

# A $W^\pm$ polarization analyzer from Deep Neural Networks

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arXiv:2102.05124

# Theoretical Motivation

## Massive vector boson final states

$$pp \rightarrow W^\pm W^\mp$$

$$pp \rightarrow W^\pm Z$$

$$pp \rightarrow ZZ$$

- Indirect approach of checking SM : polarization searches
  - Longitudinal vs. Transverse
- SM can predict polarization fraction
- Longitudinal polarization is sensitive to EWSB
- New physics can impact polarizations fractions. Ex.) new resonance, or higher dimensional operators (SMEFT)

# *W* Polarization Measurement

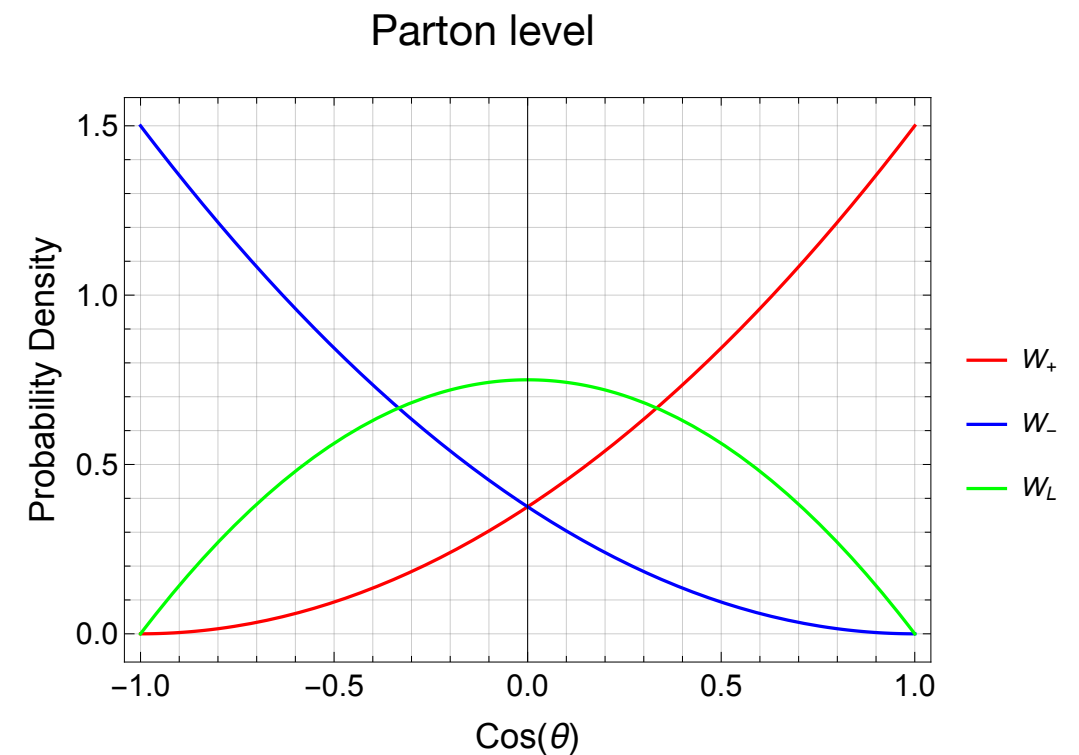
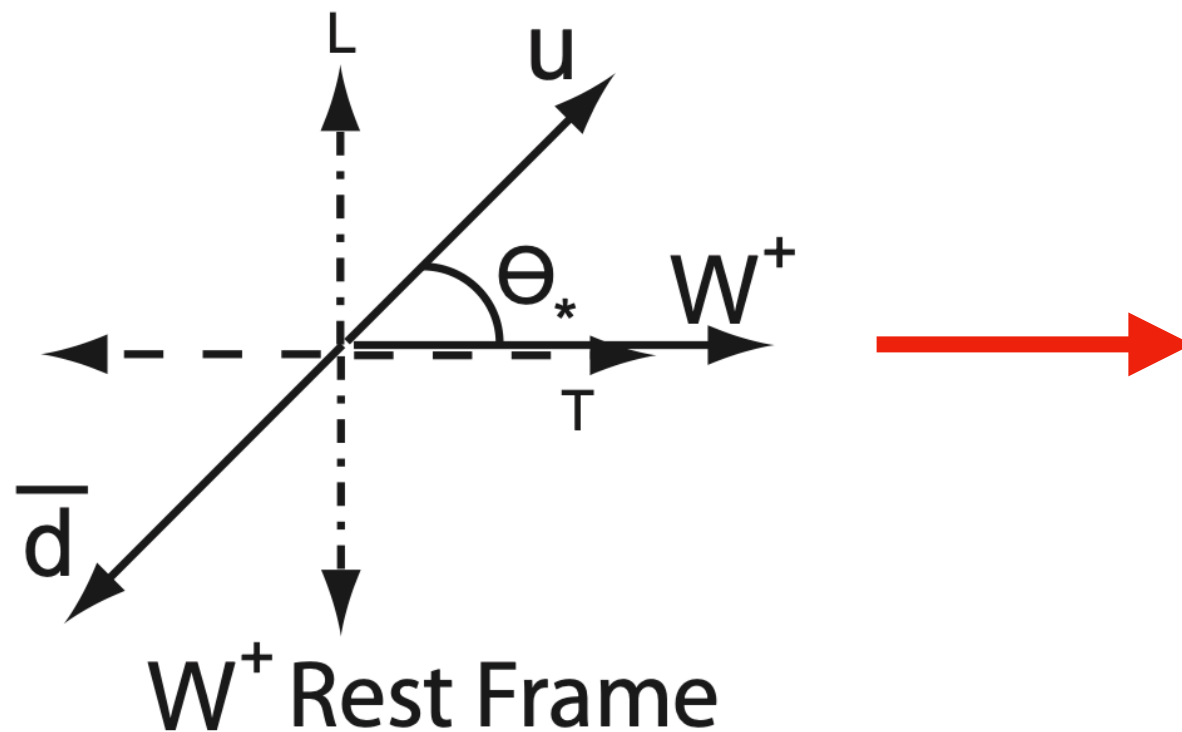
- Leptonic tagging is done
  - lower branching ratio than hadronic state
  - neutrino reconstruction
- Can we do hadronic *W* tagging?
  - Extract correct boson jet
  - QCD effect washes out parton level information
  - Possible jet substructure information can be used to tag polarization
    - N-subjettiness (S. De, V. Rentala, W. Shepherd arXiv:2008.04318v1)
  - **How does machine learning do?**

# W Polarization Measurement

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# W polarization

## Decay of W



- Since  $W$  only interacts to the left handed particles, each polarization has distinct angular distribution (or, in lab frame,  $(E_q - E_{\bar{q}})/|\vec{p}_W|$ )
- Parton level distribution can be used as a reference point for the network optimization
- Due to the deviation, it is possible to measure polarization fraction for diboson final states

# Higher dimensional operators and W/Z polarization

- SMEFT extends the SM Lagrangian by gauge invariant higher dim ( $D > 4$ ) operators

- $$\mathcal{L}_{SMEFT} = \mathcal{L}_{SM} + \sum_{D > 4}^{\text{inf}} \frac{1}{\Lambda^{D-4}} c_j^{(D)} \mathcal{O}_j^{(D)}$$

- $W^\pm/Z$  can appear in two ways:

1.  $W_{\mu\nu}^a \rightarrow$  Primarily transverse as  $\epsilon_L^\mu W_{\mu\nu}^a \sim \frac{k^\mu}{m_W} W_{\mu\nu}^a \approx 0$

2.  $D_\mu H \rightarrow$  Primarily longitudinal as Goldstone gets eaten

Relevant operators (SILH) for diboson final states

Da Liu, Lian-Tao Wang [arXiv: 1804.08688v1]

$$\mathcal{O}_W = \frac{ig}{2} \left( H^\dagger \sigma^a \overleftrightarrow{D}^\mu H \right) D^\nu W_{\mu\nu}^a$$

$$\mathcal{O}_B = \left( H^\dagger \sigma^a \overleftrightarrow{D}^\mu H \right) \partial^\nu B_{\mu\nu}$$

$$\mathcal{O}_{2W} = -\frac{1}{2} D^\mu W_{\mu\nu}^a D_\rho W^{a\rho\nu}$$

$$\mathcal{O}_{3W} = \frac{1}{3!} g \epsilon_{abc} W_\mu^{a\nu} W_{\nu\rho}^b W^{c\rho\mu}$$

We will focus on the boxed operators

$$\mathcal{O}_{HW} = ig (D^\mu H)^\dagger \sigma^a (D^\nu H) W_{\mu\nu}^a$$

$$\mathcal{O}_{HB} = ig' (D^\mu H)^\dagger (D^\nu H) B_{\mu\nu}$$

# Higher dimensional operators and W/Z polarization

$$\mathcal{O}_W = \frac{ig}{2} \left( H^\dagger \sigma^a \overleftrightarrow{D}^\mu H \right) D^\nu W_{\mu\nu}^a$$

$$\mathcal{O}_{3W} = \frac{1}{3!} g \epsilon_{abc} W_\mu^{a\nu} W_{\nu\rho}^b W^{c\rho\mu}$$

