

Improving Heavy Dirac Neutrino Prospects at Future Hadron Colliders Using Machine Learning

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Introduction

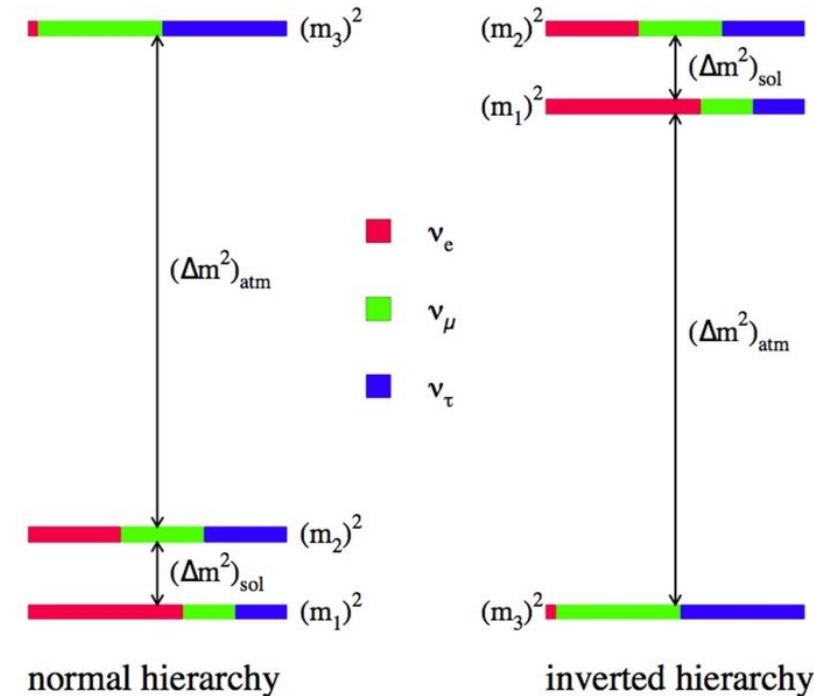
➤ Neutrino mass

- Confirmed by Neutrino Oscillation experiments
(Super-Kamiokande, Sudbury, Daya Bay, ⋯)

3 neutrino mass states.

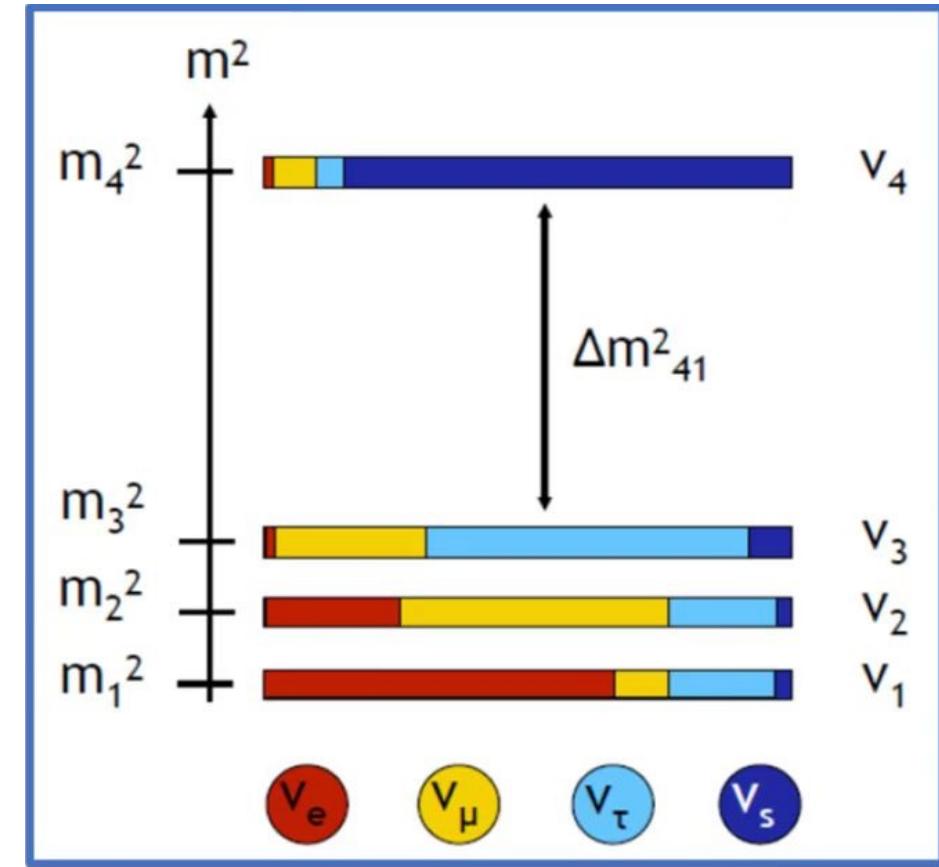
➤ In particle physics

- Beyond the Standard Model (SM) description
 - *New physics!*



Introduction

- Seesaw mechanisms provide natural explanations of the tiny neutrino masses.
- Type-I : Prof. Tsutomu Yanagida
- Type-II : many others...
- Type-III: Prof. Xiao-Gang He



