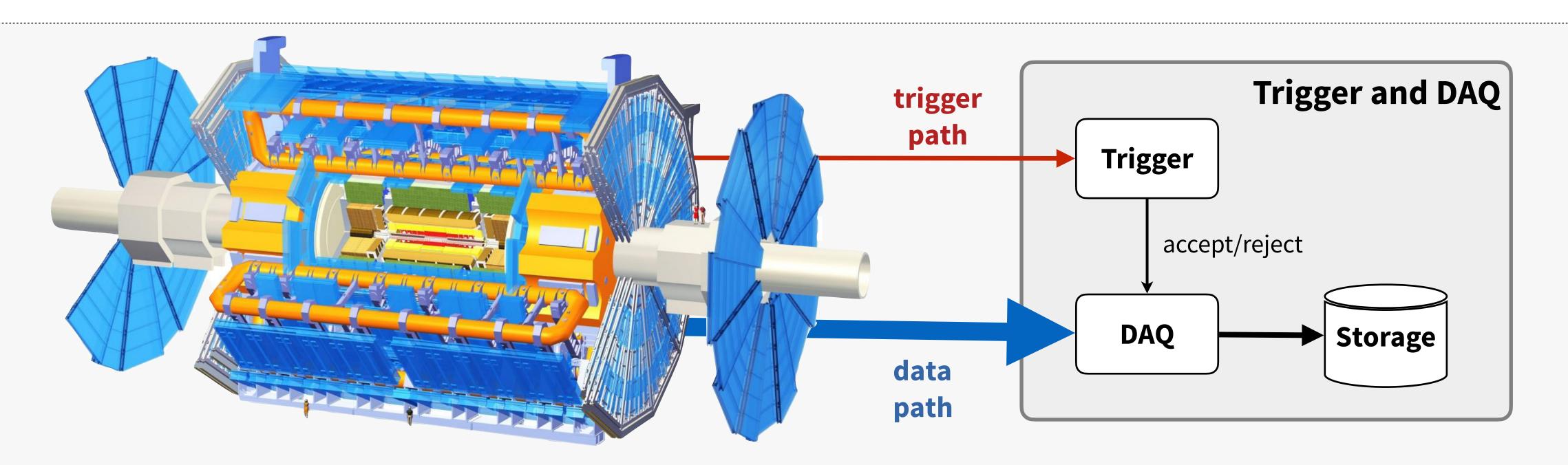








## From detector to T-DAQ



#### **Trigger path**

- From specific detectors to trigger logic
- Continuous streaming of trigger data
- Dedicated connections

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#### Data path

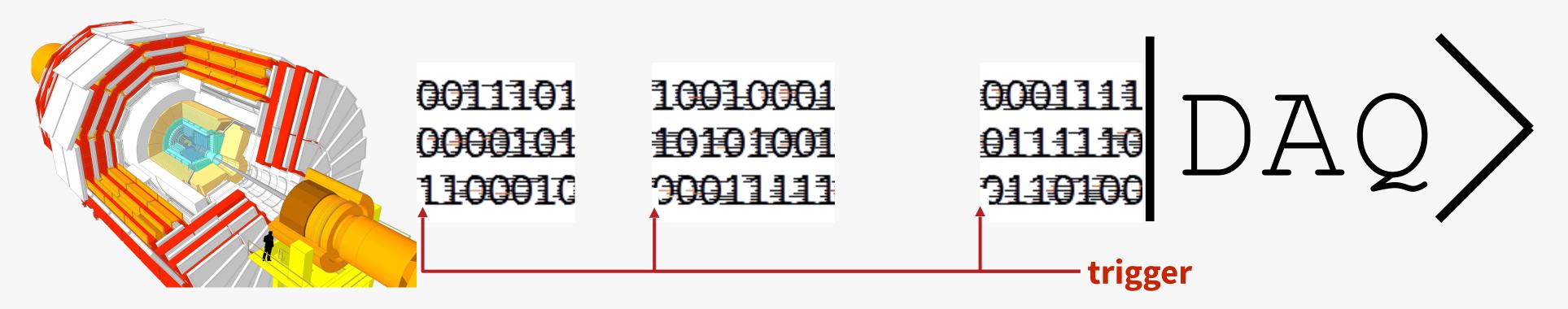
- From all the detectors to readout
- Transmission on positive trigger decision





## Readout: Triggered vs Streaming

**Triggered**: data is readout from detector only when a trigger signal is raised





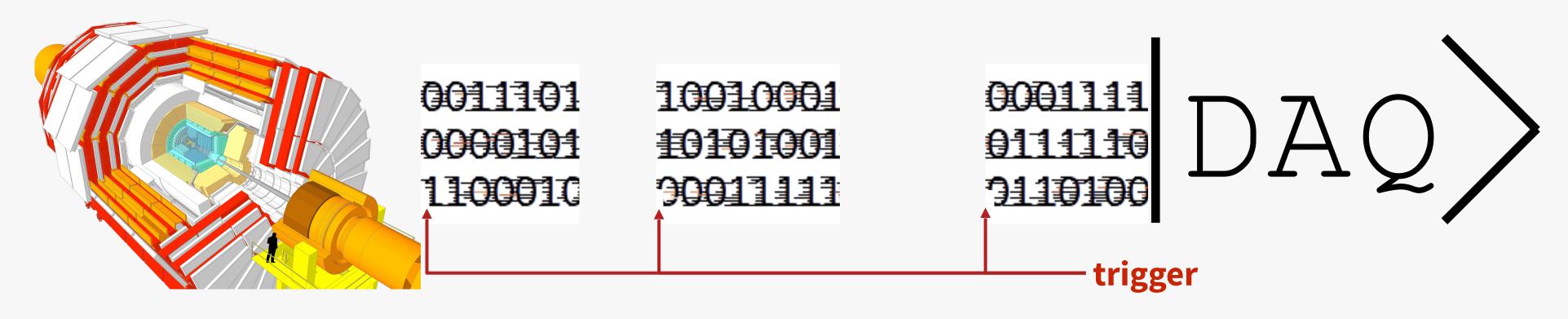






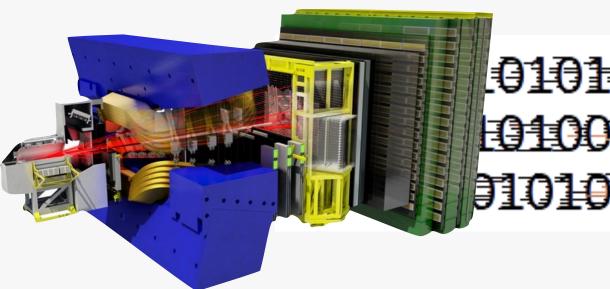
## Readout: Triggered vs Streaming

**Triggered**: data is readout from detector only when a trigger signal is raised



**Streaming**: detector pushes all its data and the downstream DAQ must keep the pace

data reduction still takes place, but post readout



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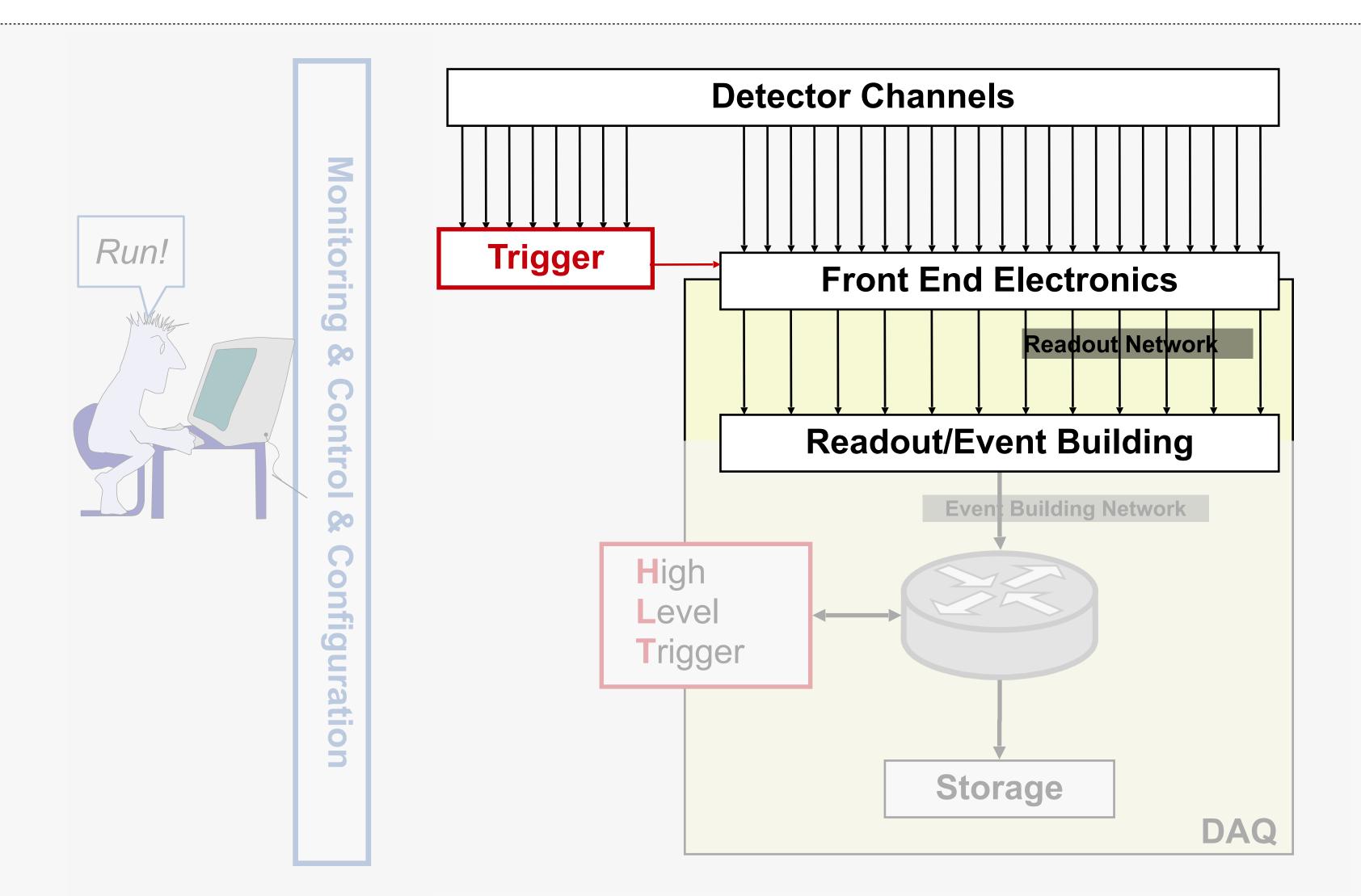






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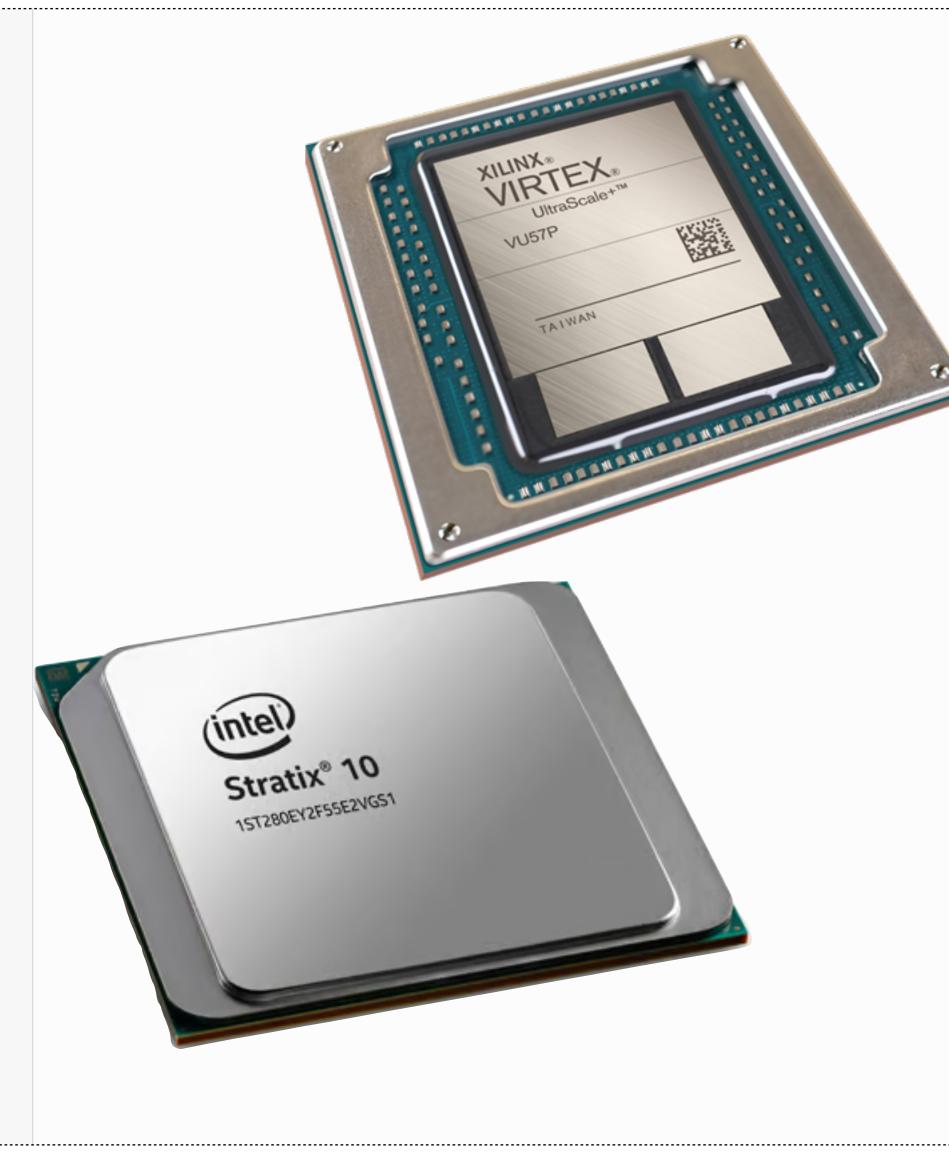








## Field Programmable Gate Arrays



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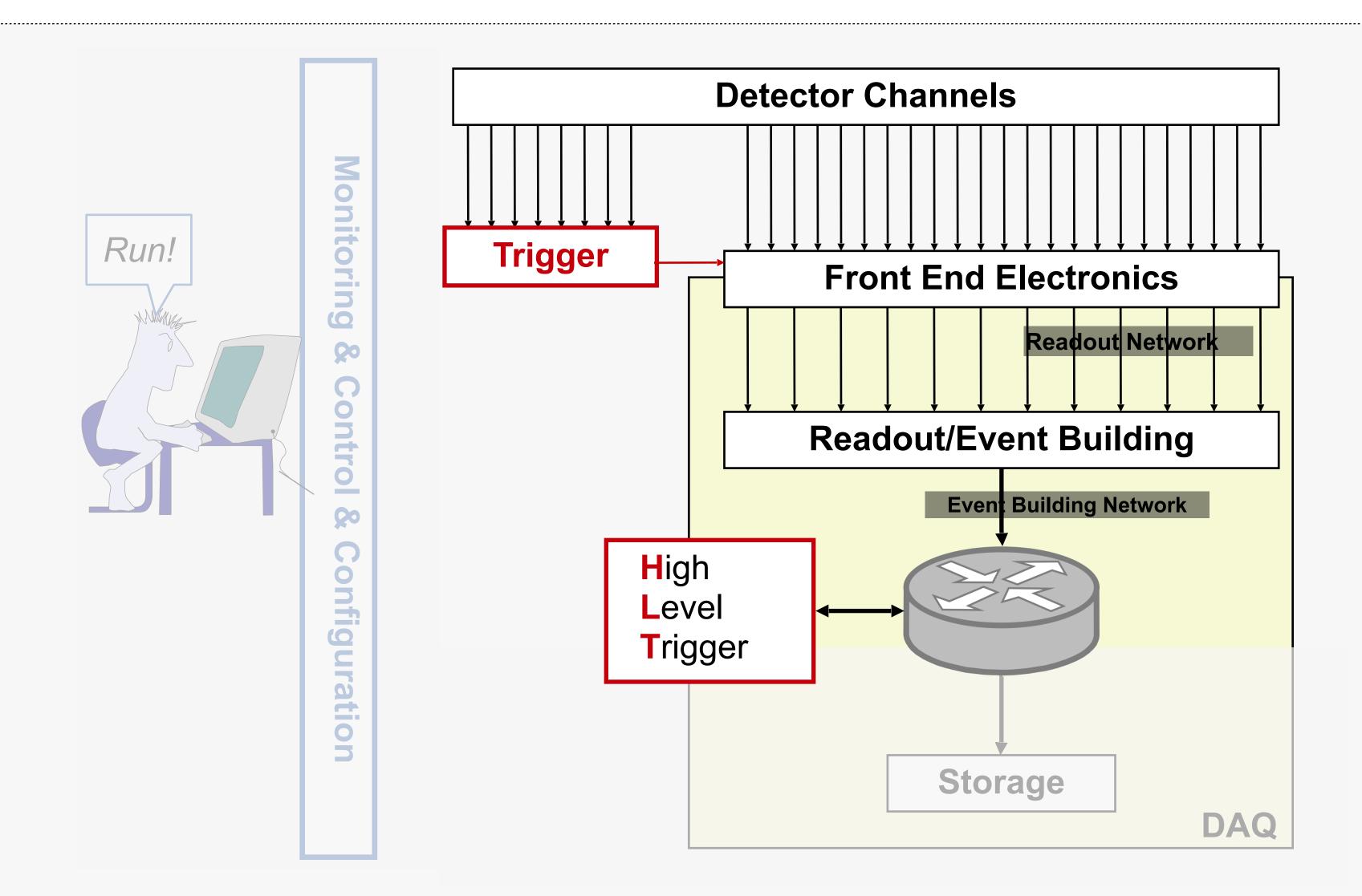
#### FPGAs have become becoming TDAQ's bread & butter

• Signal processing, data formatting, natively parallel tasks (e.g. pattern recognition), machine learning, ...

> Covered in yesterday's FPGA Programming Lecture -Dr. Kristian Harder

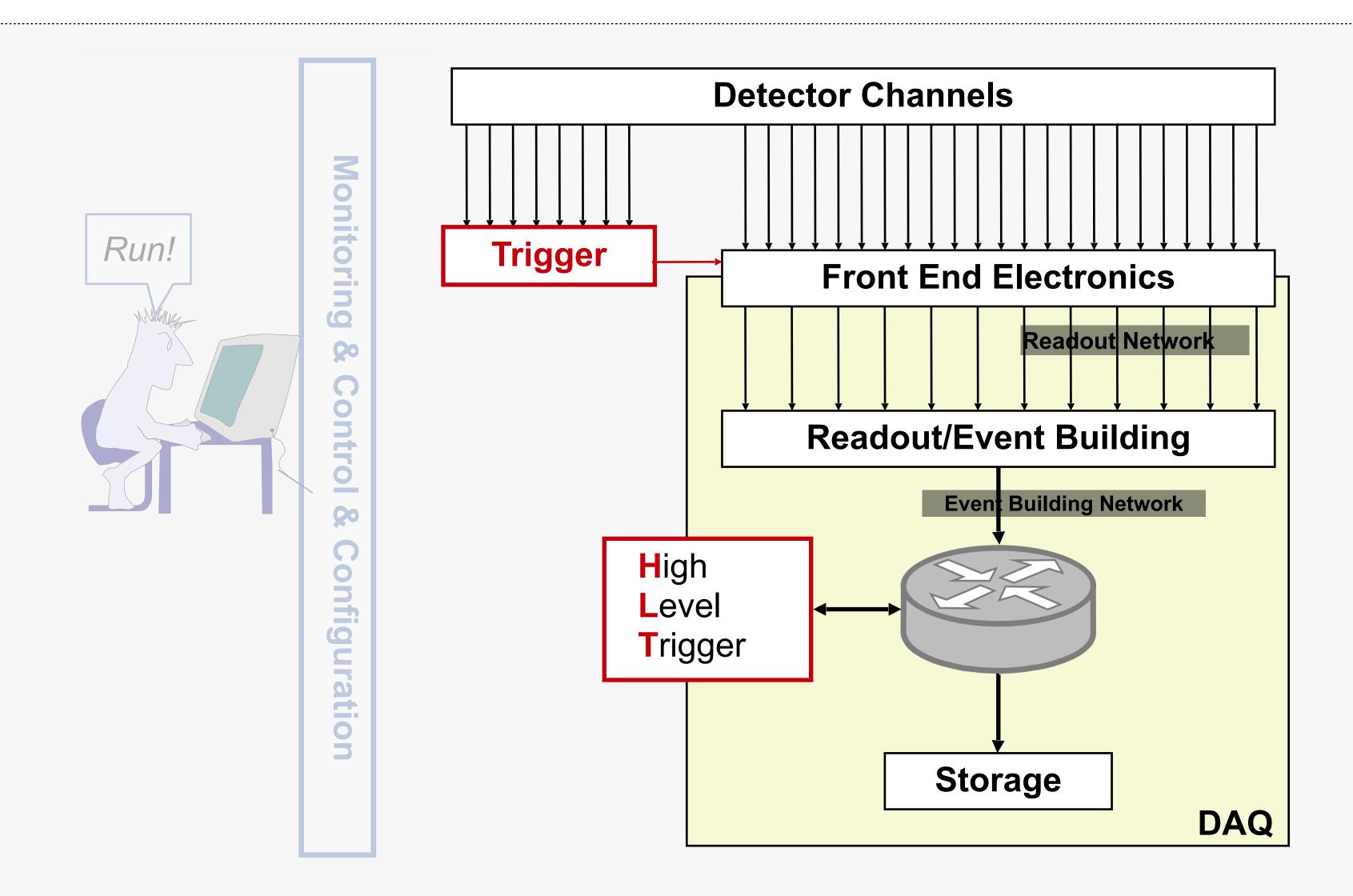






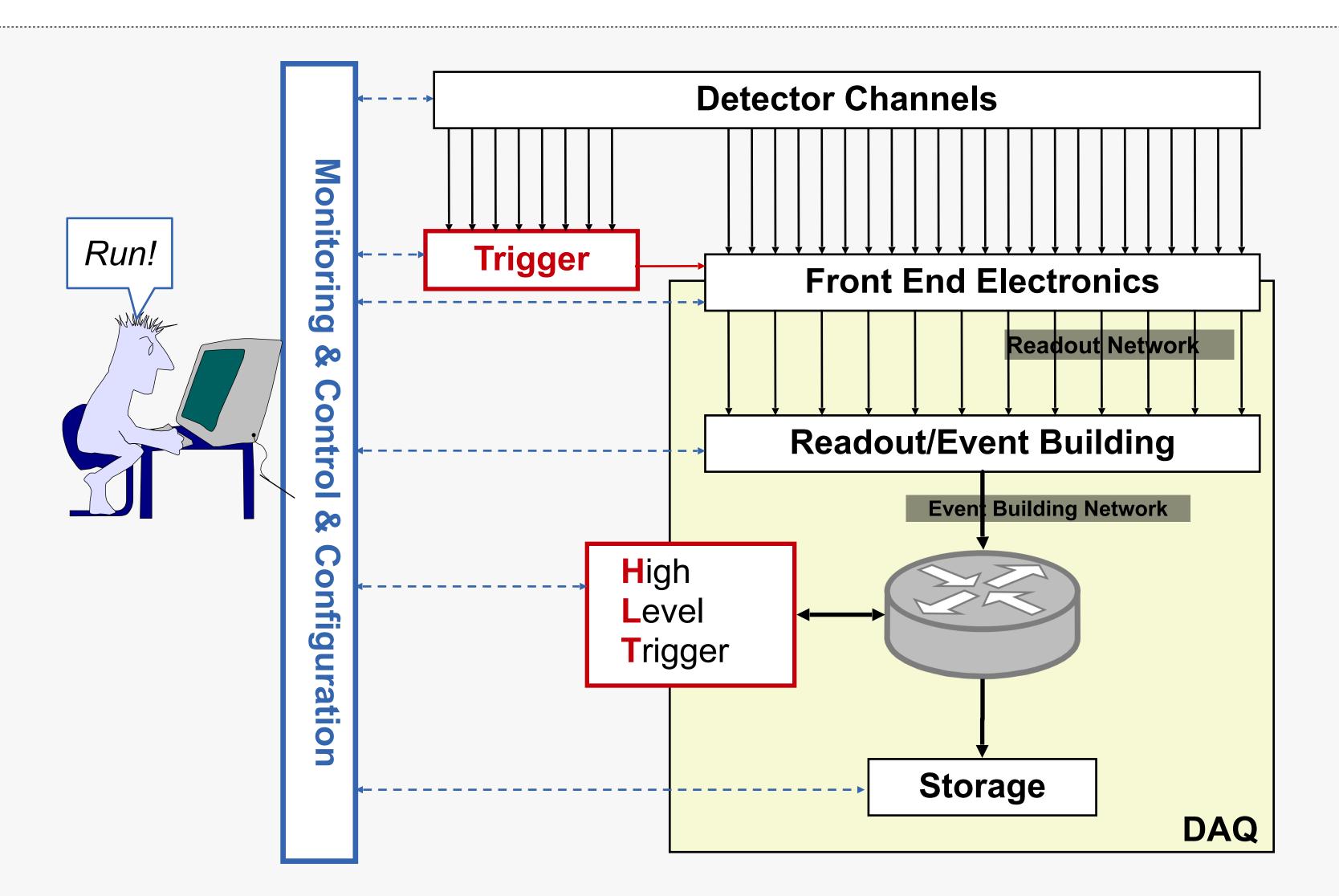








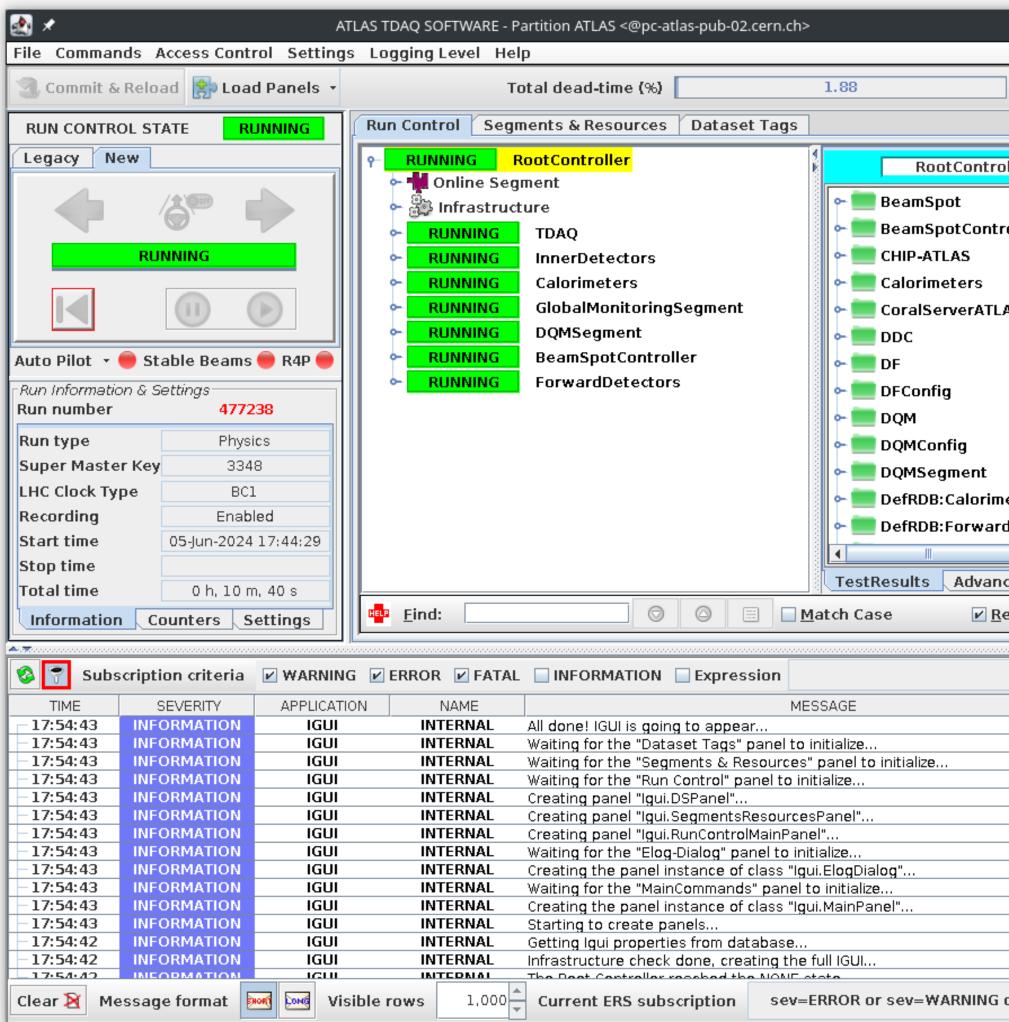








# The glue of your experiment



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#### Configuration

 Ensemble of detectors, trigger and DAQ parameters defining the system behaviour during data taking

#### Control

- Orchestrate applications participating to data taking
- Via distributed Finite State MachineMonitoring
  - Of data taking operations
    - What is going on?
    - What happened?
    - When?
  - Where?





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4. DAQ Challenges at the LHC



#### with a toy model

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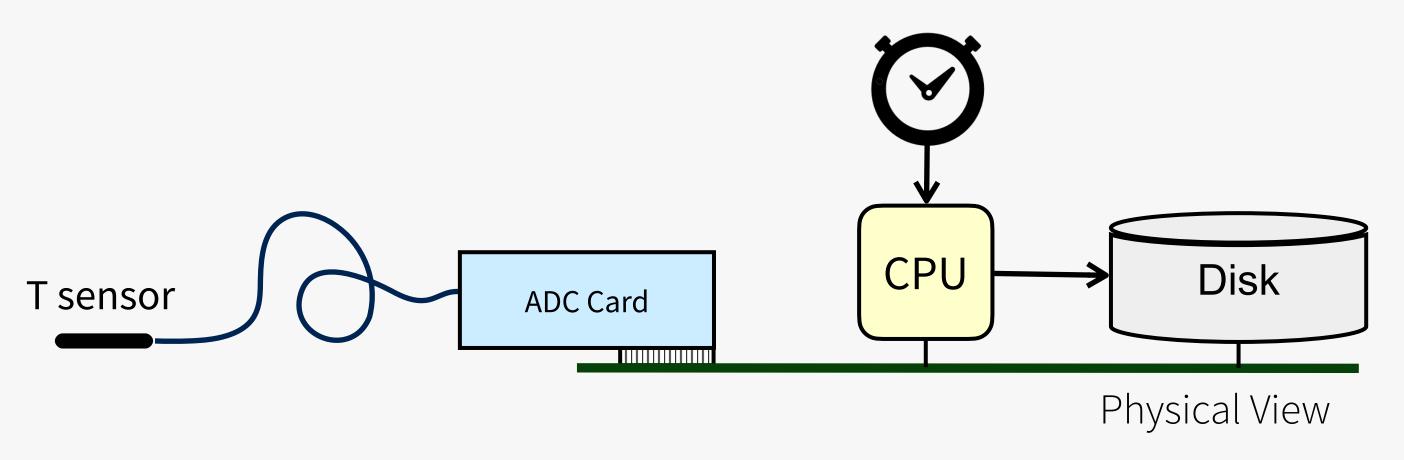
## Basic DAQ: periodic trigger

Eg: measure temperature at a fixed frequency

Clock triggered

ADC performs analog to digital conversion, digitization (our front-end electronics)

- Encoding analog value into binary representation CPU does
- Readout, Processing, Storage









## Basic DAQ: periodic trigger

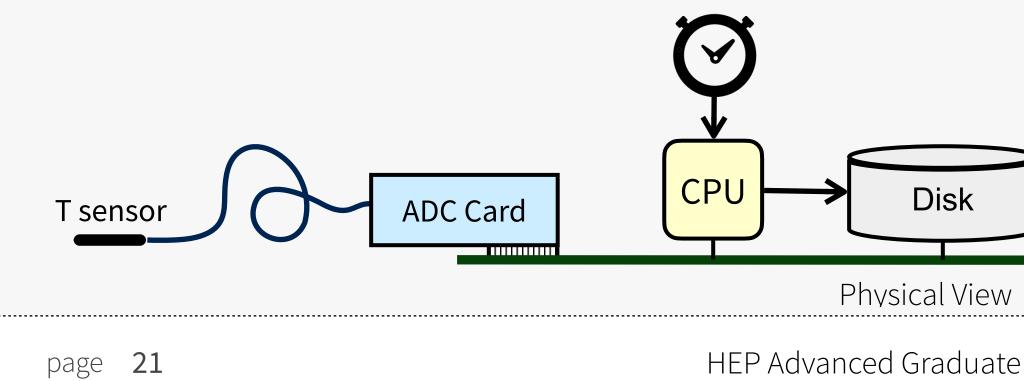
System clearly limited by the time  $\tau$  to process an "event"

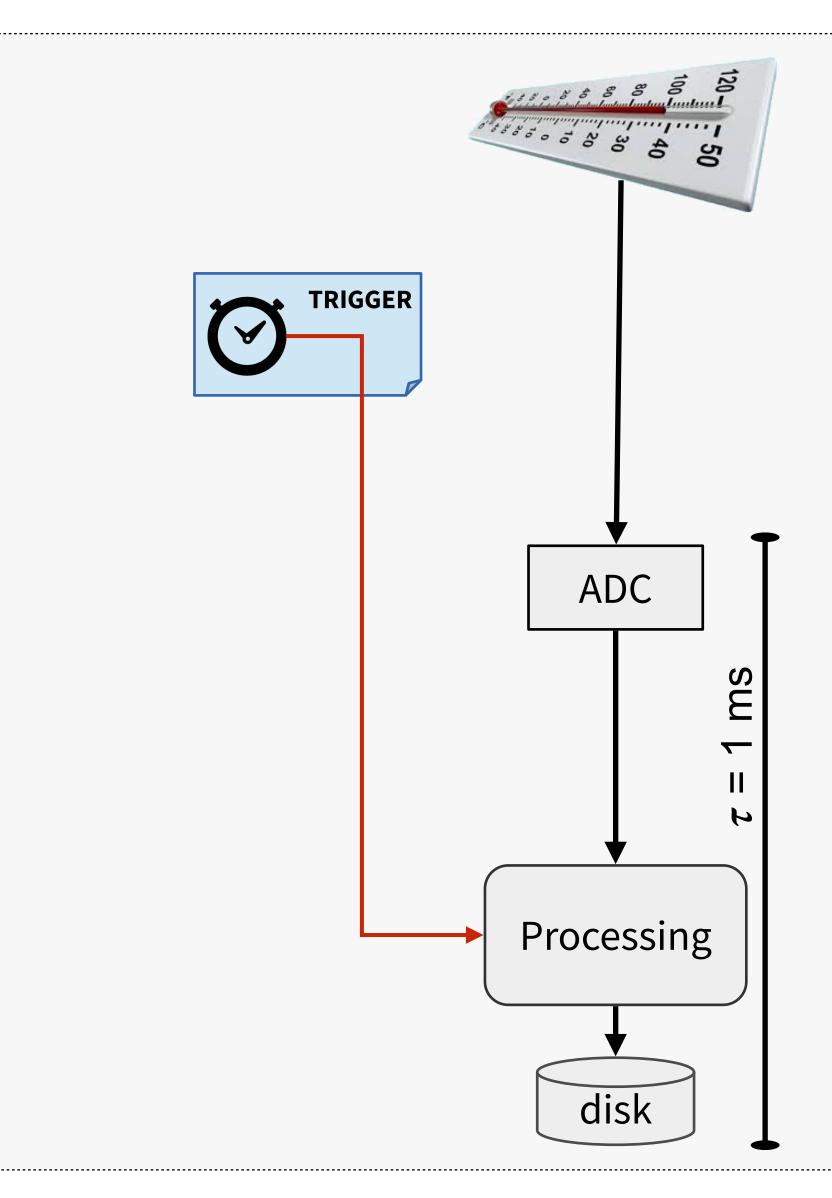
• ADC conversion + CPU processing + Storage

The DAQ maximum sustainable

rate is simply the inverse of  $\tau$ , e.g.:

• E.g.:  $\tau = 1 \text{ ms } R = 1/\tau = 1 \text{ kHz}$ 





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#### Events asynchronous and unpredictable

• E.g.: beta decay studies

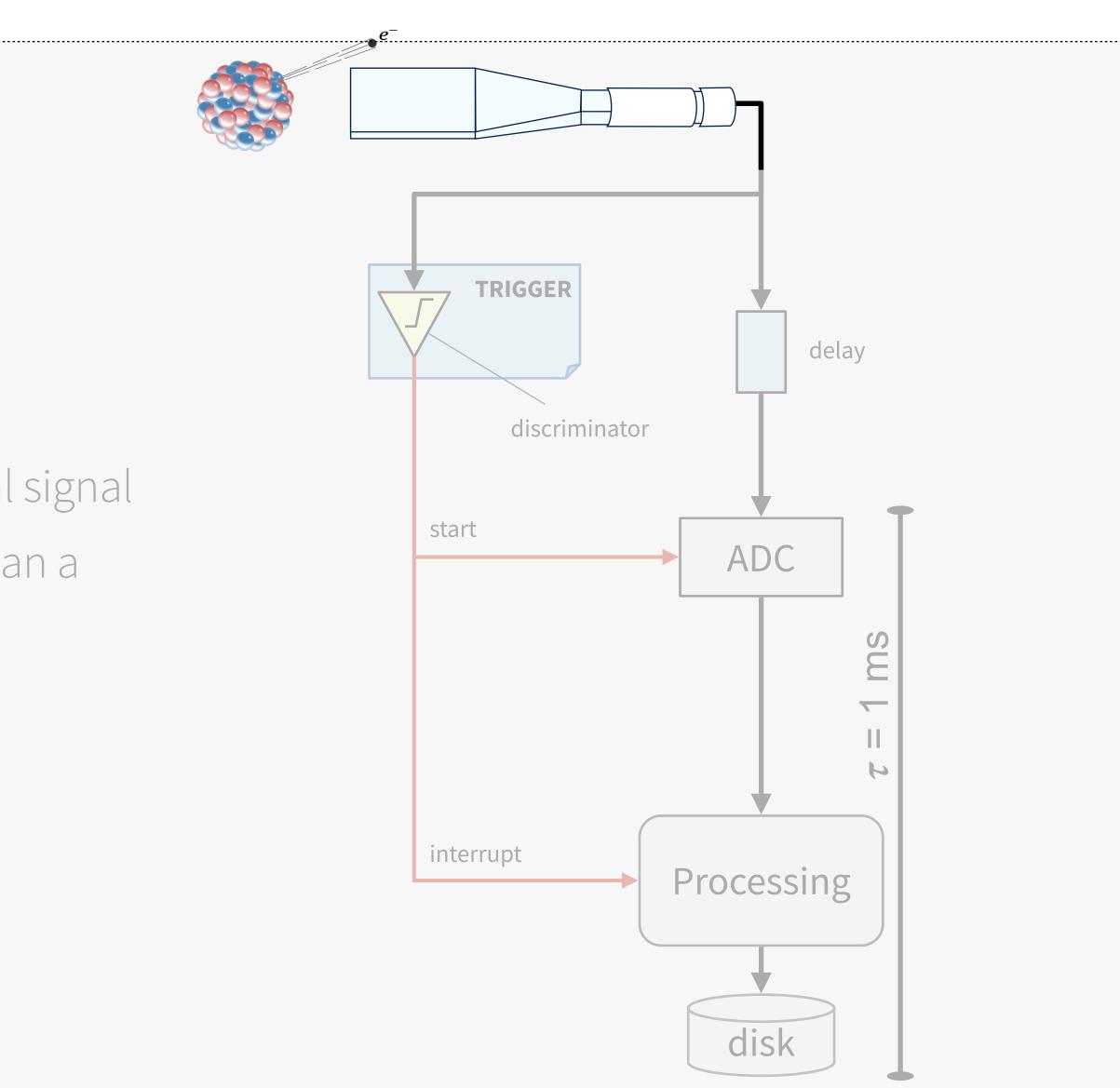
A physics trigger is needed

• **Discriminator**: generates an output digital signal if amplitude of the input pulse is greater than a given threshold

