

# Polarised boson tagging at the LHC with machine-learning methods

## Polarized Perspectives: Tagging and Learning in the SM

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CERN QTI



QUANTUM  
TECHNOLOGY  
INITIATIVE

# Amplitude-assisted tagging of longitudinally polarised bosons using wide neural networks

## Paper:

*Grossi, Incudini, Pellen, Pelliccioli* - **Eur. Phys. J. C (2023) 83:759**

**Slides:** revisited material from my co-authors Mathieu Pellen and Giovanni Pelliccioli

**THANKS**



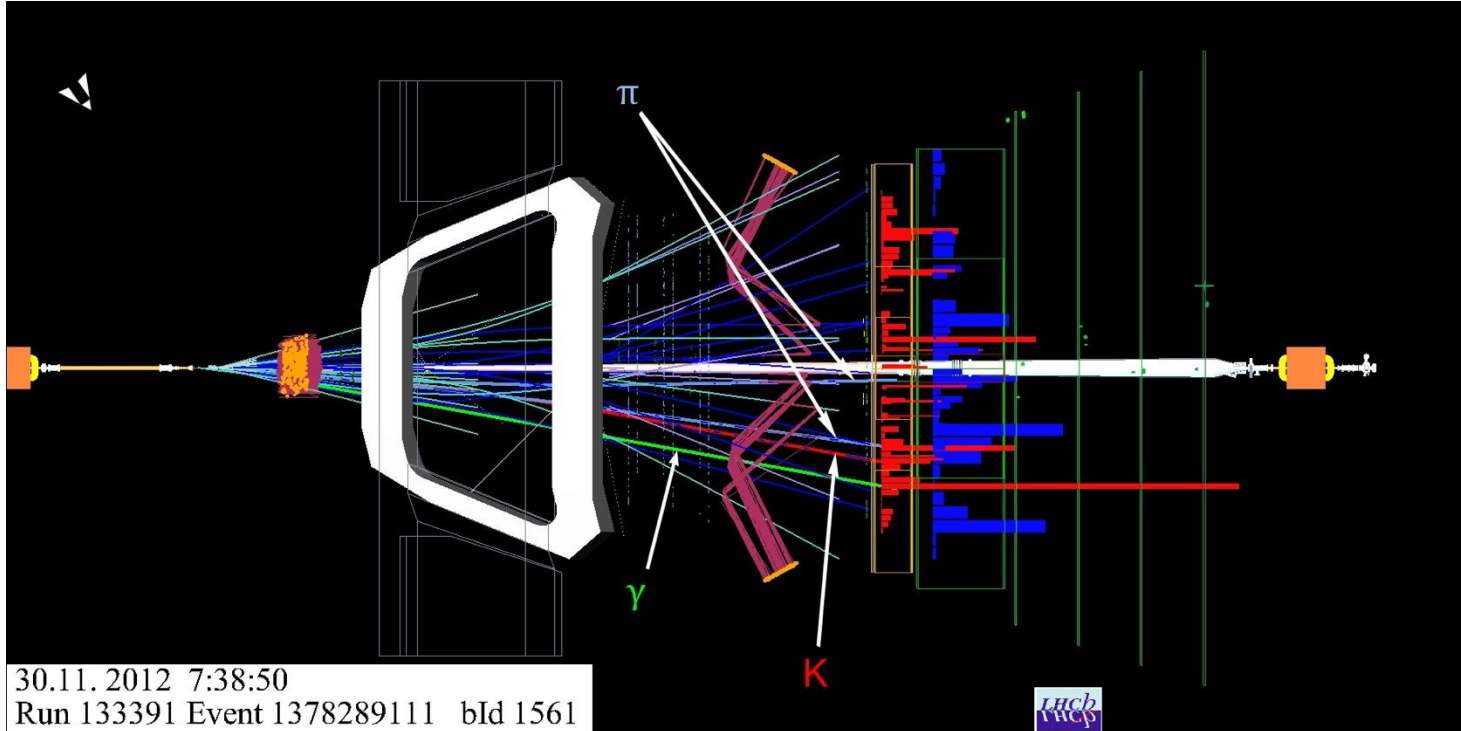
- **Problem Definition**
- **ML & Polarization Tagging**
- **Results**
- **Summary & Outlook**





- **Problem Definition**
- **ML & Polarization Tagging**
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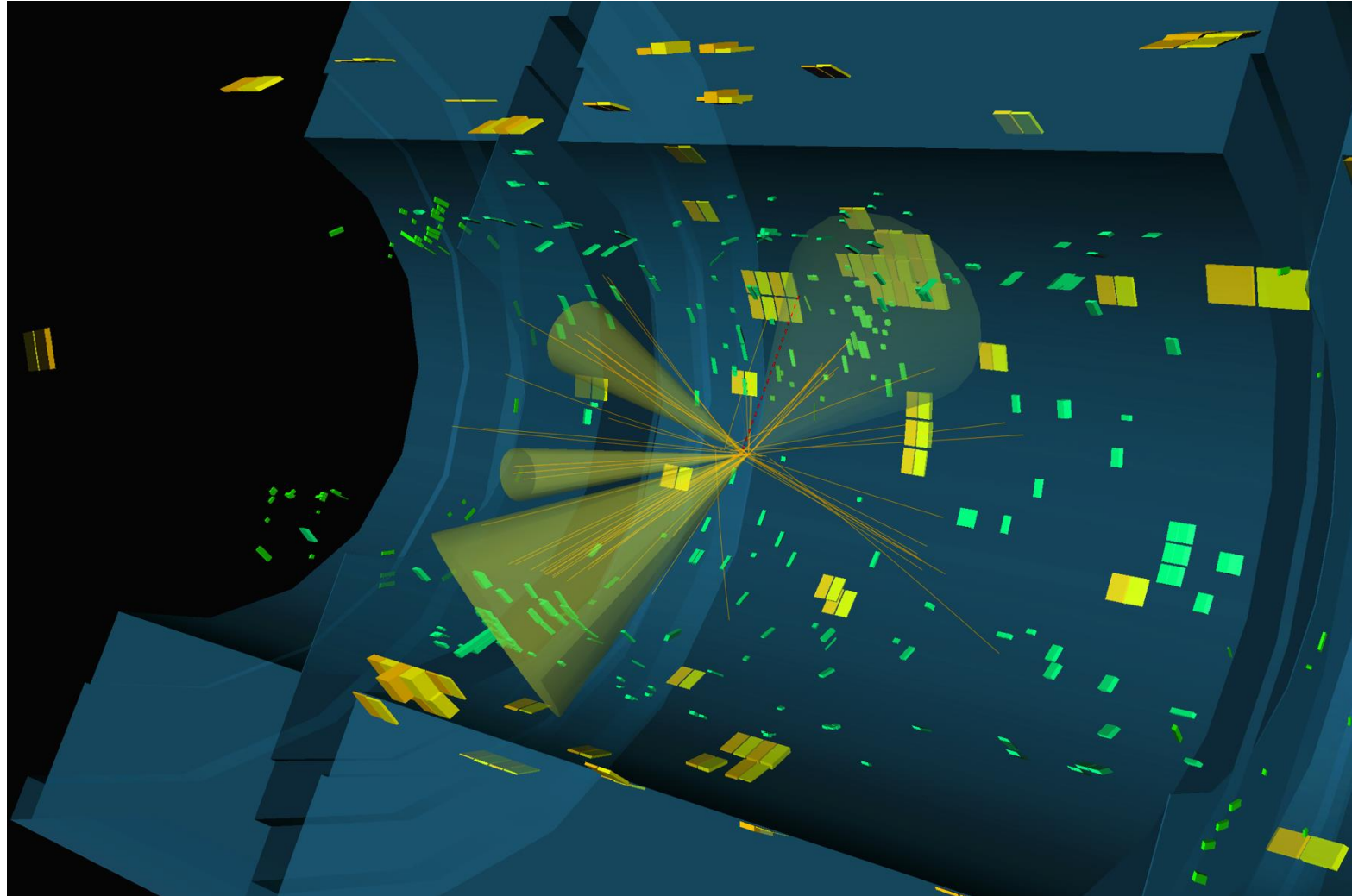