- Operators above contribute to the cross section of  $pp \rightarrow W^\pm Z$  but different impact on polarization breakdown
- Knowing polarization can better distinguish between effects
- For our analysis, we do not introduce full SMEFT operators for simplification of parameters

- Benchmark SMEFT Lagrangian :  $\mathcal{L}_{SMEFT} = \mathcal{L}_{SM} + \frac{2c_W}{m_W^2} \mathcal{O}_W + \frac{3!c_{3W}g^2}{m_W^2} \mathcal{O}_{3W}$

Convention from A. Alloul, B. Fuks, and V. Sanz, arXiv:1310.5150

- We will later predict longitudinal content of  $pp \rightarrow W^\pm(jj)Z(ll)$  for both SM and SMEFT cases

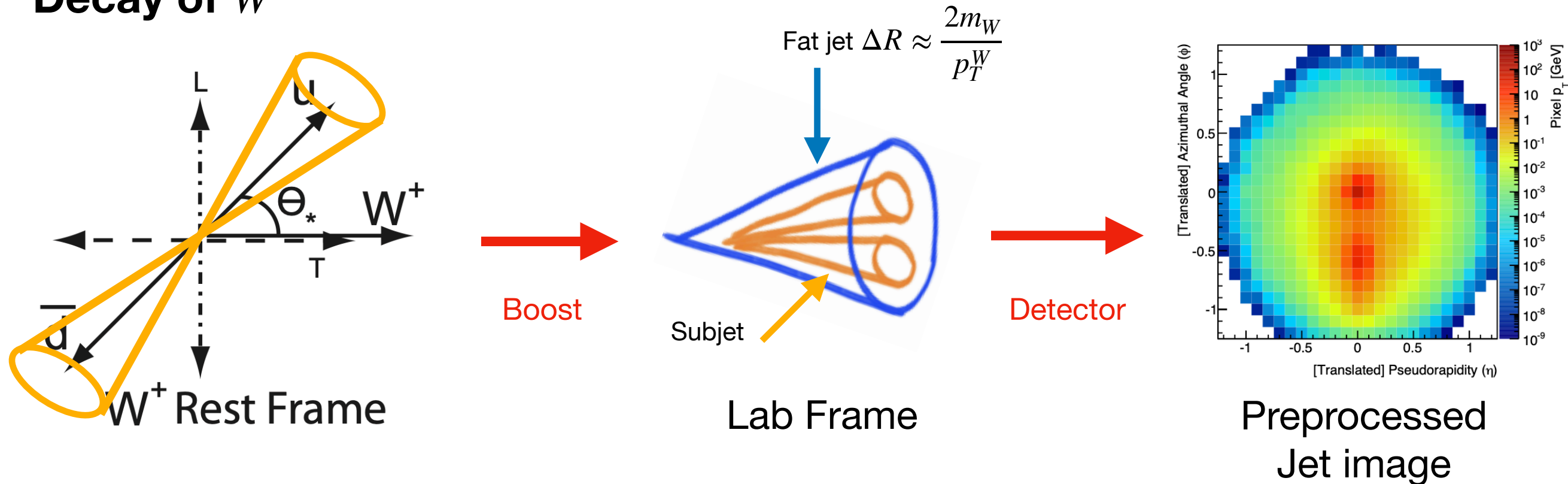
# W Polarization Measurement

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# First ingredient: Boosted $W$ Jet

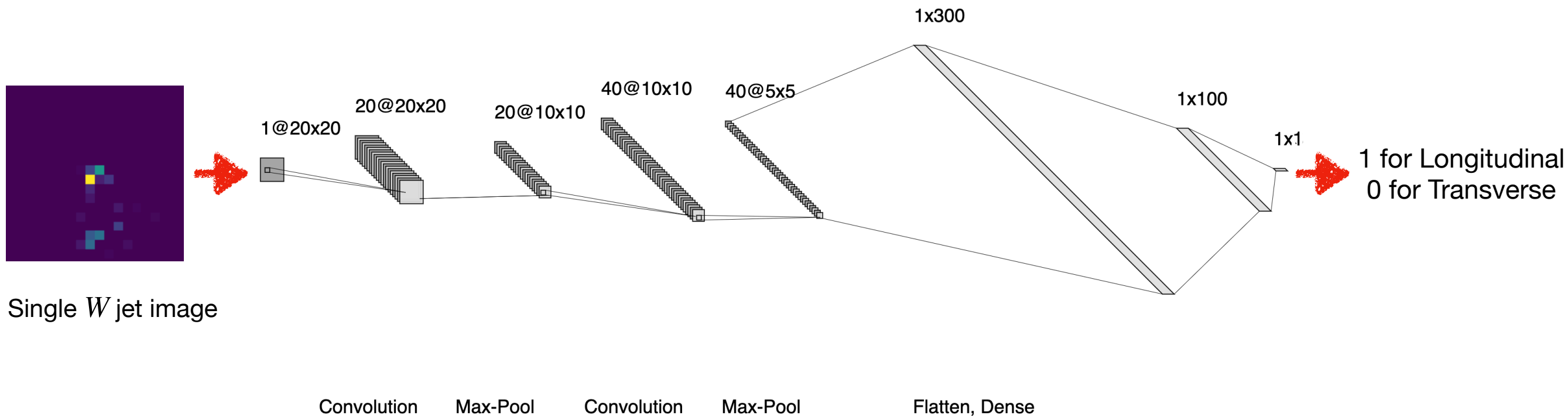
## Decay of $W$



- Quark becomes QCD jet
- Due to the boost, collimation of the jet deduces the angular distribution signature
- After boost  $\theta^* \rightarrow$  opening angle (sensitive to  $p_T$ )
- At extreme high  $p_T^W$ , subjet signature can disappear
- Particles are plotted on pixelized  $\eta - \phi$  plane and their color is determined from  $p_T$

# Convolutional Neural Network (CNN)

## Image classification



- The network is trained with simulated events (**MadGraph** + **Pythia** + **Delphes**) of boosted longitudinal and transverse  $W$ 's respectively for tagging purposes

Training Sample

$$\begin{aligned} \text{Longitudinal : } & pp \rightarrow \phi \rightarrow W^\pm W^\mp \\ \text{Transverse : } & pp \rightarrow W^\pm j \end{aligned}$$