NB: delay introduced to compensate for the

#### trigger latency

• Signal split in trigger and data paths



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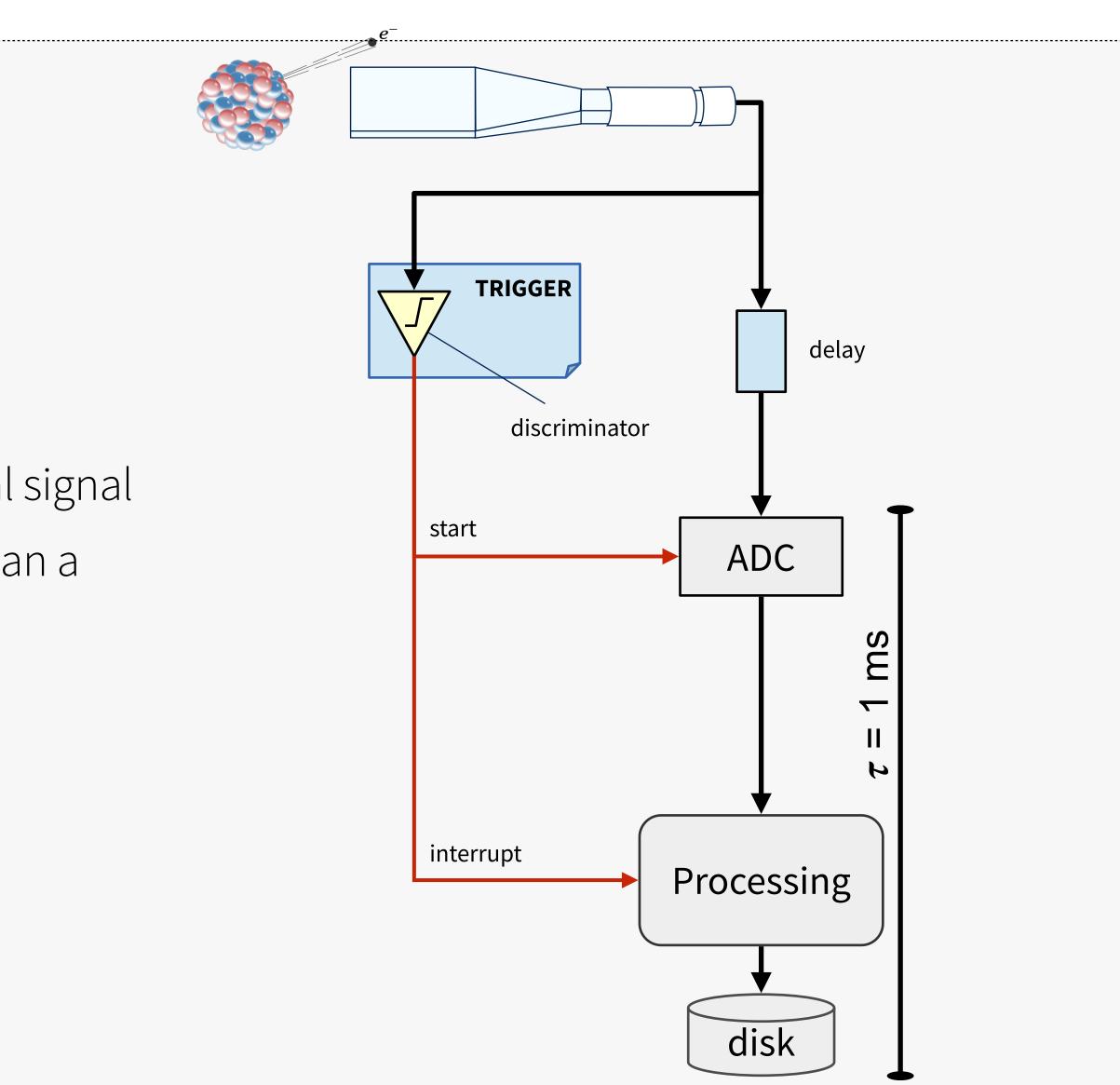
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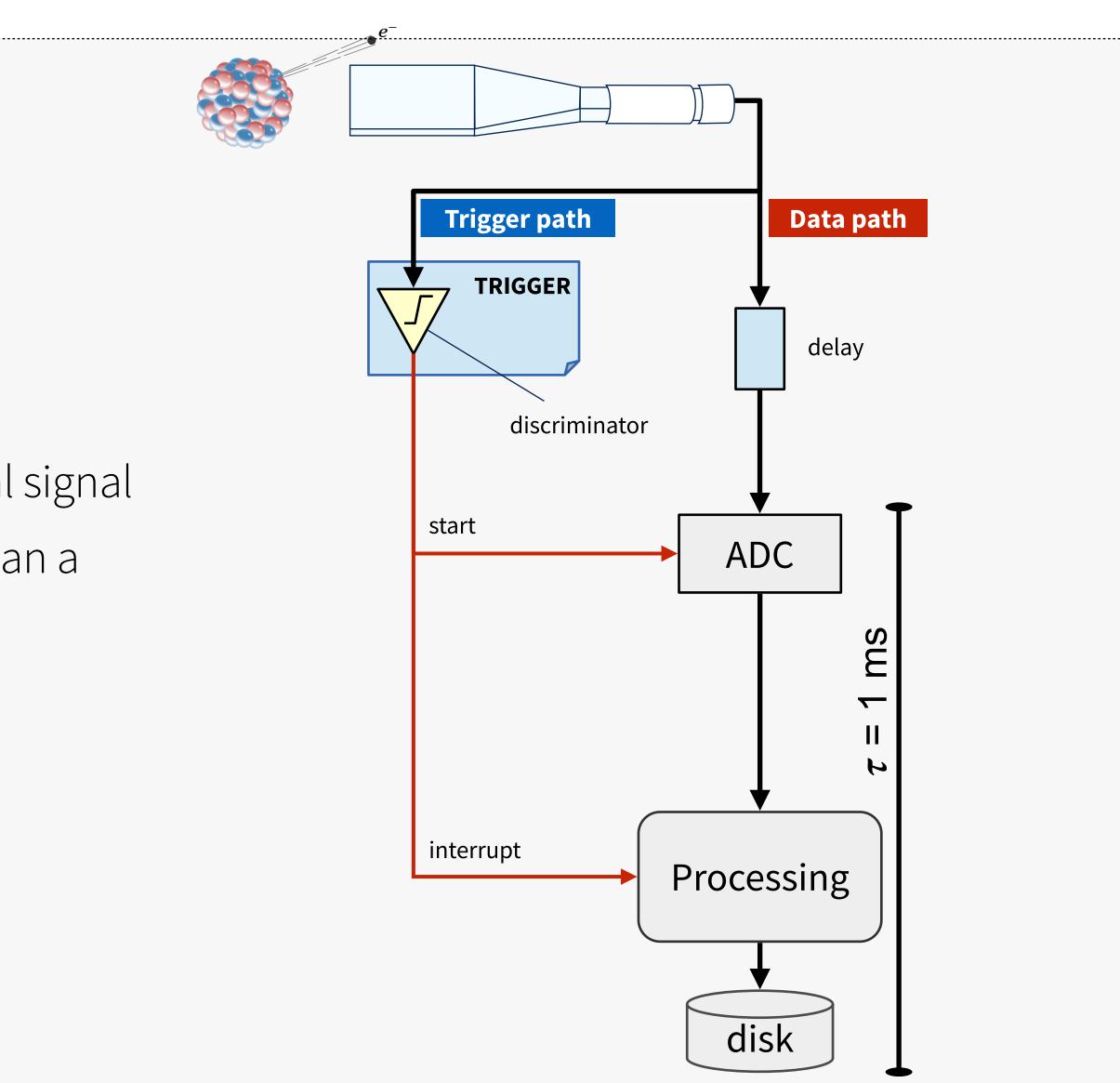
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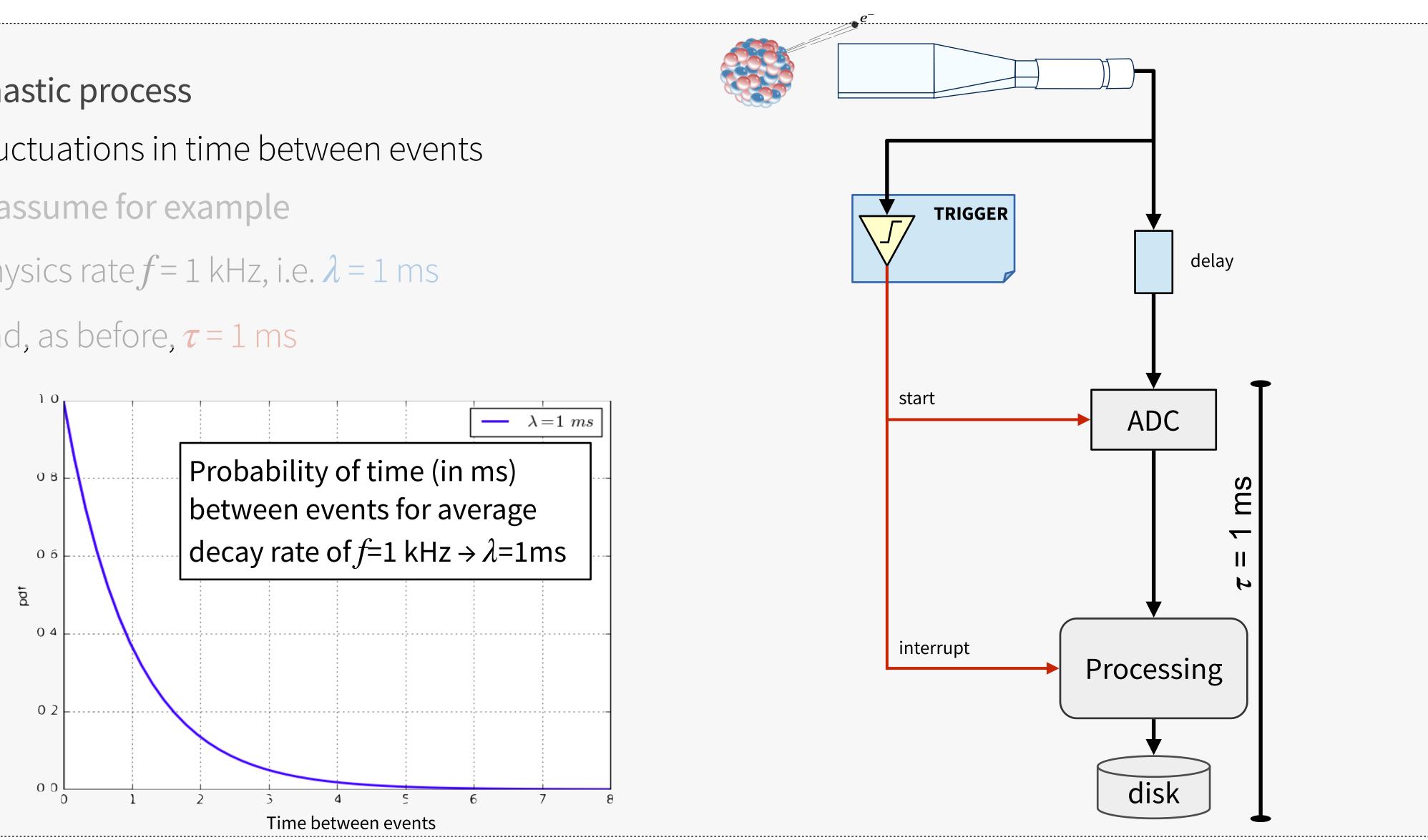
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#### Stochastic process

- Fluctuations in time between events Let's assume for example
- physics rate f = 1 kHz, i.e.  $\lambda = 1$  ms
- and, as before,  $\tau = 1 \text{ ms}$



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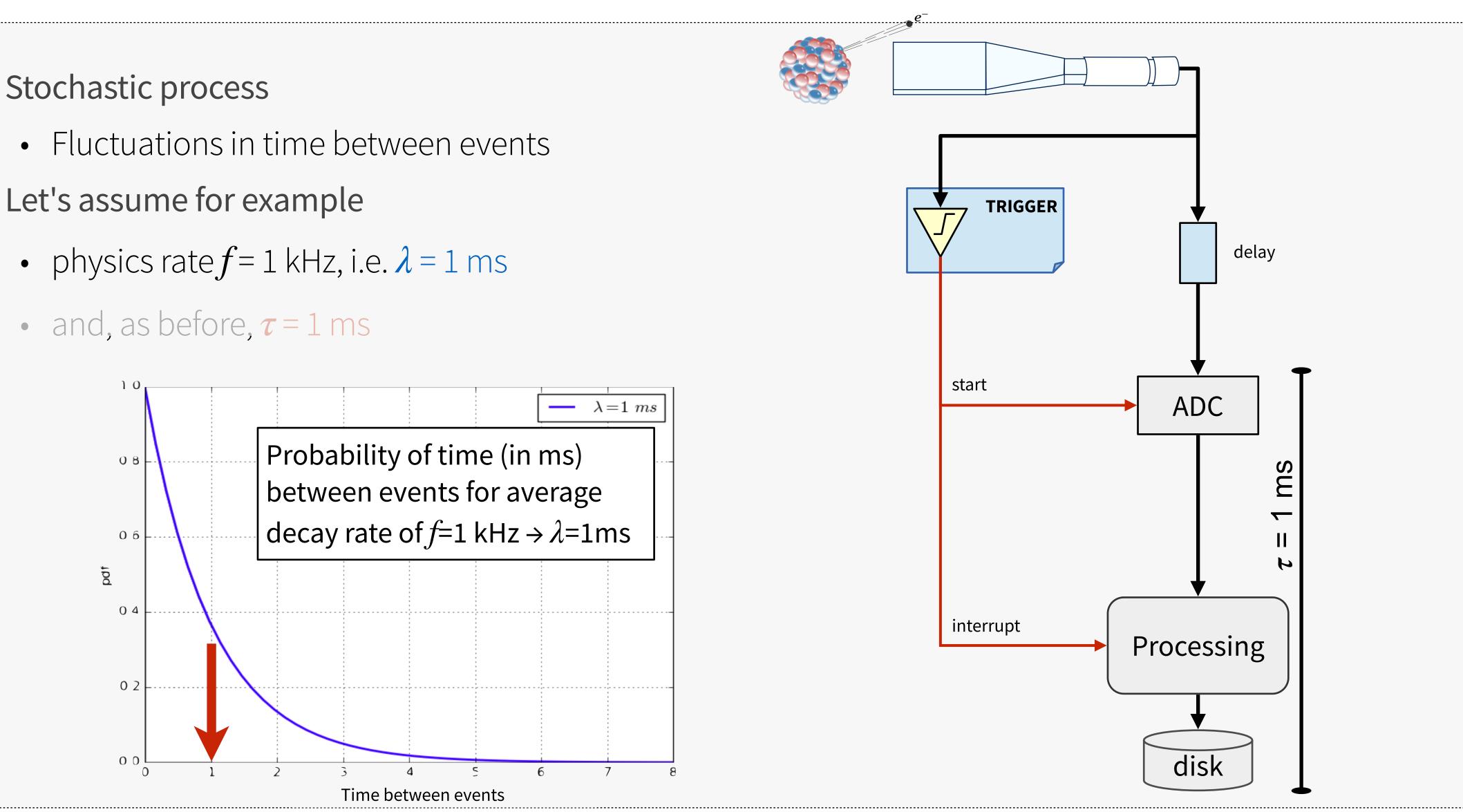


#### Stochastic process

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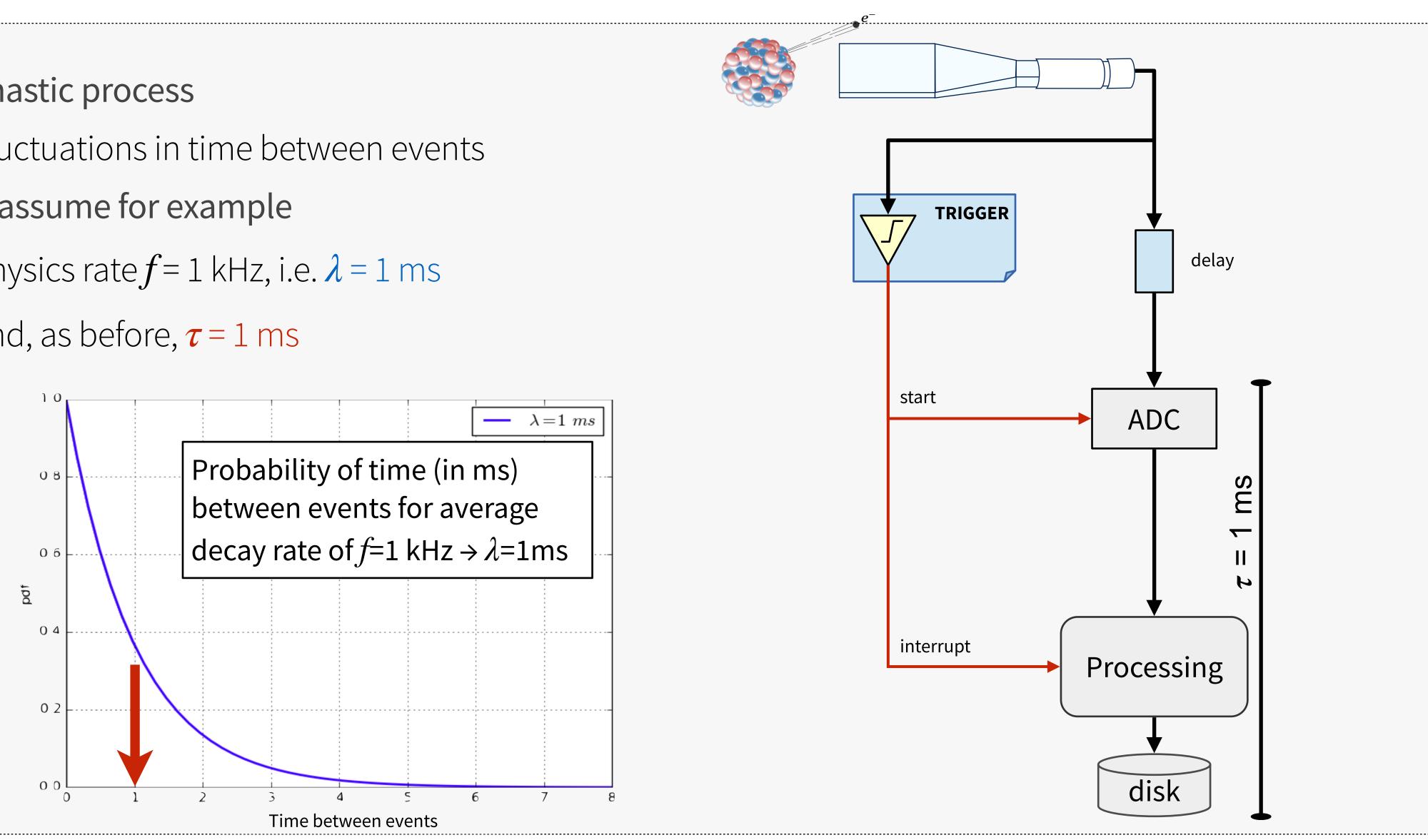


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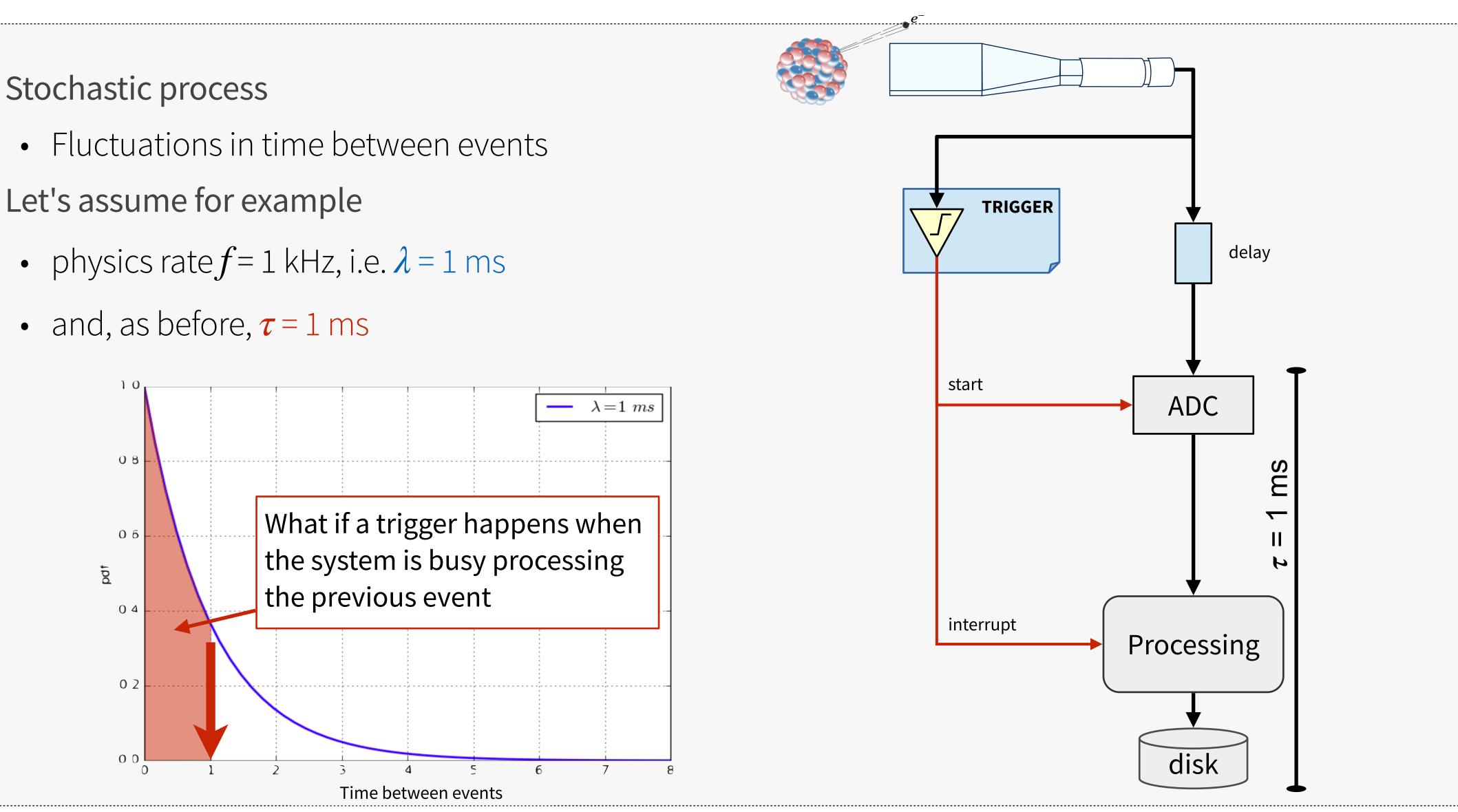
#### Stochastic process

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• Fluctuations in time between events

• and, as before,  $\tau = 1 \text{ ms}$ 



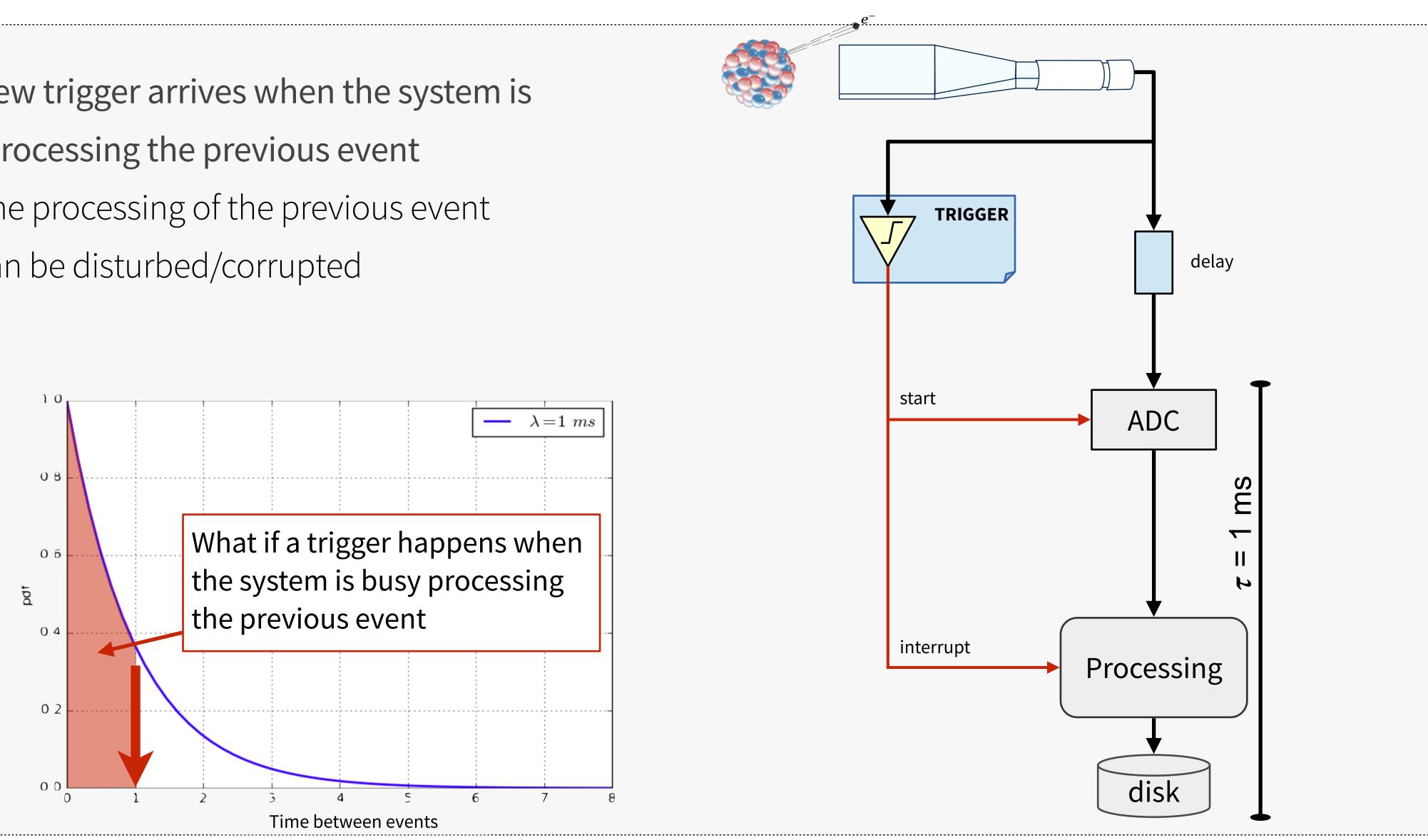




## The system is still processing

If a new trigger arrives when the system is still processing the previous event

• The processing of the previous event can be disturbed/corrupted







# Thinking...

#### For stochastic processes, our trigger and daq system needs to be able to:

- Determine if there is an "event" (**trigger**)
- Process and store the data from the event (daq)
- Have a feedback mechanism, to know if the data processing pipeline is free to process a new event:

#### busy logic









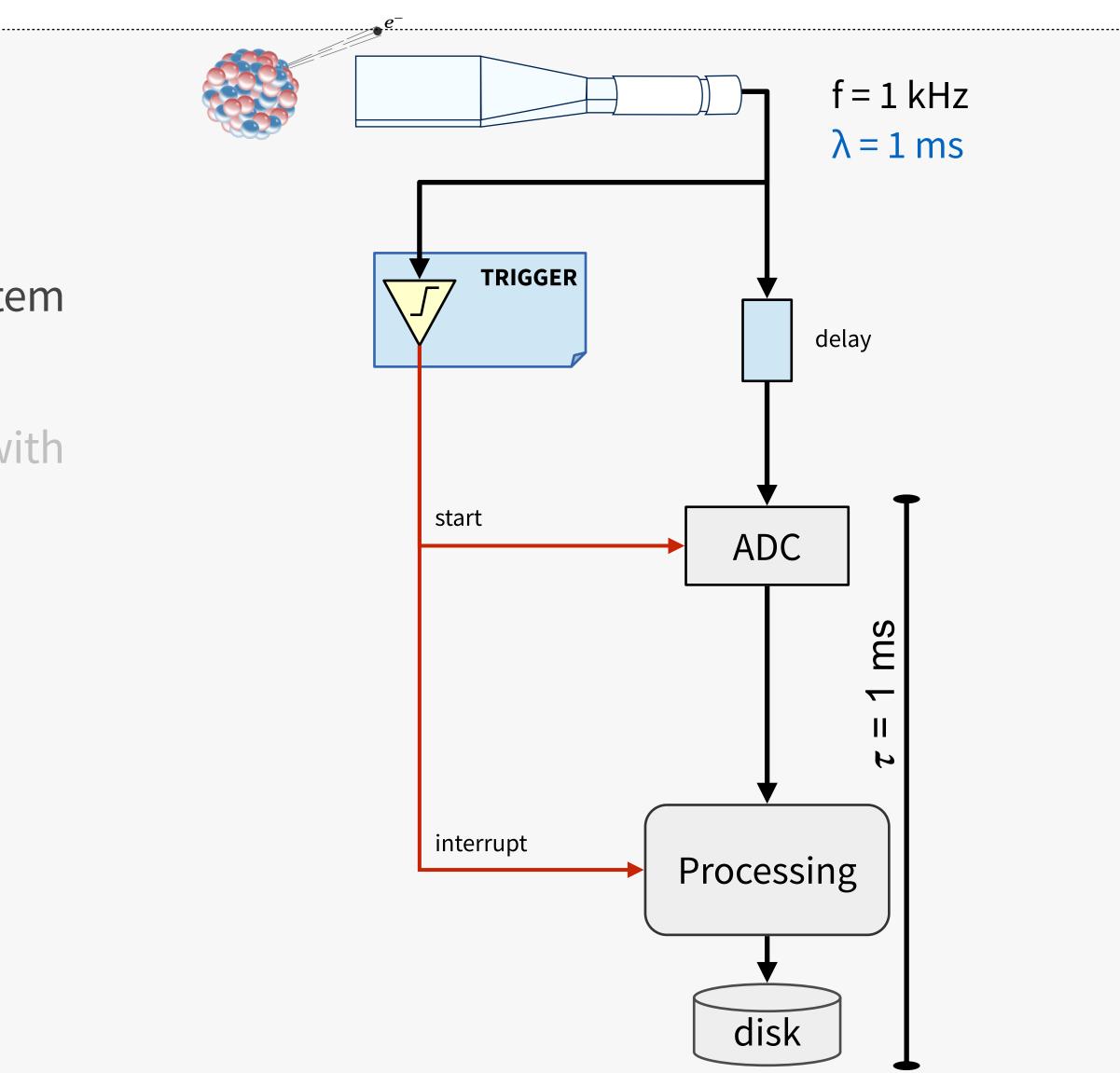


#### The **busy logic** avoids triggers while the system is busy in processing

A minimal **busy logic** can be implemented with

- an **AND** gate
- a **NOT** gate
- a flip-flop
  - **bistable** circuit that changes state (**Q**) by signals applied to the control inputs (SET, CLEAR)





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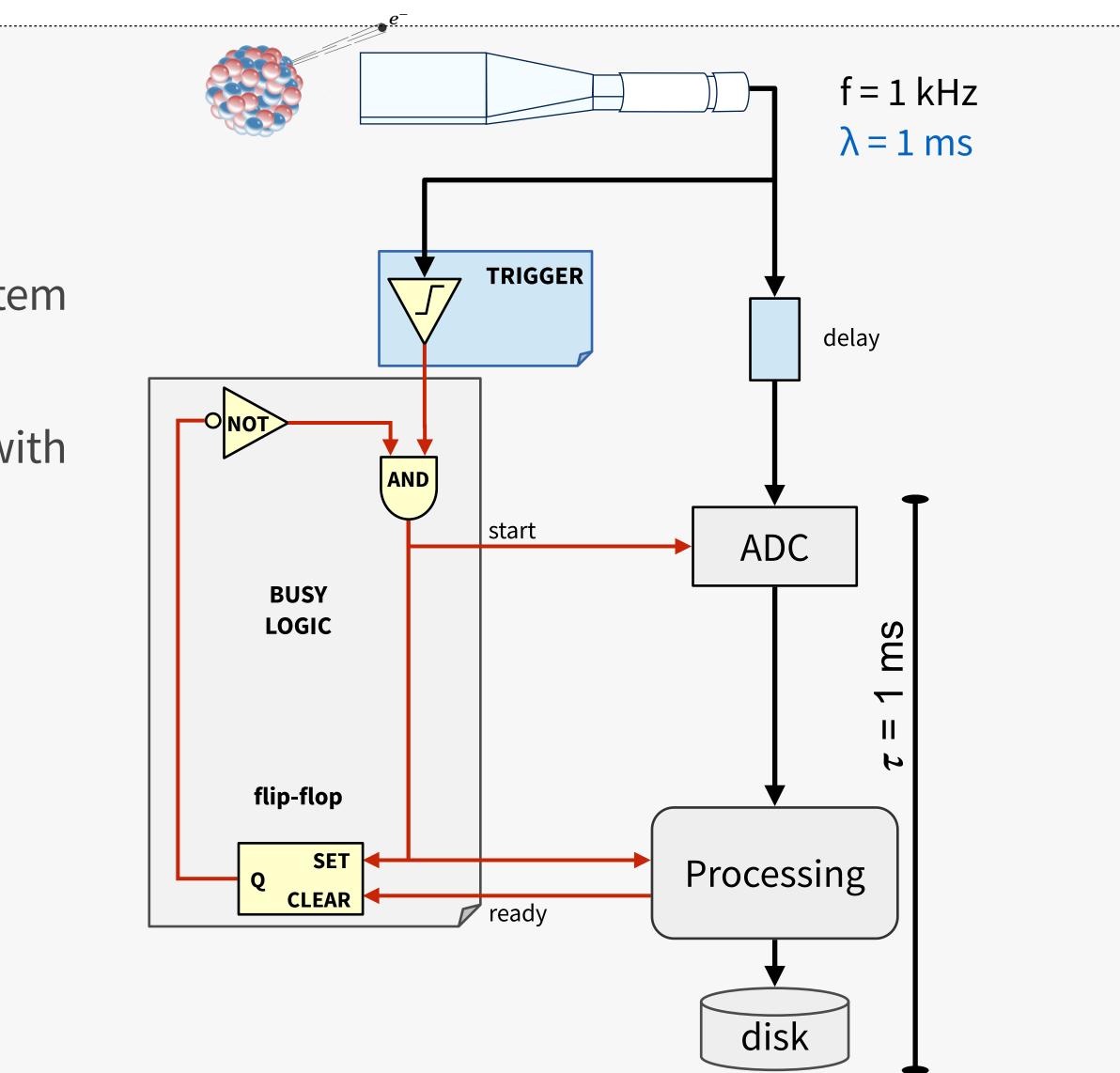


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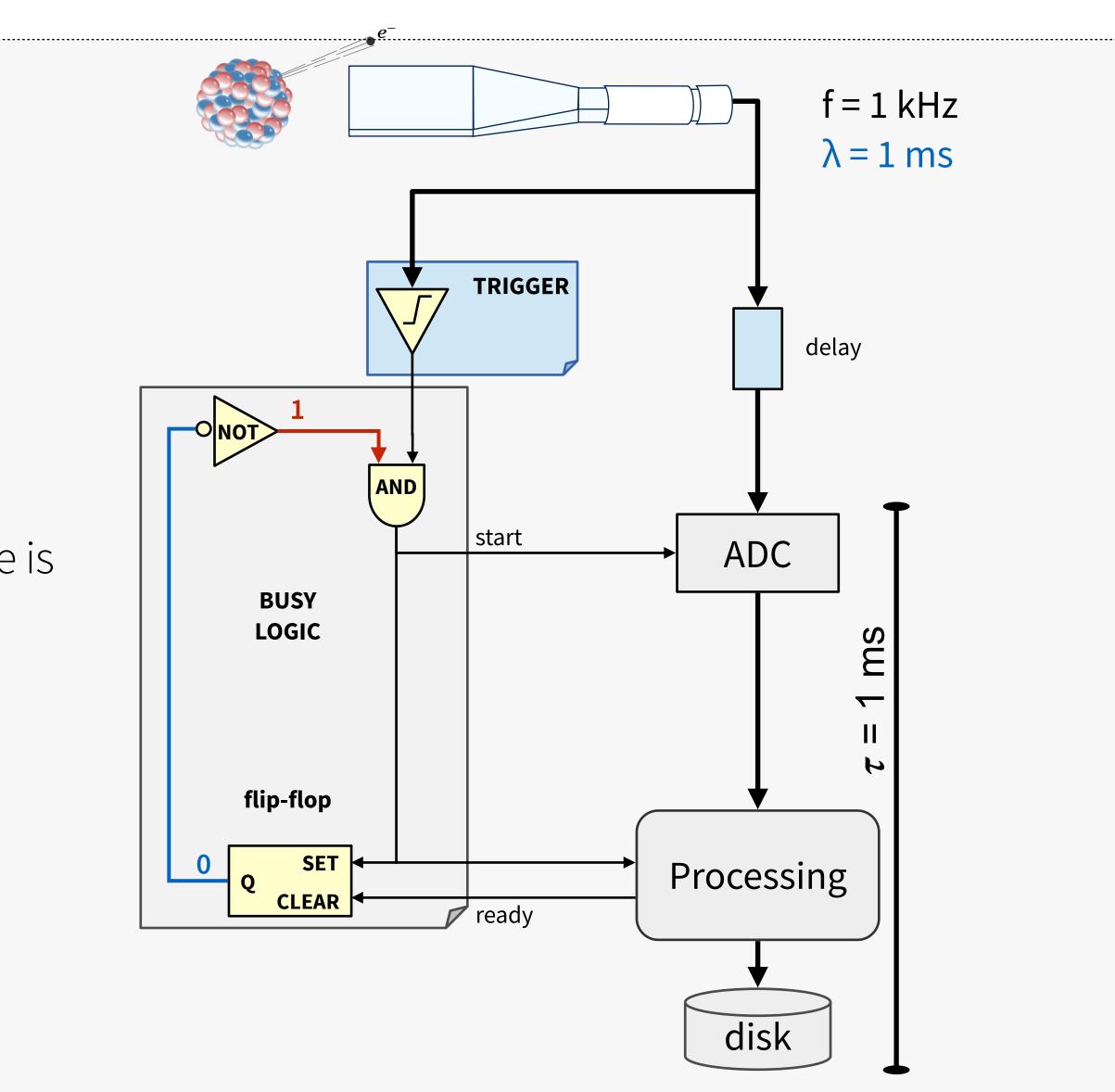


#### Start of run

- the flip-flop output is down (ground state)
- via the NOT, one of the port of the AND gate is set to up (opened)

