

## The state of Event Generators for the HL-LHC



Olivier Mattelaer

# Steane's question

Do we need  $N(N)$ LO calculations?

Which event generators can be used?



**How much is the complexity of computations going to increase?**

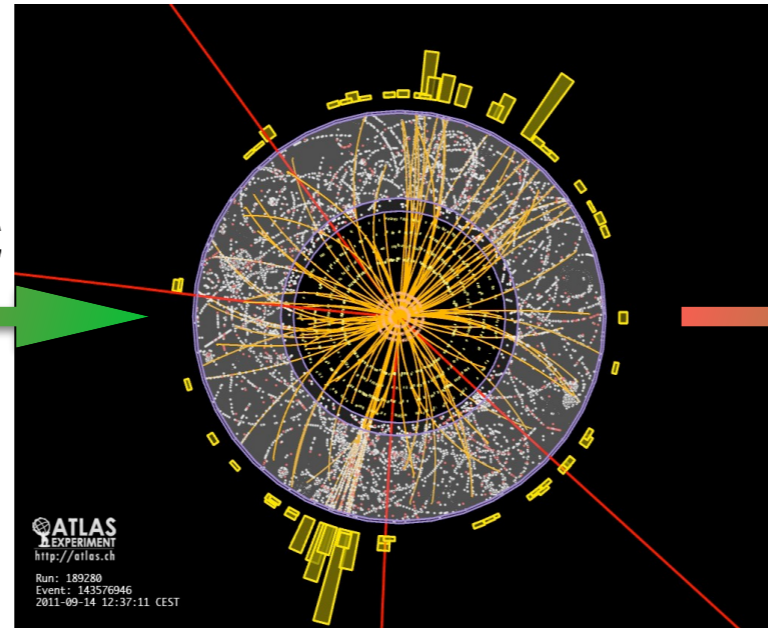
# What is Event Generator ?

The World



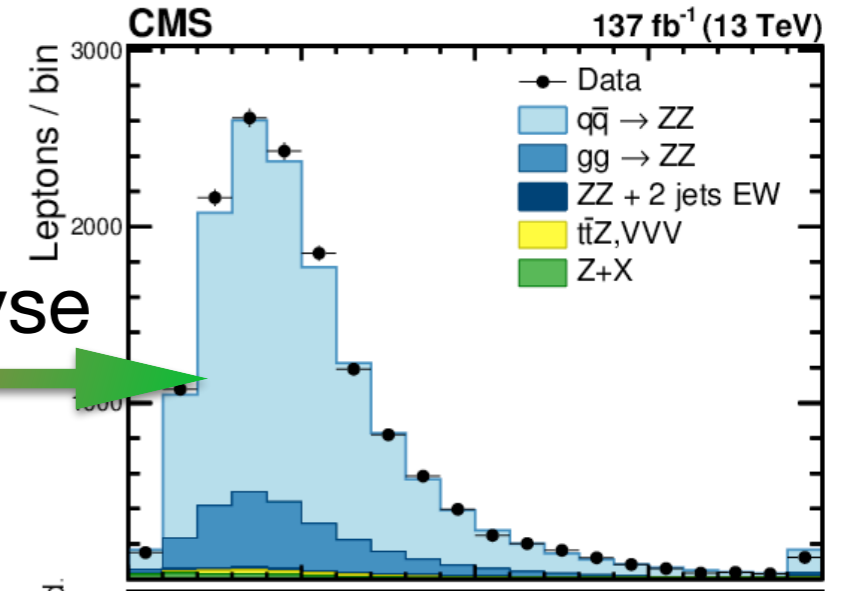
LHC

Experiment



Analyse

Measurement



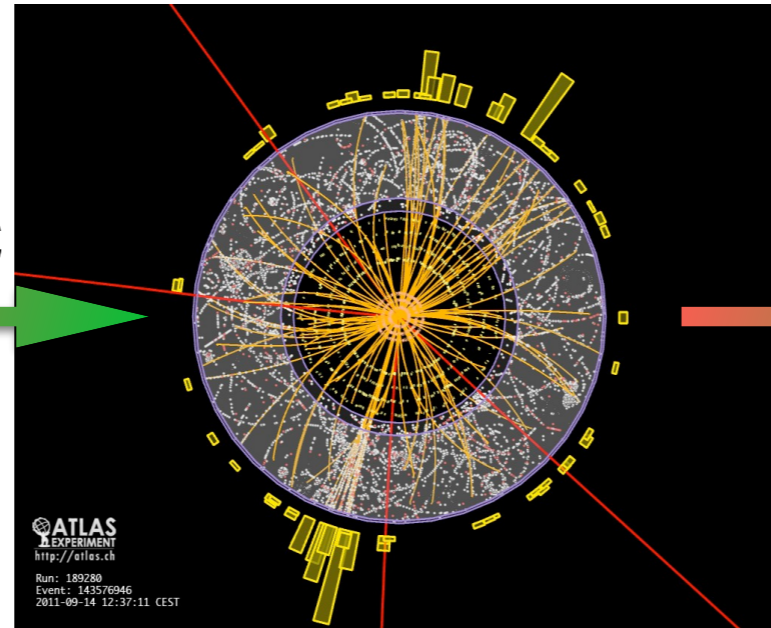
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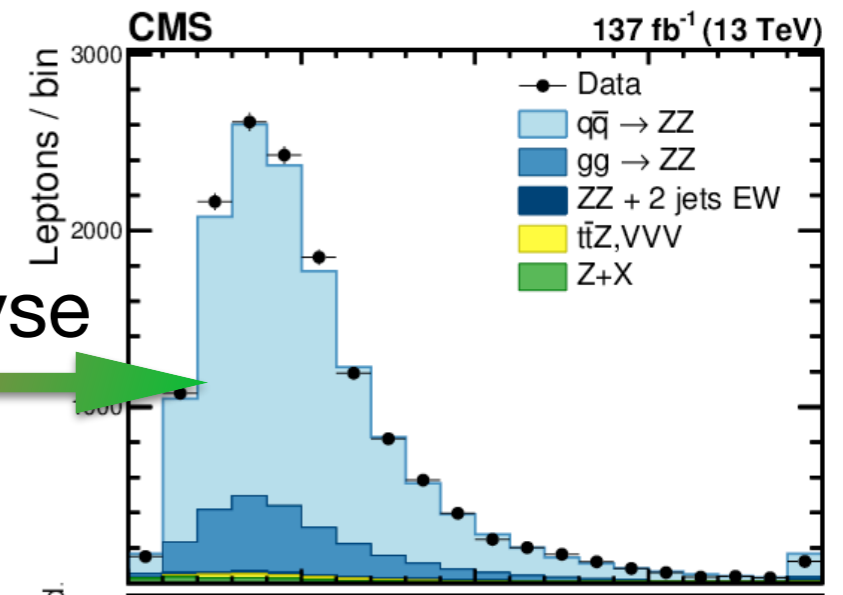
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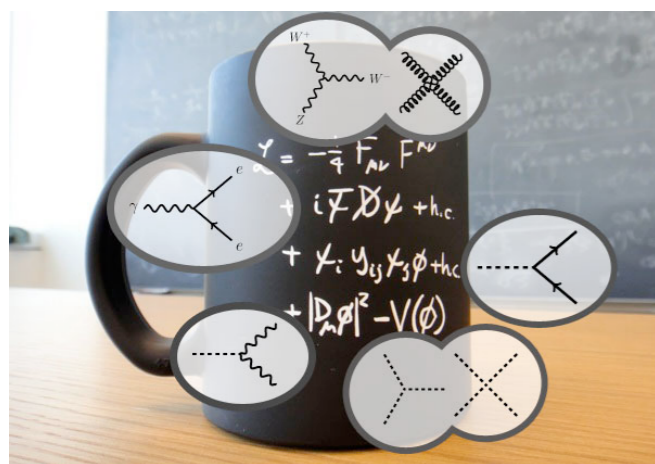


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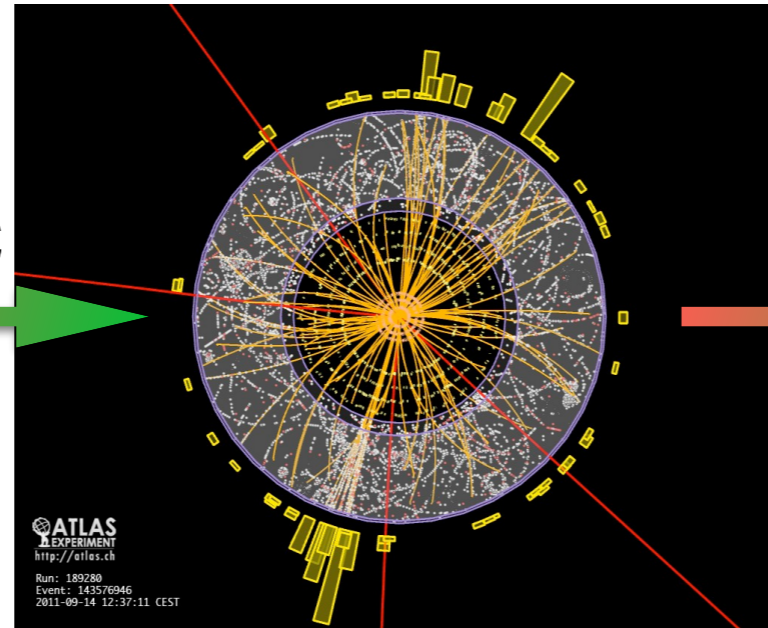
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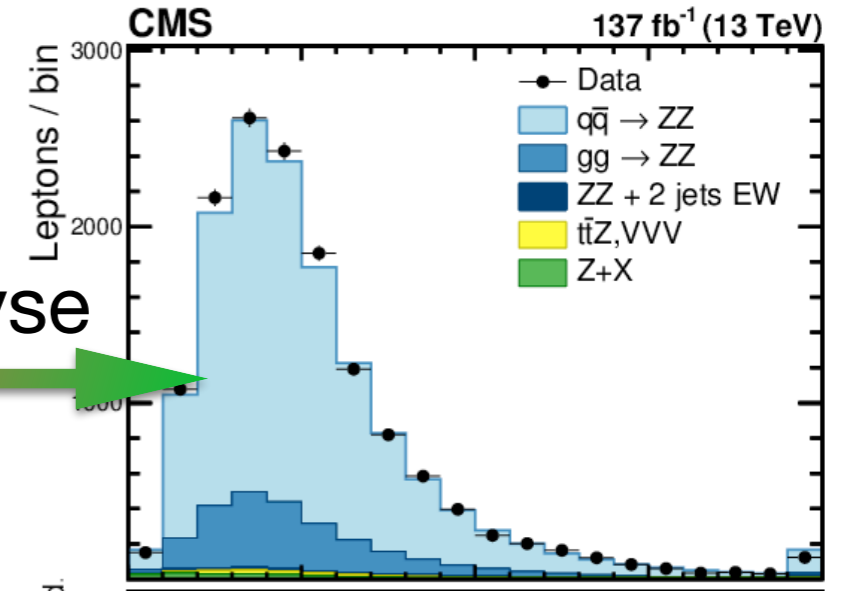
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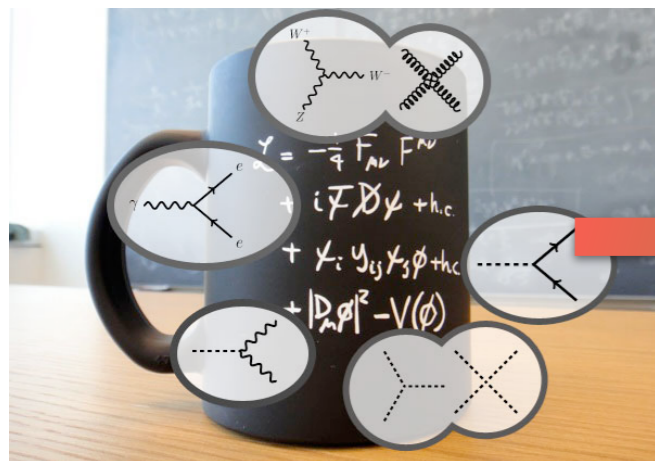


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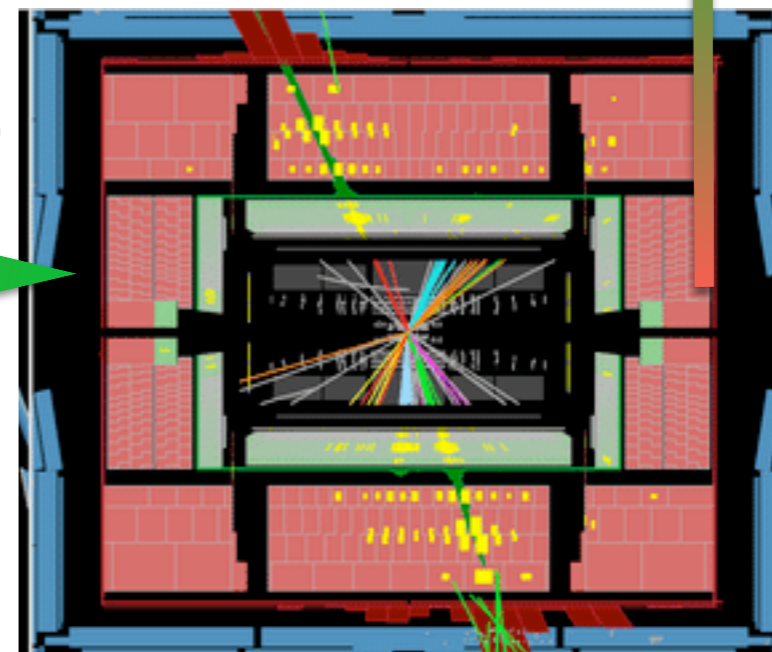
Measurement



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Analyse

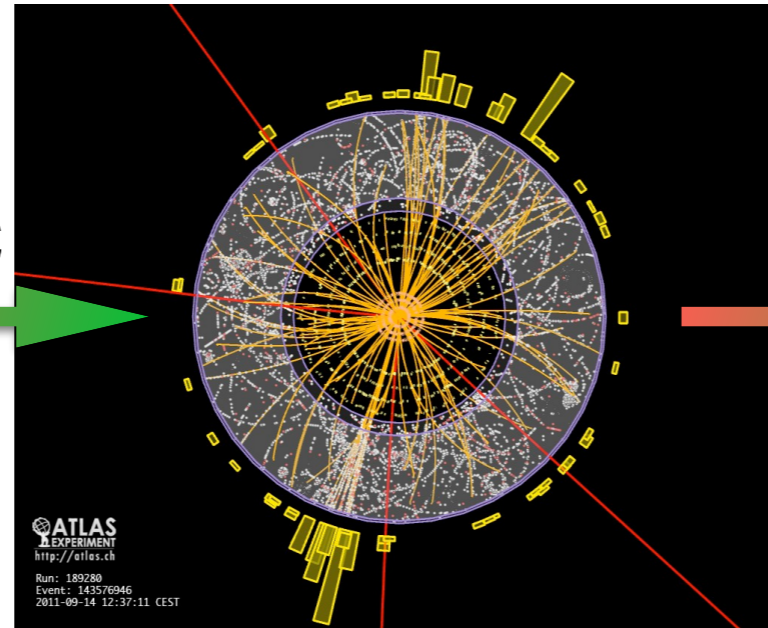
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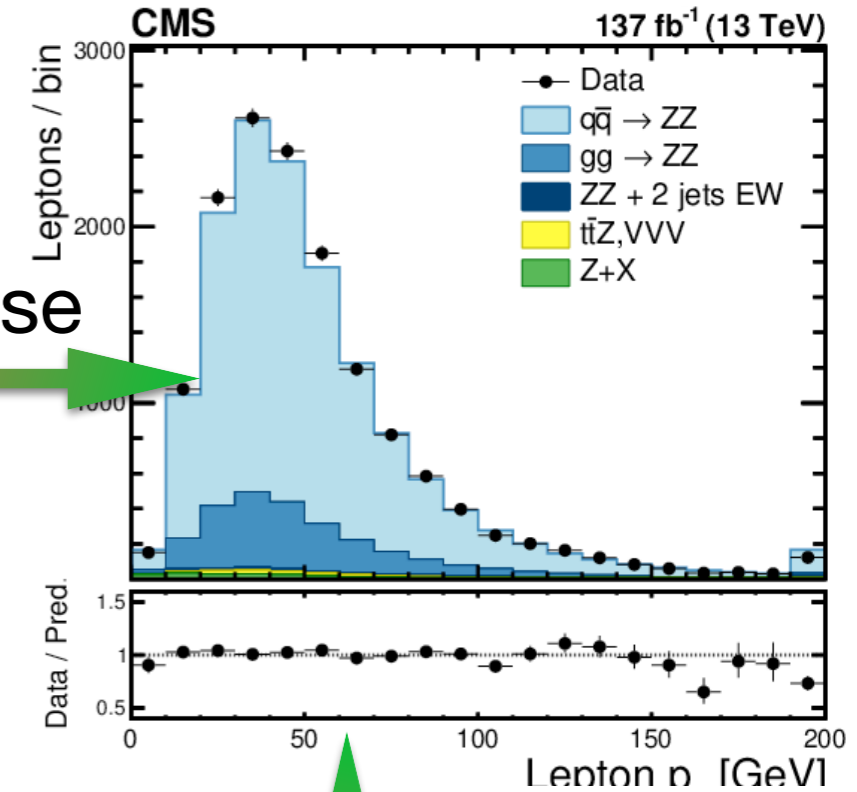
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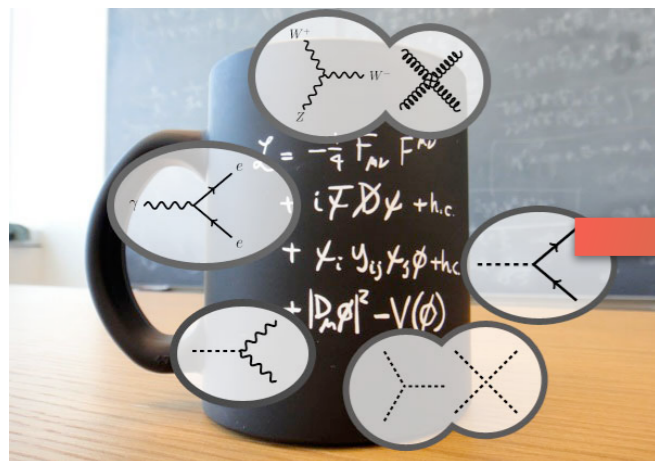
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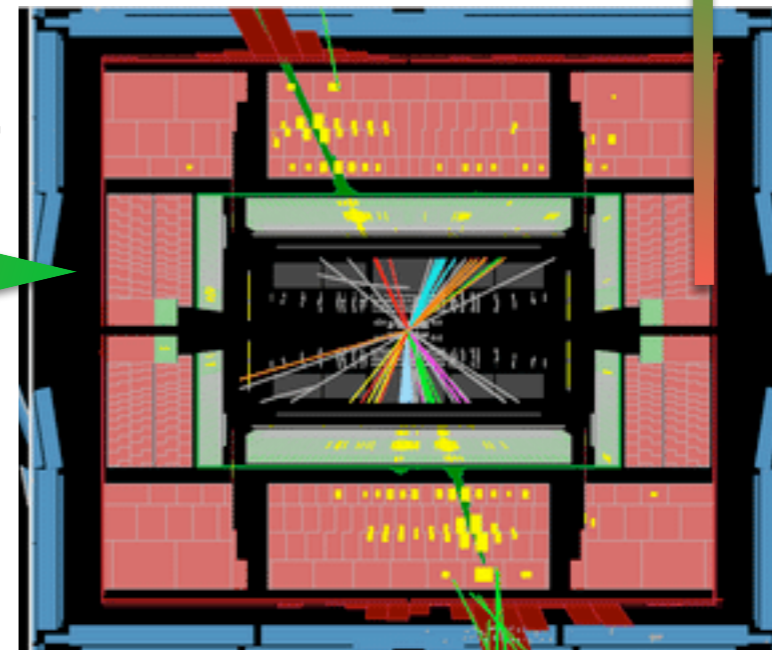


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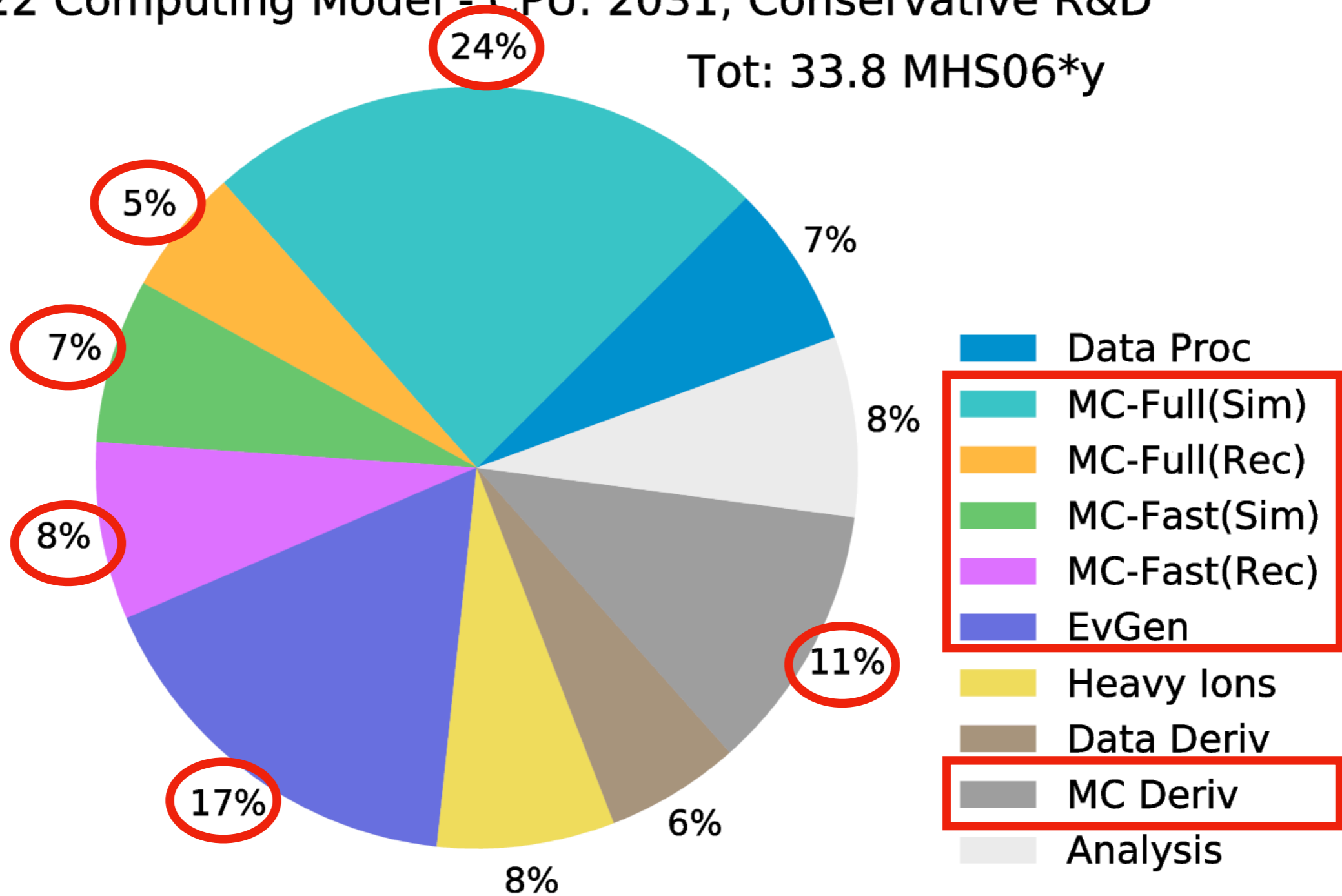


