

# ML techniques in $H^+ \rightarrow \tau\nu$ mass reconstruction

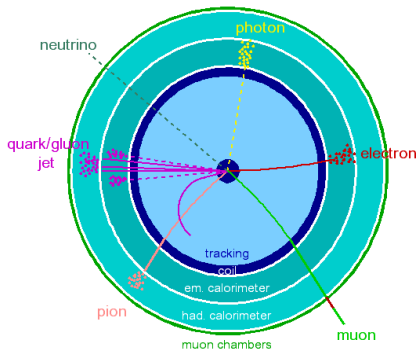
Max Isacson, Mikael Mårtensson,  
Camila Rangel Smith, Henrik Öhman



UPPSALA  
UNIVERSITET

Statistical Machine Learning, 21 april 2016

All HEP experiments have to deal with *missing (transverse) energy* (MET,  $E_T^{\text{miss}}$ ). Usually in the form of *neutrinos* escaping the detector without interacting.

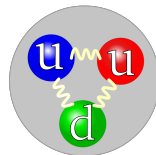


The missing energy can be reconstructed in the transverse plane, by energy conservation:

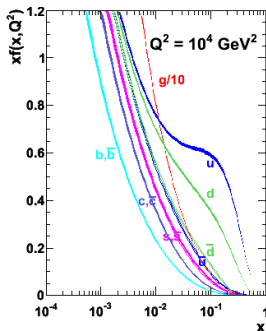
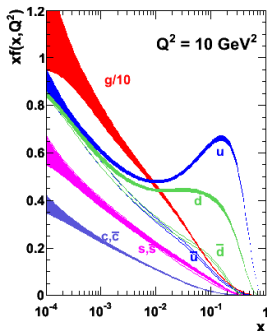
$$E_{x,y}^{\text{miss}} = - \sum E_{x,y} \quad (1)$$

However in hadron colliders the longitudinal energy information is lost.

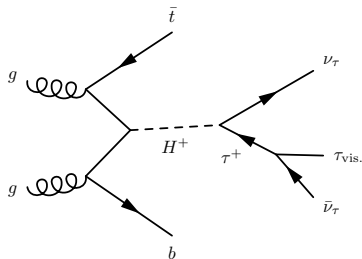
This is because we are actually colliding the *partons* inside the hadrons, which carry an unknown fraction  $x$  of the longitudinal momentum.



MSTW 2008 NLO PDFs (68% C.L.)

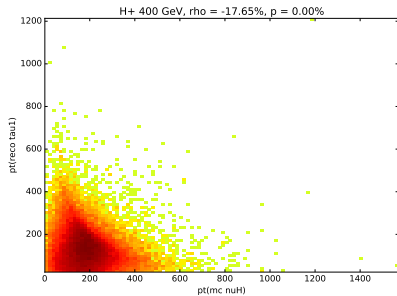
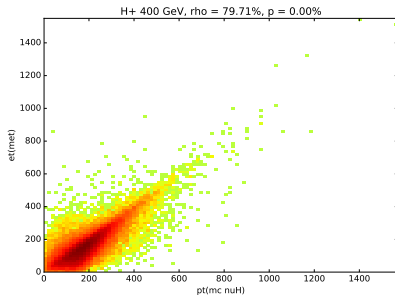
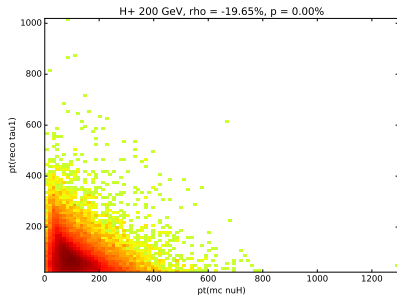
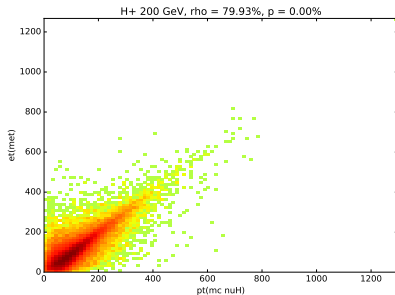


- ▶ Response of  $X$  — The ratio between a reco. quantity  $X$  and the true value  $X_{\text{truth}}$ , as a function of  $X_{\text{truth}}$ .
- ▶ Resolution of  $X$  — The distribution of  $(X_{\text{truth}} - X)/X_{\text{truth}}$ .
- ▶ Transverse mass  $m_T$  — The mass in the transverse plane between objects 1 and 2  $m_T = \sqrt{2E_T^1 E_T^2 (1 - \cos(\phi_1 - \phi_2))}$ .