#### i.e. system ready for new triggers







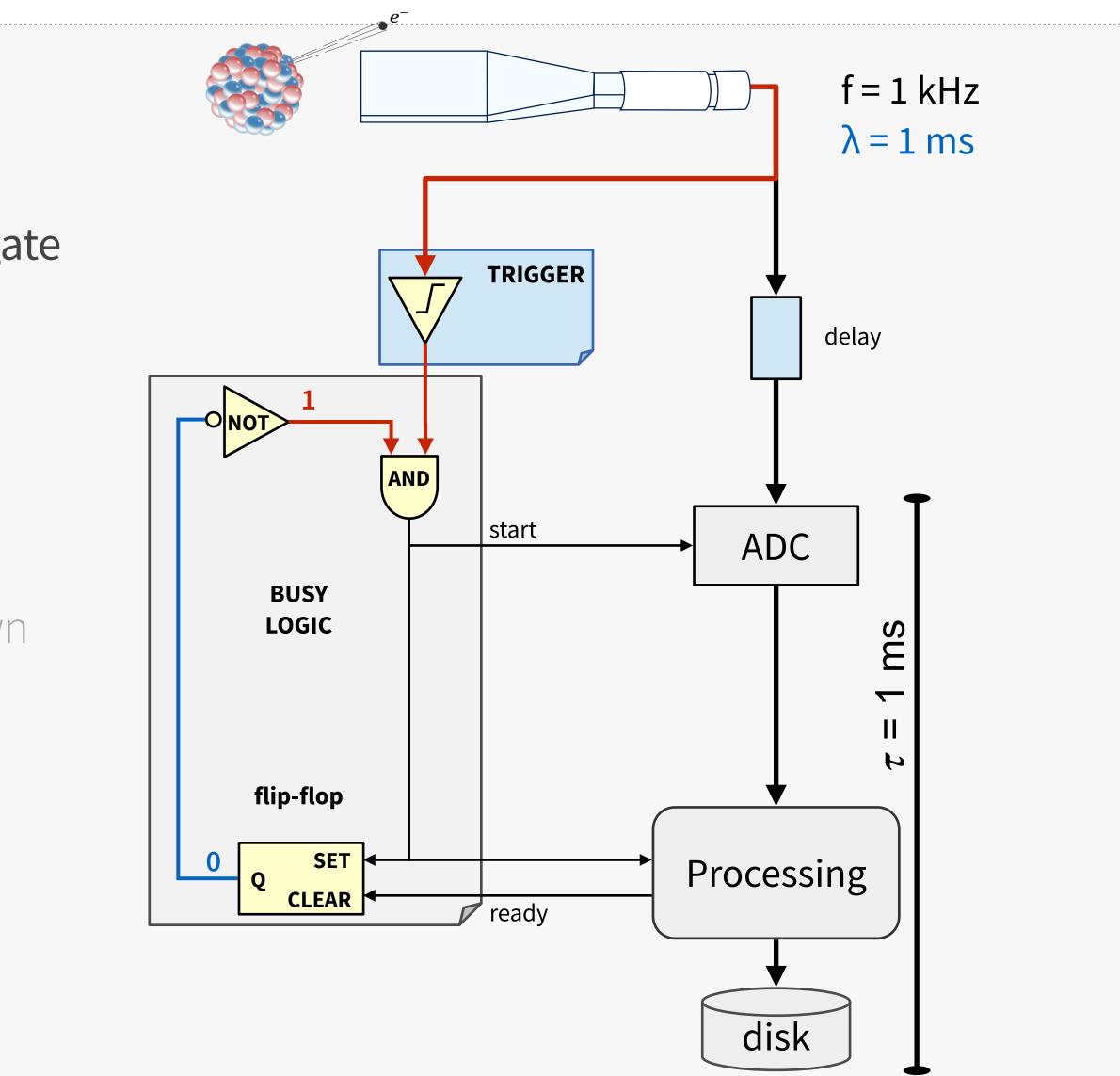


If a trigger arrives, the signal finds the AND gate open, so:

- The ADC is started
- The processing is started
- The flip-flop is flipped
- One of the AND inputs is now steadily down (closed)

Any new trigger is inhibited by the AND gate (busy)







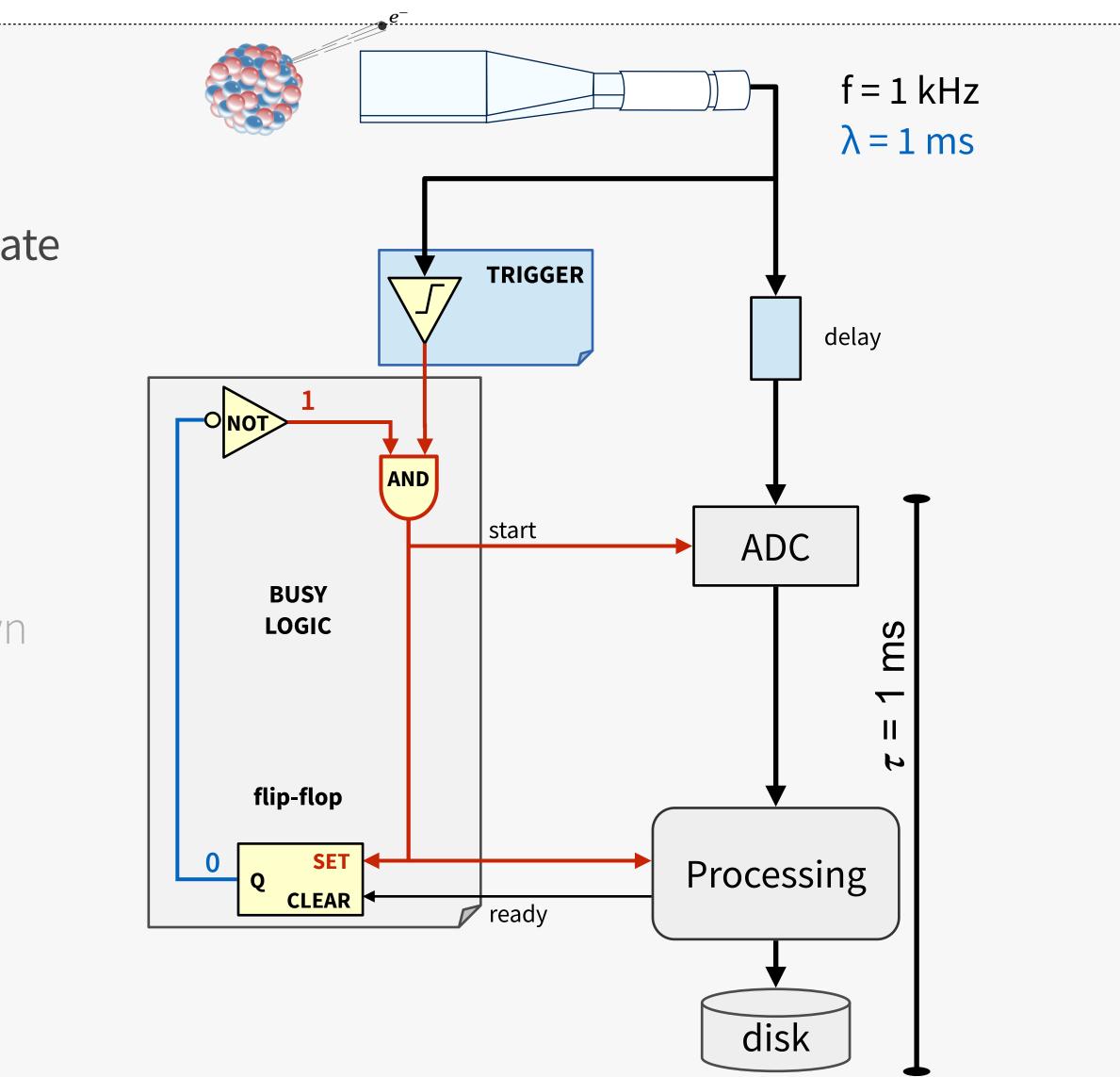


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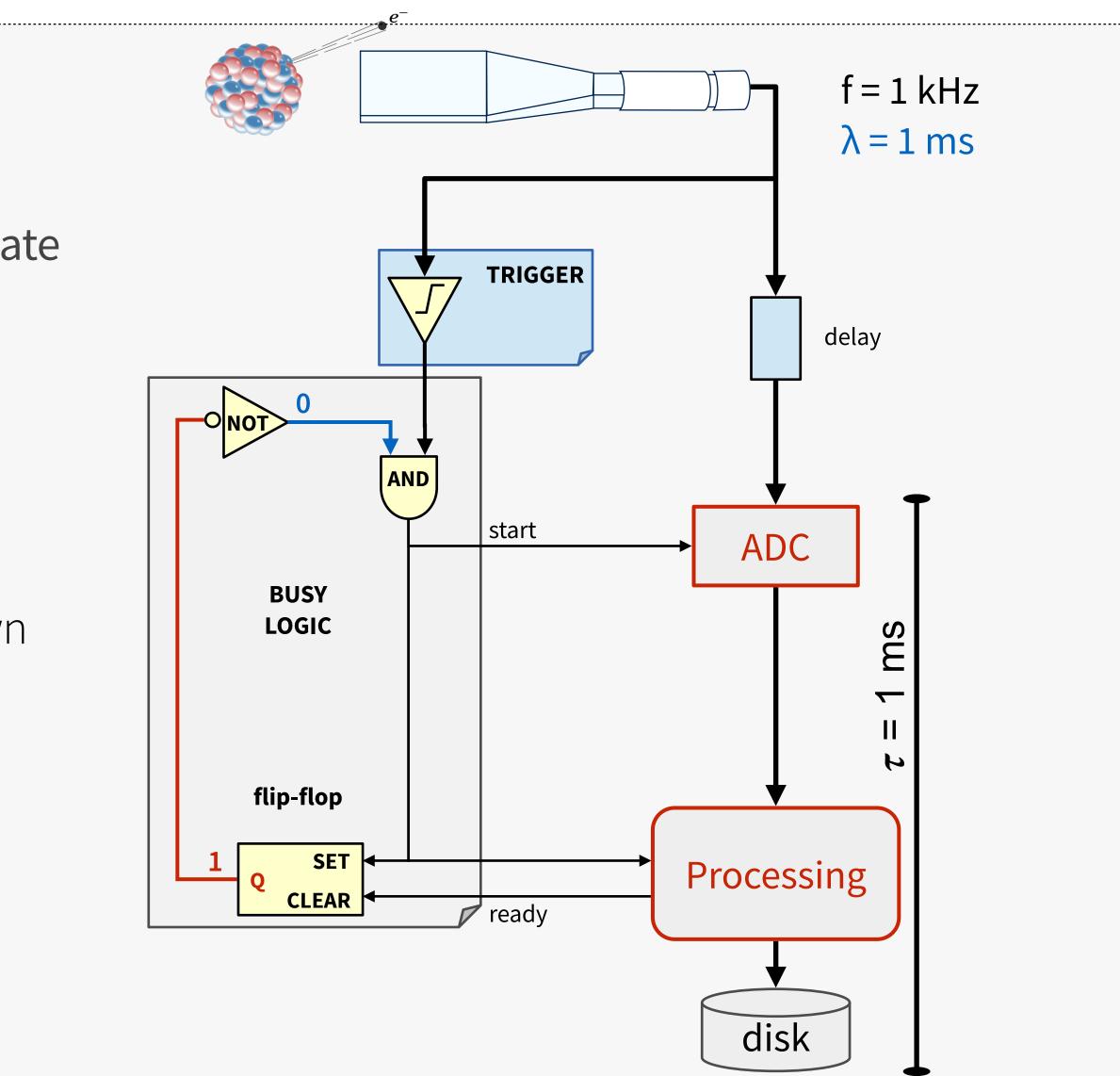


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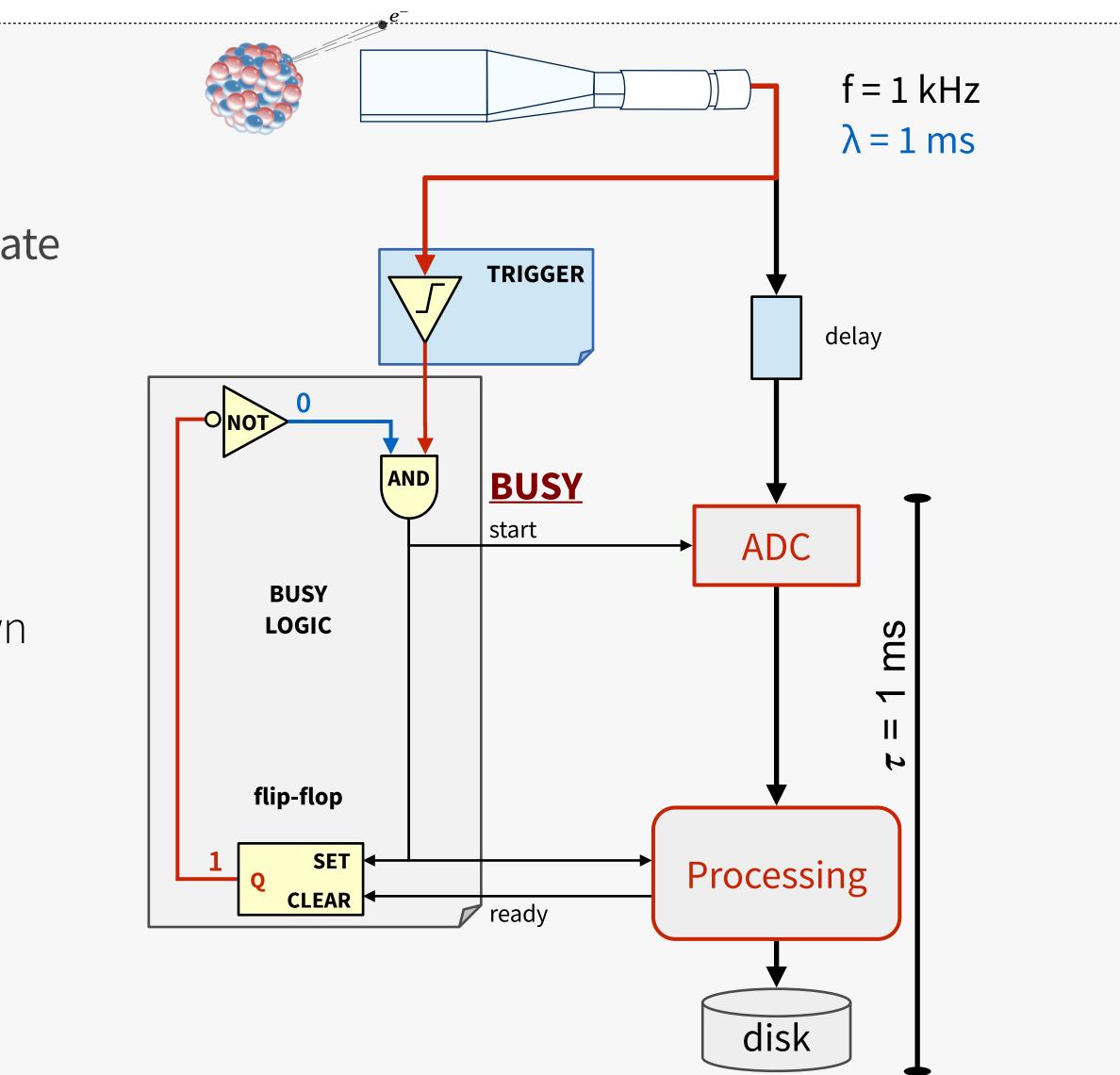


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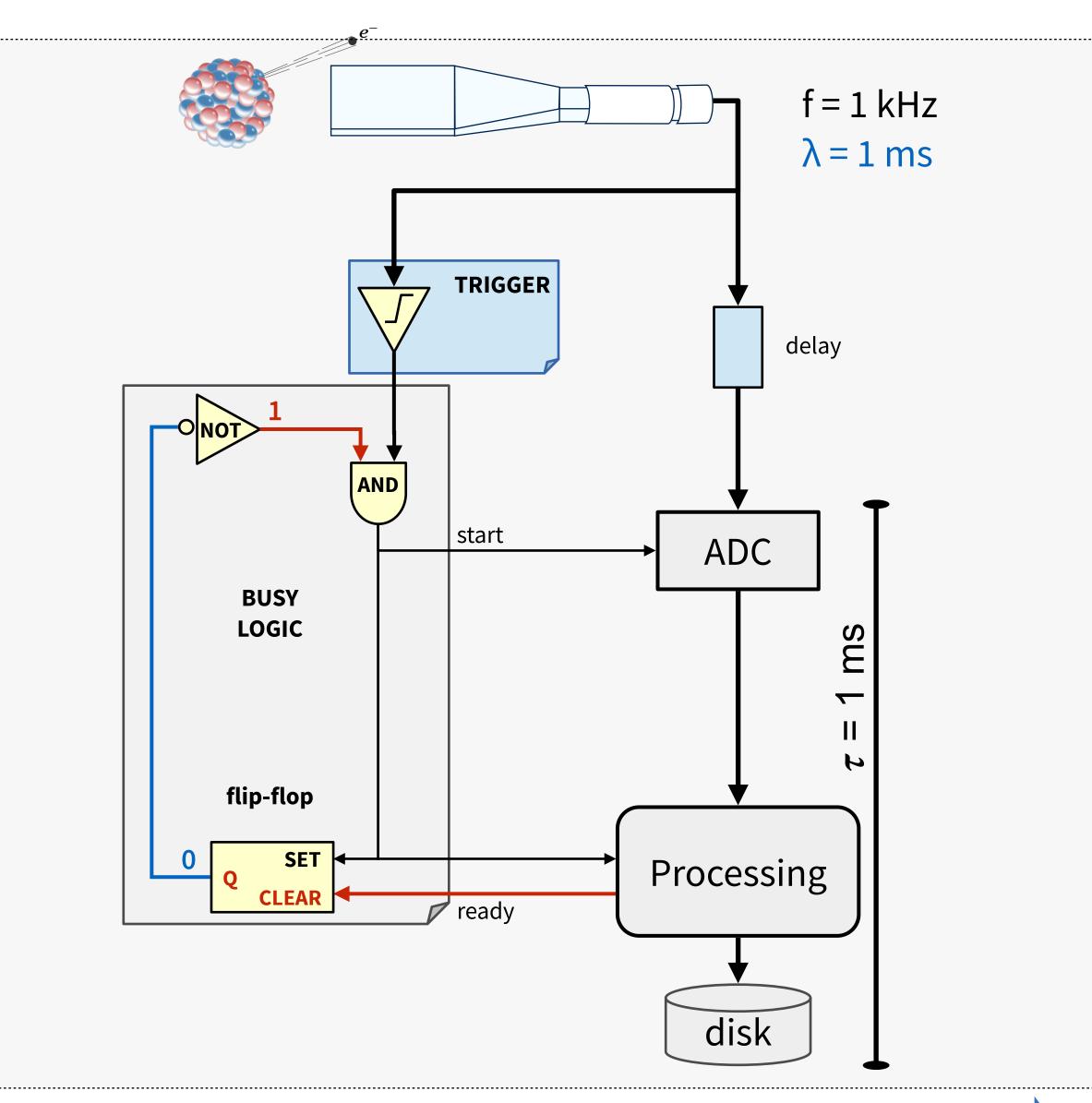
At the end of processing a ready signal is sent to the flip-flop

- The flip-flop flips again
- The gate is now opened
- The system is ready to accept a new trigger

i.e. busy logic avoids triggers while daq is busy in processing

 New triggers do not interfere w/ previous data









So the busy logic protects electronics from unwanted triggers

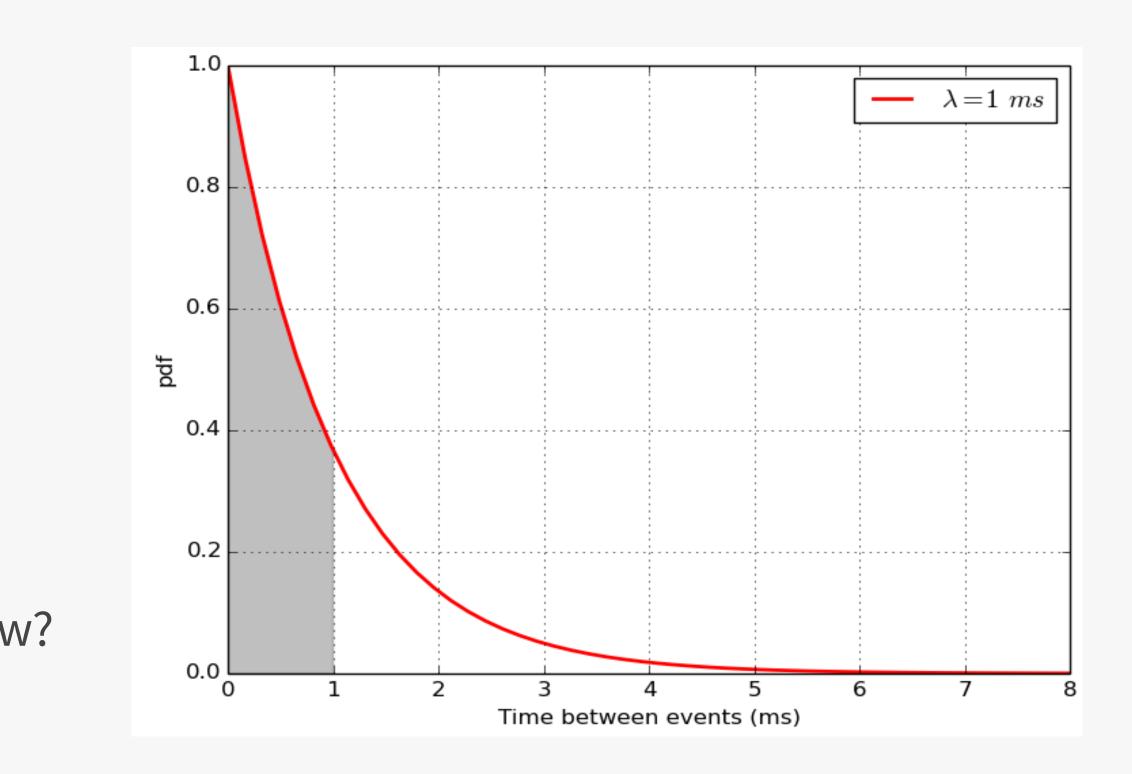
• New signals are accepted only when the system in ready to process them

What (average) DAQ rate can be achieved now?

• How much we lose with the busy logic?

Reminder: with periodic triggers and  $\tau = 1$  ms the limit was **1** kHz









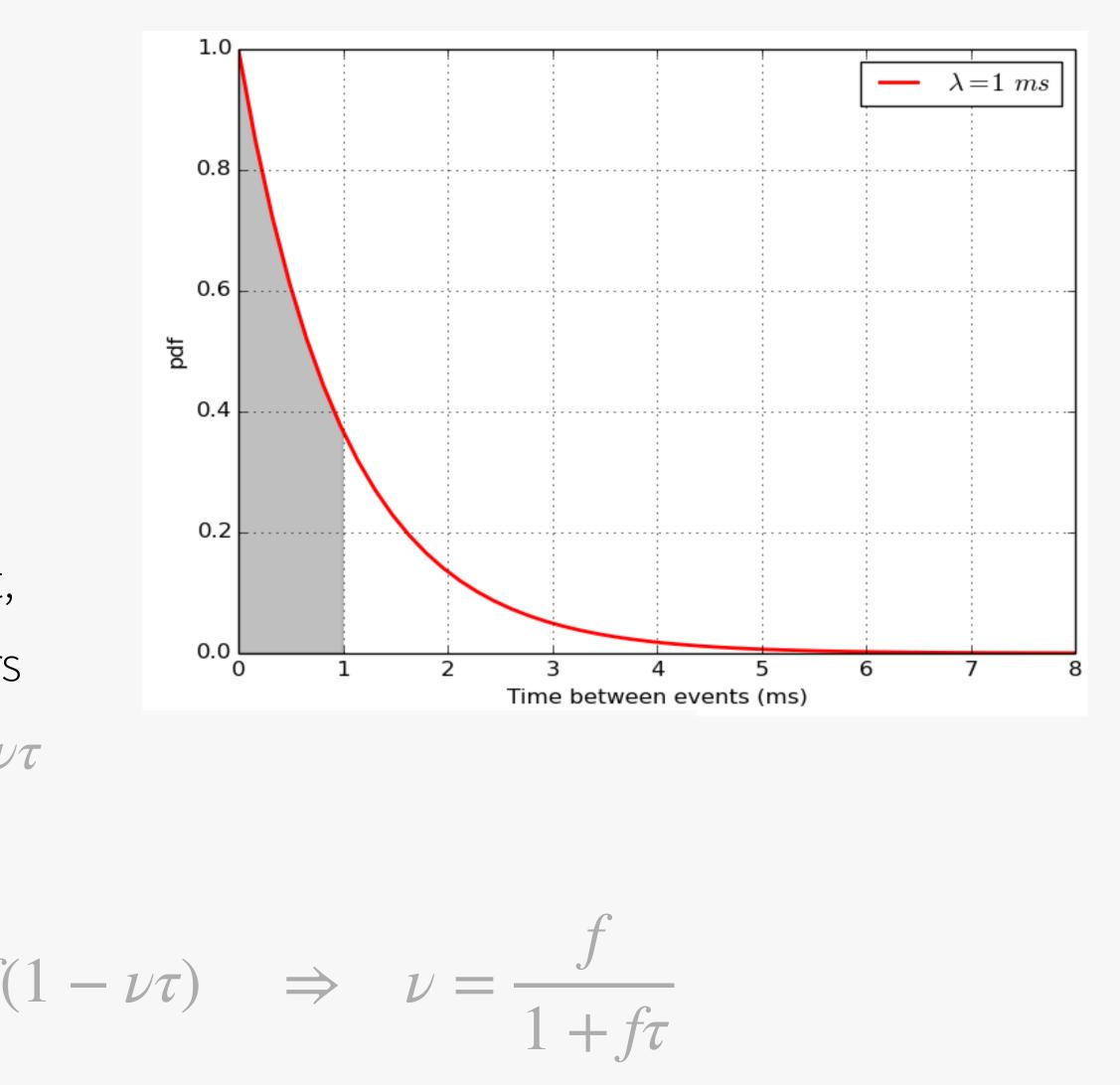
Definitions

- *f*: average rate of physics (input)
- $\nu$ : average rate of DAQ (output)
- $\tau$ : **deadtime**, needed to process an event, without being able to handle other triggers • probabilities: P[busy] =  $\nu\tau$ ; P[free] =  $1 - \nu\tau$

Therefore:

$$\nu = fP[free] \Rightarrow \nu = f($$





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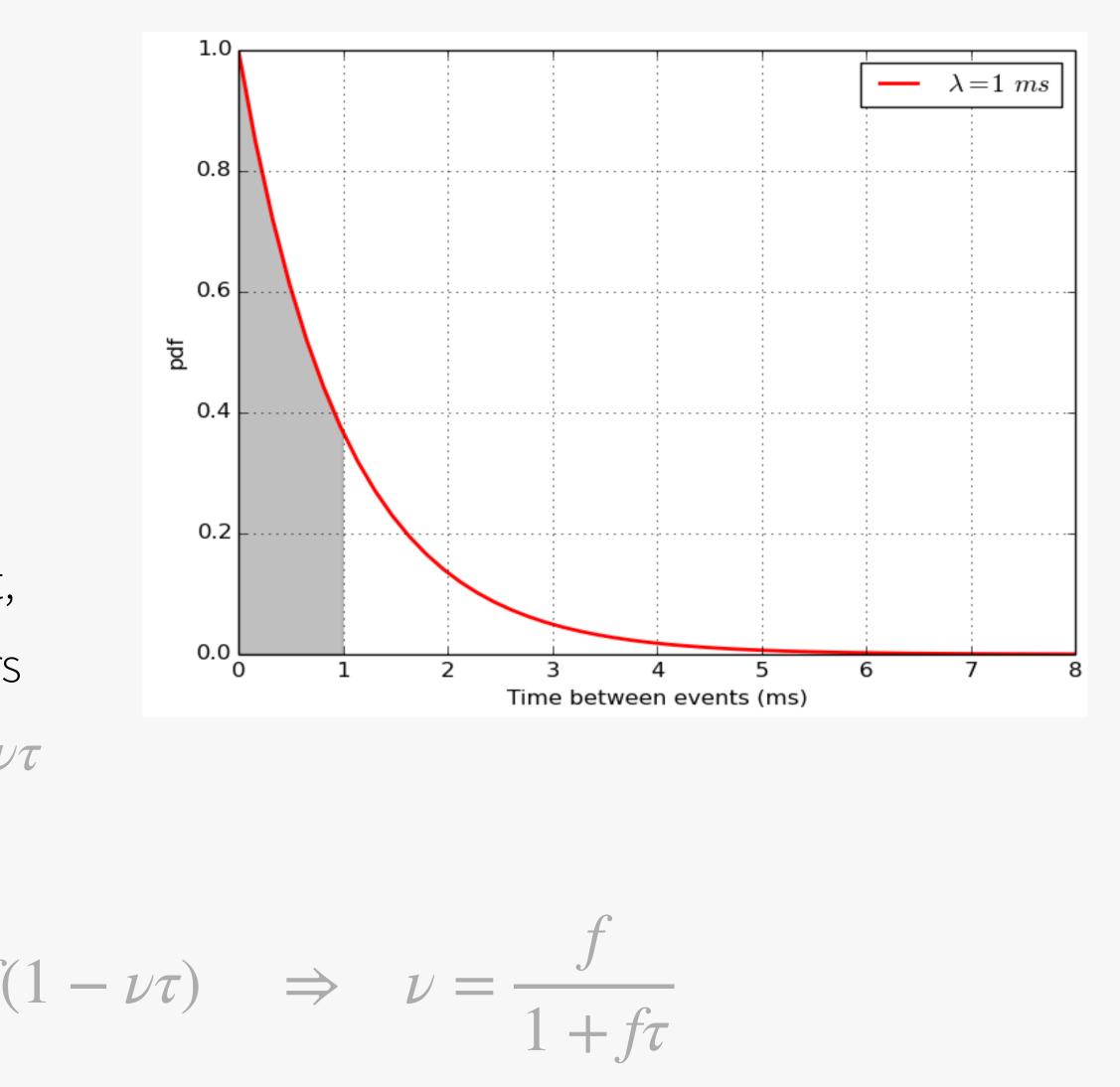




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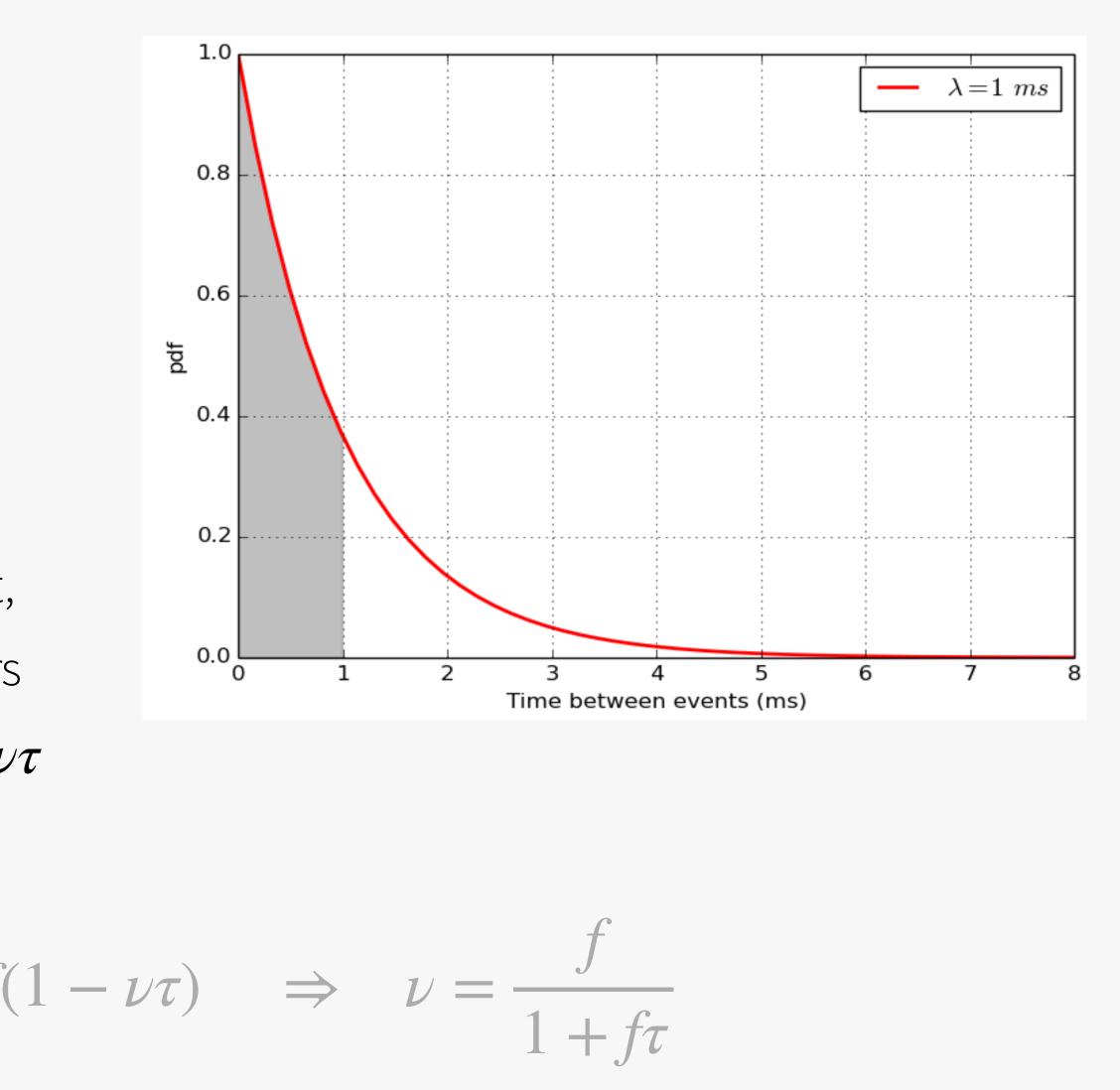




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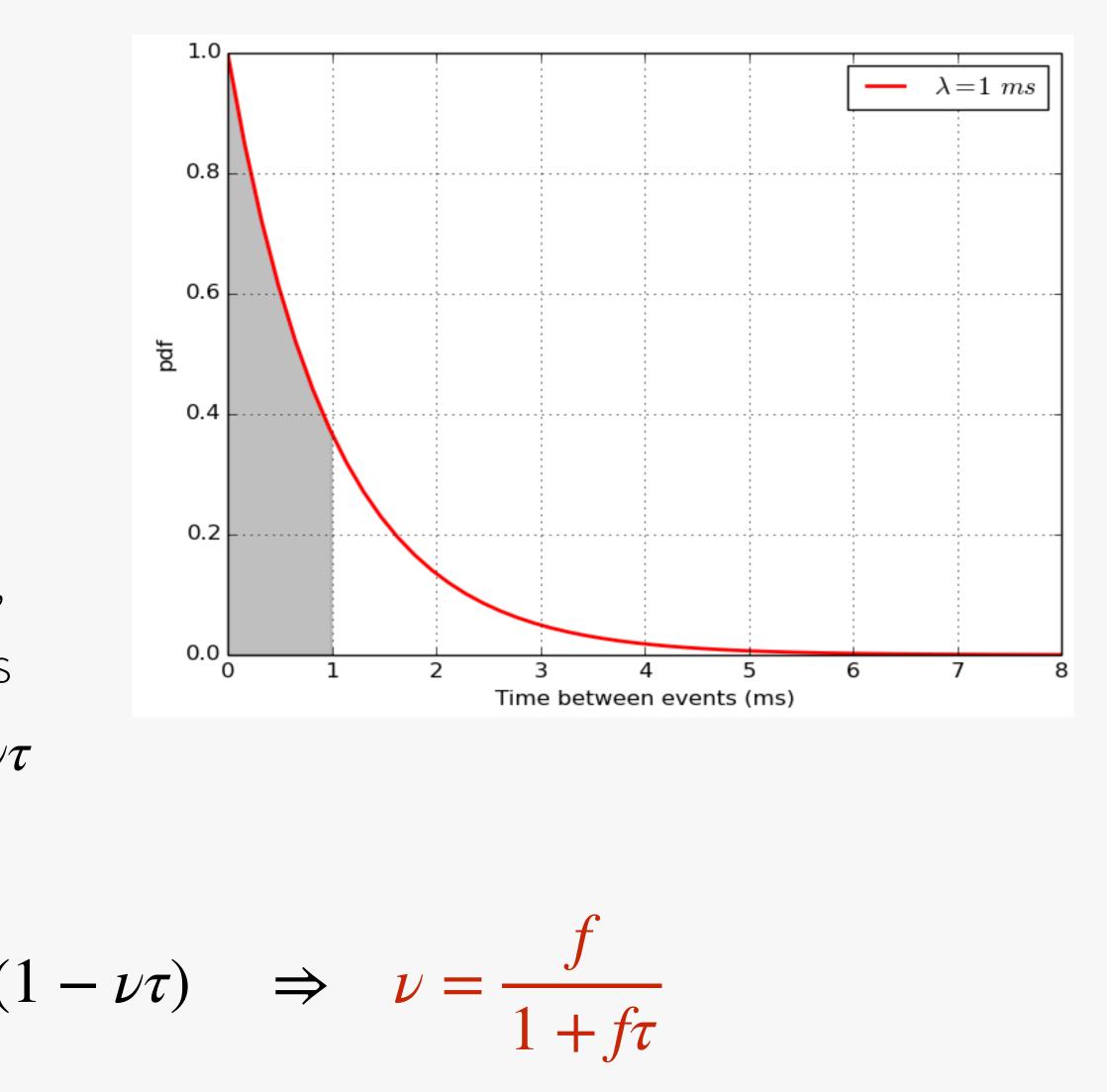




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#### Due to stochastic fluctuations

- DAQ rate always < physics rate
- Efficiency always < 100%

#### So, in our specific example

- Physics rate 1 kHz
- Deadtime 1 ms

page **44** 

$$\nu = \frac{J}{1 + f\tau} < f$$

$$\epsilon = \frac{N_{saved}}{N_{tot}} = \frac{1}{1 + f\tau} < 100\%$$

ſ

 $f = 1 \ kHz \qquad \qquad \nu = 500 \ Hz \\ \tau = 1 \ ms \qquad \qquad \epsilon = 50 \%$ 





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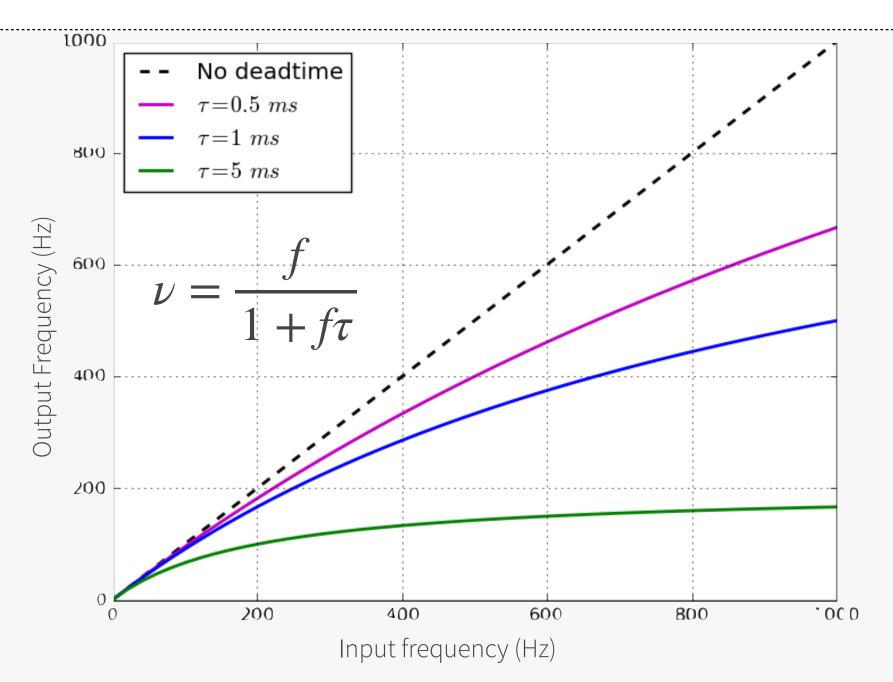
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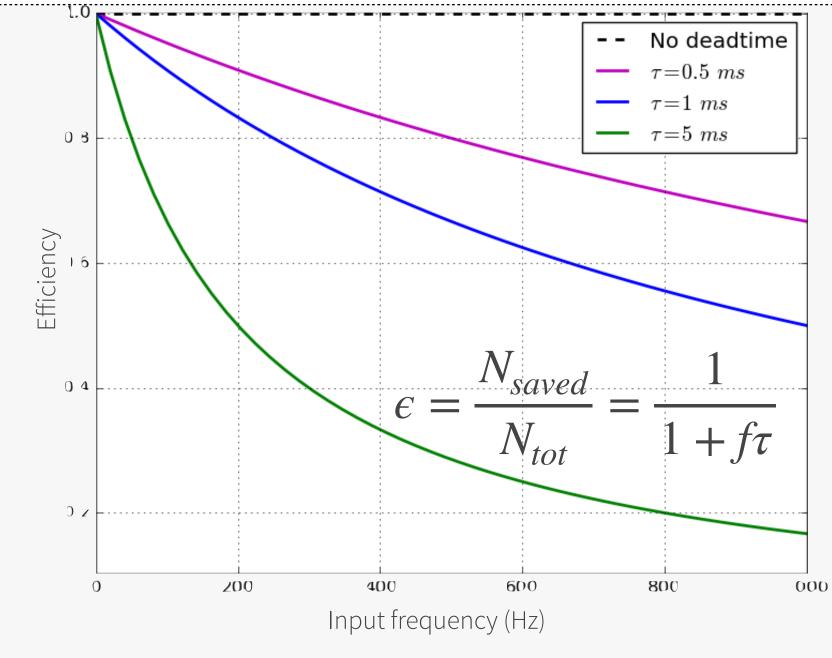






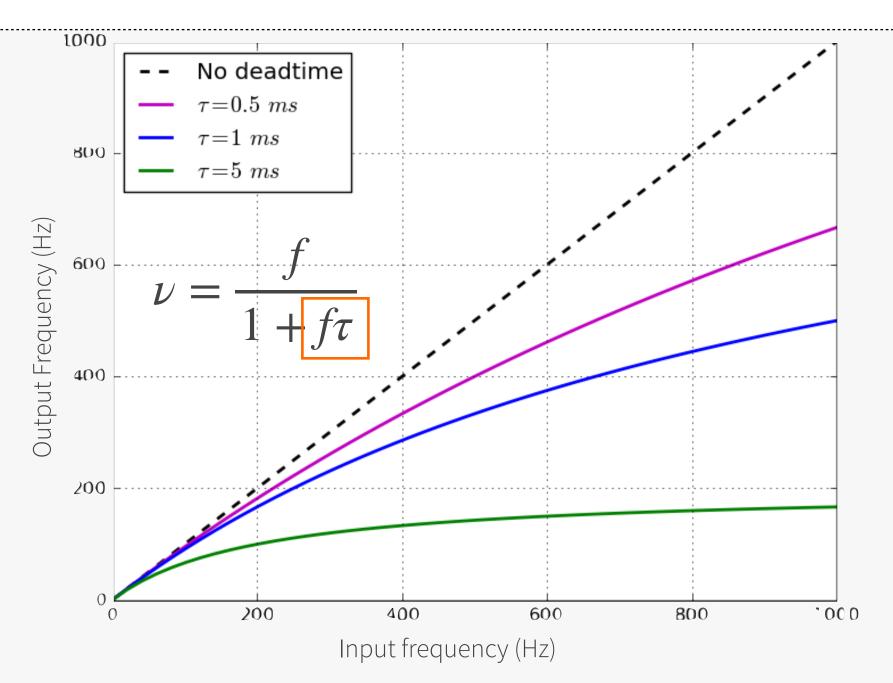
In order to obtain  $\epsilon \simeq 100\%$  (i.e.:  $\nu \simeq \tau$ )  $\rightarrow f\tau \ll 1 \rightarrow \tau \ll \lambda$ 

- E.g.:  $\epsilon \simeq 99\%$  for f = 1 kHz  $\Rightarrow \tau < 0.01$  ms  $\Rightarrow 1/\tau > 100$  kHz
- To cope with the input signal fluctuations, we have to **over-design** our DAQ system by a **factor 100**! How can we mitigate this effect?