Can we design an original method to extract polarisation?

Maybe...

Can we design an original method to extract polarisation fractions using the maximal information encoded in the **amplitude**?

**YES**



# Polarization of weak boson

- Important probes of SM gauge and (extended) Higgs sectors
- Polarisation of gauge bosons related to Electroweak symmetry breaking
- The cross-section and angular distribution of longitudinal polarizations are particularly sensitive to beyond standard model (BSM) physics
- Special interest in di-boson (inclusive, VBS, Higgs decays)

*The theoretical study and the experimental extraction of such pseudo-observables is thus of prime importance for the present and upcoming physics programme of the LHC*

# Math – Polarization of weak boson

$$\mathcal{M} = \mathcal{M}_\mu^{\mathcal{P}} \frac{i}{k^2 - M_V^2 + i\Gamma_V M_V} \left( -g^{\mu\nu} + \frac{k^\mu k^\nu}{M_V^2} \right) \mathcal{M}_\nu^{\mathcal{D}}$$

↑ PROD      ↑ DEC

$$\sum_{\lambda=1}^4 \varepsilon_\lambda^\mu(k) \varepsilon_\lambda^{\nu*}(k)$$

polarisation vectors of the massive gauge boson (not Lorentz invariant: defined in a specific frame)

generic (unpolarised) amplitude featuring a resonant gauge boson decaying into a lepton-neutrino pair

...the unpolarised cross section (measured experimentally)

$$|\mathcal{M}|^2 = |\mathcal{M}_L|^2 + |\mathcal{M}_T|^2 + 2 \operatorname{Re}(\mathcal{M}_L^* \mathcal{M}_+) + 2 \operatorname{Re}(\mathcal{M}_L^* \mathcal{M}_-)$$

POL. AMP.      interf.

Polarisation state of an unstable particle is not directly accessible in the detectors and the information about it can only be reconstructed (in a probabilistic way) from the stable decay products → pseudo-observable ('quantum')



# Polarization of weak boson

angular-coefficient extraction

We cannot directly measure polarisations of EW bosons. But...

Decay-product angular distributions reflect polarisation state of the decayed V boson

Ex: differential cross-section with no lepton cuts for  $W \rightarrow l\nu$  is:

$$\frac{1}{\sigma} \frac{d\sigma}{d\cos\theta}(W^\pm \rightarrow l^\pm \nu) = \frac{3}{4} f_0 \sin^2 \theta + \frac{3}{8} f_R (1 \pm \cos \theta)^2 + \frac{3}{8} f_L (1 \mp \cos \theta)^2$$

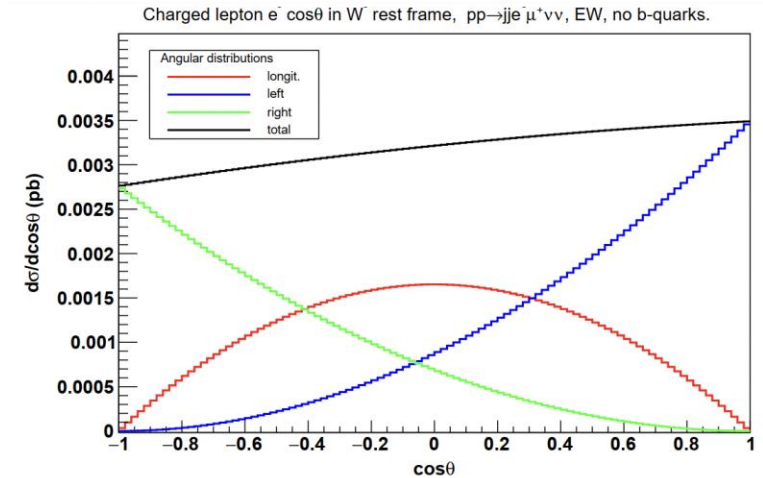
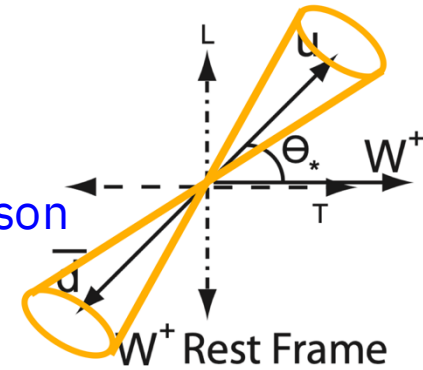
where  $f_0, f_R, f_L$  are W polarisation fractions

$\theta$  is the lepton polar angle in the W rest frame wrt  $W$  direction in the lab frame

Cross section (polarization fractions) can be written in terms of Legendre Polynomials of decay products

→ Boson polarisation can be measured as angular distributions of particles produced in the decay process from unpolarized VBS process

Decay of W



[A. Ballestrero, E. Maina, and G. Pelliccioli; W boson polarization in vector boson scattering at the LHC; JHEP 03 (2018)]

Grossi et al. 10.1140/epjc/s10052-020-08713-1



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# Machine Learning – not a newcomer

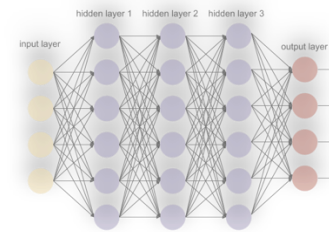
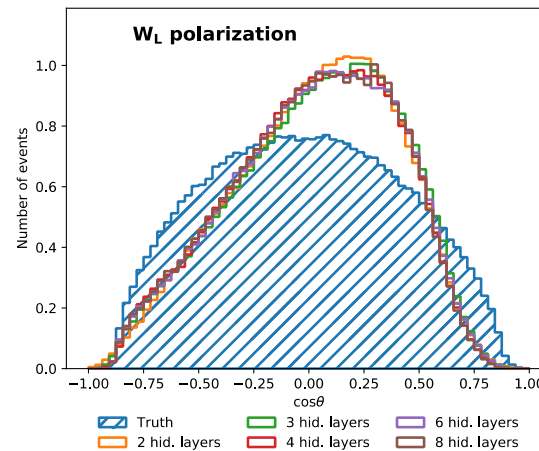
We cannot directly measure polarisations of EW bosons. But...

