In addition to using traditional cuts (Higgs combined tool) to set upper limits on Z'2HDM and Z'Baryonic models, a multivariate analysis using discriminants from a Bayesian Neural Network (BNN) was used. For an overview and example of how a Bayesian framework fits with a neural network, see Bayesian Neural Networks [1]. To capture the best results, three different Bayesian discriminants were created, each using unique variables in their discriminants. The performance of two of these discriminants is Figure 1.

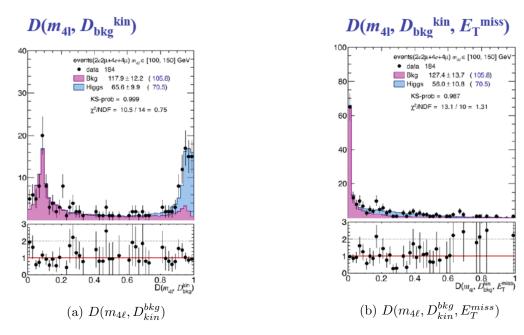


Figure 1: Performance of Discriminants

From this preliminary result, it appears that the $D(m_{4\ell}, D_{kin}^{bkg})$ can better differentiate between Higgs events and background events. Nevertheless, both were used to set upper limits.

To produce the limits for each model, a Bayesian approach was again used, this time on μ . Bayesian

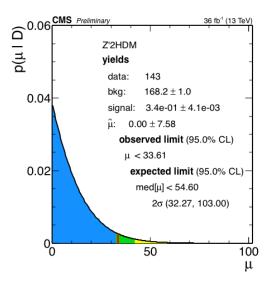


Figure 2: Upper limit on Z'2HDM with $m_{Z'} = 600 GeV, m_{A0} = 300 GeV$

statistics rely on given prior probabilities to predict the probability of data fitting a model and, in turn, of a model fitting data. To keep this analysis as general as possible and avoid any unintended bias, a flat prior was used. This Bayesian analysis was done on each individual model (with varied masses of Z' and it's decay products) for a 95% confidence interval. An example of an individual limit on μ is shown in Figure 2.

Combining the observed μ with the branching ratio and cross sections for each model has yielded upper limits, with 95% confidence, on Z'2HDM and Z'baryonic models. In addition to the three BNN discriminants $D(E_T^{miss}), D(m_{4\ell}, D_{kin}^{bkg}), D(m_{4\ell}, D_{kin}^{bkg}, E_T^{miss})$ used to set upper limits, E_T^{miss} and $P_{T4\ell}$ were also used in the Bayesian analysis to set limits. The upper limits differed depending on which variable was used to set them, and these differences can be seen in Table 1 for Z'Baryonic $m_{Z'} = 95$, $m_{\chi} = 50$. Note that these are the limits before being weighted by their specific branching ratio and cross section. Smaller limits are better.

Value	E_T^{miss}	$P_{T4\ell}$	$D(E_T^{miss})$	$D(m_{4\ell}, D_{kin}^{bkg})$	$D(m_{4\ell}, D_{kin}^{bkg}, E_T^{miss})$
Observed	18.231	25.764	41.898	104.778	37.358
-2σ	18.713	26.003	42.280	103.086	37.913
$-\sigma$	20.624	28.943	52.733	134.193	47.448
Expected	26.121	33.710	70.289	172.855	63.037
$+\sigma$	33.386	42.715	102.172	257.193	84.345
$+2\sigma$	42.389	51.528	136.672	324.874	115.923

Table 1: μ values for Z'Baryonic $m_{Z'} = 95$, $m_{\chi} = 50$

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The table shows us that discriminants without kinematic variables do are not sensible to limit our models. Instead, traditional approaches like E_T^{miss} and $P_{T4\ell}$ set the lowest limits, along with a well made discriminant $D(m_{4\ell}, D_{kin}^{bkg}, E_T^{miss})$ based off of these variables. At this time, systematic uncertainties associated with these variables are not included in these limits, and that those uncertainties will propagate to the discriminants. The BNN itself, however, does not have any systematic uncertainty associated with it. These limits are easier to visualize graphically. Below in Figure 3 are limits on Z' $m_{A0} = 300$ and Z'Baryonic $m_{\chi} = 1$ (the Z'Baryonic model used for the combined channel monoHiggs analysis) made with E_T^{miss} as the limiting variable.

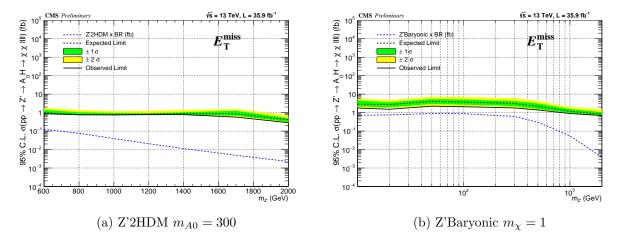


Figure 3: Upper limits on models using E_T^{miss}

Here we can see the limits on various mass models with the given m_{A0} and m_{χ} produced. The goal of the analysis is to get the limits under 1, since is this the exclusion range (where we can say that a model is so unlikely that it no longer merits consideration). From these plots, it is apparent that the models centered around a higher mass Z' are getting closer to the exclusion range, although they are not there yet. With more events from the HL-LHC, the limits on many of these models will move toward and below the threshold for exclusion. At this time, systematic uncertainties are not considered in the limits, so the limits are being recalculated.

What follows is a graphical comparison of the limits gathered through a Bayesian analysis on several variables. Most notable will be the comparison between limits produced by E_T^{miss} , $P_{T4\ell}$, and $D(m_{4\ell}, D_{kin}^{bkg}, E_T^{miss})$.

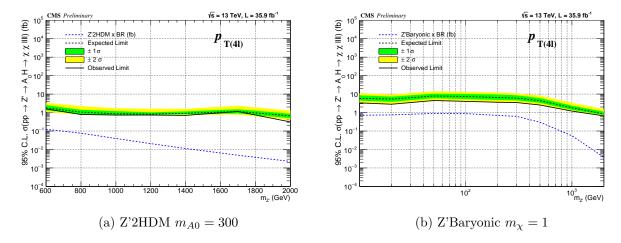


Figure 4: Upper limits on models using $P_{T4\ell}$

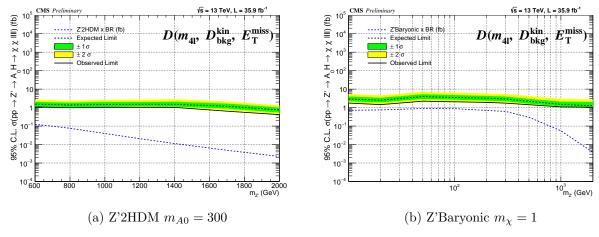


Figure 5: Upper limits on models using $D(m_{4\ell}, D_{kin}^{bkg}, E_T^{miss})$

A quick analysis of these limits shows that the plots for each model retain the same shape, despite the variable used to form the limits. The discriminant does soften the limit curve when compared to the single kinematic factors, which is to be expected. This comparison does, to some degree, shed light on which variable is the most effective for setting limits. While E_T^{miss} is the most commonly accepted kinematic variable, it's high systematic uncertainty leaves room for $P_{T4\ell}$ to be an almost

equally, of not more effective, variable in limit setting. The possibility of $P_{T4\ell}$ as a useful limiter is currently an area of current interest and will continue to be explored by creating discriminants with $P_{T4\ell\ell}$ instead of and alongside E_T^{miss} , and testing their performance.

References

 $[1]\,$ L. Demortier, S. Jain, and H. B. Prosper. Reference priors for high energy physics. , 82(3):034002, August 2010.