Inverse seesaw

➤ The Yukawa Lagrangian is given by [1,2]

$$-\mathcal{L}_Y = Y_{\alpha\beta} \bar{L}_\alpha \Phi N_{R,\beta} + M_{N,\alpha\beta} \bar{S}_{L,\alpha} N_{R,\beta} + \frac{1}{2} \mu_{S,\alpha\beta} \bar{S}_{L,\alpha} S_{L,\beta}^C + \text{H. c.}$$

$L_\alpha = (\nu_\alpha, \ell_\alpha)^T$ - Standard Model lepton doublet

Φ - Standard Model Higgs doublet

$S_L^C \equiv S_L^T C^{-1}$ - charge conjugate of S_L

M_N - Dirac mass term

μ_S - Majorana mass term

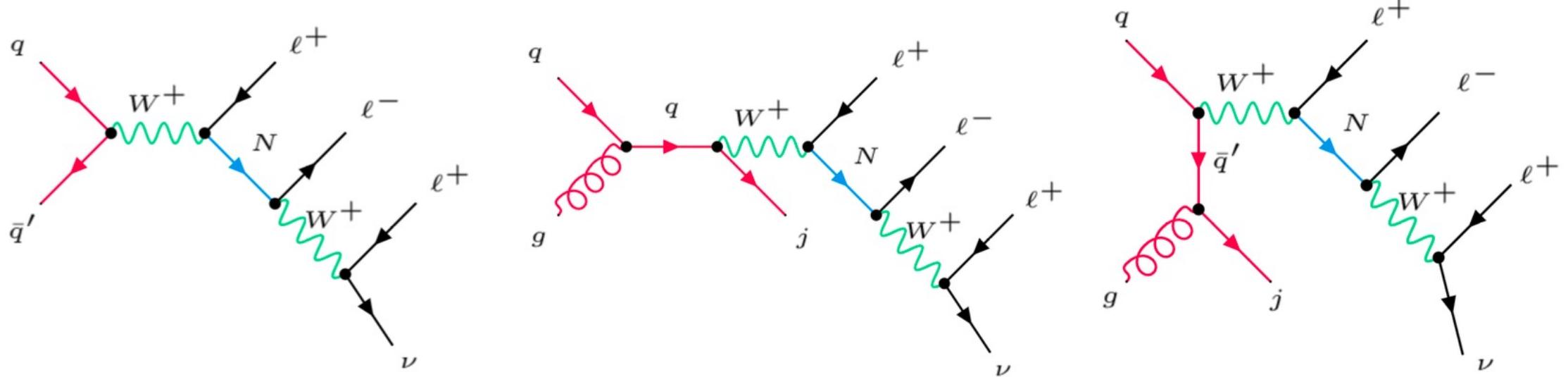
[1] R.N. Mohapatra, PRL 56 (1986) 561

[2] R.N. Mohapatra and J.W.F. Valle, PRD 34 (1986) 1642

Inverse seesaw

➤ In the limit of $\|\mu_S\| \ll \|M_N\|$ (with $\|\mu_S\| \equiv \sqrt{\text{tr}(x^\dagger x)}$), the primary signature of heavy neutrinos at the hadron colliders are [3,4]:

$$pp \rightarrow \ell_\alpha^\pm N \rightarrow \ell_\alpha^\pm \ell_\beta^\mp W^\pm \rightarrow \ell_\alpha^\pm \ell_\beta^\mp \ell_\gamma^\pm \nu$$



[3] C. Degrande *et al.*, PRD 94 (2016) 053002 (arXiv:1602.06957)

[4] arXiv:1408.0983, 1706.02298

Inverse seesaw

➤ The heavy neutrino N will decay [5,6]:

$$\Gamma(N \rightarrow \ell^- W^+) = \frac{\alpha_W |V_{\ell N}|^2}{16} \frac{m_N^3}{m_W^2} \left(1 - \frac{m_W^2}{m_N^2}\right)^2 \left(1 + \frac{m_W^2}{m_N^2}\right)$$

$$\Gamma(N \rightarrow \nu_\ell Z) = \frac{\alpha_W |V_{\ell N}|^2}{32 \cos^2 \theta_W} \frac{m_N^3}{m_Z^2} \left(1 - \frac{m_Z^2}{m_N^2}\right)^2 \left(1 + \frac{m_Z^2}{m_N^2}\right)$$

$$\Gamma(N \rightarrow \nu_\ell h) = \frac{\alpha_W |V_{\ell N}|^2}{32} \frac{m_N^3}{m_W^2} \left(1 - \frac{m_h^2}{m_N^2}\right)^2$$