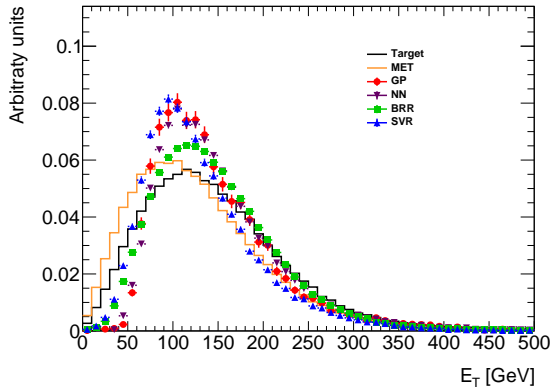


- ▶ Most of the missing energy will be carried by  $\nu_H$  — The neutrino from the Higgs decay.
- ▶ The features used are the various kinematics of the visible part of this process.
- ▶ We consider 3 masses for the  $H^+$ : 200, 300, and 400 GeV

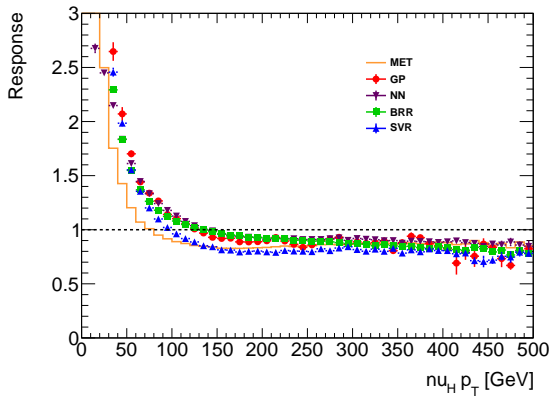
# Predictors

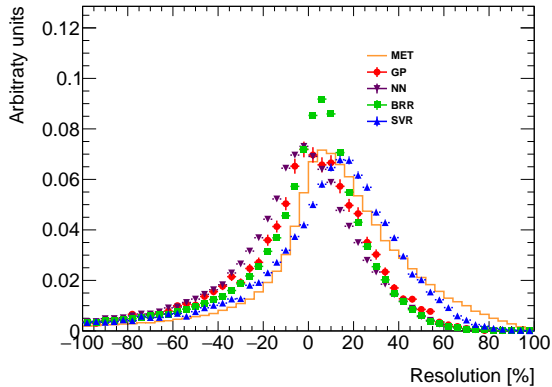


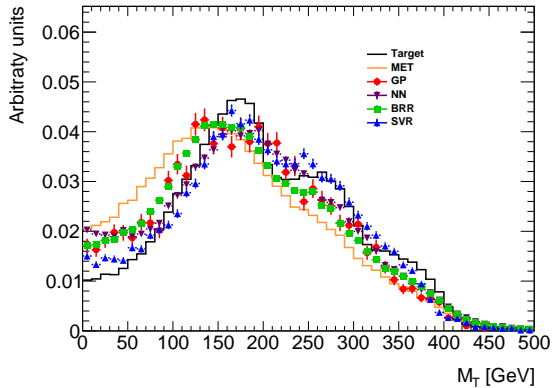
- ▶ Gaussian process
  - ▶ Neural network
  - ▶ Bayesian ridge regression
  - ▶ Support vector regression
- 
- ▶ The target for the training is always the  $p_T$  of  $nu_H$ .
  - ▶ We consider 4 measures of quality, the distribution of the predicted  $p_T$ , the response, the resolution, and the reconstructed  $m_T$  using the predicted  $p_T$ .
  - ▶ These are compared to the standard MET.











## Conclusions

- ▶ Of the tested methods the Neural net performs the best.
- ▶ However, all methods fails to reconstruct the spectrum at low  $p_T$ .

## Plans

- ▶ We plan to continue this in the future – Will be of interest to HBSM searches at ATLAS and other hadron experiments.
- ▶ The performance will probably be better when training against the full 4-vector of  $v_H$ .
- ▶ More complicated variables could also be included in the training.