

# Highlights from CMS experiment and prospects for Run3

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# This talk in a nutshell

- ◉ *With Run3 starting, we are still in the process of analysing Run2 data*
  - ◉ *There are interesting (but still small) discrepancies to watch for)*
  - ◉ *More could come*
- ◉ *Pursuing the search process is a must*
- ◉ *With x2 more data coming at ~ same energy, maximal gain from exploring (also) new directions*
- ◉ *Alternative data taking could be crucial, to extend our search to the data that we normally throw away*
  - ◉ *Scouting*
  - ◉ *Parking*
  - ◉ *Anomaly detection*
  - ◉ *...*

# This talk in a nutshell

◉ *With Run3 starting, we are still in the process of analysing Run2 data*

◉ *There are interesting (but still small) discrepancies to watch for)*

◉ *More could come*

◉ *Pursuing the search for new physics*

◉ *With new*

**I am focusing on searches today, for lack of time  
A lot more happening with Higgs, top, EW, QCD and  
Heavy ions**

◉ *Alternative channels  
that we normally throw away*

◉ *Scouting*

◉ *Parking*

◉ *Anomaly detection*

◉ *...*



# A little bit of history

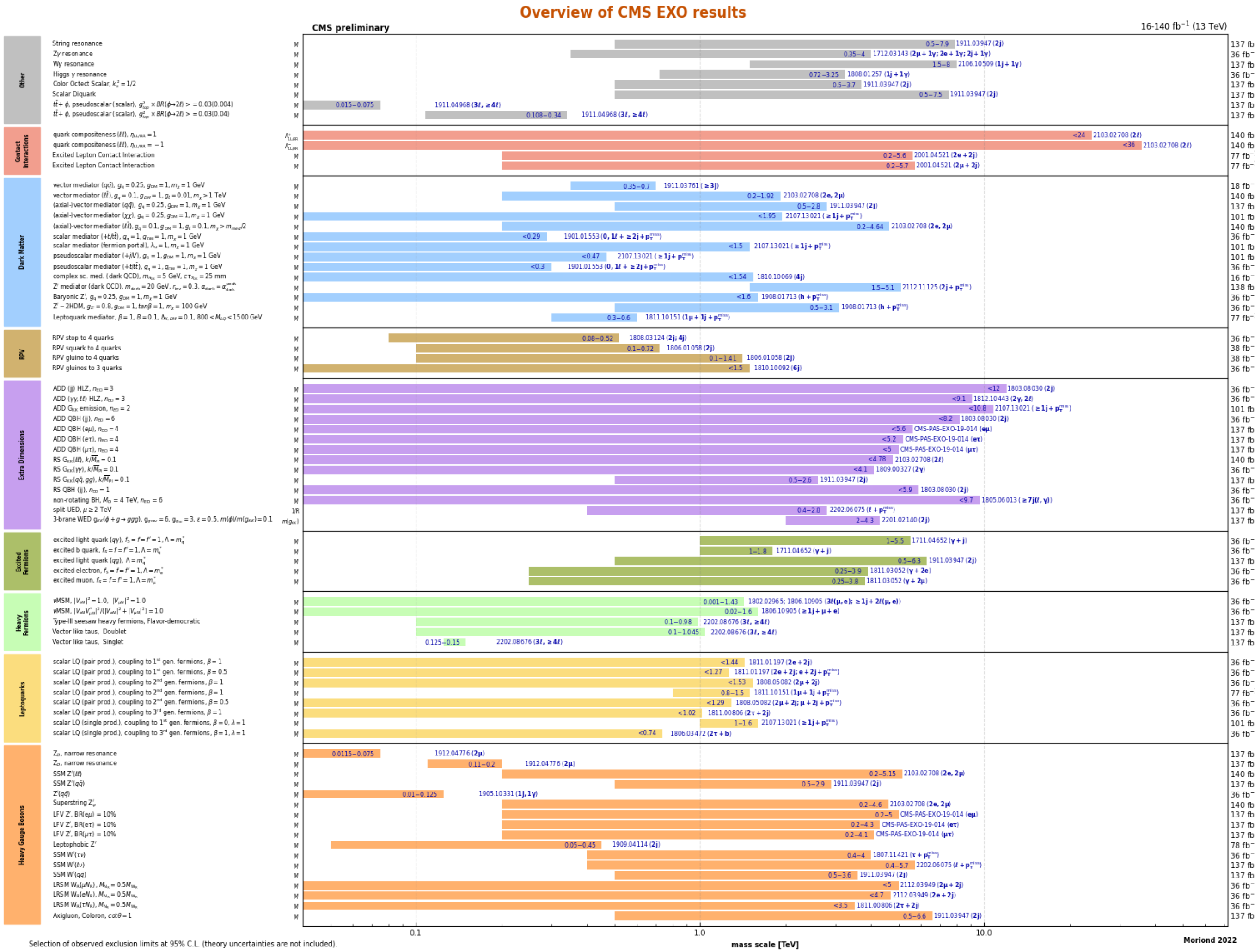
● Run 1 (7 TeV/8TeV) was a triumphant journey towards the Higgs discovery

● In Run 2 we shifted attention to new physics, thanks to energy increase (to 13 TeV)

● No discovery reported, but

● several analyses reported 3-4 sigma (local) excesses that should be monitored with more data

● many analyses are still ongoing



Selection of observed exclusion limits at 95% C.L. (theory uncertainties are not included).





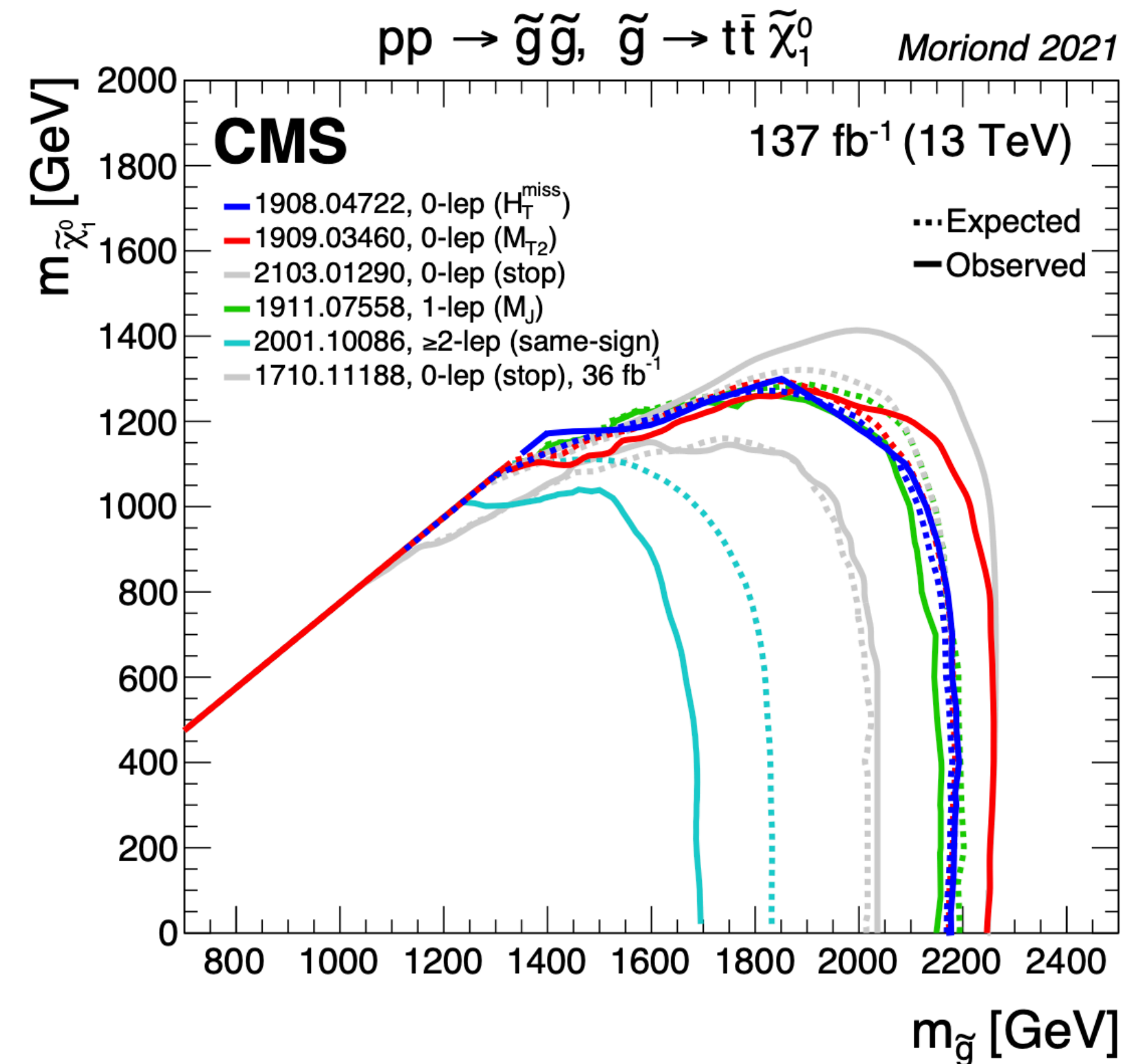
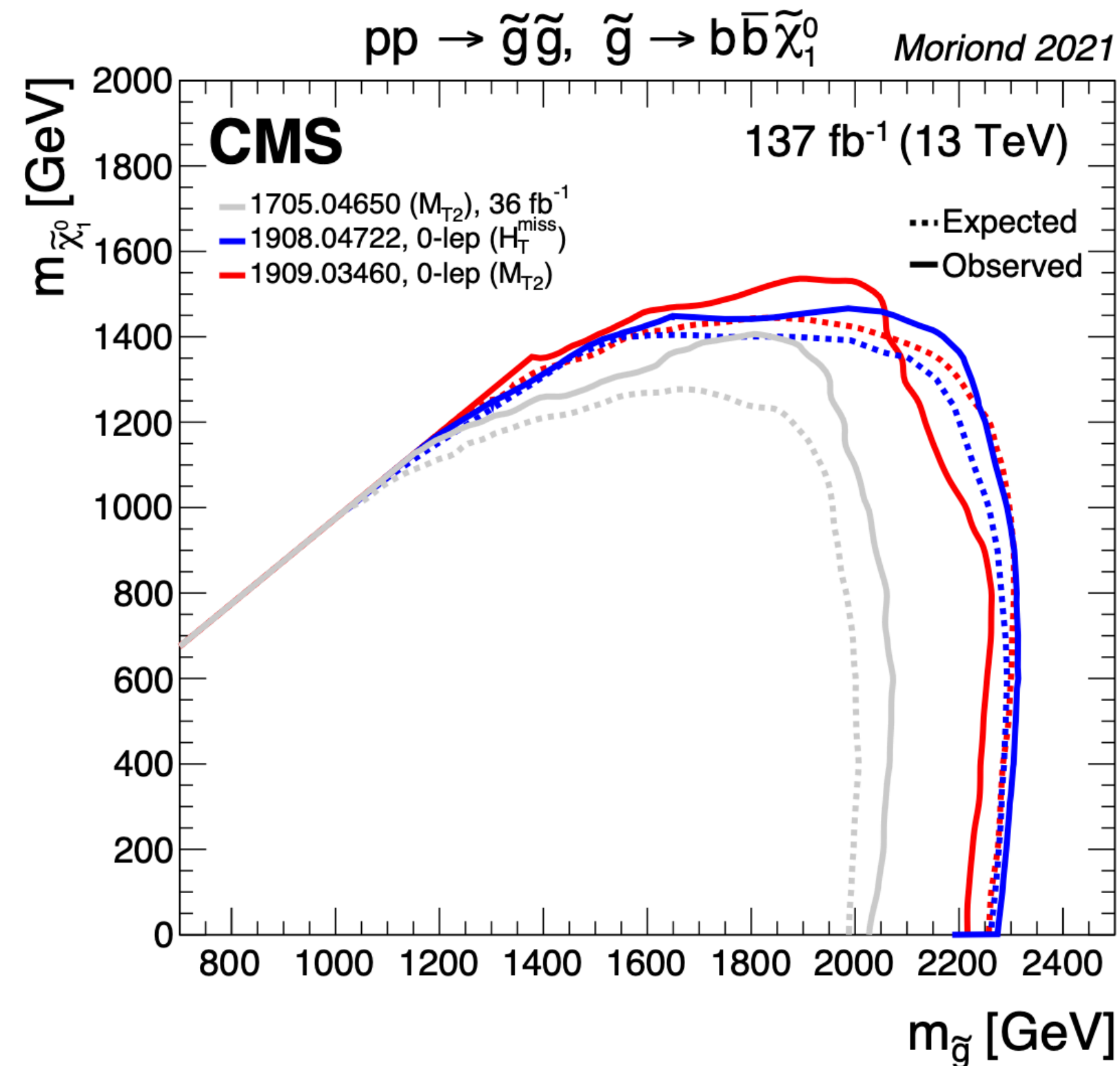
# Let's keep in mind...

⦿ Exclusion plots come with assumptions and warnings

⦿ For instance, SUSY limits use simplified models

⦿ Not a complete model, BR=100% typically assumed, Cross section computed assuming all other sparticles are decoupled

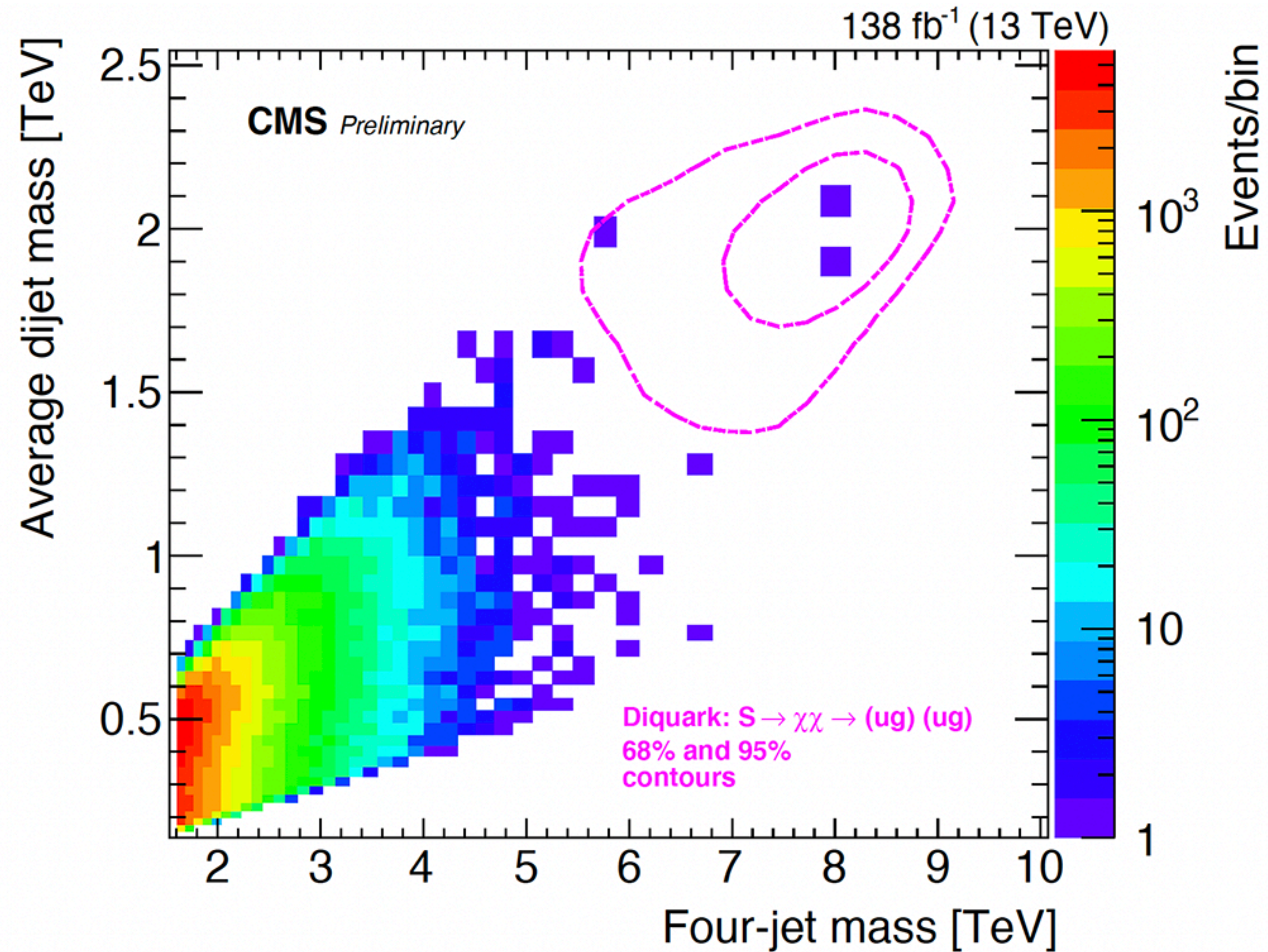
⦿ The actual exclusion might be weaker





# Some interesting excess

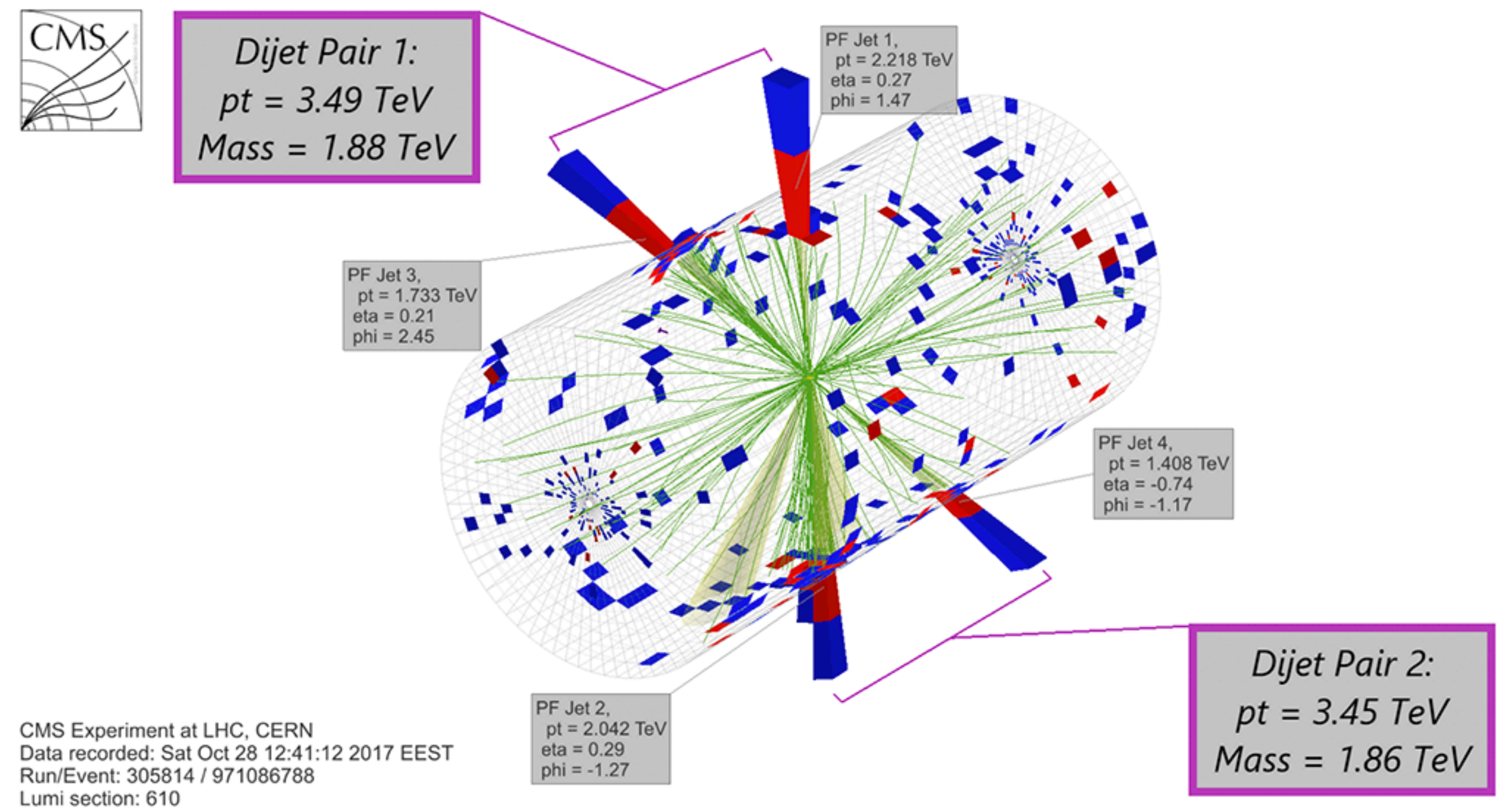
- ⦿ *The dijet resonance search identified an excess of a few events on the tail*
- ⦿ *When inspected, these events revealed a common structure*
  - ⦿ *two jet pairs, merged into single jets with  $M \sim 2$  TeV*
  - ⦿ *The dijet system had mass above 6 TeV*
- ⦿ *Nothing yet significant, but something to watch*





# Some interesting excess

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- *Nothing yet significant, but something to watch*