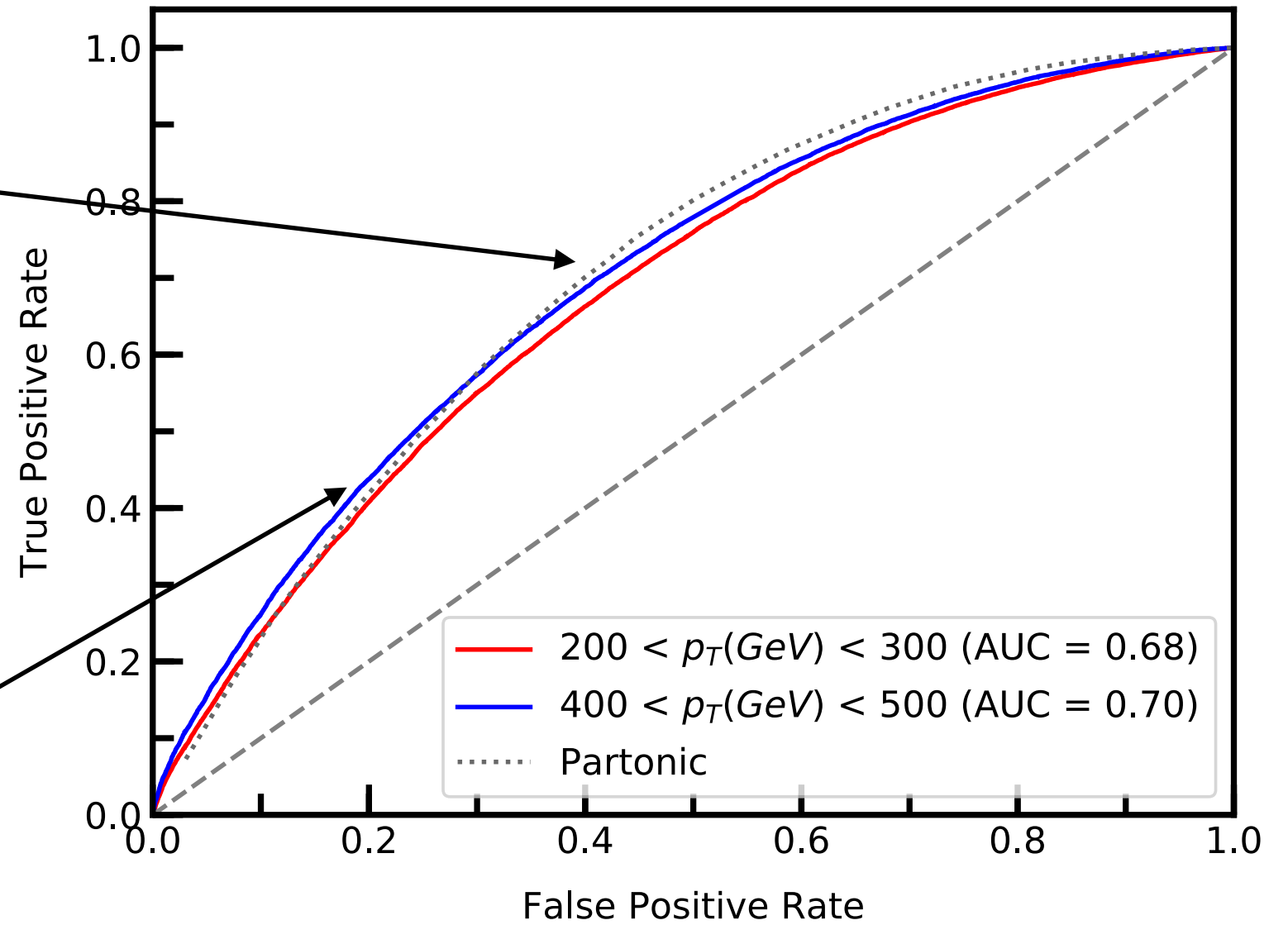
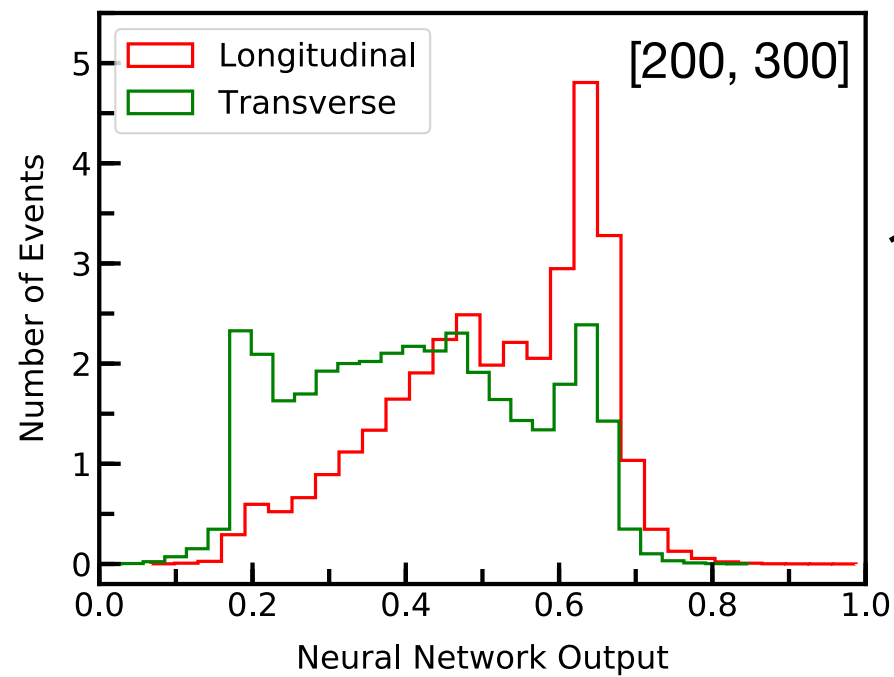
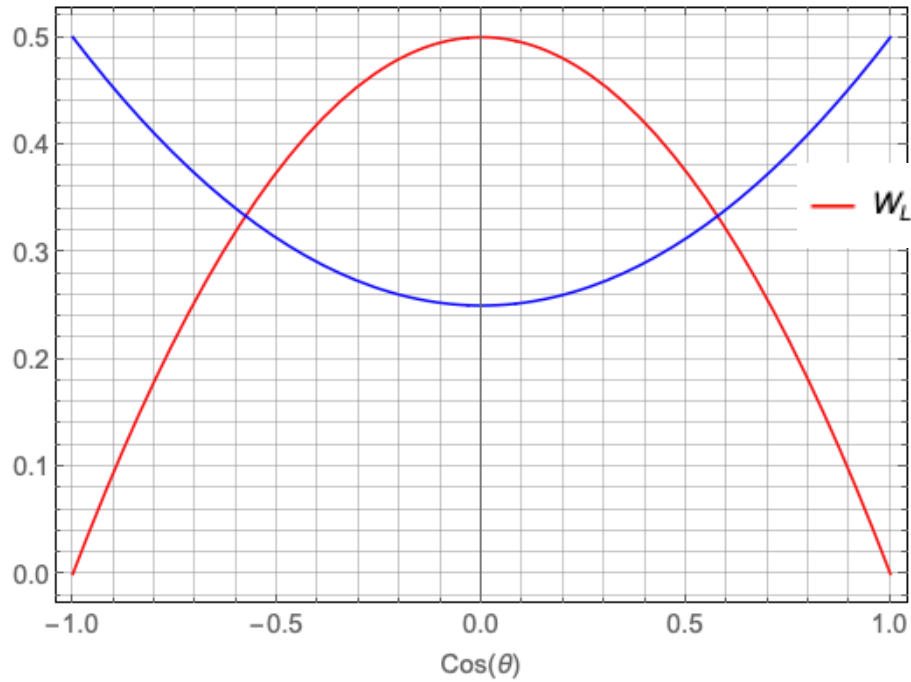
Testing Sample

$$pp \rightarrow W^\pm(j)Z(l^+l^-)$$

- We did not consider any underlying events ( looking into ideal scenarios as first study )

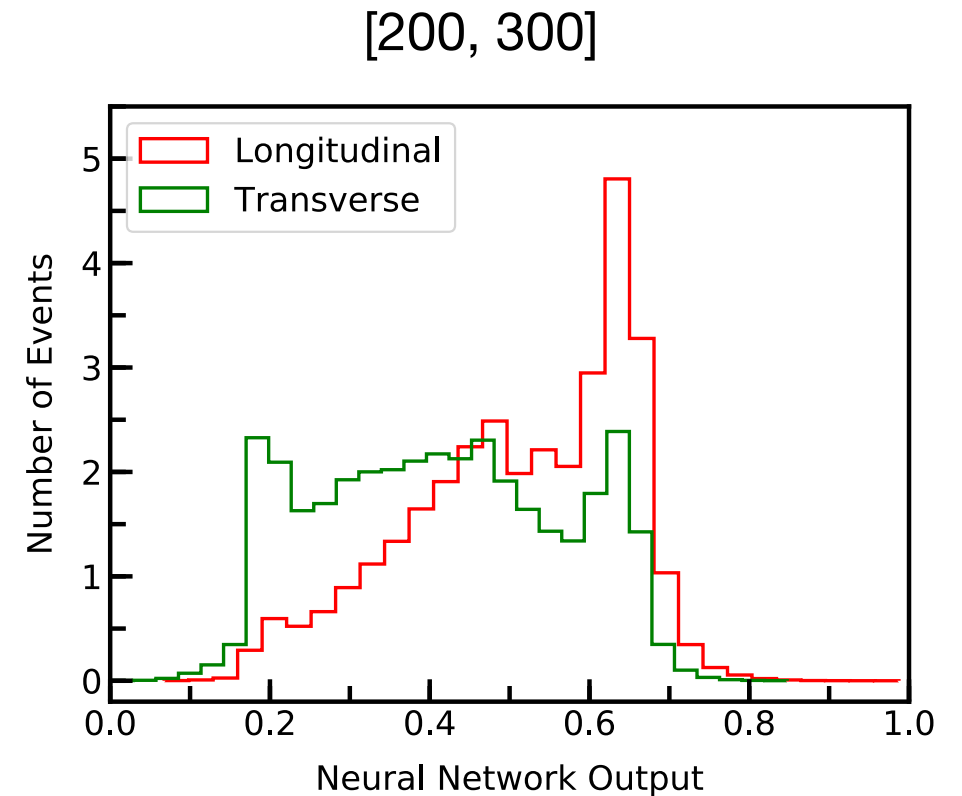
- Depending on  $p_T^W$ , images are separated into 2 bins: [200,300] and [400,500] since for fat jet,  $\Delta R \approx \frac{2m_W}{p_T^W}$

# ROC Curve



# Trained Network Quality Check

- Checking distribution can tell us how good the separation between two polarization
- **Inhibits potential event by event tagging because of large overlap**