# Computing planning I

**ATLAS Preliminary**

2022 Computing Model - CPU: 2031, Conservative R&D

Tot: 33.8 MHS06\*y



The full comparison to "theory" takes ~75%

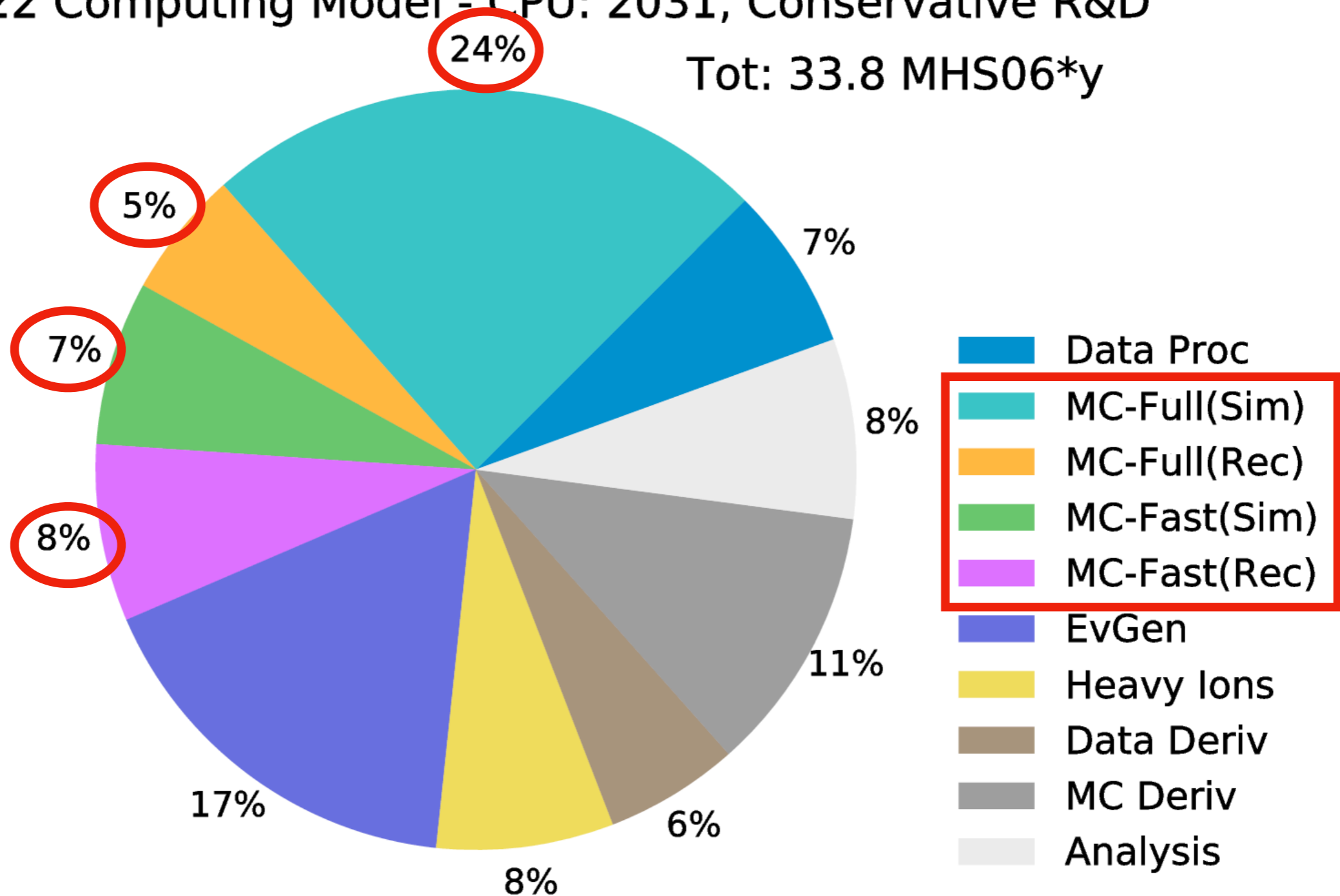


# Computing planning I

**ATLAS Preliminary**

2022 Computing Model - CPU: 2031, Conservative R&D

Tot: 33.8 MHS06\*y



Detector Simulation of such events takes ~45%

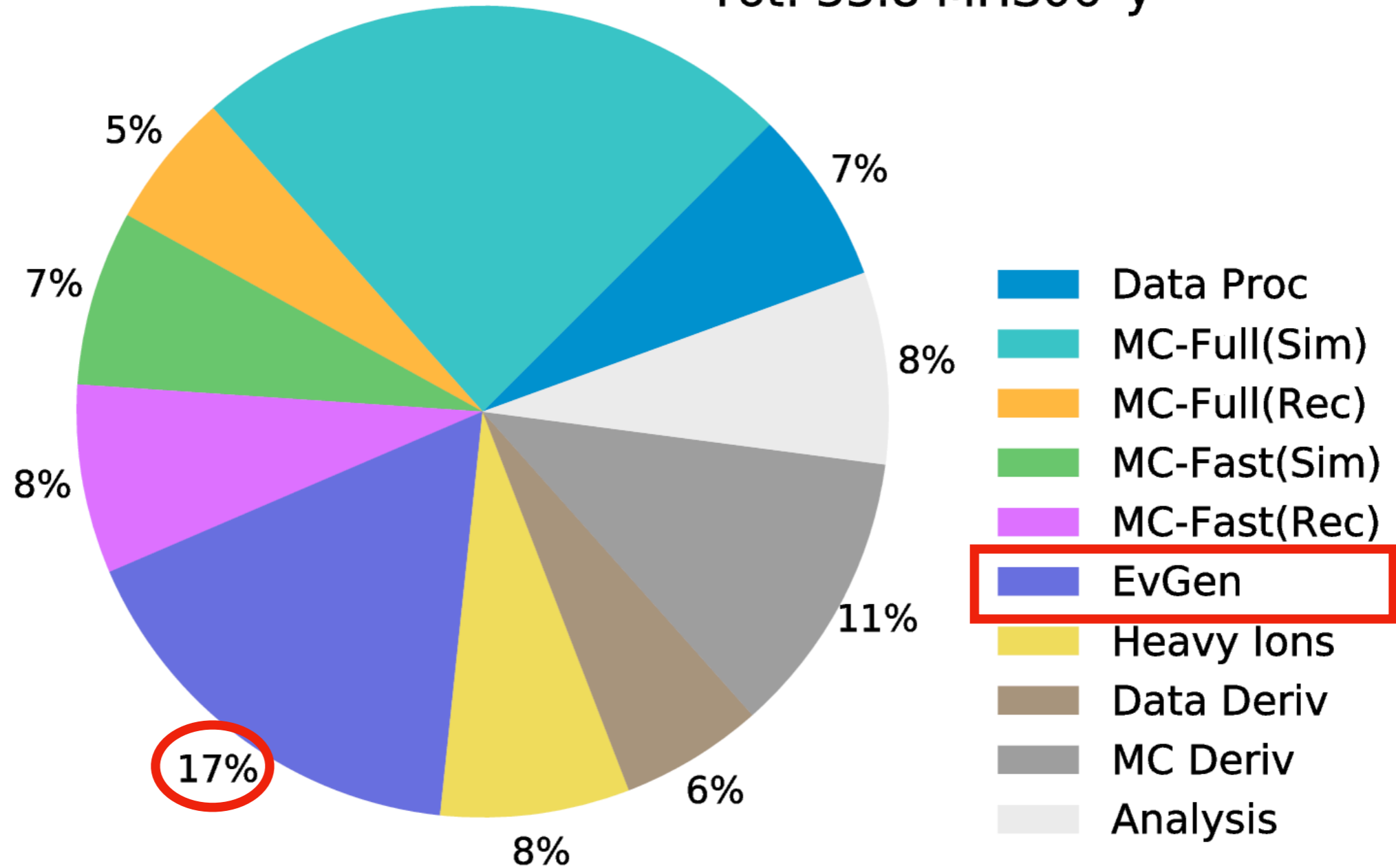
# Computing planning I

**ATLAS Preliminary**

2022 Computing Model - CPU: 2031, Conservative R&D

24%

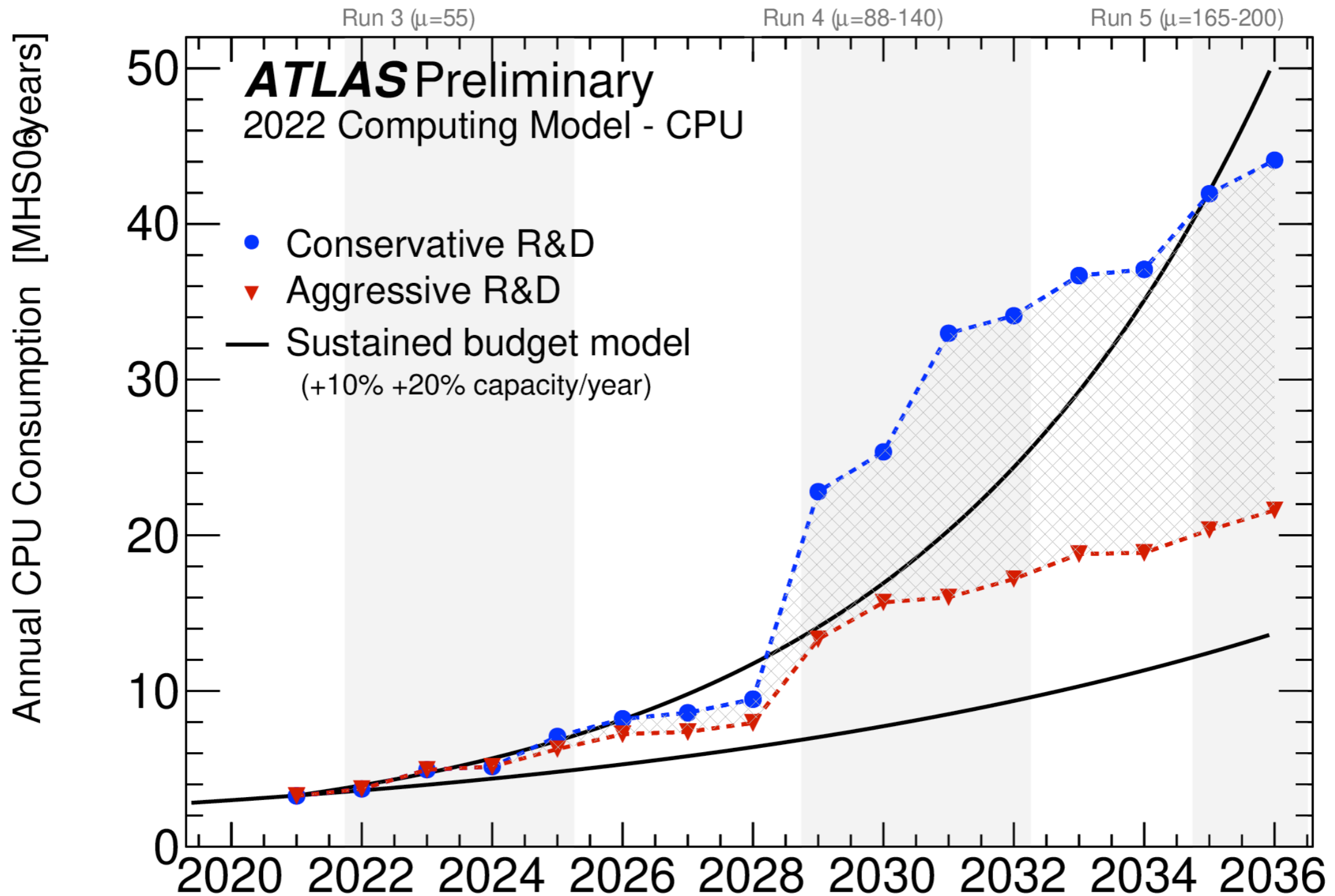
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Pure simulation of events at 17%

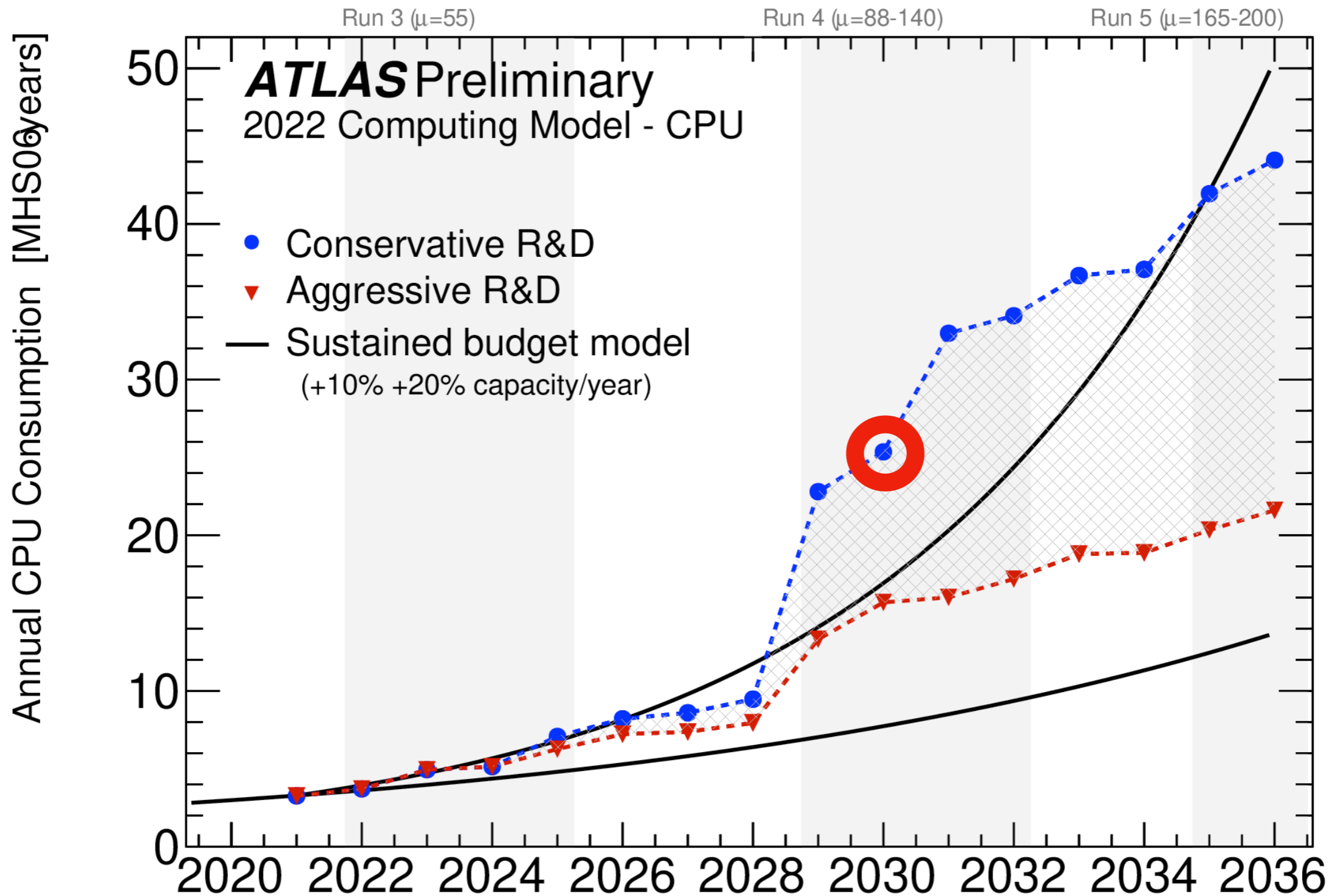


# Computing planning II



~ 2029 is the most critical period

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# Theory Need for HL



- We want **MORE, MORE and MORE**
  - **More** accurate (NLO/NNLO)
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- We need **cross-checking**
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Do we really need more?

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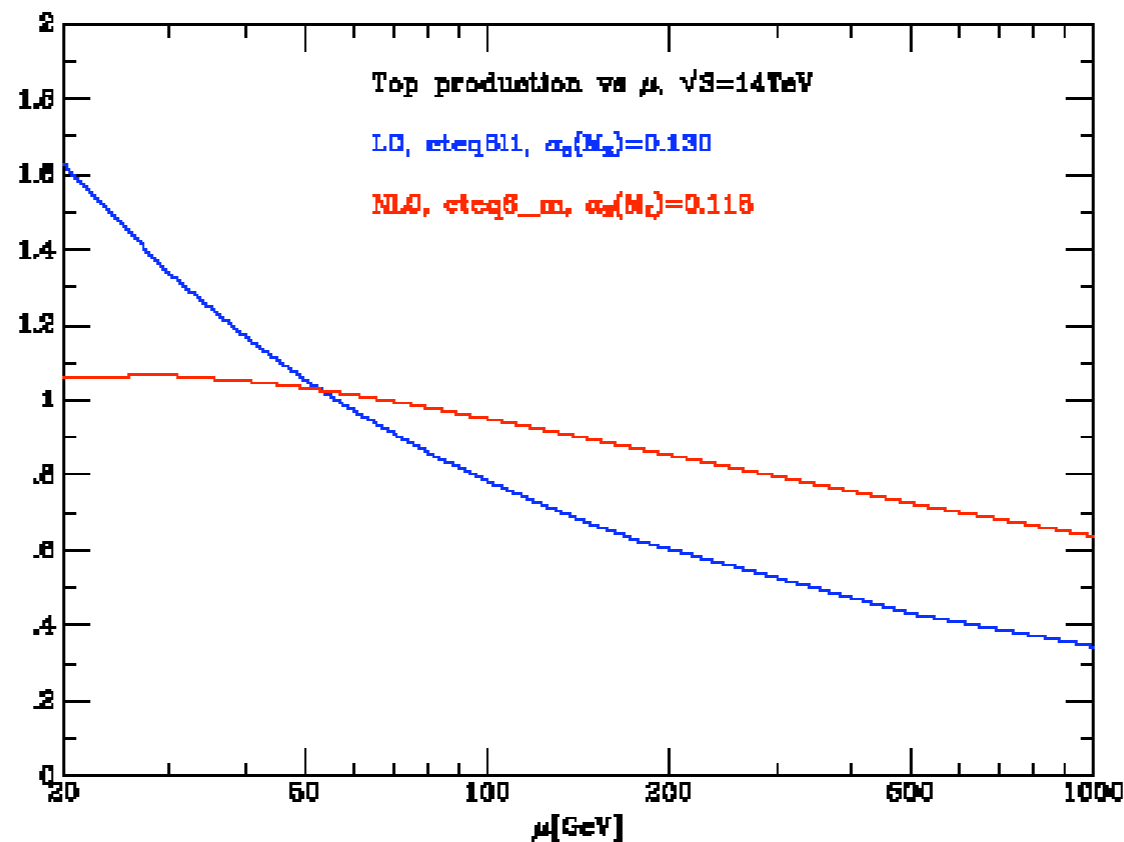
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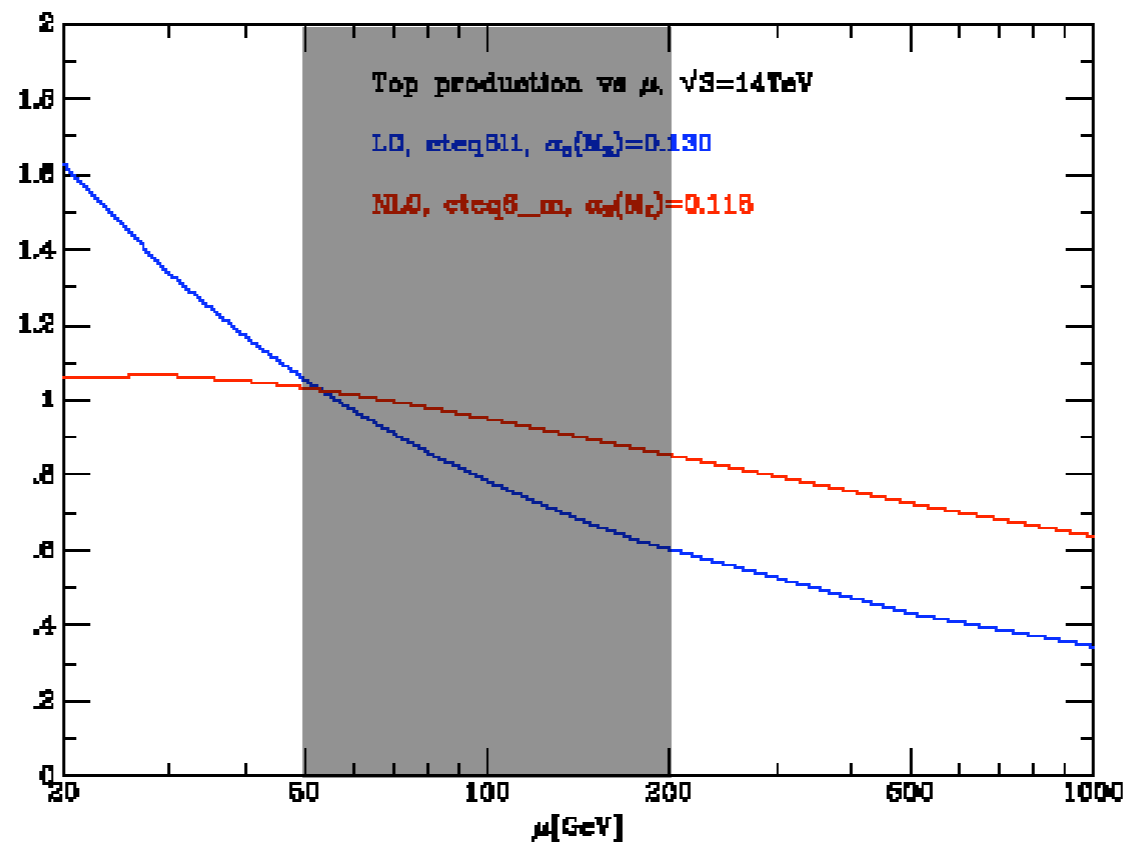
# Improved predictions

$$\hat{\sigma} = \sigma^{\text{Born}} \left( 1 + \frac{\alpha_s}{2\pi} \sigma^{(1)} + \left( \frac{\alpha_s}{2\pi} \right)^2 \sigma^{(2)} + \left( \frac{\alpha_s}{2\pi} \right)^3 \sigma^{(3)} + \dots \right)$$