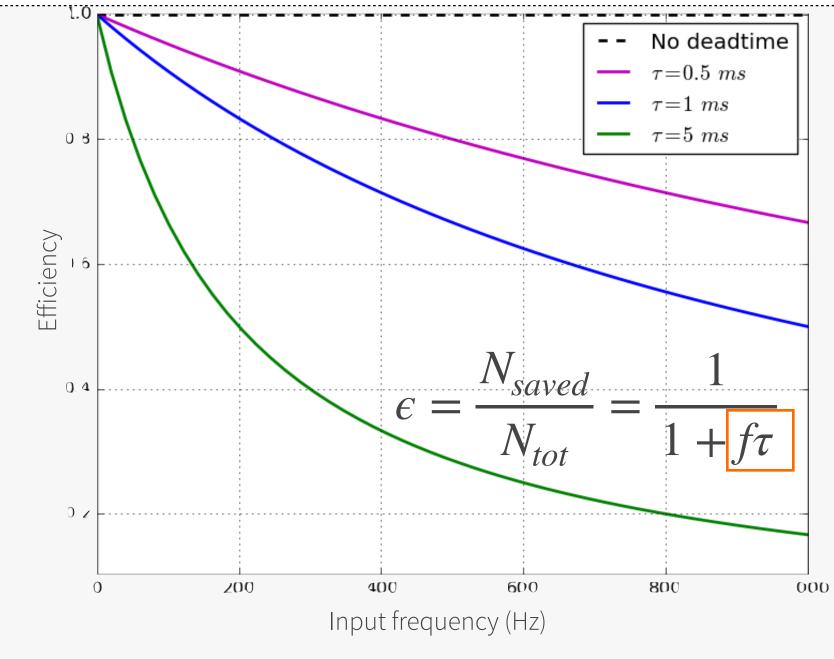






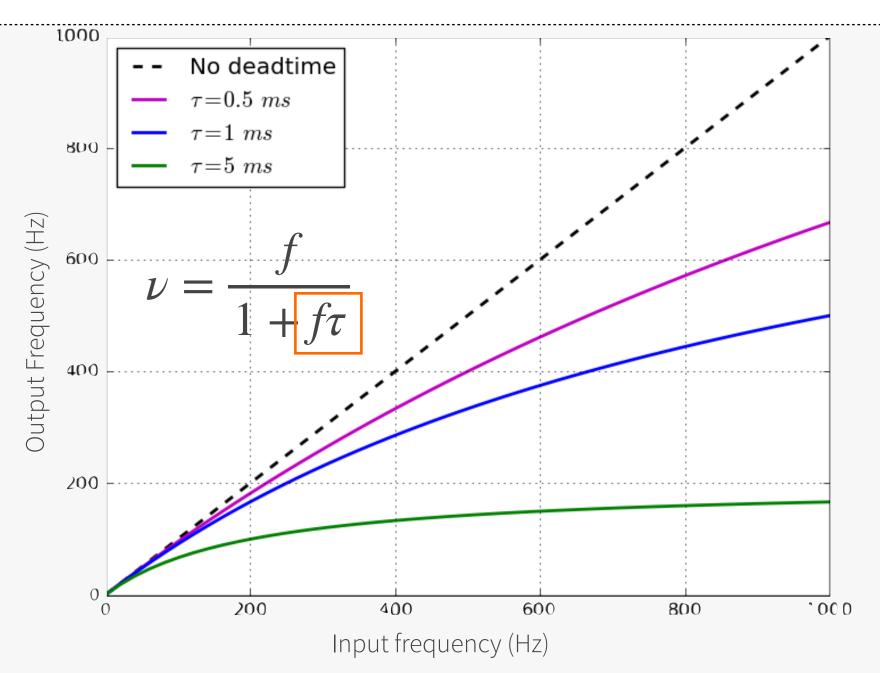
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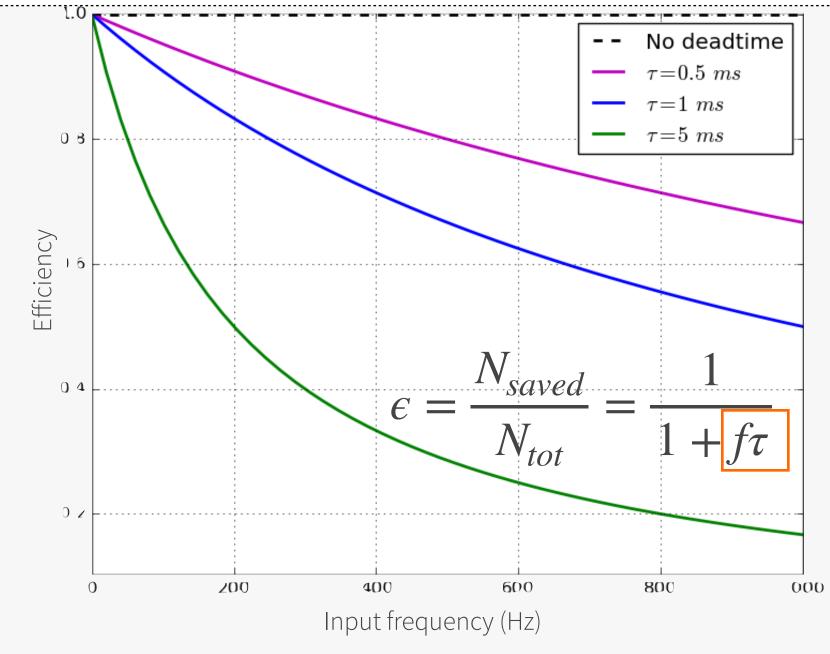






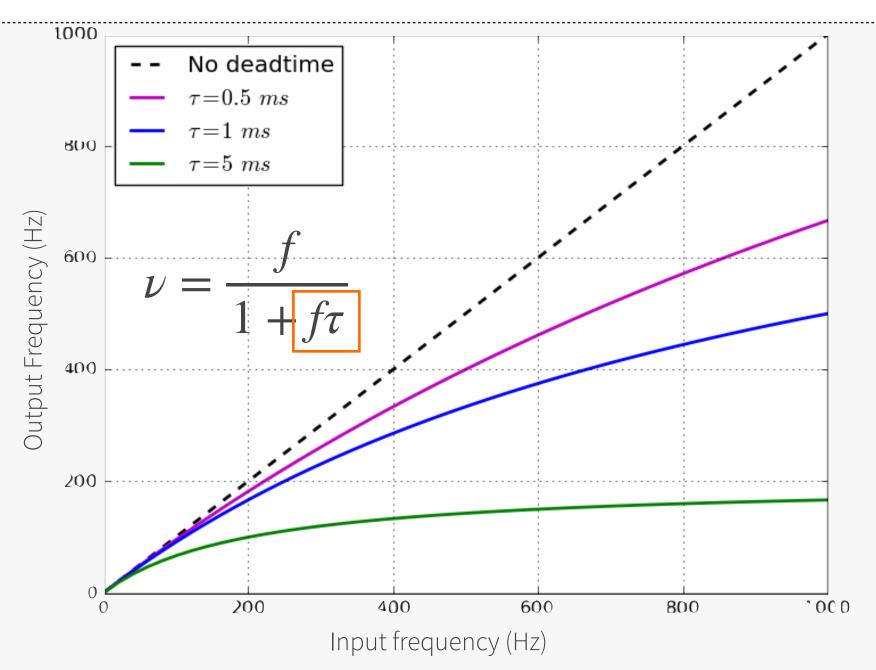
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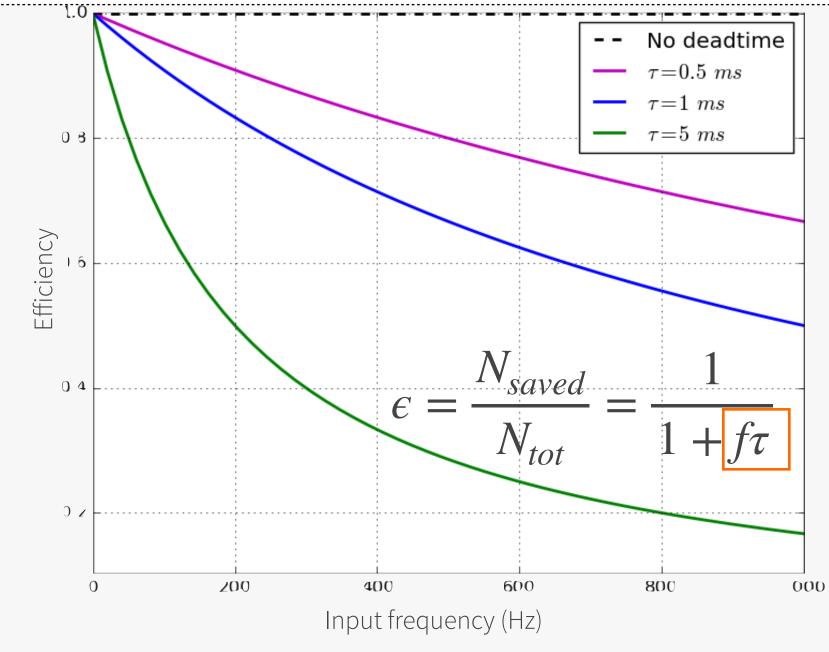






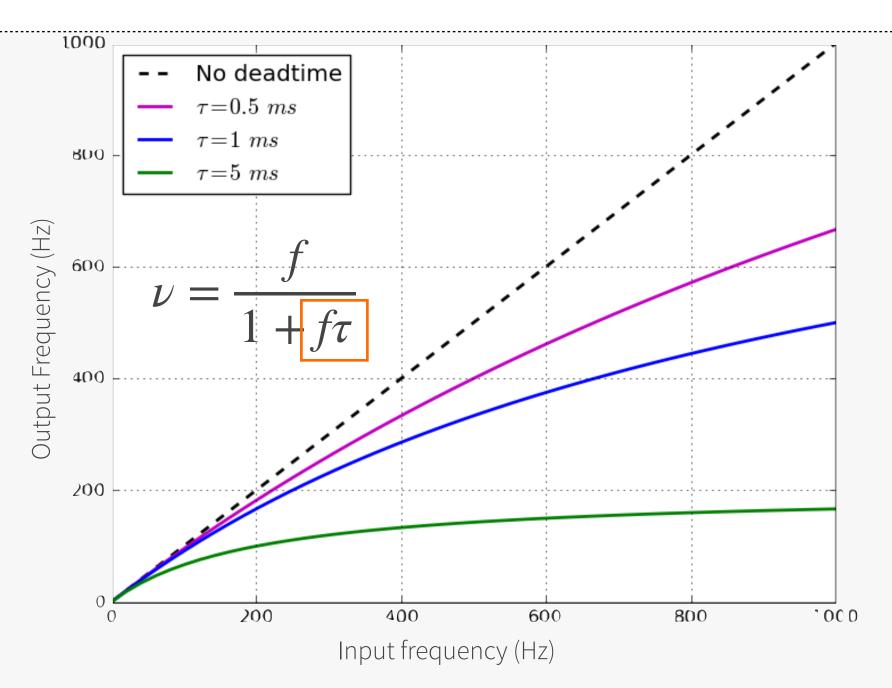
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08

Efficiency

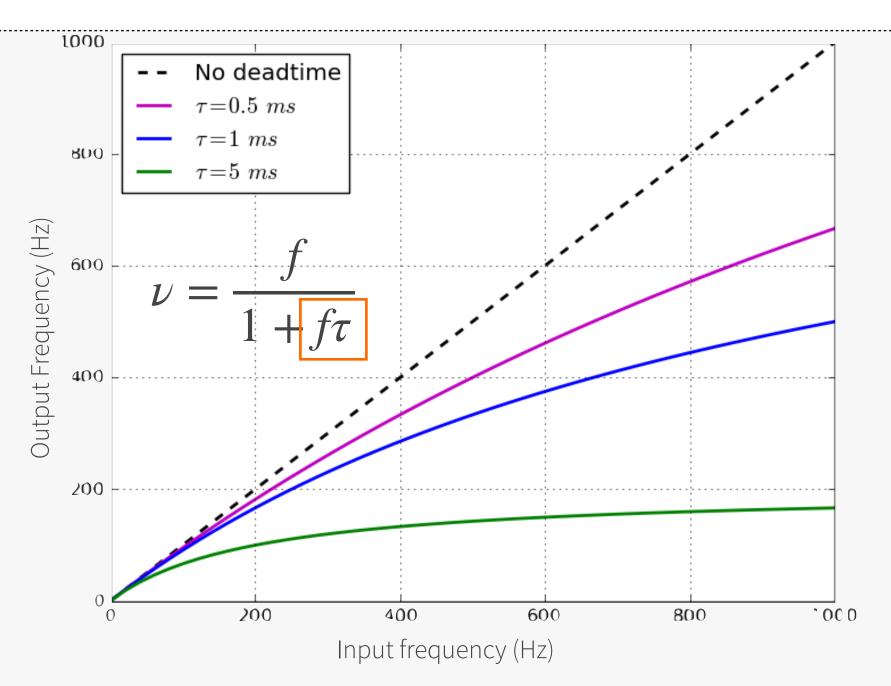
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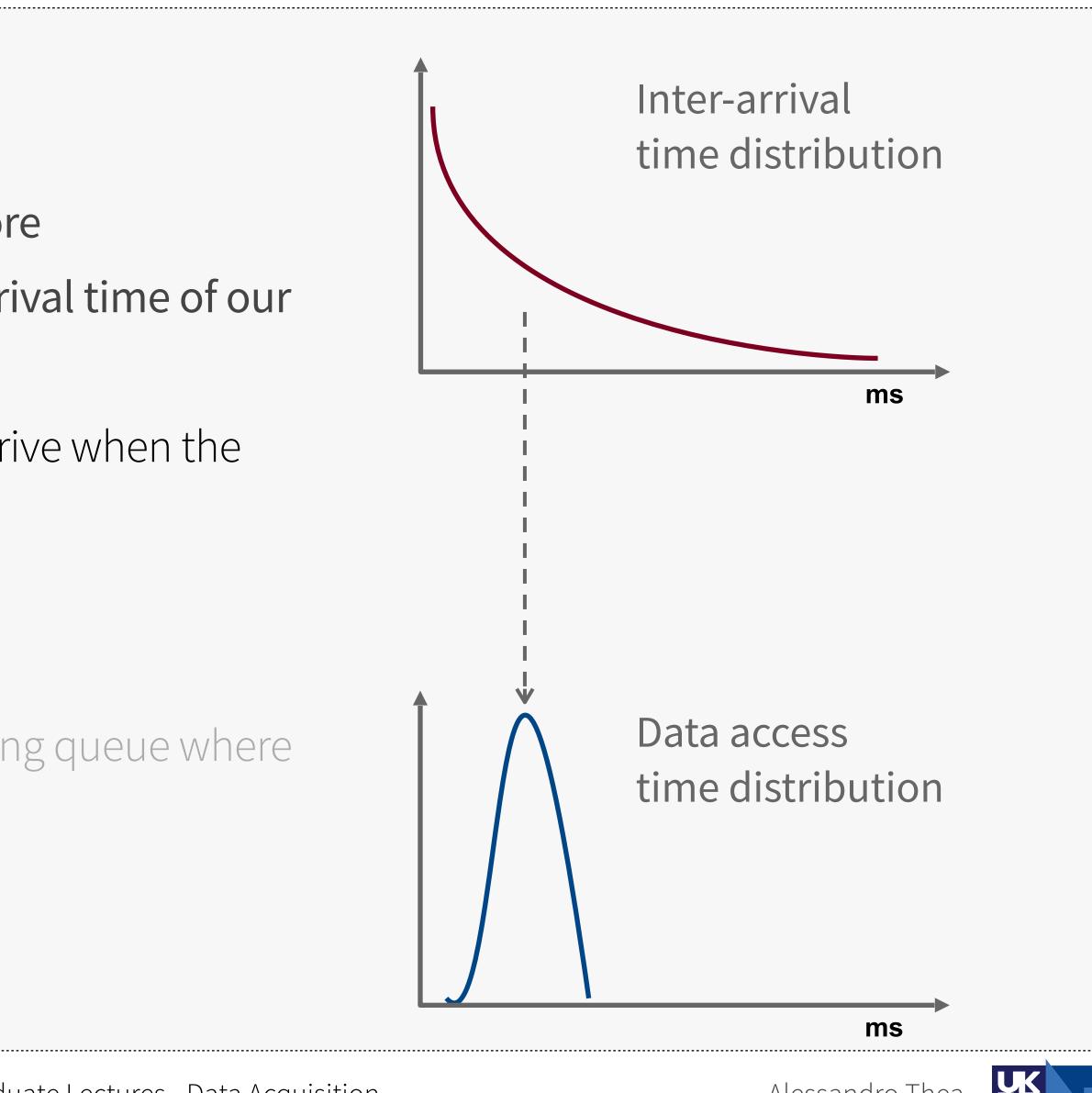
What if we were able to make the system more deterministic and less dependent on the arrival time of our signals?

- Then we could ensure that events don't arrive when the system is busy
- This is called **de-randomization**

How it can be achieved?

• by **buffering** the data (introducing a holding queue where it can wait to be processed)





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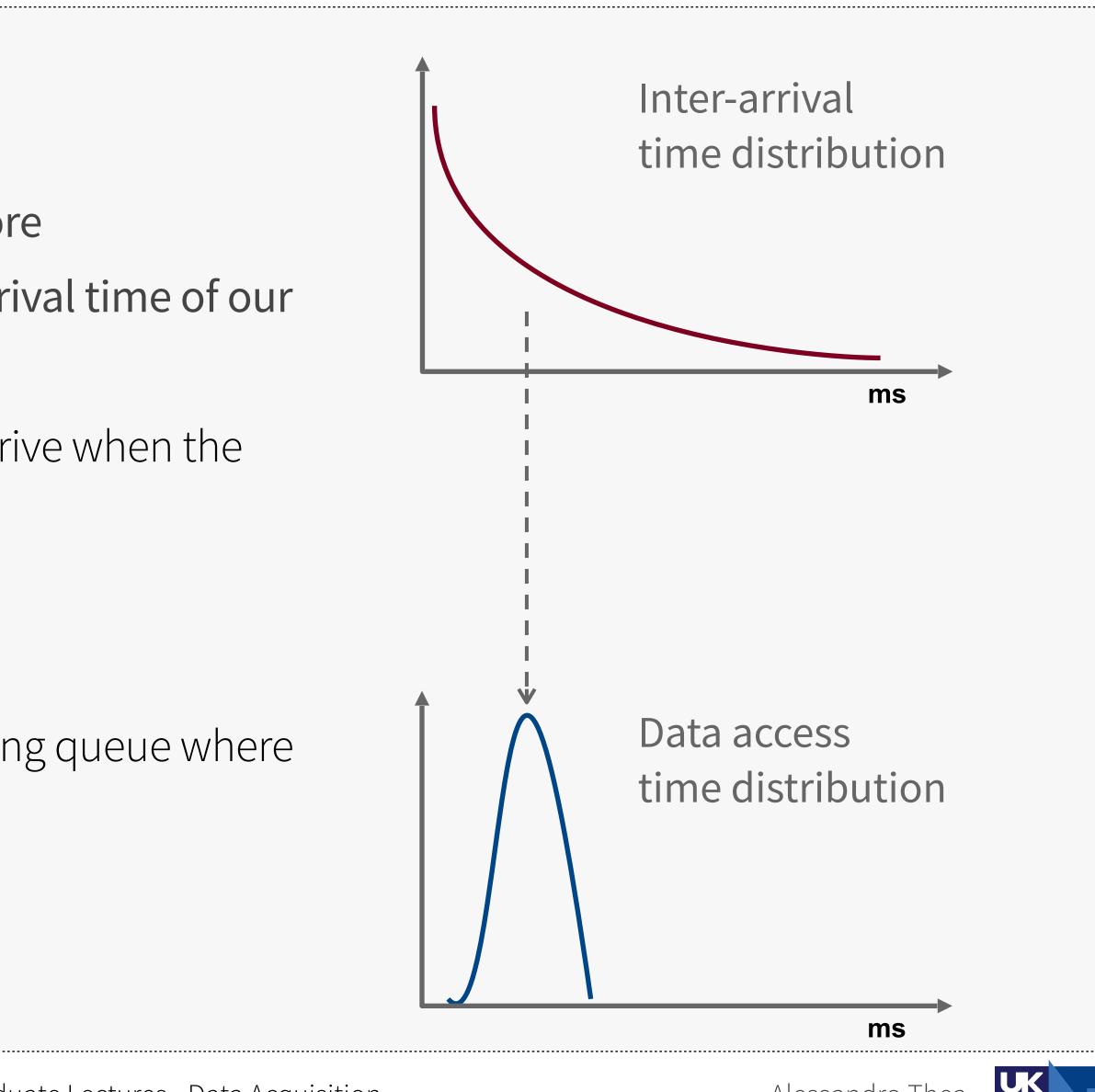


What if we were able to make the system more deterministic and less dependent on the arrival time of our signals?

- Then we could ensure that events don't arrive when the system is busy
- This is called **de-randomization**

How it can be achieved?

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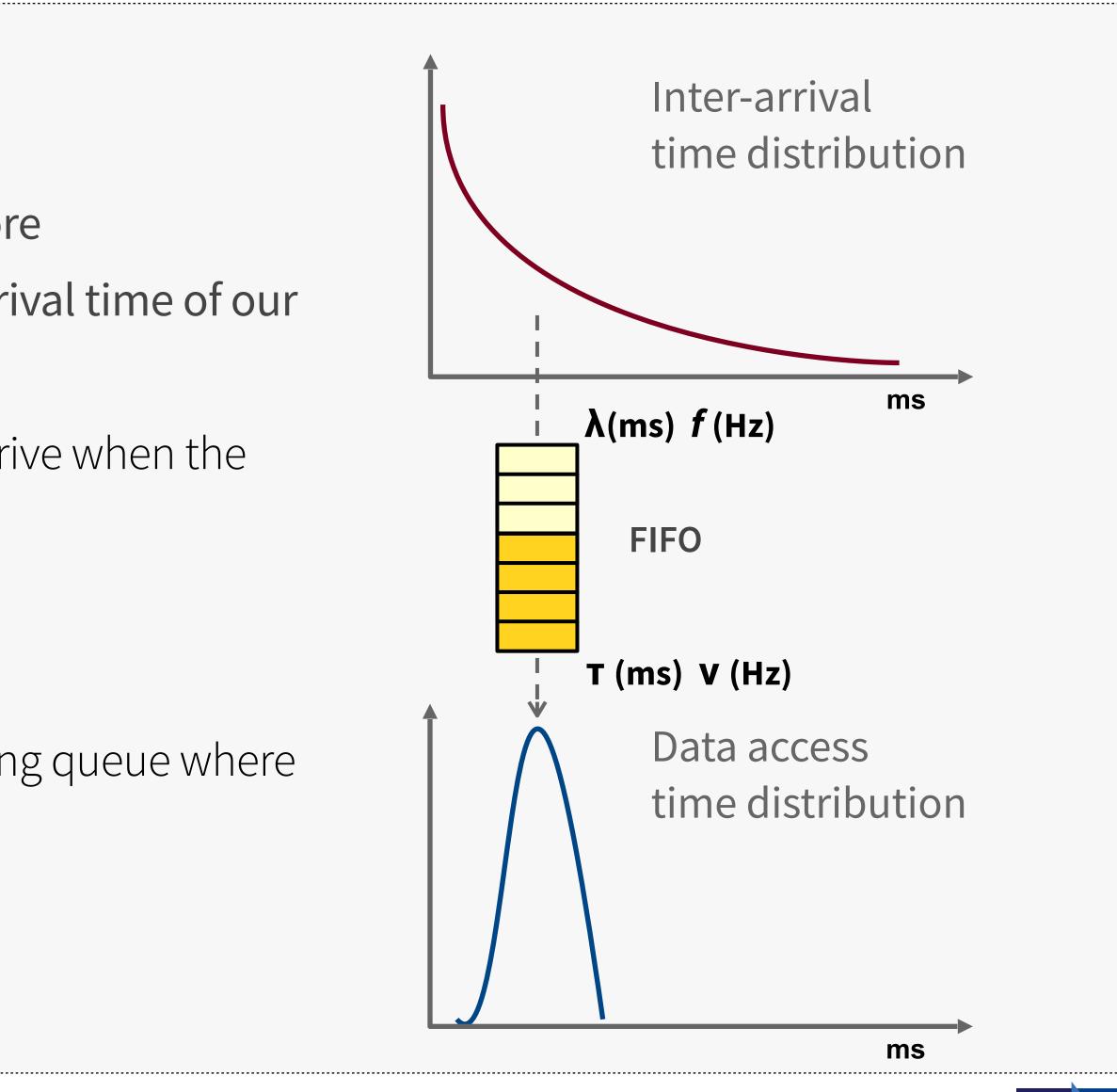


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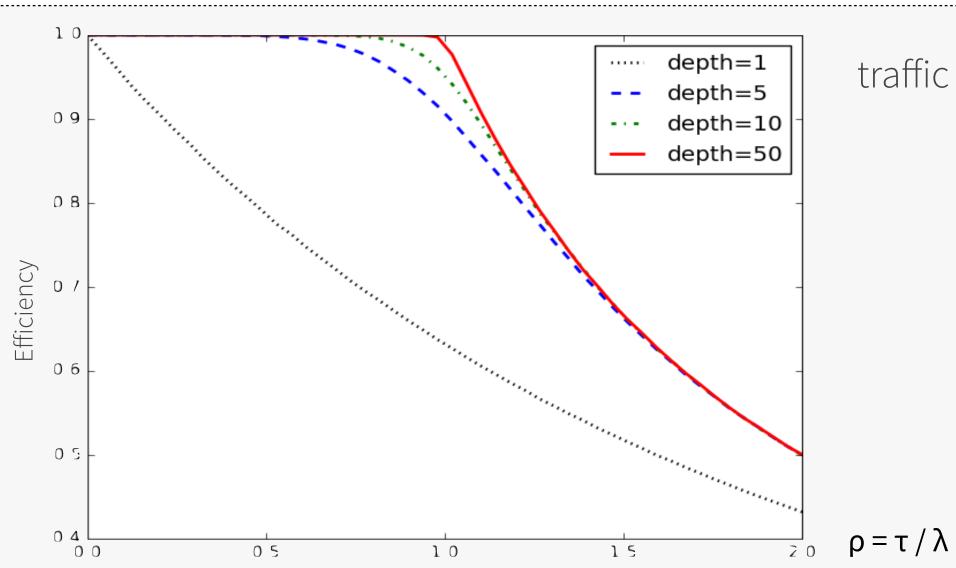
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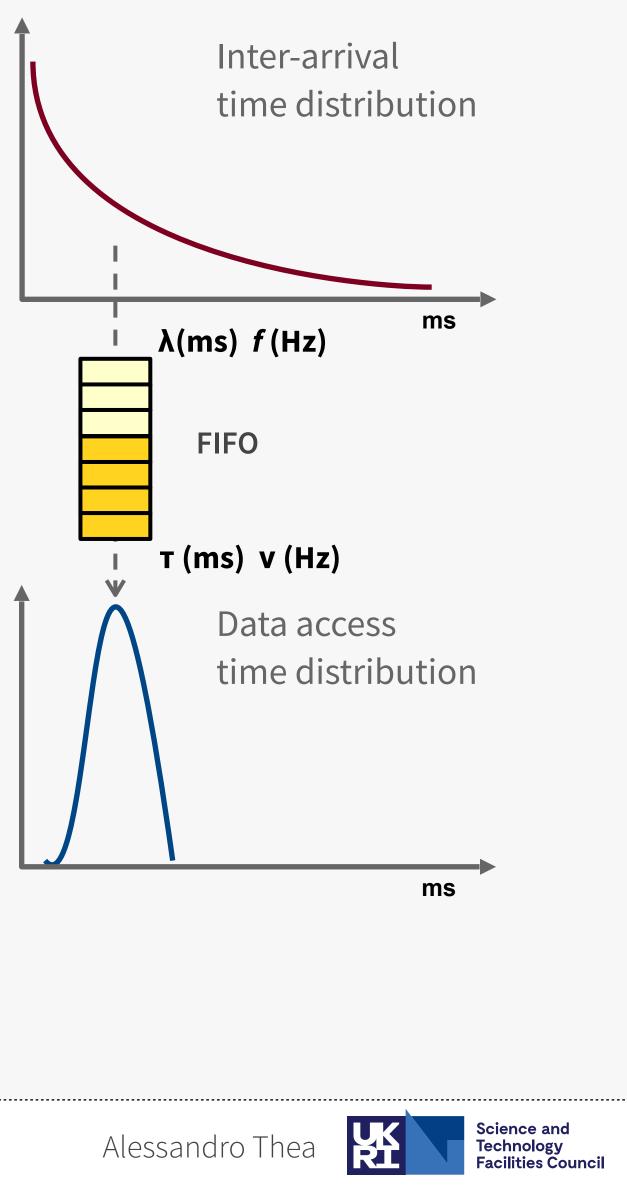


Efficiency vs traffic intensity for different queue depths

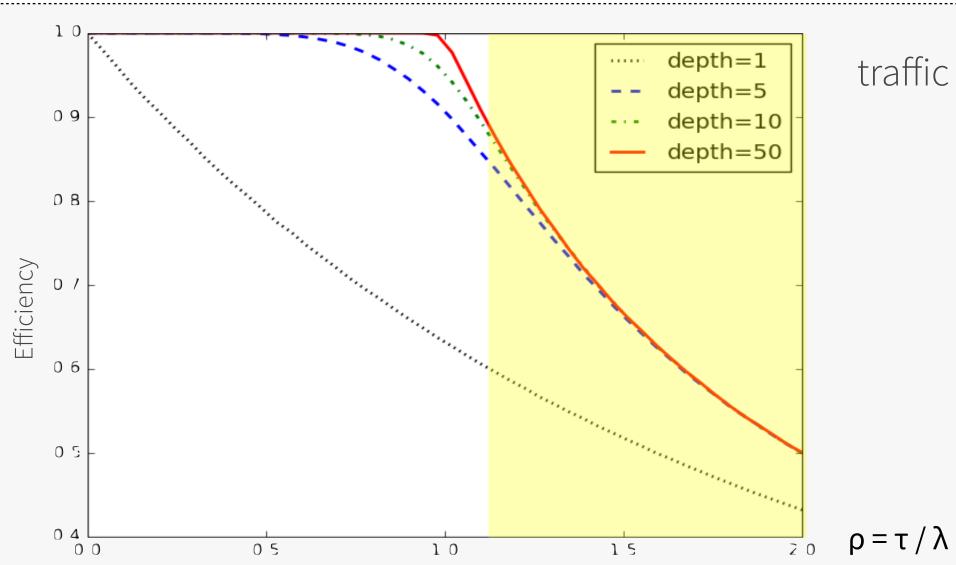
- the output is over-designed ( $\tau \ll \lambda$ ) • ρ≪1:
- $\rho > 1$ : the system is overloaded ( $\tau > \lambda$ )
- $\rho \sim 1$ : using a queue, high efficiency obtained even w/ moderate depth Analytic calculation possible for very simple systems only
- Otherwise MonteCarlo simulation is required

traffic intensity :  $\rho = \tau/\lambda$ 





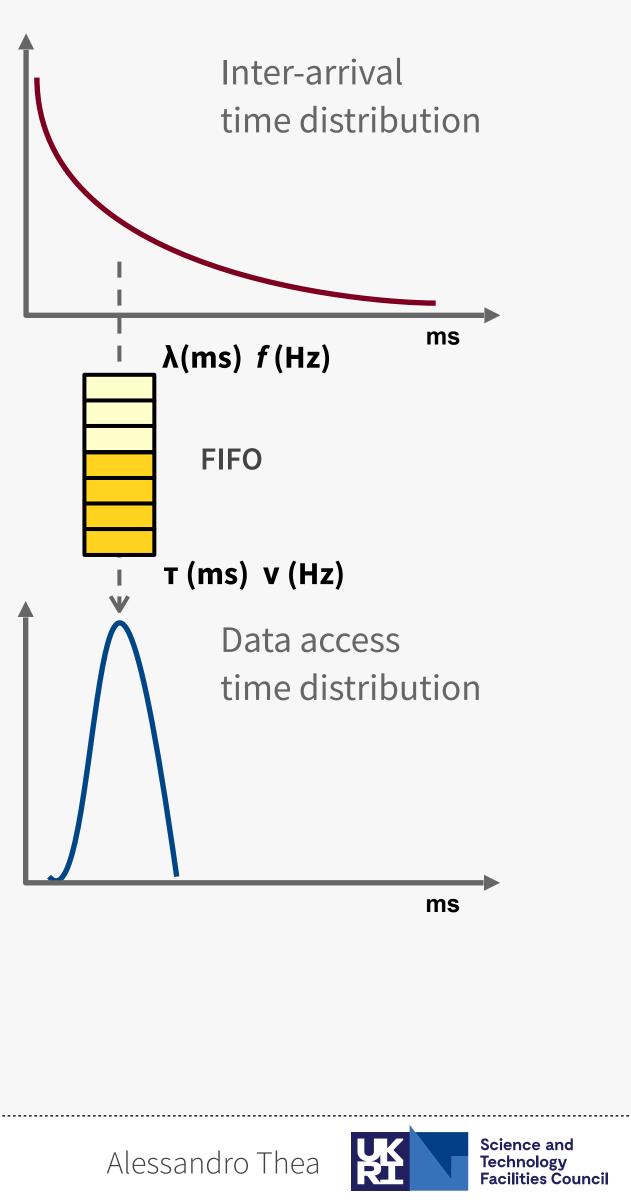




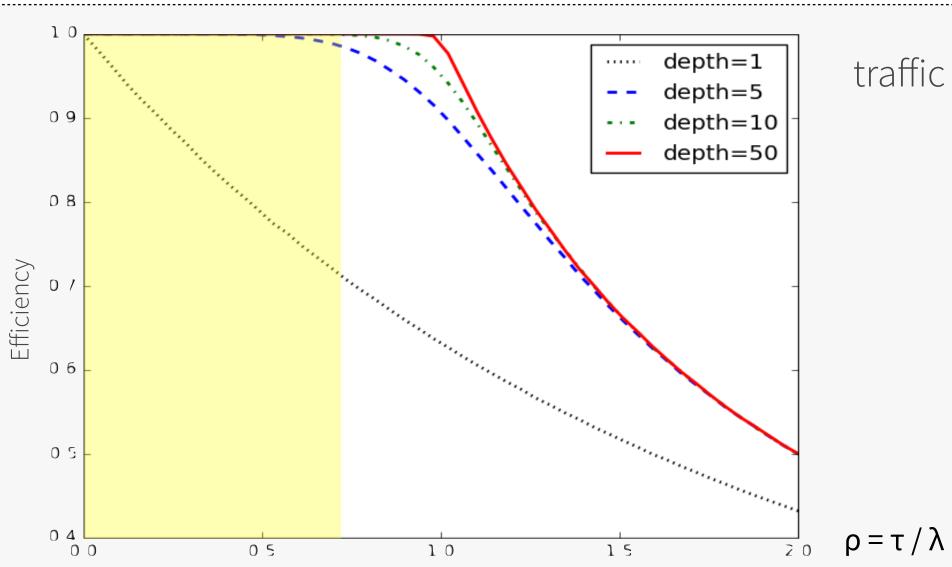
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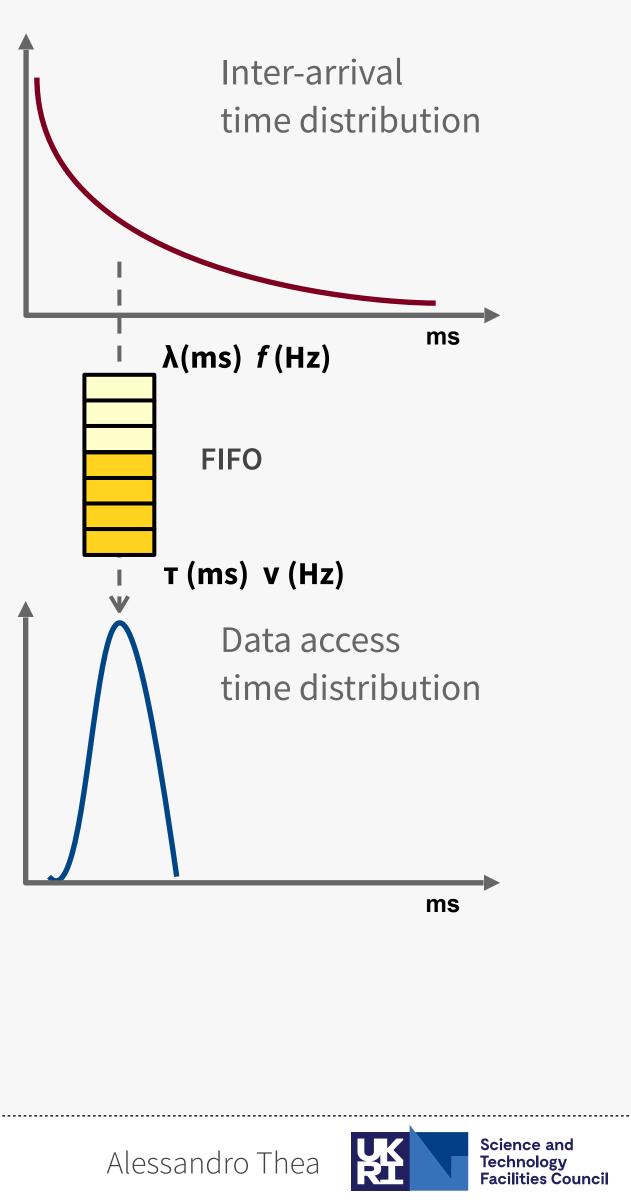