- Various **neural-network** strategies used for **polarisation extraction** from leptonic decays in **VBS** events (lower BR, neutrino reco) [Searcy et al. 1510.01691, Lee et al. 1812.07591, 1908.05196, Grossi et al. 2008.05316, Li et al. 2010.13281, 2109.09924]
- Machine-learning (ML) approaches also used to extract polarisations from **hadronic decays** (extract correct boson jet, QCD effect) [Kim Martin 2102.05124].

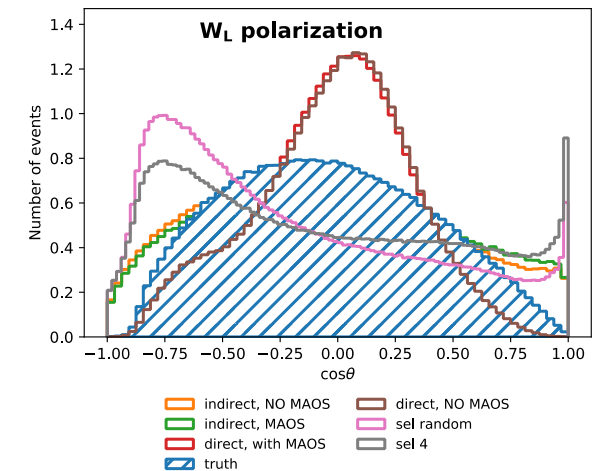
Mostly use **kinematical features** to reconstruct **polarisation-sensitive angles** or directly extract **polarisation fractions** with DNN score.

The VBS case...  $\cos(\theta)$

- Semi-leptonic:  $pp \rightarrow j\mu^+\nu_\mu$



- Full-leptonic:  $pp \rightarrow jj e^+ \nu_e \mu^+ \nu_\mu$



Grossi et al. 10.1140/epjc/s10052-020-08713-1

# Machine Learning – not a newcomer

- We can perform fits of LHC data with polarised templates

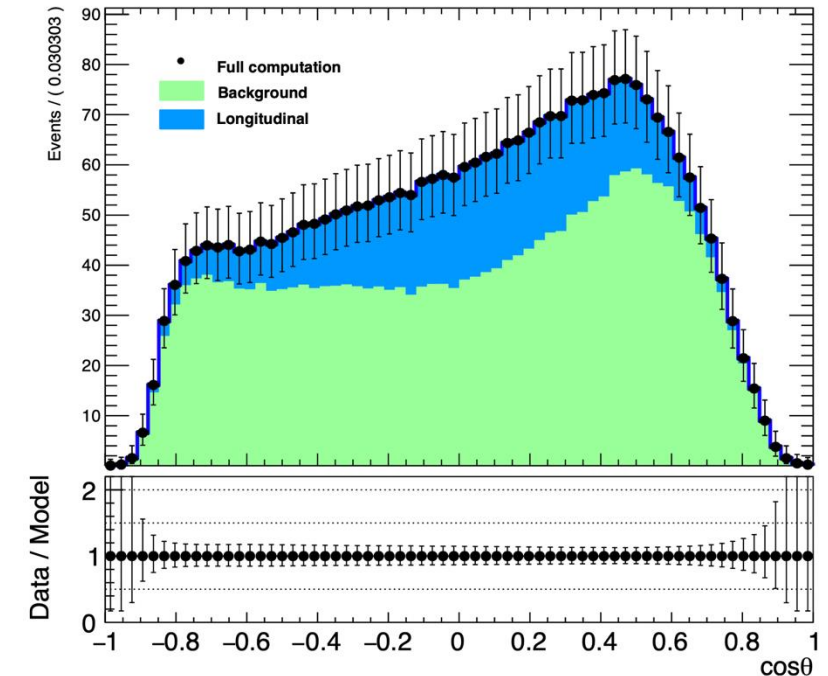
→ extraction of parameters based on **theory input**

## Open questions

- Should the two signatures be fitted separately or together?
  - How should theoretical uncertainties be taken into account in the fit?
  - How should one define the overall uncertainty on the fit of the polarisation fractions?
  - What are the optimal observables?
- Answers in collaboration with experimental collaborations

## Drawbacks

- Restricted to given observables
- Interpretation at the integrated level



Grossi et al. [10.1140/epjc/s10052-020-08713-1](https://arxiv.org/abs/10.1140/epjc/s10052-020-08713-1)

# Polarization tagging

New idea for polarisation extraction:

*'Amplitude-assisted tagging of longitudinally polarised bosons using wide neural networks'*

use **AMPLITUDE** to extract parameters for the theory



- no restricted to given observables (no obs)
- event-by-event interpretation (fully differential)
- can be applied to any physics problem that can be cast in one single, bounded, ratio ( next)

Not to be confused with  
(Similar/connected ideas):

- matrix-element method  
[Kondo; J. Phys. Soc. Jap. 57 (1988) 4126-4140 / 60 (1991) 836-844.]
- optimal-observable method  
[Diehl, Nachtmann; Z. Phys. C 62 (1994) 397-412, hep-ph/9603207], [Janot; 1503.01325]
- **MELA** (Matrix Element Likelihood Approach)  
[Gao, Gritsan, Melnikov, Schulze, et al.; 1001.3396, 1208.4018, 1309.4819, 1606.03107]

# Polarization tagging

How to tag an unpolarised event as longitudinal.

$$f_L(\mathcal{O}) = \frac{d\sigma_L}{d\mathcal{O}} / \frac{d\sigma_{unp}}{d\mathcal{O}}$$

At the **event-by-event**/phase-space-point level, at leading order is equivalent to  $r_L = \frac{|M_L|^2}{|M|^2}$

longitudinal

unpolarized

$r_L$  would be the optimal observable

BUT, we don't measure it

TODO: Unpolarised events  $\rightarrow$  reweight/sample according to truth  $r_L \rightarrow$  longitudinal events (*the probability for an event to be longitudinally polarised*)

# Polarization tagging - using wide neural networks

- $r_L$  requires knowledge of all momenta (initial and final), partonic process and the PDF (+ PS)
- experimentally we have access to **partial information** about the unpolarised process (final-state only, no flavour structure).