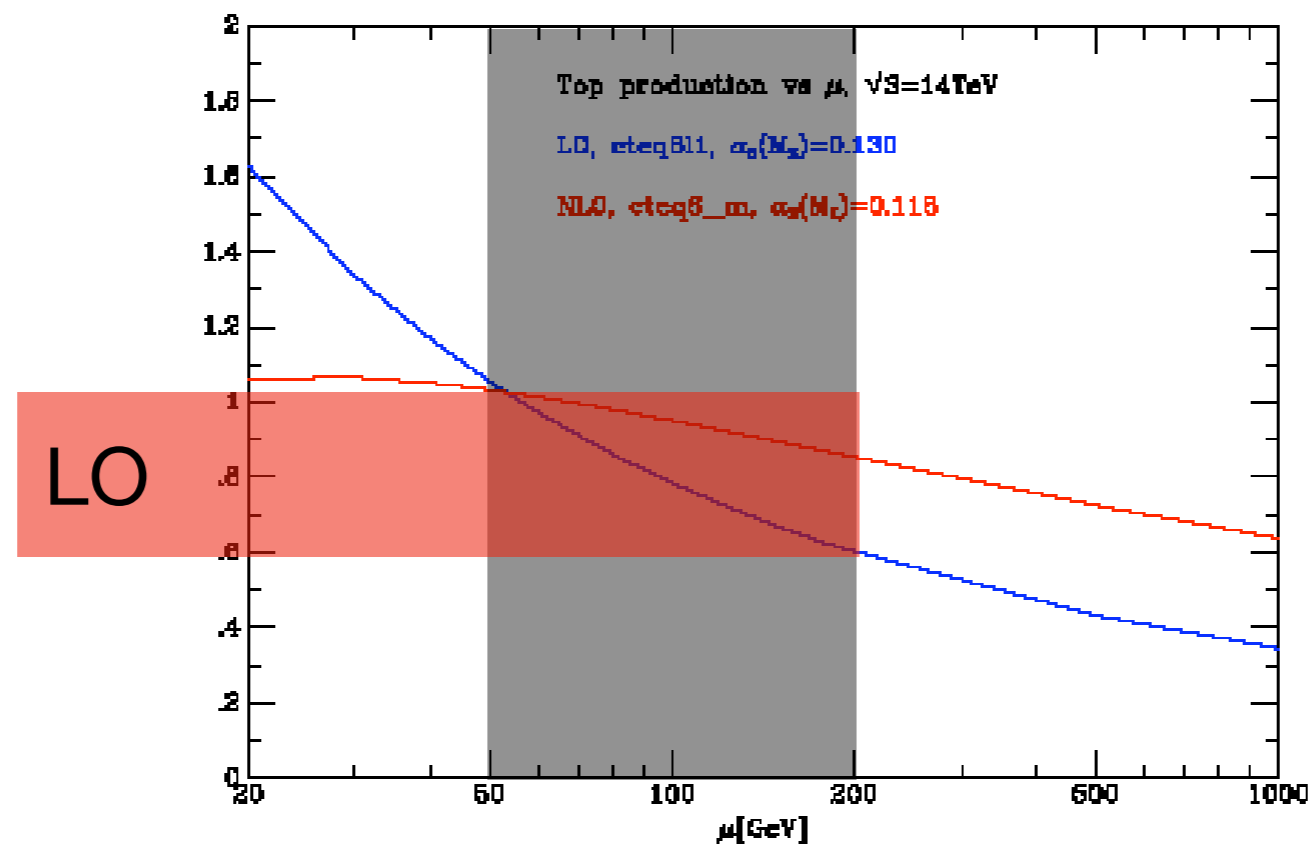
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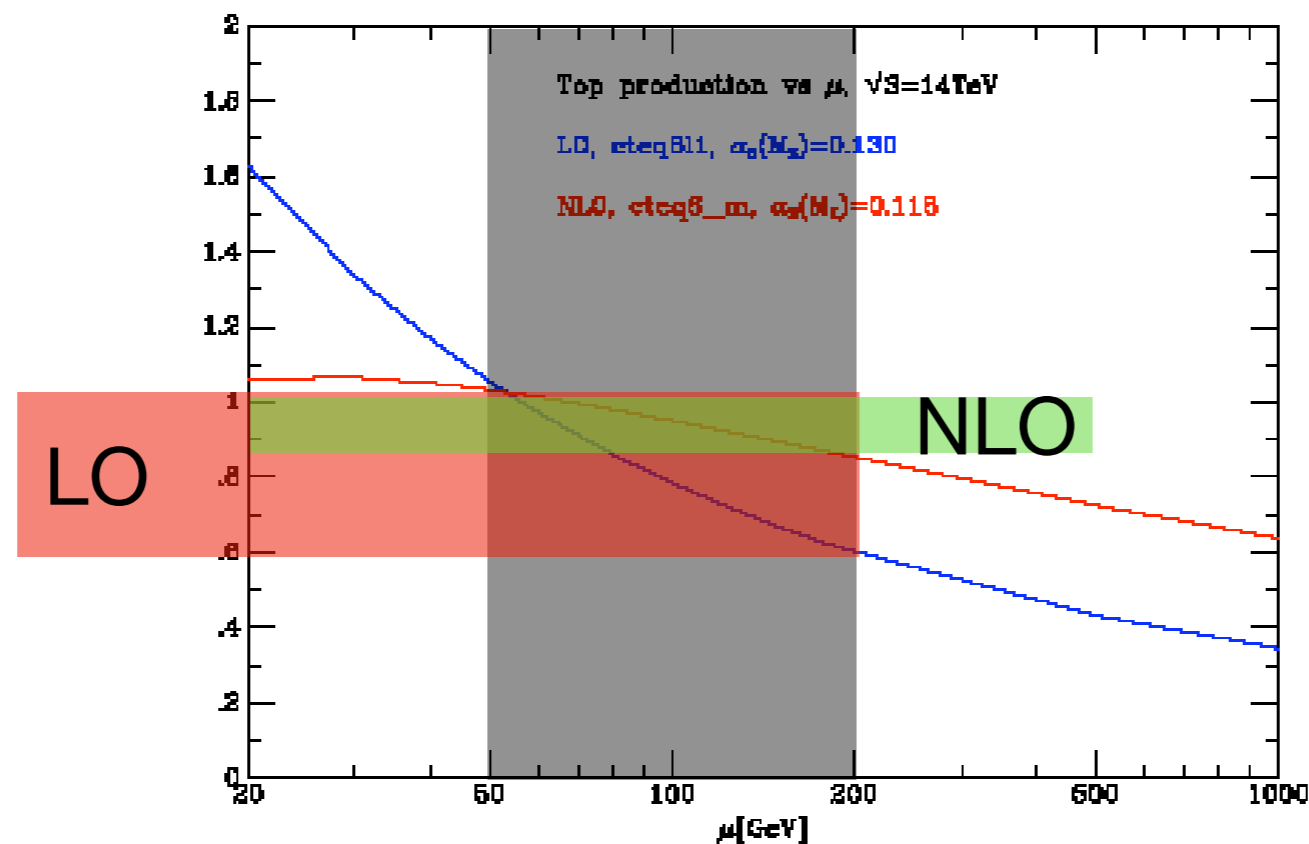
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  - Poor accuracy (error at  $\sim 40\%$ )

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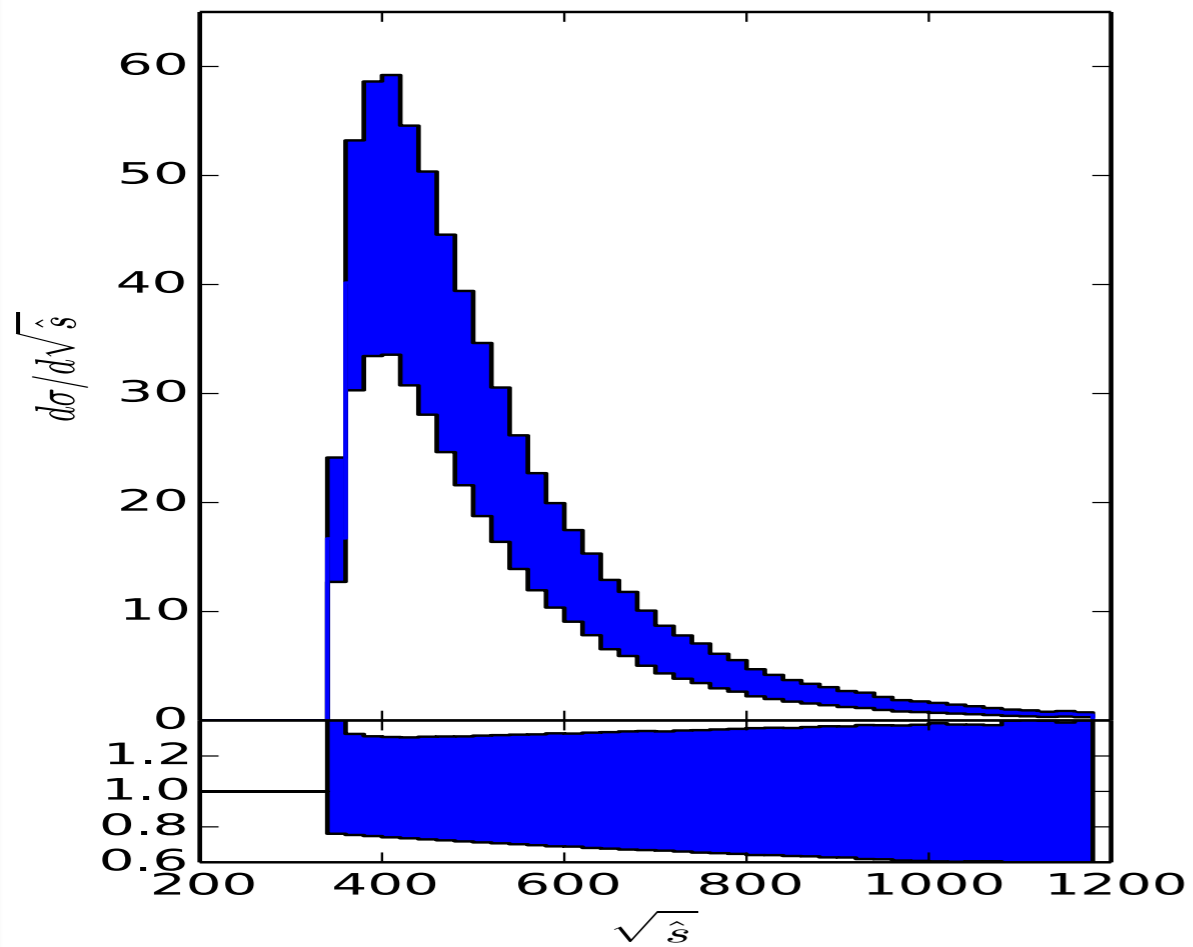


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# Is LO useful?

Error on LO prediction (top quark pair)

Error on Normalisation

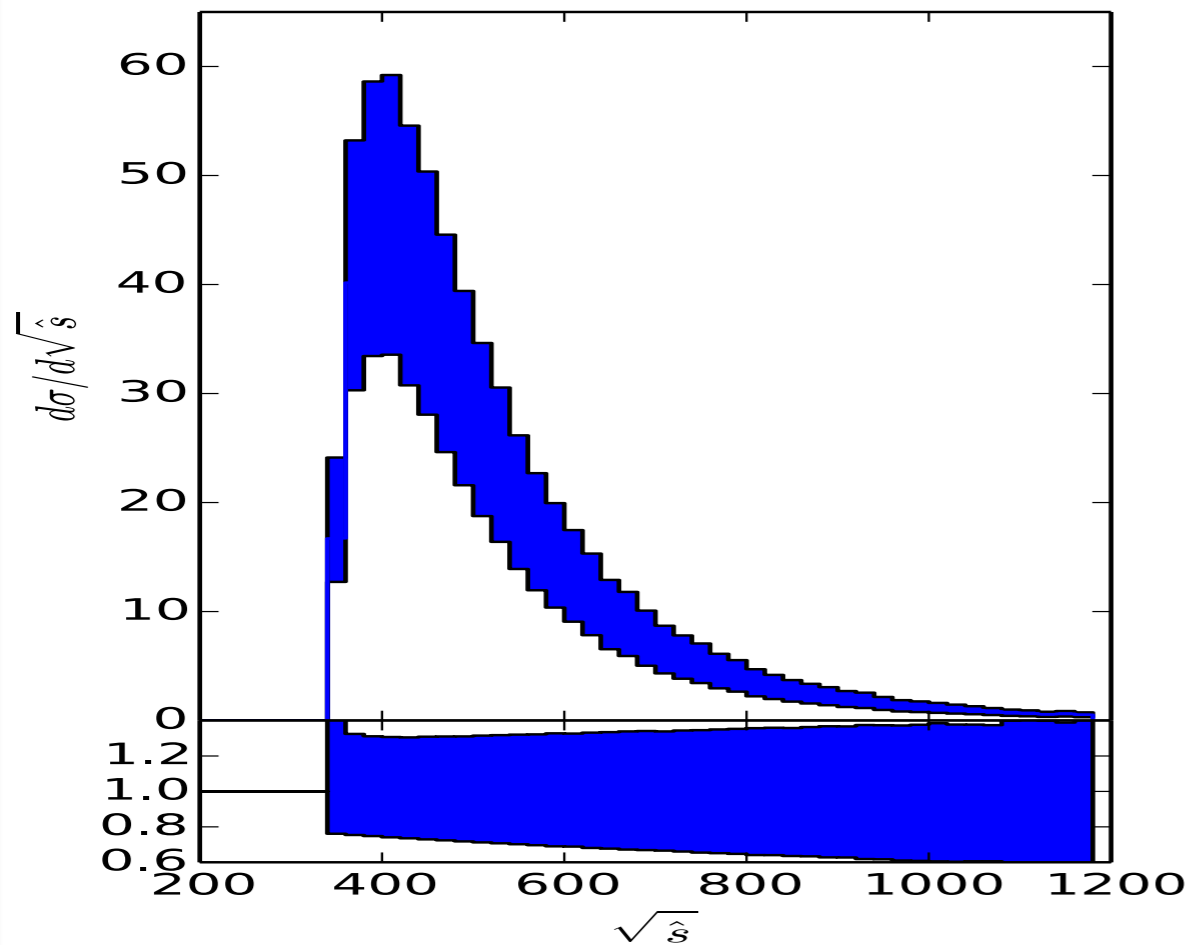




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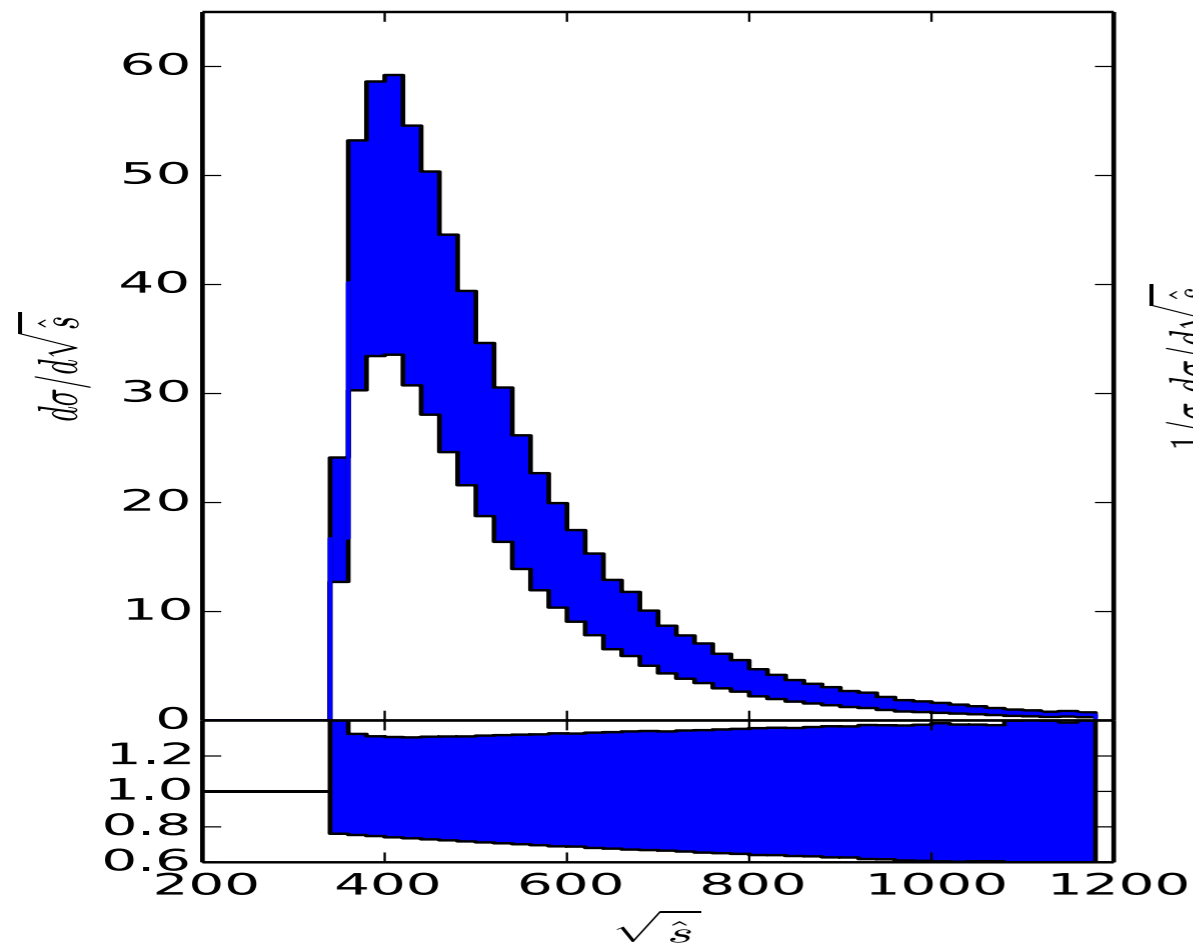


Too Large Error (30-50%)

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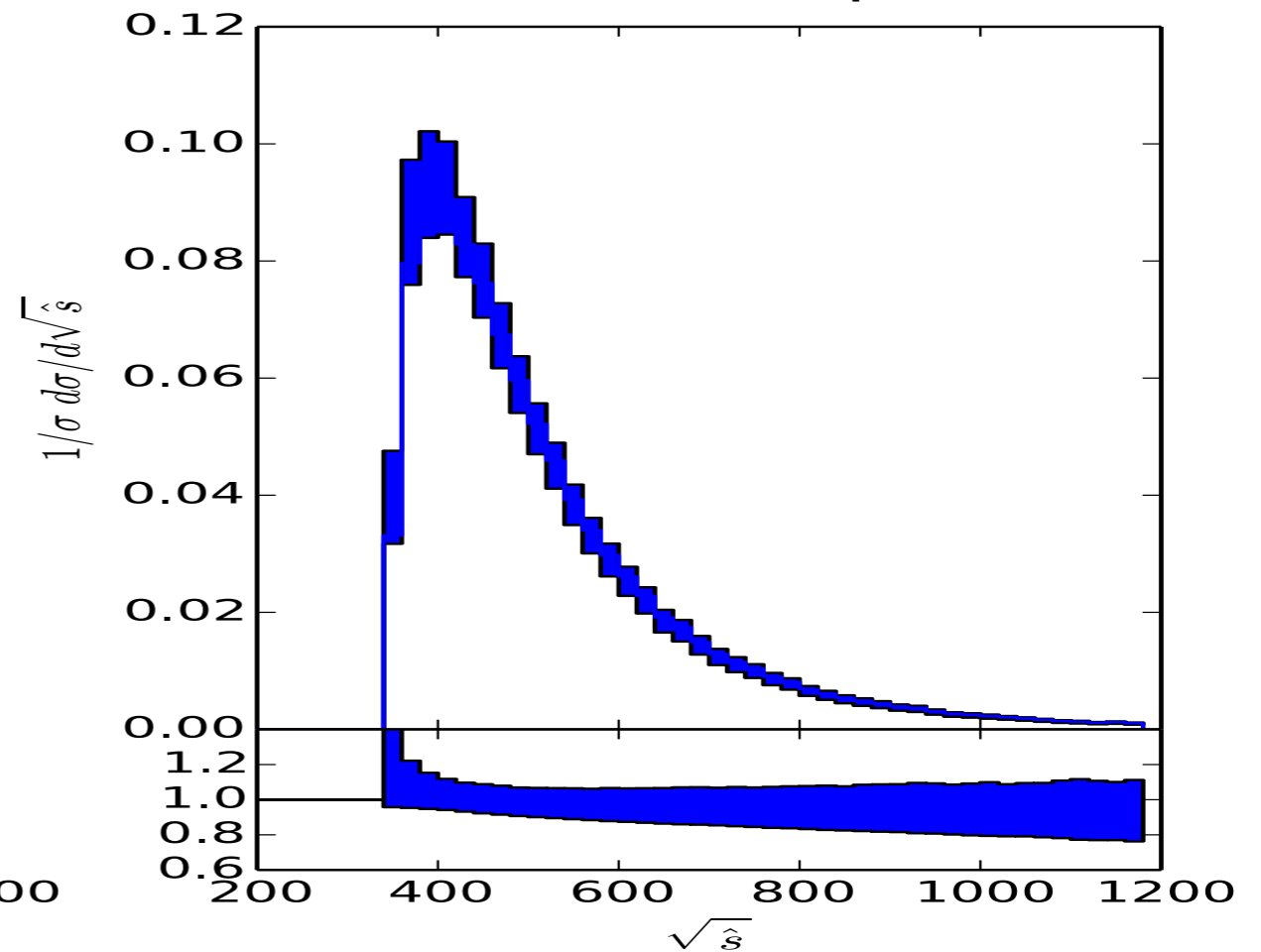
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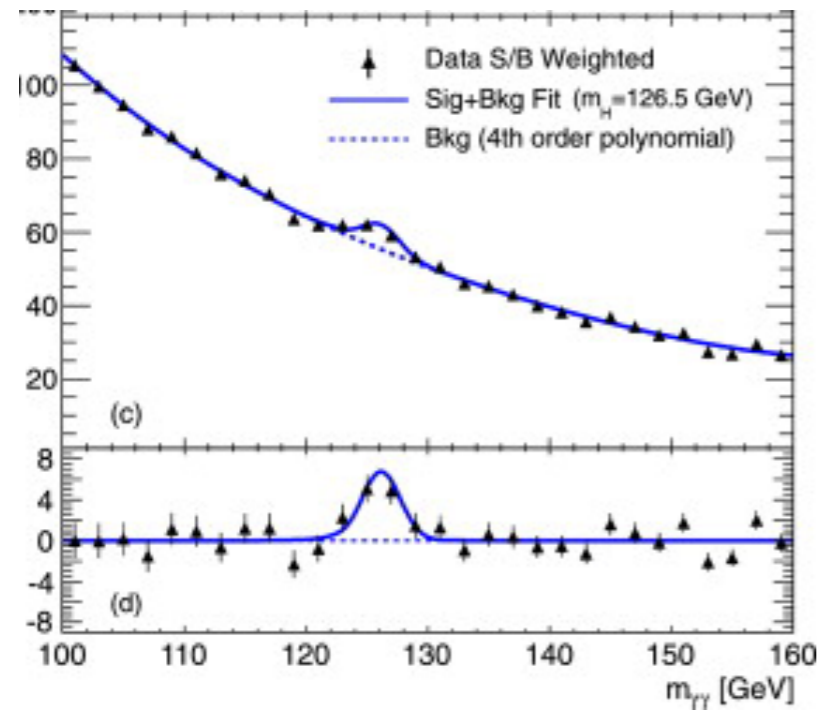
### Error on Shape



More reasonable (~10%)

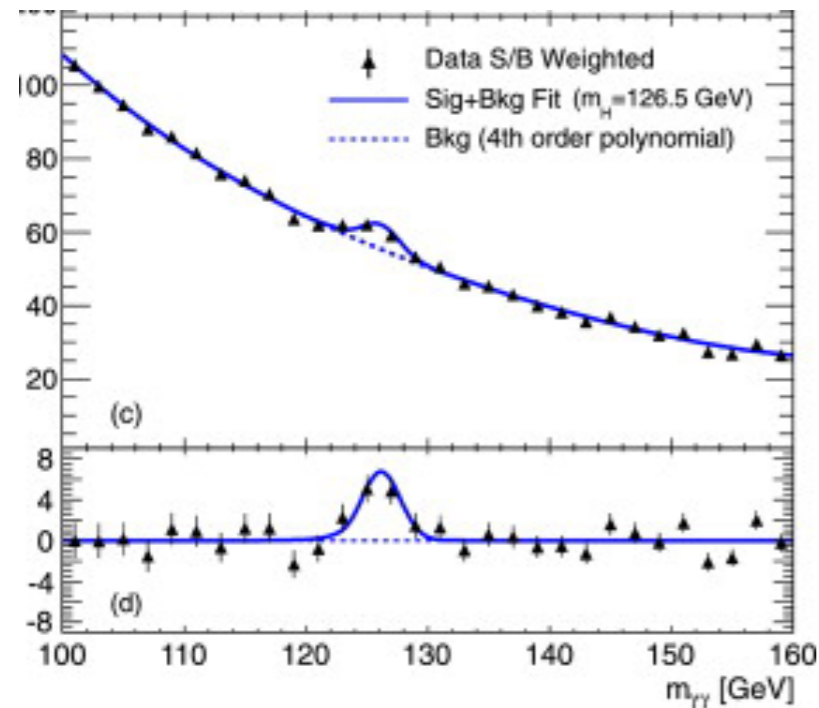
# Kind of measurement

## Peak



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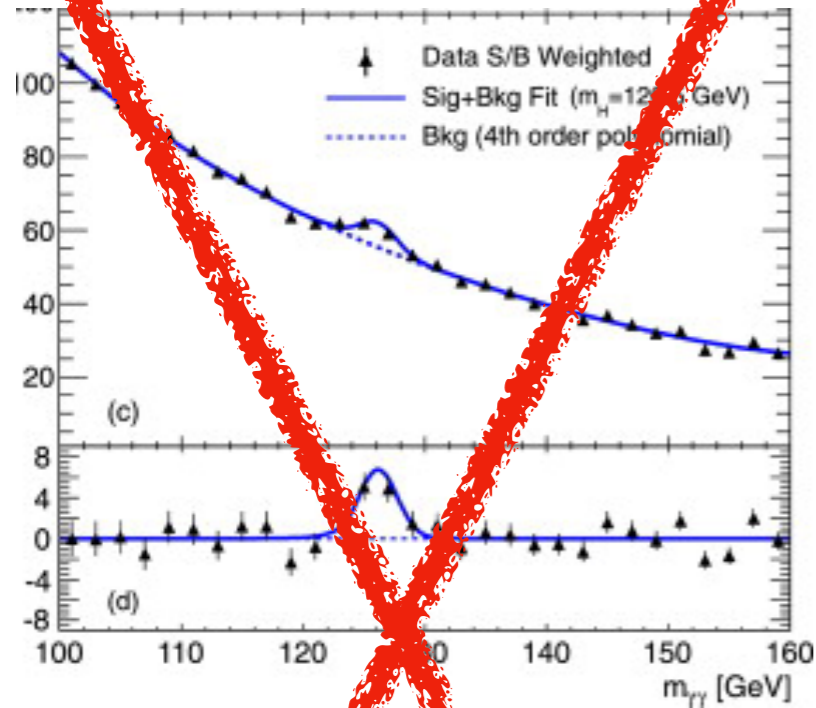