[5] A. Pilaftsis, Z. Phys. C 55 (1992) 275

[6] W. Buchmuller and C. Greub, Nucl. Phys. B 363 (1991) 345

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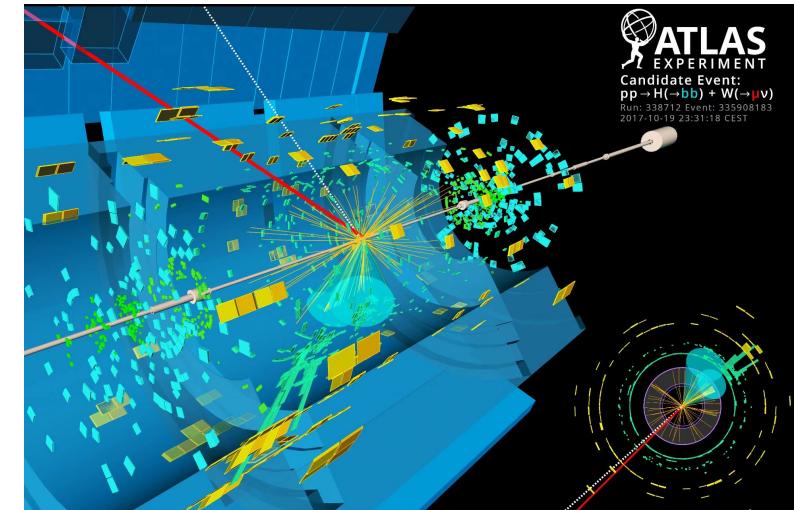
Collider analysis – Signal generation

- The final states of the signature in the detector:
electrons, muons, jets, ⋯

- Measurements of each particle in the detector:

- p_T - transverse momentum
- η - pseudorapidity
- ϕ - azimuthal angle

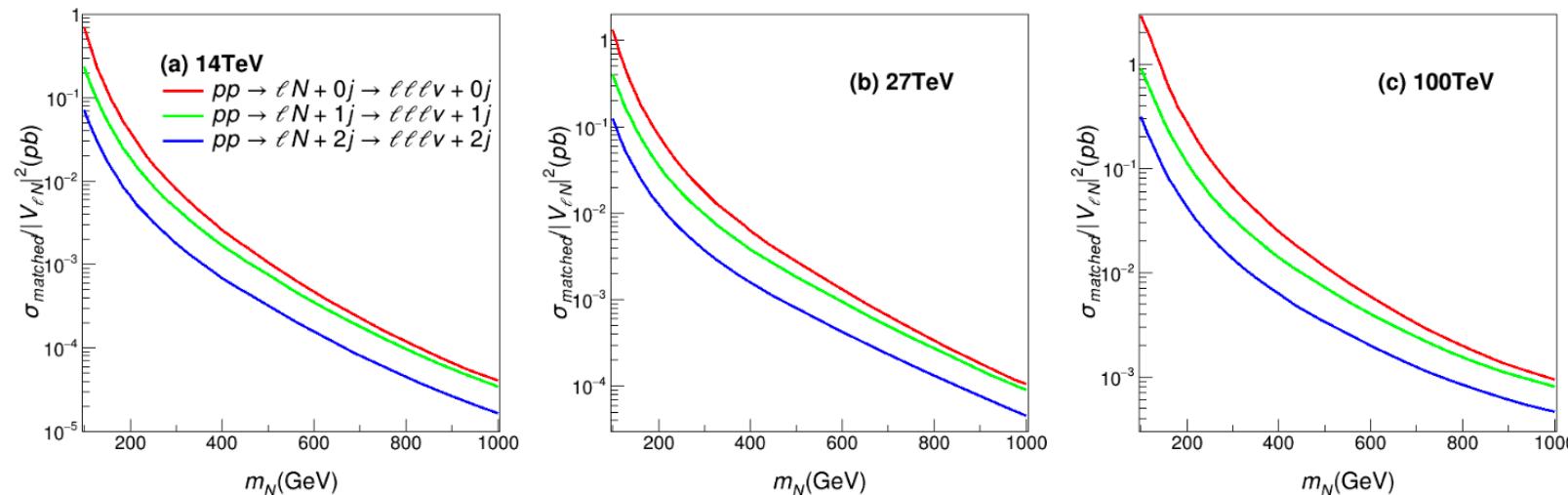
- Geometrical acceptance cuts from ATLAS: $p_T^{\ell,leading} > 20 \text{ GeV}$



\sqrt{s} (TeV)	electron		muon		jet	
	p_T (GeV)	$ \eta $	p_T (GeV)	$ \eta $	p_T (GeV)	$ \eta $
14	>10	< 2.47	>10	< 2.7	> 20	< 4.5
27	>10	< 2.47	>10	< 2.7	> 30	< 4.5
100	>15	< 2.47	>15	< 2.7	> 45	< 4.5

Collider analysis – Signal generation

- Cross sections of trilepton signal process



- Cross sections increase with higher center of mass energies.
- Cross sections with 2jets are much smaller than those with 1jet. It is safe to neglect events with higher jet multiplicity $n_j \geq 3$.