➡ bypass this lack of information by using a neural network (NN)

# Polarization tagging – setting

- Z+j production at the LHC at  $\sqrt{s} = 13.6$  TeV
- Use MADGRAPH5\_AMC@NLO for checks  
[Alwall, et al.; 1405.0301], [Buarque Franzosi, Mattelaer, Ruiz, Shil; 1912.01725]
- Use RECOLA [Actis et al.; 1605.01090] for  $r_L$  computation
- Use PYTHIA [Sjöstrand et al.; 1410.3012] for PS

→ *Generation set-up*

$$p_{T,j} > 10 \text{ GeV}, \quad |y_j| < 5, \quad \text{and} \quad 76 \text{ GeV} < M_{\mu^+\mu^-} < 106 \text{ GeV}$$

→ *Inclusive set-up*

$$p_{T,j} > 20 \text{ GeV}, \quad |y_j| < 4, \quad \text{and} \quad 81 \text{ GeV} < M_{\mu^+\mu^-} < 101 \text{ GeV}$$

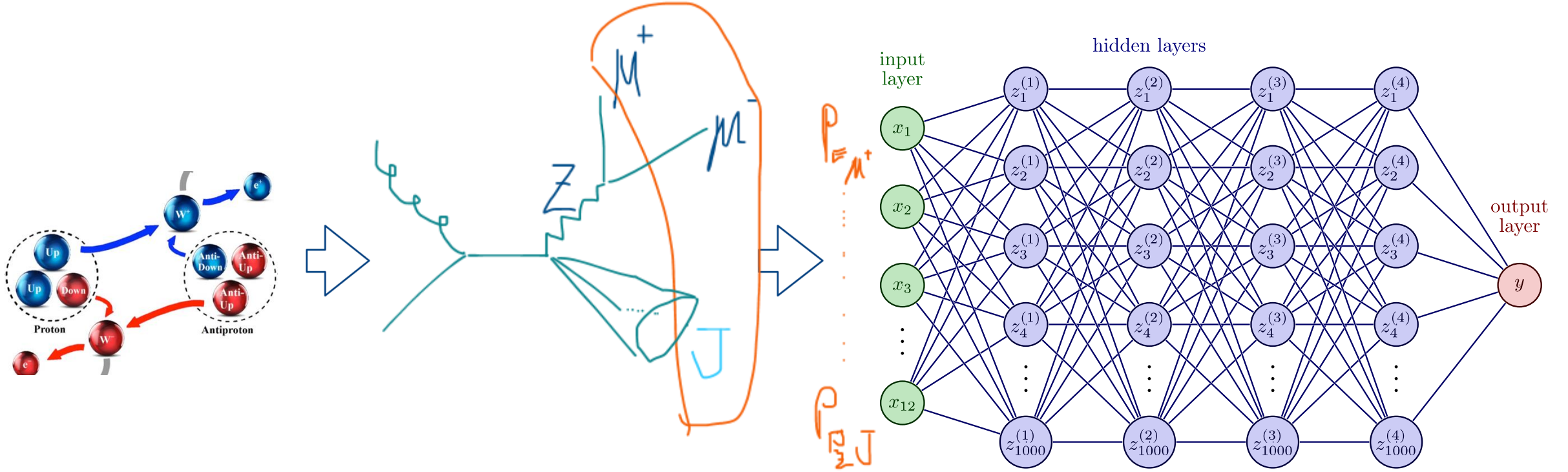
☞ No cuts on Z-boson decay products

→ *Fiducial set-up*

$$p_{T,\mu^\pm} > 20 \text{ GeV} \quad \text{and} \quad |y_{\mu^\pm}| < 2.7$$



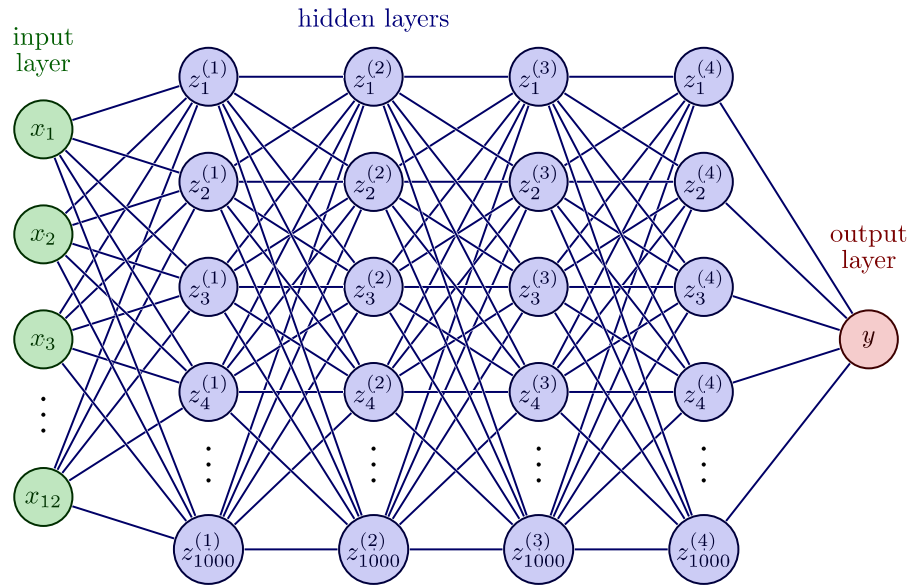
# Polarization tagging - using wide neural networks



$$p_{\mu^+}, p_{\mu^-}, p_j; r_L$$

$$p_{\mu^+}, p_{\mu^-}, p_j; \tilde{r}_L$$

# Polarization tagging - using wide neural networks



*PyTorch library and RMSprop algorithm  
for gradient-descent optimisation*

## Physics DETAILS

- NN for fixed-order events
- NN for parton-showered events
- training in inclusive setup, testing in inclusive and fiducial one.

## NN DETAILS

- Optimal-model width-to-depth ratio found to be 1000 : 5  $\rightarrow$  the NN is rather wide.