“EASY”

Background directly measured from **data**.  
Theory needed only for parameter extraction and validation

# Kind of measurement

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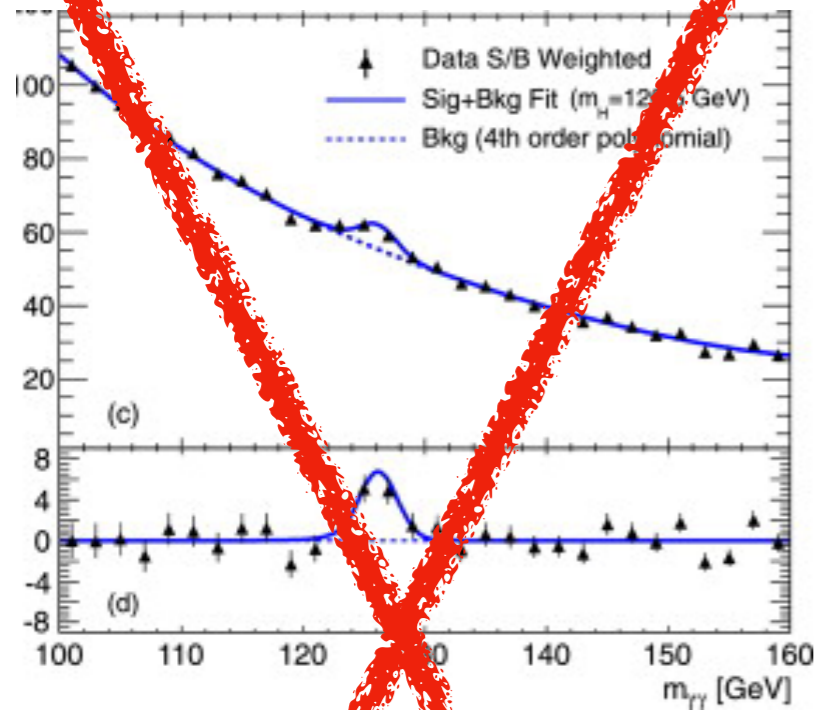
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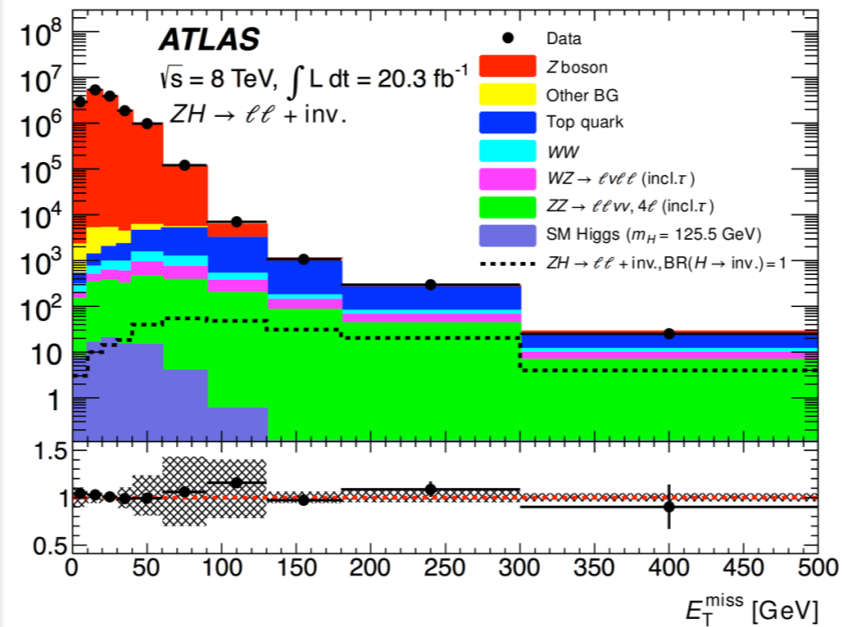
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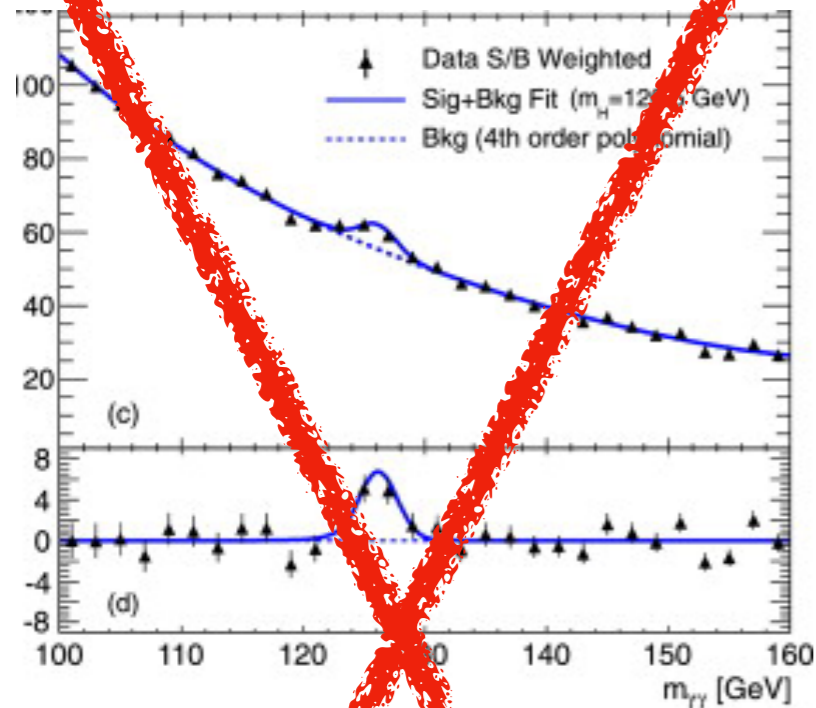
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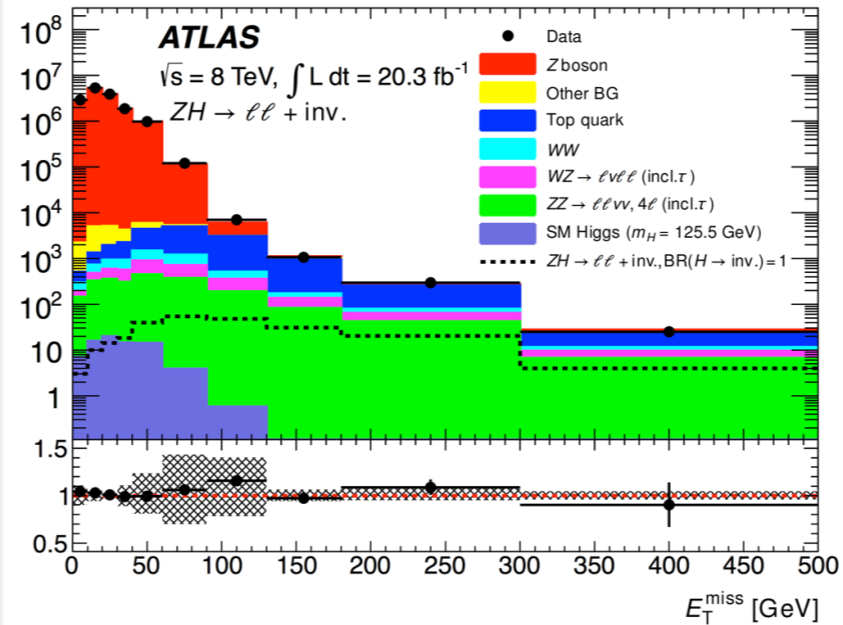
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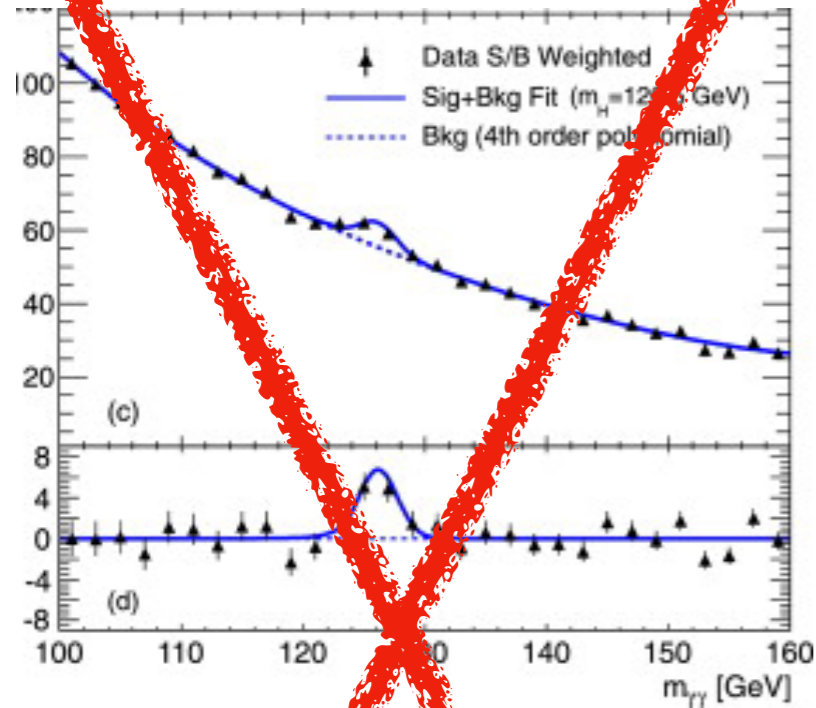


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Background **SHAPE** needed.  
Flexible MC for both signal and background validated and tuned to data

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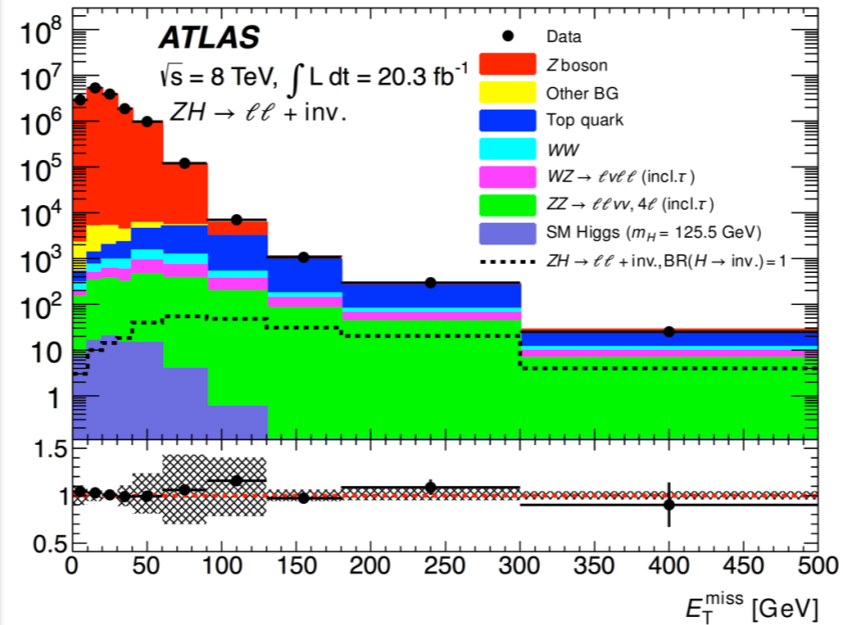
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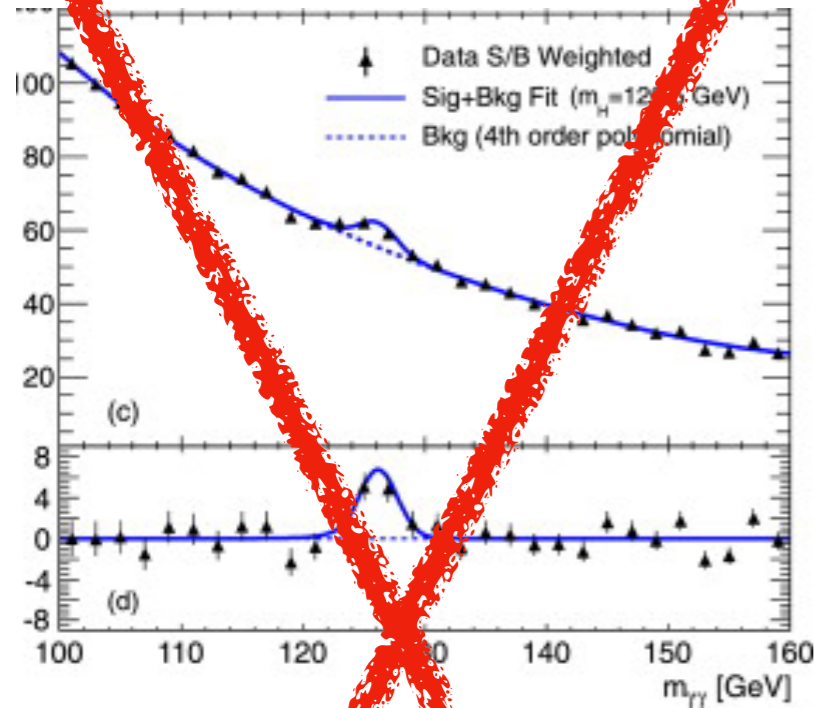
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LO  
NLO for HL

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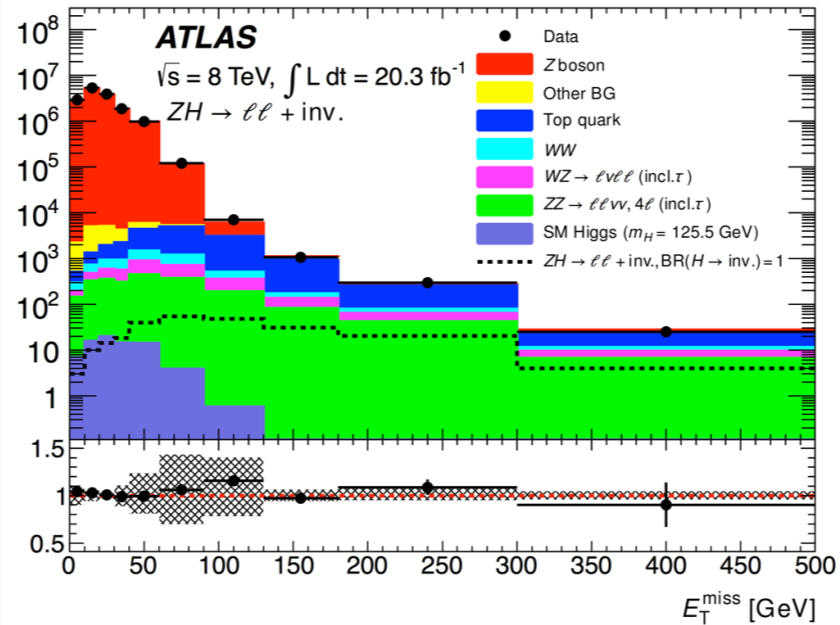
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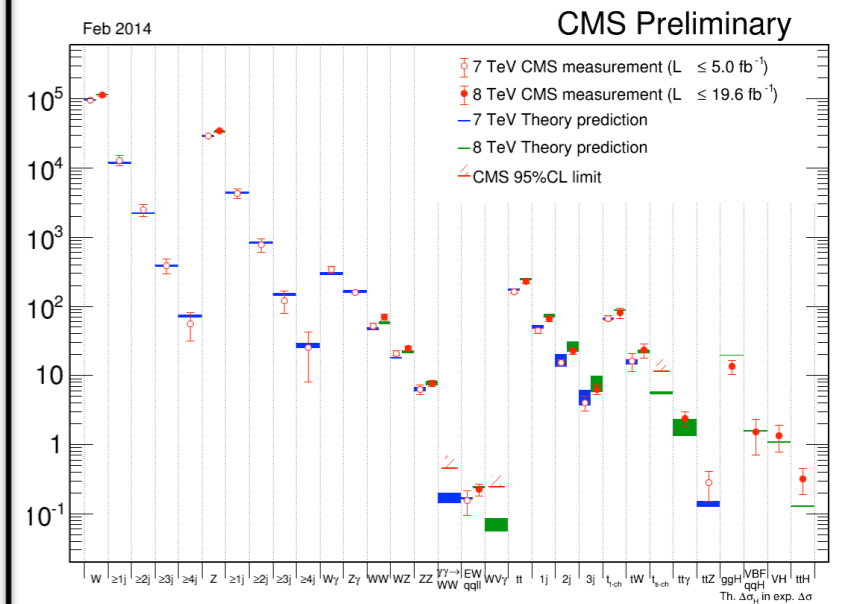


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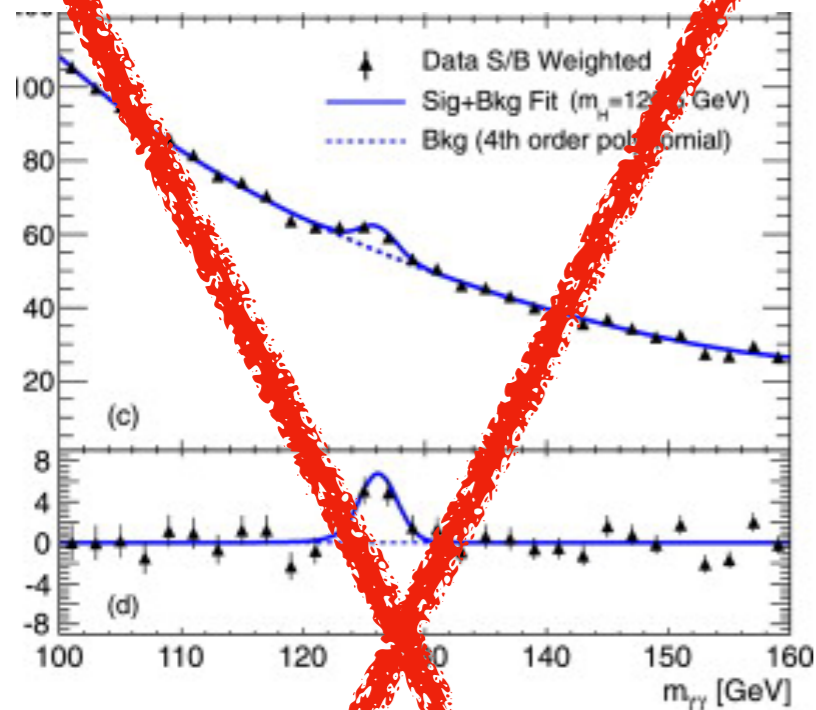
## Rate





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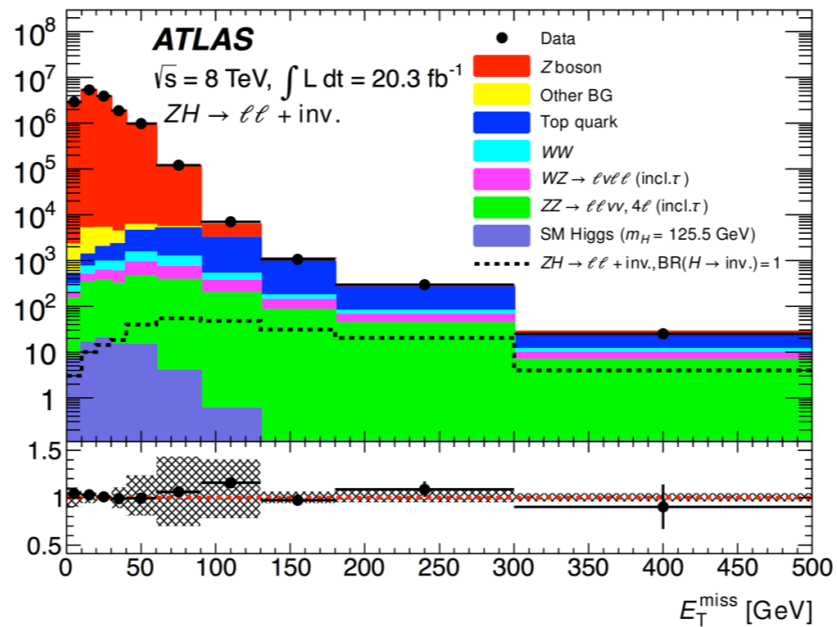
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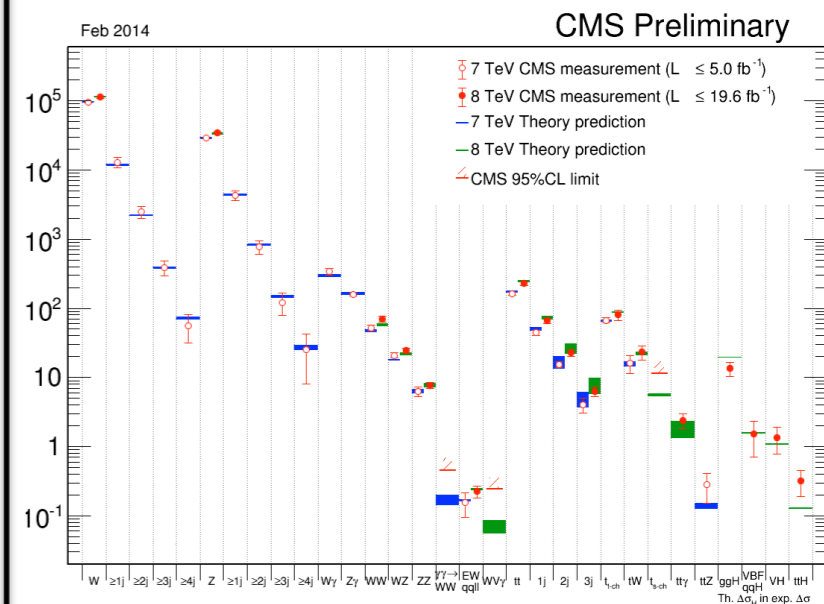


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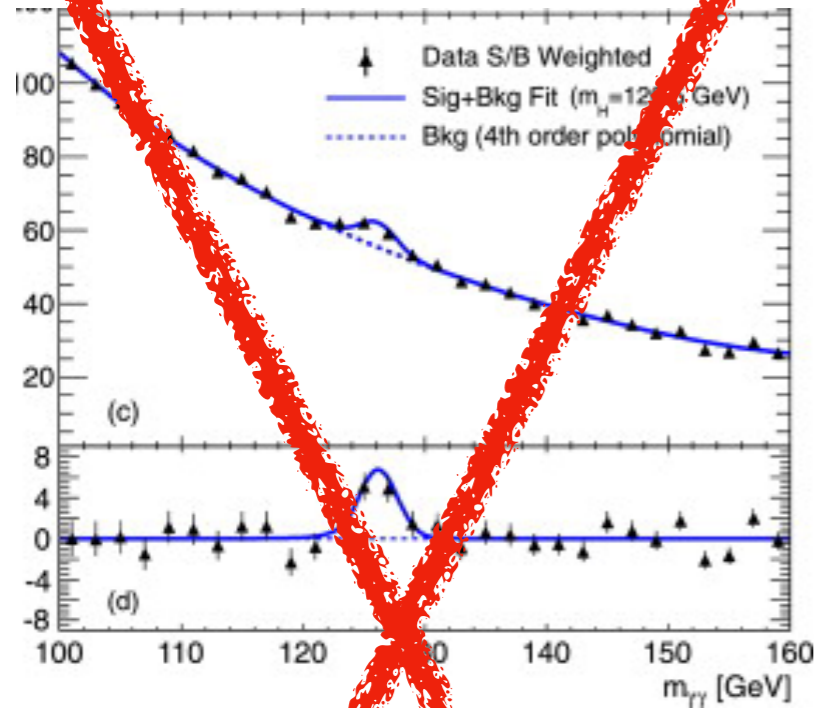


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 Complicated interplay of best simulations and data