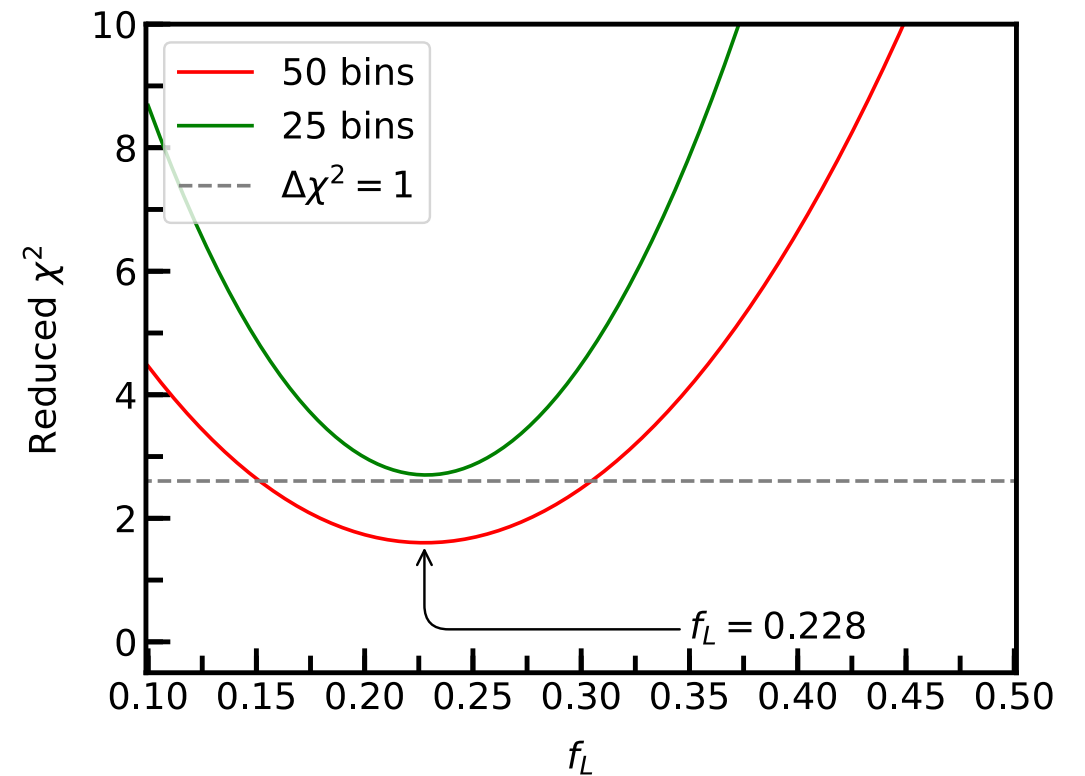
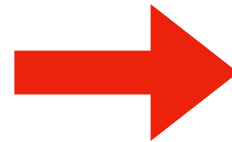
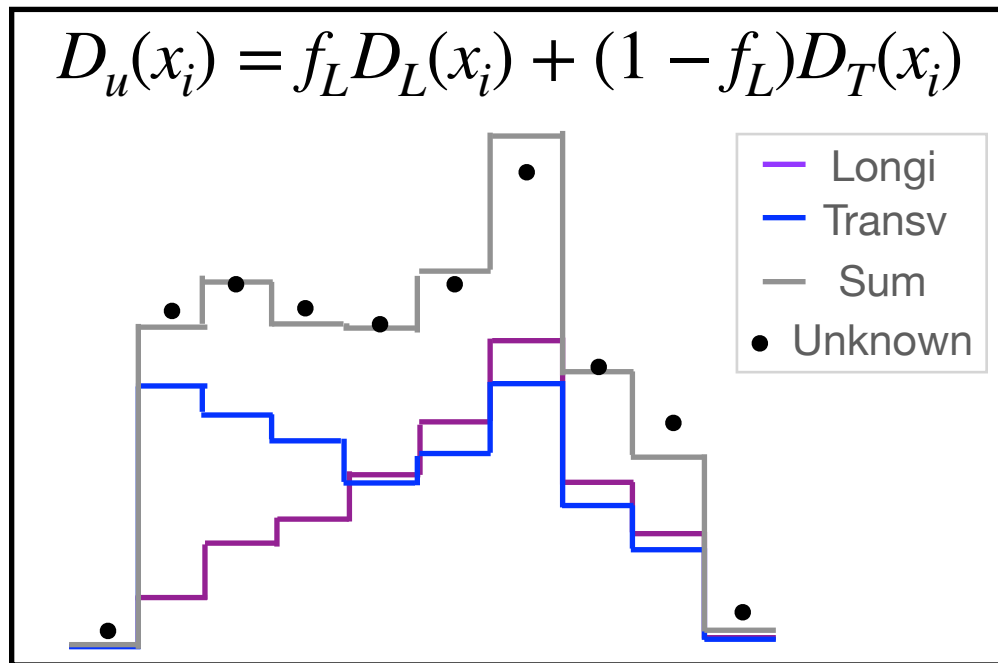


Putting decision threshold would contain large contamination

- As a result, we decide to use pure polarization distributions as template to identify the polarization content in given event collection
  - randomly select number of jet images from unknown sample → polarization fraction

# Longitudinal fraction ( $f_L$ )

## Template fit method



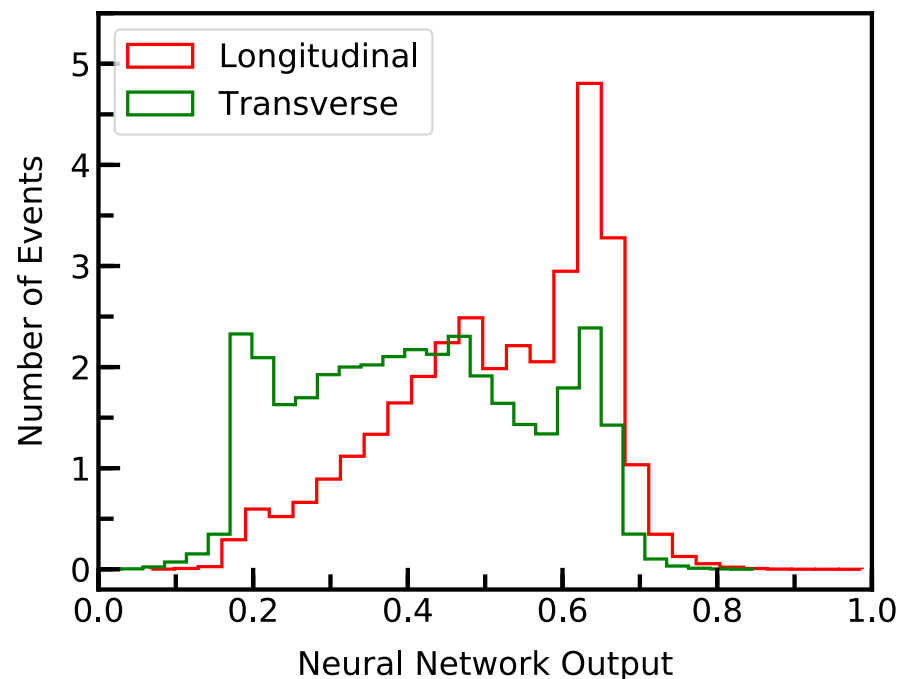
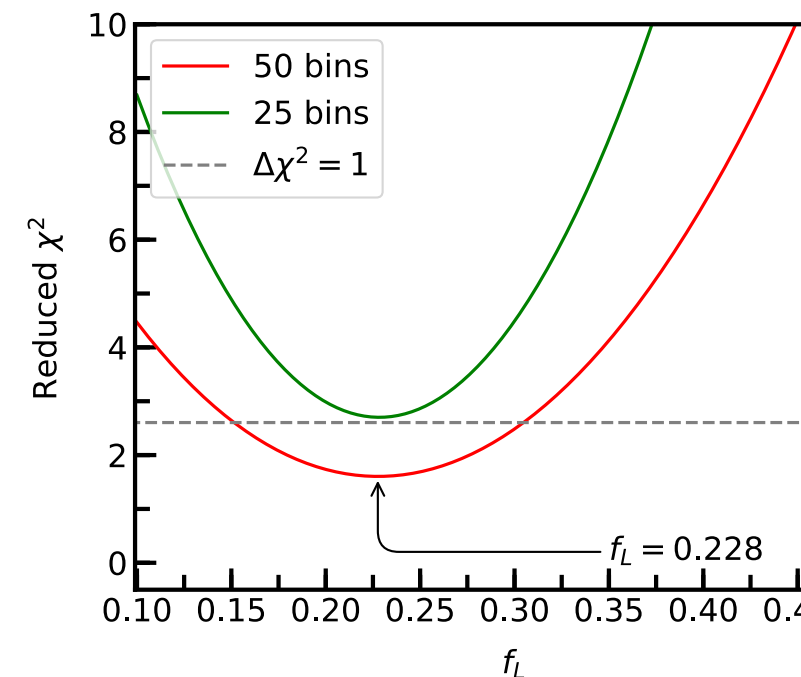
- Consider each pure polarization histogram as “template” that can be applied to the unknown sample
- Fit quality is determined by  $\chi^2$  distance test

$$\chi^2(f_L) = \sum_{i=1}^B \frac{(O_i - N_s(f_L L_i + (1 - f_L) T_i))^2}{N_s(f_L L_i + (1 - f_L) T_i)}$$