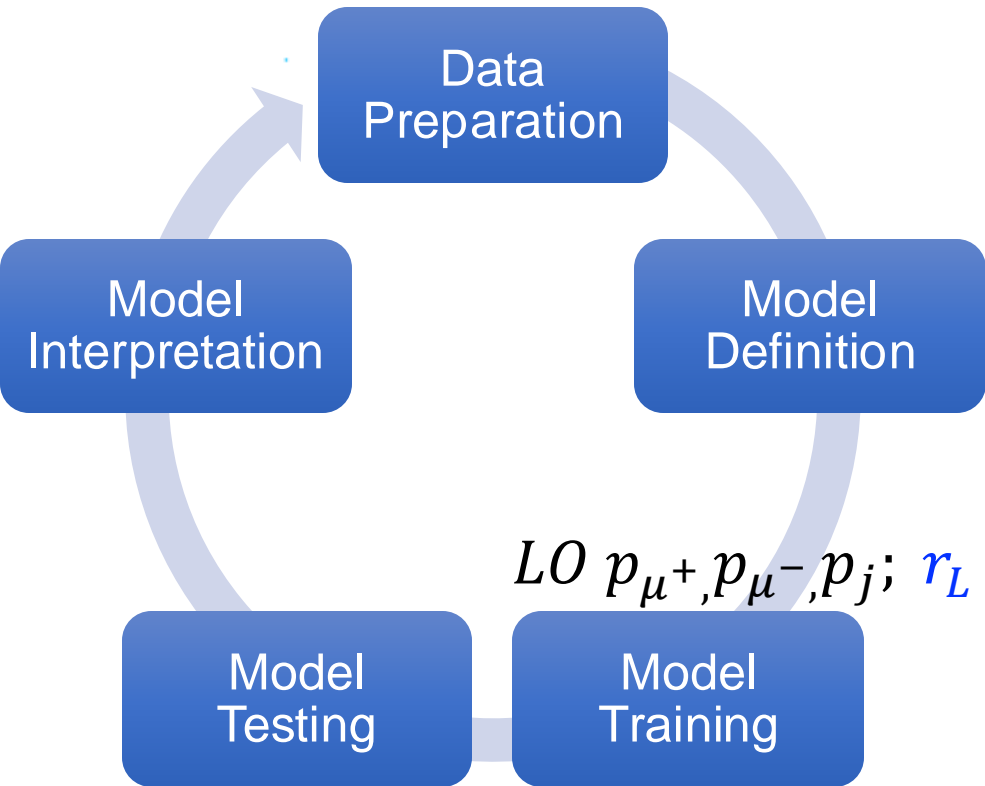
*shallow and wide are easier to train (no chaotic dynamics), but potential lost of non-linear expressivity (i.e. linear model).*



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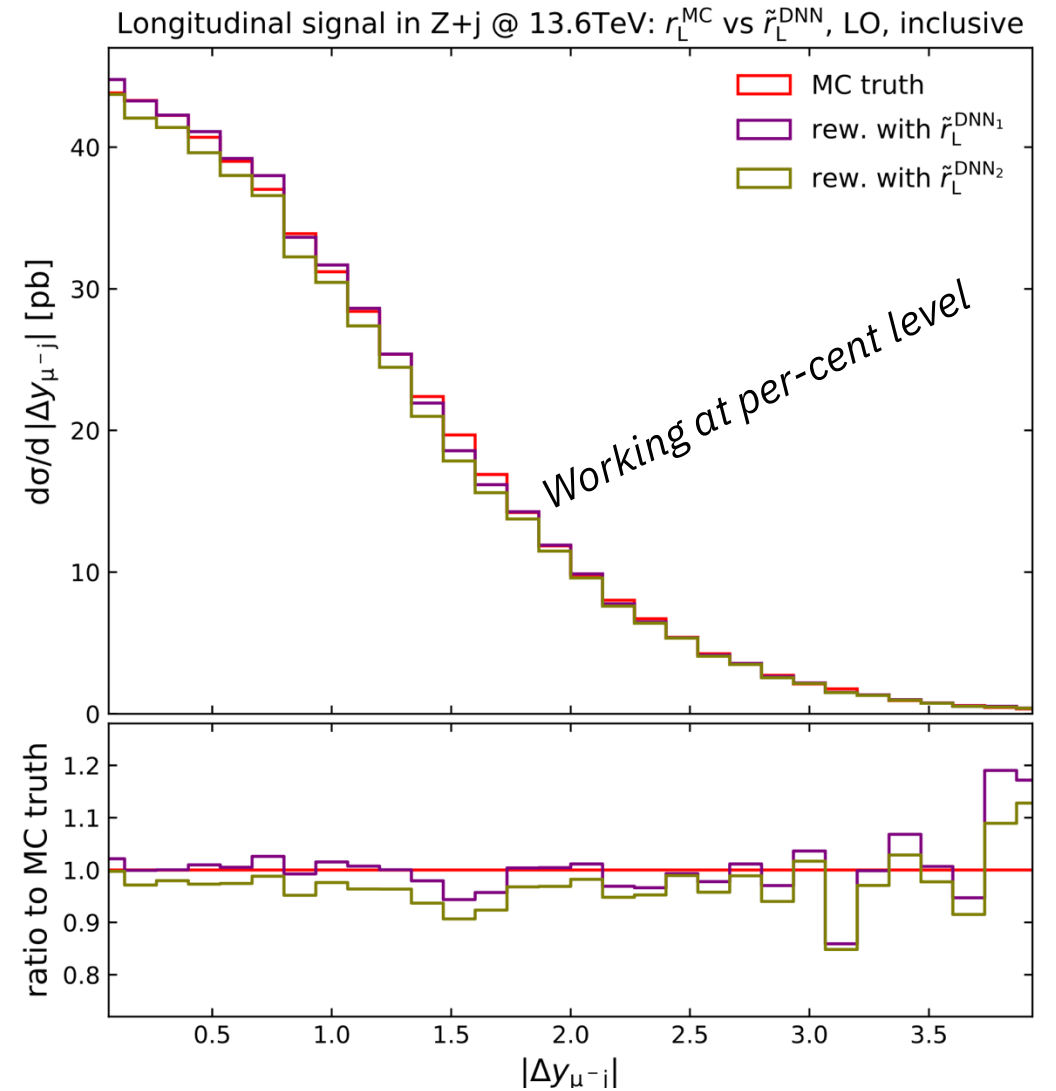