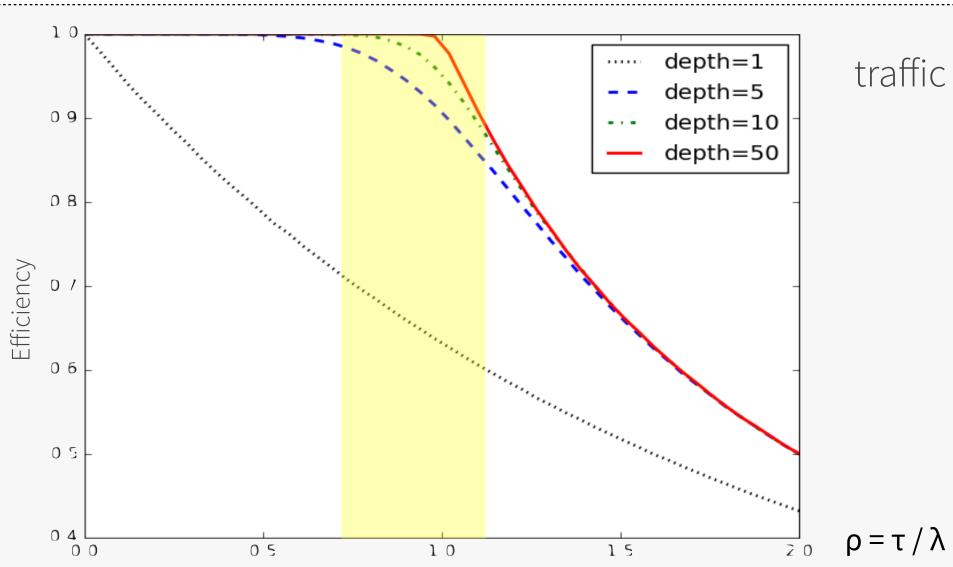
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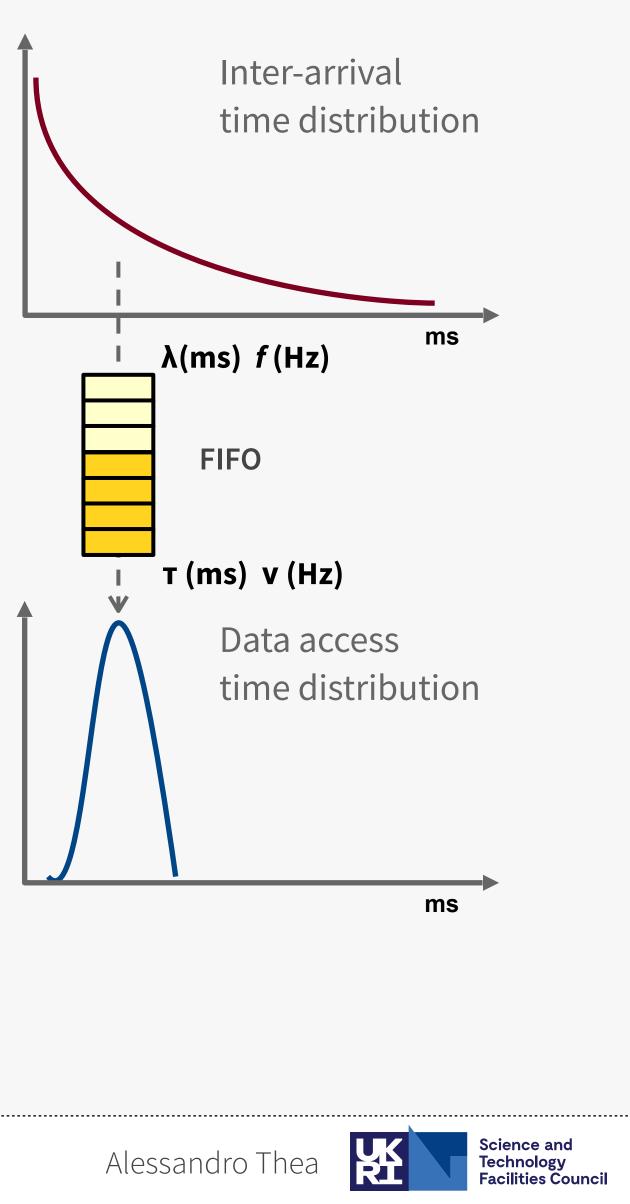




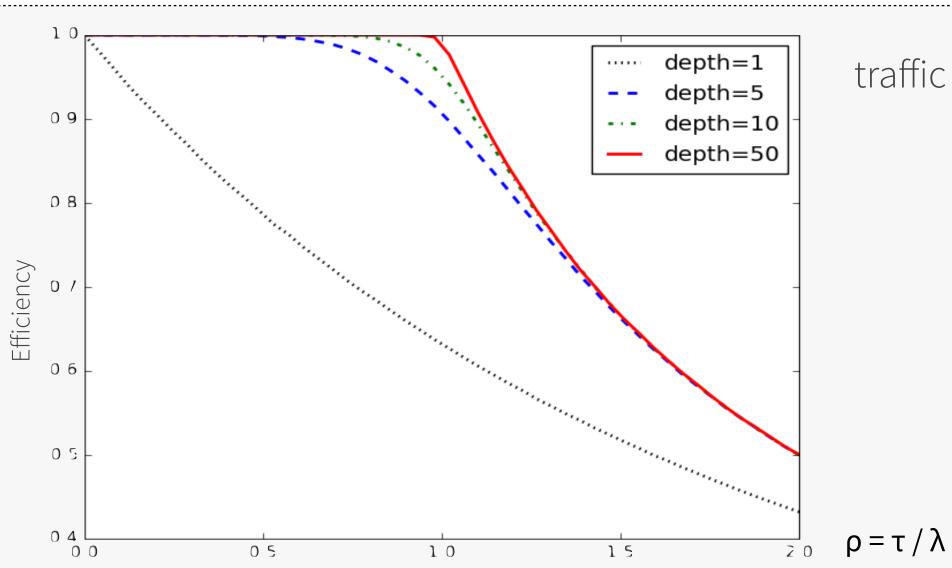
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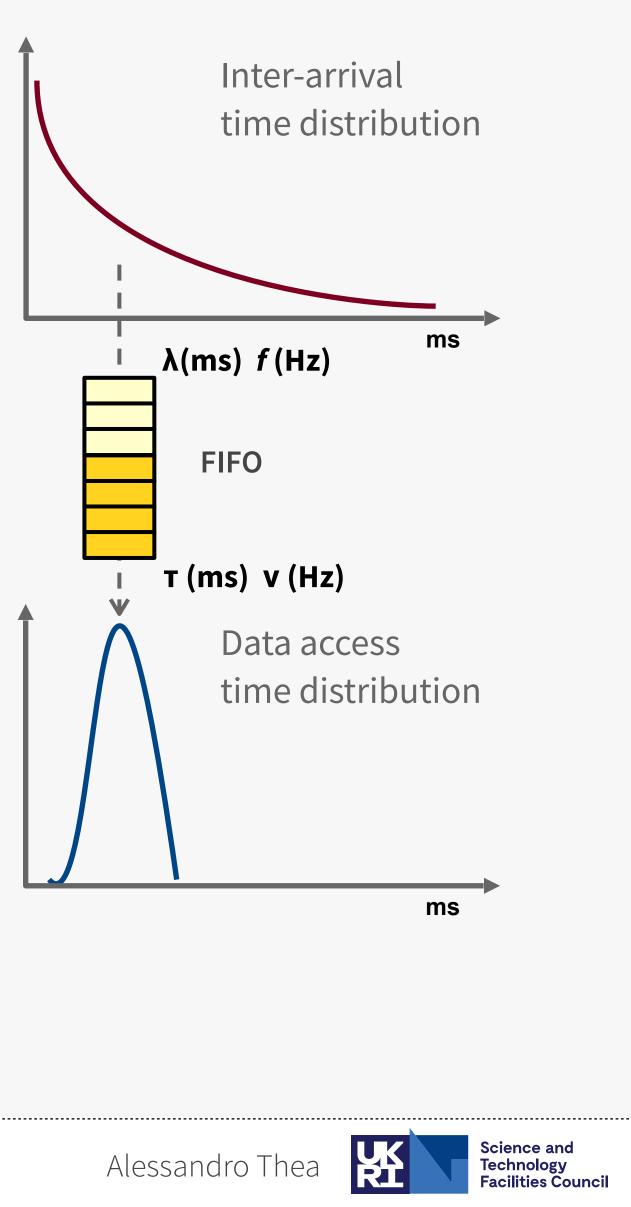




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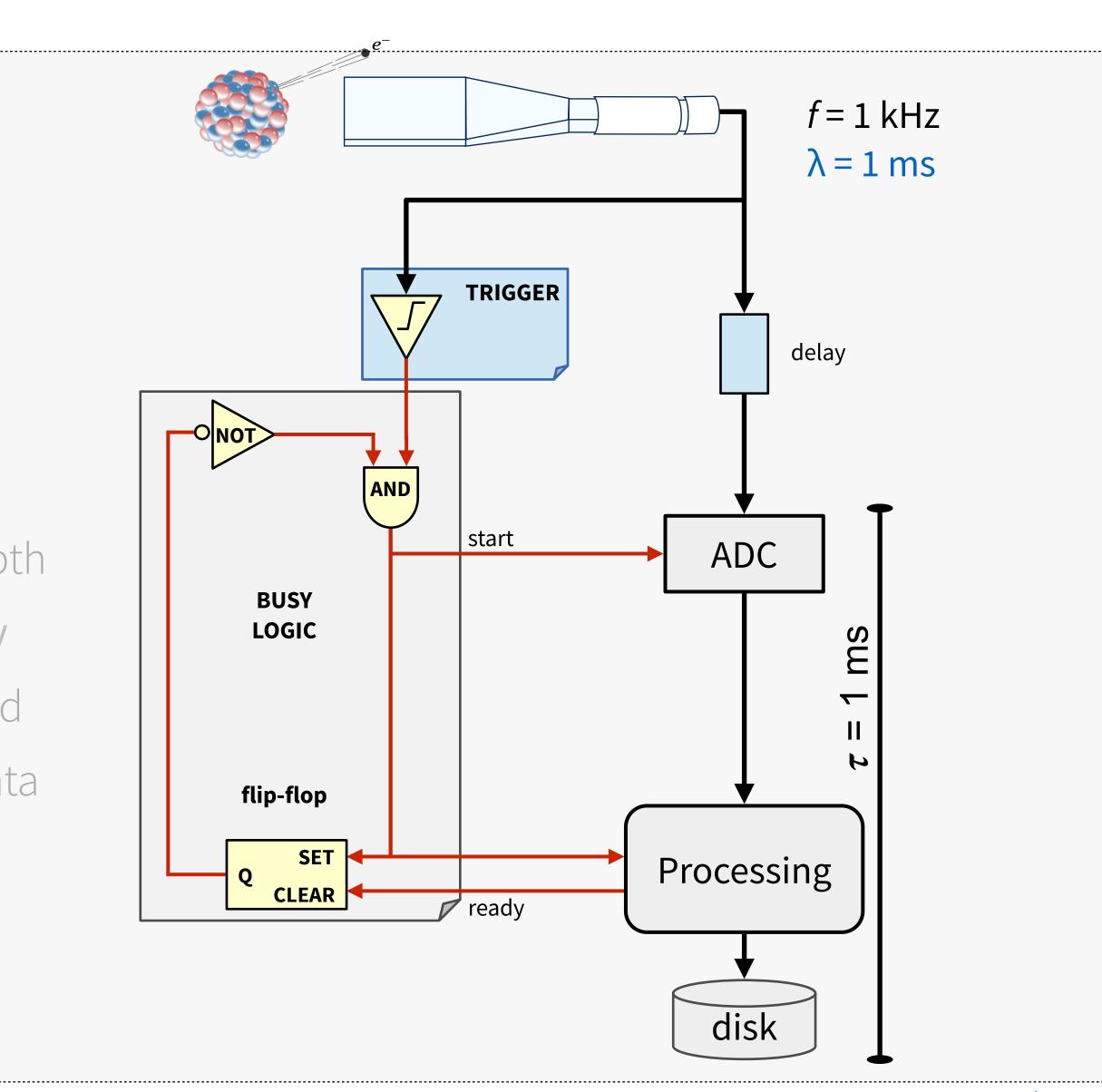




#### Input fluctuations can be absorbed and smoothed by a queue

- A FIFO can provide a ~steady and de-randomized output rate
- The effect of the queue depends on its depth Busy is now defined by the buffer occupancy
- Processor pulls data from the buffer at fixed rate, separating the event receiving and data processing steps





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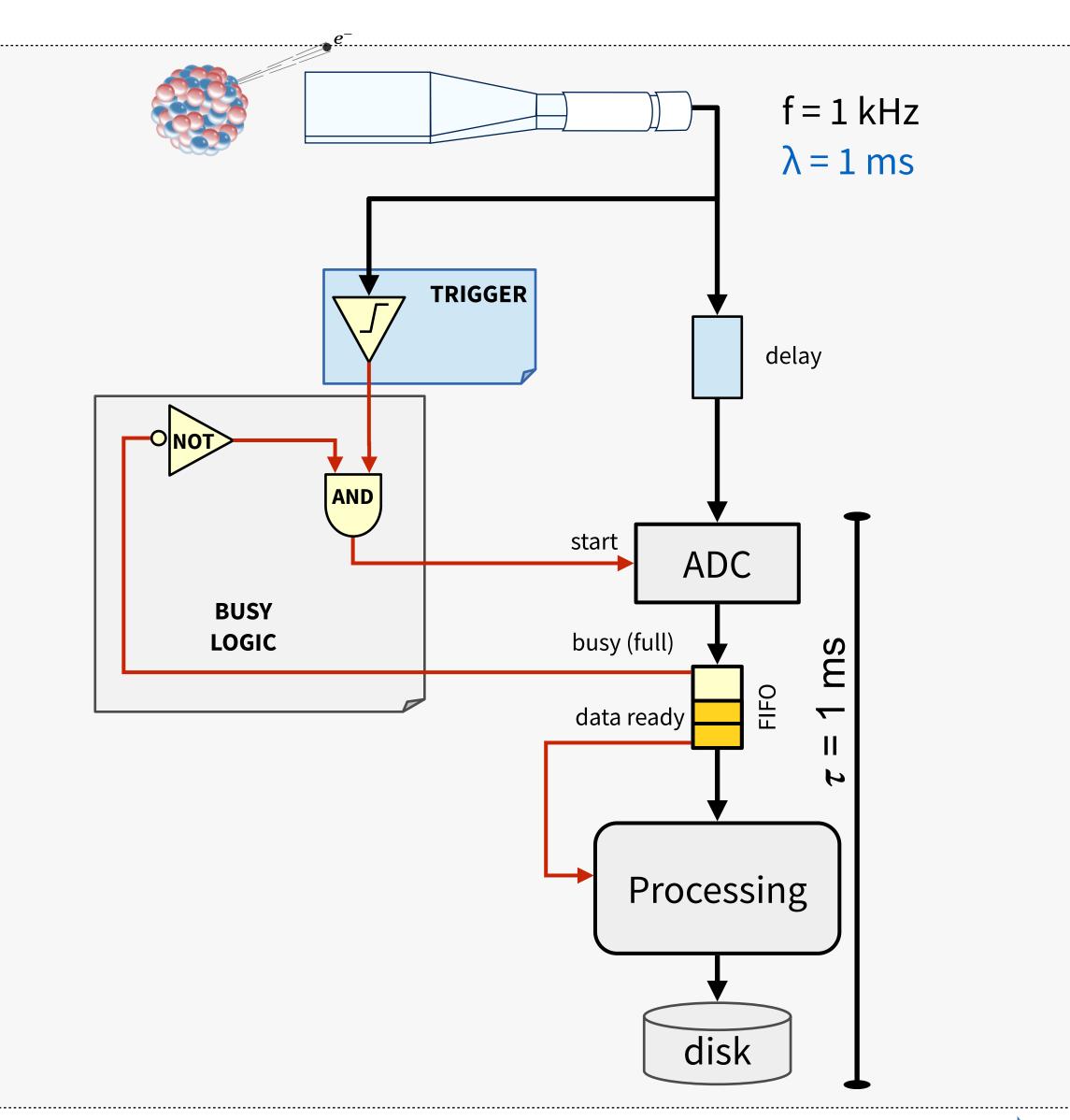




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The FIFO decouples the low latency front-end from the data processing

Minimize the amount of "unnecessary" fast components

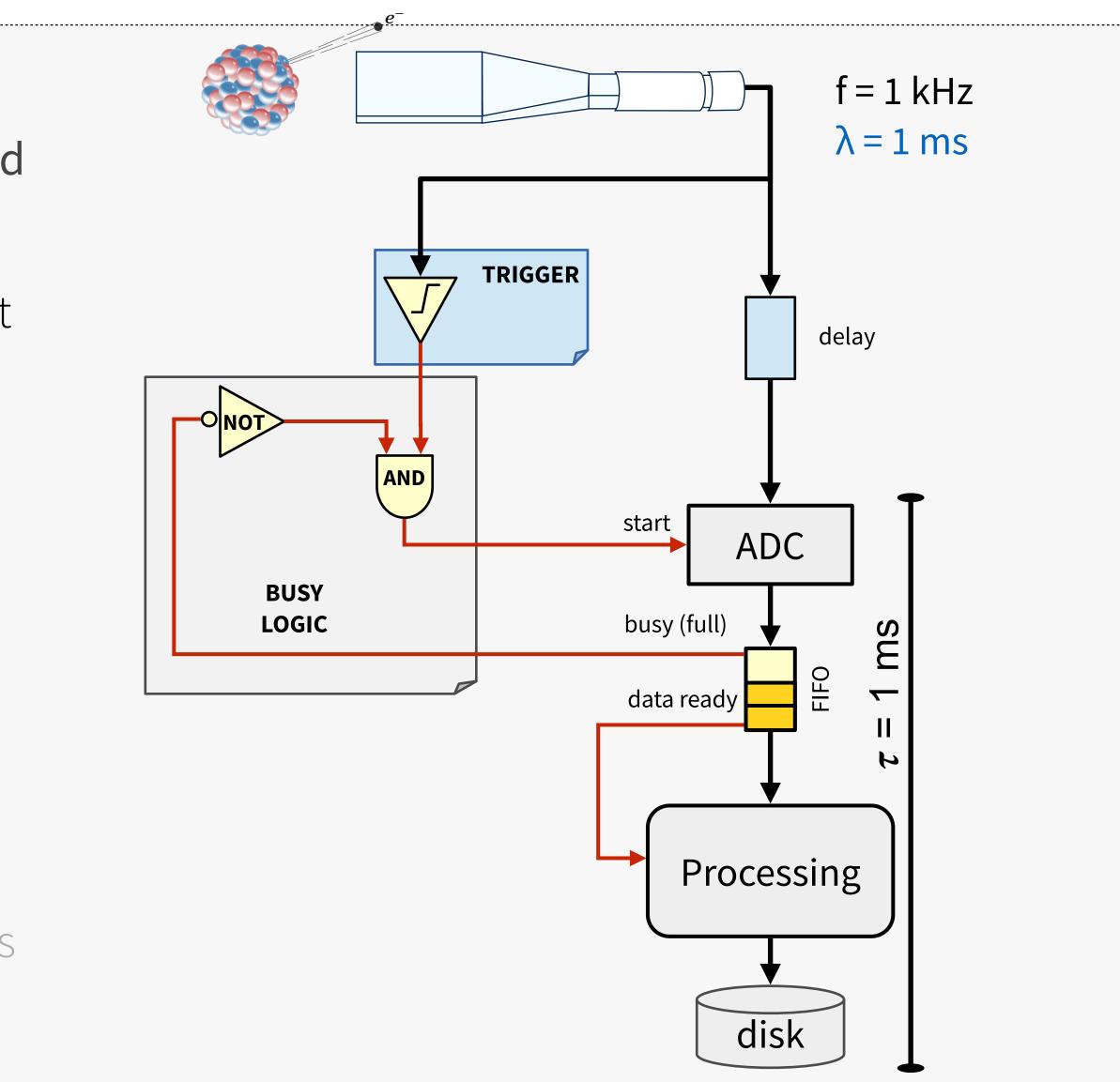
~100% efficiency w/ minimal deadtime achievable if

- ADC can operate at rate  $\gg f$
- Data processing and storage

operate at a rate  $\sim f$ 

Could the delay be replaced with a "FIFO"?

• Analog pipelines, heavily used in LHC DAQs







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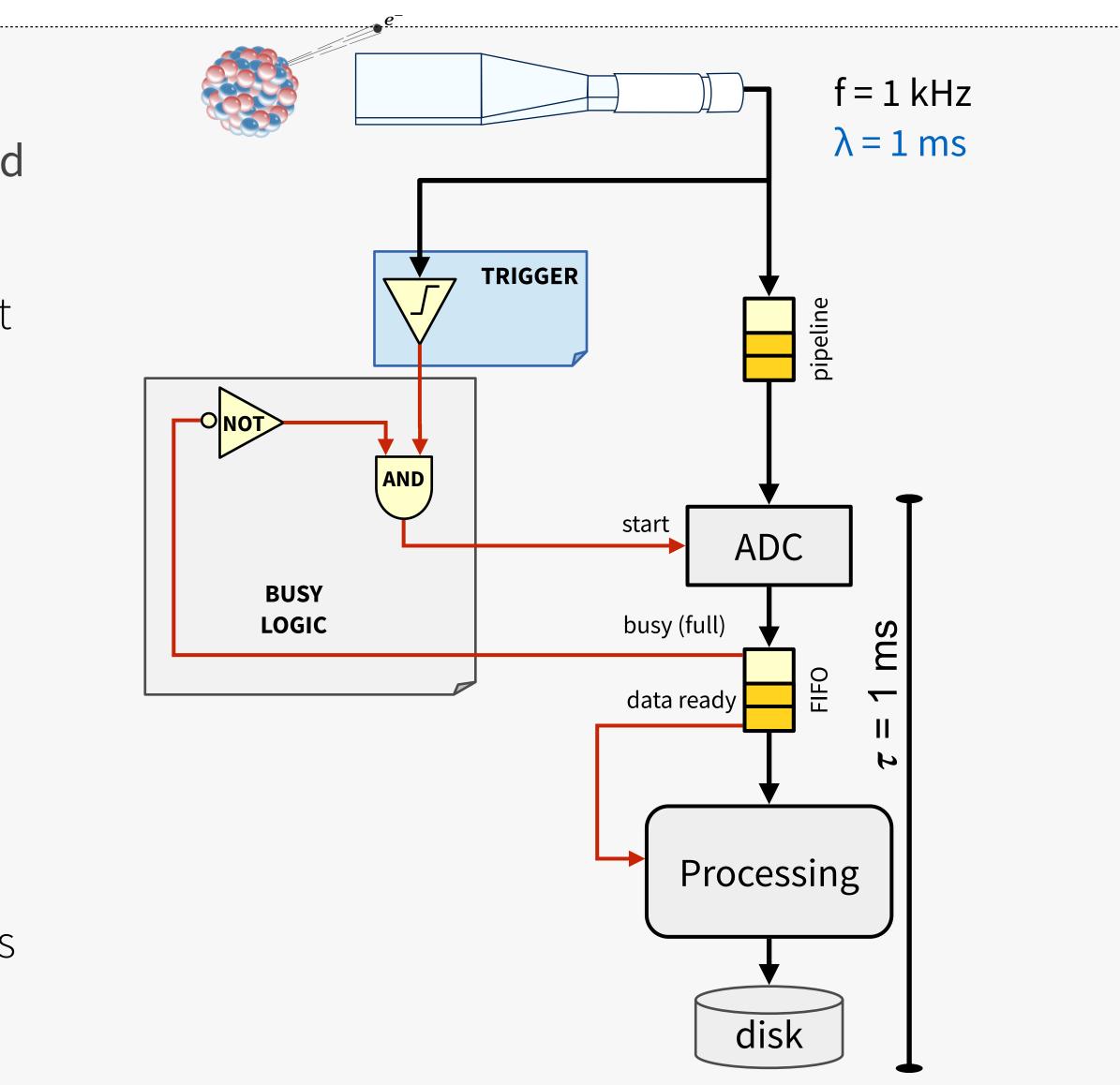
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# Collider setup

Do we need de-randomization buffers also in collider setups?

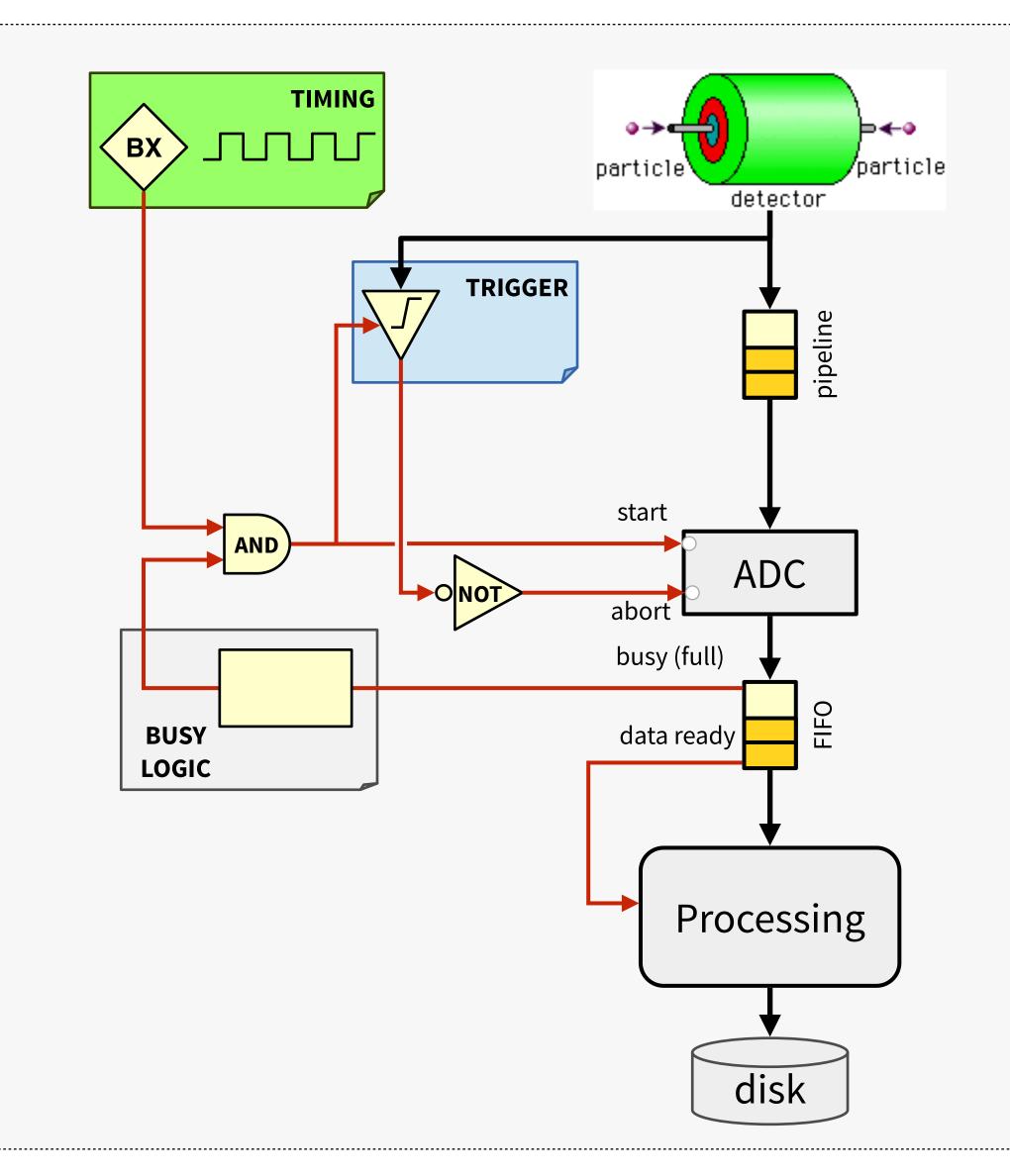
- Particle collisions are synchronous
- But the time distribution of triggers is random: interesting events are unpredictable

De-randomization still needed

More complex busy logic to protect buffers and detectors

- Eg: accept n events every m bunch crossings
- Eg: prevent some dangerous trigger patterns









# Collider setup

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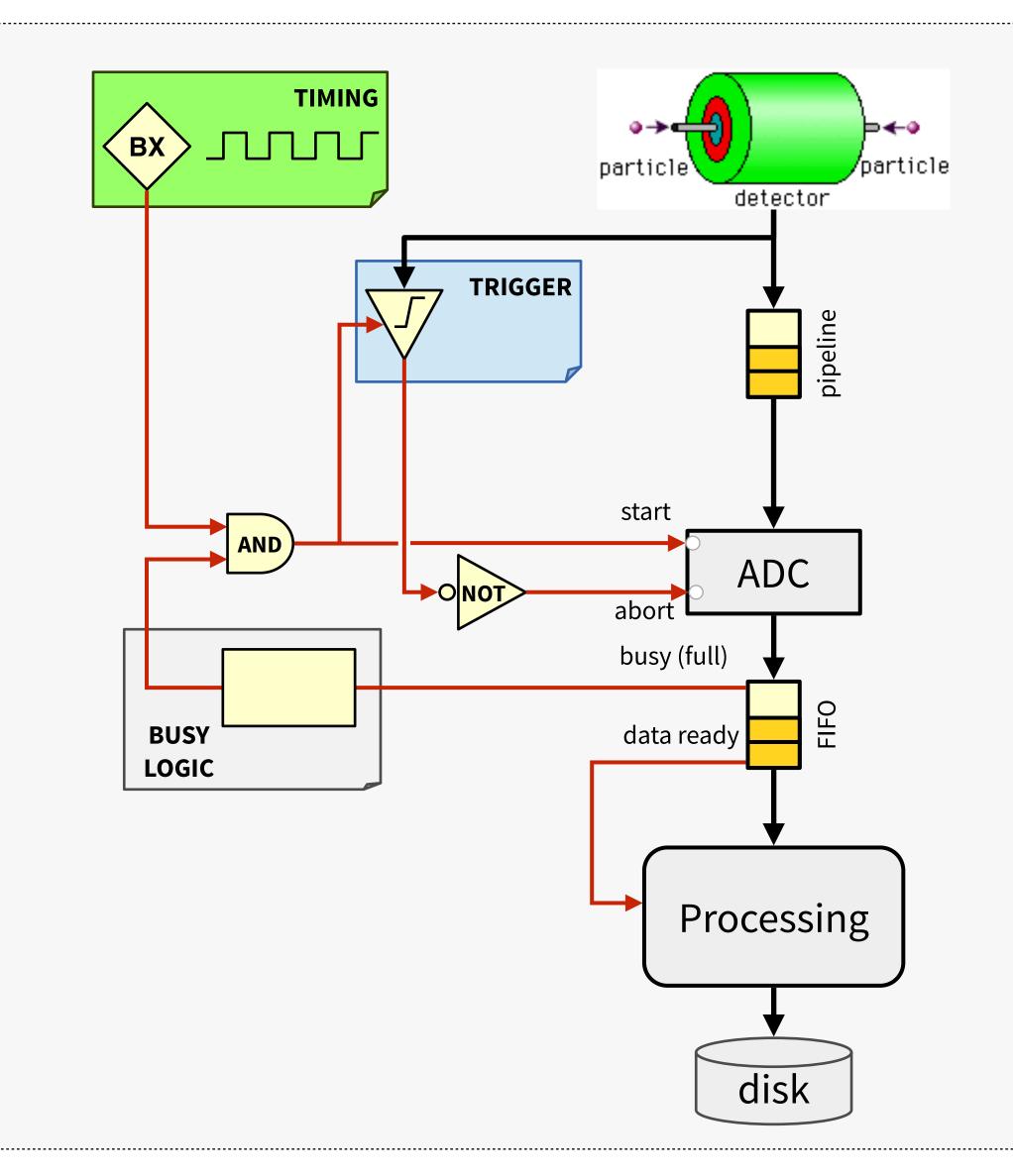
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#### Outline

1. Introduction1.1. What is DAQ?1.2. System architecture

2. Basic DAQ concepts
2.1. Digitization, Latency
2.2. Deadtime, Busy, Backpressure
2.3. De-randomization

#### 3. Scaling up

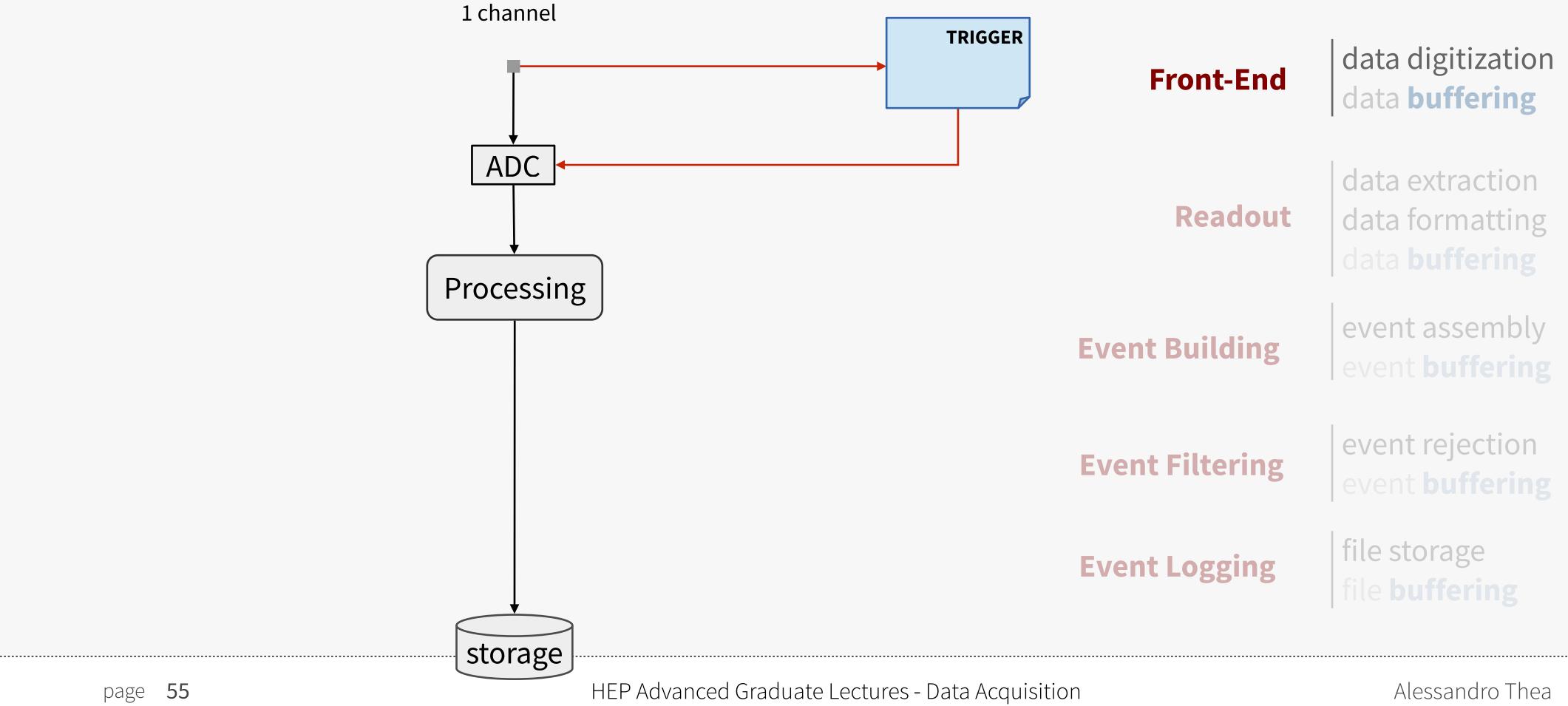
- 3.1. Readout and Event Building
- 3.2. Buses vs Network