Collider analysis – Signal generation

- Assuming either $|V_{eN}| \neq 0$ or $|V_{\mu N}| \neq 0$, in the charge space, the three leptons of an event can be either $+ - +$ or $- + -$. All the resultant trilepton states are

mixing	trilepton states	signs ($\pm \mp \pm$)
V_{eN}	eee	$e^\pm e^\mp e^\pm$
	$ee\mu$	$e^\pm e^\mp \mu^\pm$
$V_{\mu N}$	$\mu\mu e$	$\mu^\pm \mu^\mp e^\pm$
	$\mu\mu\mu$	$\mu^\pm \mu^\mp \mu^\pm$

Collider analysis – background generation

➤ The main background:

$$pp \rightarrow ZW^\pm \rightarrow \ell_1^\pm \ell_2^\mp \ell_3^\pm \nu$$

➤ Pre-selection:

- The total number of energetic charged leptons is exactly 3, i.e. $n_e + n_\mu = 3$.
- The invariant mass of the two leading leptons should be larger than 12 GeV, i.e. $m_{\ell_1 \ell_2} > 12 \text{ GeV}$.
- The number of jets is not larger than 2, i.e. $n_j \leq 2$.
- The missing transverse momentum $E_T^{miss} = | - \sum_{\nu_i} \vec{p}_T(\nu_i) |$ is larger than 20 GeV, i.e. $E_T^{miss} > 20 \text{ GeV}$.

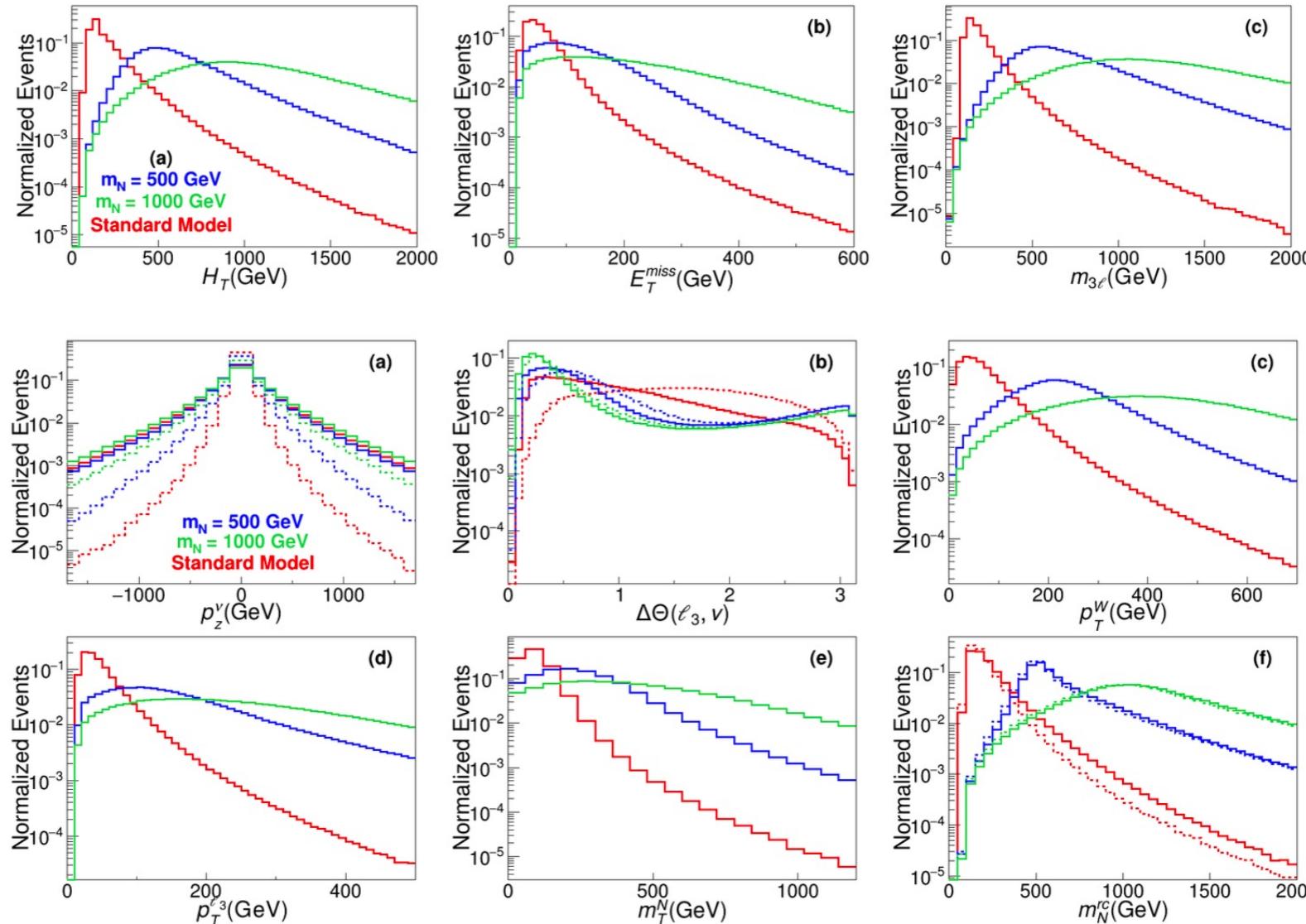
Collider analysis – background generation