# Some interesting excess

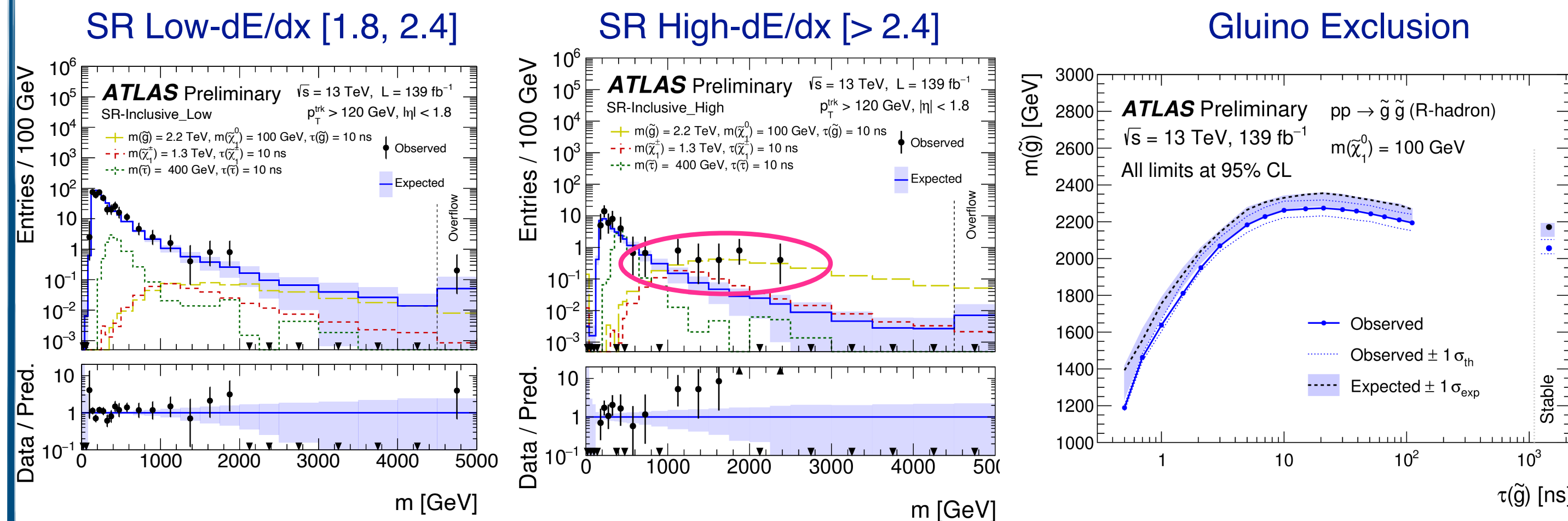
- Excess seen in ATLAS  $dE/dx$  analysis. CMS still analysing Run2 data + will repeat the search in Run3

Talk by Julia Gonski @Moriond

[SUSY-2018-42](#)

## Pixel $dE/dx$ Result

- Excess (3.6 $\sigma$  local, 3.3 $\sigma$  global)** in high  $dE/dx$  SR ( $> 2.4$ ) with for target mass hypothesis of 1.4 TeV
  - ! Cross check of candidate tracks with TileCal and MS time-of-flight variables was consistent with  $\beta = 1$ , therefore **not consistent with LLP hypothesis**
- Set limits on gluino, chargino, stau hypotheses
  - Max sensitivity for  $\tau \sim 10$ -30 ns
  - Gluino R-hadrons with mass  $< 2.27$  TeV excluded with  $\tau = 20$  ns and LSP mass = 100 GeV





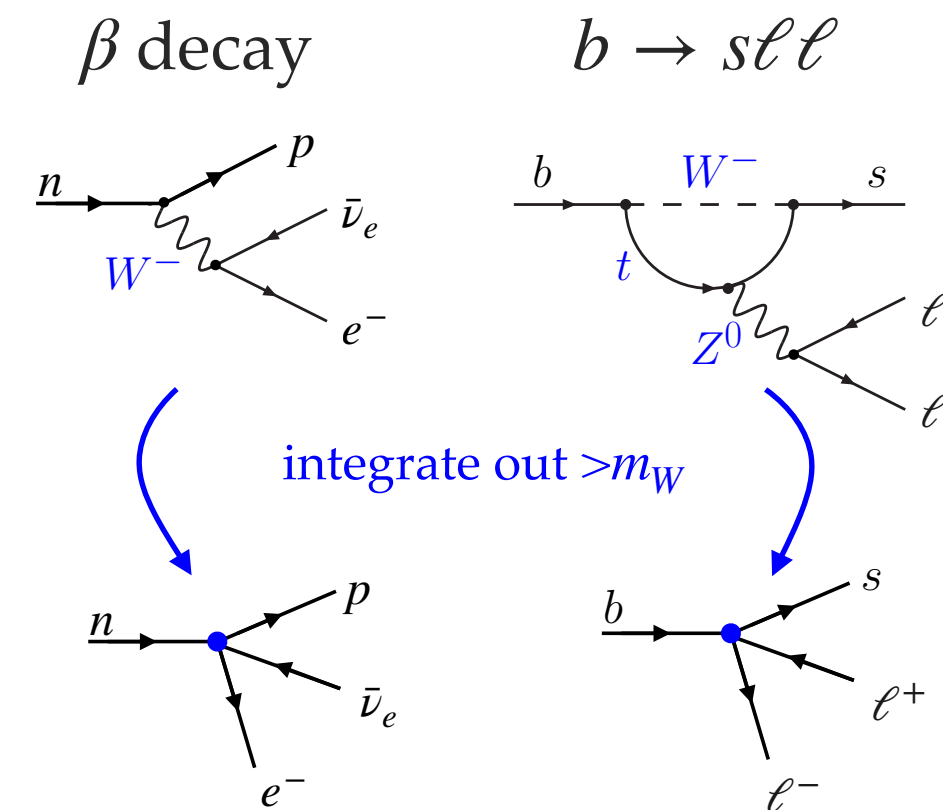
# And the biggest excitement

● LHCb anomalies point to breaking of Lepton Flavor Universality

● In  $\mu/e$  ratio of  $b \rightarrow s l l$

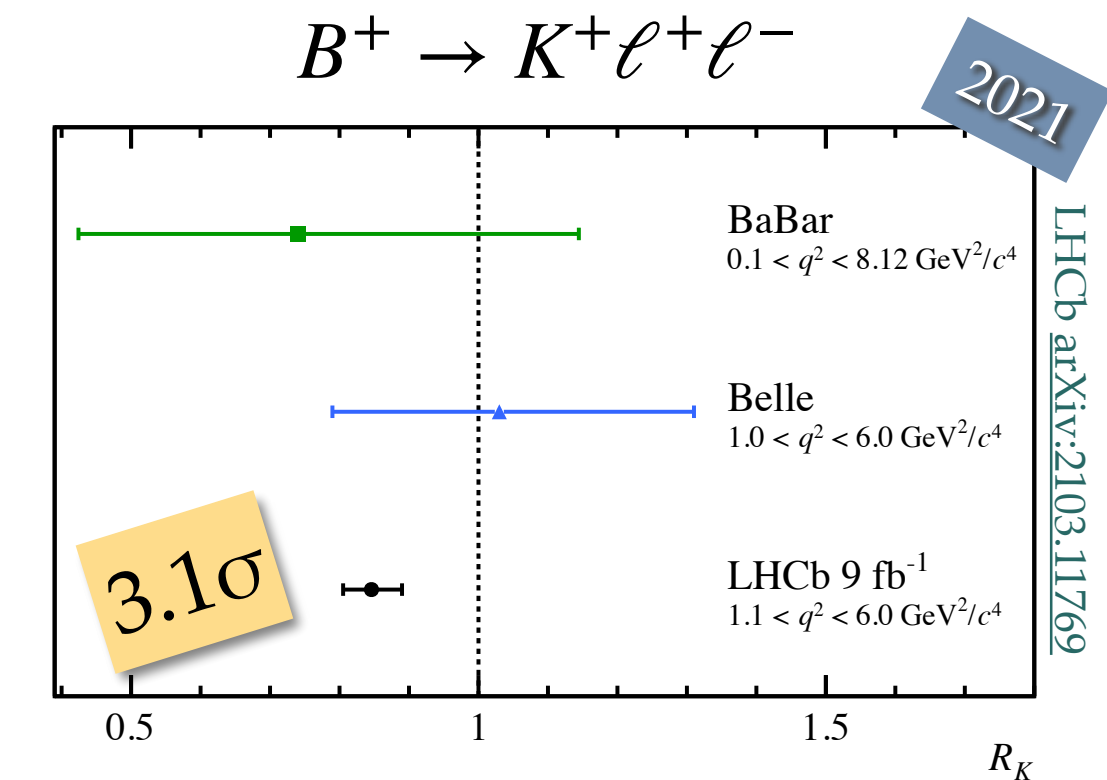
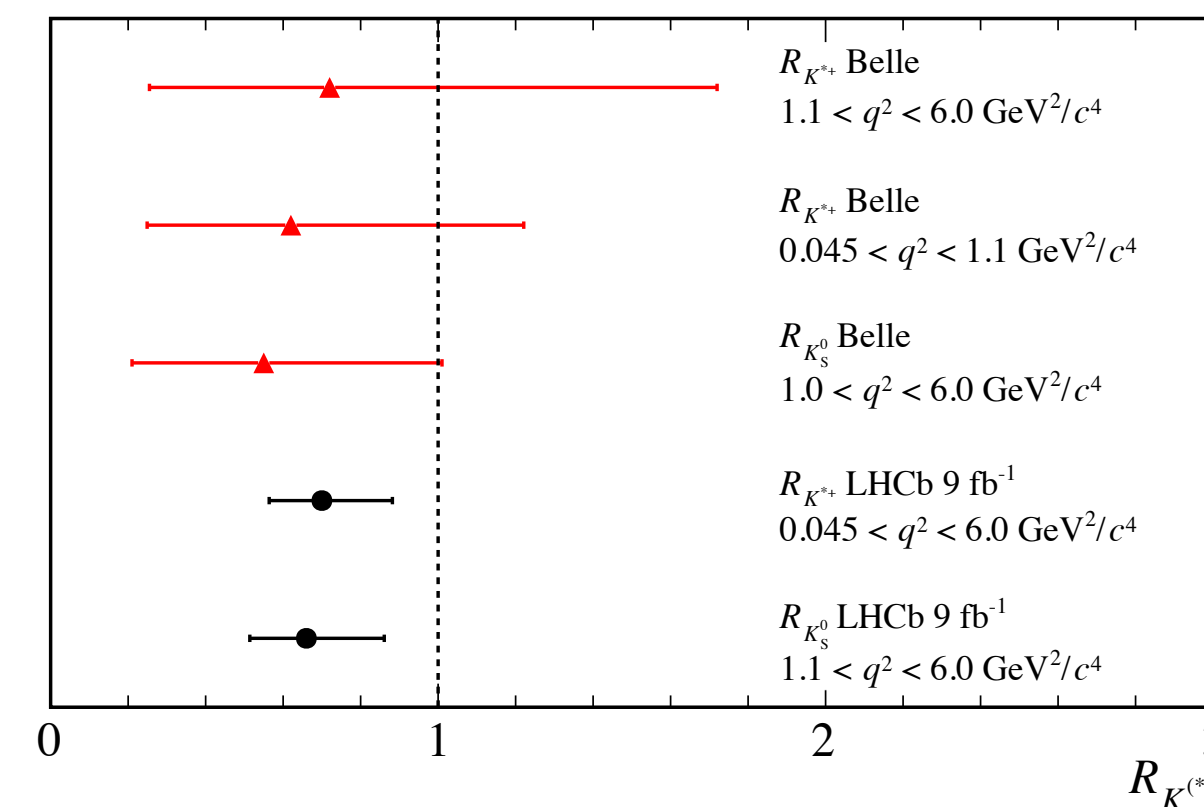
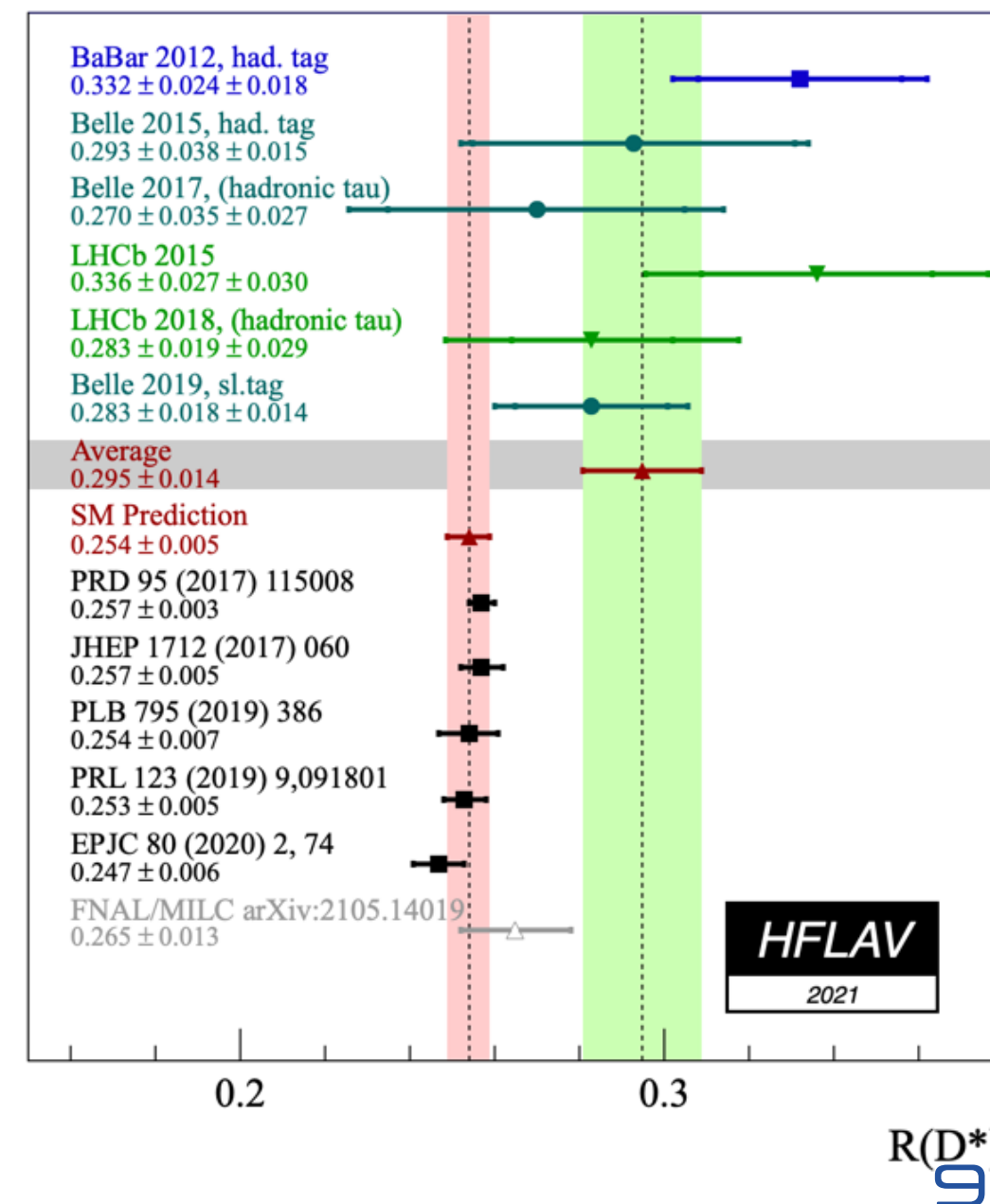
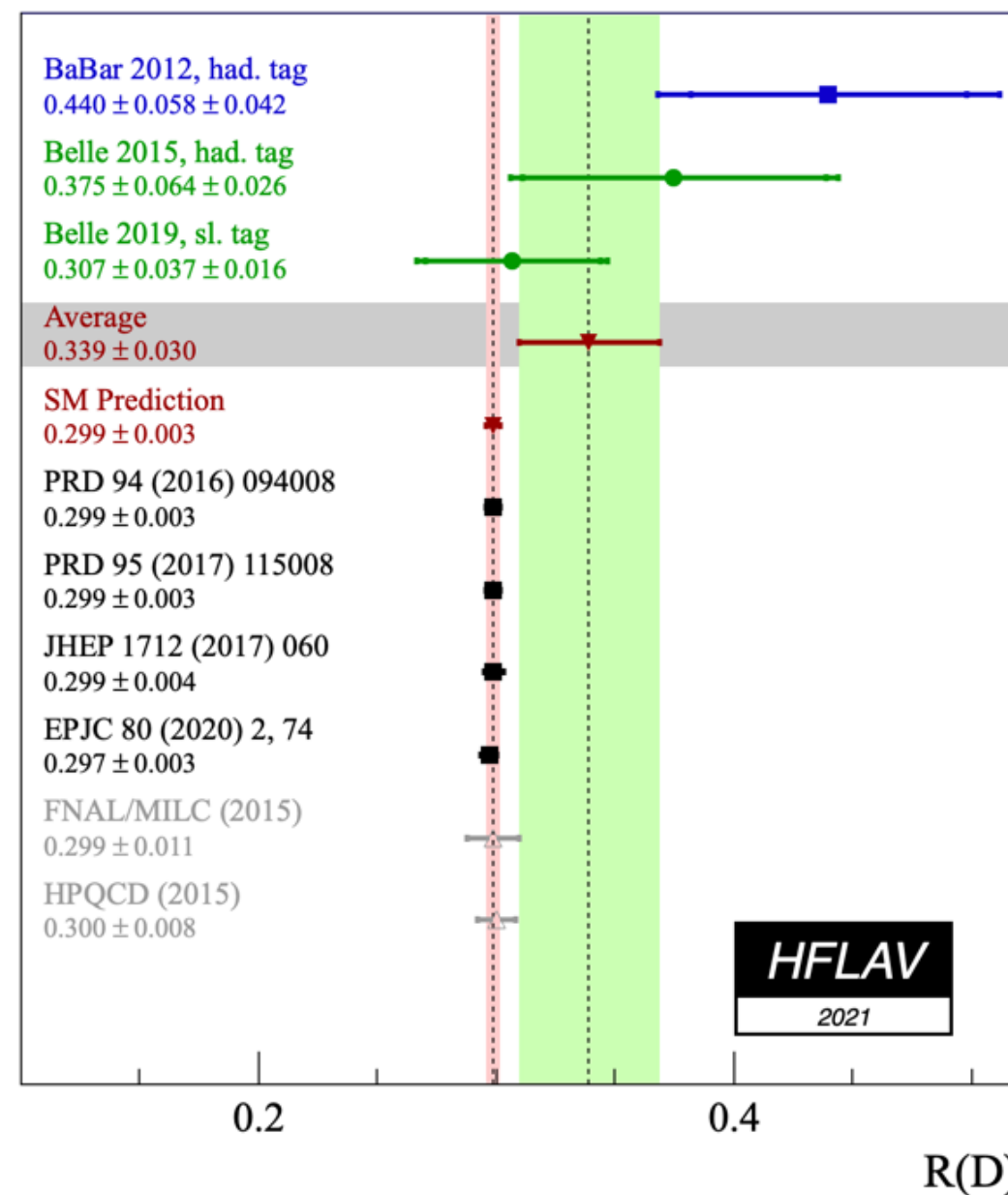
● In  $\tau/\mu$  ratio of  $b \rightarrow c l \nu$  rates

● Confirmation from LHCb at larger statistics + independent experiment



$$\mathcal{H}_{\text{eff}} = \frac{G_F}{\sqrt{2}} V_{tb} V_{ts}^* \sum_i C_i O_i$$

$$R_H = \frac{\int_{q_{\text{min}}^2}^{q_{\text{max}}^2} \frac{d\mathcal{B}(B \rightarrow H\mu^+\mu^-)}{dq^2} dq^2}{\int_{q_{\text{min}}^2}^{q_{\text{max}}^2} \frac{d\mathcal{B}(B \rightarrow He^+e^-)}{dq^2} dq^2} \stackrel{\text{SM}}{\cong} 1$$