4. DAQ Challenges at the LHC







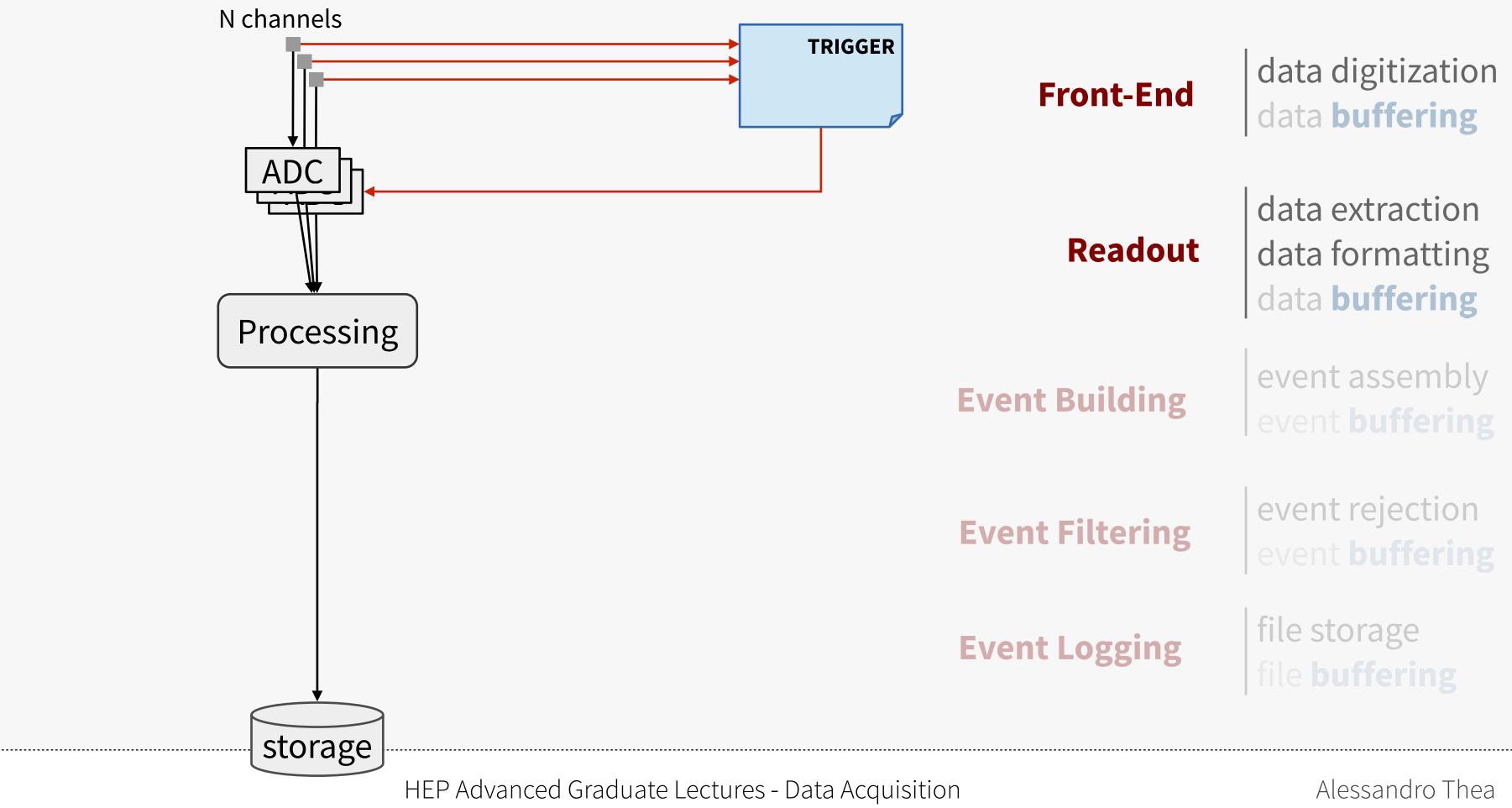






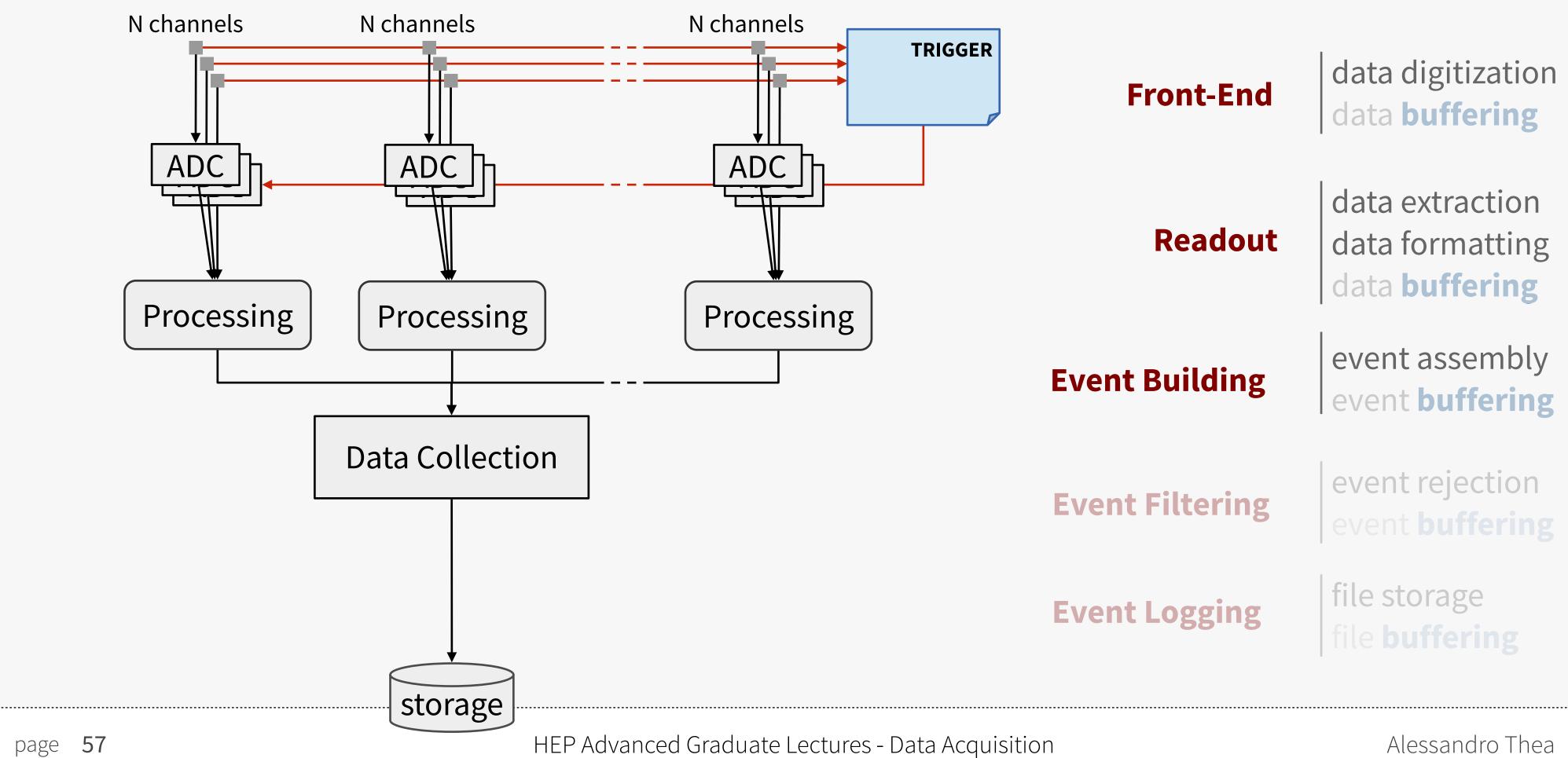
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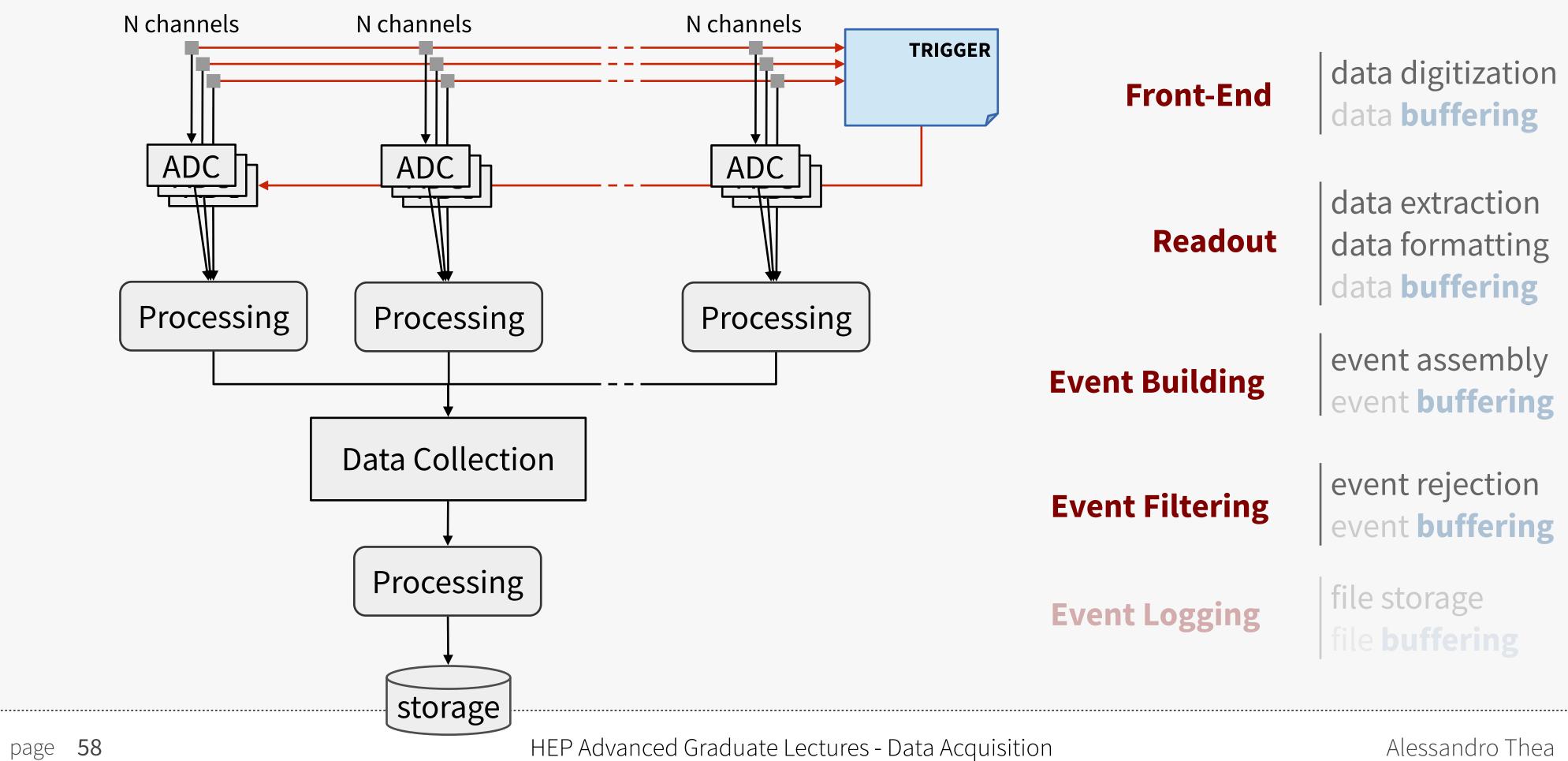










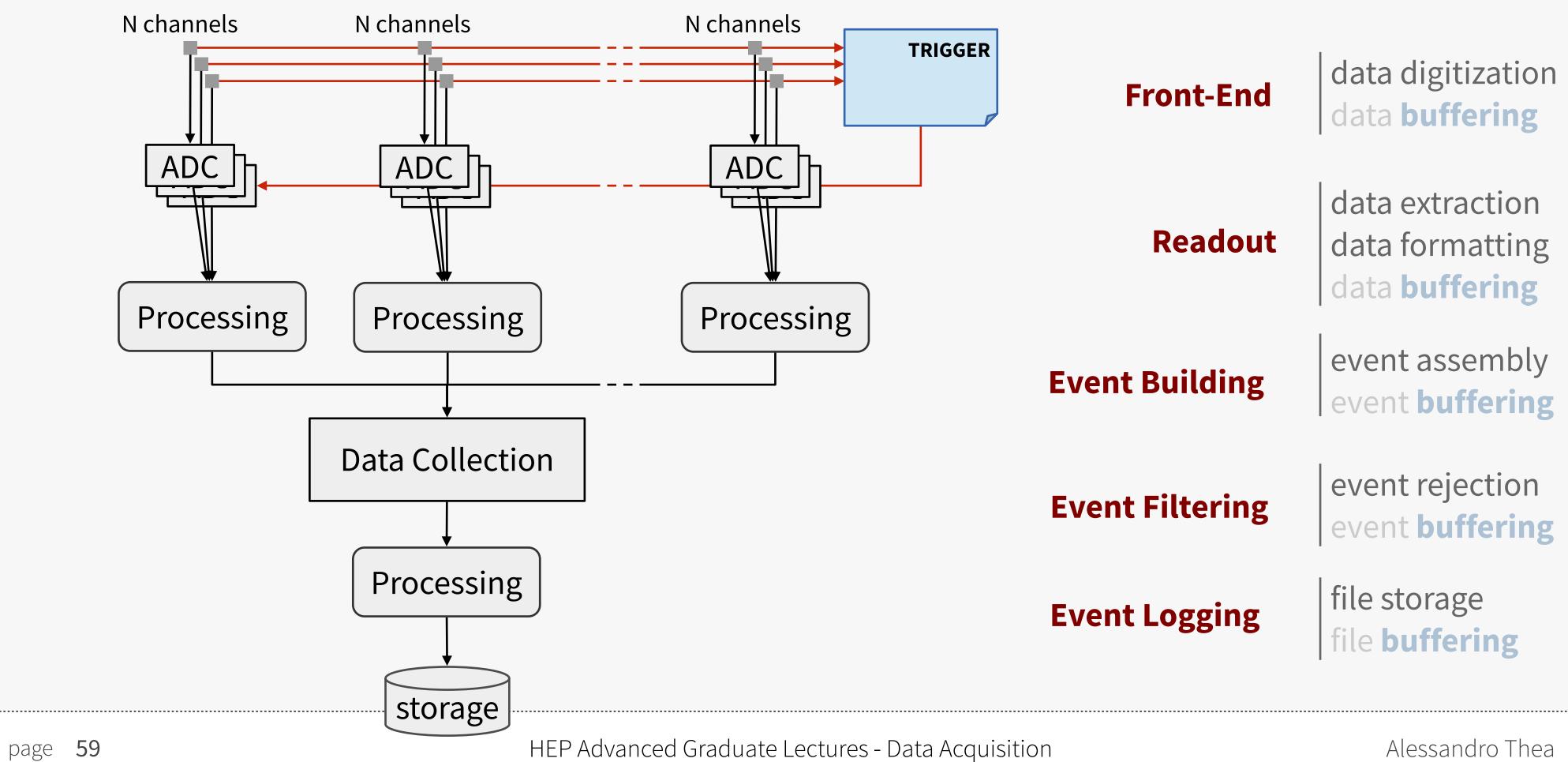






## Adding more channels

Adding more channels requires a hierarchical structure committed to the data handling and conveyance



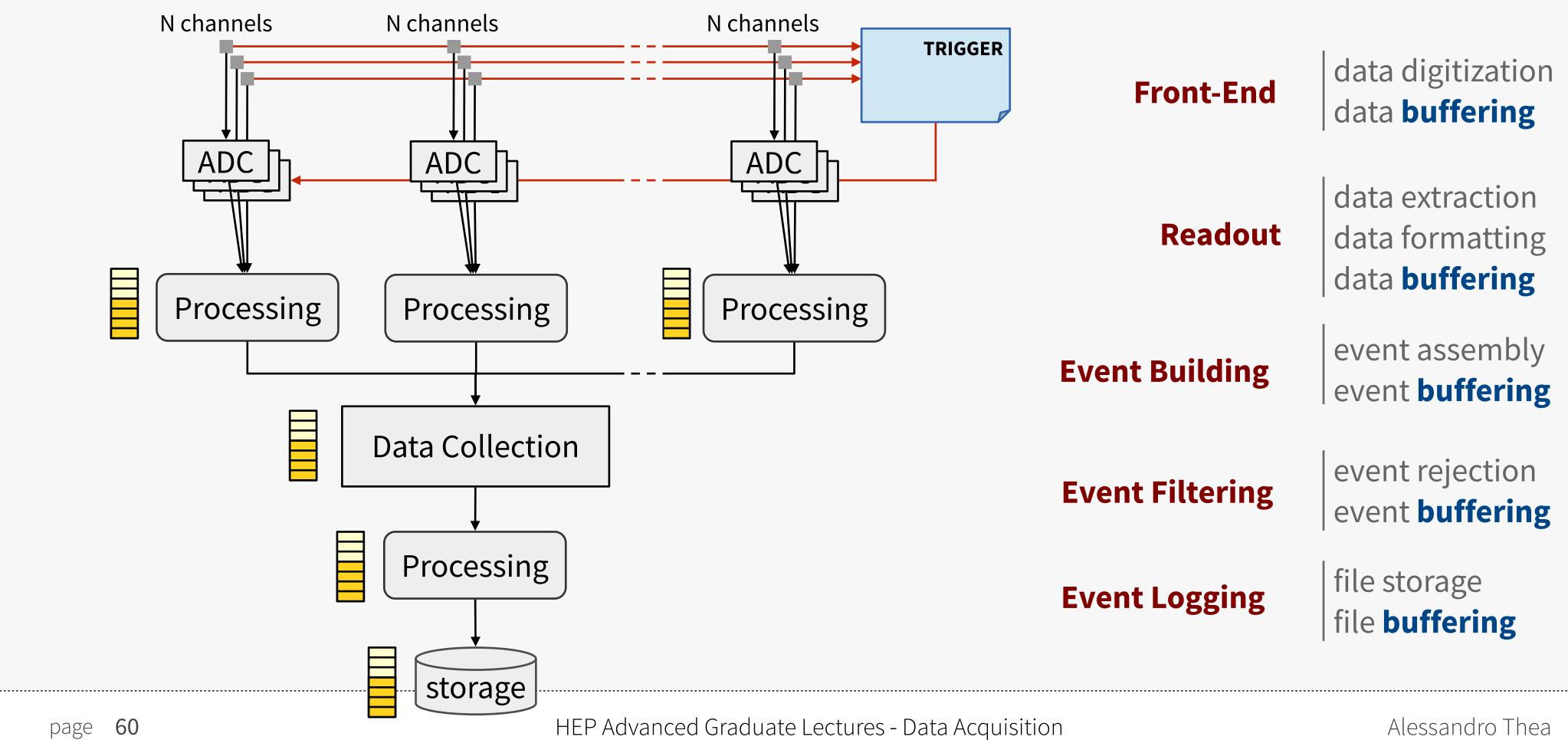




## Adding more channels

### **Buffering** usually needed at every level

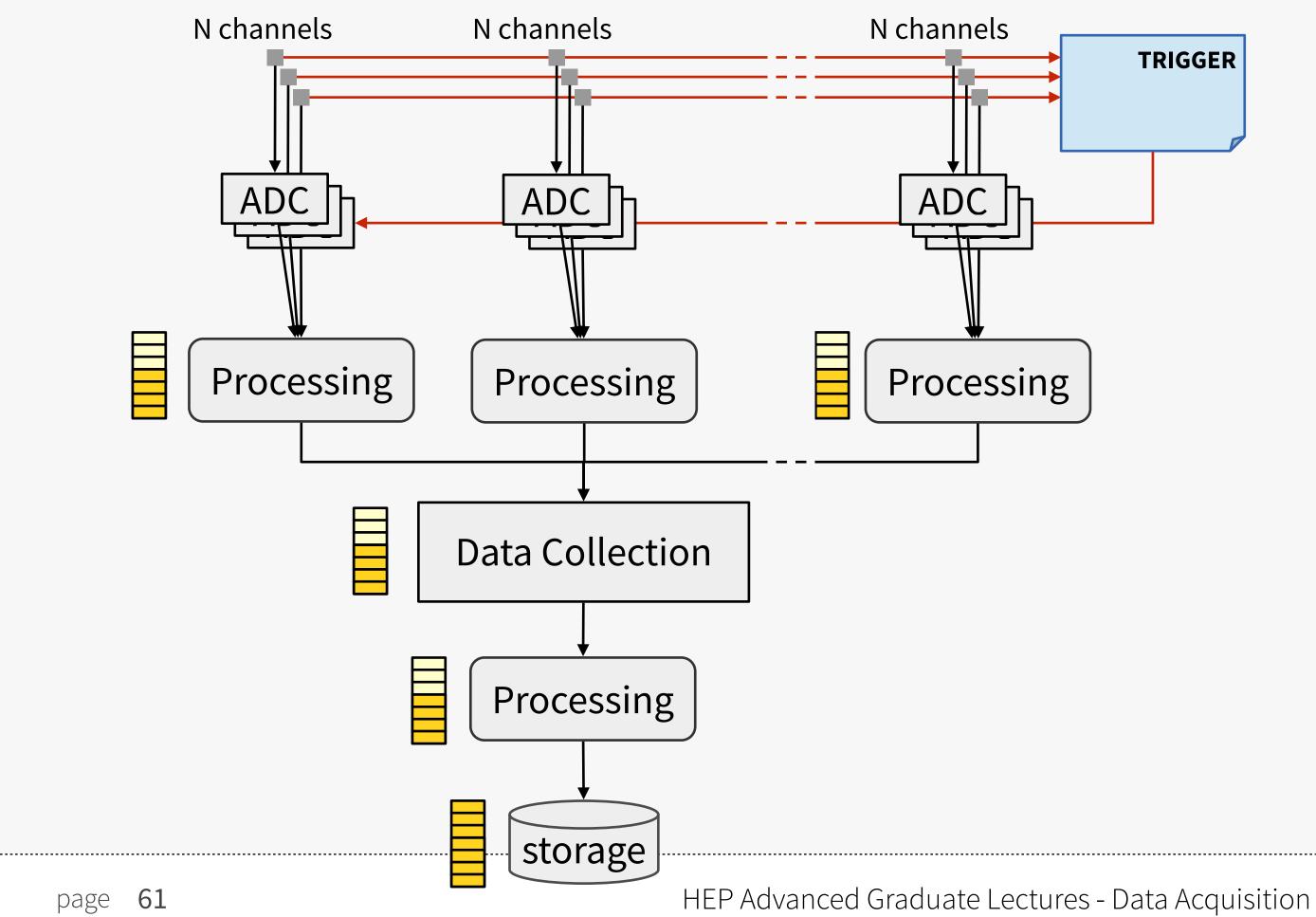
• DAQ can be seen as a multi level buffering system







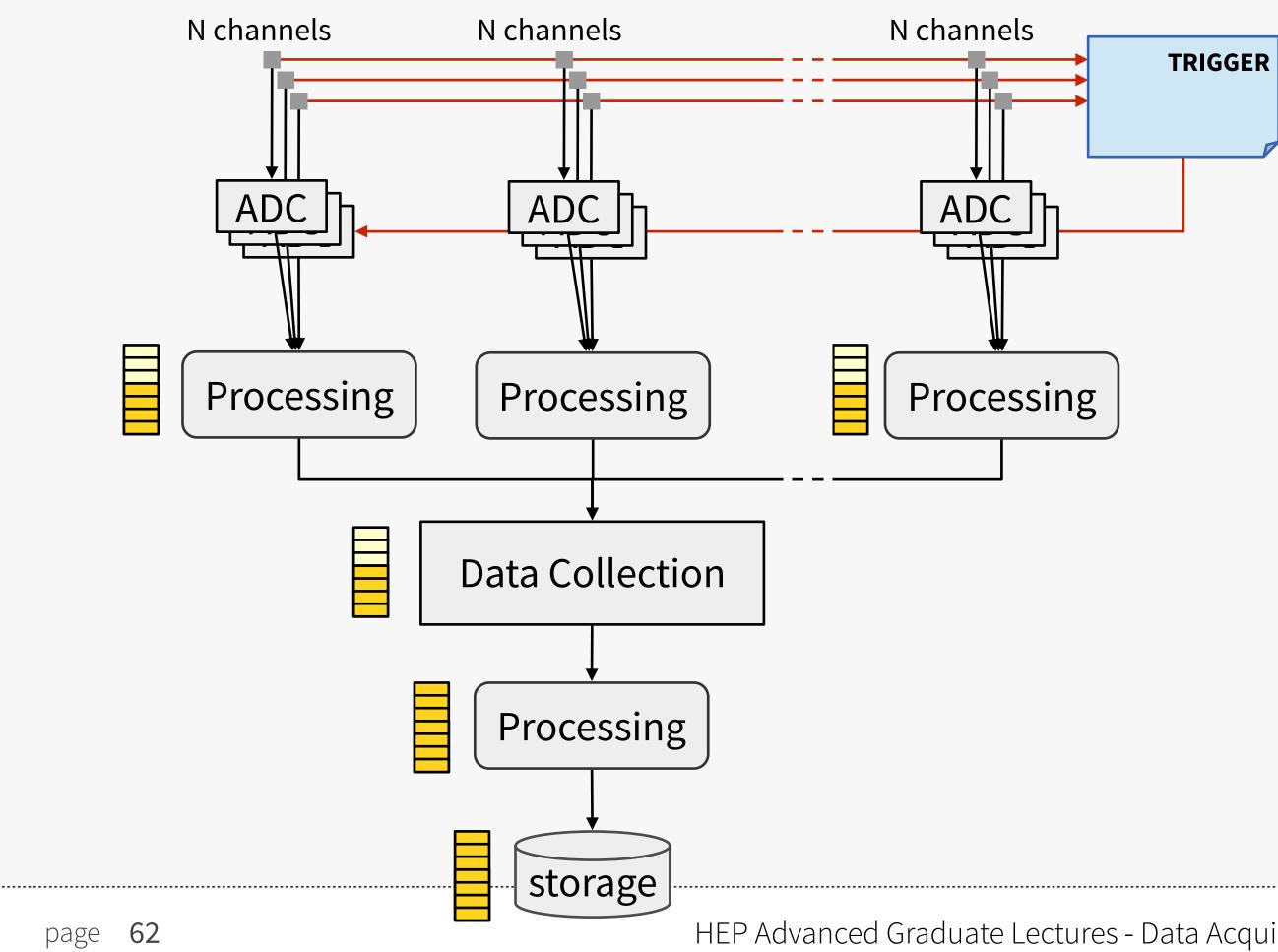
#### If a system/buffer gets saturated







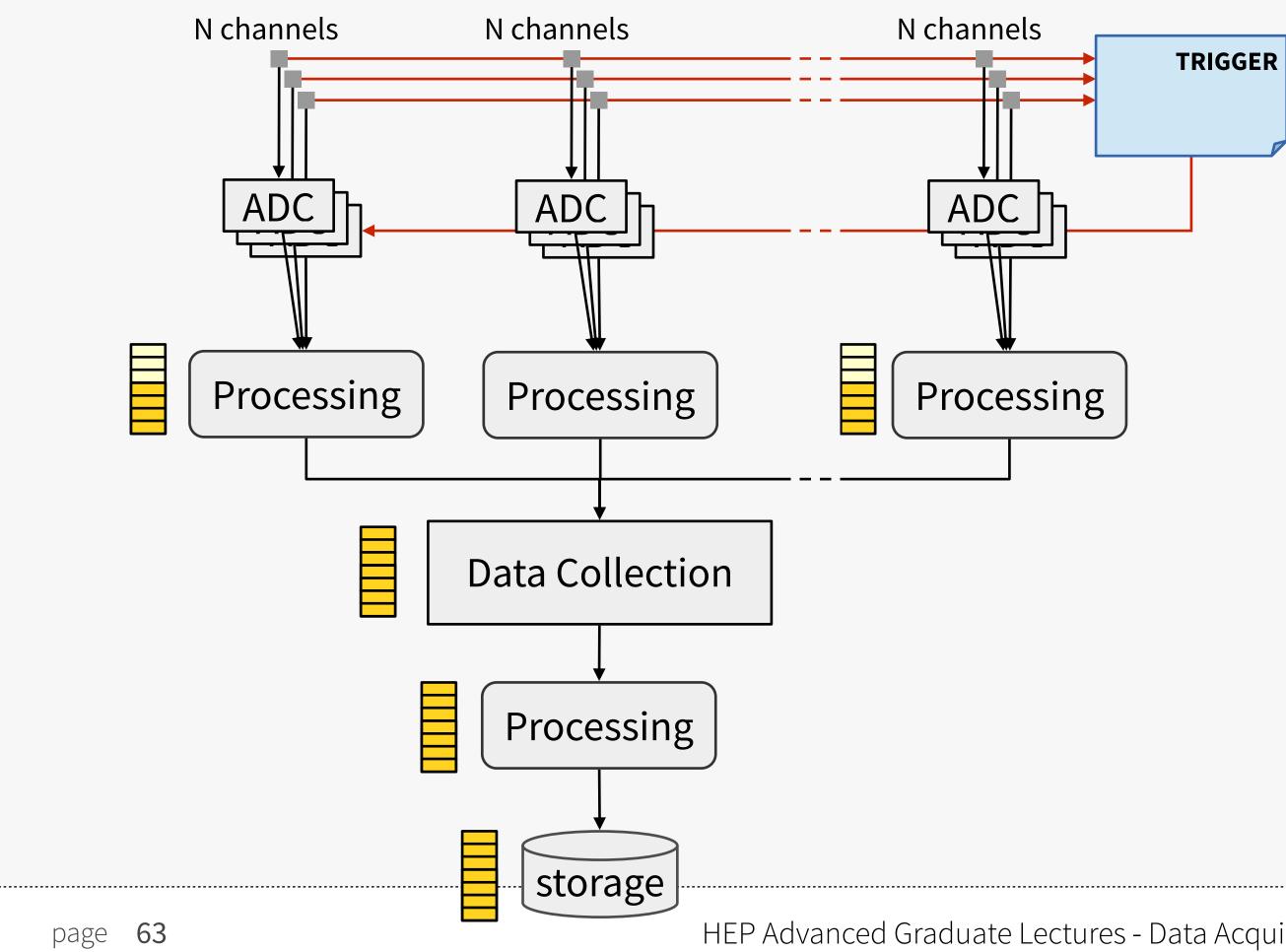
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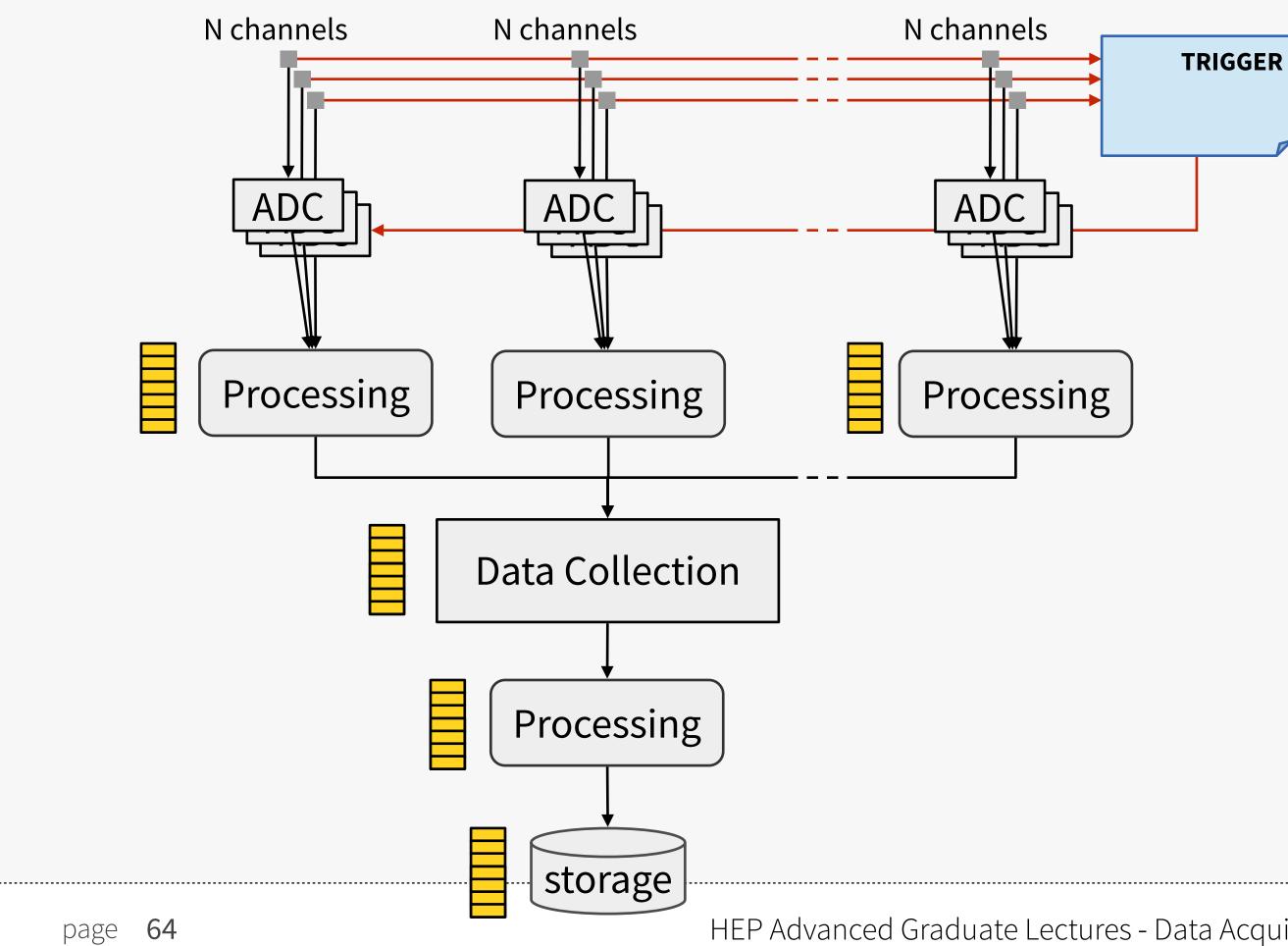
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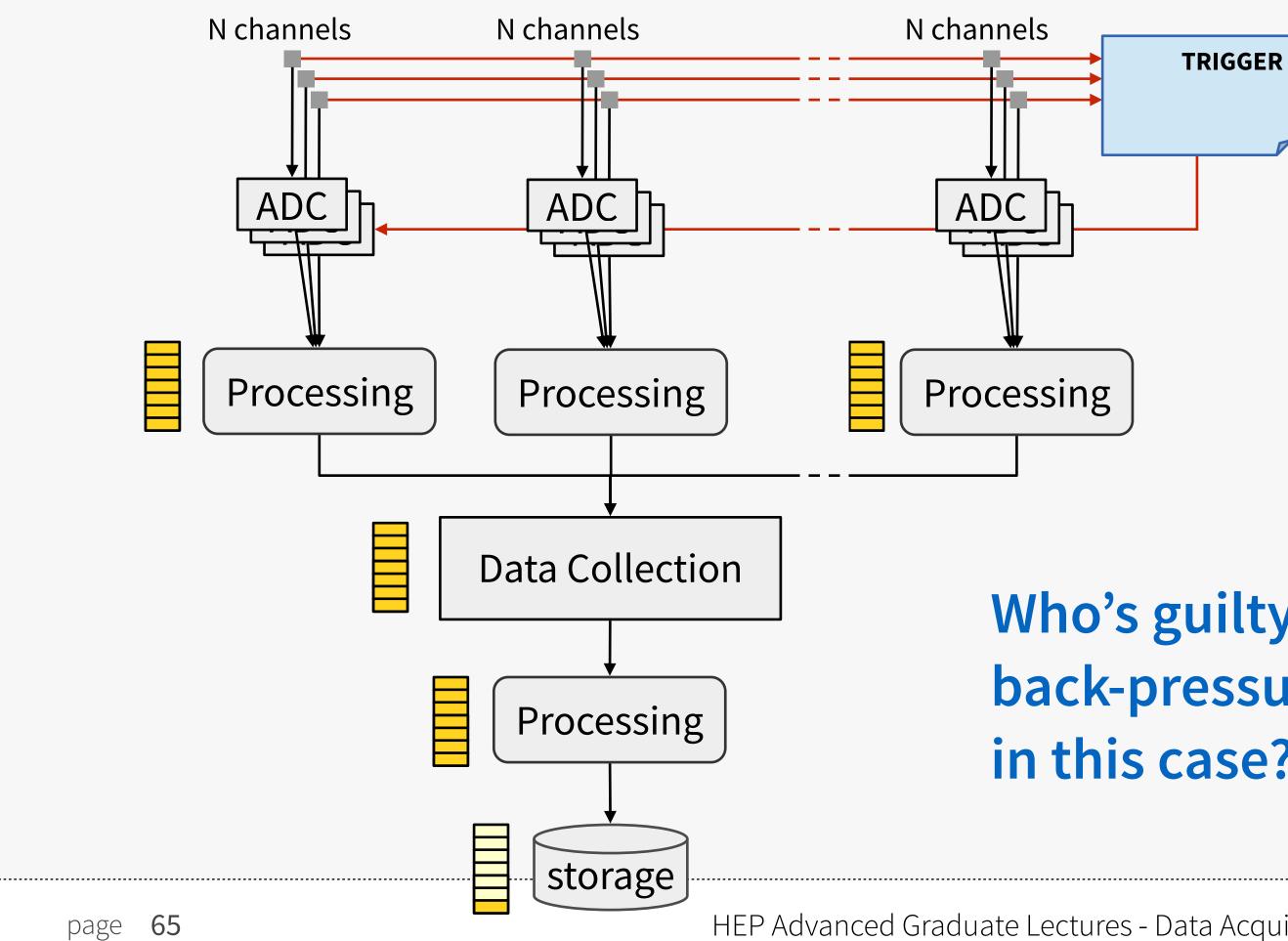
- Up to exert **busy** to the trigger system
- **Debugging**: where is the source of back-pressure?
  - follow the buffers occupancy via the monitoring system





#### If a system/buffer gets saturated

• the "pressure" is propagated upstream (**back-pressure**)



- Up to exert **busy** to the trigger system
- **Debugging**: where is the source of back-pressure?
  - follow the buffers occupancy via the monitoring system

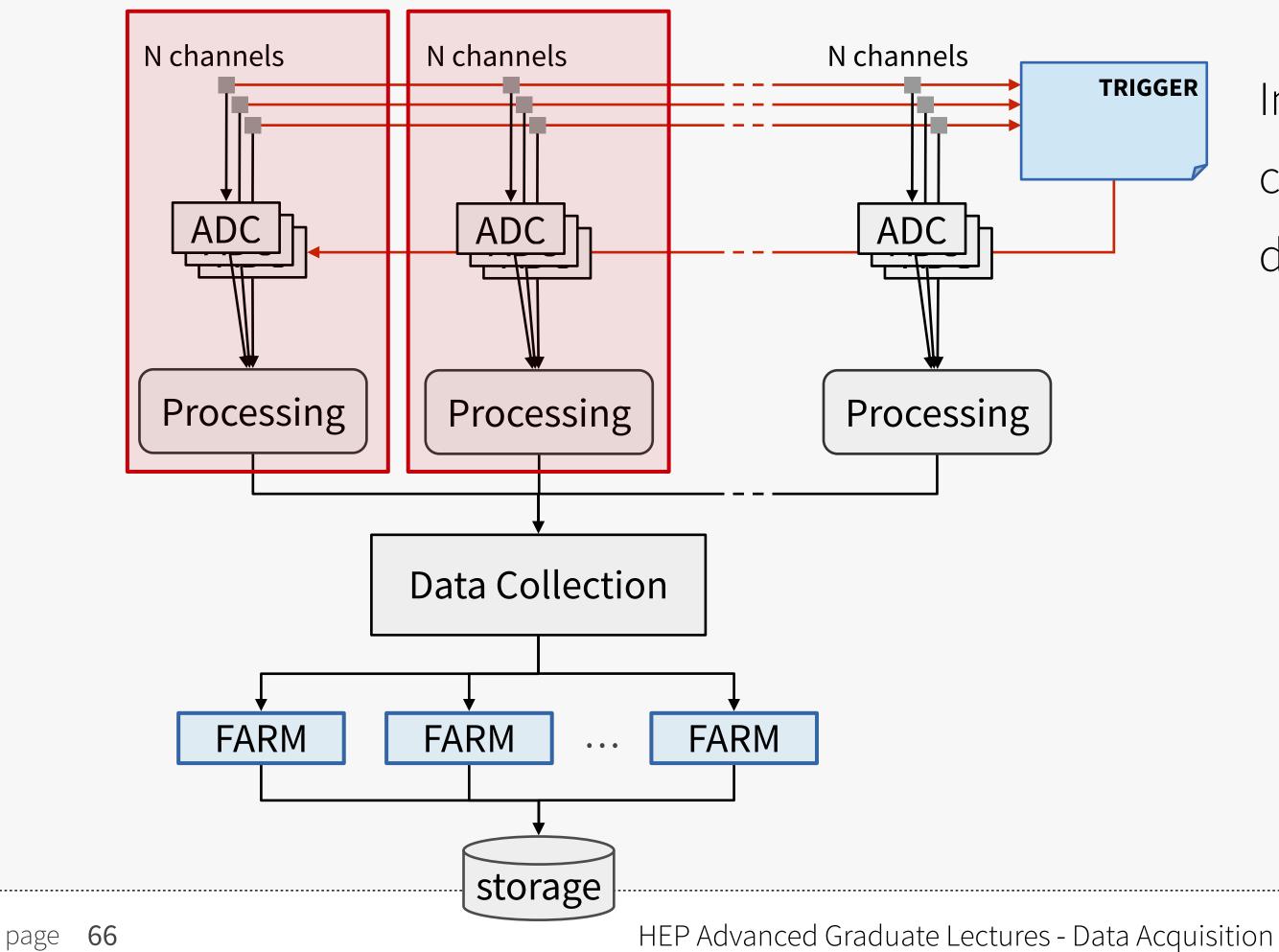
### Who's guilty of back-pressure in this case?