# Simpler Method

## Network output average method

- Template fitting method depends on finding “sweet spot” for  $f_L$ 
  - number of bins
  - find minimum  $\chi^2(f_L)$
- Simplify by treating output distribution as probability distribution



$$\int x dx (D_u(x) = f_L D_L(x) + (1 - f_L) D_T(x))$$

$$\langle x_u \rangle = f_L \langle x_L \rangle + (1 - f_L) \langle x_T \rangle$$

$$f_L = \frac{\langle x_u \rangle - \langle x_T \rangle}{\langle x_L \rangle - \langle x_T \rangle}$$

Confirmed that both yield the same result

# SM Prediction Result

$$pp \rightarrow W^\pm Z$$

$p_T$ range	truth $\sigma_L/\sigma_{tot}$	predicted $f_L$
<b>[200,300]</b>	0.265	$0.259 \pm 0.013$
<b>[400,500]</b>	0.304	$0.300 \pm 0.033$

- Truth  $f_L$  is obtained from polarization enforced feature of MadGraph
- At both  $p_T$ , predicted values are accurate with enough precision
- At high  $p_T$ , larger uncertainty comes from lower statistics
- Error is estimated from pseudo experiments
- CNN can predict well with SM case but need to test more (SMEFT extension)

# SMEFT Extension (Scenario 1)

## Shift cross section and polarization fraction

	$p_T$ range	$\sigma(pp \rightarrow W^\pm(jj)Z(\ell\ell))$ (fb)	truth $\sigma_L/\sigma_{tot}$	predicted $f_L$
SM	$200 \text{ GeV} \leq p_T \leq 300 \text{ GeV}$	6.67	0.265	$0.259 \pm 0.013$
	$400 \text{ GeV} \leq p_T \leq 500 \text{ GeV}$	0.35	0.304	$0.300 \pm 0.033$

	$p_T$ range	$\sigma(pp \rightarrow W^\pm Z)$ (fb)	truth $\sigma_L/\sigma_{tot}$	predicted $f_L$	
SM + single operator	$O_W$	$200 \text{ GeV} \leq p_T \leq 300 \text{ GeV}$	6.93	0.311	$0.297 \pm 0.010$
		$400 \text{ GeV} \leq p_T \leq 500 \text{ GeV}$	0.42	0.439	$0.391 \pm 0.033$
	$O_{3W}$	$200 \text{ GeV} \leq p_T \leq 300 \text{ GeV}$	6.58	0.258	$0.254 \pm 0.011$
		$400 \text{ GeV} \leq p_T \leq 500 \text{ GeV}$	0.50	0.198	$0.181 \pm 0.043$

- Benchmark Wilson Coefficient values
  - $c_W = 10^{-3}$
  - $c_{3W} = 3 \times 10^{-3}$
- If cross section measurement does not match to SM, polarization measurement can be a key to spot the dominance



# SMEFT Extension (Scenario 2)

Equal cross section but shift polarization fraction

SM

$p_T$ range	$\sigma(pp \rightarrow W^\pm(jj)Z(\ell\ell))$ (fb)	truth $\sigma_L/\sigma_{\text{tot}}$	predicted $f_L$
$200 \text{ GeV} \leq p_T \leq 300 \text{ GeV}$	6.67	0.265	$0.259 \pm 0.013$
$400 \text{ GeV} \leq p_T \leq 500 \text{ GeV}$	0.35	0.304	$0.300 \pm 0.033$

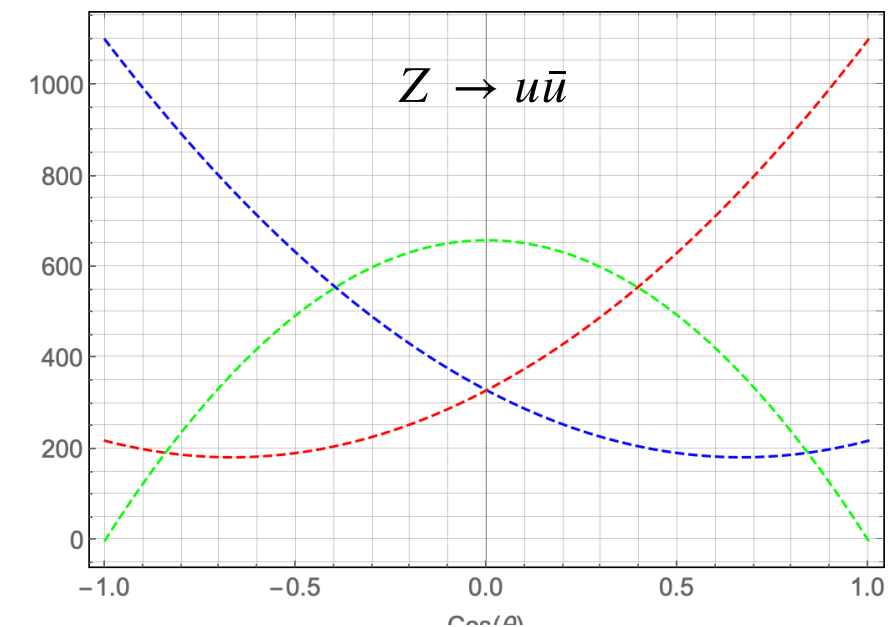
SM +  $\mathcal{O}_W$  +  $\mathcal{O}_{3W}$

$p_T$ range	$\sigma(pp \rightarrow W^\pm Z)$ (fb)	truth $\sigma_L/\sigma_{\text{tot}}$	predicted $f_L$
$200 \text{ GeV} \leq p_T \leq 300 \text{ GeV}$	6.68	0.202	$0.207 \pm 0.011$
$400 \text{ GeV} \leq p_T \leq 500 \text{ GeV}$	0.34	0.285	$0.282 \pm 0.044$

- Two Wilson coefficients are tuned to keep cross section the same but shift  $f_L$
- Even though cross section agrees with SM, polarization measurement can be a way to capture BSM signatures

# Conclusion/Discussion

- In this initial study, analysis using network's output average values can help to predict  $f_L$  even for hadronic  $W$
- Network prediction can catch  $f_L$  deviations originated from dim 6 operators
- With or without cross section shift, polarization measurement can clear out degeneracies between EFT operators
- Future directions
  - Possible applicability on  $Z$  jets
  - $W^\pm$  vs.  $Z$  vs. QCD (adding more realistic components)
  - Further optimization of network



# Different Network Results

## MaxOut & ResNet

	truth	$f_L$ CNN	$f_L$ MaxOut	$f_L$ ResNet
SM	0.265	$0.259 \pm 0.013$	$0.287 \pm 0.011$	$0.259 \pm 0.012$
$O_W$	0.311	$0.297 \pm 0.010$	$0.321 \pm 0.010$	$0.295 \pm 0.009$
$O_{3W}$	0.258	$0.254 \pm 0.011$	$0.282 \pm 0.012$	$0.257 \pm 0.011$

Low pT

	truth	$f_L$ CNN	$f_L$ MaxOut	$f_L$ ResNet
SM	0.304	$0.300 \pm 0.033$	$0.323 \pm 0.026$	$0.301 \pm 0.034$
$O_W$	0.439	$0.391 \pm 0.033$	$0.407 \pm 0.025$	$0.414 \pm 0.034$
$O_{3W}$	0.198	$0.181 \pm 0.043$	$0.250 \pm 0.026$	$0.194 \pm 0.032$