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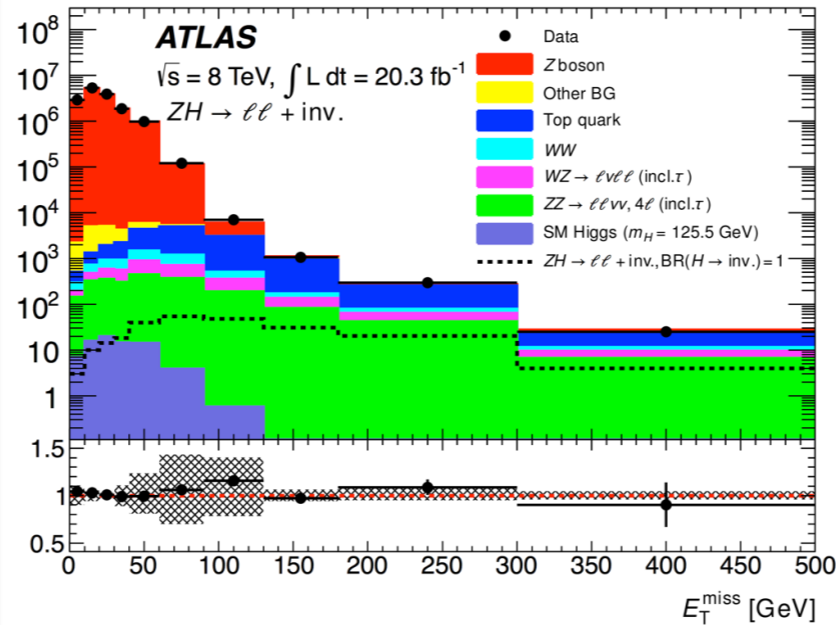
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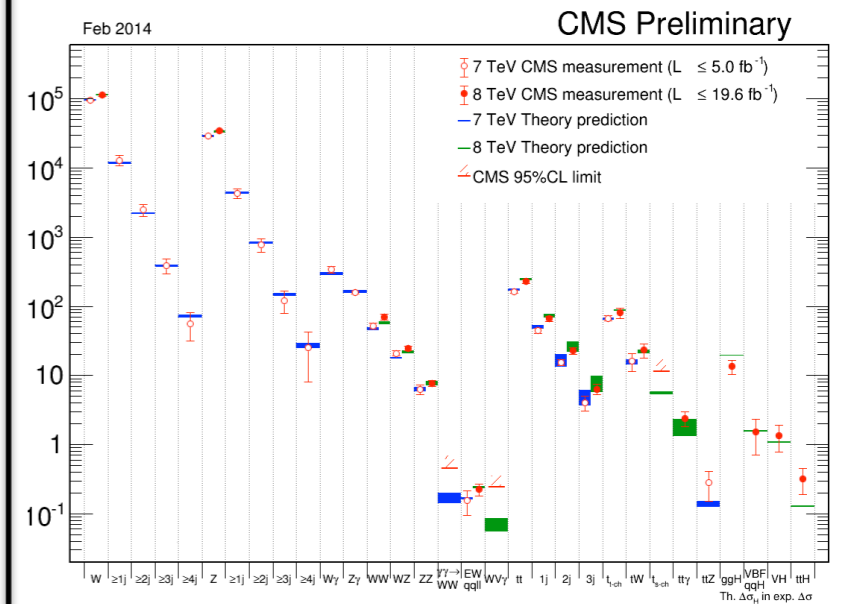


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NLO required  
often NNLO for HL



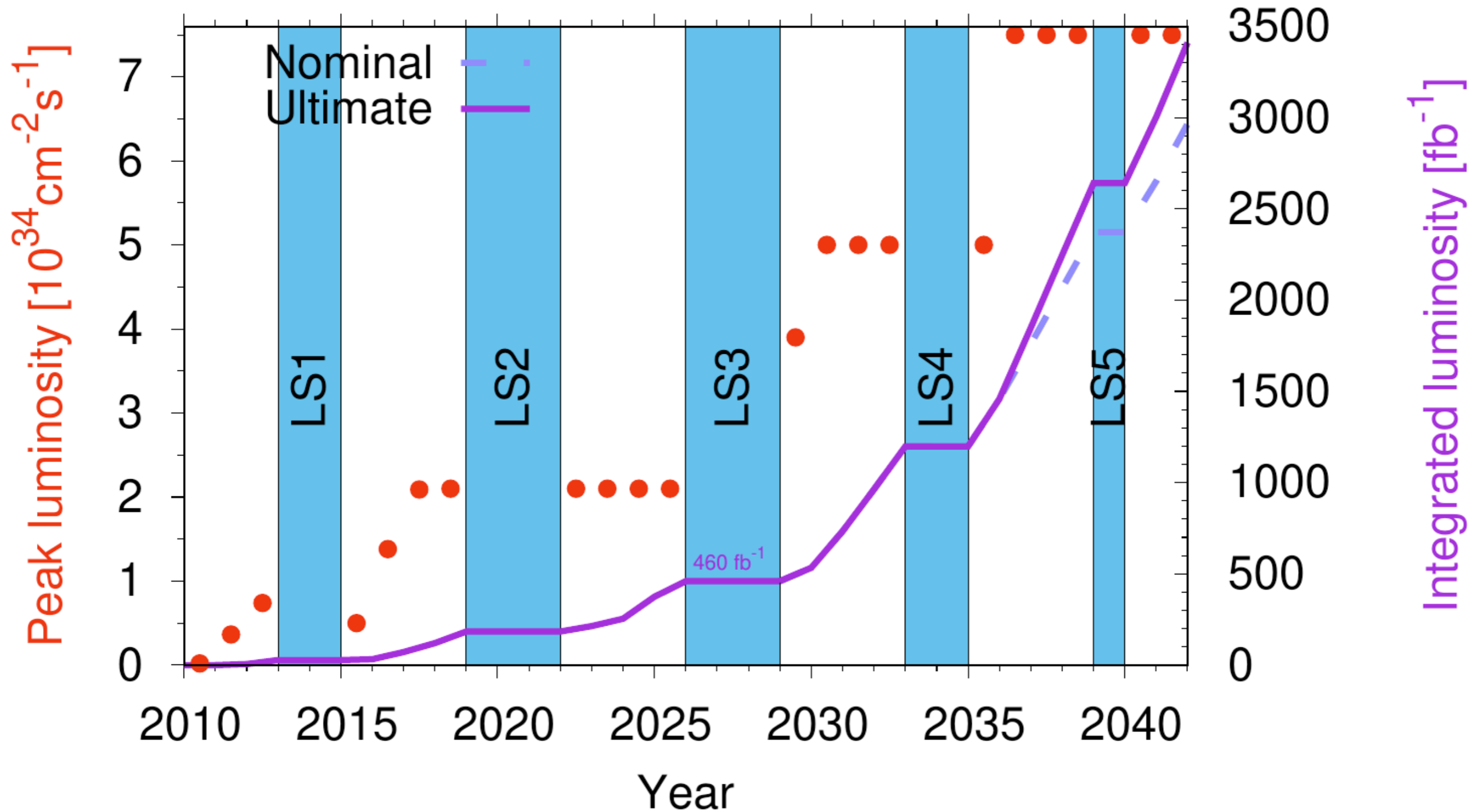
# Plan



- We want **MORE, MORE and MORE**
  - **More** accurate (NLO/NNLO)
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Do we really need more Events?

# HL-LHC plan



We are at half time of the LHC, we will collect 6 times more data

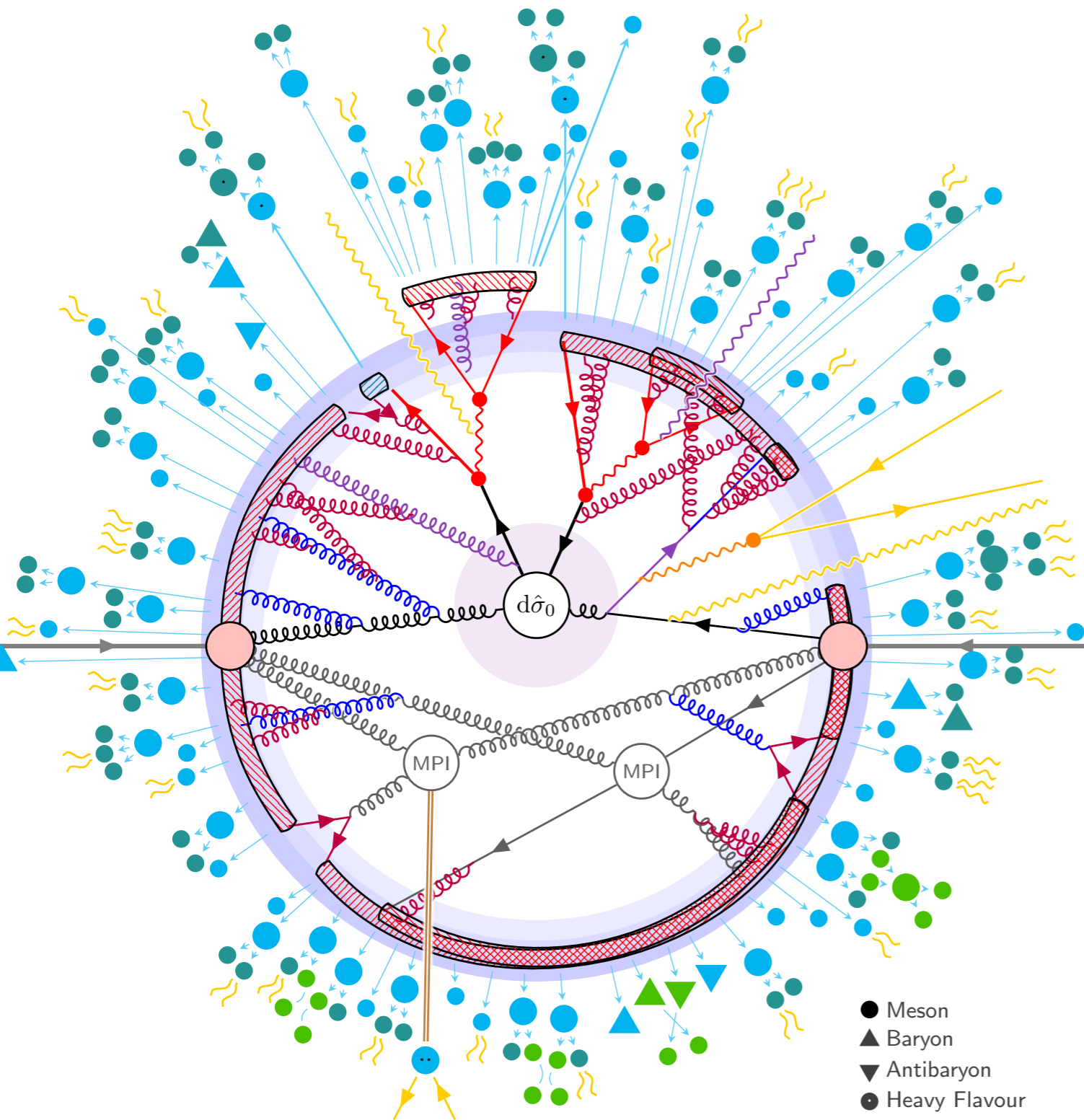
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Do we really need more Complexity?

# Where do we start PS?

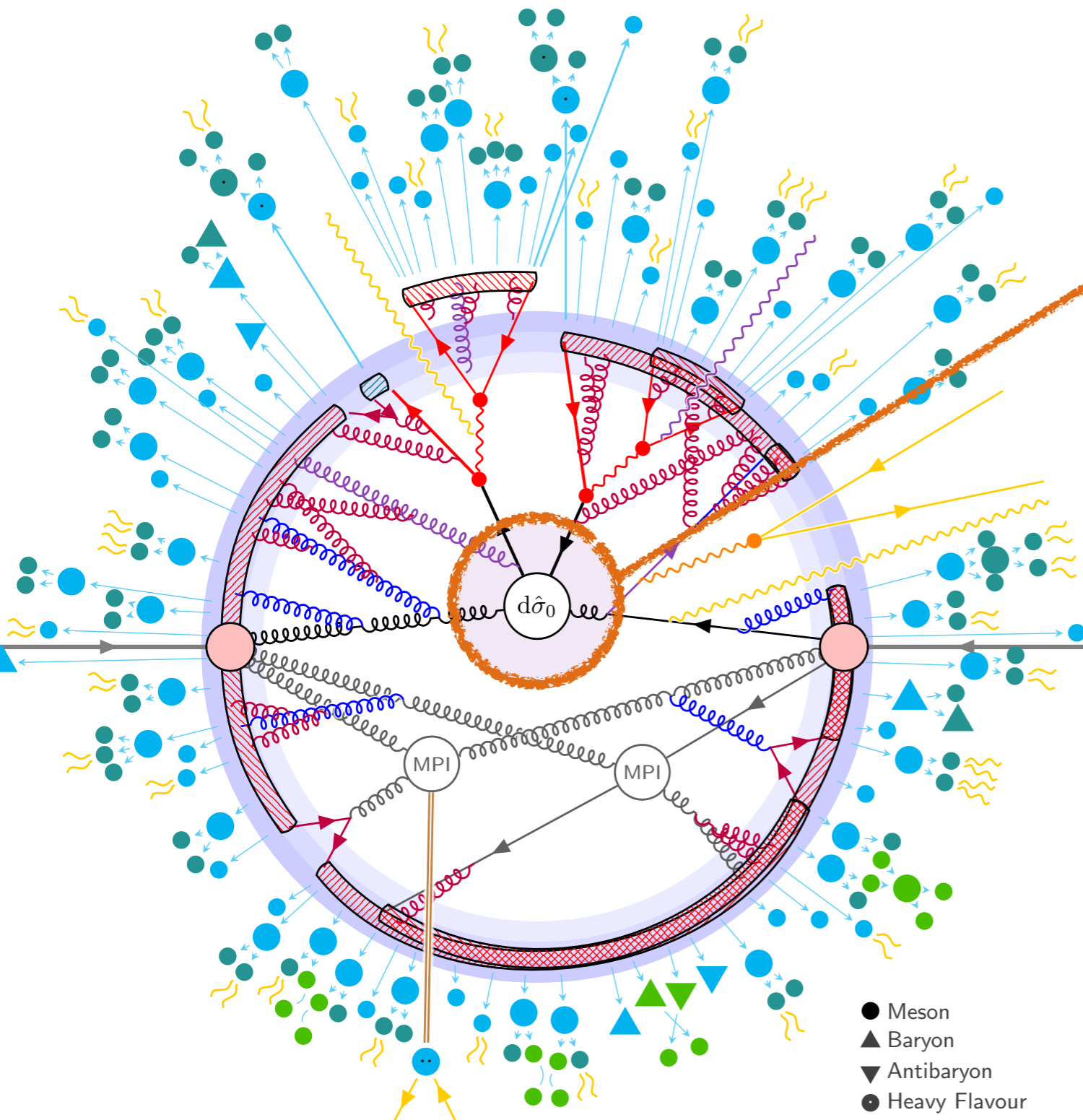


From pythia8 manual

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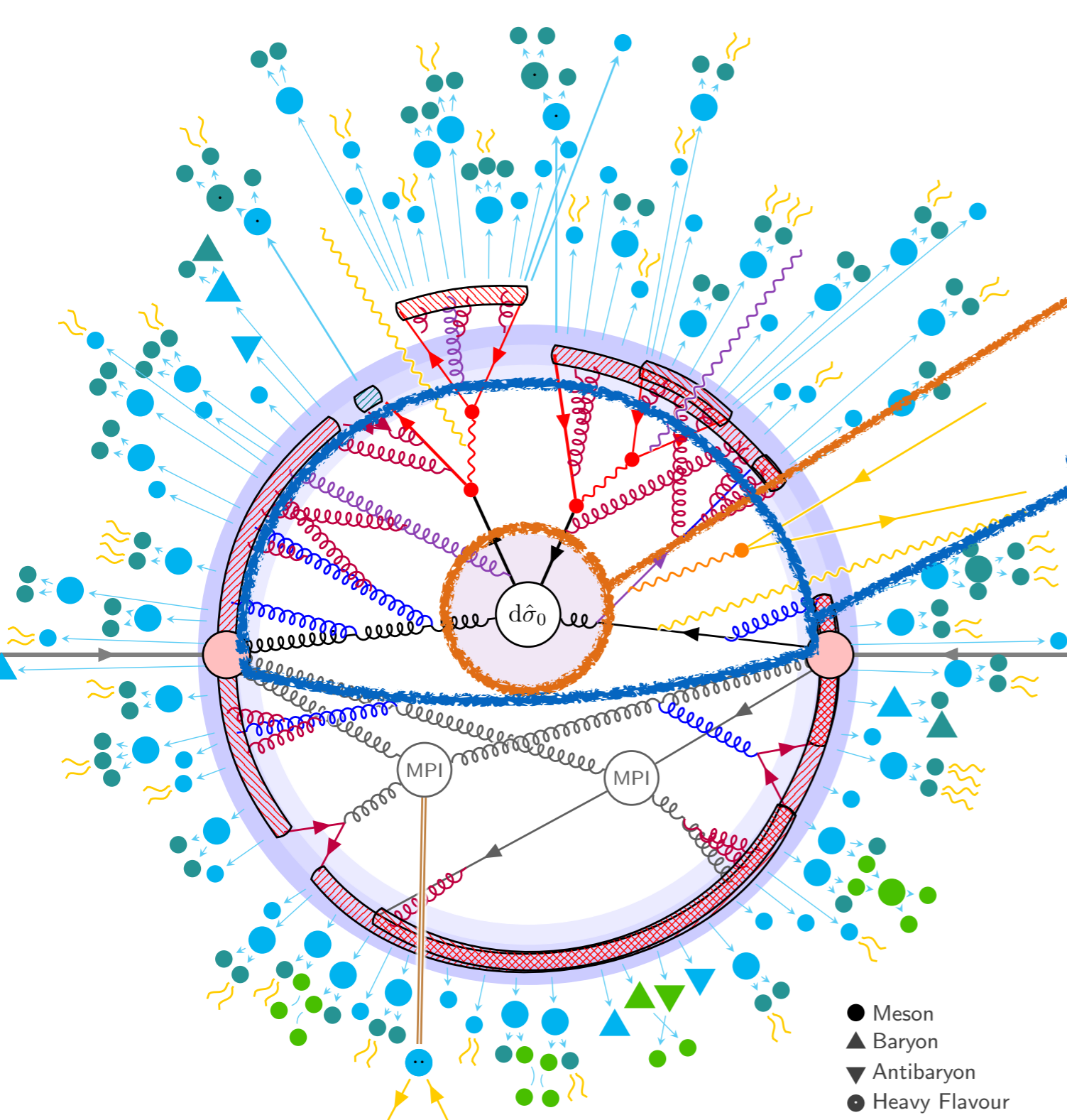
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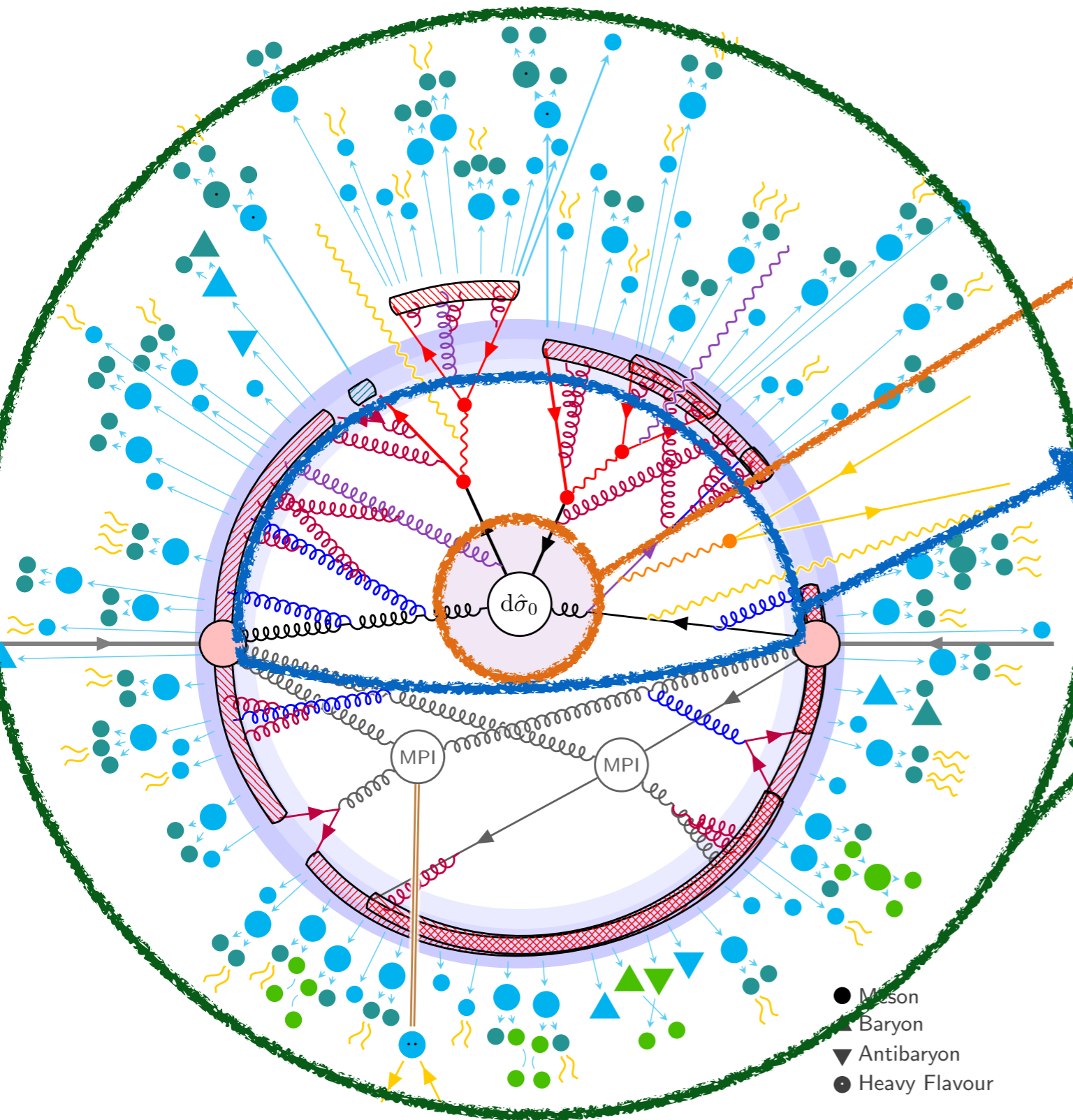
## Parton-Shower