- Cross sections of all possible trilepton final state processes after pre-selections:

process	cross section (fb)	
	0-jet	1-jet
$ZW \rightarrow \ell\ell\nu$	29.10	20.50
$\ell\ell\nu$ (off-shell + interference)	1.65	0.84
4ℓ	1.56	1.25

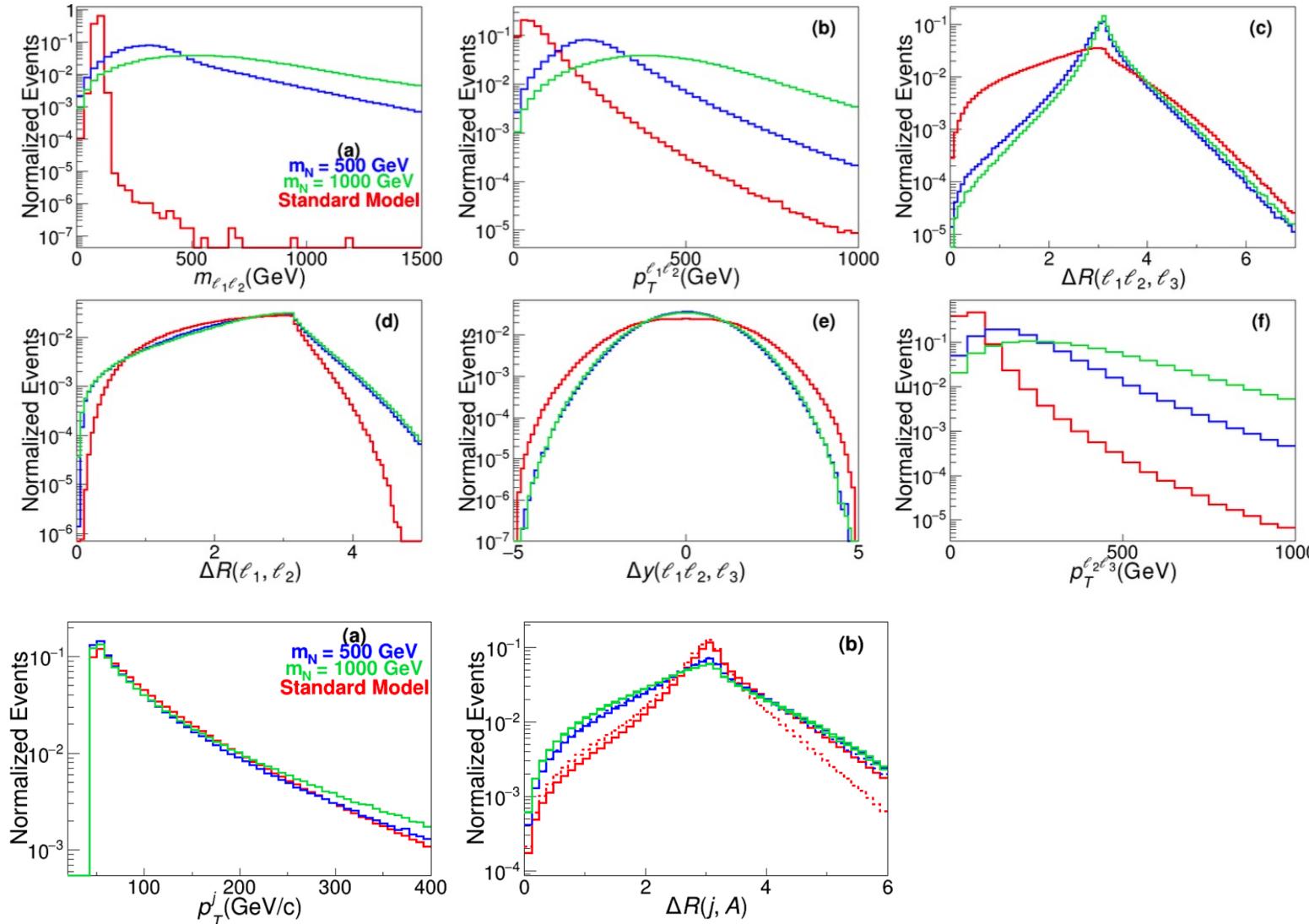
- The contributions of the off-shell and four-lepton processes to the standard model backgrounds are small. We can just scale the ZW background by 1.1 to take the other two into account.

Collider analysis – Feature observables



Variables reconstructed from the 3-lepton measurements.