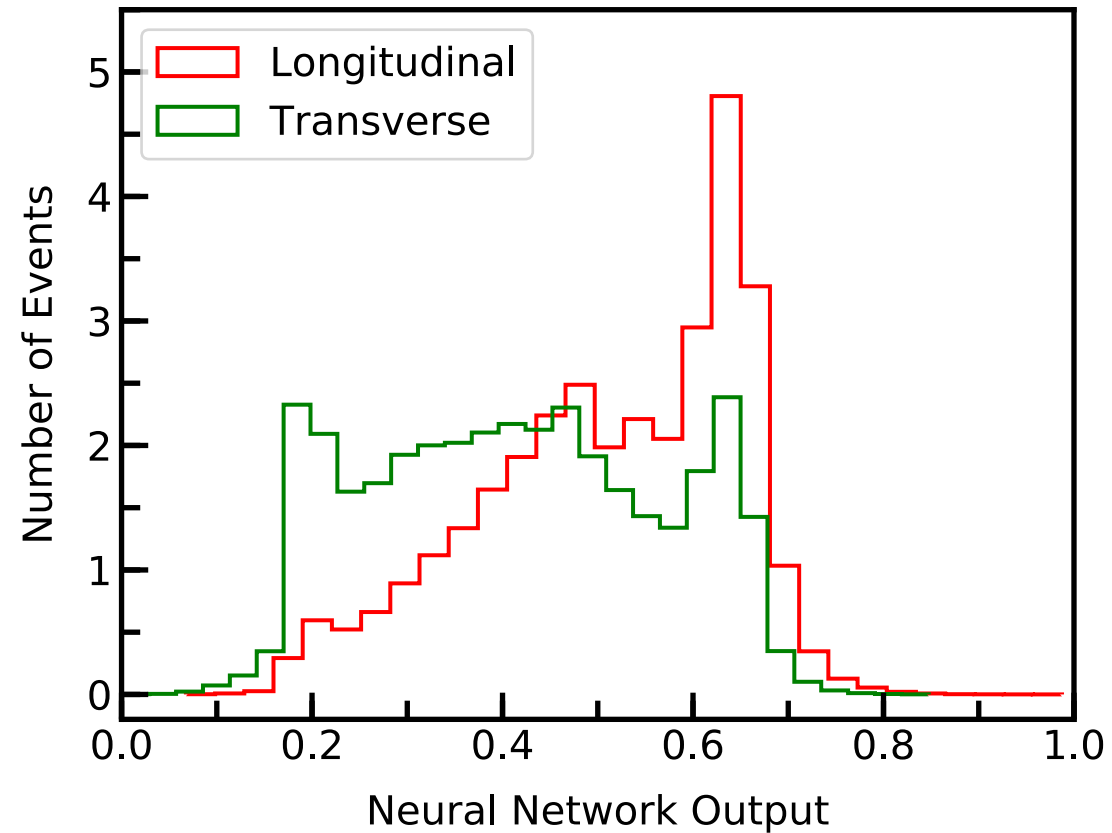
High pT

# Backup slides

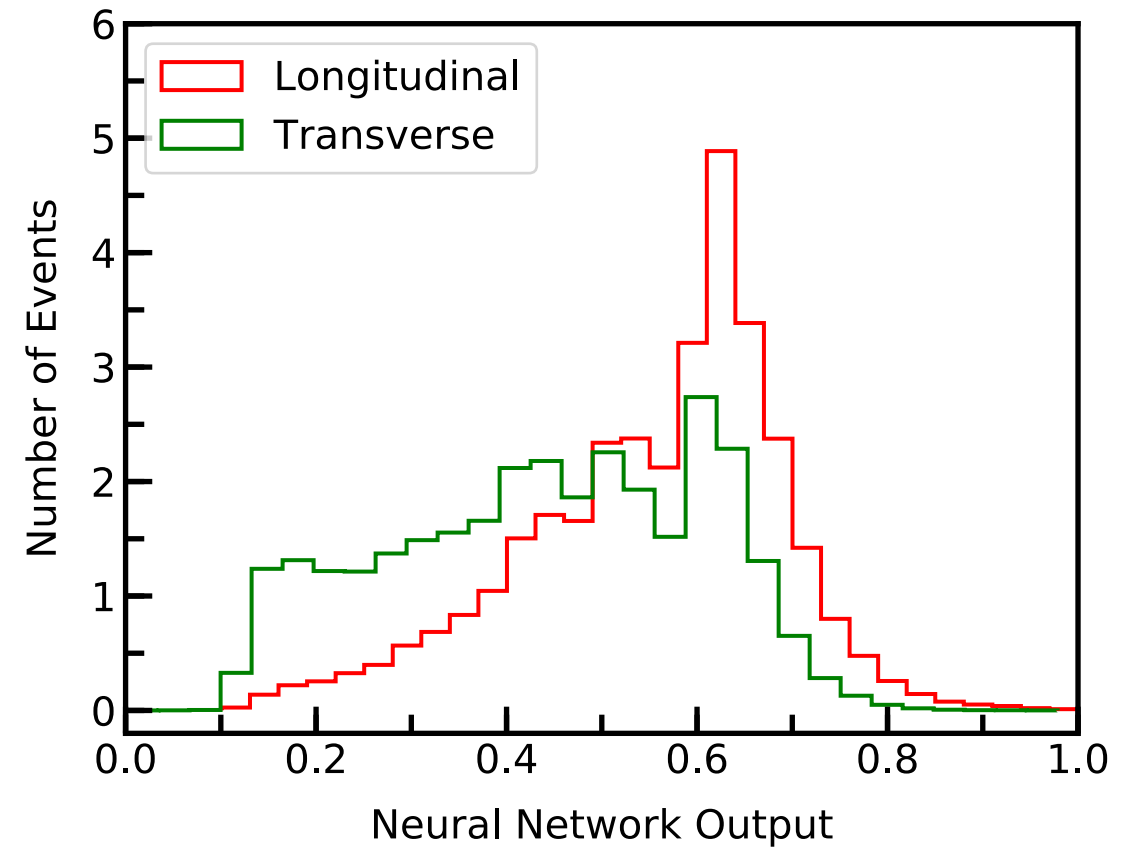
# Training Quality

## Distribution check

[200, 300]

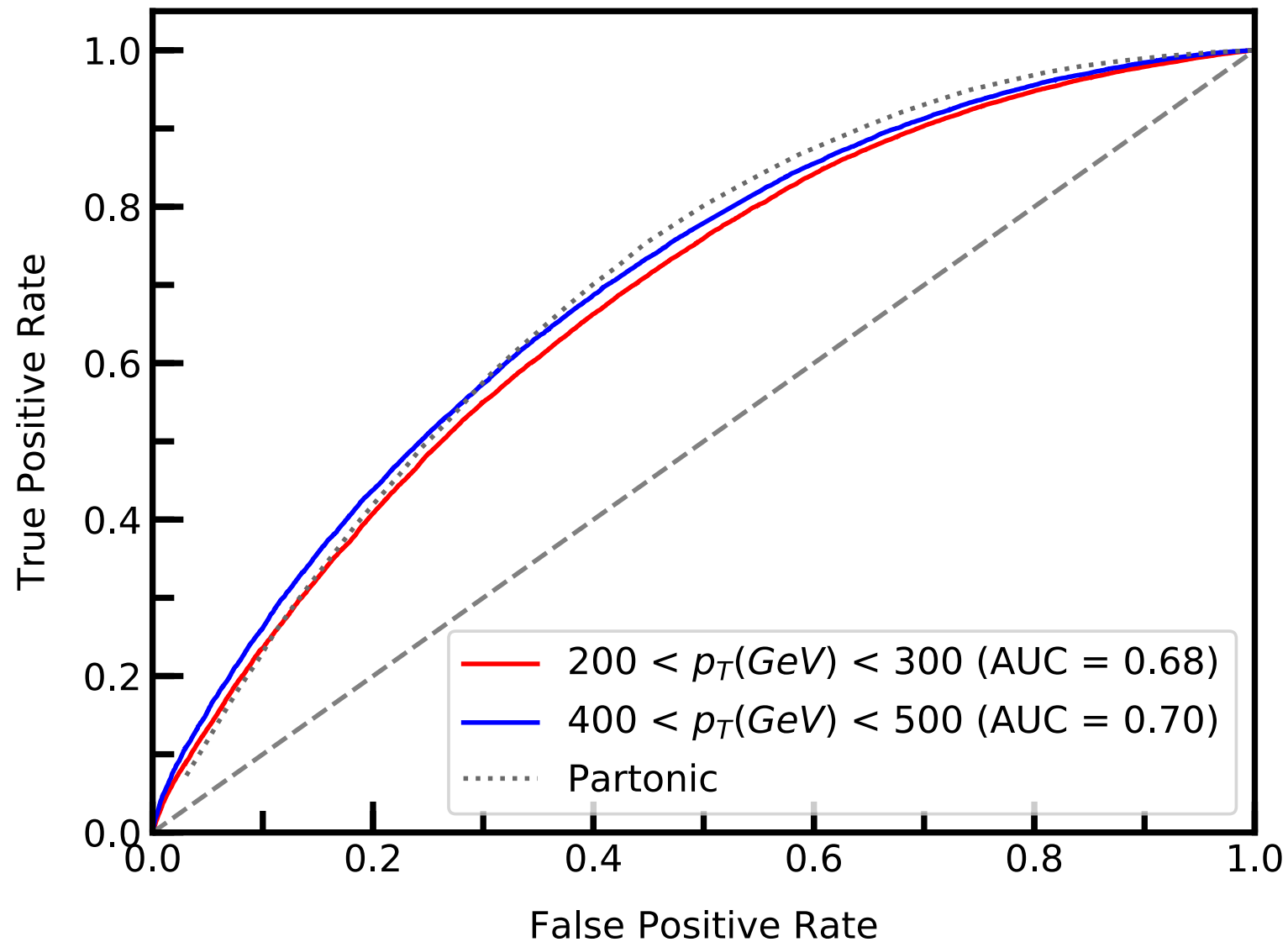


[400, 500]