- Markov-Chain
- Lot of branching

- Meson
- ▲ Baryon
- ▼ Antibaryon
- Heavy Flavour



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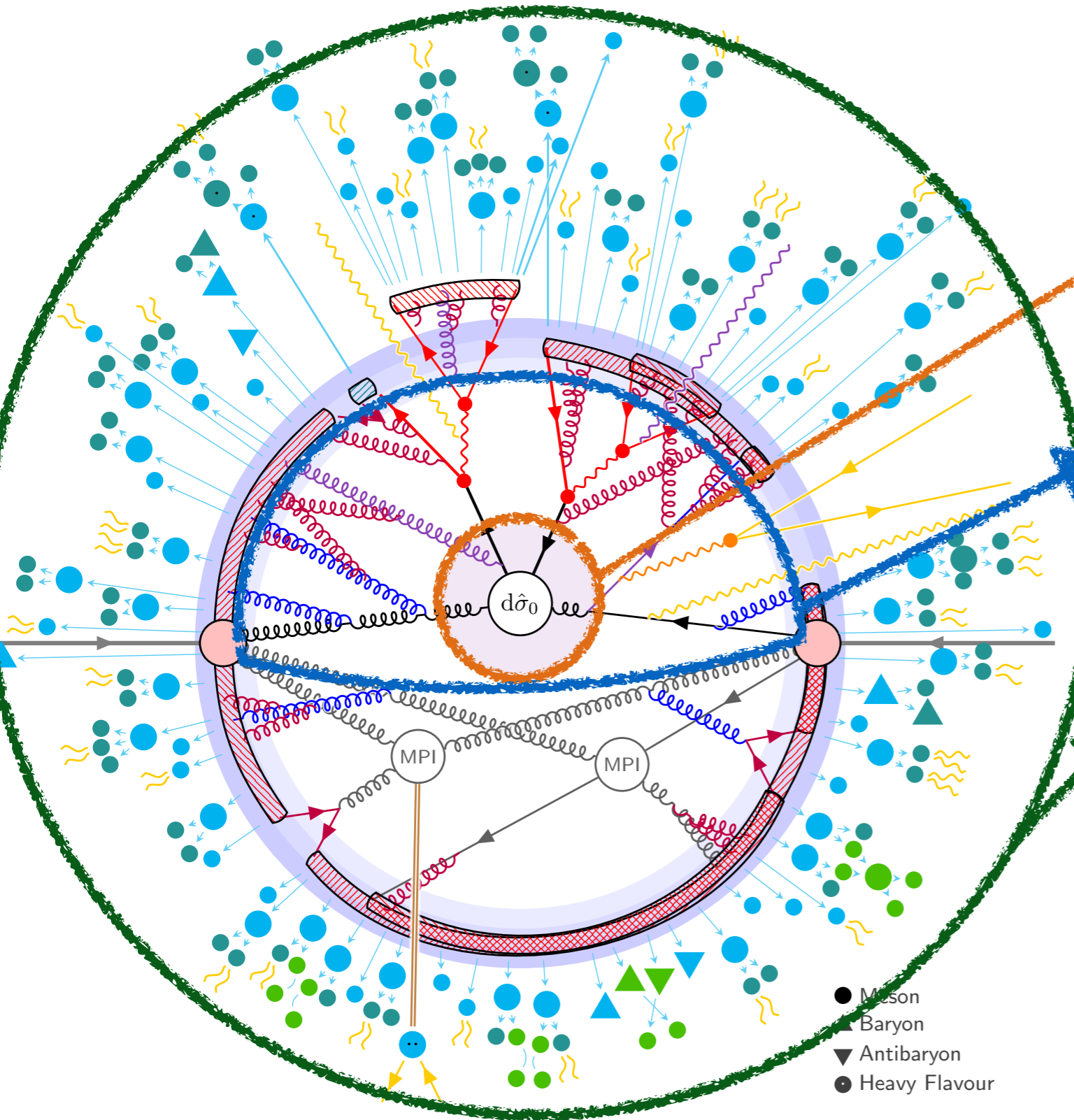
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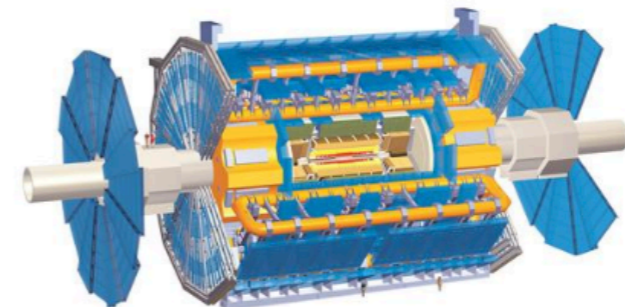
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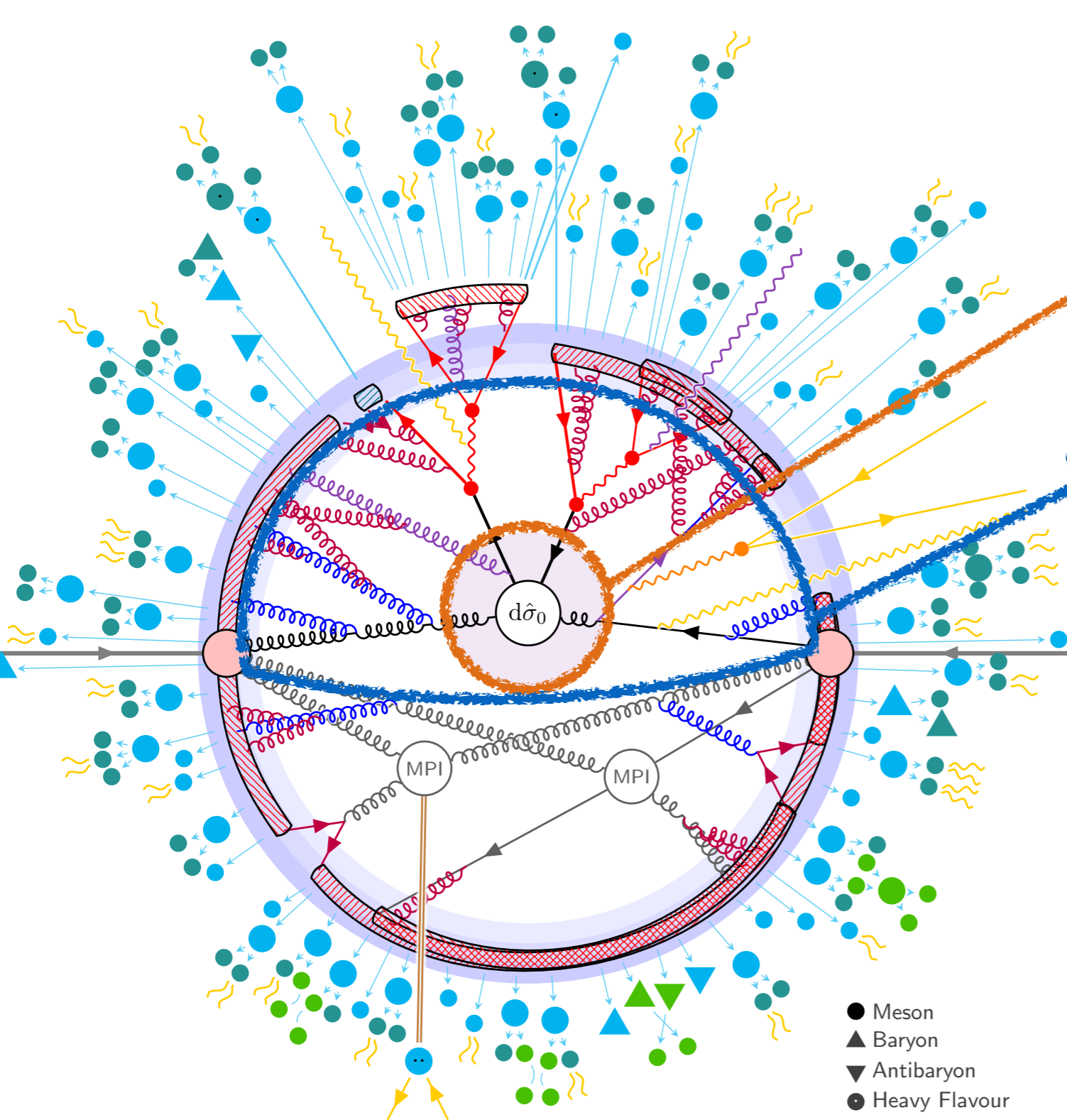
- Model Based

## Detector simulation



From pythia8 manual

# Where do we start PS?



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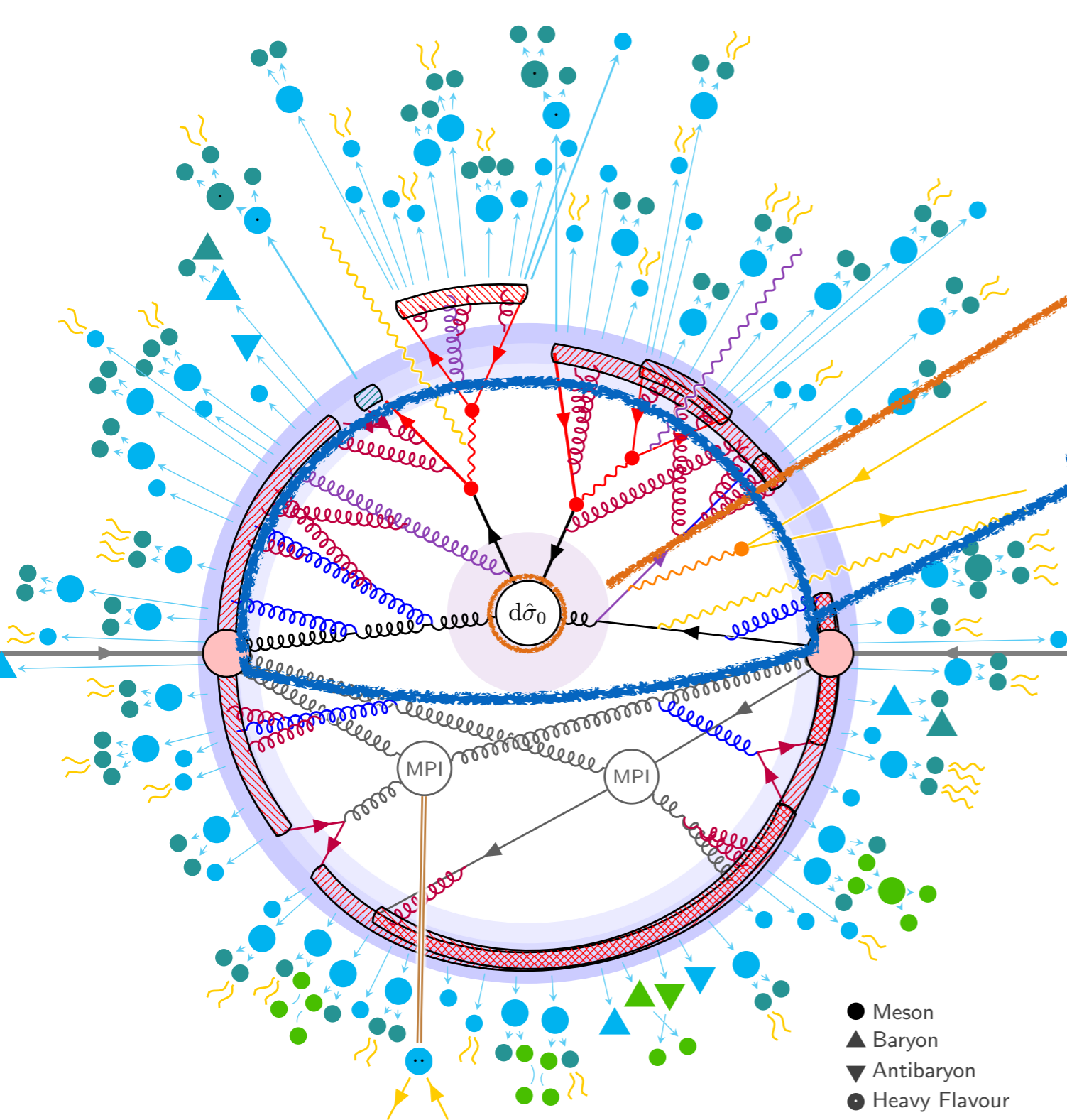
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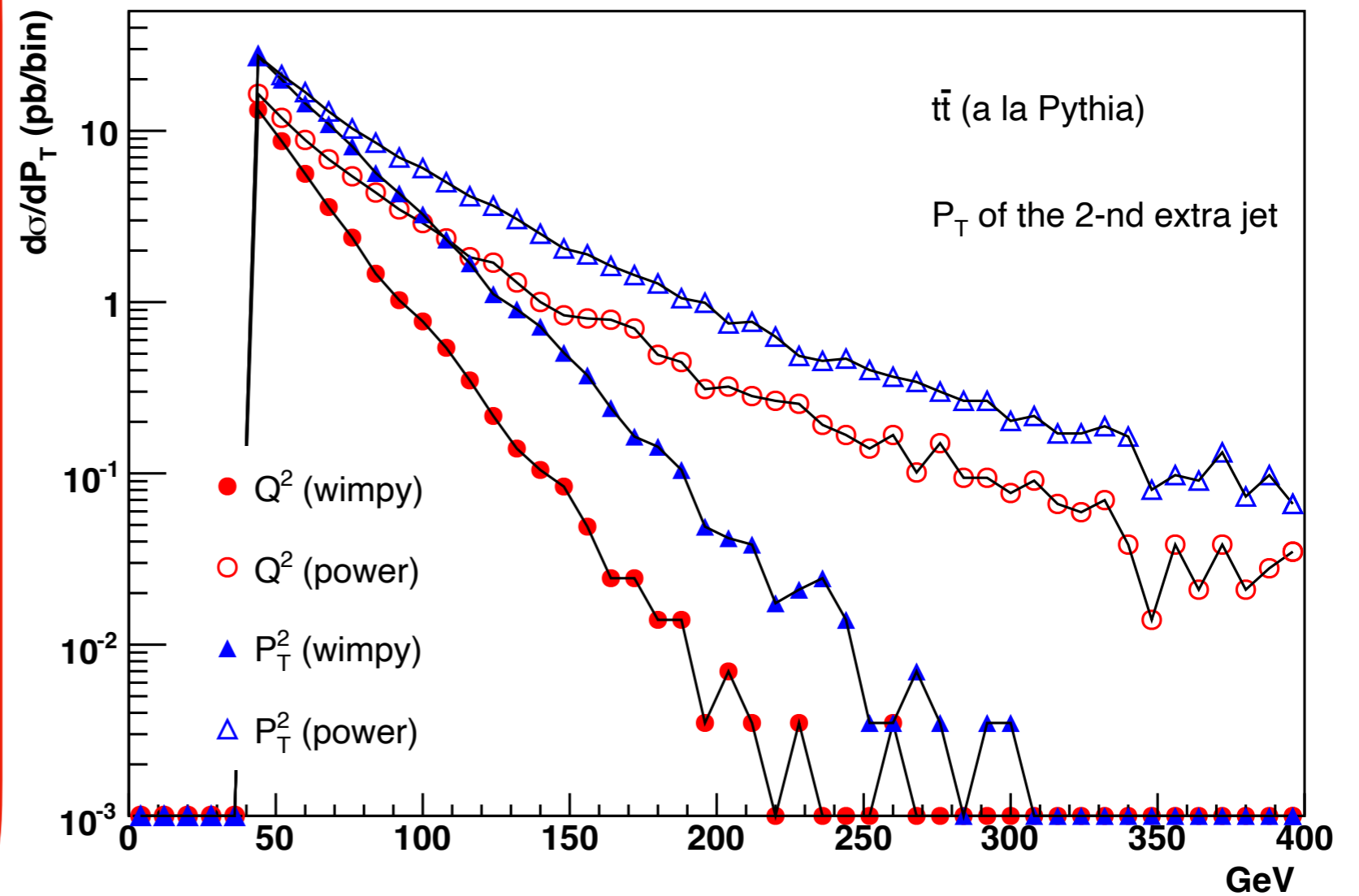
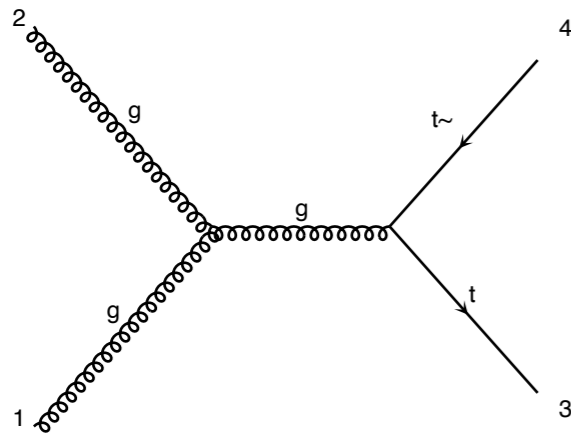
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# PS alone vs matched samples

If you stop too early: too large dependence in the Parton-Shower

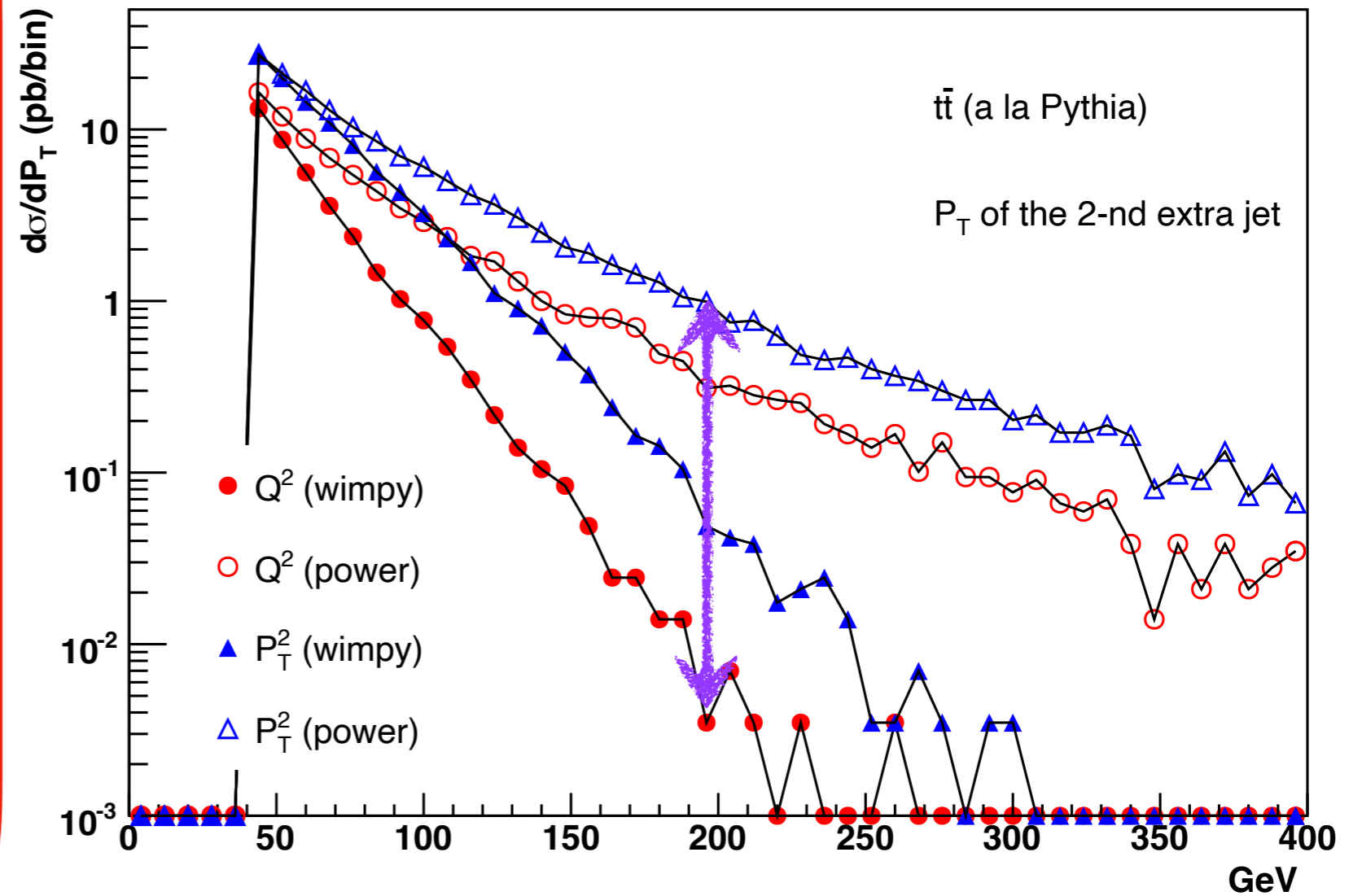
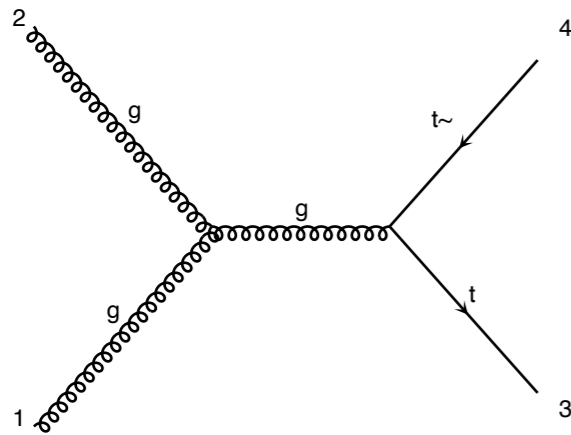
Sample generated



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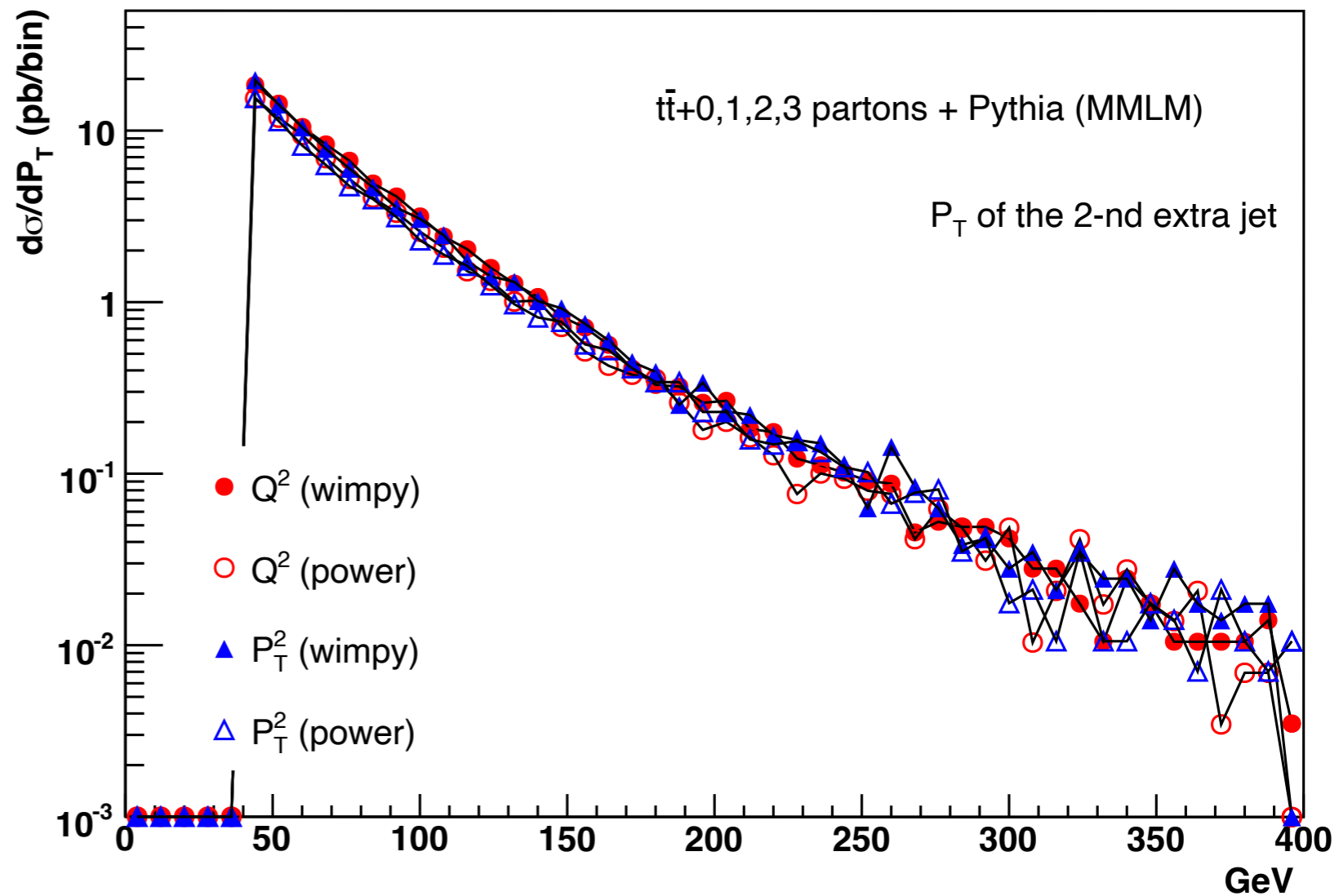
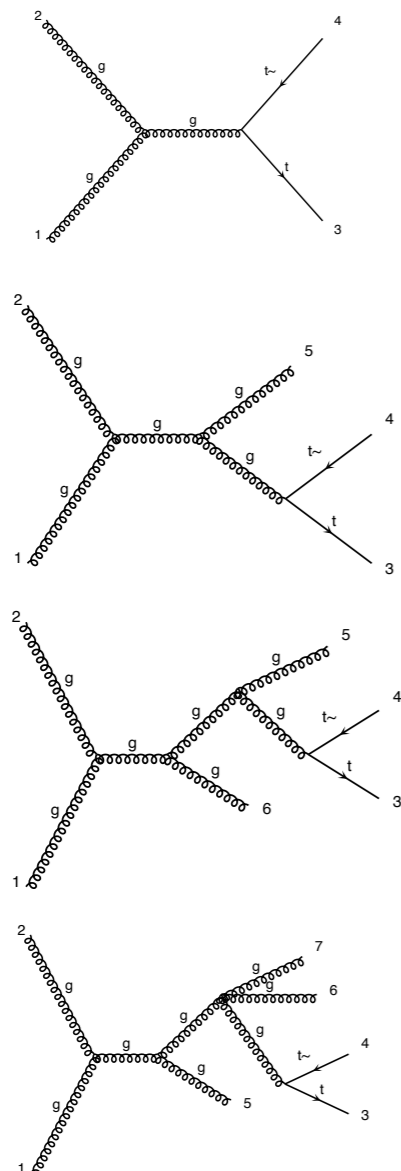
Two order of magnitude in the prediction



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Sample generated

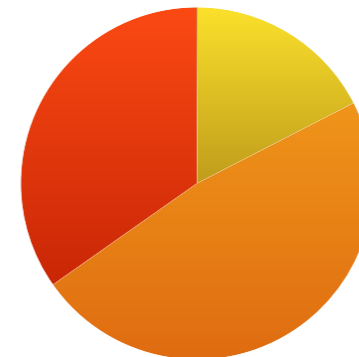
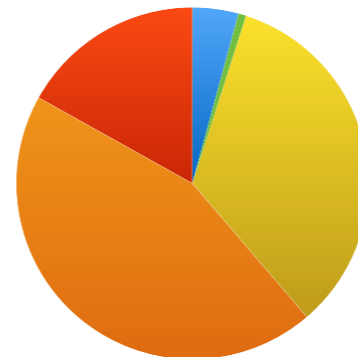
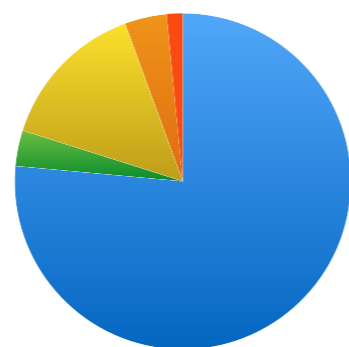


Error dominated by other factor !!