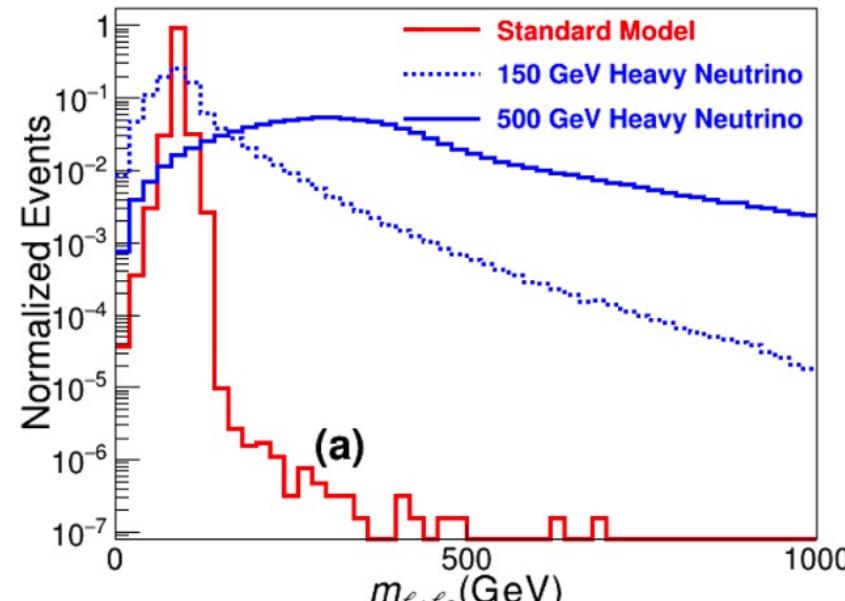
Collider analysis – Feature observables



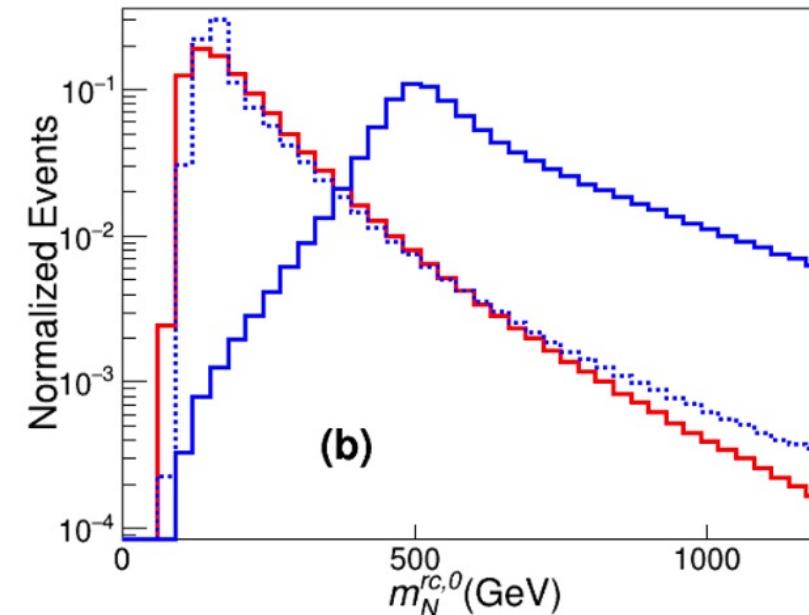
Variables reconstructed from the 3-lepton measurements.

Collider analysis – Feature observables

- Variables with the largest separation power:



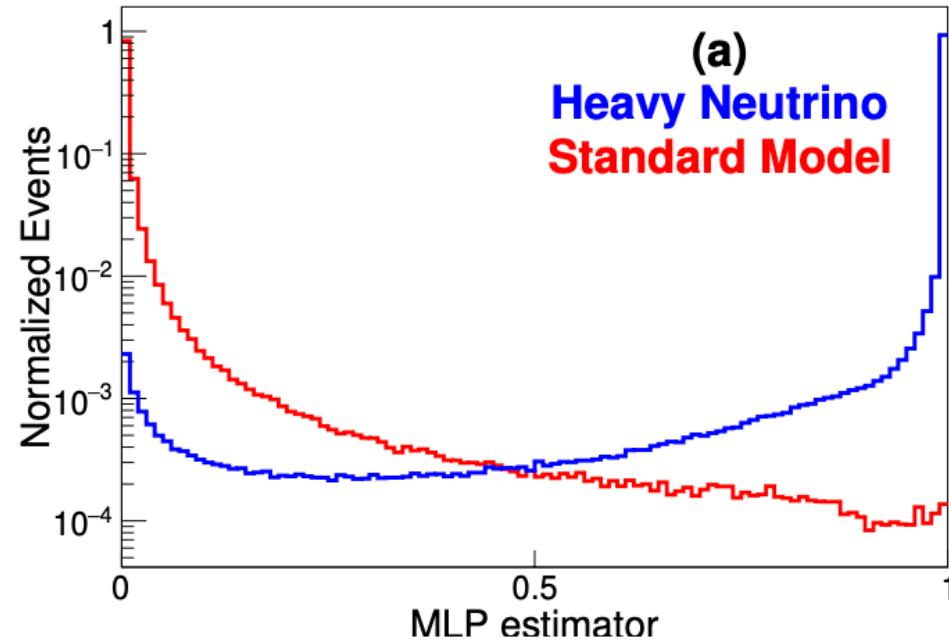
Z mass



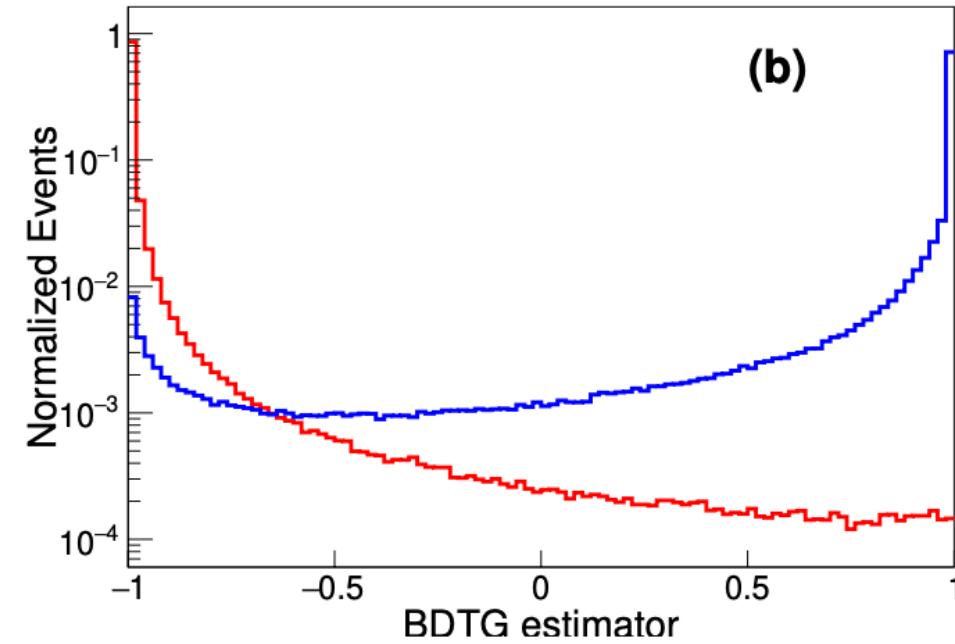
Heavy neutrino mass
(calculated from W
mass assumption) $W^\pm \rightarrow \ell_3^\pm \nu$

Collider analysis – Machine Learning (ML)

- The ML estimator distributions for $m_N = 500$ GeV:



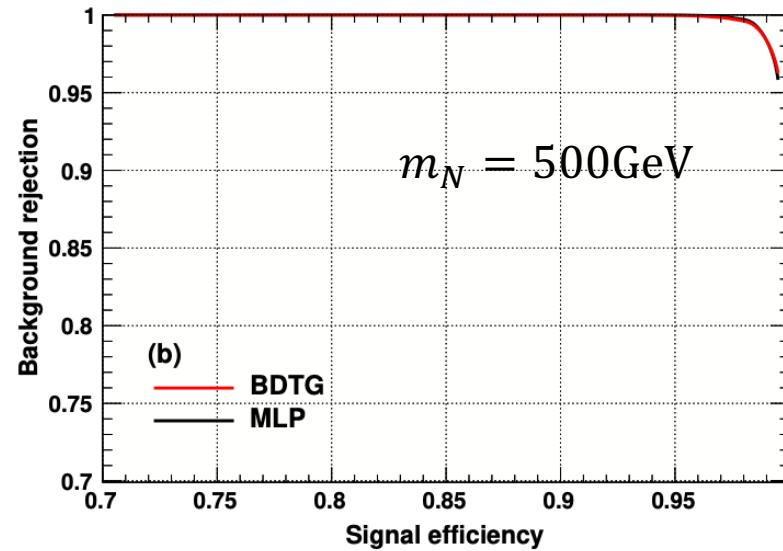
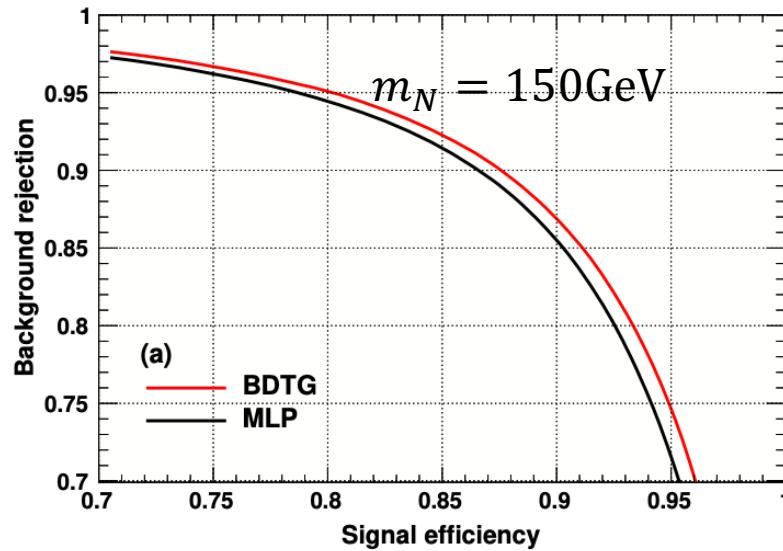
Multi-Layer Perceptron (MLP)



Boosted Decision Tree with Gradient boosting (BDTG)