# Polarization tagging - LO



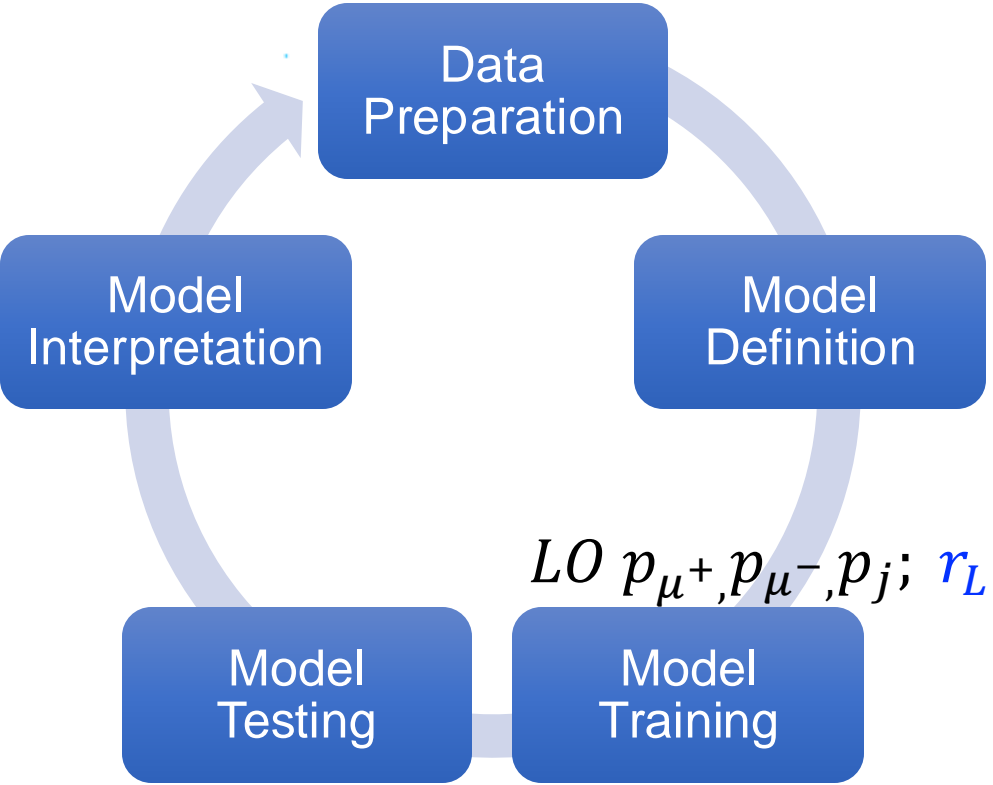
$LO\ p_{\mu^+}, p_{\mu^-}, p_j; r_L$

$LO\ p_{\mu^+}, p_{\mu^-}, p_j; \tilde{r}_L$



NB: the NN-training stage is tailored to the specific choice of polarisation frame.