# Code Complexity

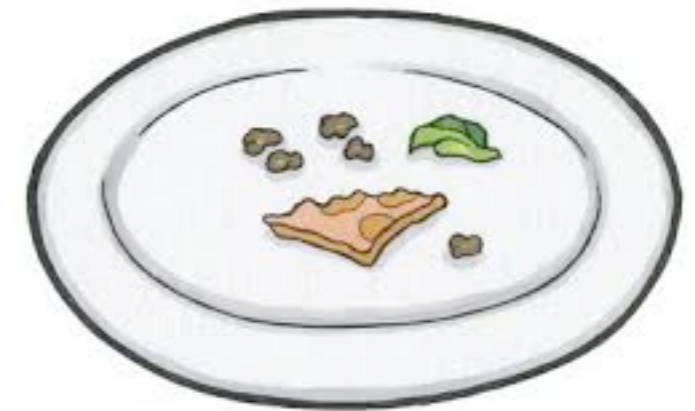
	$gg \rightarrow t\bar{t}$	$gg \rightarrow t\bar{t}gg$	$gg \rightarrow t\bar{t}ggg$
<b>madevent</b>	13G	470G	11T
<b>matrix1</b>	3.1G (23%)	450G (96%)	11T (>99%)

- color
- amplitude
- int/propagator
- external
- not ME



Complexity raise factorially square

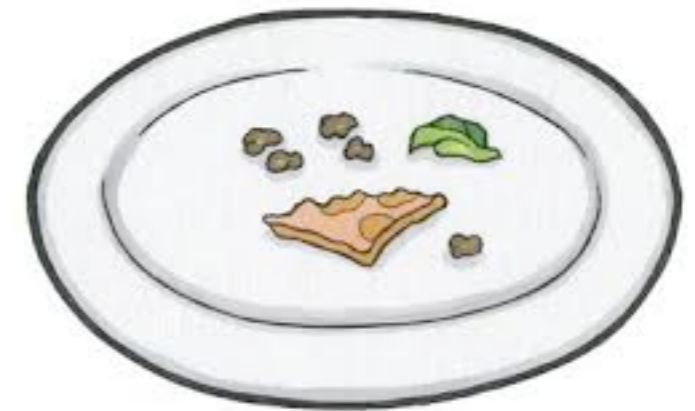
# Is it sustainable ?



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Can we do More with Less ?

# Is it sustainable ?



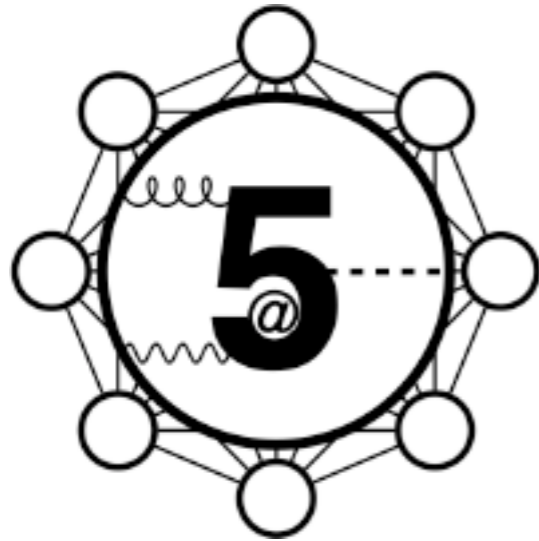
shutterstock.com · 274008899

Can we do More with Less ?

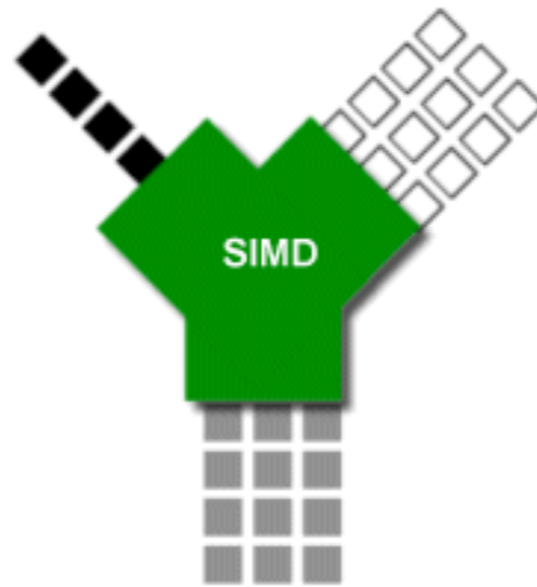
**Yes We Can!**



# Type of Solution



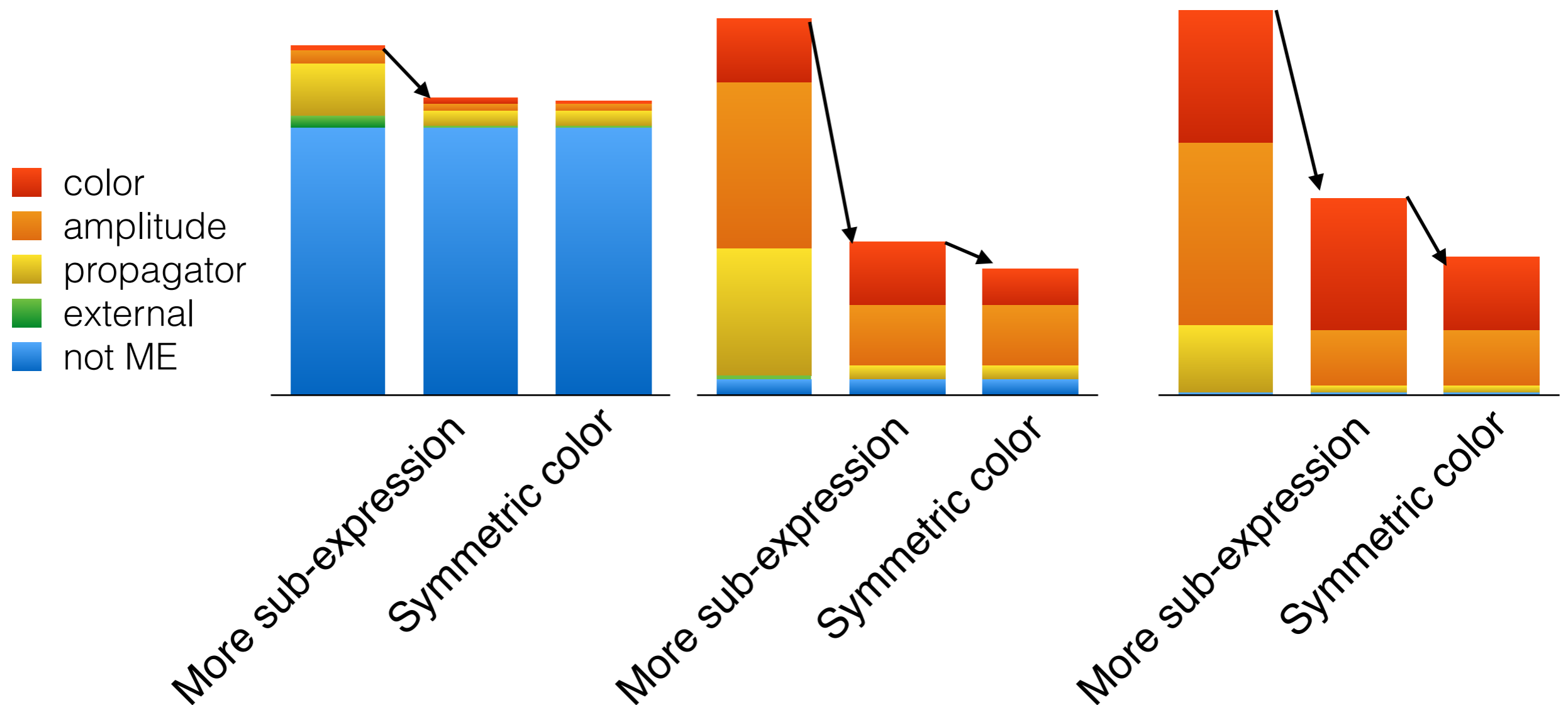
- We need better **code efficiency**
- More efficient **algorithm**
- More efficient use of **CPU/GPU**





# Better evaluation Method

	$gg \rightarrow t\bar{t}$		$gg \rightarrow t\bar{t}gg$		$gg \rightarrow t\bar{t}ggg$	
	Instructions	Reduction	Instructions	Reduction	Instructions	Reduction
<b>madevent</b>	11G	15%	180G	62%	5T	55%
<b>matrix1</b>	1G (9.3%)	68%	160G (90%)	64%	4.9T (98%)	55%

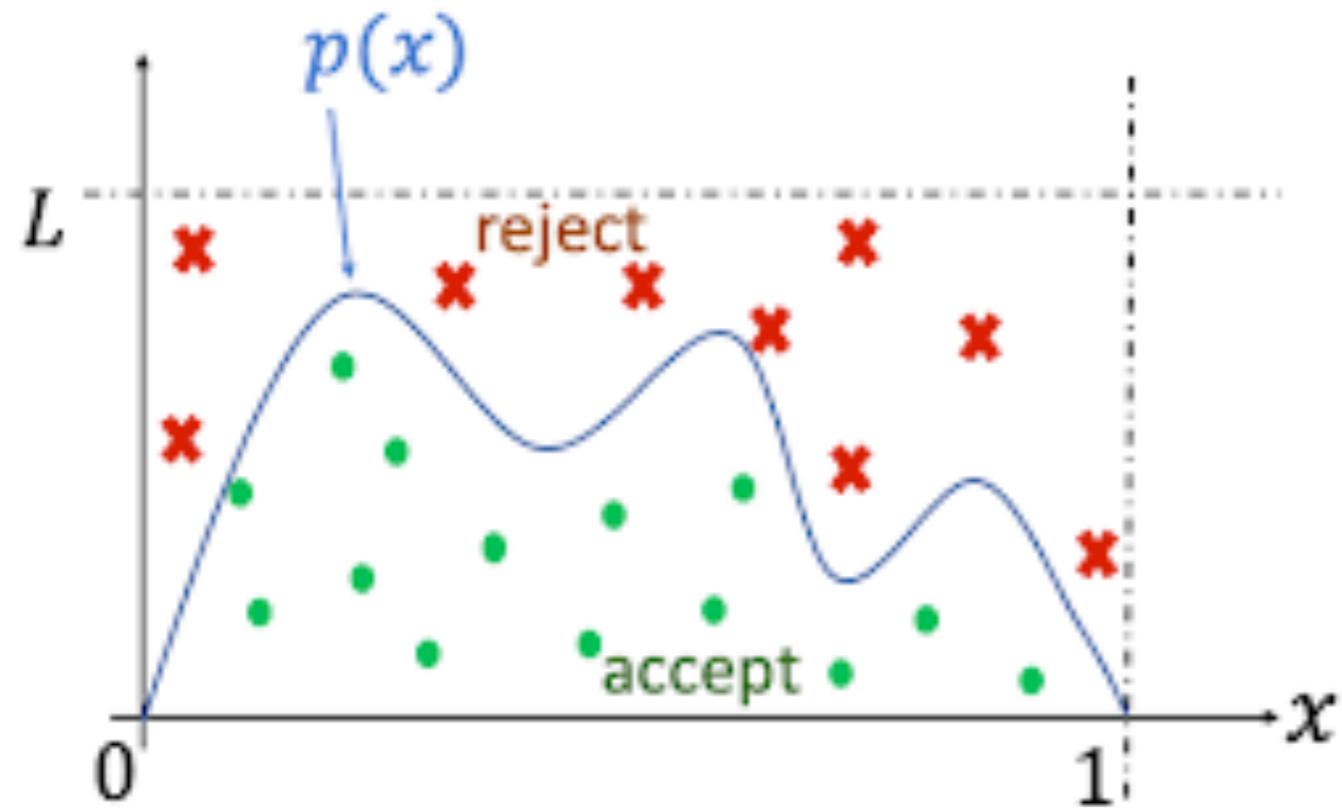


In Collaboration with Kiran Ostrelenk ([2102.00773](https://arxiv.org/abs/2102.00773))



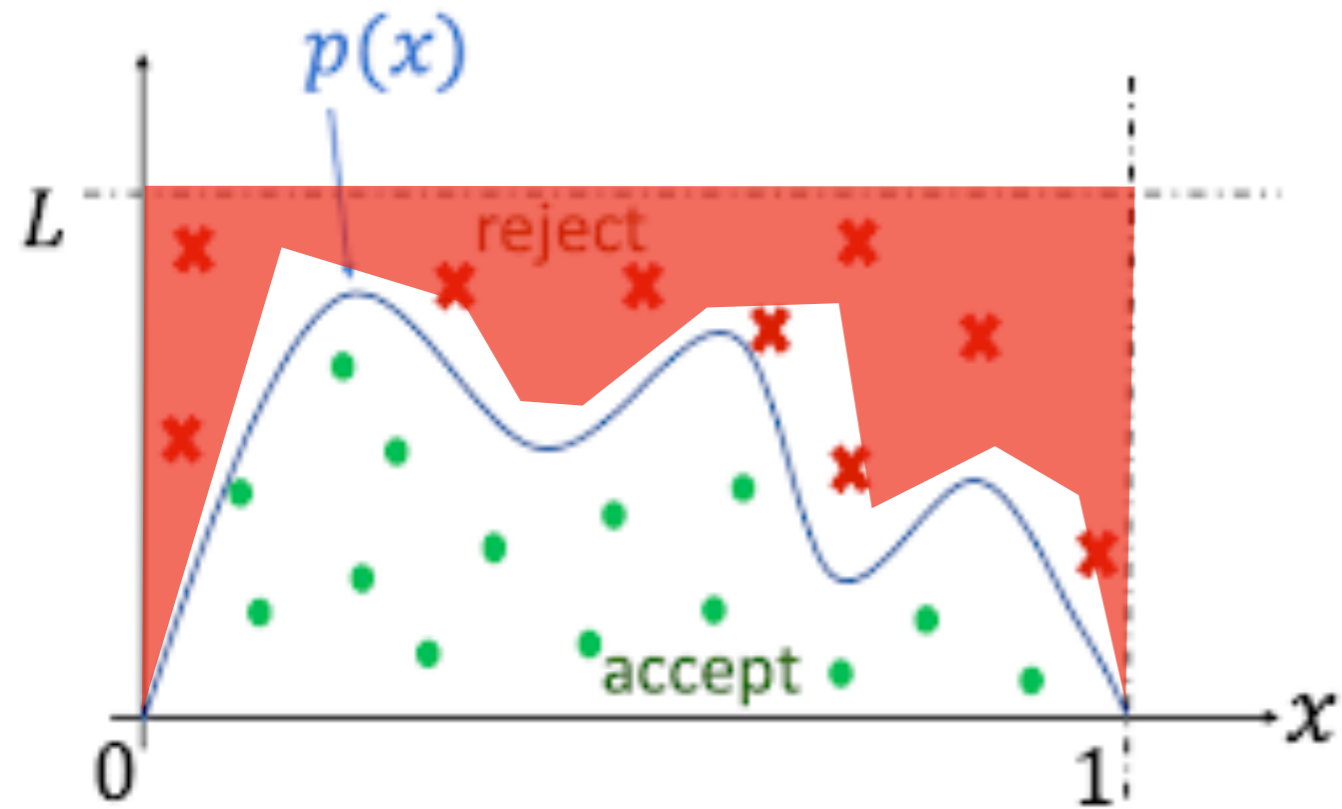
# Machine Learning

Monte-Carlo generation is based on accept/reject



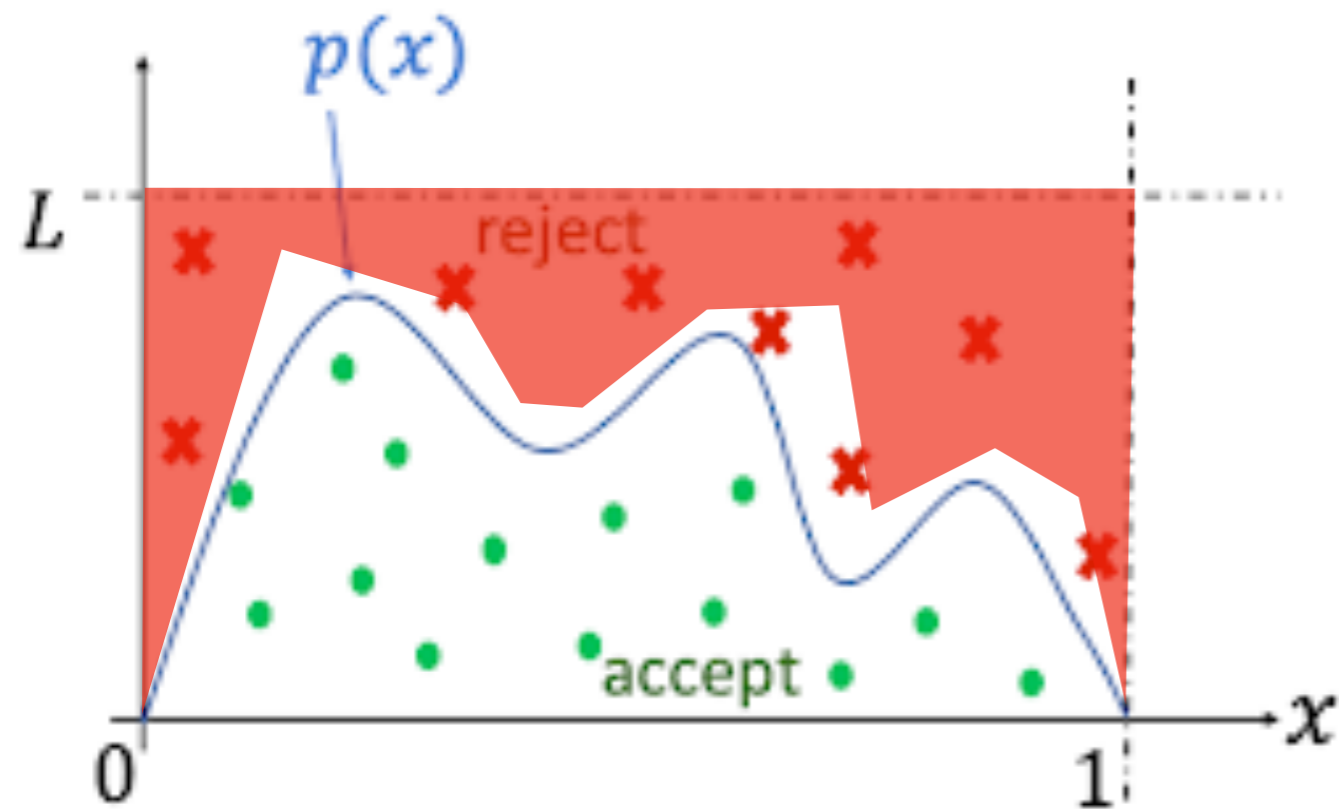
# Machine Learning

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# Machine Learning

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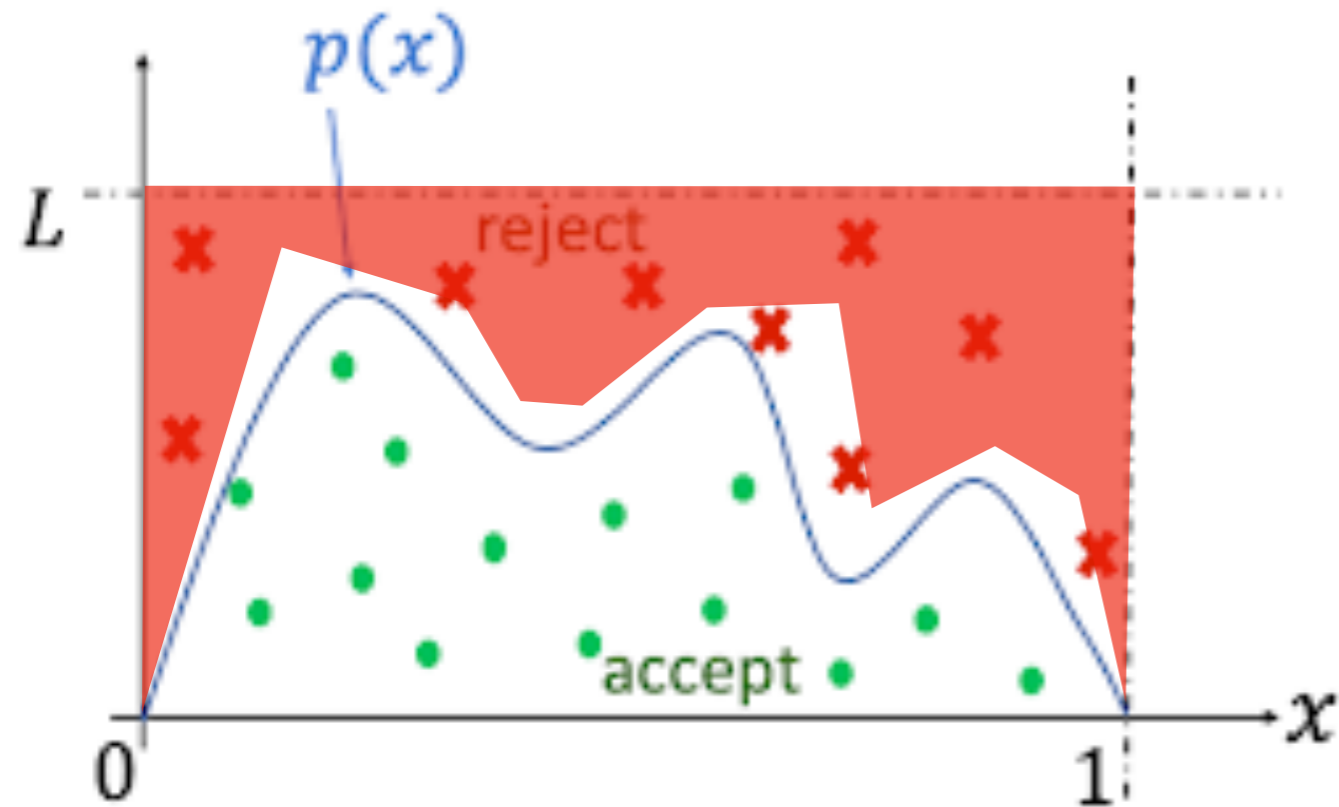