# Training Quality

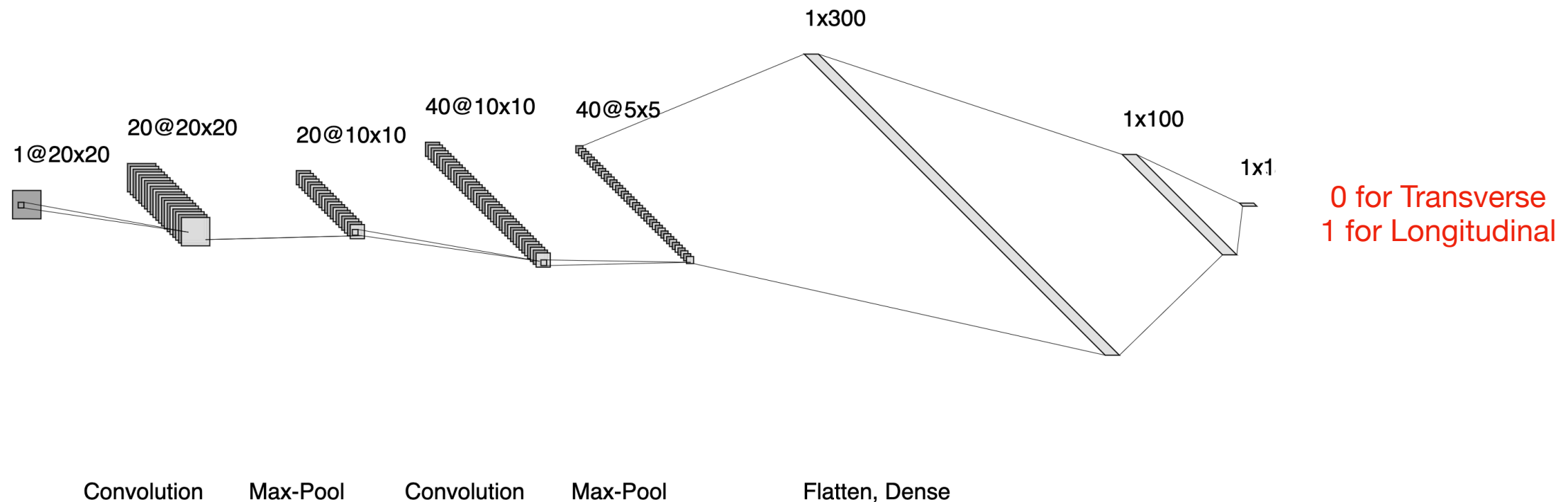
## Distribution check



- Checking distribution can tell us how good the separation between Logi and trans is.
- Inhibits potential event by event tagging since accuracy is  $\sim 60\%$
- Ensemble distribution checking to find longitudinal fraction ( $f_L$ )

# CNN Training

## Structure and training information



Ordinary CNN structure : Convolution - Flatten - Dense

pT bin	Training/Validation	Validation accuracy
[200, 300]	340k/85k	63%
[400, 500]	236k/59k	64%

# Kinematic Cut Effect

$$\frac{1}{\sigma} \frac{d\sigma}{d\cos\theta^*} = \frac{3}{8} (1 - \cos\theta^*)^2 f_L + \frac{3}{8} (1 + \cos\theta^*)^2 f_R + \frac{3}{4} \sin^2\theta^* f_0,$$

W rest frame

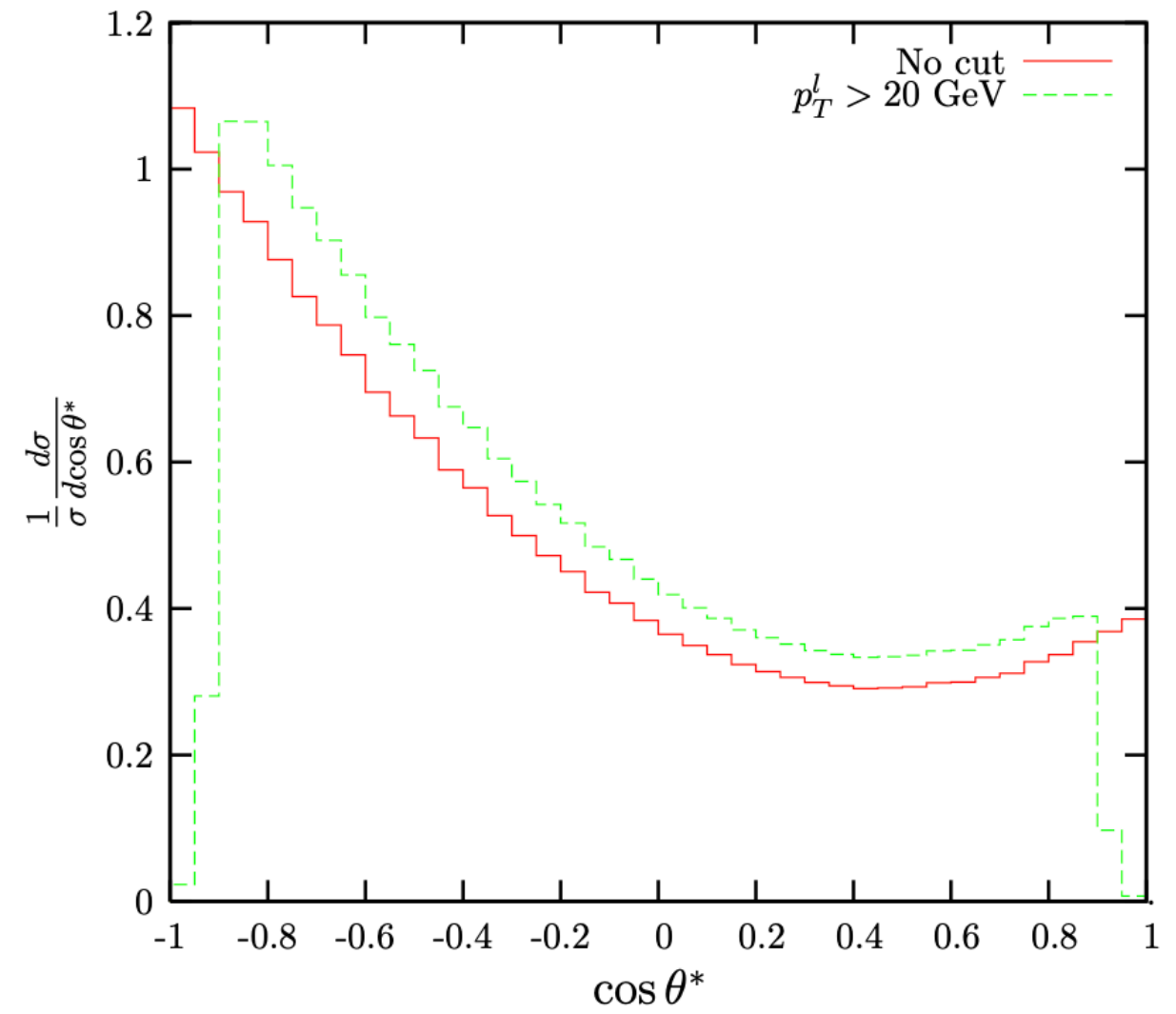
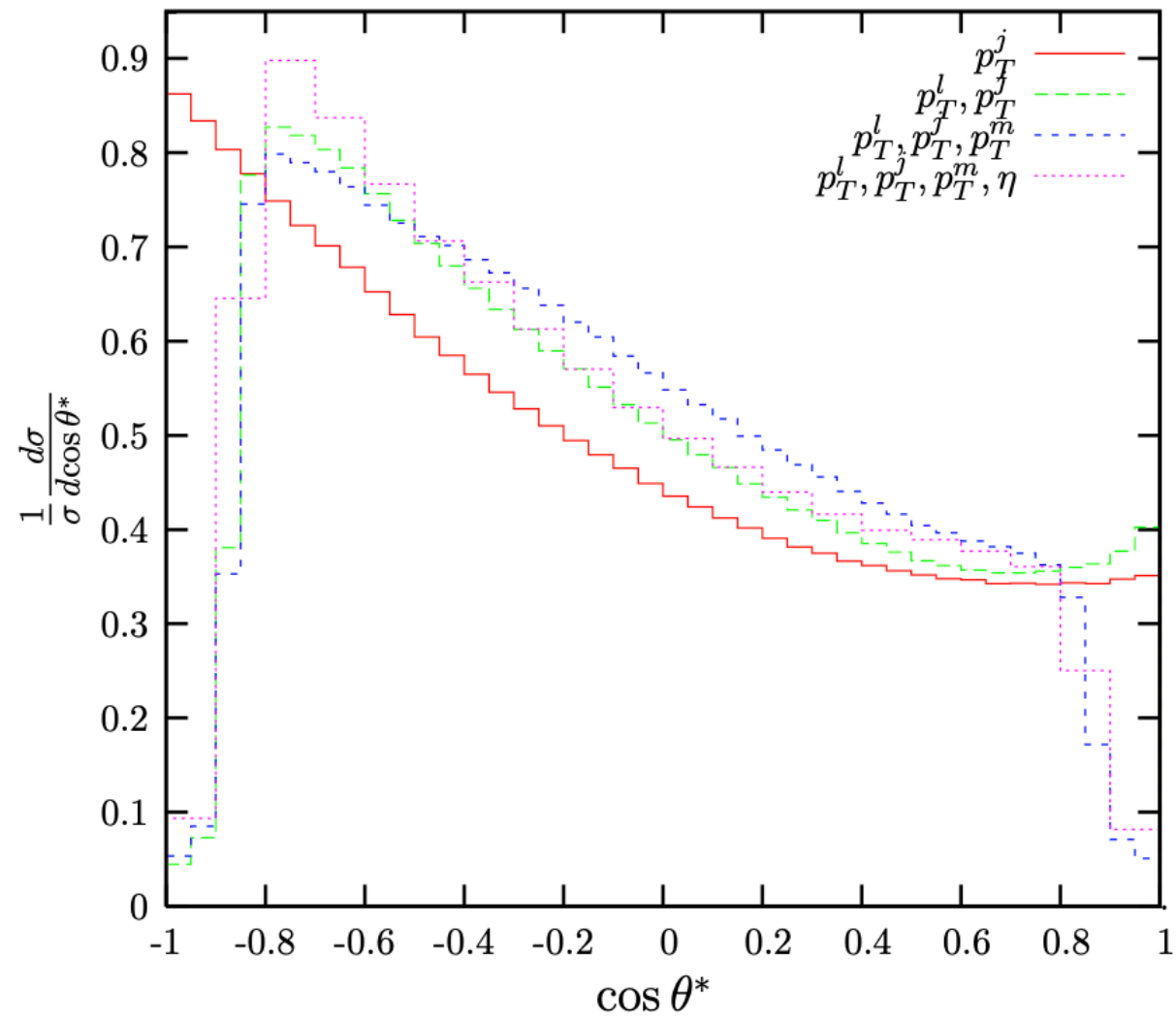
$$\frac{1}{\sigma} \frac{d\sigma}{d\cos\theta^* d\phi^*} = \frac{3}{16\pi} \left[ (1 + \cos^2\theta^*) + A_0 \frac{1}{2} (1 - 3\cos^2\theta^*) + A_1 \sin 2\theta^* \cos\phi^* \right. \\ \left. + A_2 \frac{1}{2} \sin^2\theta^* \cos 2\phi^* + A_3 \sin\theta^* \cos\phi^* + A_4 \cos\theta^* \right],$$