## Building blocks

Reading out data or building events out of many channels requires many components



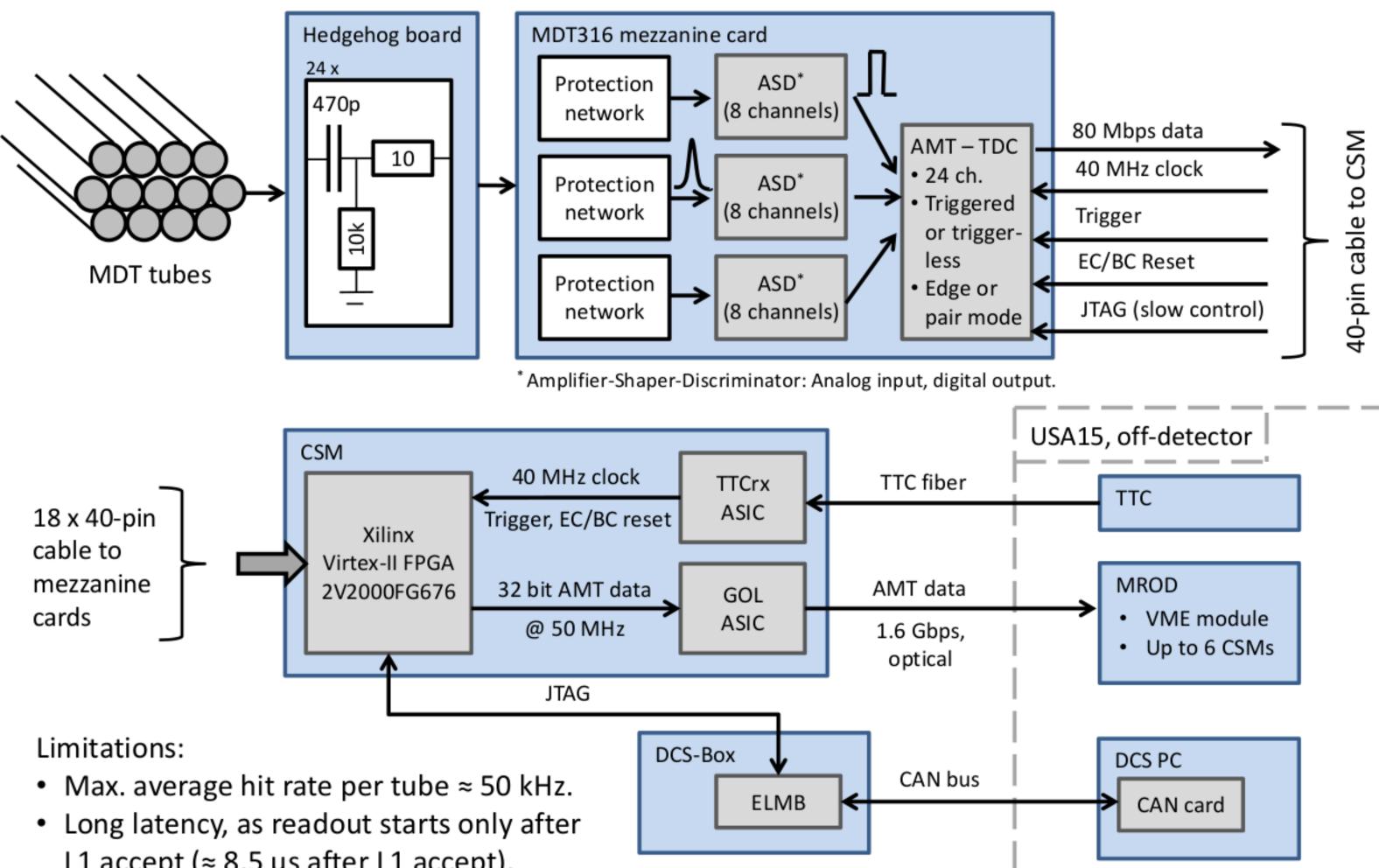
In the design of our hierarchical datacollection system, it's convenient to define "building blocks"

- Readout crates
- HLT racks
- event building groups
- daq slices





## Front End electronics



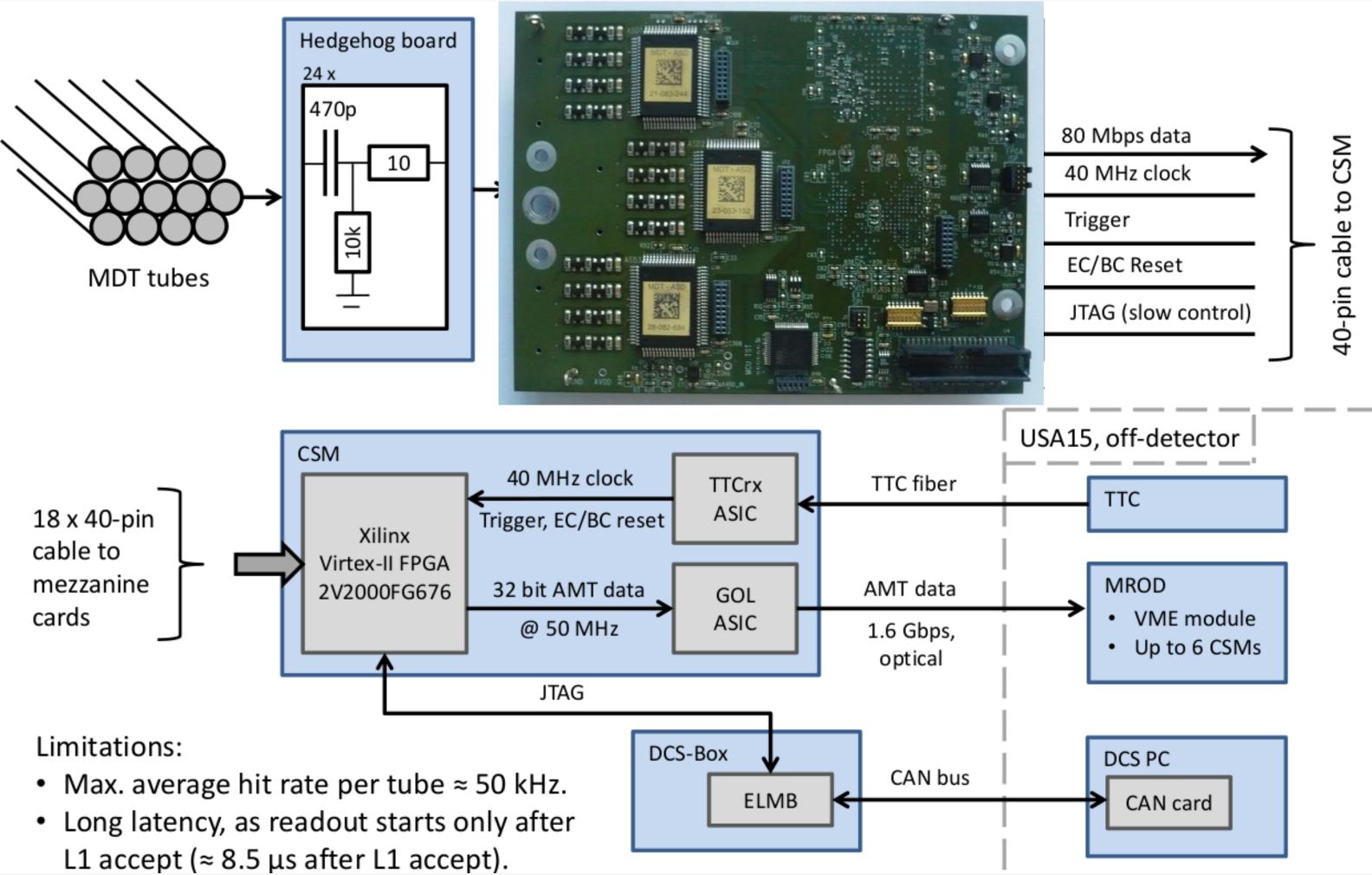
- L1 accept (≈ 8.5 µs after L1 accept).

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## Front End electronics

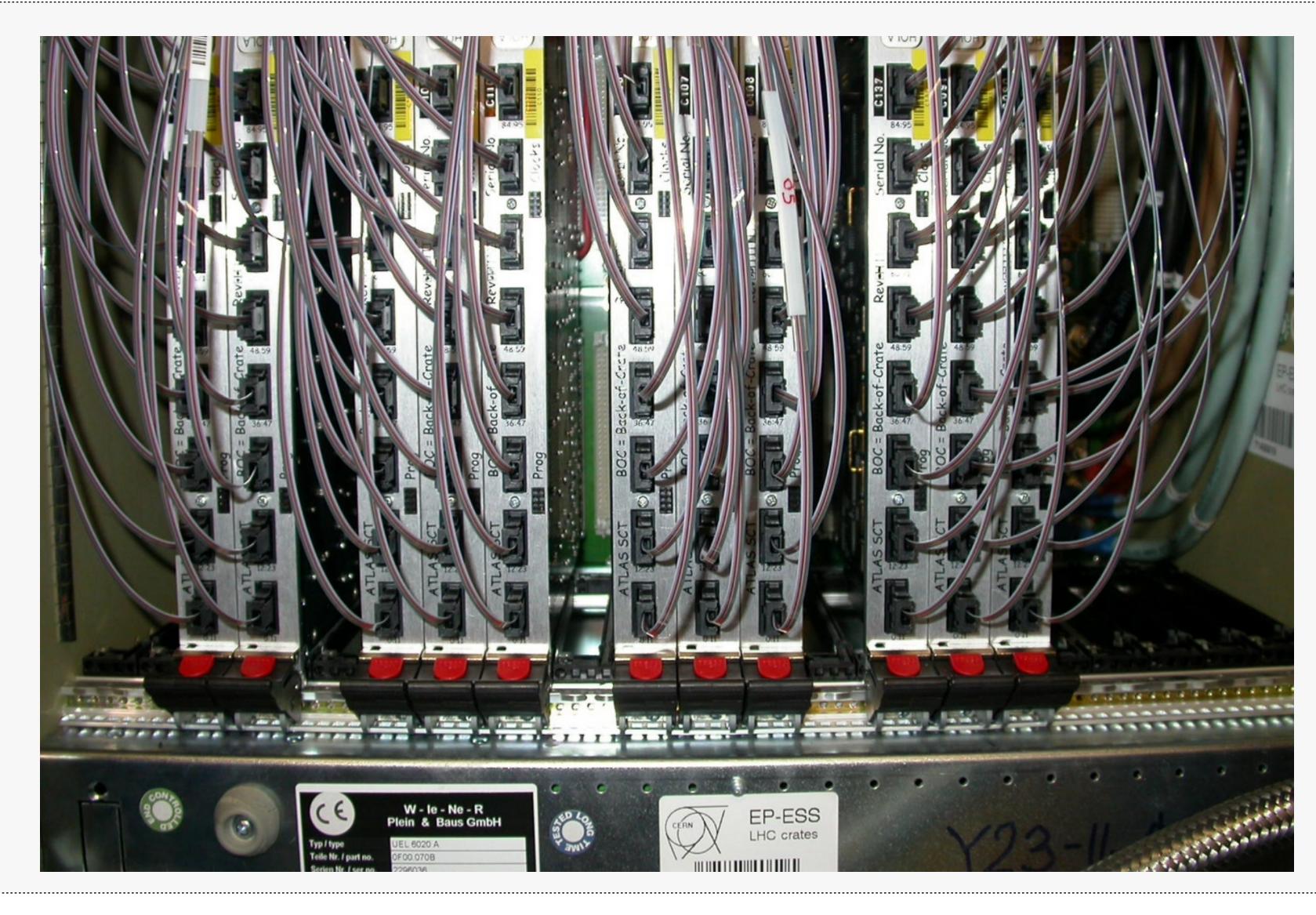


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## Readout Boards (Counting Room)



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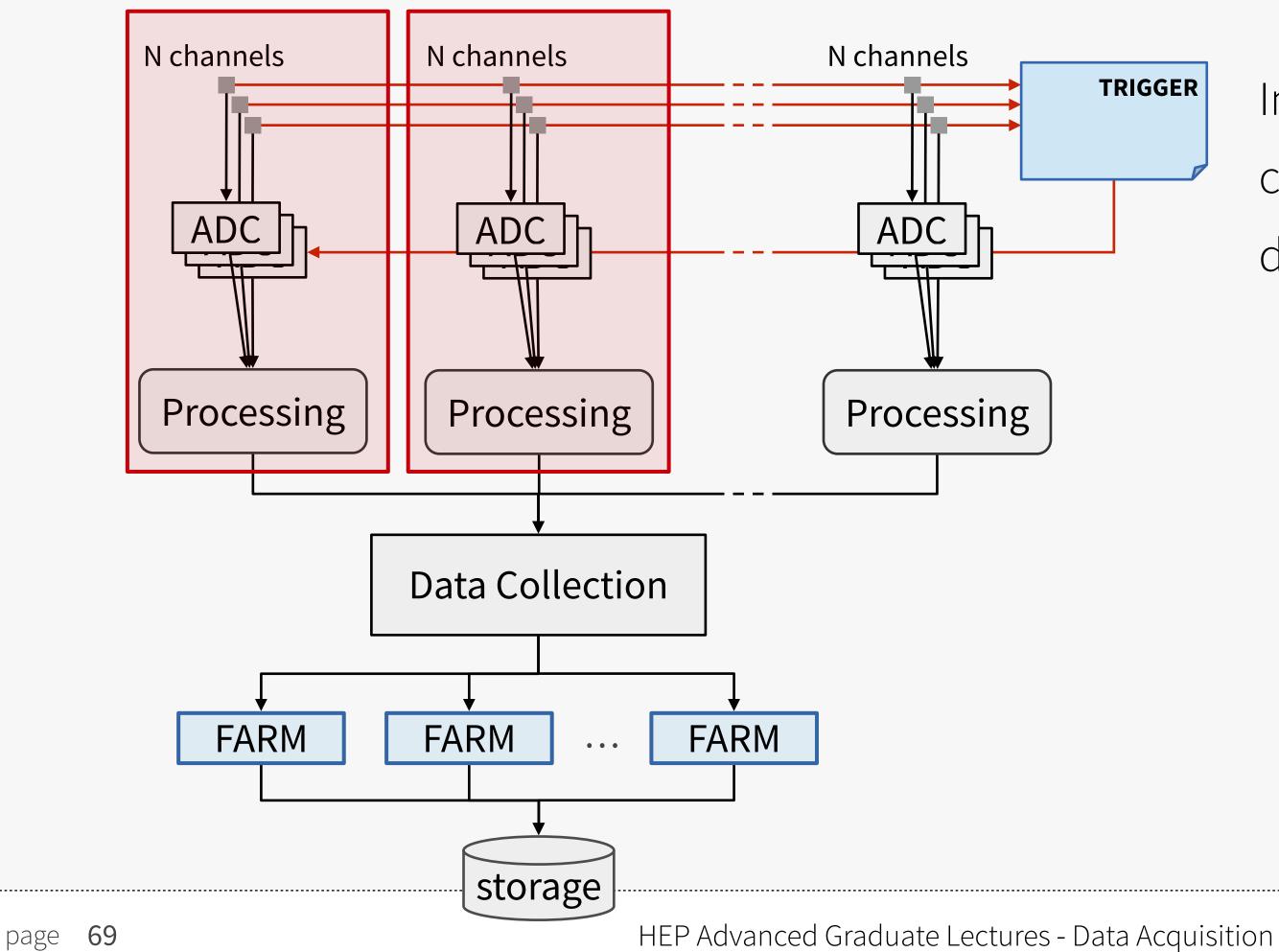






## Building blocks

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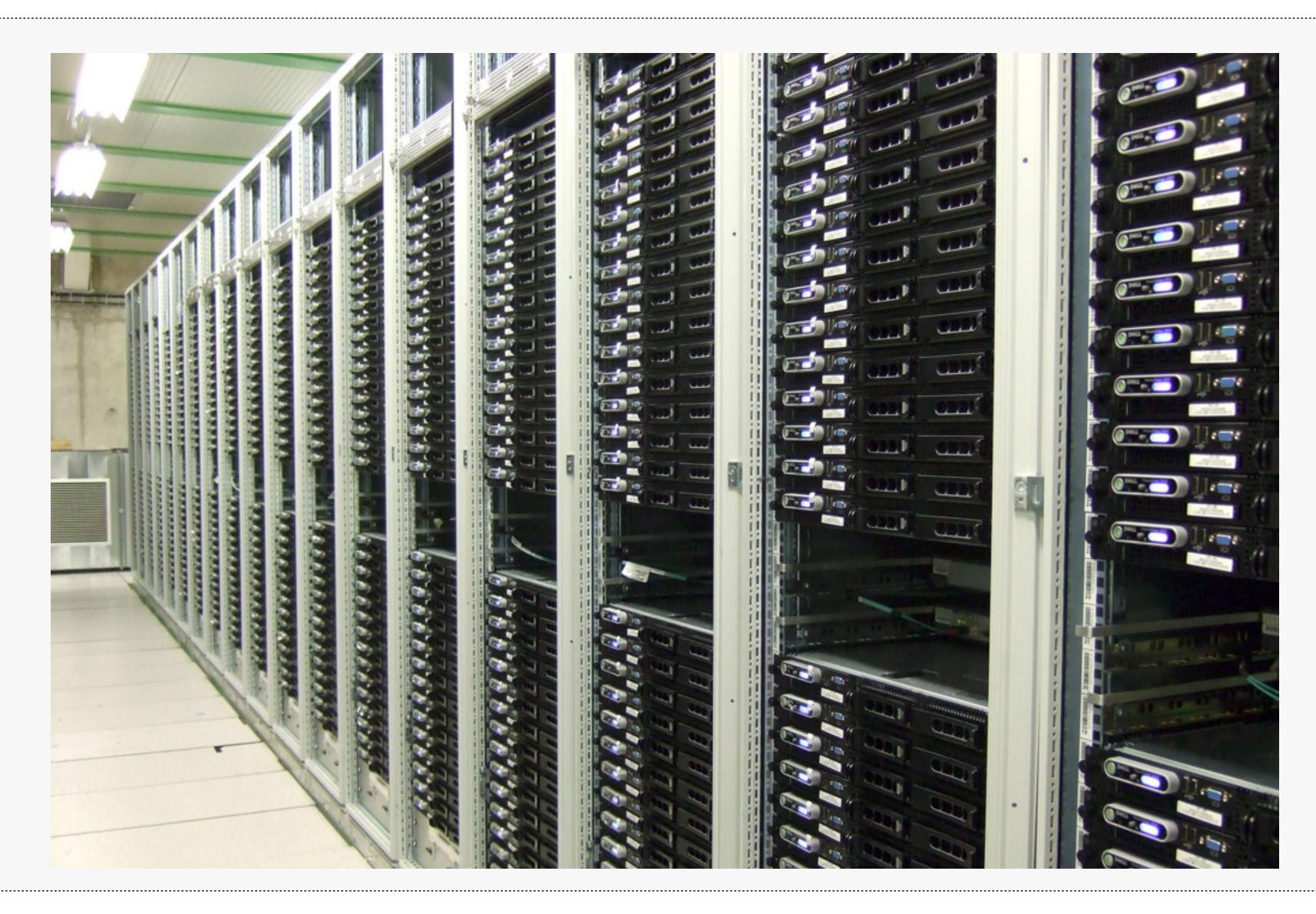
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## Farm (@surface)



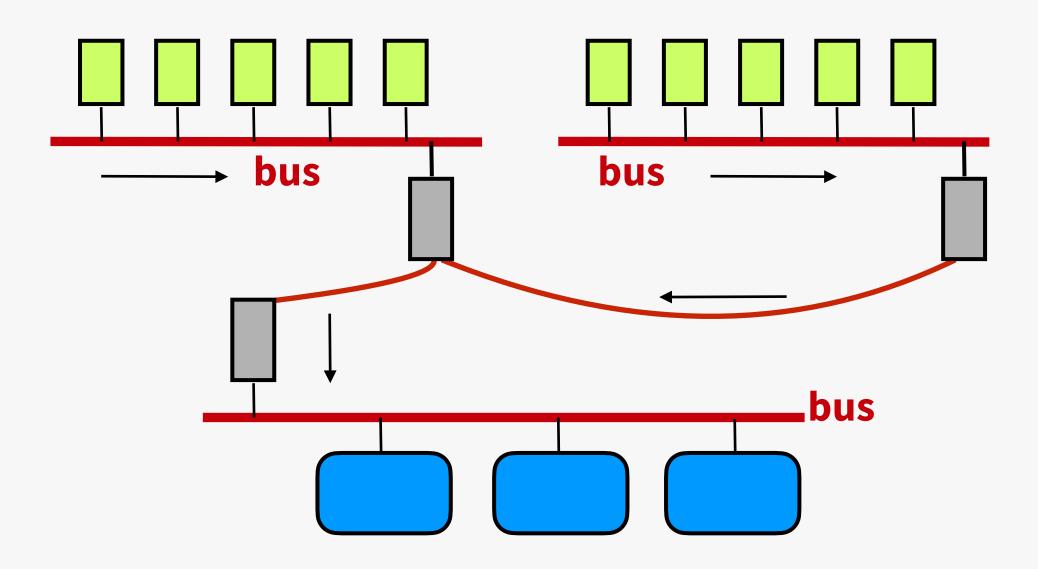
page **70** 





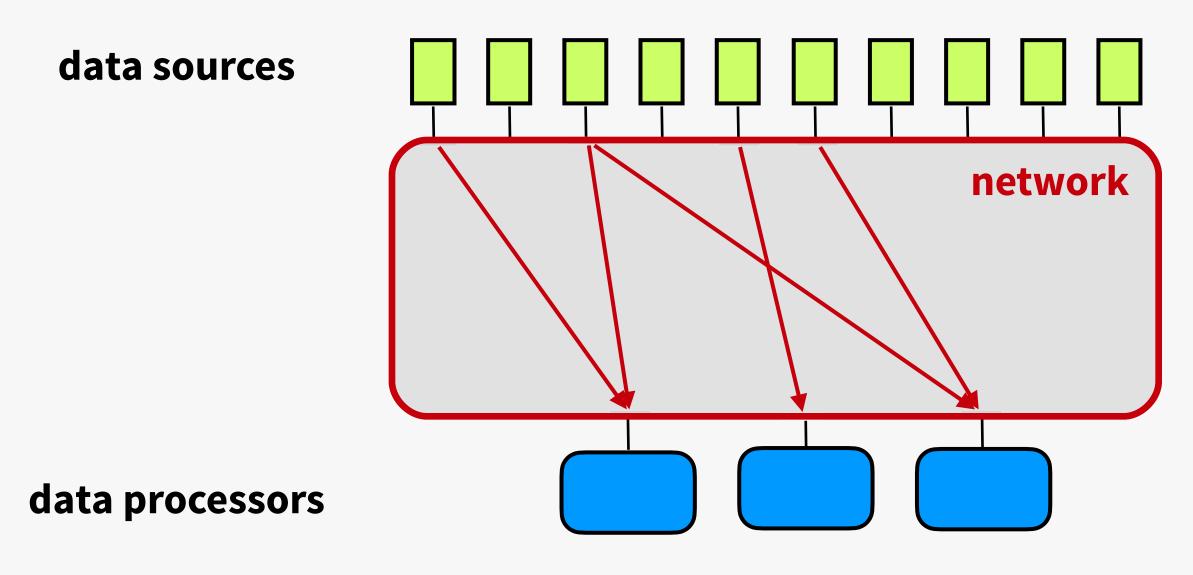
## Readout Topology

- How to connect data sources and data destinations?
- Two main classes: **bus** or **network**



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# How to organize interconnections inside the building blocks and between building blocks?







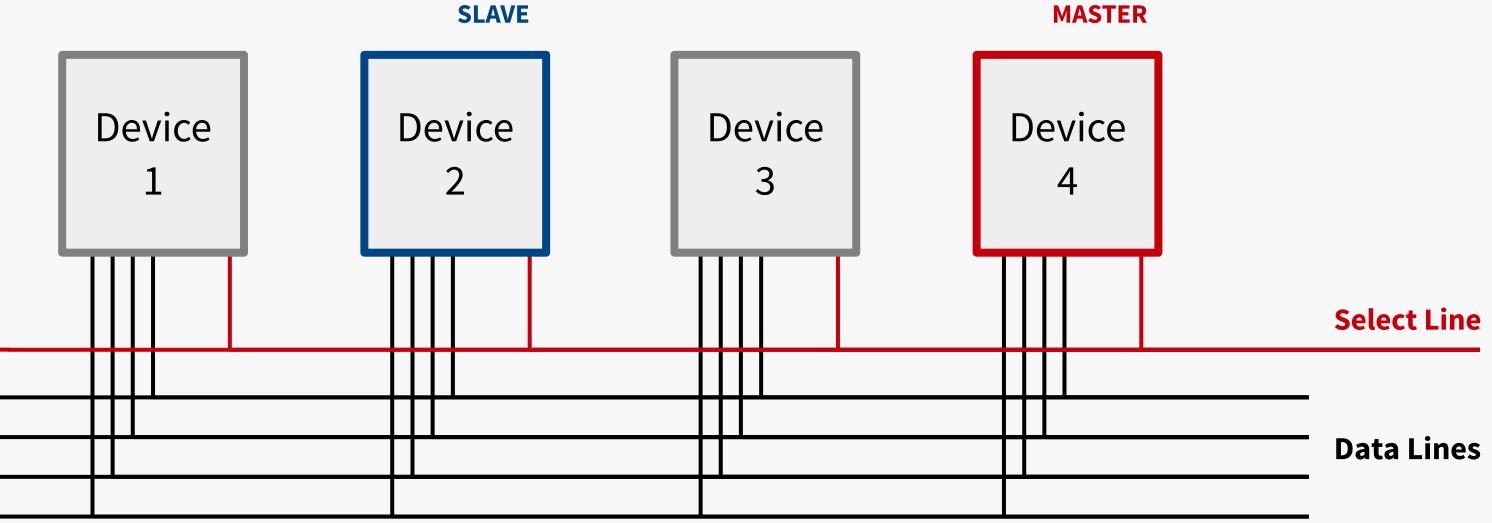
### Buses

#### Devices connected via a **shared bus**

• Bus  $\rightarrow$  group of electrical lines

Sharing implies **arbitration** 

- Devices can be **master** or **slave**
- Devices can be addresses (uniquely identified) on the bus



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### E.g.: SCSI, Parallel ATA, VME, PCI ...

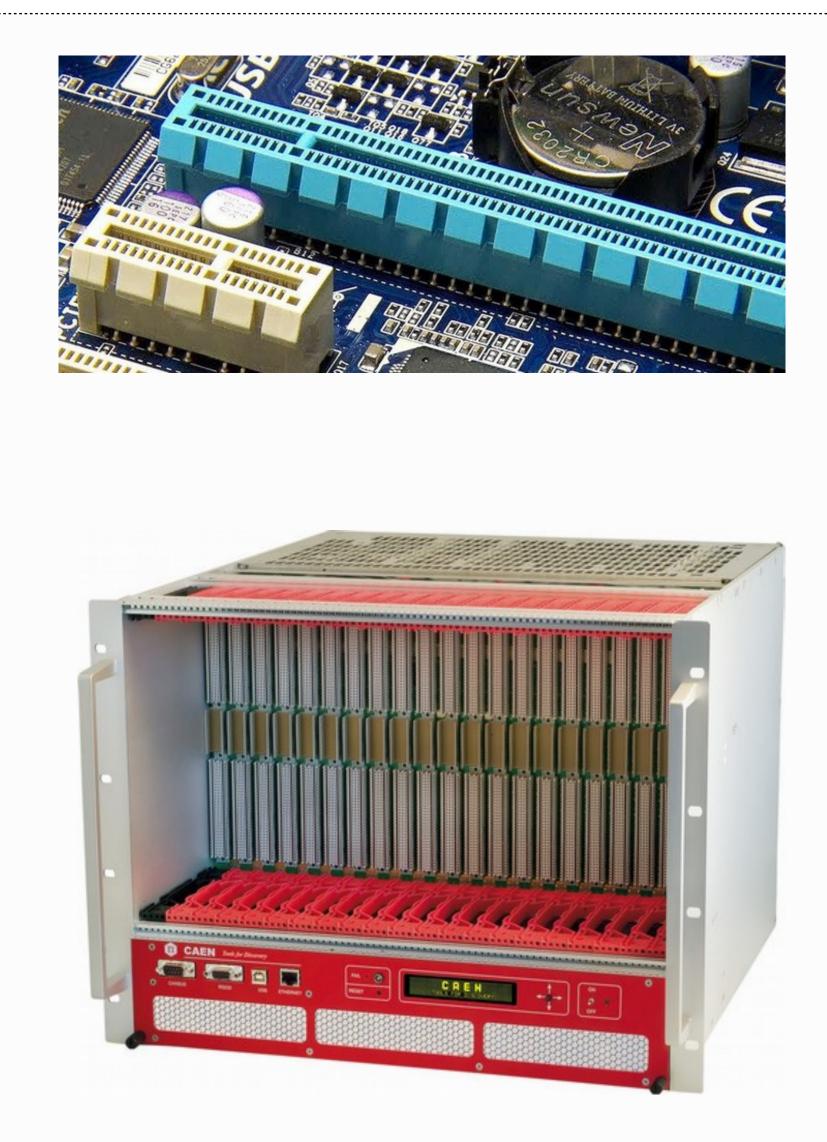
• local, external, crate, long distance, ...

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## Bus examples (some)

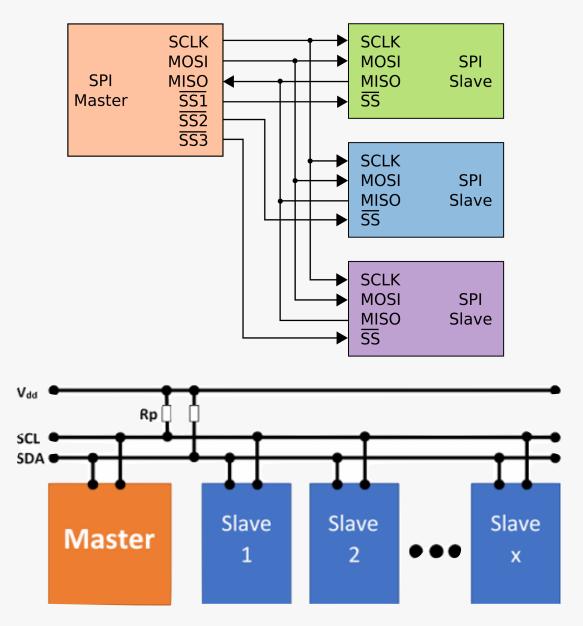


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- I2C
- SPI
  - **7**
- UART
- PCI express

- VME
- μTCA
- ATCA









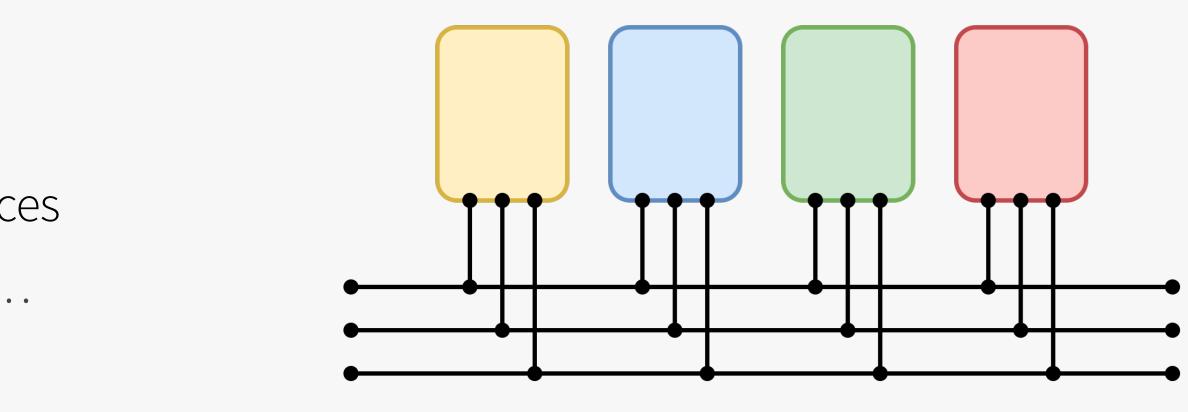
## Bus facts

### Simple :-)

- Fixed number of lines (bus-width)
- Devices have to follow well defined interfaces
  - Mechanical, electrical, communication, ...

### **Scalability issues :-(**

- Bus bandwidth is shared among all the devices
- Maximum bus width is limited
- Maximum number of devices depends on bus length
- Maximum bus frequency is inversely proportional to the bus length
- On the long term, second order "effects" may limit the scalability of your system





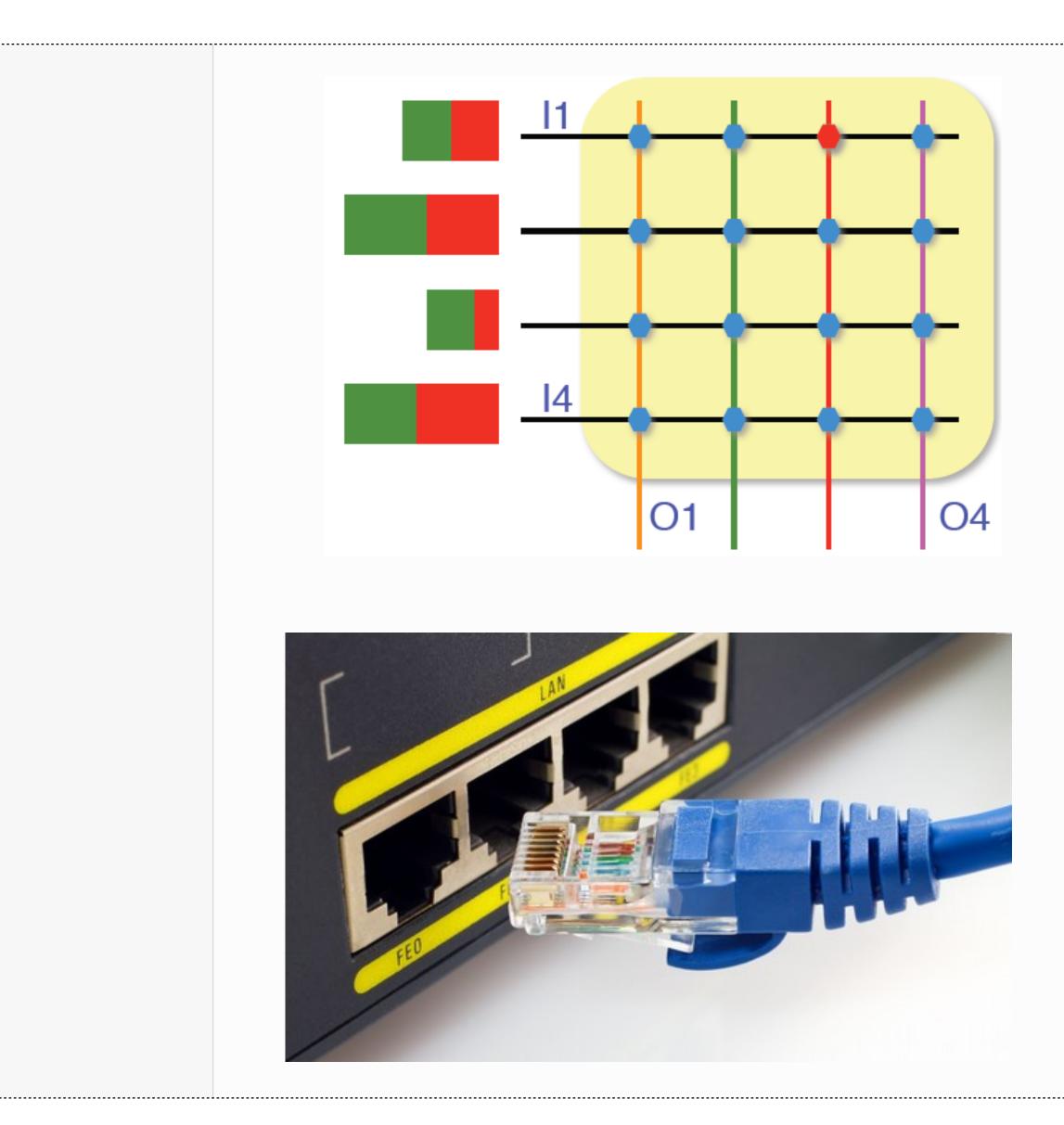




Some second order "effects"

Shanna all



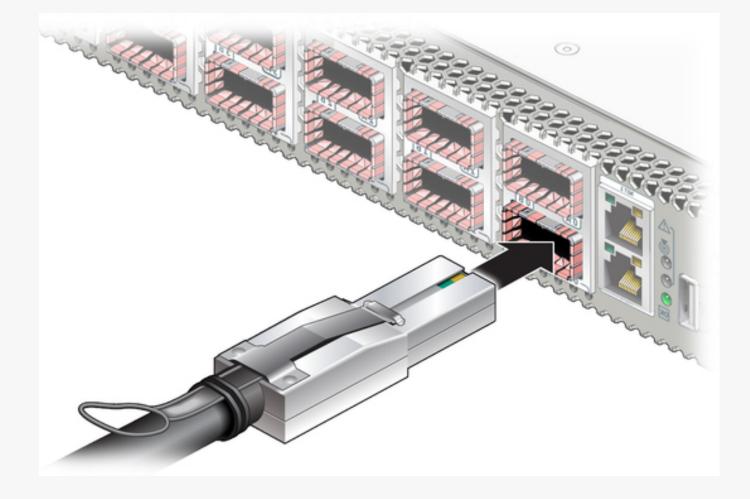




#### All devices are equal (peers)

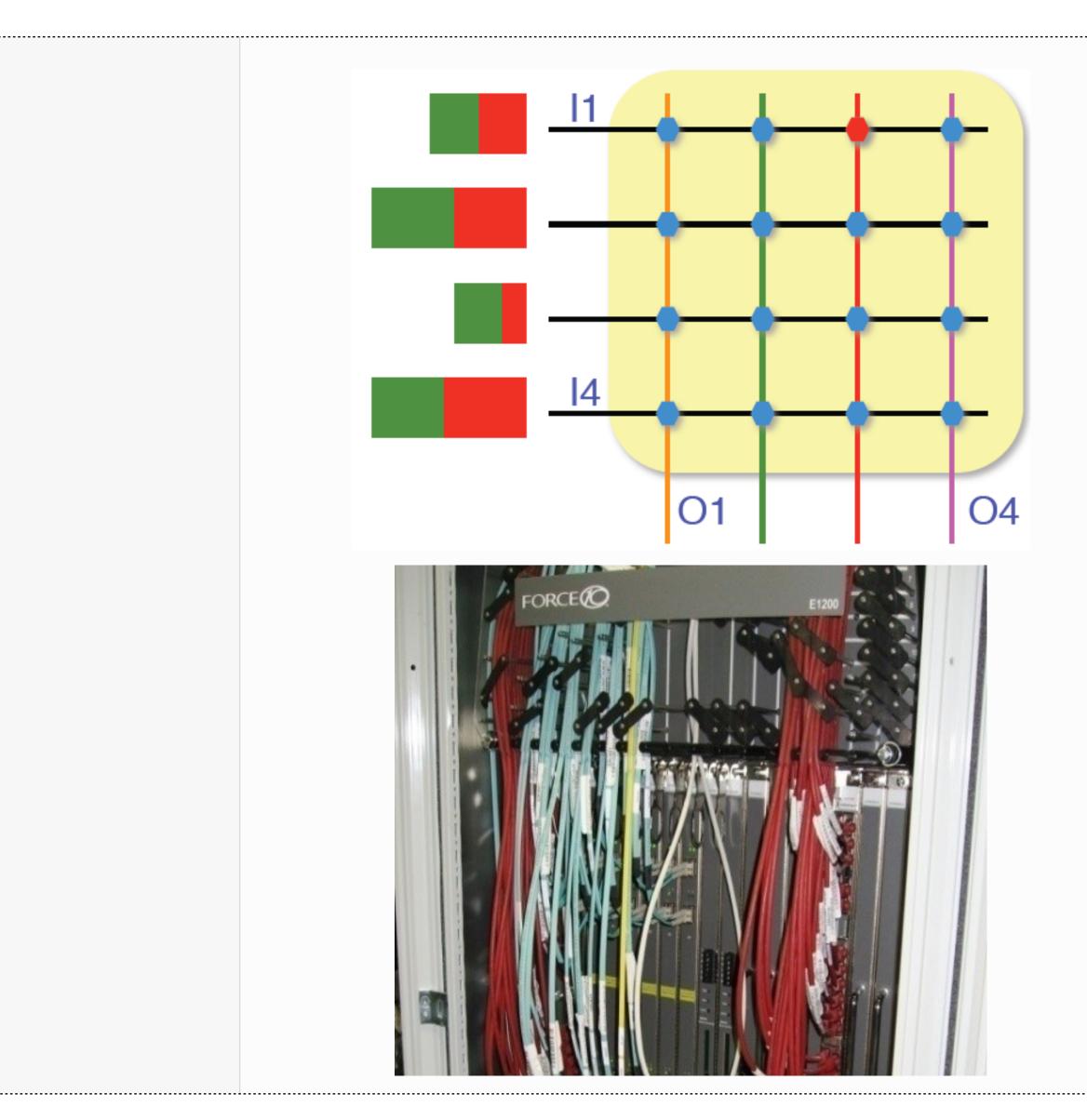
- They **communicate directly** with each other via messages
  - No arbitration
  - Bandwidth guaranteed
- Not just copper: optical, wireless

Eg: Telephone, Ethernet (1G, 10G, 100G, 400G), ...