Machine Learning is key to improve our knowledge of the function

This is a perfect fit for modern AI

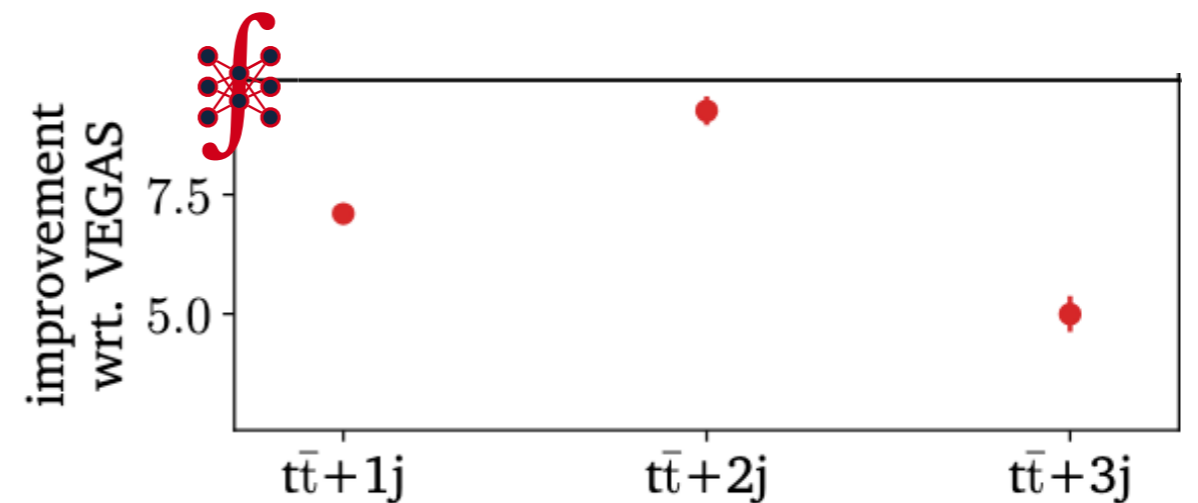
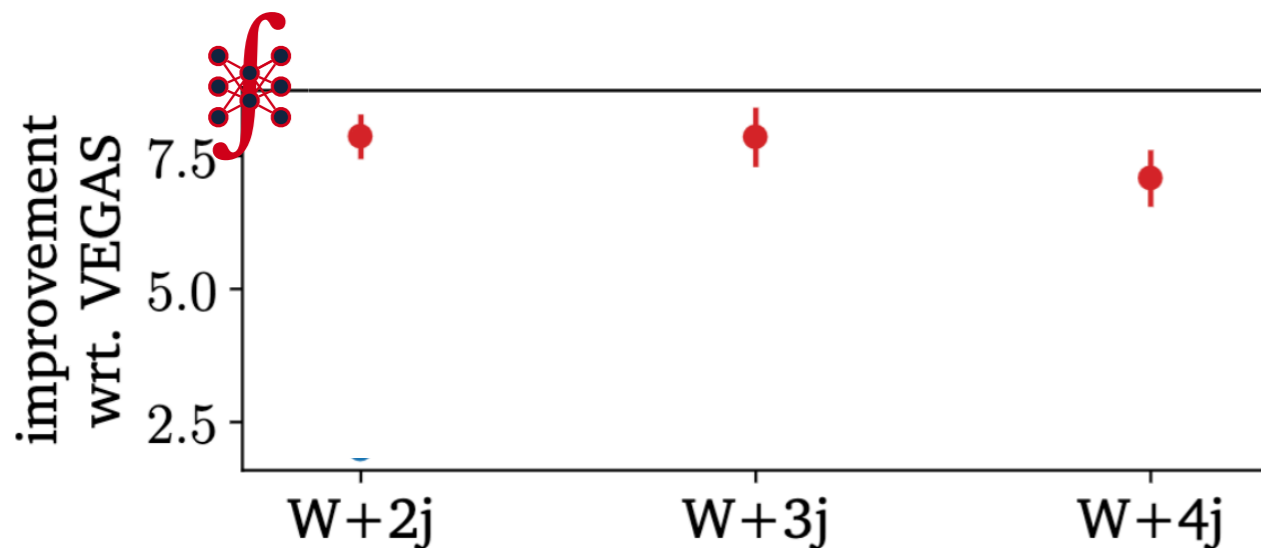
# Machine Learning

Monte-Carlo generation is based on accept/reject



Machine Learning is key to improve our knowledge of the function

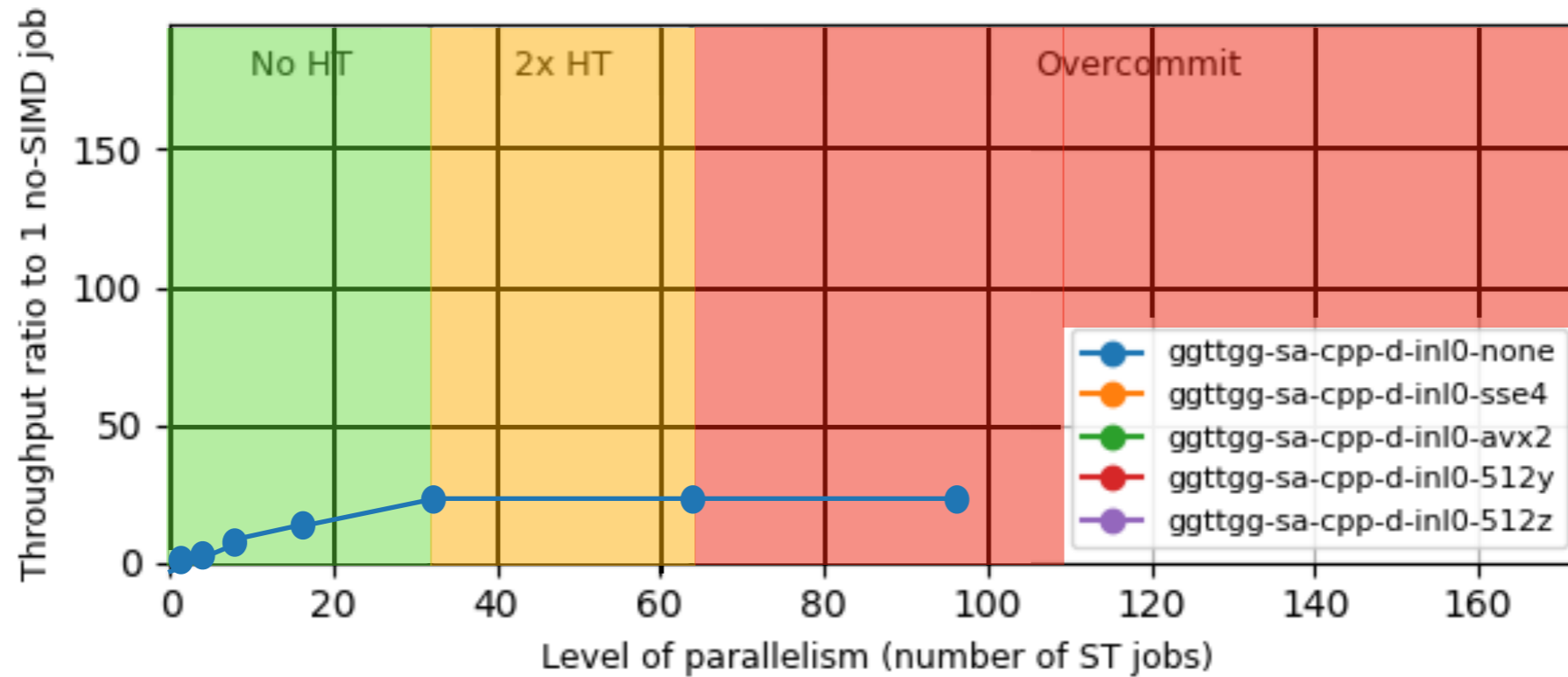
This is a perfect fit for modern AI



MadNIS collaboration: 2311.01548

# SIMD

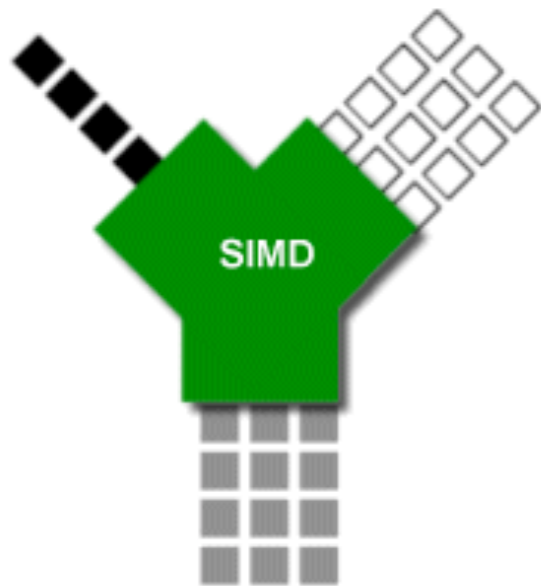
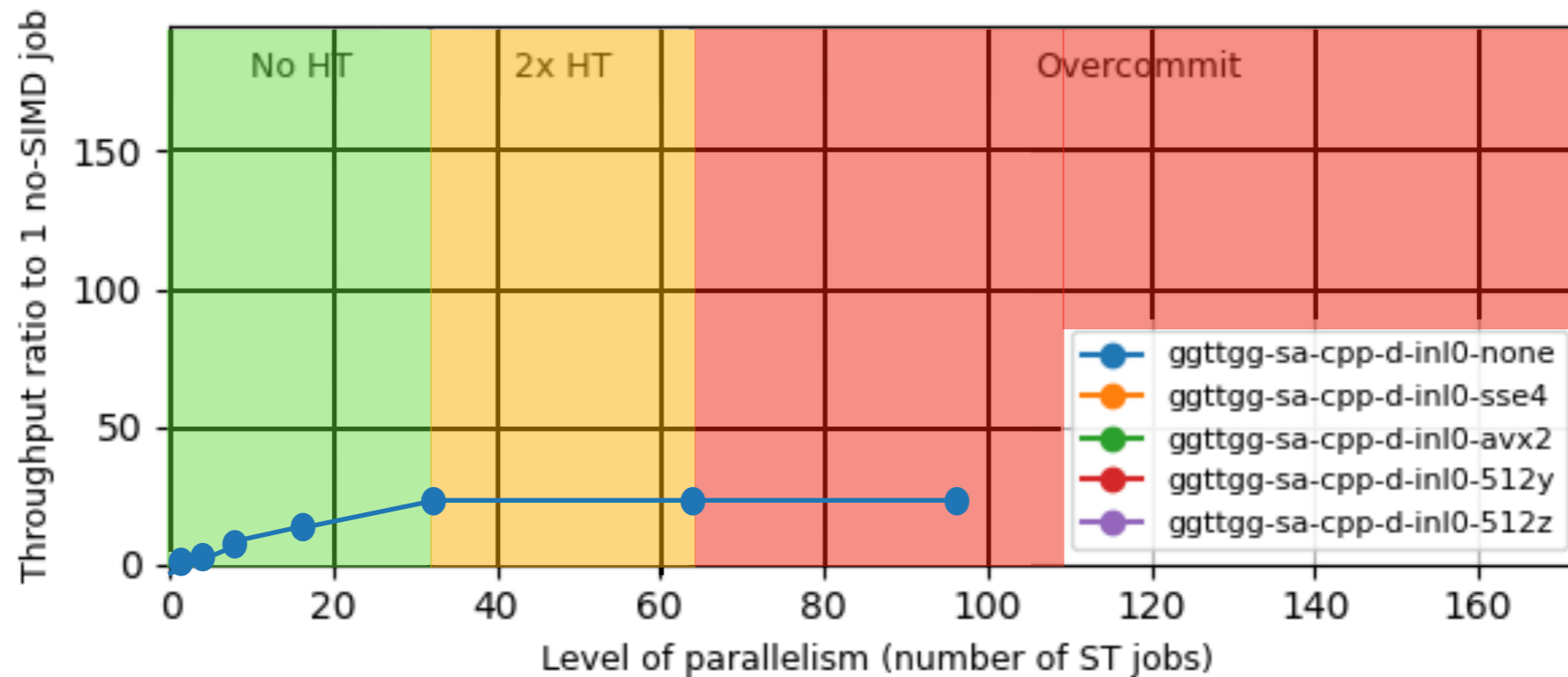
ore 2.1GHz Xeon Gold 6130 with 2x HT) for 10 cycles



See Andrea Valassi talk for more details

# SIMD

ore 2.1GHz Xeon Gold 6130 with 2x HT) for 10 cycles



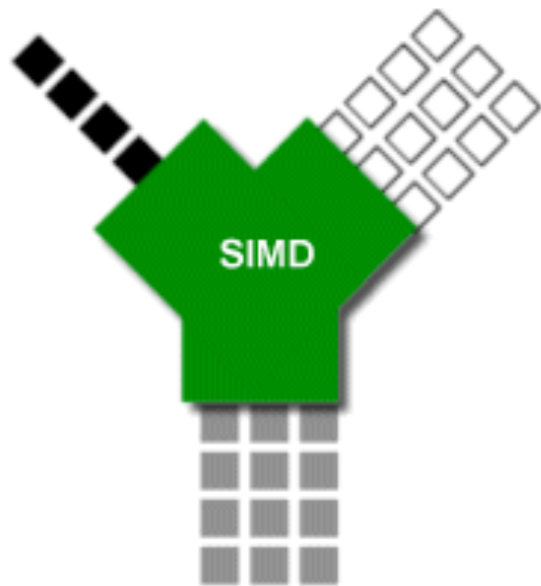
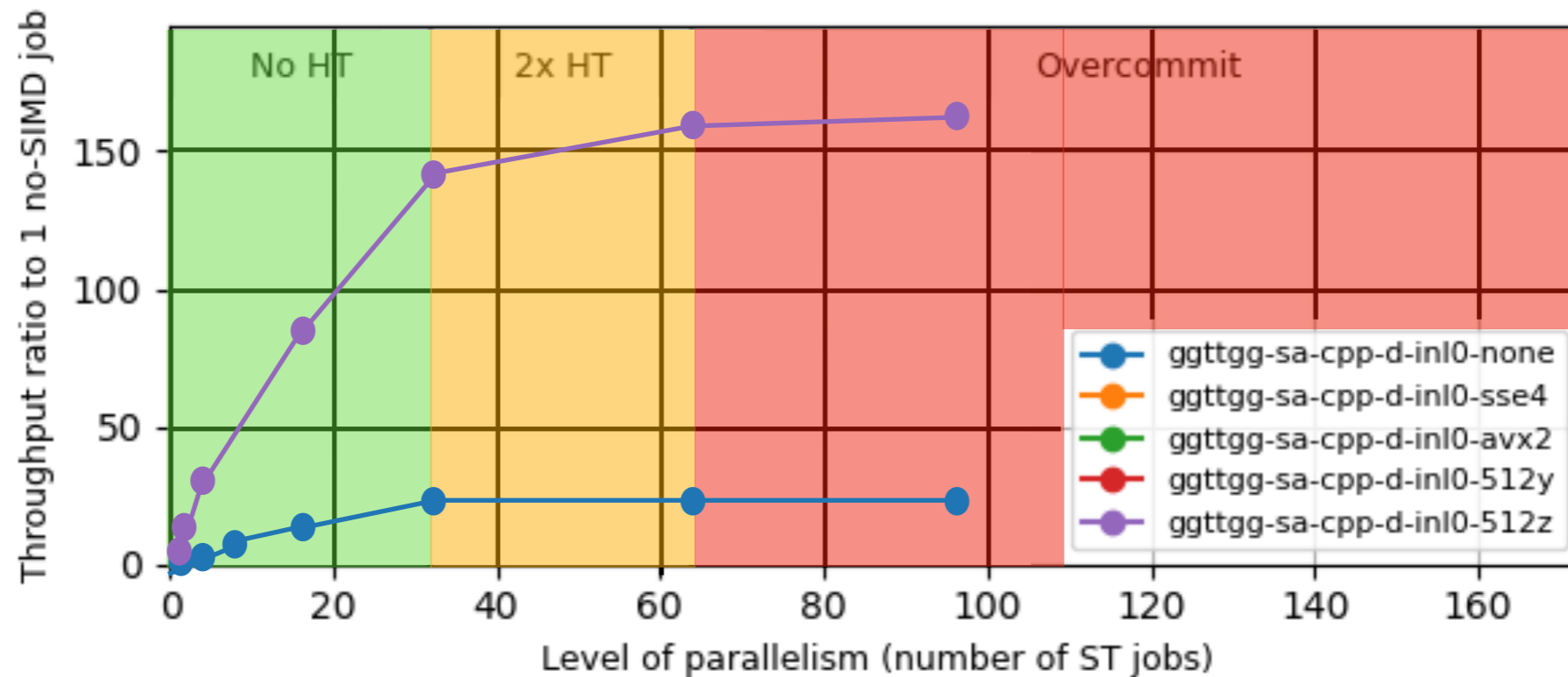
- Matrix-Element are naturally in lock step parralelism
- Perfect for SIMD (and GPU)

See Andrea Valassi talk for more details



# SIMD

Core 2.1GHz Xeon Gold 6130 with 2x HT for 10 cycles



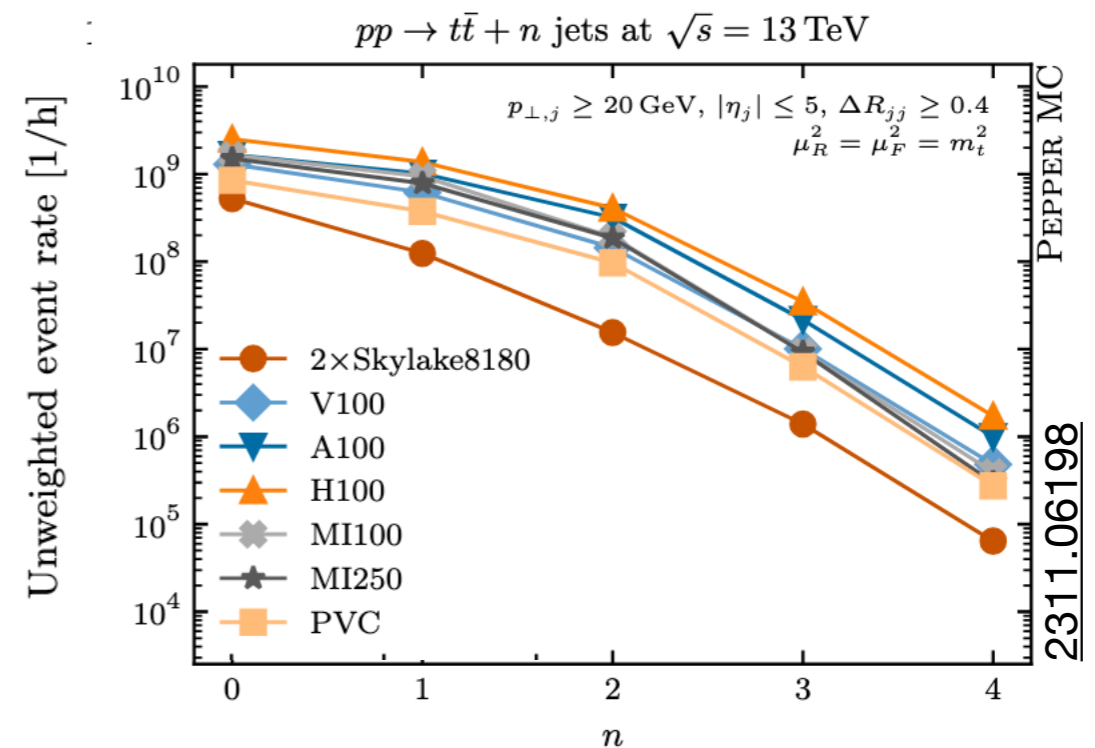
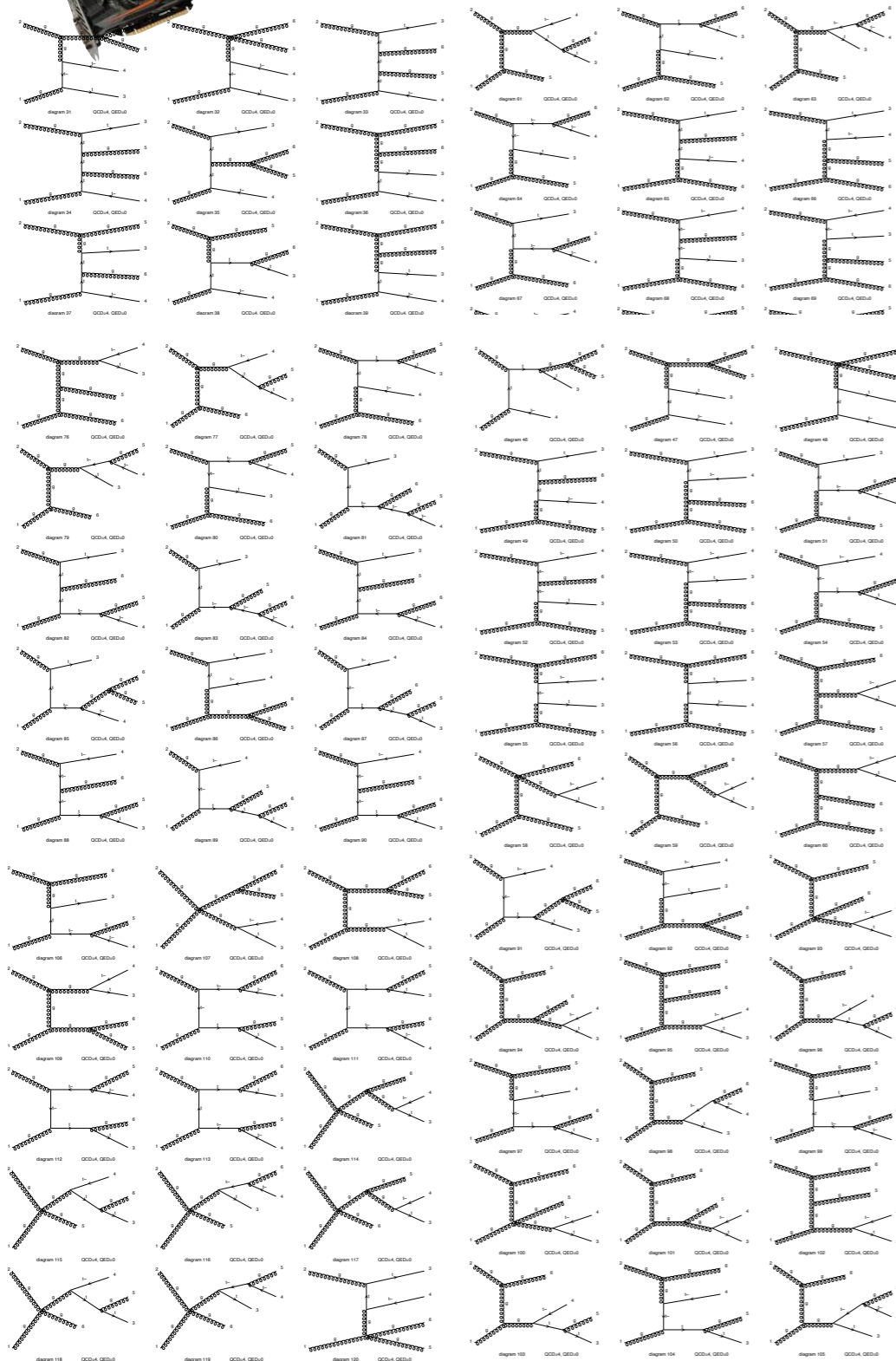
- Matrix-Element are naturally in lock step parralelism
- Perfect for SIMD (and GPU)
- Near to perfect speed-up (7x instead)

See Andrea Valassi talk for more details



# GPU

- Any LO code now capable to use GPU
- MadGraph and Pepper
- Large Speed-up reported
- 100 times faster
- NLO port in progress



See Andrea Valassi and Zenny Wettersten talk for more details

# Conclusion

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- Theoretician would like more and more FLOPS
- Experiments will need more events/accuracy/complexity
- FLOPS budget will be under tension in 3 years
  
- The IT community is helping towards more efficiency
- The theory community is starting to acknowledge effort for better hardware/software.
  
- Remember that we can not rely on a single tool.