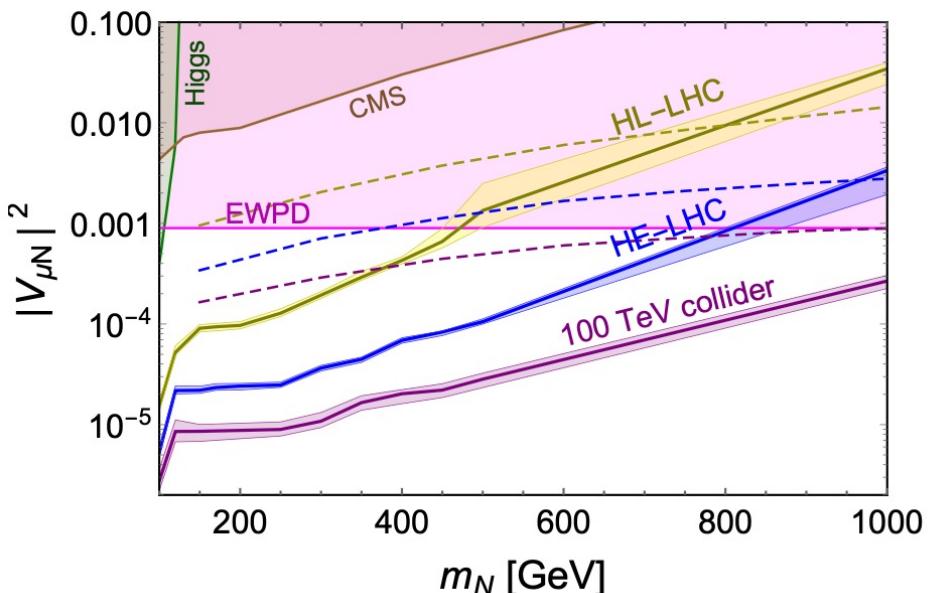
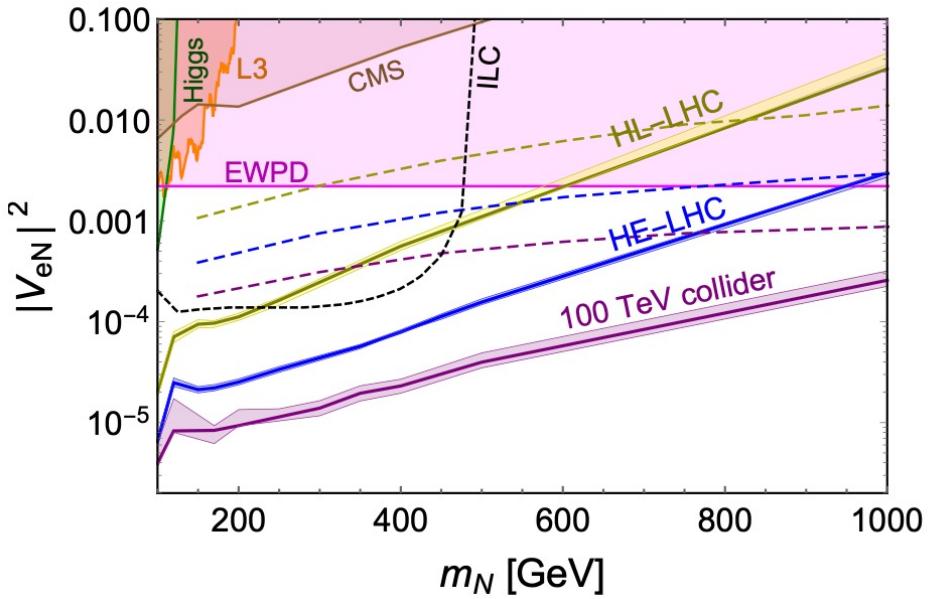
Collider analysis – Machine Learning

- The signal efficiency and background rejection power:



- BDTG and MLP methods show compatible results.
- When the heavy neutrino mass goes to a higher value, the separation becomes better.

Results



- Sensitivities of the heavy-light neutrino mixing $|V_{eN}|^2$ (upper) and $|V_{\mu N}|^2$ (lower) at 95% C.L. .
- With machine learning methods, $|V_{lN}|^2$ can be improved up to $O(10^{-6})$ for heavy neutrino mass $m_N = 100$ GeV and $O(10^{-4})$ for $m_N = 1$ TeV.

$\sqrt{s} = 14$ TeV(3ab^{-1}), 27 TeV(15ab^{-1}), and 100 TeV(30ab^{-1})

Summary

- By using the machine learning methods, we study the sensitivities of heavy pseudo-Dirac neutrino N in the inverse seesaw at the high-energy hadron colliders.
- We use either the Multi-Layer Perceptron or the Boosted Decision Tree with Gradient Boosting to analyze the kinematic observables and optimize the discrimination of background and signal events.
- It is found that the reconstructed Z boson mass and heavy neutrino mass play crucial roles in separating the signal from backgrounds.
- The prospects of heavy-light neutrino mixing $|V_{lN}|^2$ (with $l = e, \mu$) are estimated by using machine learning at the hadron colliders with $\sqrt{s} = 14$ TeV, 27 TeV, and 100 TeV, and it is found that $|V_{lN}|^2$ can be improved up to $O(10^{-6})$ for heavy neutrino mass $m_N = 100$ GeV and $O(10^{-4})$ for $m_N = 1$ TeV.

Thank you !