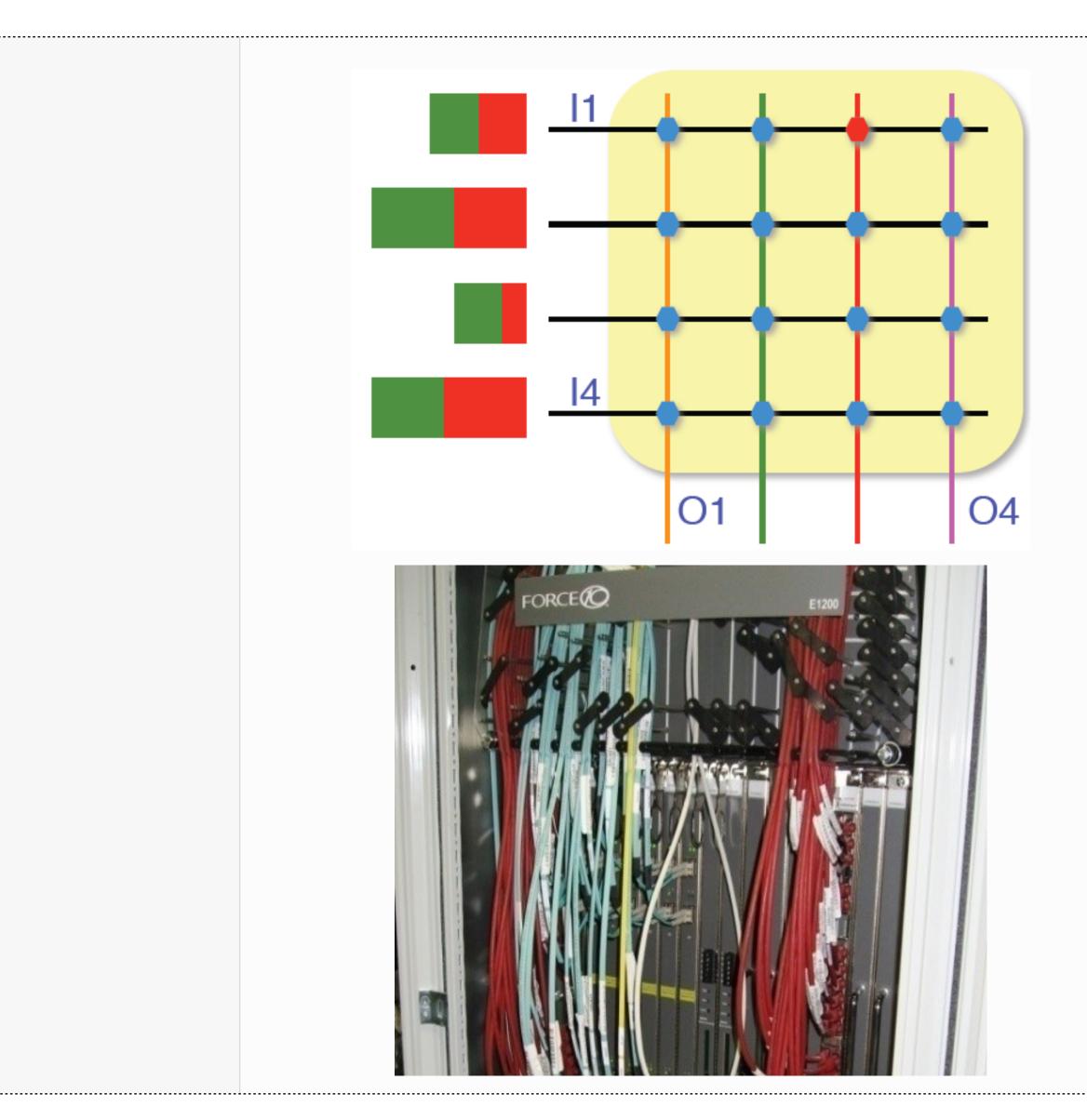
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In switched networks, **switches** move messages between sources and destinations

- Finds the right path between endpoints
   How congestions (two messages with the same destination at the same time) are
   handled?
- The key is ....







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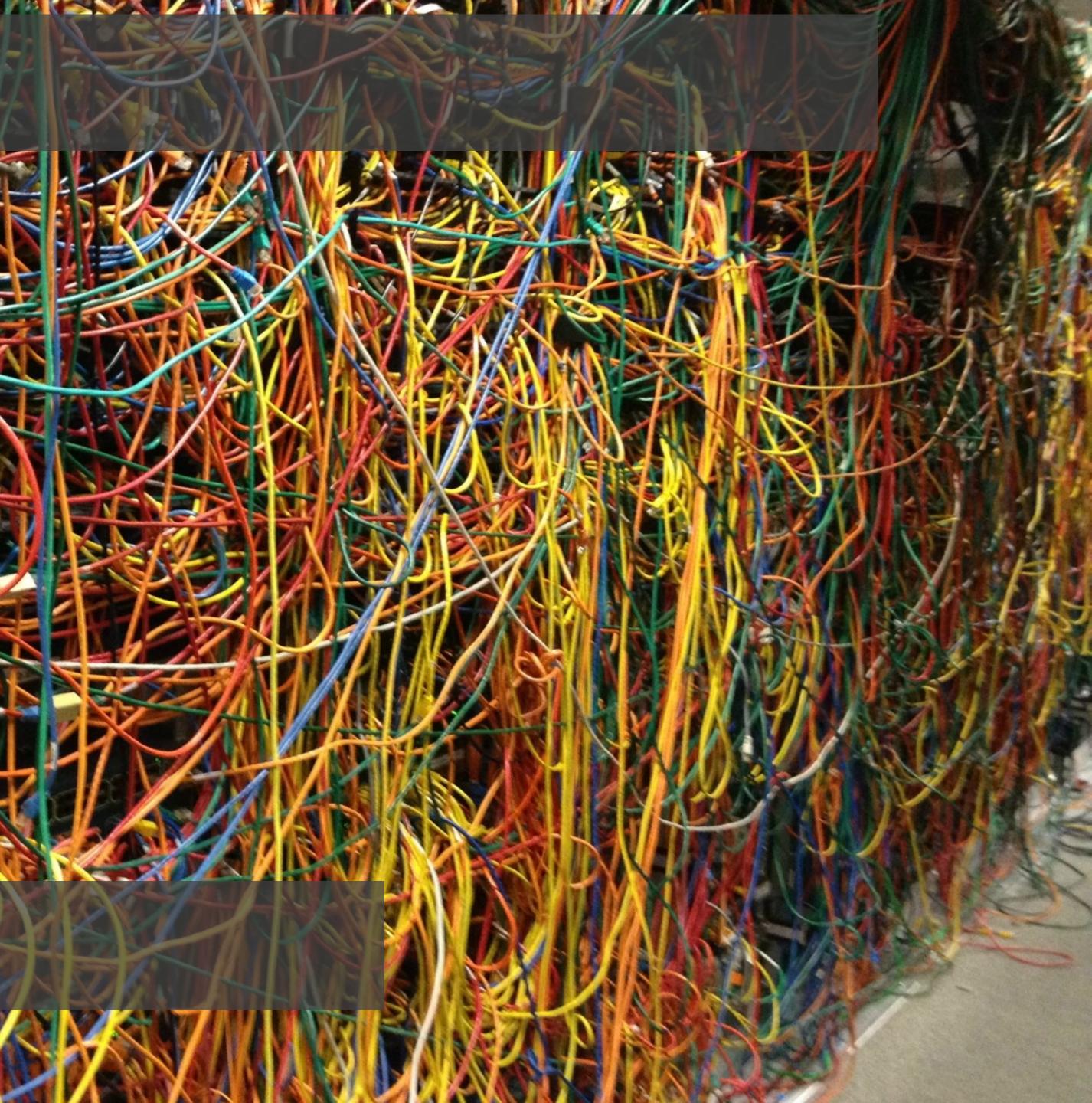
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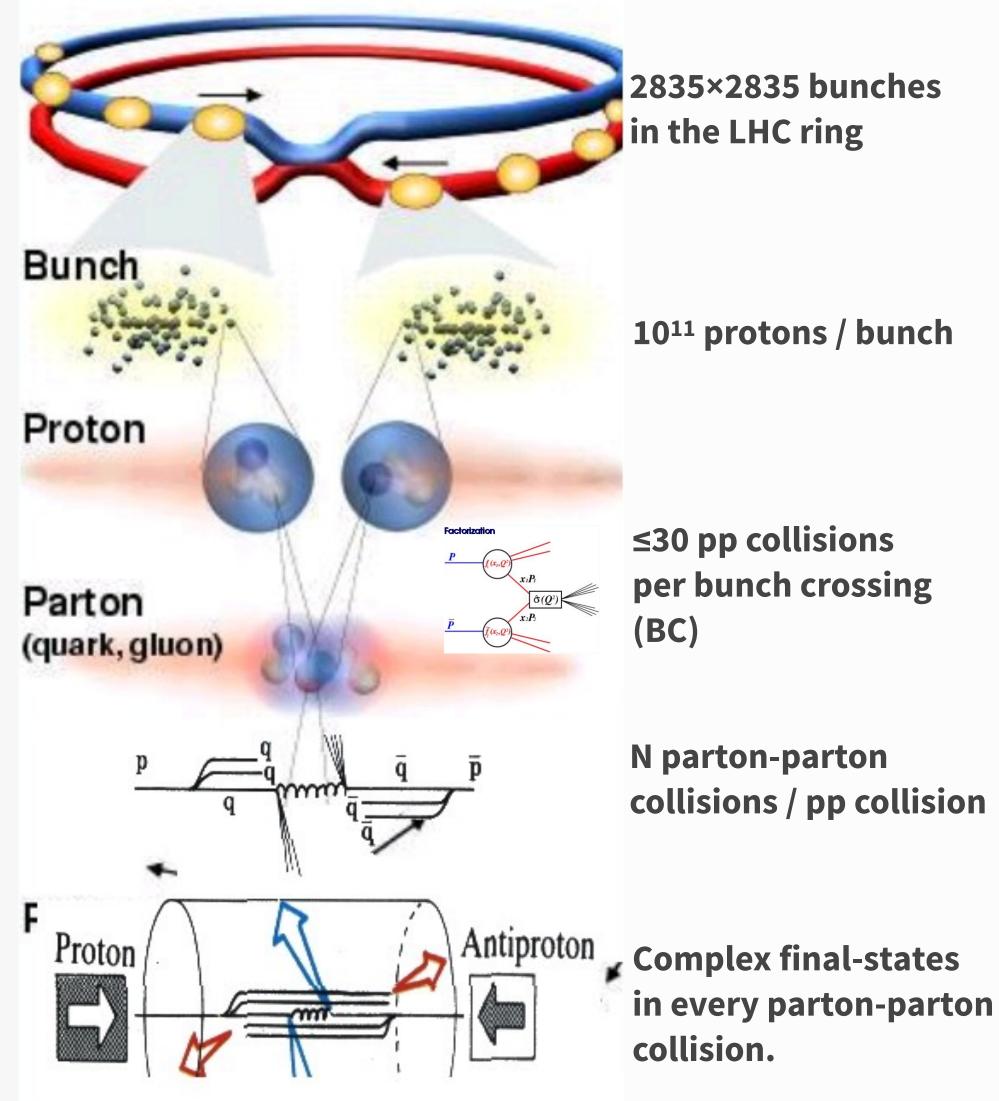




Cable management is still a thing.



## LHC and its products



Design parameters

 $E_{cms} = 14 \text{ TeV}$  $L = 10^{34} / \text{cm}^2 \text{ s}$ BC clock = 40 MHz

 $R = \sigma_{in} \times L$ 

Interesting processes **extremely rare**, high Luminosity is essential

#### • Close collisions in space and time

- Large proton bunches (1.5x10<sup>11</sup>)
- Fixed frequency: 40MHz (1/25ns)

Protons are **composite particles** 

abundant low energy interactions

Few rare high-E events overwhelmed in abundant low-E environment

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## LHC Detectors Challenges

Huge

- O(10<sup>6</sup>-10<sup>8</sup>) channels
- ~1 MB event size for pp collisions
  - ► 50 MB for pb-pb collisions (Alice)
- Need huge number of connections

#### Fast and slow detectors

- Some detectors readout requires >25 ns and integrate more than one bunch crossing's worth of information
  - e.g. ATLAS LArg readout takes ~400 ns

#### Online, what is lost is lost forever

• Need to monitor selection - need very good control over all conditions



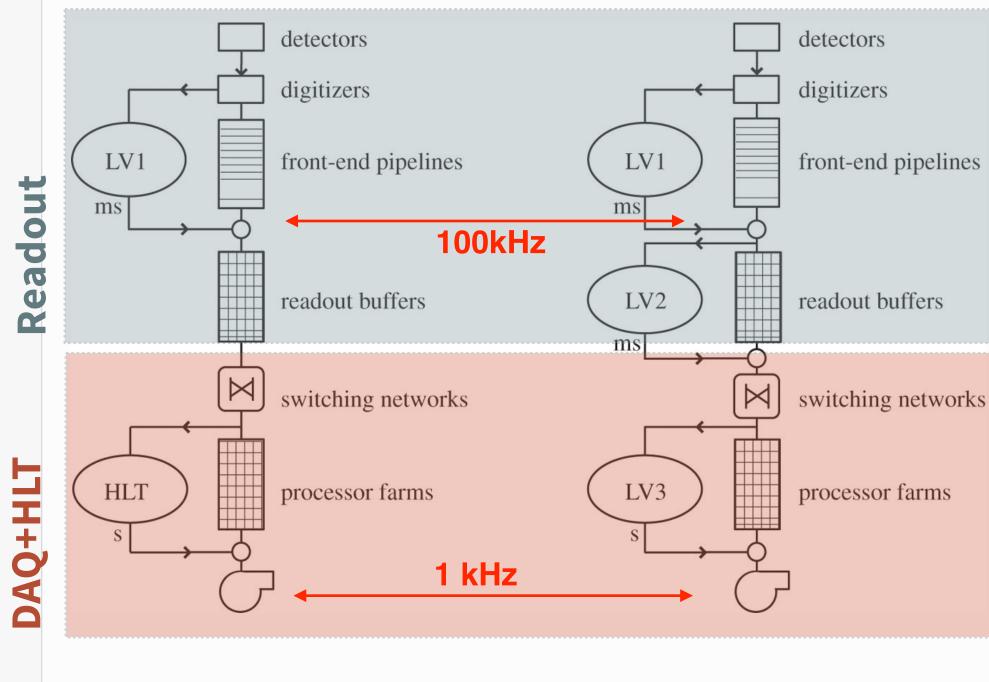
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## HLT/DAQ requirements

- Robustness and redundancy
- Scalability to adapt to Luminosity, detector evolving conditions
- Flexibility (> 10-years lifetime)
- Based on commercial products
- Limited cost



### ATLAS/CMS Example (Run 1)

- 1 MB/event at 100 kHz for O(100ms) HLT latency
  - Network: 1 MB\*100 kHz = 100 GB/s
  - HLT farm:  $100 \text{ kHz}^{*}100 \text{ ms} = O(10^{4}) \text{ CPU cores}$
- Intermediate steps (level-2) to reduce resources, at cost of complexity (at ms scale)

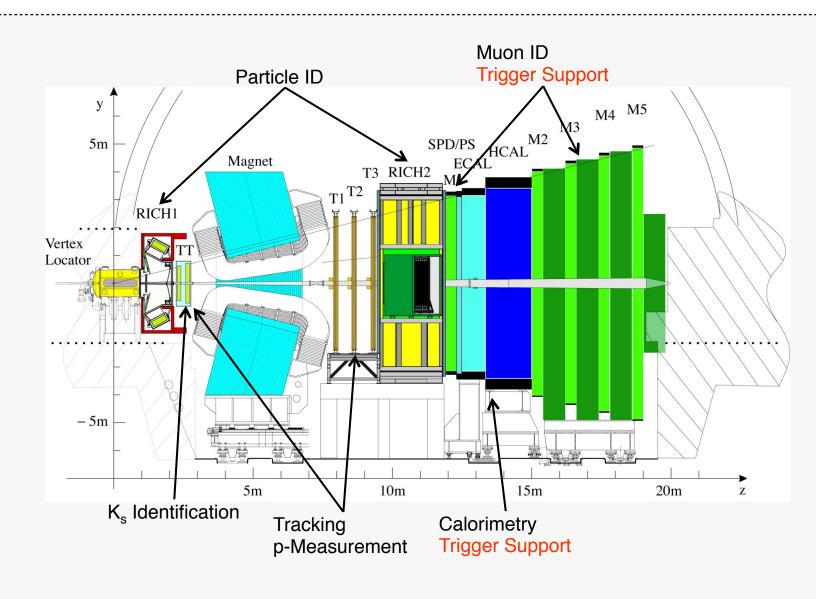
Prefer COTS hardware: PCs (linux based), Ethernet protocols, standard LAN, configurable devices

See S.Cittolin, DOI: 10.1098/rsta.2011.0464





## LHCb TDAQ Architecture



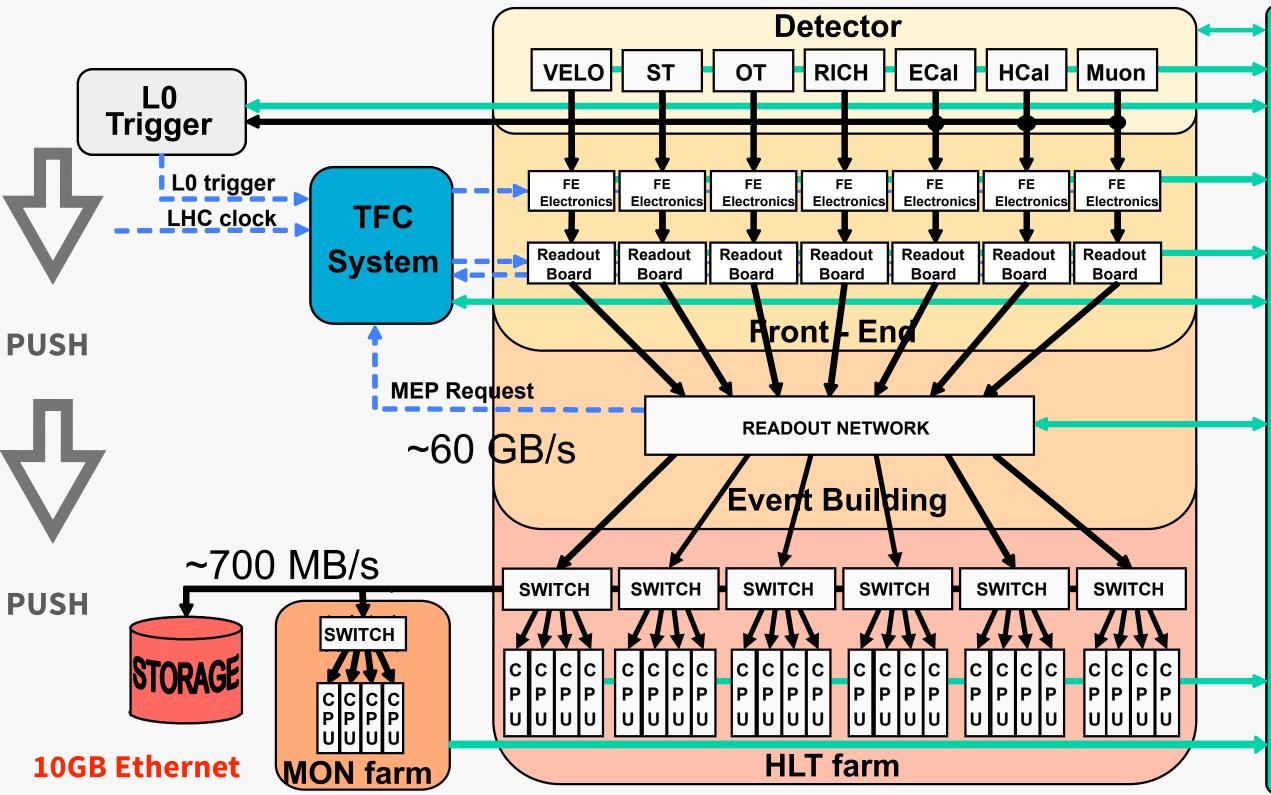
Single forward arm spectrometer  $\rightarrow$ reduced event size

- Average event size 60 kB
- Average rate into farm 1 MHz
- Average rate to tape ~12 kHz

Small event, at high rate

• optimised transmission





- **Event data**
- Timing and Fast Control Signals
- **Control and Monitoring data**









## ALICE

### 19 different detectors

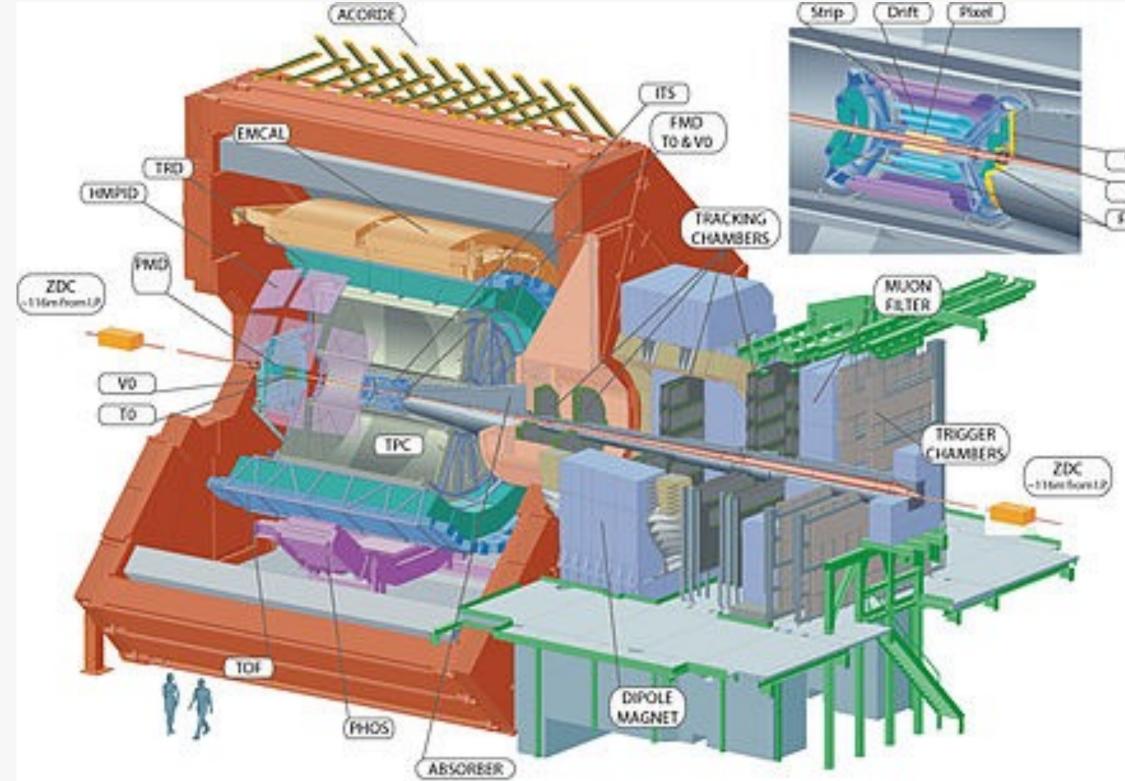
- with high-granularity ant timing information
- Time Projection Chamber (TPC): very high occupancy, slow response
   Large event size (> 40MB)
- TPC producing 90% of data

Challenges for the TDAQ design:

- detector readout: up to ~50 GB/s
- low readout rate: max 8 kHz
- storage: 1.2 TB/s (Pb-Pb)





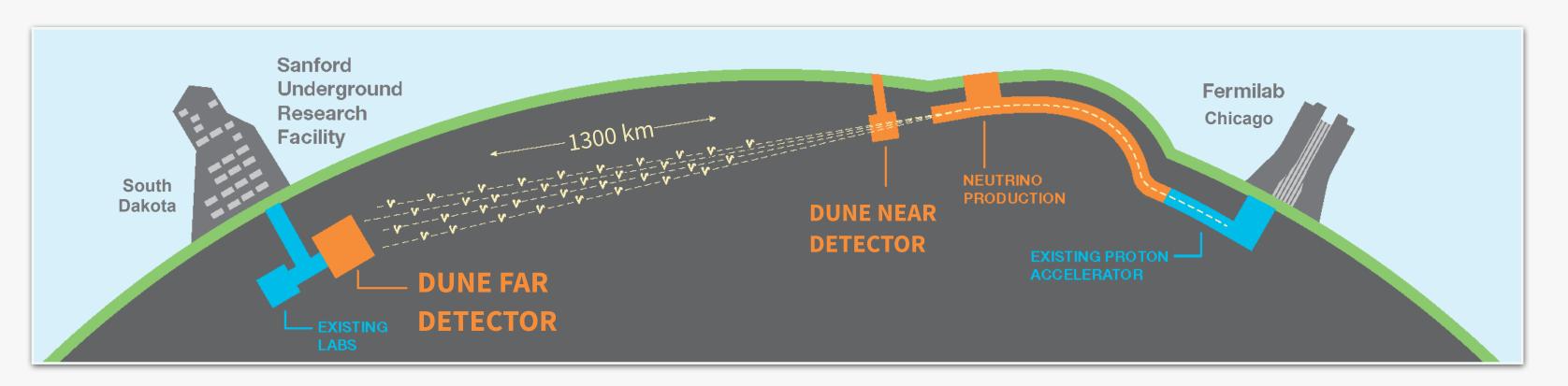


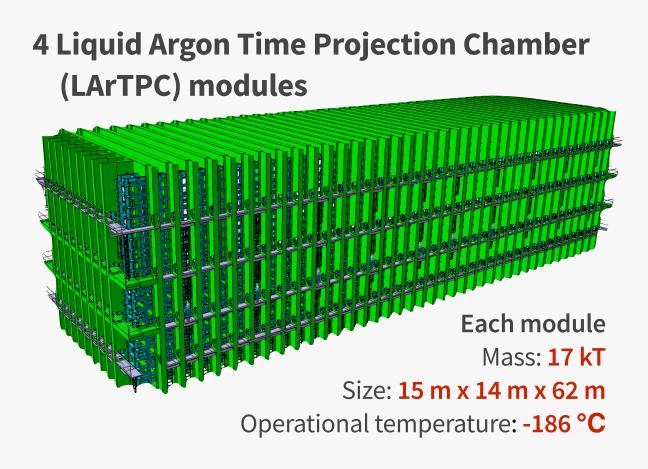




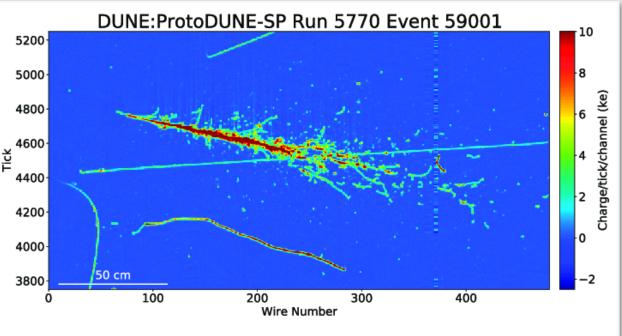


## The Deep Underground Neutrino Experiment - DUNE





#### High-resolution imaging detector



Leading edge, world class neutrino experiment with high profile physics programme

• Nature of neutrinos, supernova collapse, proton decay searches

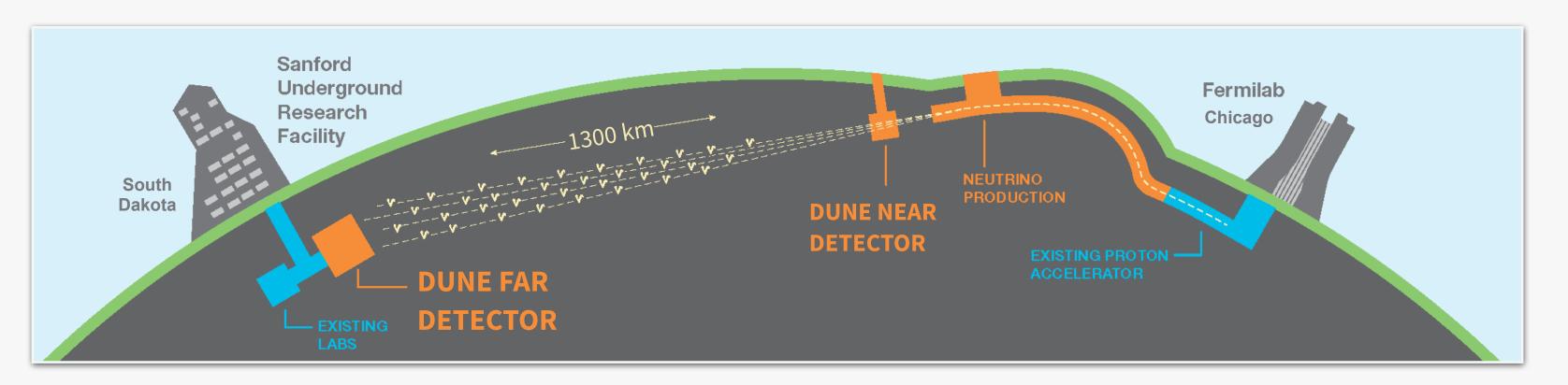
### **Gigantic Far Detector**

Huge target mass AND high resolution imaging

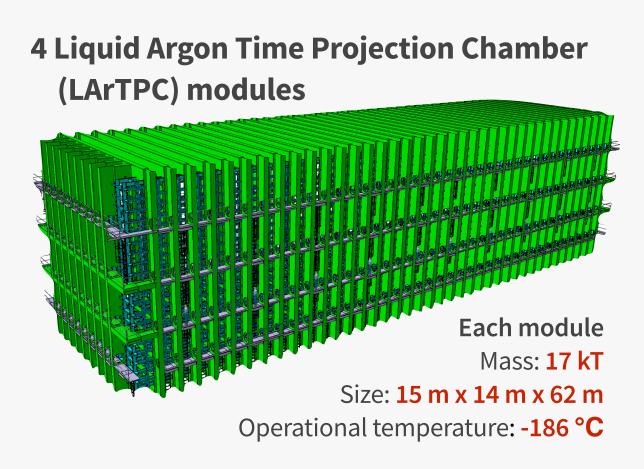


spatial resolution: 5mm

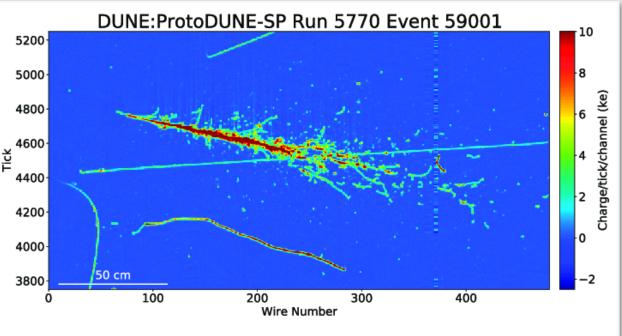
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#### **TDAQ: no quick access and no large host lab in the vicinity!**



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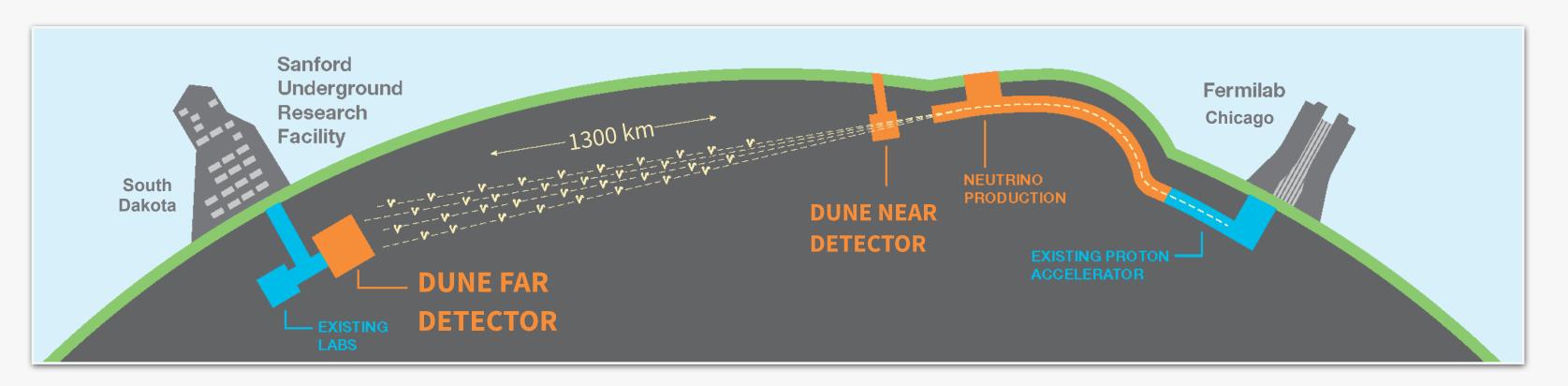
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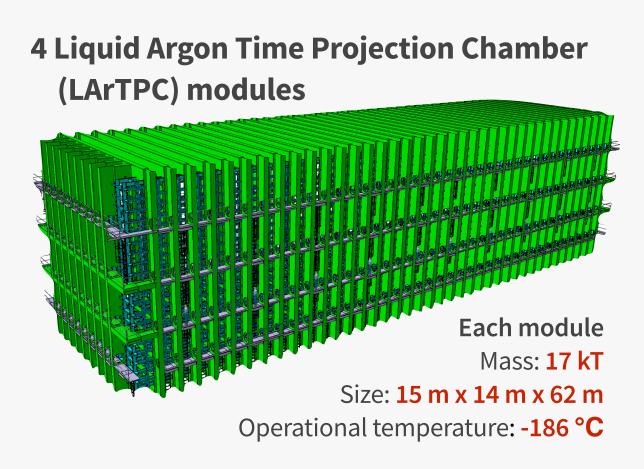


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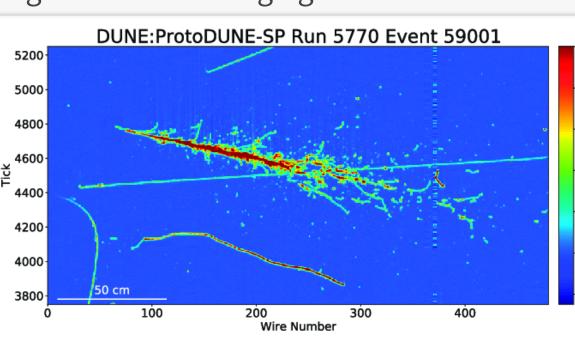
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### **Gigantic Far Detector**

Huge target mass AND high resolution imaging

**TDAQ: 4 independent instances, synchronized to a** common clock, supporting different detector technologies

spatial resolution: 5mm



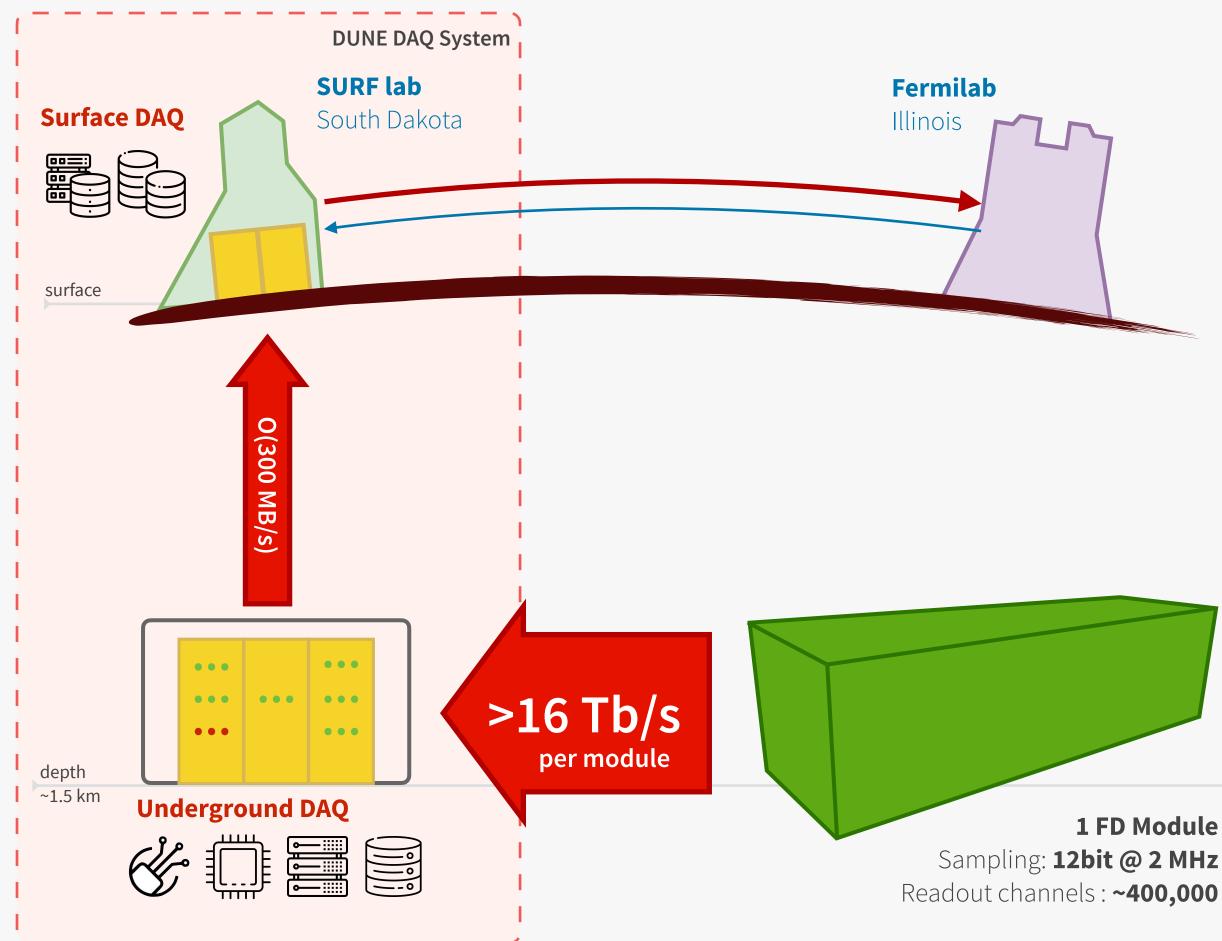
## The DUNE Data AcQuisition System - DAQ

#### **DUNE DAQ System**

- Collects large amount of streaming data from detectors
- Selects only interesting interactions
- ▶ Buffers the **full data stream** for ~100s for supernova physics
- Deliver selected interactions to permanent storage

Unique challenges

- ► High data rate, high uptime
- Remote experimental site
- Deep underground in an active mine



## Summary

This lecture is just an introduction about data acquisition

- DAQ (& Trigger) is a complex and fascinating topic, combining very different expertise
- More details on **Trigger** and **FPGAs** in the following lectures

Covered the principles of a simple data acquisition system

- Basic elements: trigger, derandomiser, FIFO, busy logic
- Scaling to multi-channel, multi-layer systems
- How data is transported
  - Bus versus network

### A (very) brief overview of LHC experiments + DUNE DAQ systems

• Similar architectures, different optimisations driven by detector requirements



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