# And the biggest excitement

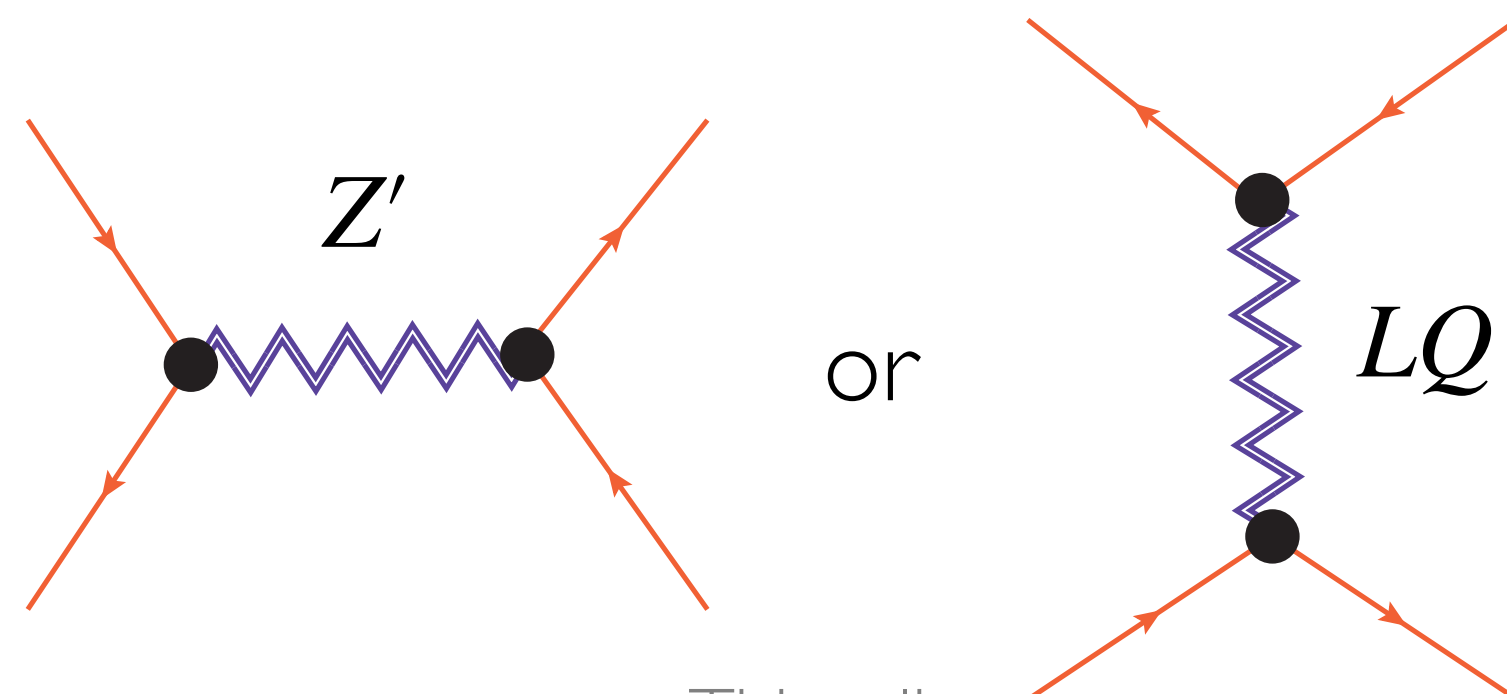
Several implications at high  $p_T$

Leptoquarks

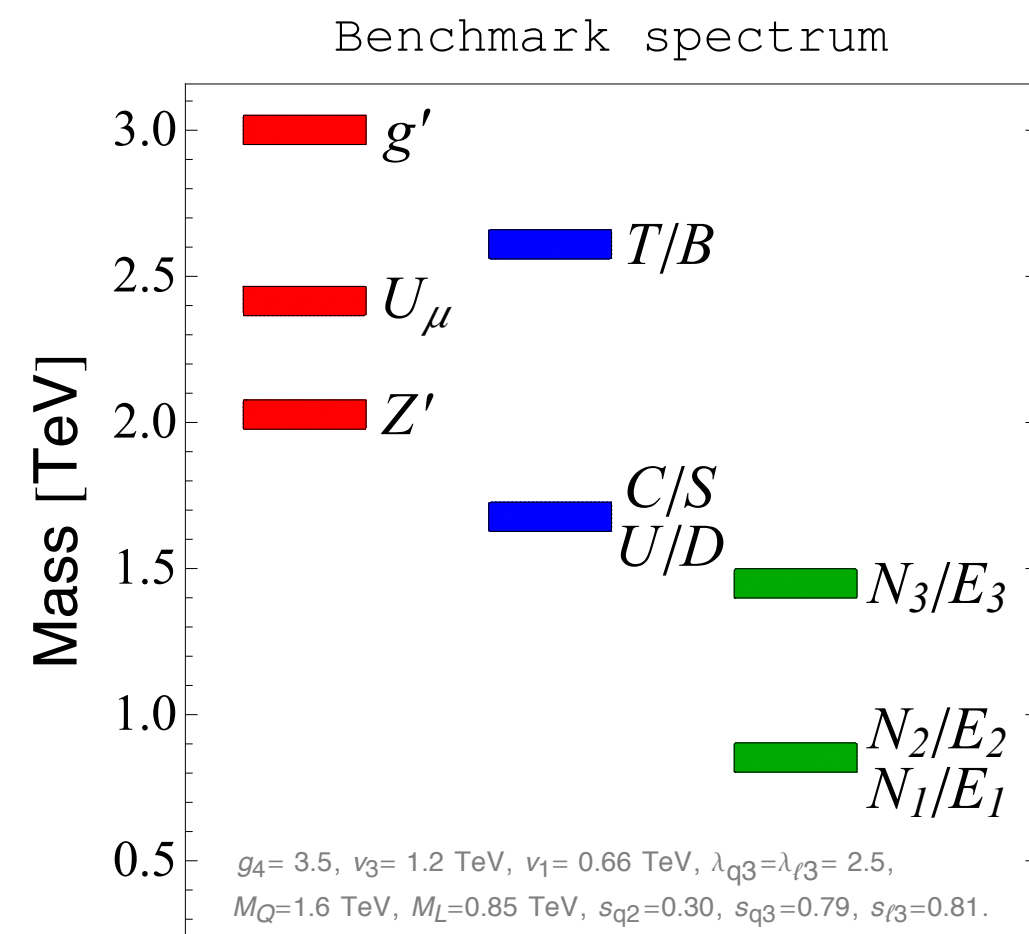
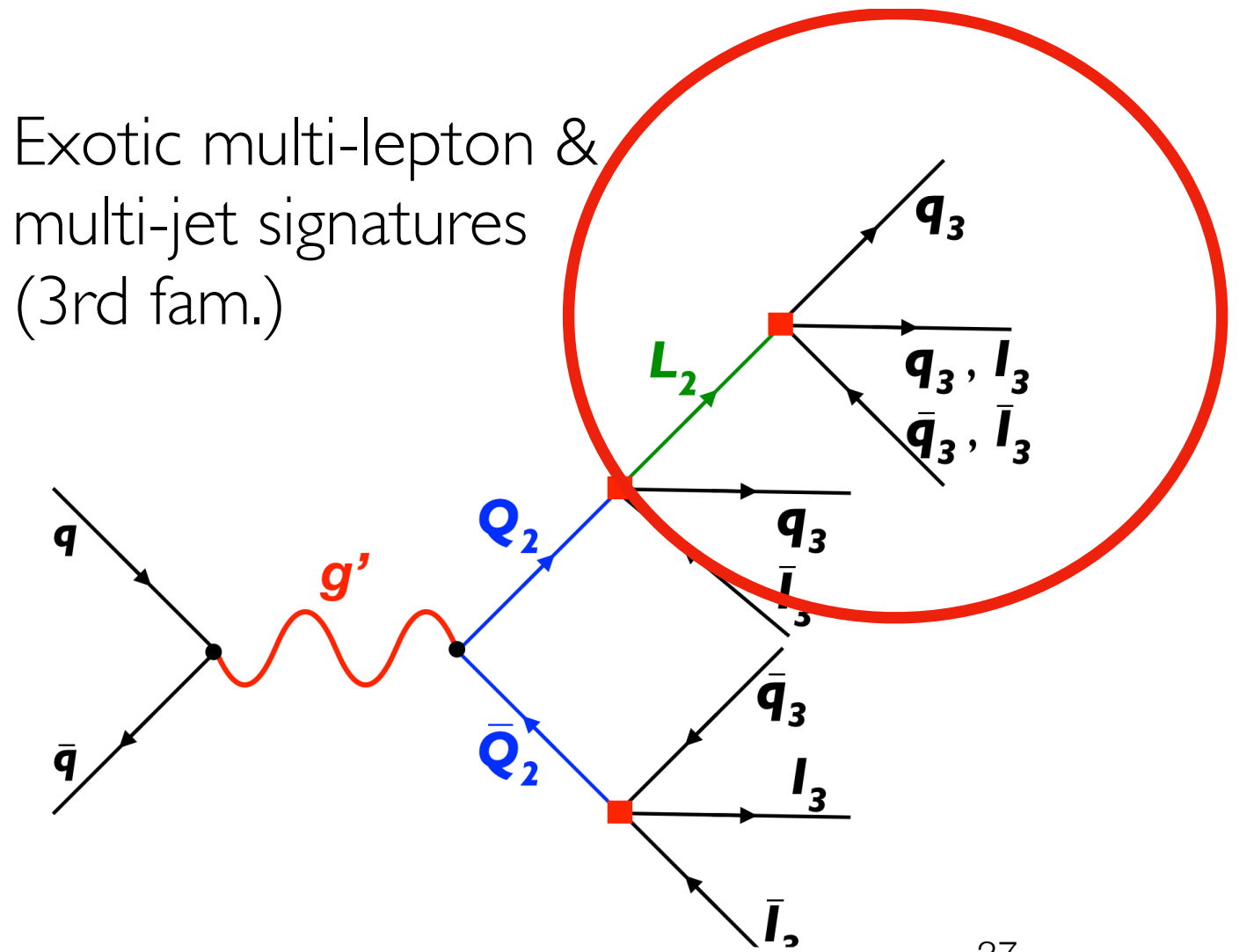
$Z'$

Long decay chains to final states with 3rd generation particles

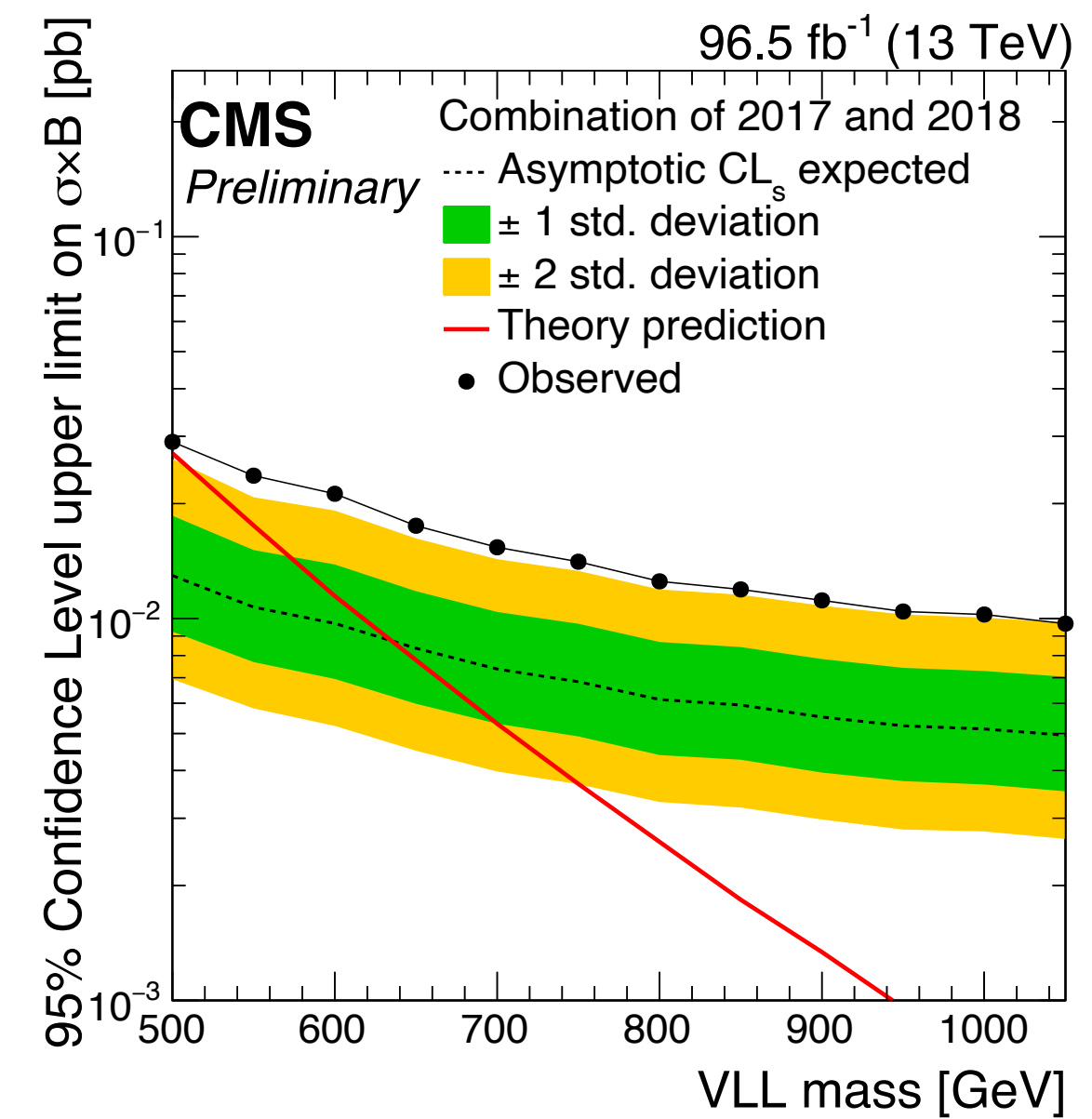
Mild excess observed ( $2.8\sigma$  of significance)



Exotic multi-lepton & multi-jet signatures (3rd fam.)



Di Luzio, Fuentes-Martin, AG, Nardecchia, Renner; 1808.00942



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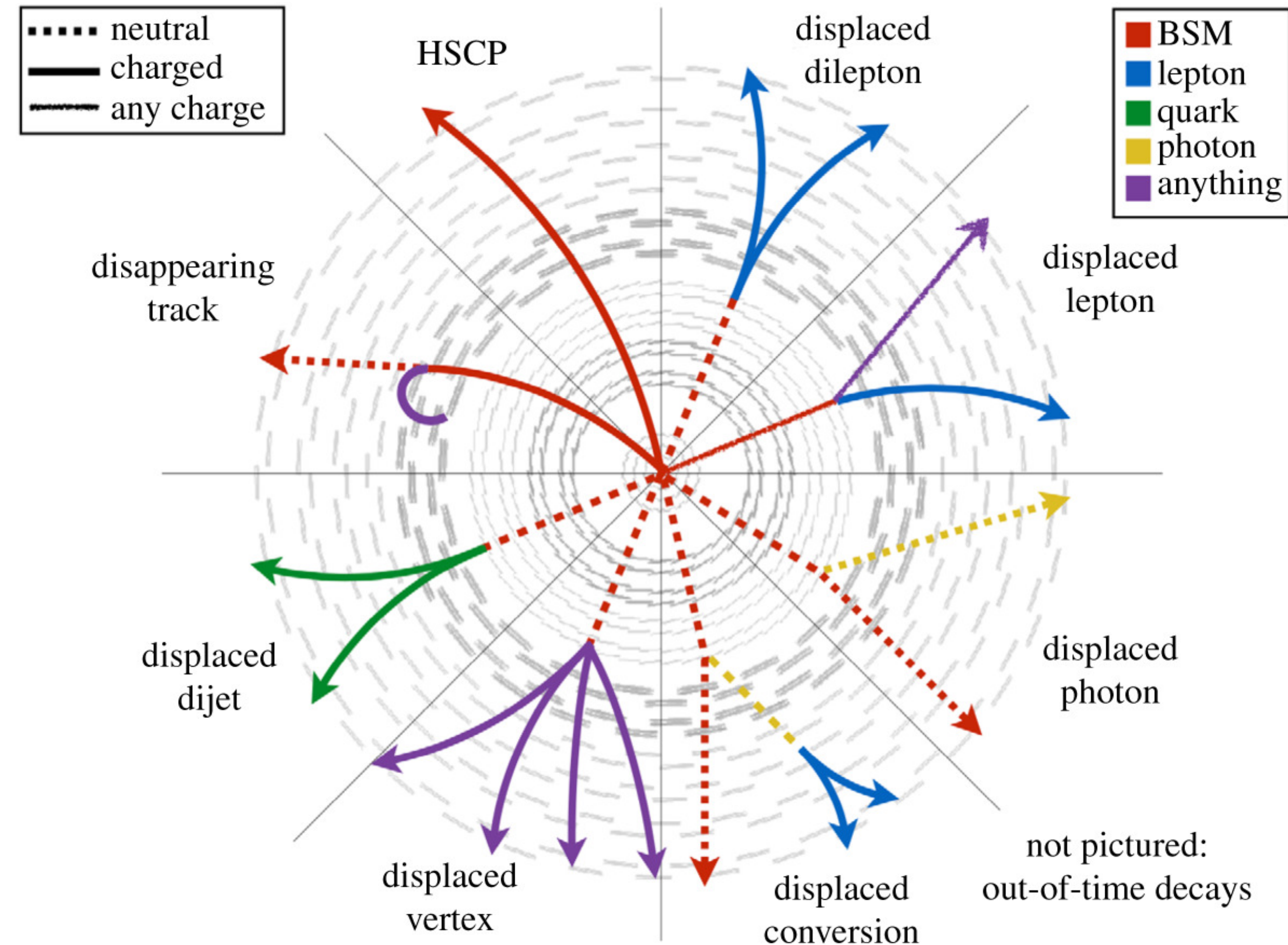
# What should we expect from Run3

## ● Luminosity doubling

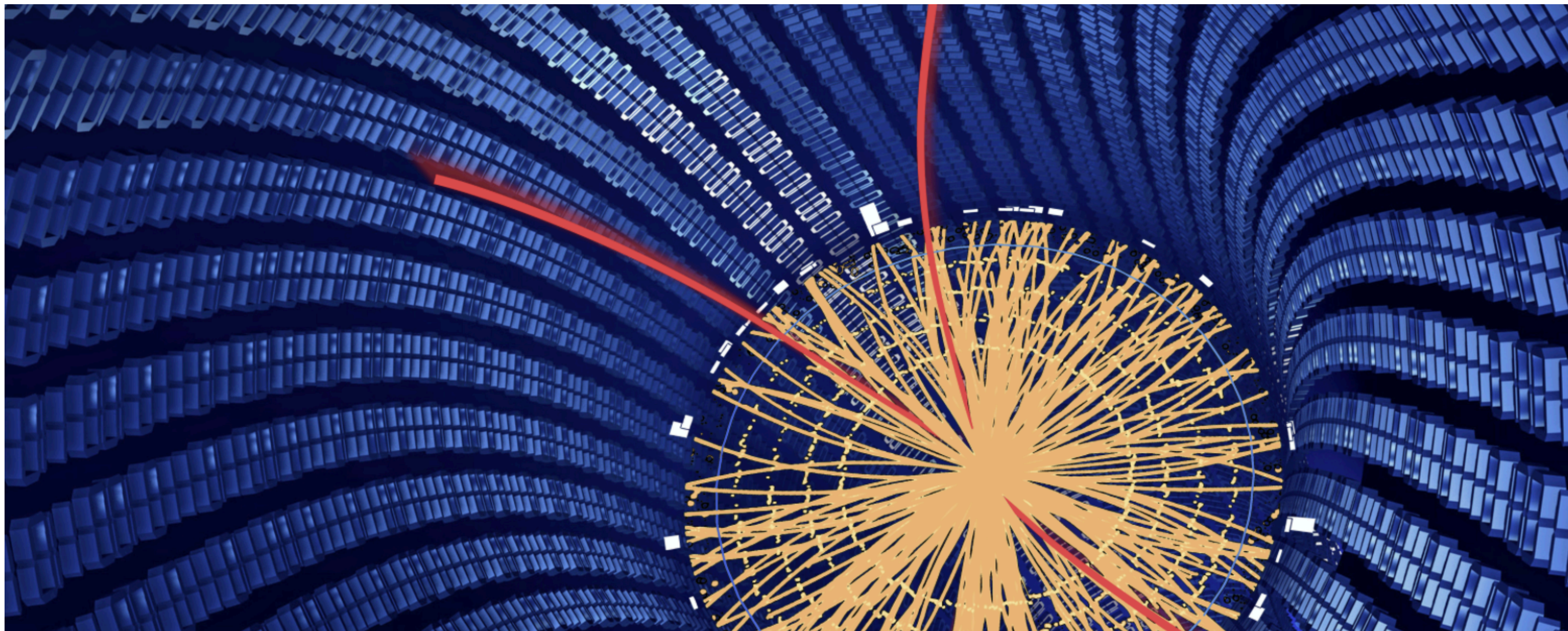
- can improve current results by at least  $\sqrt{2}$
- can investigate previous excesses with comparable datasets

## ● New trigger opportunities

- Improved algorithms and hardware let us do more than in the past
- New triggers imply new (i.e., unexplored) territory, basically a new experiment
- For instance, a big push in this direction from long-lived particle searches



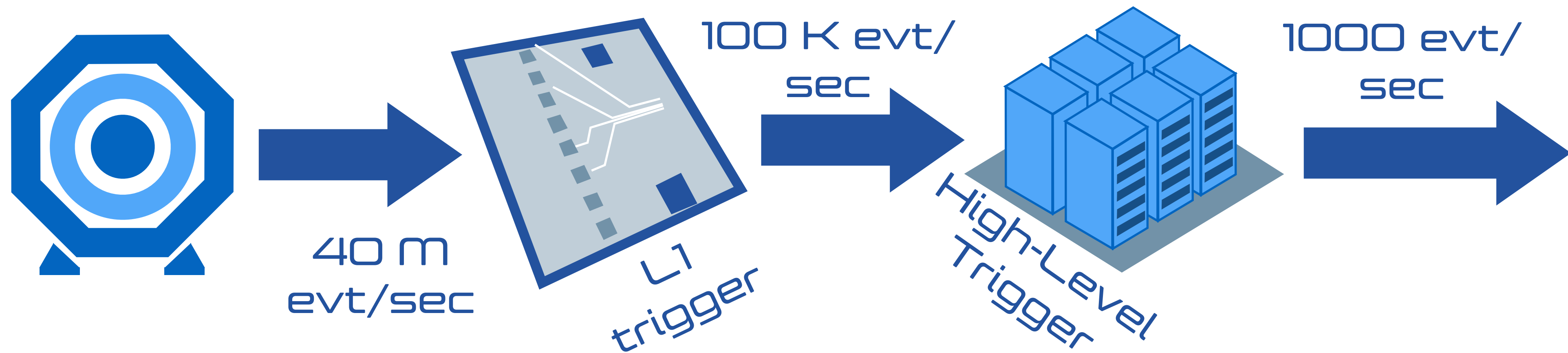




# Alternative Data Taking Strategies: scouting



# The LHC Big Data Problem

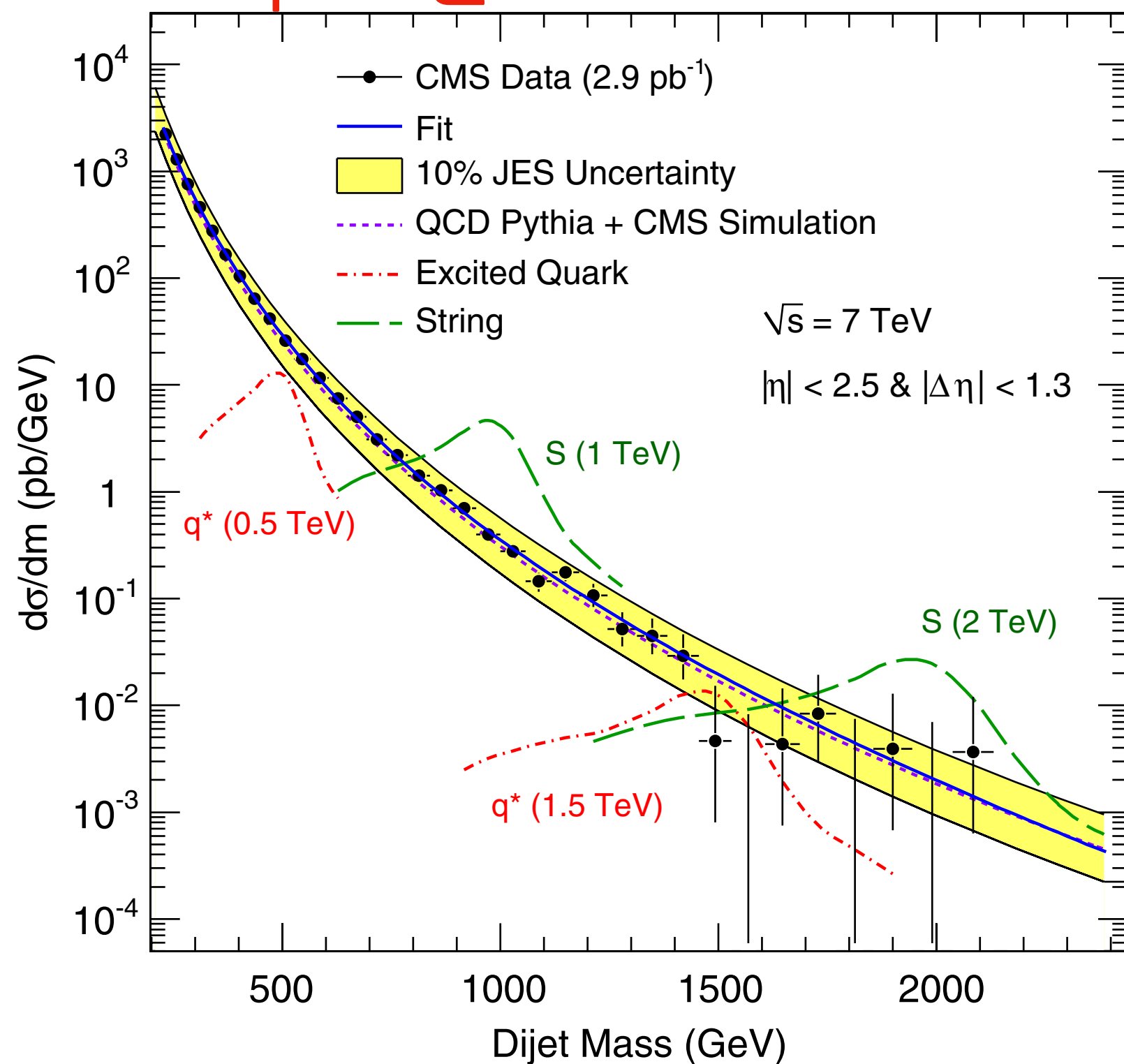


- ◎ *LHC produces more events than what we can store*
- ◎ *We then filter them using trigger*
- ◎ *The menu is a negotiation between different physics topics and the rate of the corresponding processes*
- ◎ *Many studies are trigger limited*