# Polarization tagging LO + PS

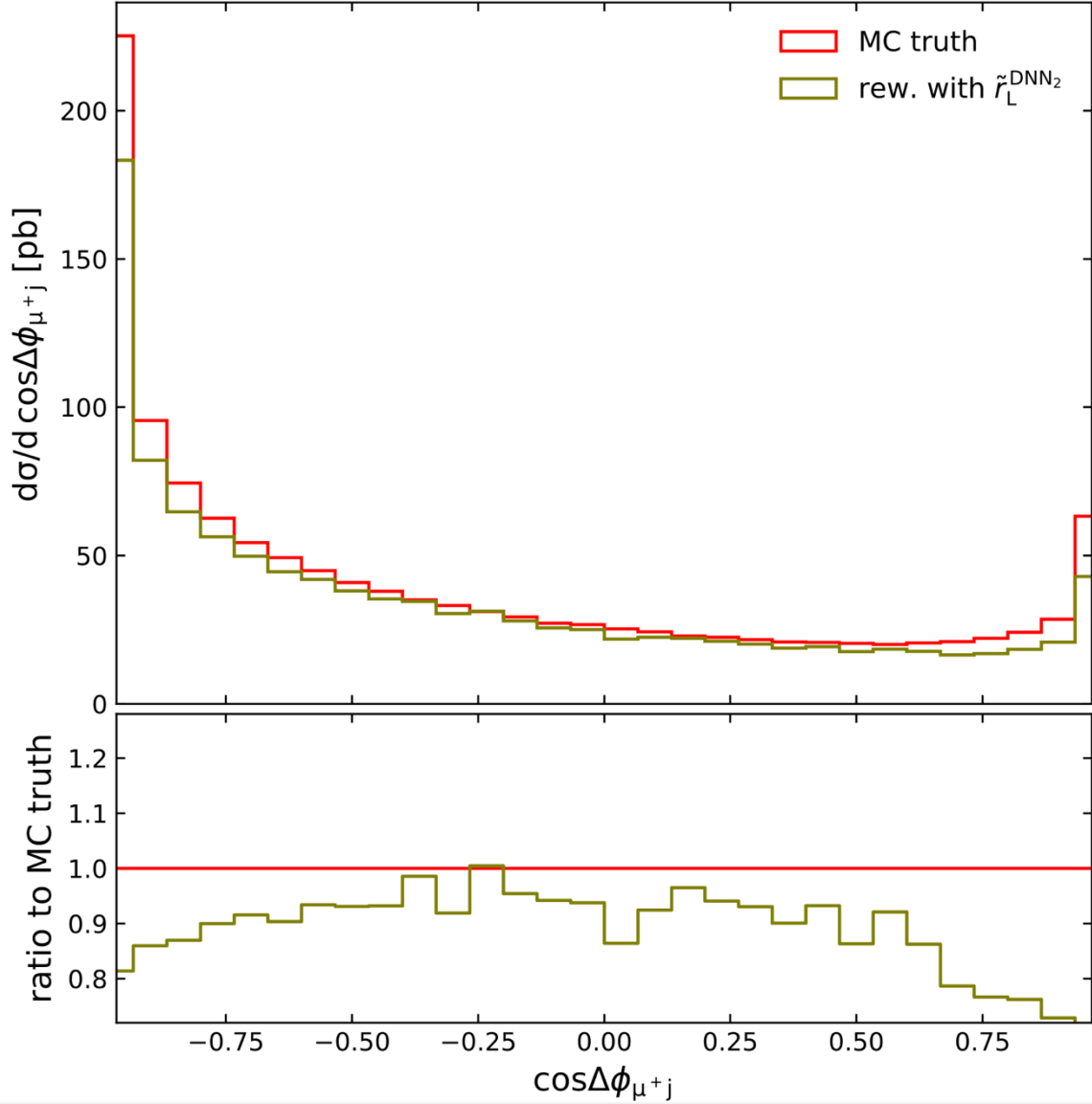


*LO*  $p_{\mu^+}, p_{\mu^-}, p_j; r_L$

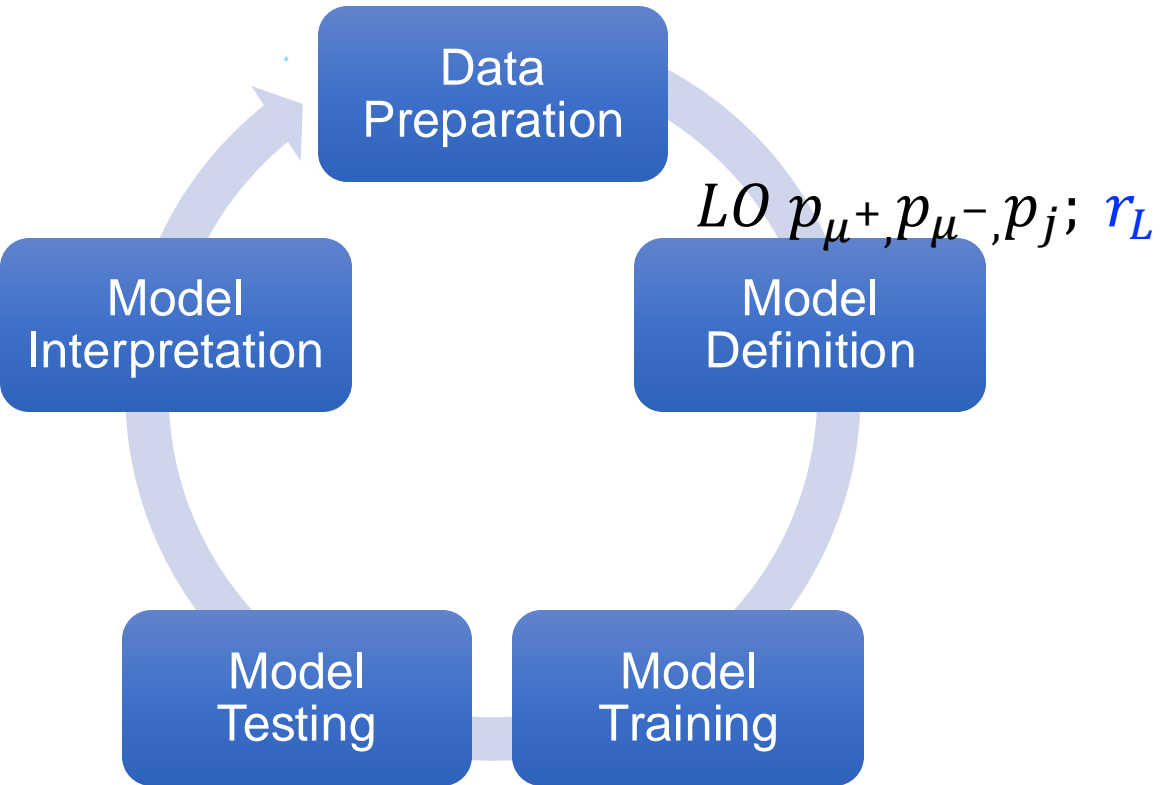
*LO + PS*  $p_{\mu^+}, p_{\mu^-}, p_j; \tilde{r}_L$

Failing! (describing PS corrections instead of polarisation) → Retraining with  $r_L$  computed before PS!

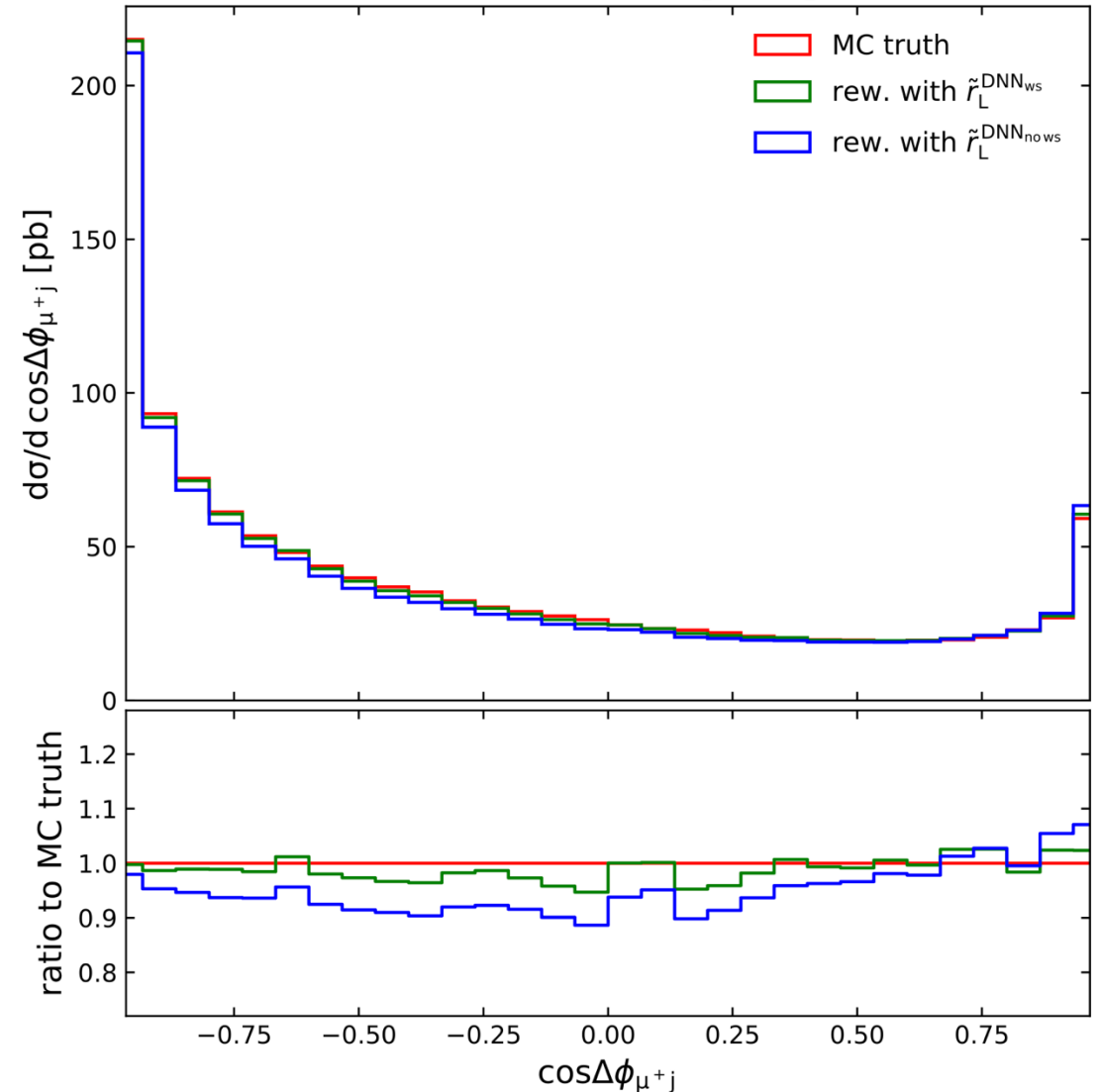
Longitudinal signal in Z+j @ 13.6TeV:  $r_L^{MC}$  vs  $\tilde{r}_L^{DNN}$ , LOPS, inclusive