W at LHC

- Integrating over  $\phi^*$  will give the same result but kinematic cut can change



# Kinematic Cut Effect



# Kinematic Cut Effect

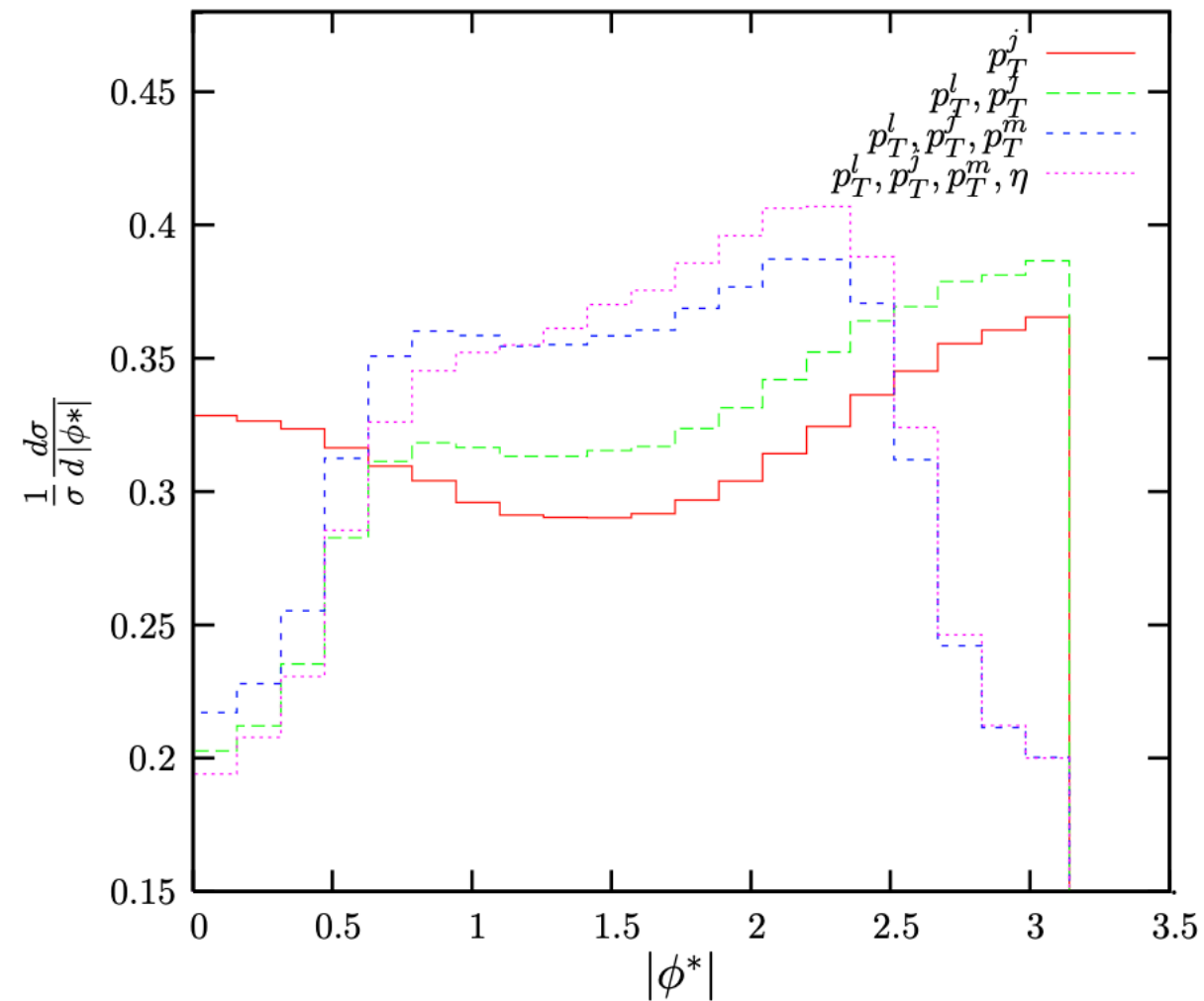


Figure 9: Normalised azimuthal angle distributions for a set of different selection cuts imposed on final-state leptons and jets for  $W^+ + 1$  jet production at 7 TeV.

# Uncertainty

## Small experiments

- From large test set, we randomly select subset ( $N$  number of events) to obtain  $f_L$
- $N$  is determined from expected number of events at particular luminosity
- At current LHC luminosity  $\sim 2000$  events at low  $p_T$  and 200 events at high  $p_T$
- At High Lumi LHC  $\sim 20k$  events at low  $p_T$  and 2k events at high  $p_T$
- By iterating the process, we can obtain average value with standard deviation

	<b>300 fb<sup>-1</sup></b>	<b>3000 fb<sup>-1</sup></b>
<b>[200,300]</b>	0.044	0.010
<b>[400,500]</b>	0.130	0.033

# Experimental Results

## ATLAS result

ATLAS Result ( $36\text{fb}^{-1}$ )

	Data	$f_0$		MATRIX	
		POWHEG+PYTHIA			
$W^+$ in $W^+Z$	$0.26 \pm 0.08$	$0.233 \pm 0.004$		$0.2448 \pm 0.0010$	
$W^-$ in $W^-Z$	$0.32 \pm 0.09$	$0.245 \pm 0.005$		$0.2651 \pm 0.0015$	
$W^\pm$ in $W^\pm Z$	$0.26 \pm 0.06$	$0.2376 \pm 0.0031$		$0.2506 \pm 0.0006$	
$Z$ in $W^+Z$	$0.27 \pm 0.05$	$0.225 \pm 0.004$		$0.2401 \pm 0.0014$	
$Z$ in $W^-Z$	$0.21 \pm 0.06$	$0.235 \pm 0.005$		$0.2389 \pm 0.0015$	
$Z$ in $W^\pm Z$	$0.24 \pm 0.04$	$0.2294 \pm 0.0033$		$0.2398 \pm 0.0014$	

ATLAS Collaboration [arXiv:1902.05759]

1. Previous attempts from ATLAS collaboration to measure polarization with leptonic final states
  - Leptonic final state: small branching ratio
  - Complication in  $\nu$  reconstruction
2. If we can use hadronic  $W$ , we gain more statistics but need to deal with hadronic jets