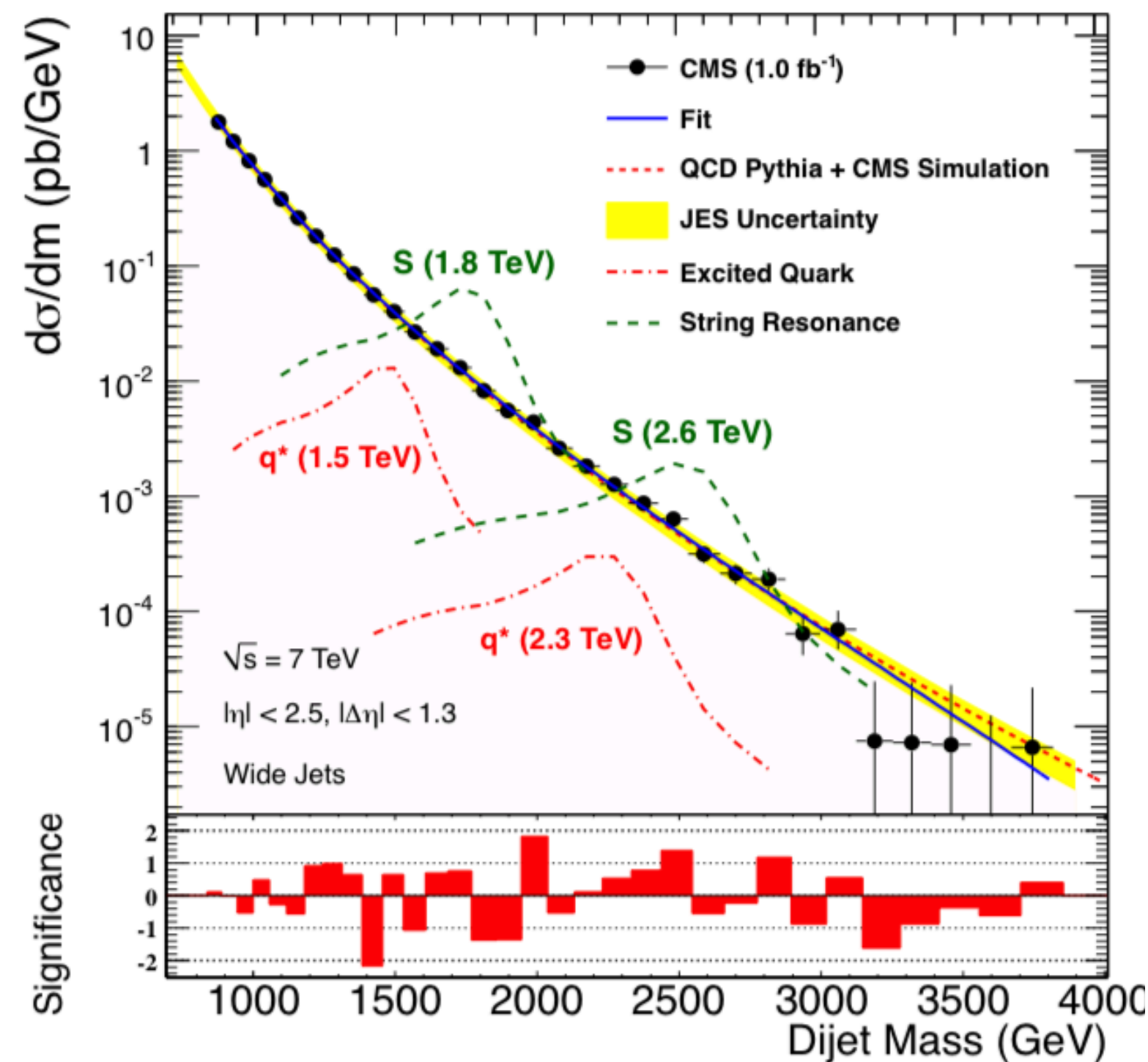
# Example: Dijet resonance searches

- Most of LHC events have two jets
- If one wants to look for dijet resonances, a hard cut on jet energy has to be applied
- For this reason, we cannot keep sensitivity to low-mass resonances

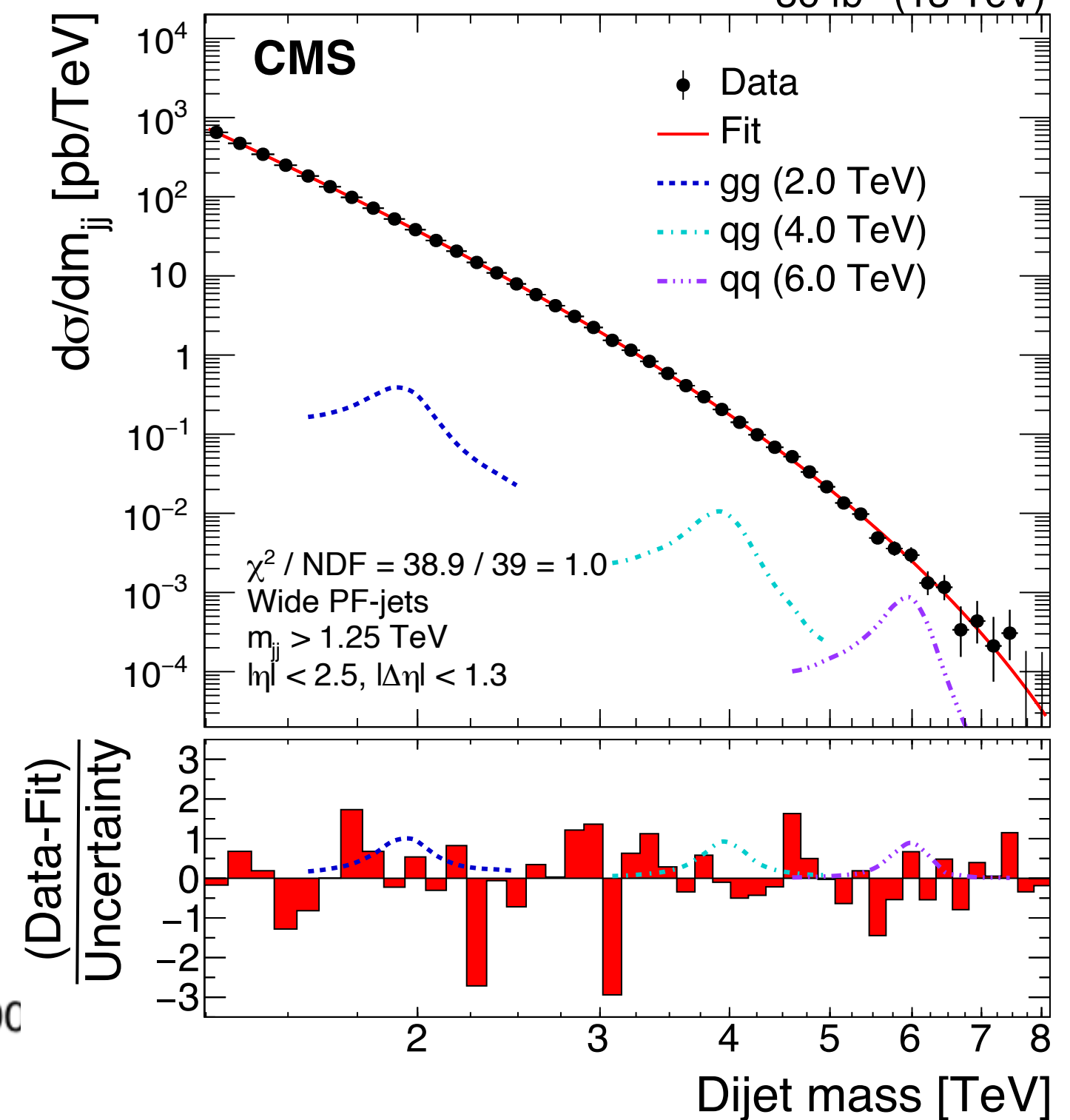
3 pb<sup>-1</sup> @7 TeV in 2010



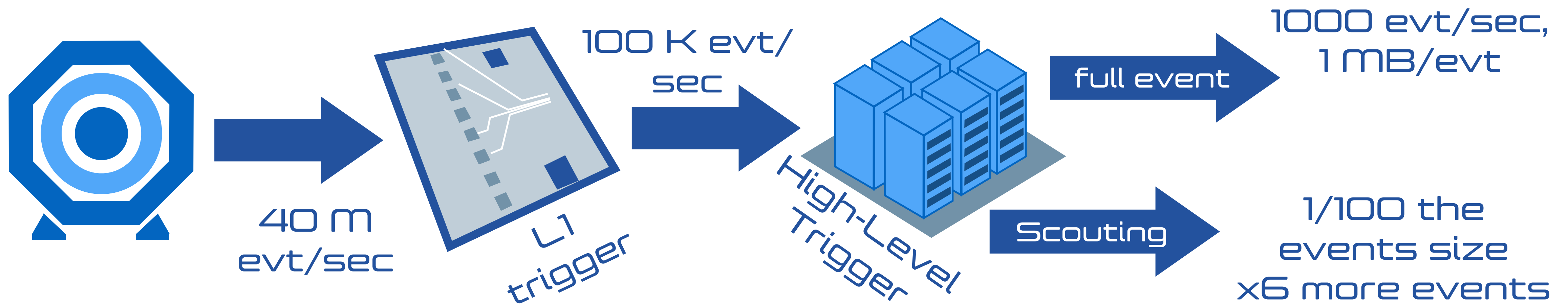
1 fb<sup>-1</sup> @7 TeV in 2011



36 fb<sup>-1</sup> @13 TeV in 2016



# The Scouting Streams



◎ *Trade-off between # of events & event size*

◎ *write HLT objects (four-momenta, etc...) rather than full event content*

◎ *write many more events (essentially HLT passthrough, limited by L1)*

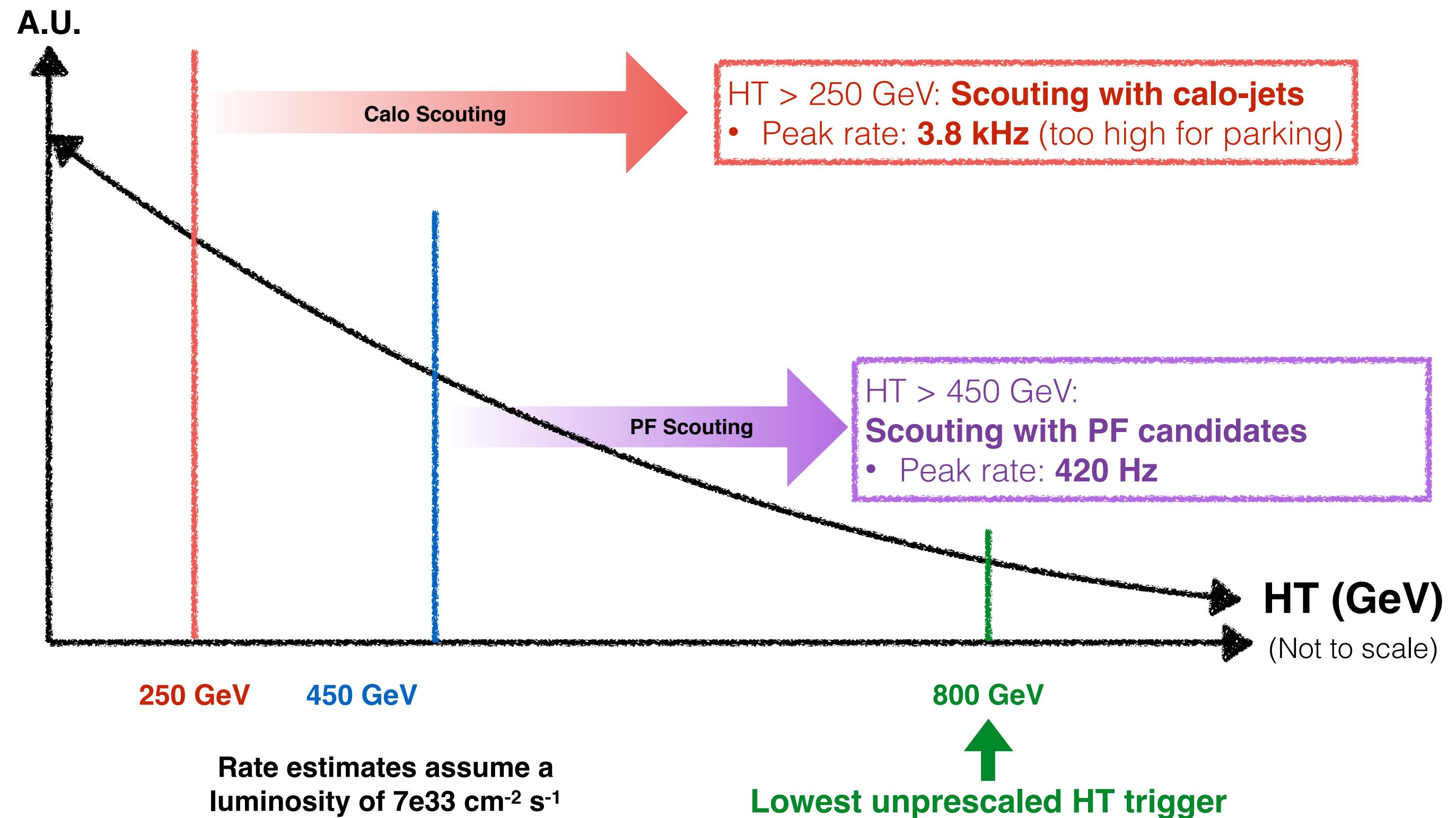


# DiJet Scouting Streams

● *DiJet Data Scouting strategy in Run2 used two tiers*

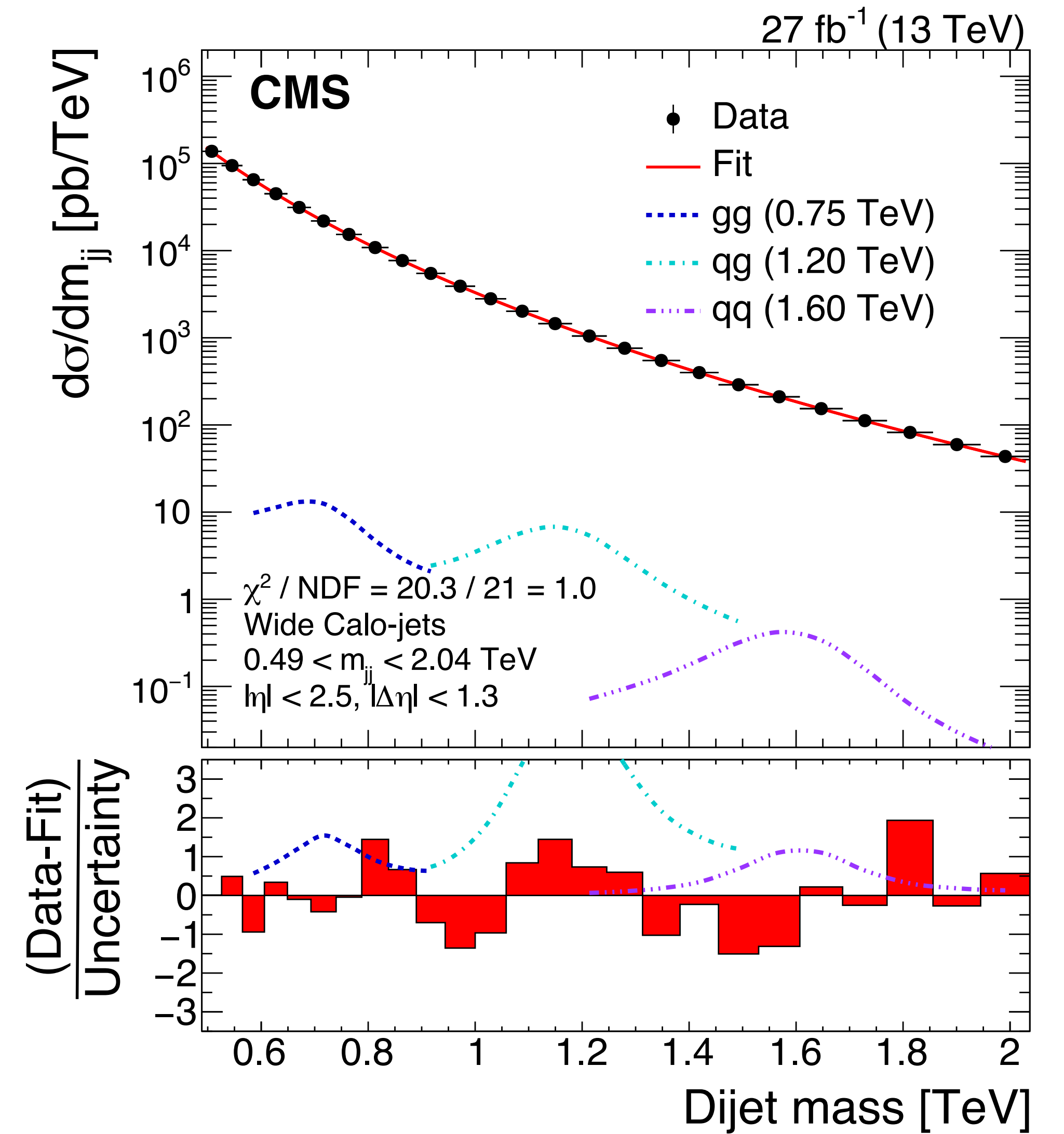
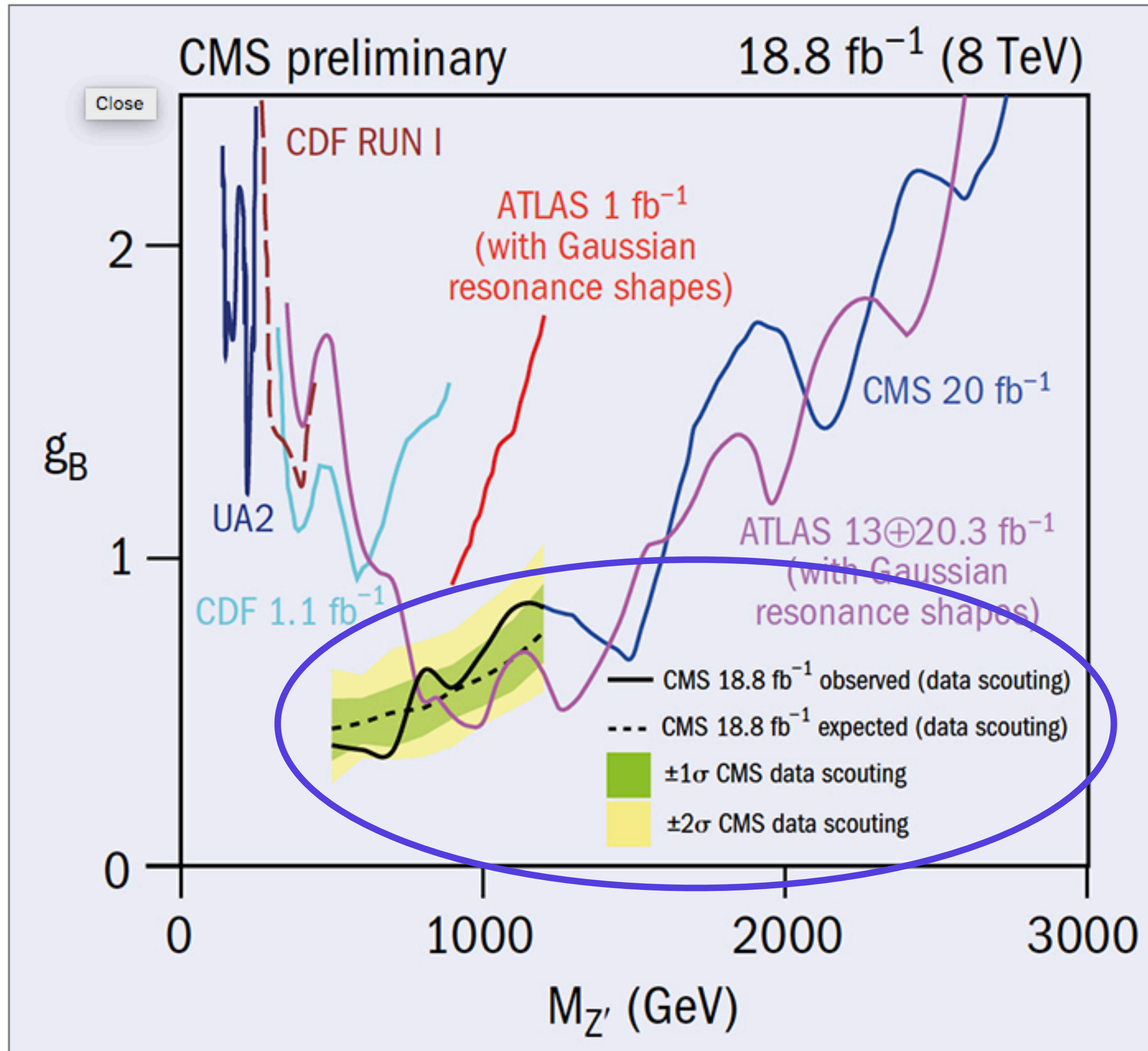
● *A calo-scouting stream with minimal event content, to go as down as possible (limited by L1 trigger)*