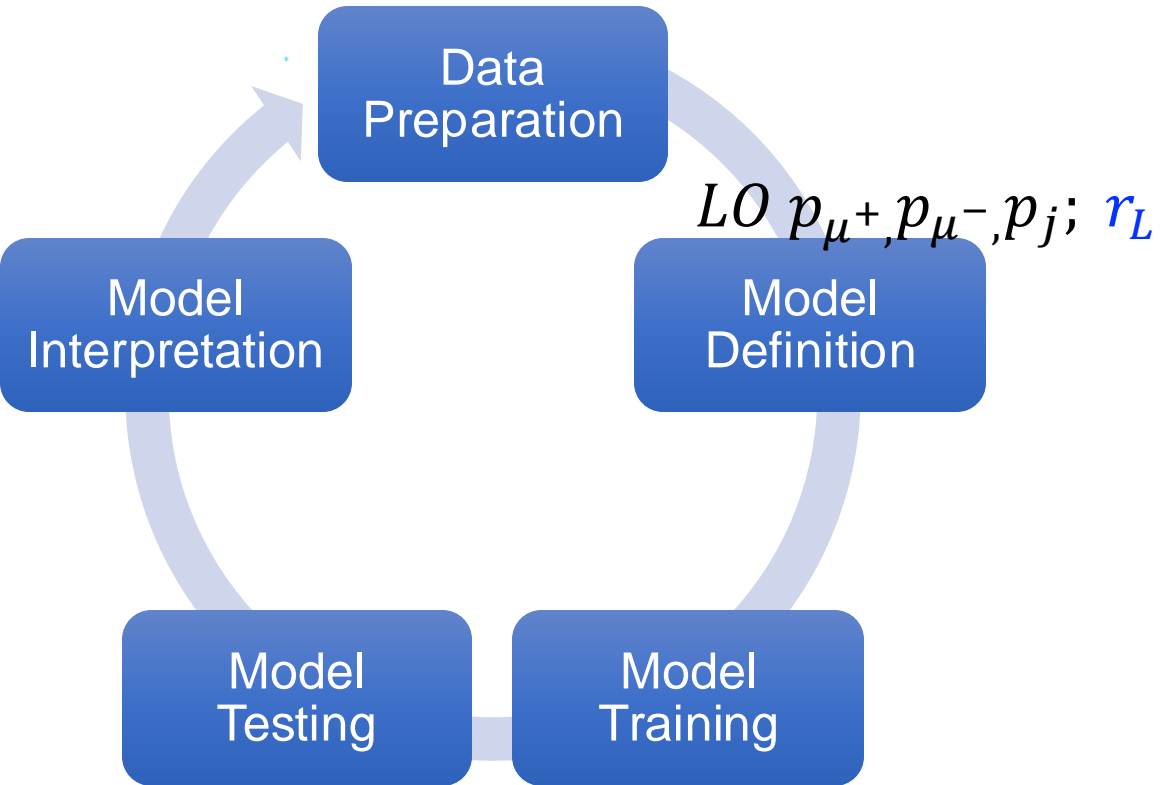
# Polarization tagging – Warm-start



Longitudinal signal in Z+j @ 13.6TeV:  $r_L^{MC}$  vs  $\tilde{r}_L^{DNN}$ , LOPS, inclusive



# Polarization tagging – Warm-start



- Warmup training gives better results
- Initial conditions of the LO+PS learning is set by LO learning
- **Satisfactory performance** for integrated fractions (1% level), angular observables and bulk of  $p_T$  ones (dedicated training needed in least-populated phase-space regions).
- LO+PS is better reproduced than LO (less  $r_L \sim < 0$ )

$LO + PS p_{\mu^+}, p_{\mu^-}, p_j; \tilde{r}_L$



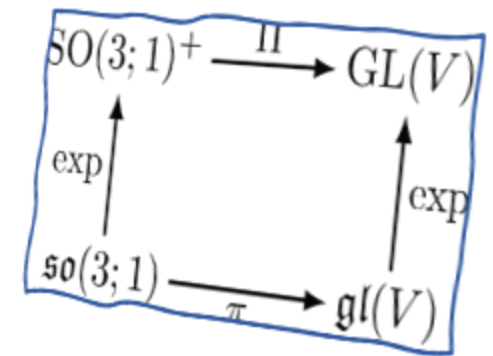
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# Discussion and Outlook

- Propose an original method to extract polarisation fractions using the amplitude
- The method can be applied to any physics problem that can be cast in one single, bounded, ratio (in the present application  $rL$ ) which can be approximately reconstructed using incomplete information thanks to NN methods
- Applicable at different energy?
- Higher orders
  - Possible reweighting with virtual and real amplitudes
- Use physics-informed approaches, enforcing symmetries conservation?  
BUT there is no guarantees



# Observation of quantum entanglement in top quark pair production in proton-proton collisions at $\sqrt{s} = 13$ TeV

CMS Collaboration  
6 June 2024

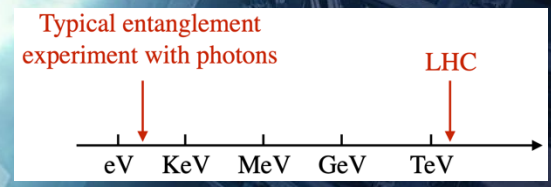
Submitted to *Reports on Progress in Physics*

**Abstract:** Entanglement is an intrinsic property of quantum mechanics and is predicted to be exhibited in the particles produced at the Large Hadron Collider. A measurement of the extent of entanglement in top quark-antiquark ( $t\bar{t}$ ) events produced in proton-proton collisions at a center-of-mass energy of 13 TeV is performed with the data recorded by the CMS experiment at the CERN LHC in 2016, and corresponding to an integrated luminosity of  $36.3 \text{ fb}^{-1}$ . The events are selected based on the presence of two leptons with opposite charges and high transverse momentum. An entanglement-sensitive observable  $D$  is derived from the top quark spin-dependent parts of the  $t\bar{t}$  production density matrix and measured in the region of the  $t\bar{t}$  production threshold. Values of  $D < -1/3$  are evidence of entanglement and  $D$  is observed (expected) to be  $-0.480^{+0.026}_{-0.029}$  ( $-0.467^{+0.026}_{-0.029}$ ) at the parton level. With an observed significance of 5.1 standard deviations with respect to the non-entangled hypothesis, this provides observation of quantum mechanical entanglement within  $t\bar{t}$  pairs in this phase space. This measurement provides a new probe of quantum mechanics at the highest energies ever produced.

# Observation of quantum entanglement in top-quark pairs using the ATLAS detector

ATLAS Collaboration

We report the highest-energy observation of entanglement, in top-antitop quark events produced at the Large Hadron Collider, using a proton-proton collision data set with a center-of-mass energy of  $\sqrt{s} = 13$  TeV and an integrated luminosity of  $140 \text{ fb}^{-1}$  recorded with the ATLAS experiment. Spin entanglement is detected from the measurement of a single observable  $D$ , inferred from the angle between the charged leptons in their parent top- and antitop-quark rest frames. The observable is measured in a narrow interval around the top-antitop quark production threshold, where the entanglement detection is expected to be significant. It is reported in a fiducial phase space defined with stable particles to minimize the uncertainties that stem from limitations of the Monte Carlo event generators and the parton shower model in modelling top-quark pair production. The entanglement marker is measured to be  $D = -0.547 \pm 0.002$  (stat.)  $\pm 0.021$  (syst.) for  $340 < m_{t\bar{t}} < 380$  GeV. The observed result is more than five standard deviations from a scenario without entanglement and constitutes both the first observation of entanglement in a pair of quarks and the highest-energy observation of entanglement to date.



The background is a dark blue gradient with several glowing blue lines and a network of nodes. The nodes are small blue dots connected by thin blue lines, forming a complex web-like structure. The lines are thicker and more prominent, curving across the frame. The overall aesthetic is futuristic and technological.

**CERN QTI**

**<https://quantum.cern/>**