● *A PF-scouting stream, going as down as HLT let us run the PF sequence. More complex event content and more flexibility (substructure, etc)*



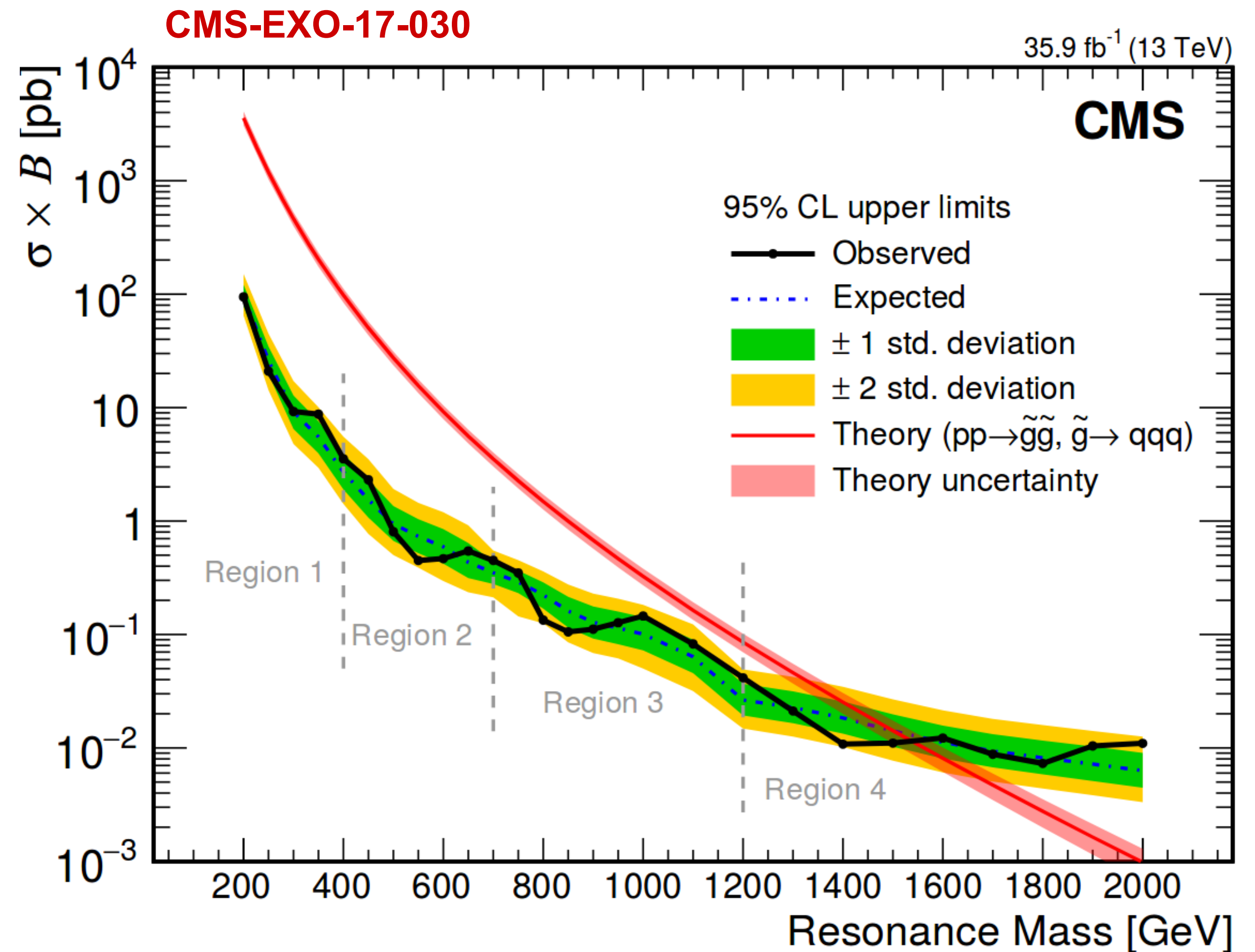


# DiJet Scouting Streams



# More than dijet

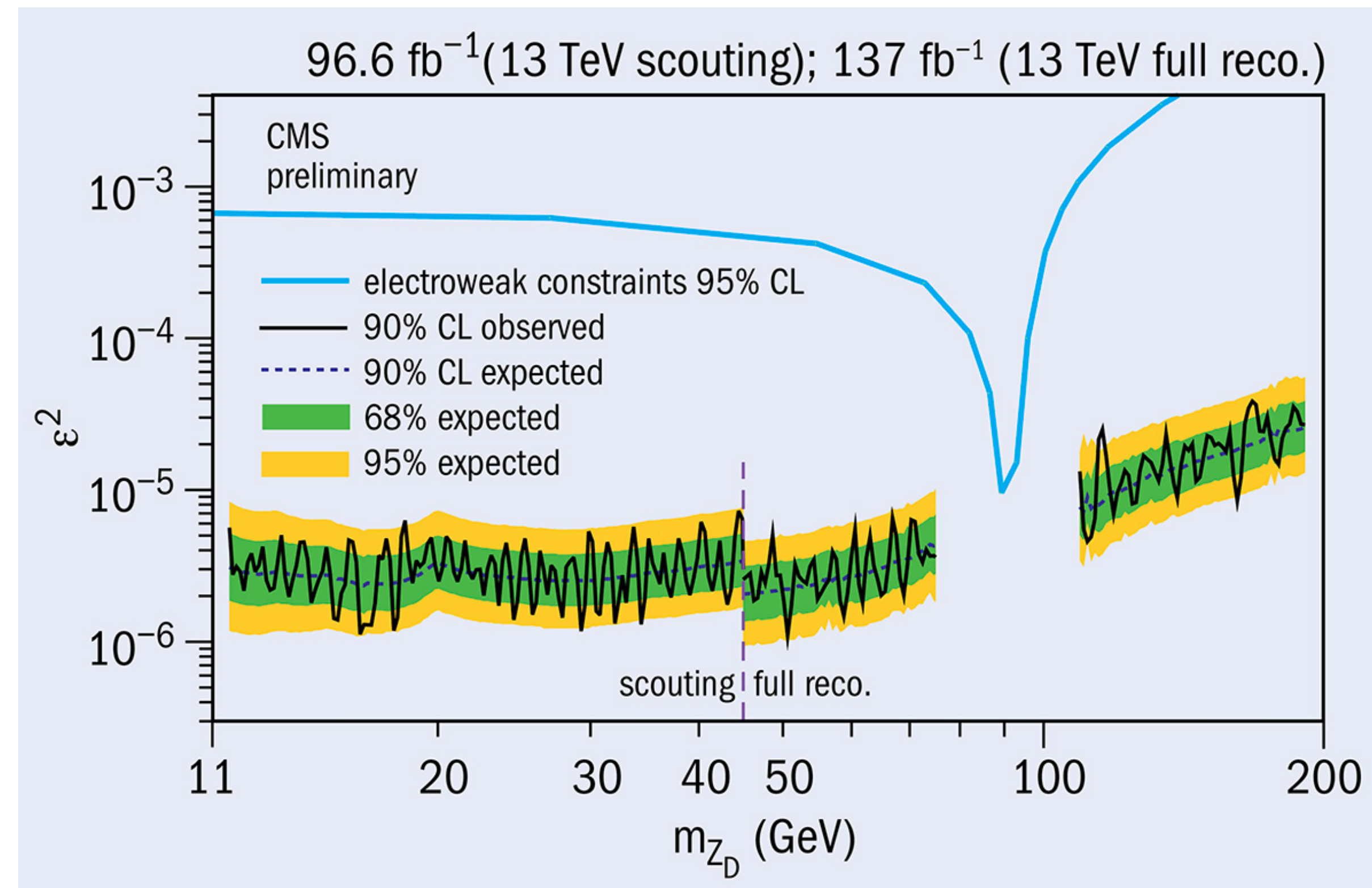
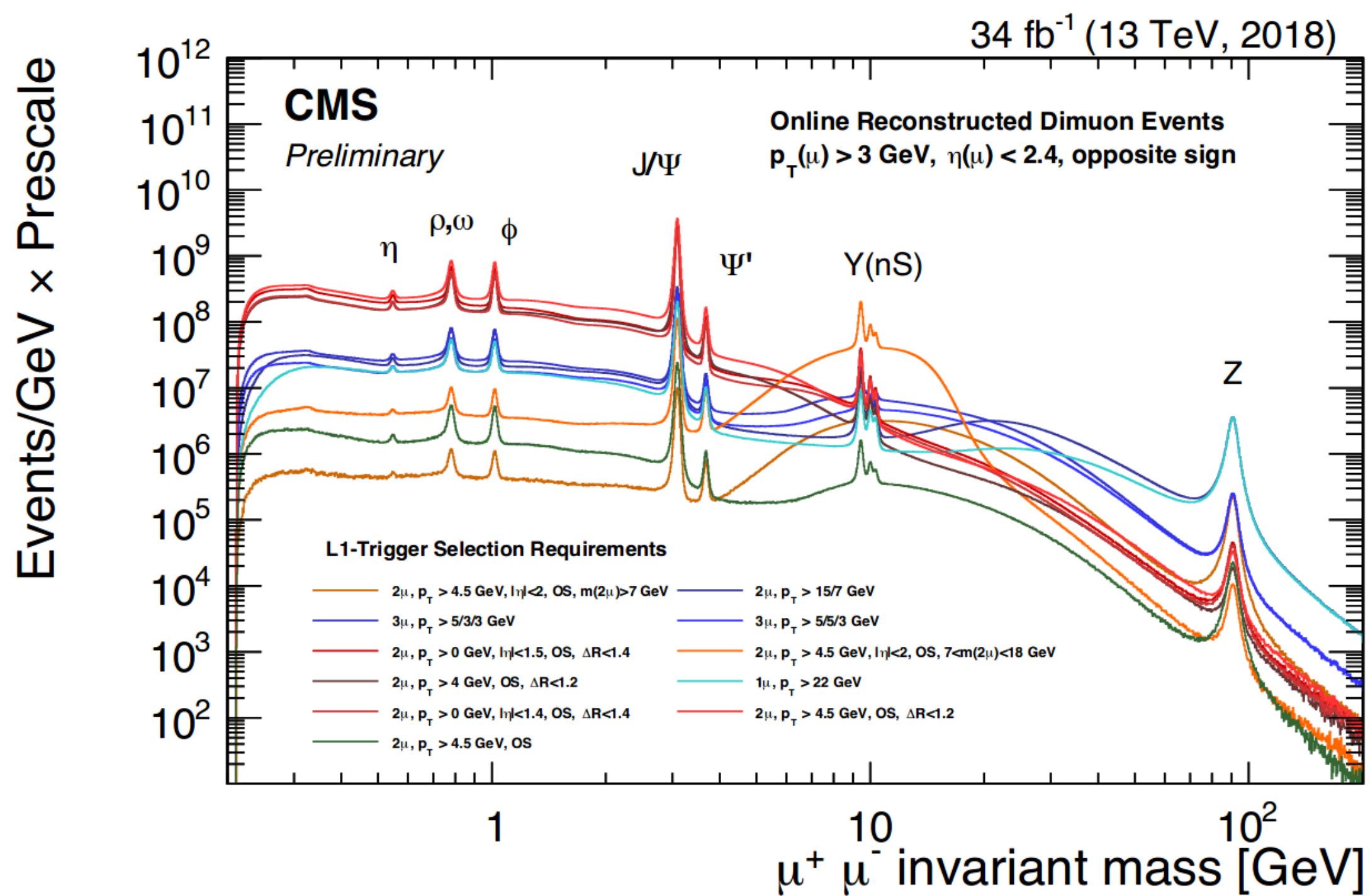
- ◎ Same approach followed for other final states
- ◎ multiple jet resonances (e.g., trijet in
- ◎ jet substructure
- ◎ ...
- ◎ In particular, the use of PF (with full access to individual particles) showed a great potential, which so far was barely explored





# Dimuon Scouting Stream

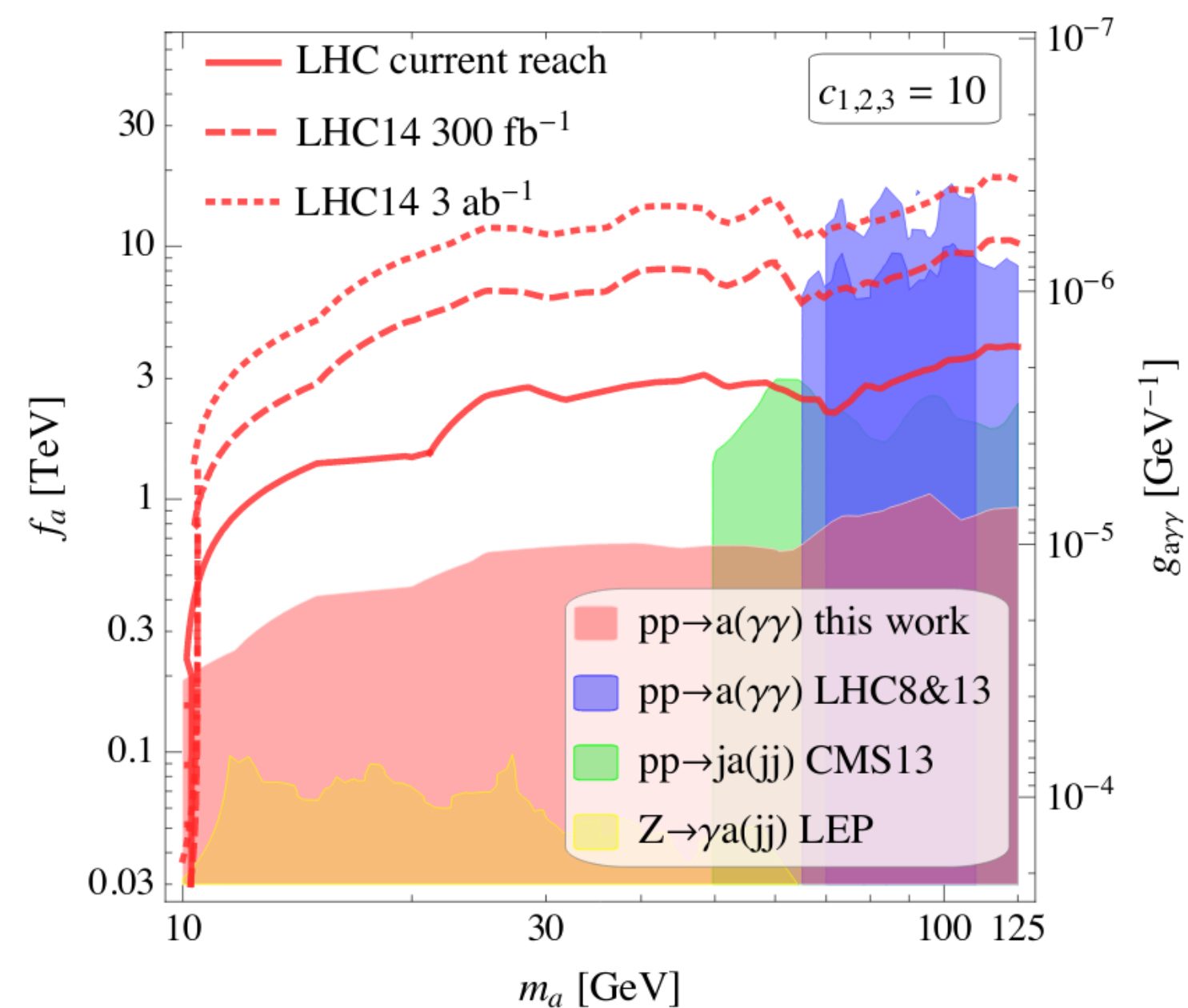
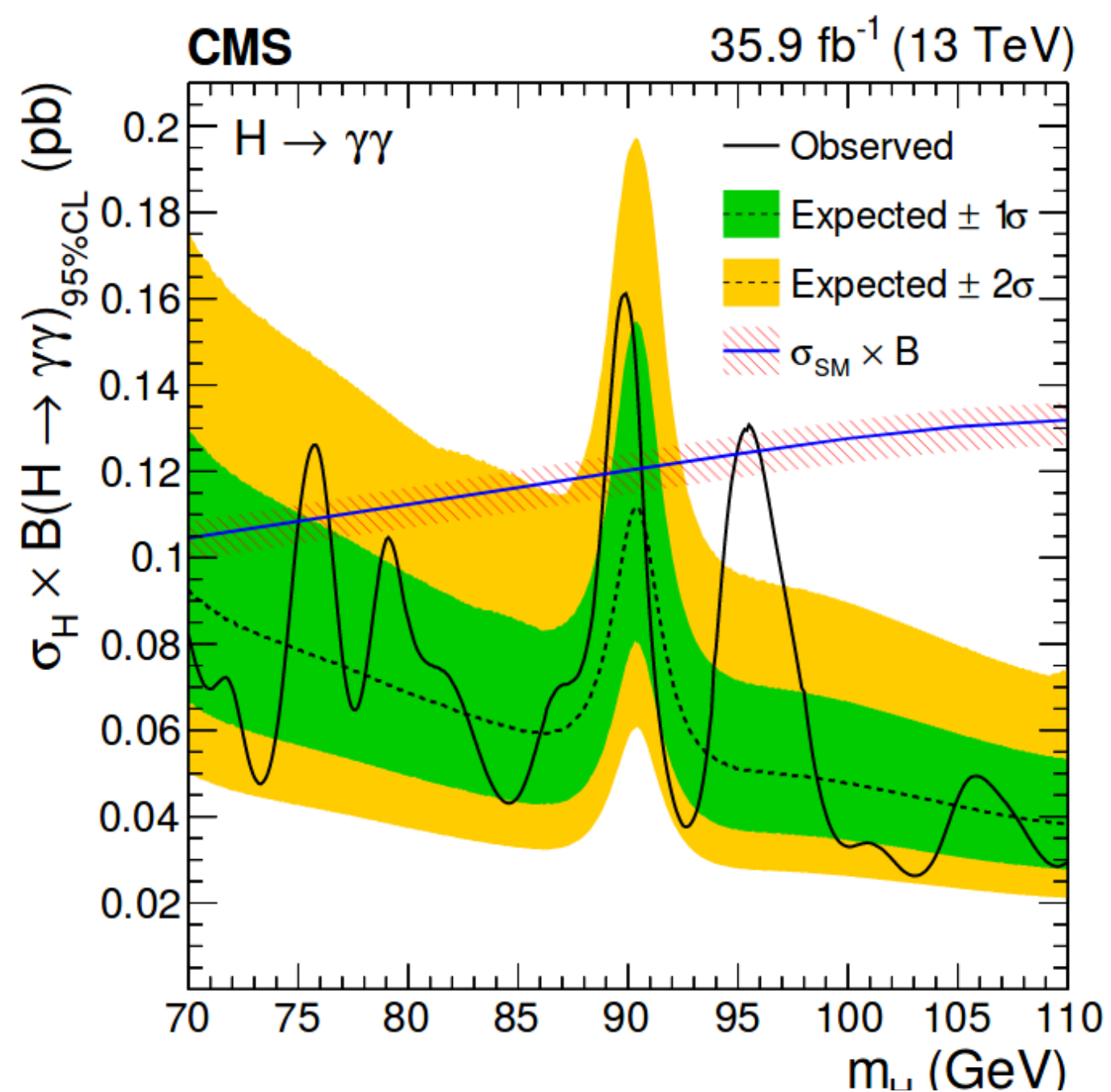
- Starting in 2015, the same idea was applied to dimuon events, to probe light dimuon resonances (prompt and displayed)
- Similar to LHCb turbostream analysis
- Probed competitive Dark Photon parameter space



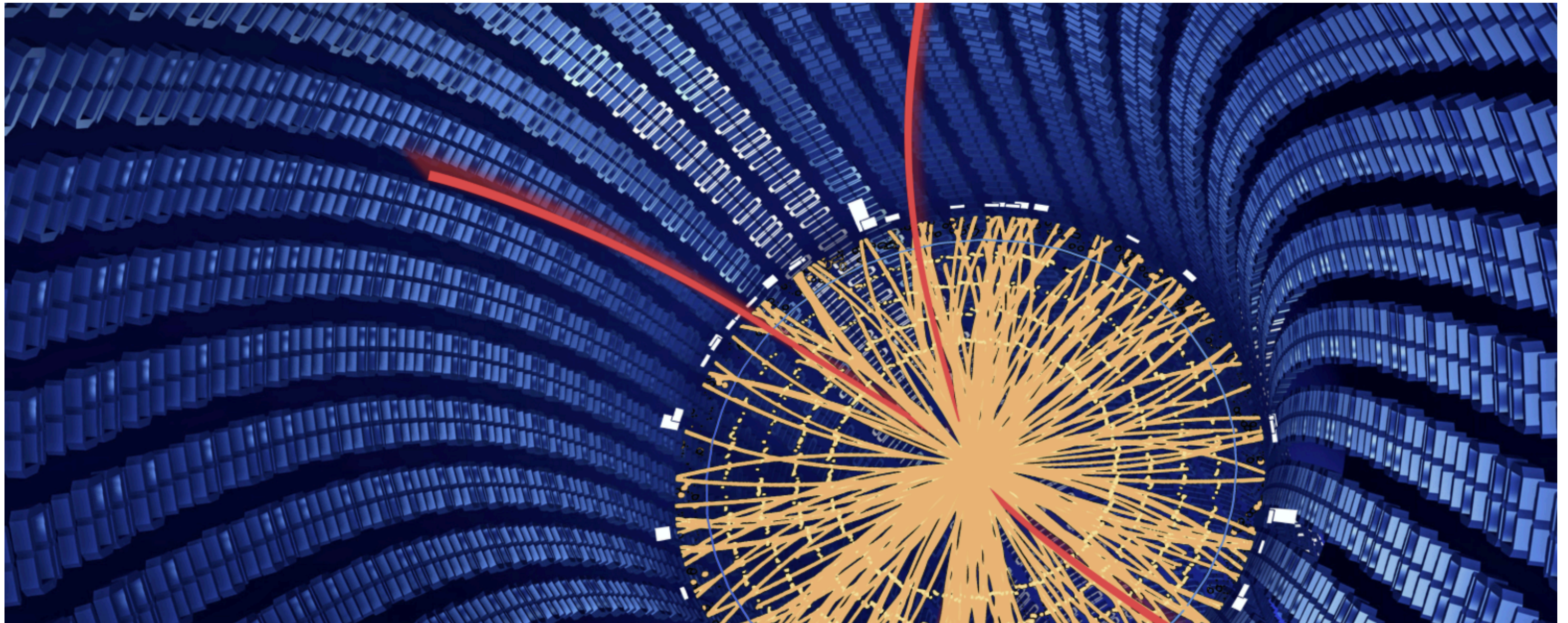


# Scouting plans for Run3

- Improved Trigger farm (heavy computing now on GPUs) makes it possible to run PF algorithm on a much larger dataset
- Can use PF scouting on  $O(10K)$  evt/sec, storing an event content including all PF particles
- Could boost the physics case of the scouting stream



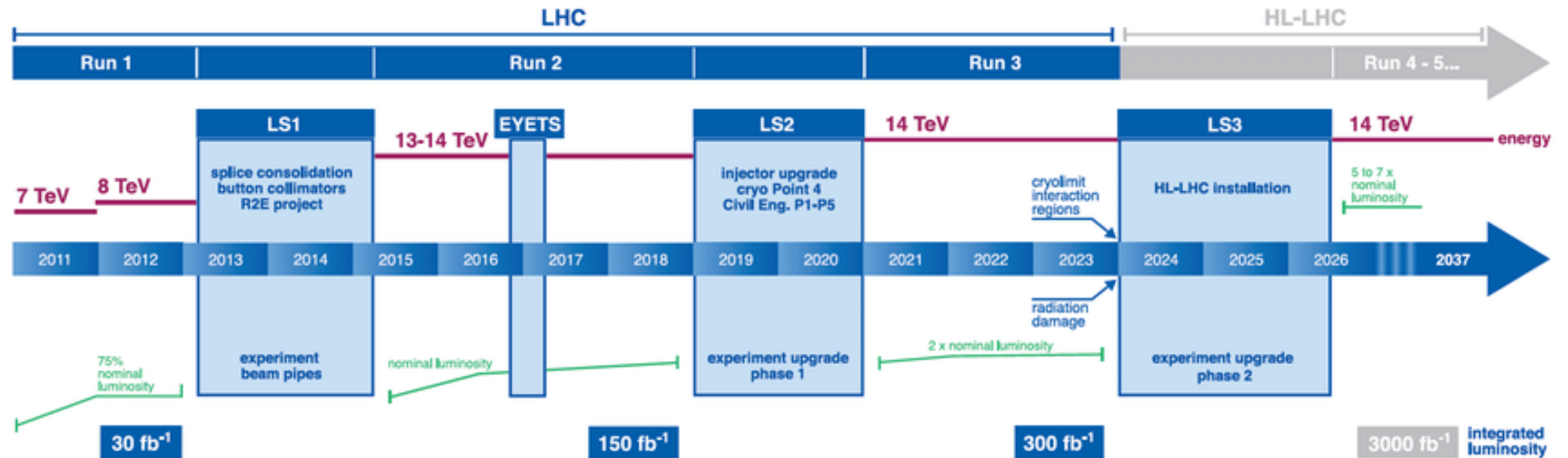




# Alternative Data Taking Strategies: parking

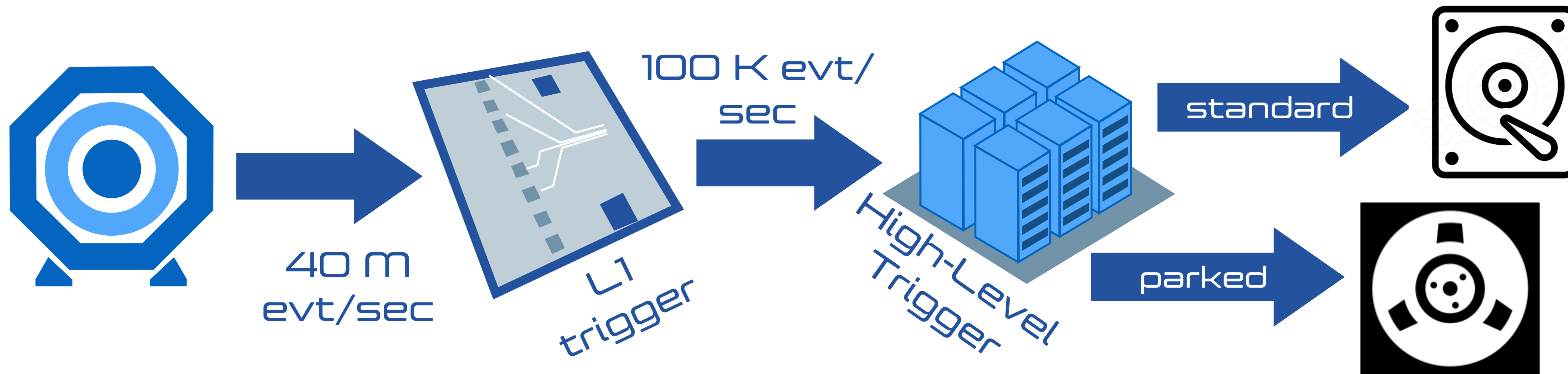


# LHC schedule: Runs and LeSs



- ⊙ *When we take data, we saturate our computing facilities with the data we take*
- ⊙ *We could take more, but then we don't have enough computing power*
- ⊙ *During long shutdowns, our CPUs are under-utilised (e.g., for MC production)*

# Data Parking

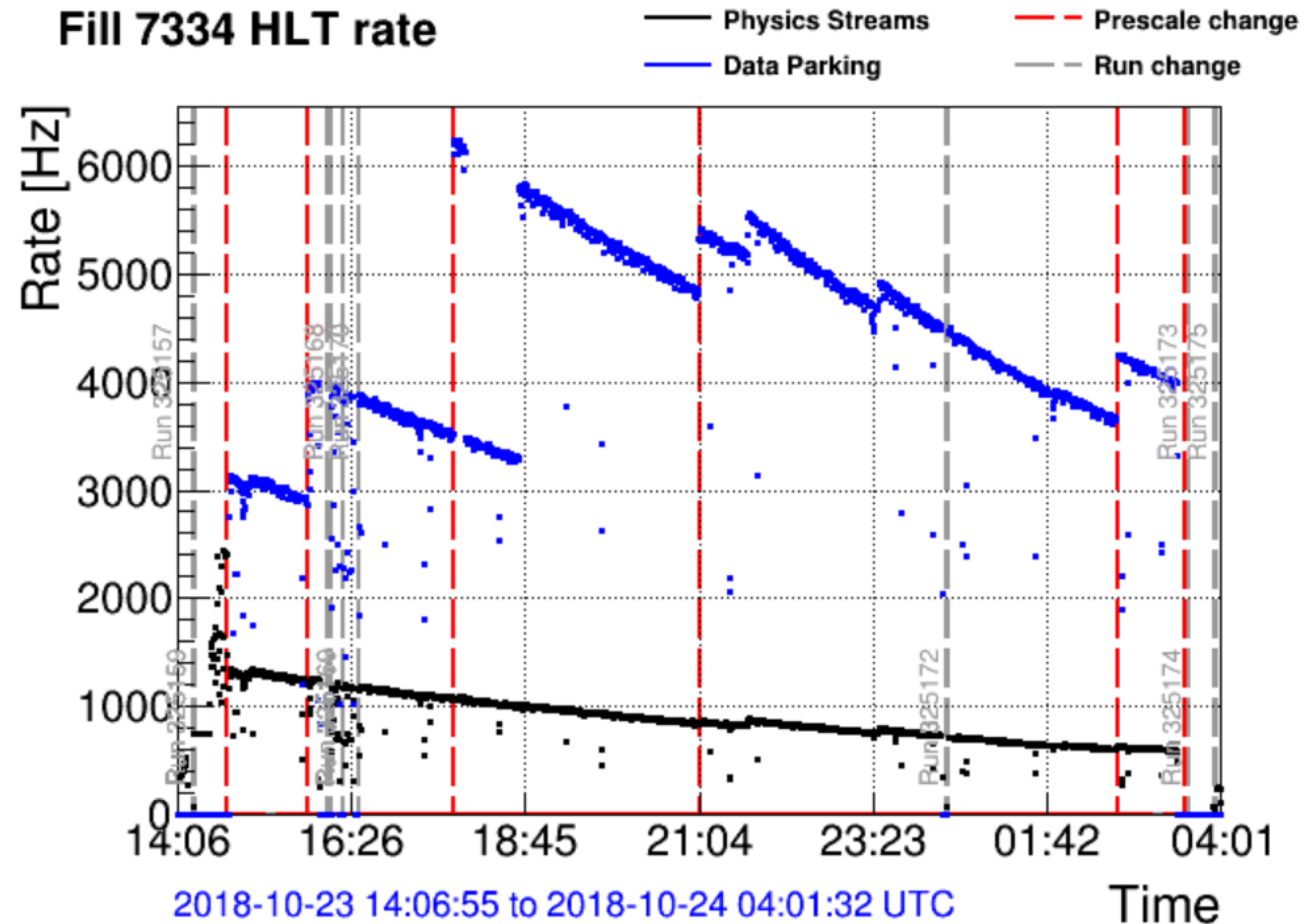


- ⊙ *Since Run 1, we open up the triggers during the last year before a LS*
- ⊙ *We send the events on tape (cheaper than disk) and reconstruct them during the shutdown*
- ⊙ *We target specific physics use cases*
  - ⊙ *Inclusive/invisible Higgs in Run 1*
  - ⊙ *B physics in Run 2*

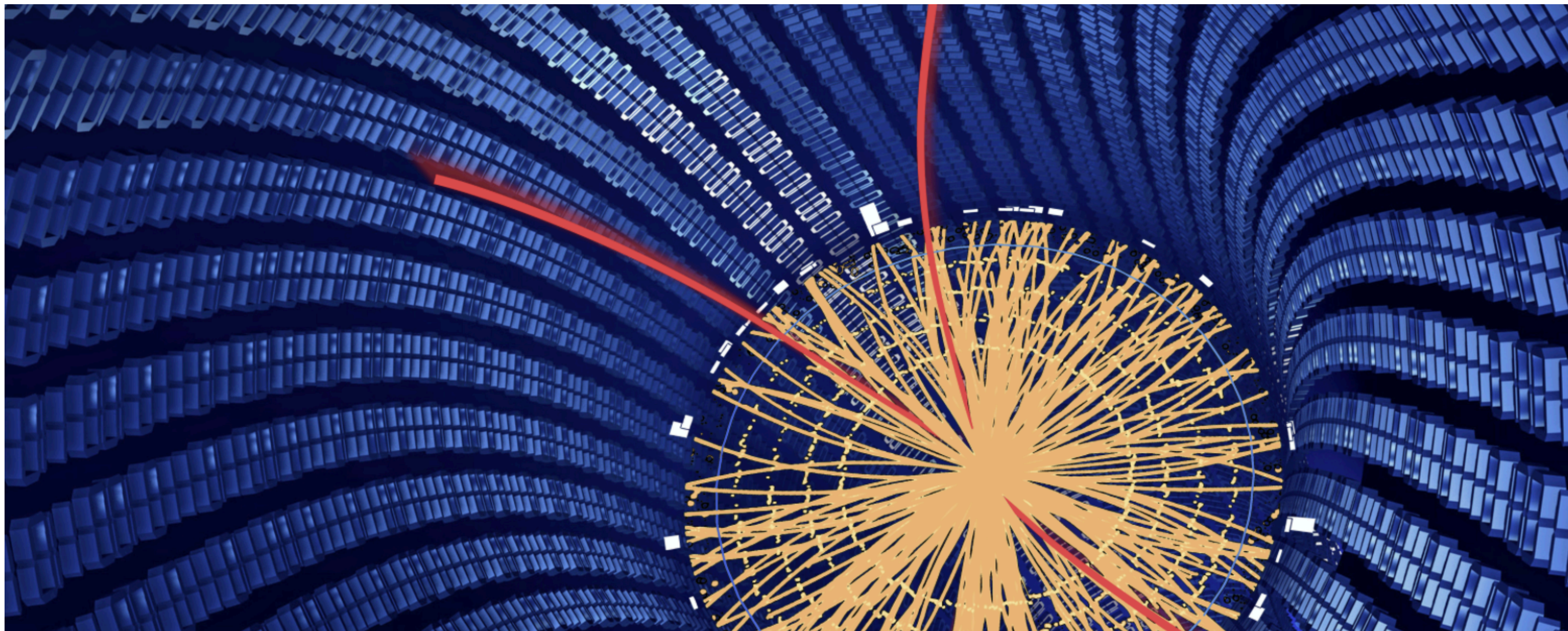


# B-physics Parking

- ⦿ *Saturated available rate with displaced muon trigger*
- ⦿ *Selected ~80% pure sample of BB mesons*
  - ⦿ *one leg biased by trigger*
  - ⦿ *one leg trigger-selection free*
- ⦿ *Used to establish a physics program targeting the LHCb anomalies*
  - ⦿ *Studies ongoing*
  - ⦿ *More to come in Run3*





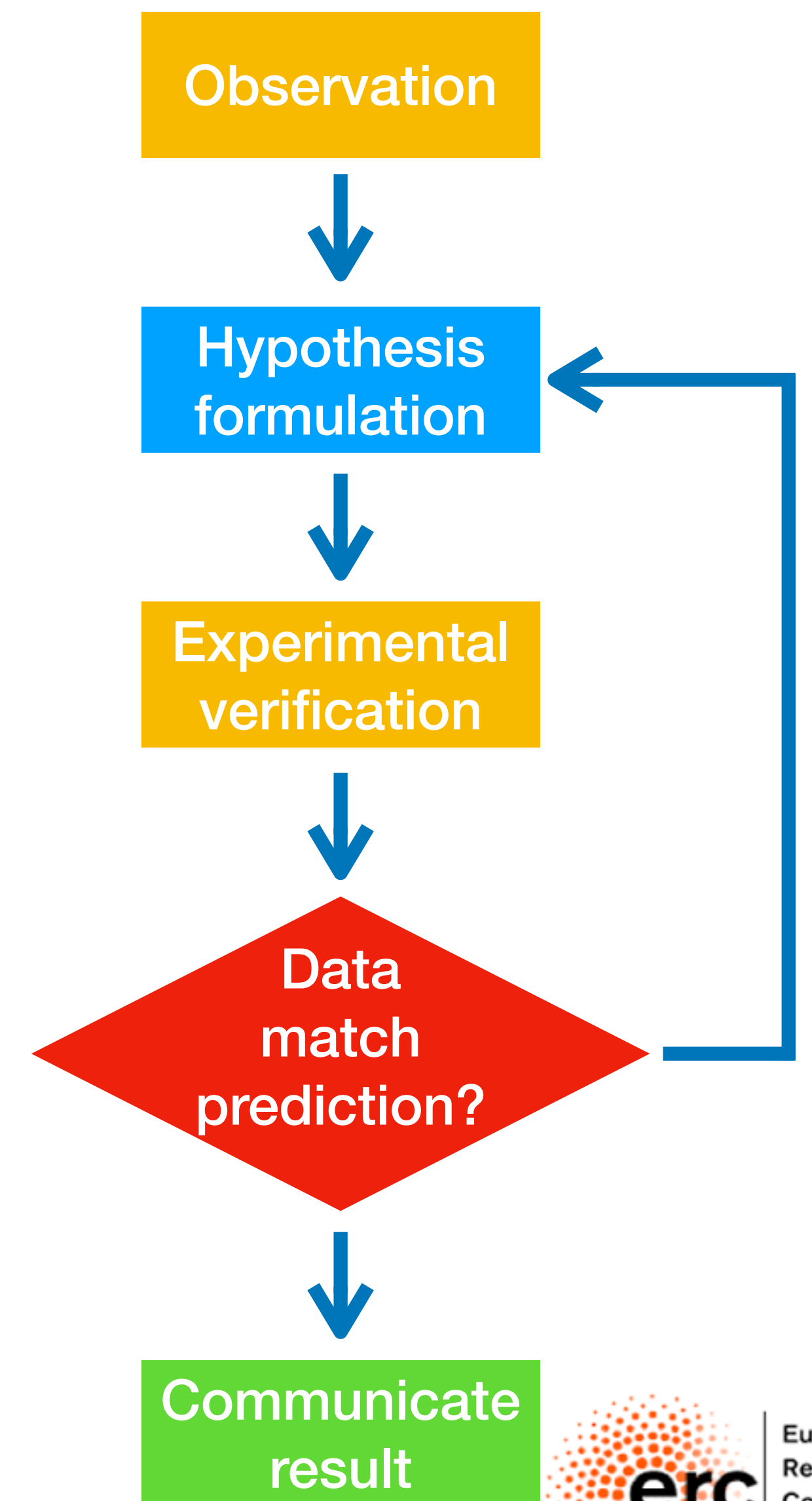


# Alternative Data Taking Strategies: anomaly detection



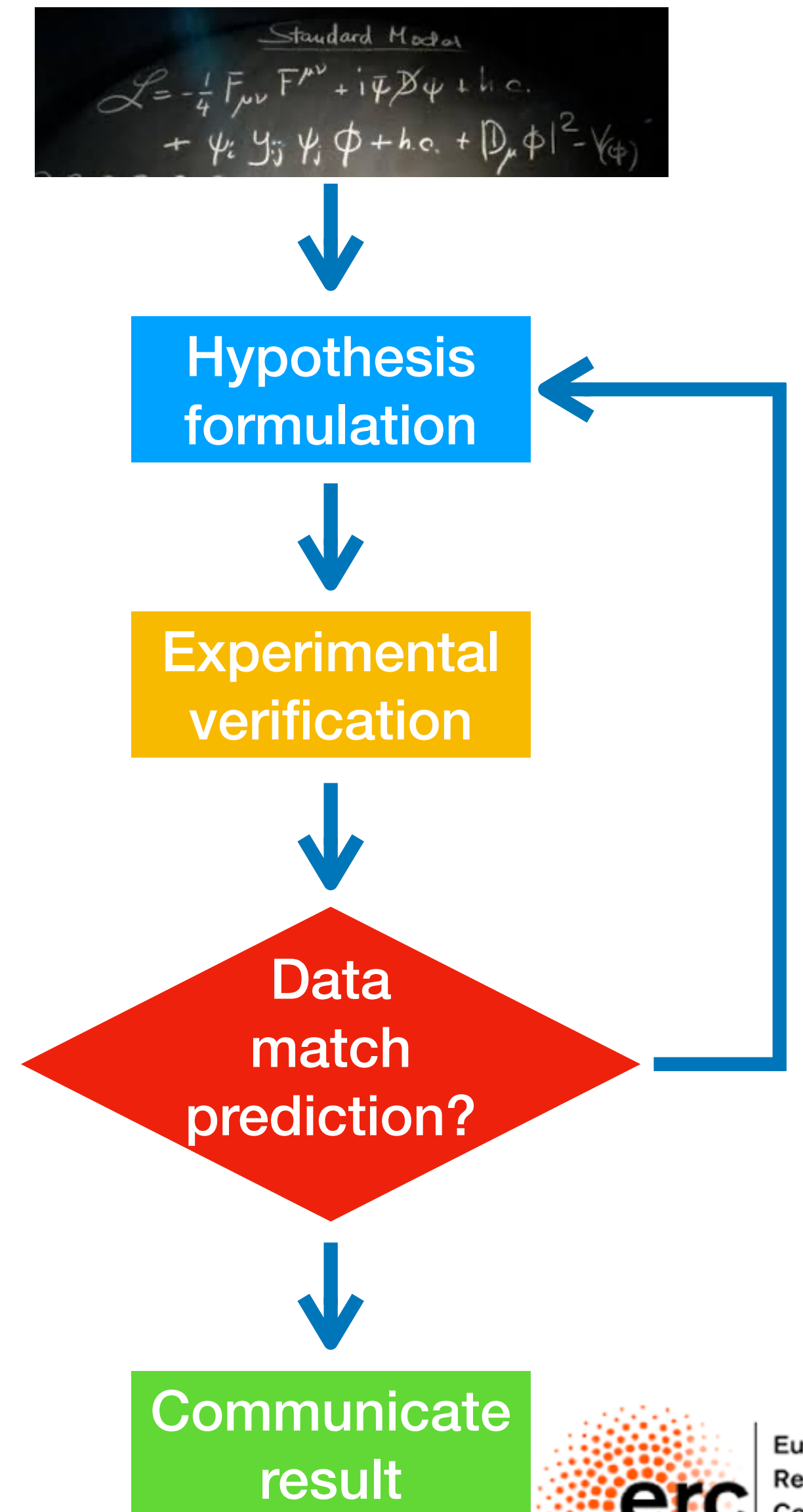
# HEP searches in LHC era

- *The first step of the scientific method consists in observing nature*



# HEP searches in LHC era

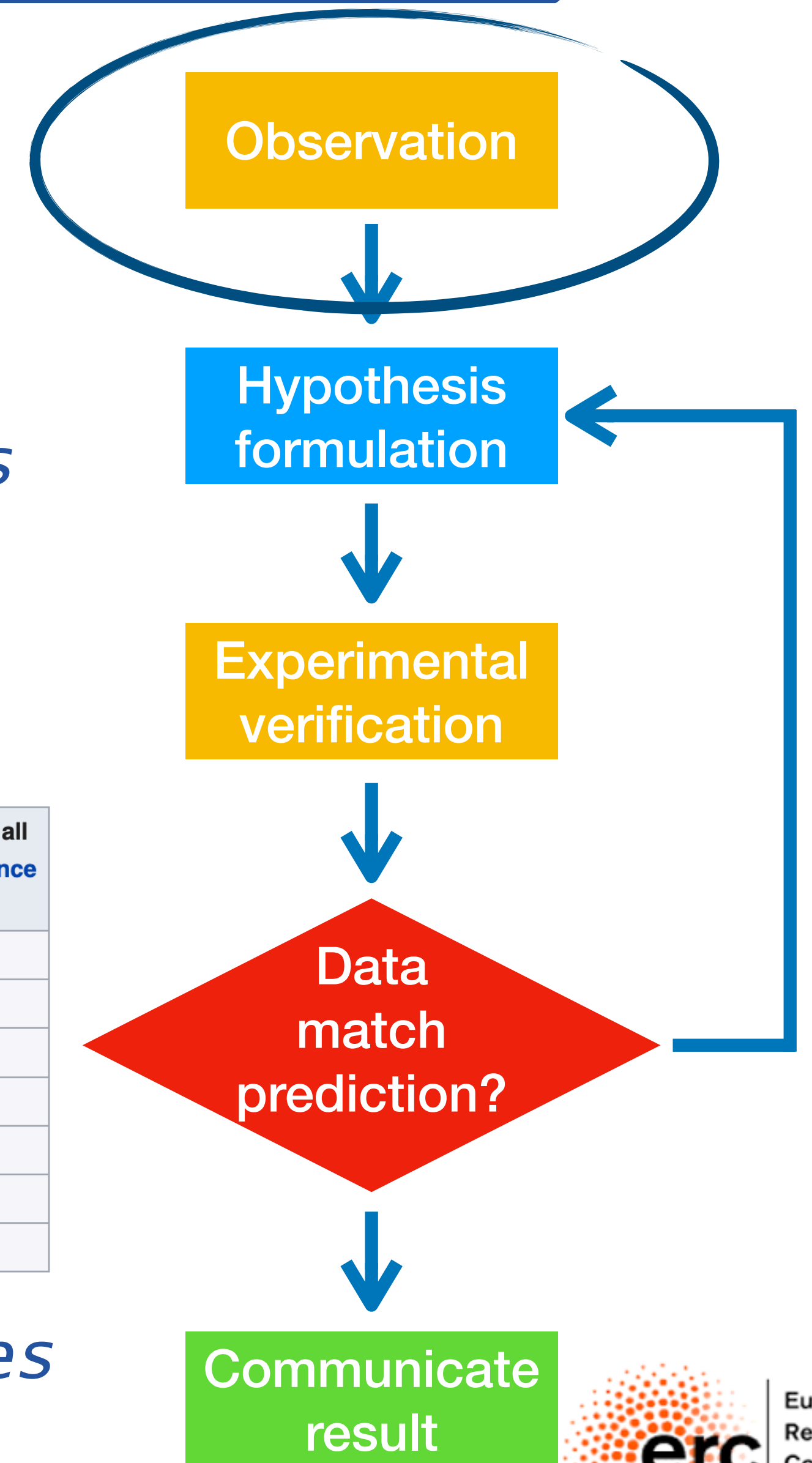
- ⊙ *The first step of the scientific method consists in observing nature*
- ⊙ *In the last 40 years, our starting point has been the SM (which was put together from experimental information collected in the 70s)*
- ⊙ *We are victim of our success:*
  - ⊙ *We use data mainly to confirm our hypothesis*
  - ⊙ *We have lost the value of learning from data*
  - ⊙ *Not by chance, we totally endorsed blind analysis as the ONLY way to search*





# Learning from Data

- ⊙ Rather than specifying a signal hypothesis upfront, we could start looking at our data first
- ⊙ Based on what we see (e.g., clustering alike objects) we could formulate a signal hypothesis
- ⊙ *EXAMPLE: star classification was formulated on empirical characteristics*

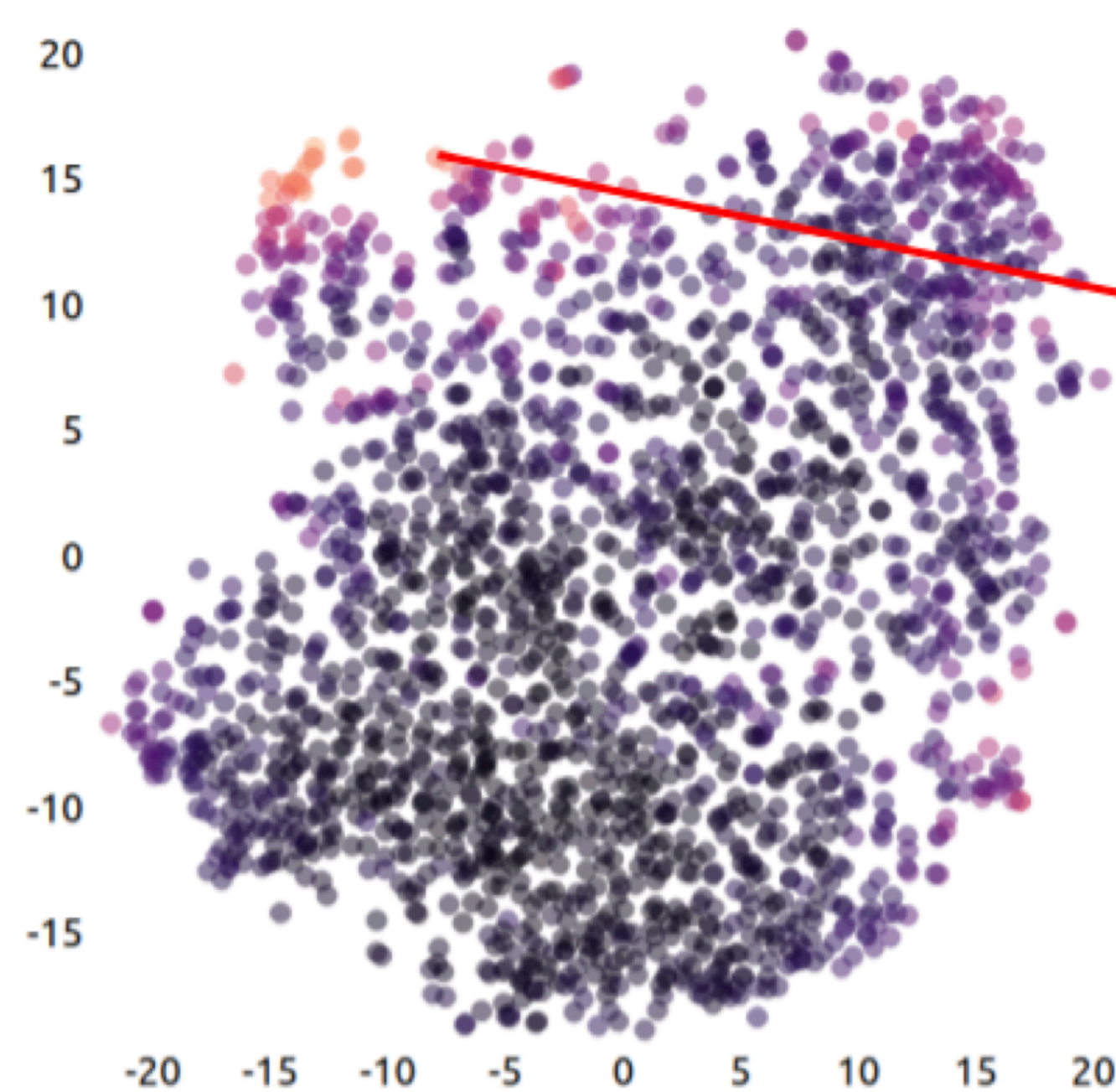


Class	Effective temperature <sup>[1][2]</sup>	Vega-relative chromaticity <sup>[3][4][a]</sup>	Chromaticity (D65) <sup>[5][6][3][b]</sup>	Main-sequence mass <sup>[1][7]</sup> (solar masses)	Main-sequence radius <sup>[1][7]</sup> (solar radii)	Main-sequence luminosity <sup>[1][7]</sup> (bolometric)	Hydrogen lines	Fraction of all main-sequence stars <sup>[8]</sup>
O	≥ 30,000 K	blue	blue	≥ 16 $M_{\odot}$	≥ 6.6 $R_{\odot}$	≥ 30,000 $L_{\odot}$	Weak	~0.00003%
B	10,000–30,000 K	blue white	deep blue white	2.1–16 $M_{\odot}$	1.8–6.6 $R_{\odot}$	25–30,000 $L_{\odot}$	Medium	0.13%
A	7,500–10,000 K	white	blue white	1.4–2.1 $M_{\odot}$	1.4–1.8 $R_{\odot}$	5–25 $L_{\odot}$	Strong	0.6%
F	6,000–7,500 K	yellow white	white	1.04–1.4 $M_{\odot}$	1.15–1.4 $R_{\odot}$	1.5–5 $L_{\odot}$	Medium	3%
G	5,200–6,000 K	yellow	yellowish white	0.8–1.04 $M_{\odot}$	0.96–1.15 $R_{\odot}$	0.6–1.5 $L_{\odot}$	Weak	7.6%
K	3,700–5,200 K	light orange	pale yellow orange	0.45–0.8 $M_{\odot}$	0.7–0.96 $R_{\odot}$	0.08–0.6 $L_{\odot}$	Very weak	12.1%
M	2,400–3,700 K	orange red	light orange red	0.08–0.45 $M_{\odot}$	≤ 0.7 $R_{\odot}$	≤ 0.08 $L_{\odot}$	Very weak	76.45%

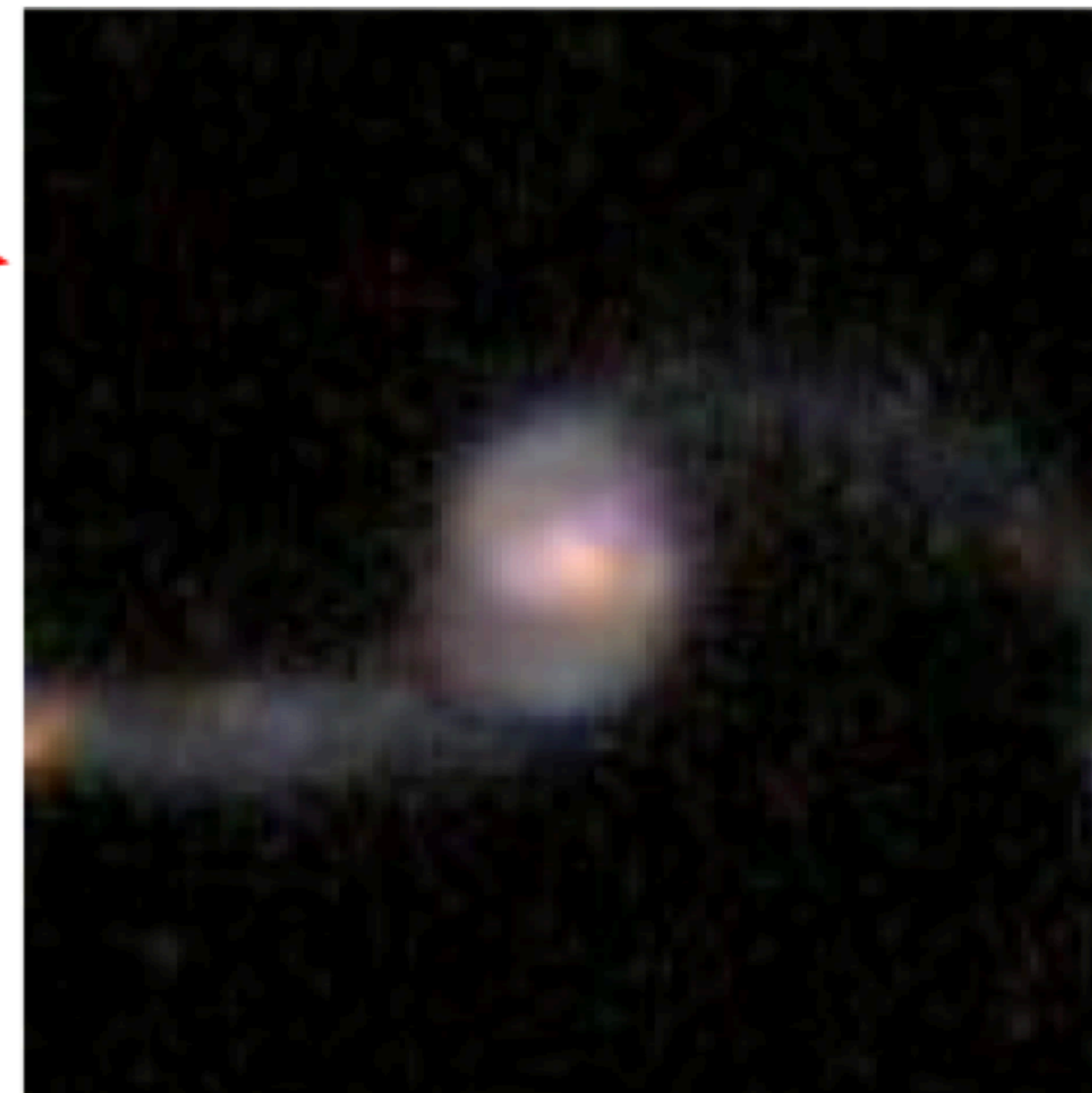
- ⊙ Afterwords, the connection to physics properties (temperature) was understood

# Learning from Anomalies

- ⦿ *Anomaly detection is one kind of data mining technique*
- ⦿ *One defines a metric of “typicality” to rank data samples*
- ⦿ *Based on this ranking, one can identify less typical events, tagging them as anomalies*
- ⦿ *By studying anomalies, one can make hypotheses on new physics mechanisms*



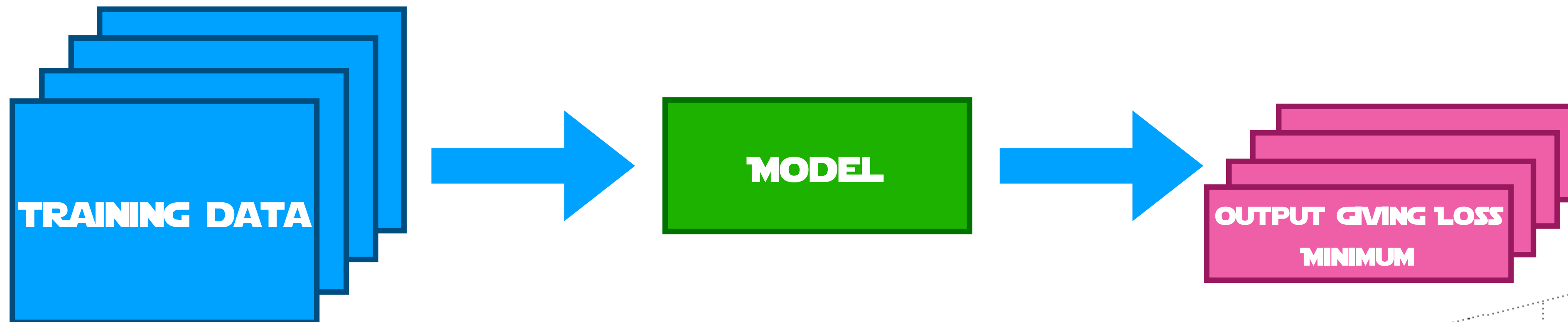
**Object ID: 960415**



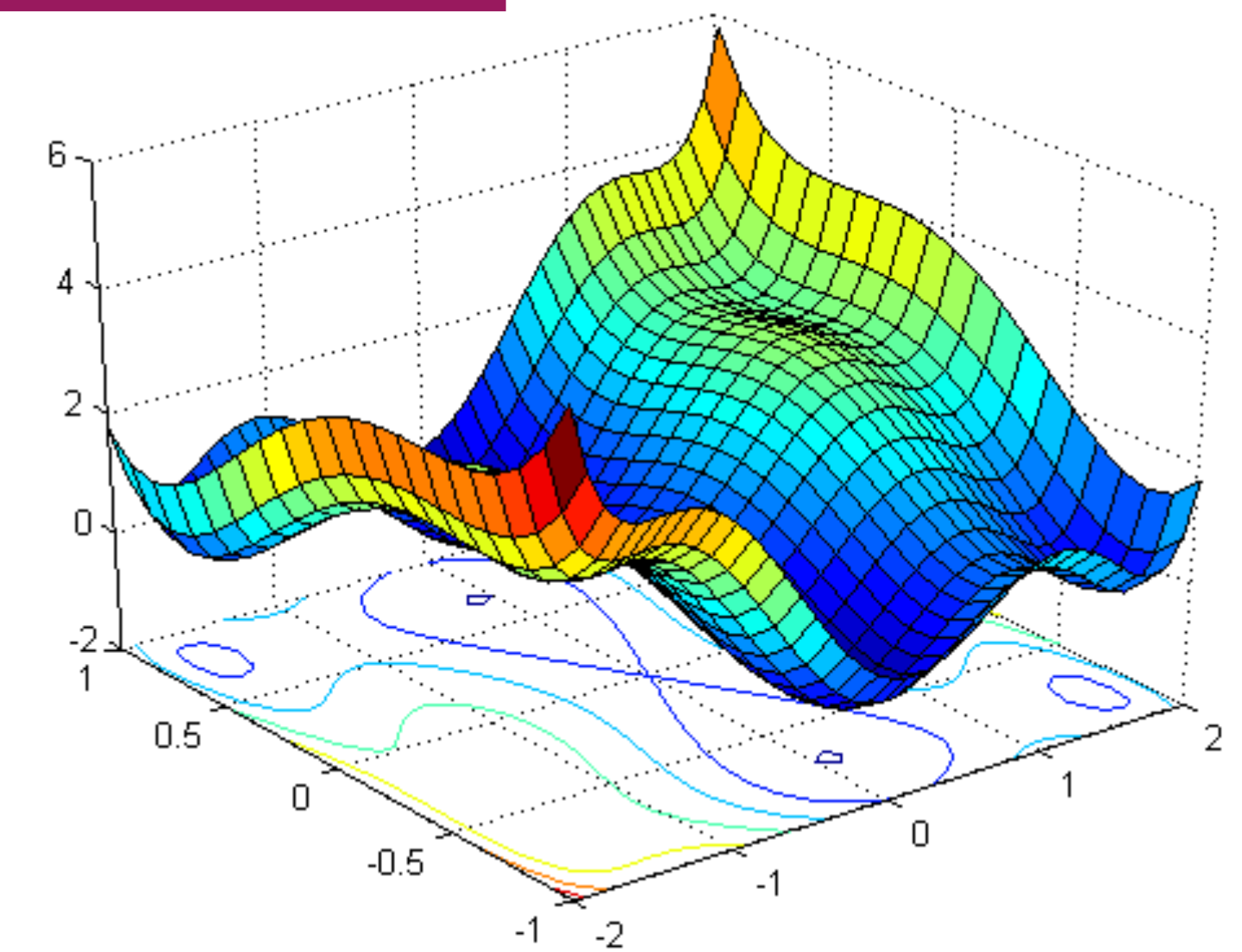
29 Anomaly Score: 4.470837



# Unsupervised learning



- ⦿ *A training dataset  $x$*
- ⦿ *No target  $y$*
- ⦿ *A model providing an output  $y$  at the minimum of the loss*
- ⦿ *A loss function of  $x$  and  $y$  specifying the task*
- ⦿ *e.g., clustering: group similar objects together*



# Autoencoders

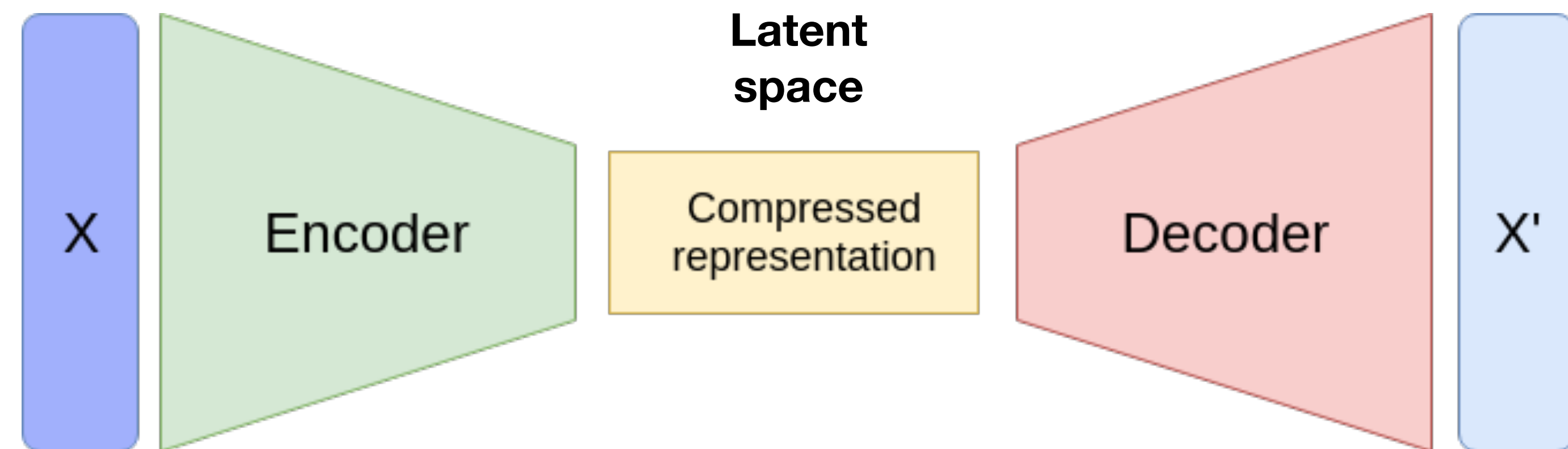
- Autoencoders are networks with a typical “bottleneck” structure, with a symmetric structure around it

- They go from  $\mathbb{R}^n \rightarrow \mathbb{R}^n$

- They are used to learn the identity function as  $f^{-1}(f(x))$

where  $f: \mathbb{R}^n \rightarrow \mathbb{R}^k$  and  $f^{-1}: \mathbb{R}^k \rightarrow \mathbb{R}^n$

- Autoencoders are essential tools for unsupervised studies



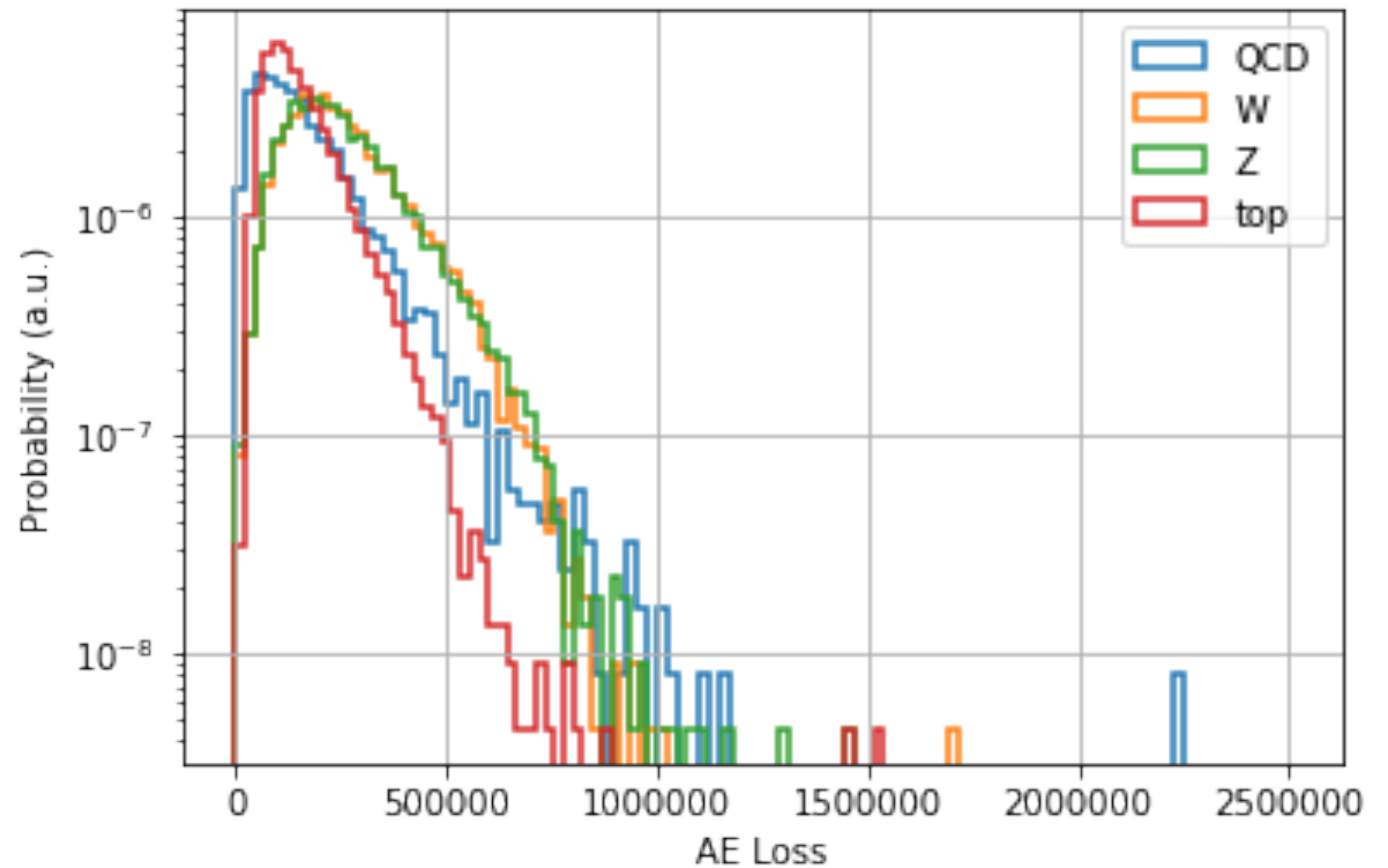


# Anomaly detection

- An autoencoder can reproduce new inputs of the same kind of the training dataset

  - The distance between the input and the output will be small
- If presented an event of some new kind (anomaly), the encoding-decoding will tend to fail

  - In this circumstance, the loss (=distance between input and output) will be bigger

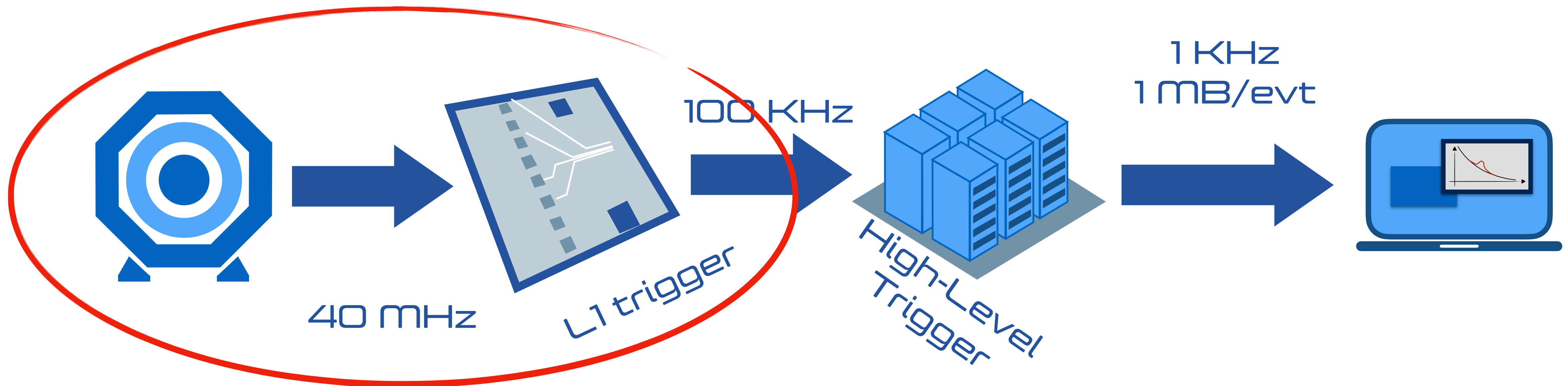


T. Heindel et al., <https://arxiv.org/abs/1808.08979>  
M. Farina et al., <https://arxiv.org/abs/1808.08992>



# Offline might be too late

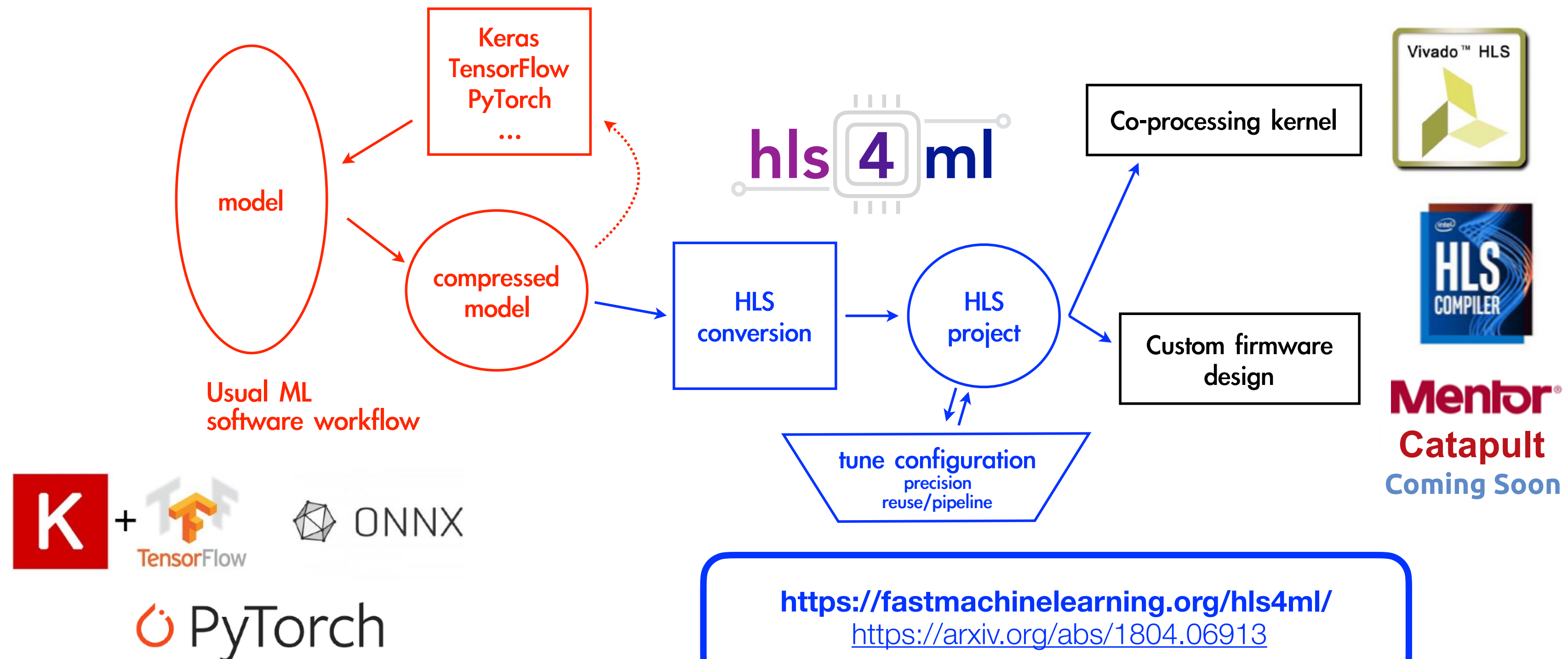
- ⦿ *With 40M beam crossings/seconds and 1000 stored, we might just be writing the wrong events*
- ⦿ *If we want to take action on a “plan B” path, this has to happen at the trigger*





# HLS4ML: NN to FPGAs

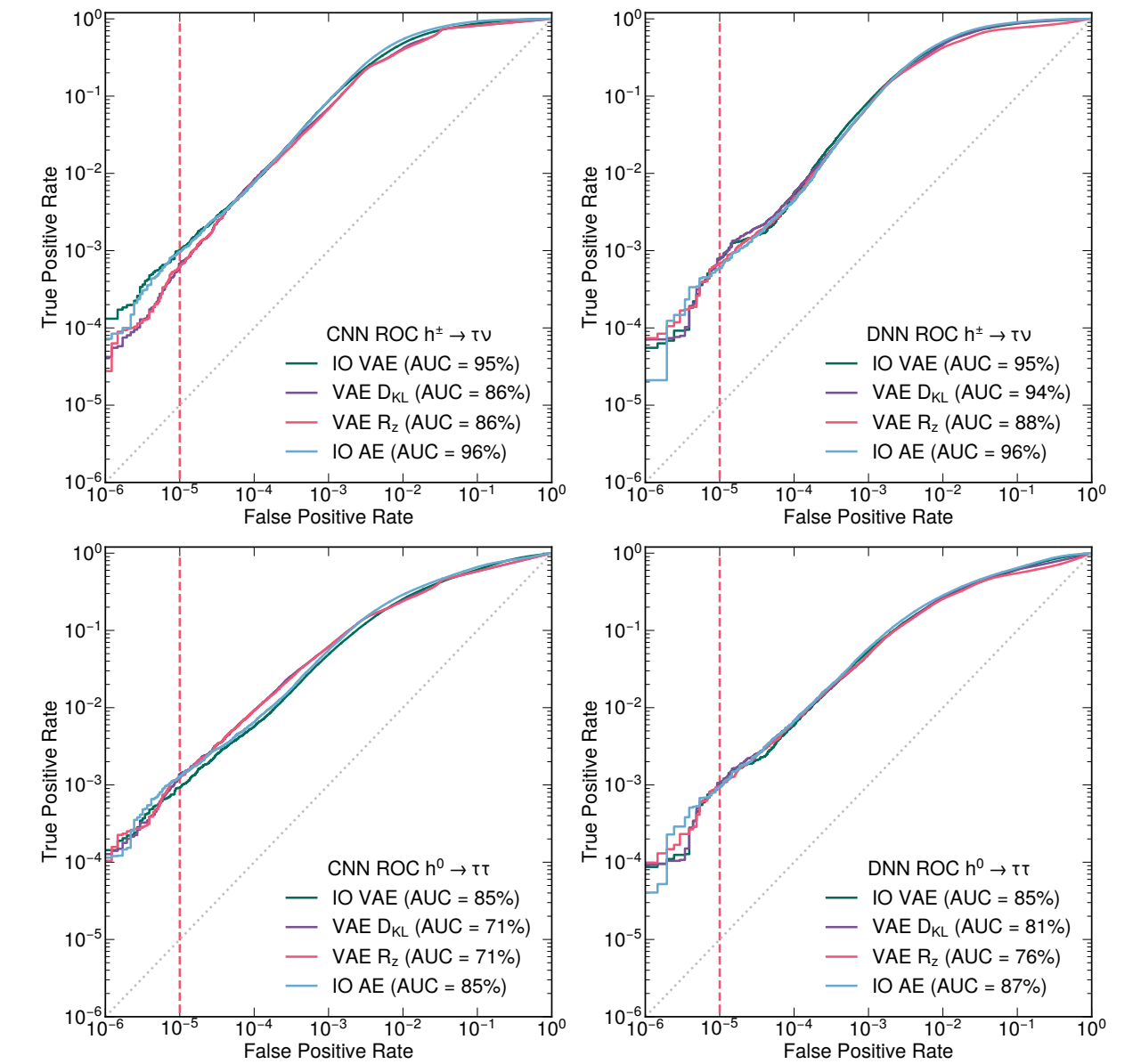
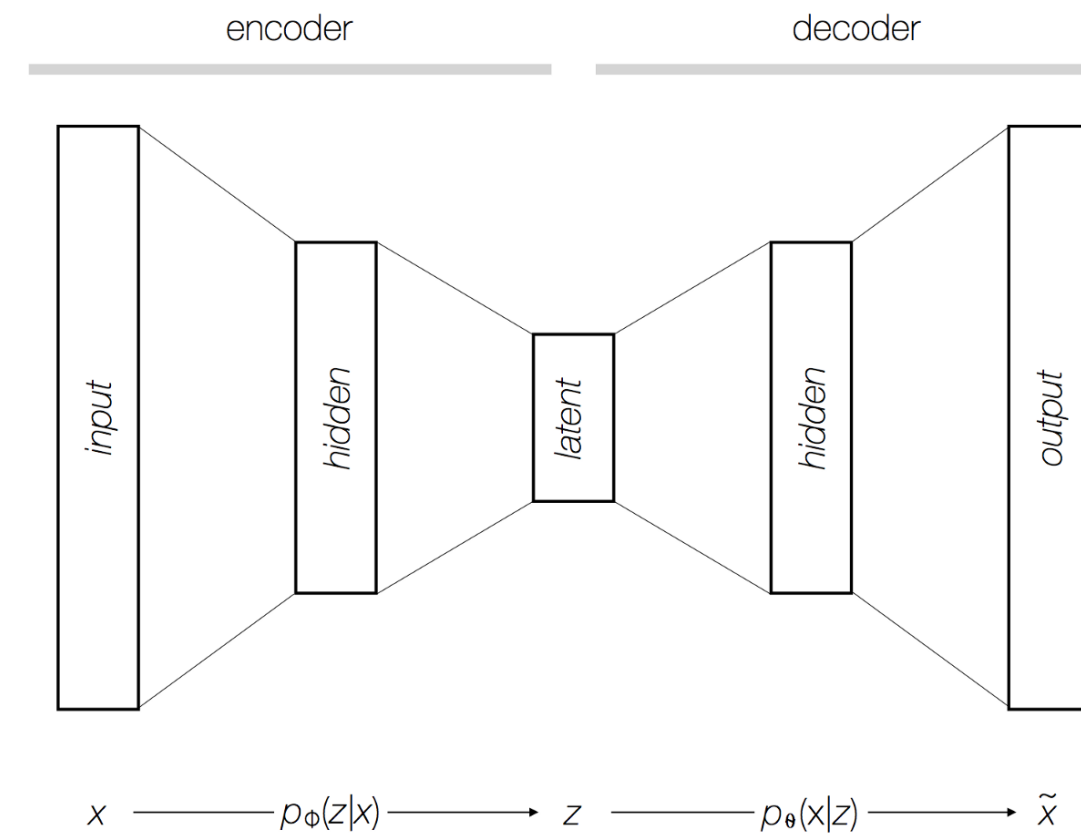
- *HLS4ML aims to be this automatic tool*
  - *reads as input models trained on standard DeepLearning libraries*
  - *comes with implementation of common ingredients (layers, activation functions, etc)*
  - *Uses HLS softwares to provide a firmware implementation of a given network*
  - *Could also be used to create co-processing kernels for HLT environments*





# Anomaly Detection on FPGA

- Autoencoders can provide event discrimination for anomaly detection, even with the little information available at L1

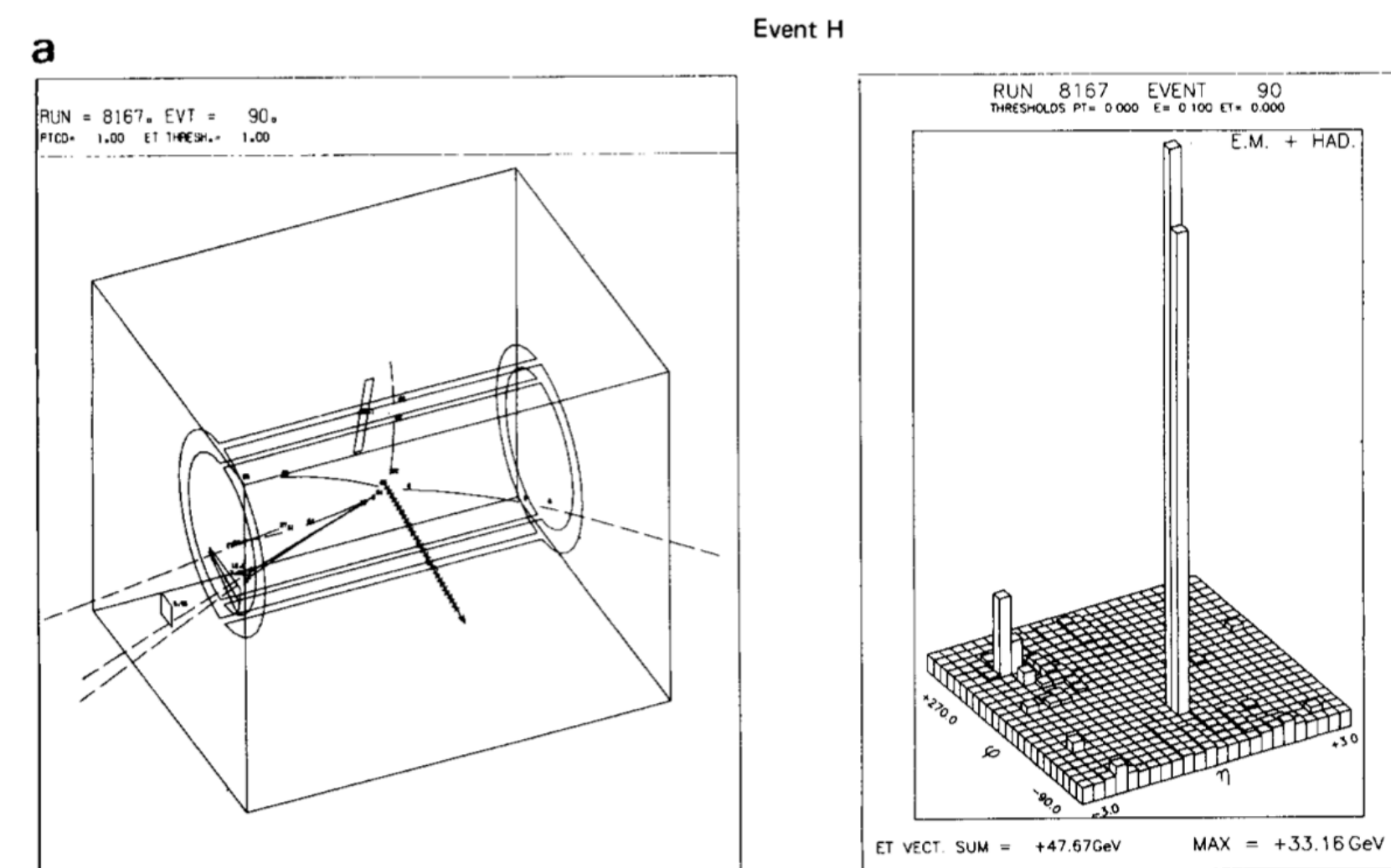


Model	DSP [%]	LUT [%]	FF [%]	BRAM [%]	Latency [ns]	II [ns]
DNN AE QAT 8 bits	2	5	1	0.5	130	5
CNN AE QAT 4 bits	8	47	5	6	1480	895
DNN VAE PTQ 8 bits	1	3	0.5	0.3	80	5
CNN VAE PTQ 8 bits	10	12	4	2	365	115



# Anomaly Dataset

- *The minimal final deliverable is a dataset of anomalous events*
- *Could be used to look for recurrent topologies*
- *By visual inspections, these “clusters” could inspire new searches on other (future) datasets*
- *Not very far away from how our field used to operate before big-data computing solutions took over*





# Much more than BSM program

- ◎ *LHC experiments will deliver much more than this*
  - ◎ *More precise  $W$  and top mass measurements*
  - ◎ *EFT operator analyses from differential cross section measurements in Higgs, top, and EW processes*
  - ◎ *Broader reach to Flavor physics*
  - ◎ *More exciting Heavy Ion physics*
- ◎ *Stay tuned...*



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# Backup Slides



# Network compression techniques

- Pruning: identify which part of the network is really relevant and remove the rest (makes network smaller)
- Quantization: use limited precision for numerical representations (save resources)
- Reuse: dilute the network inference on multiple clock cycles, trading off resource needs for latency

