

Use of Machine Learning for gamma/hadron separation with HAWC

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Abstract

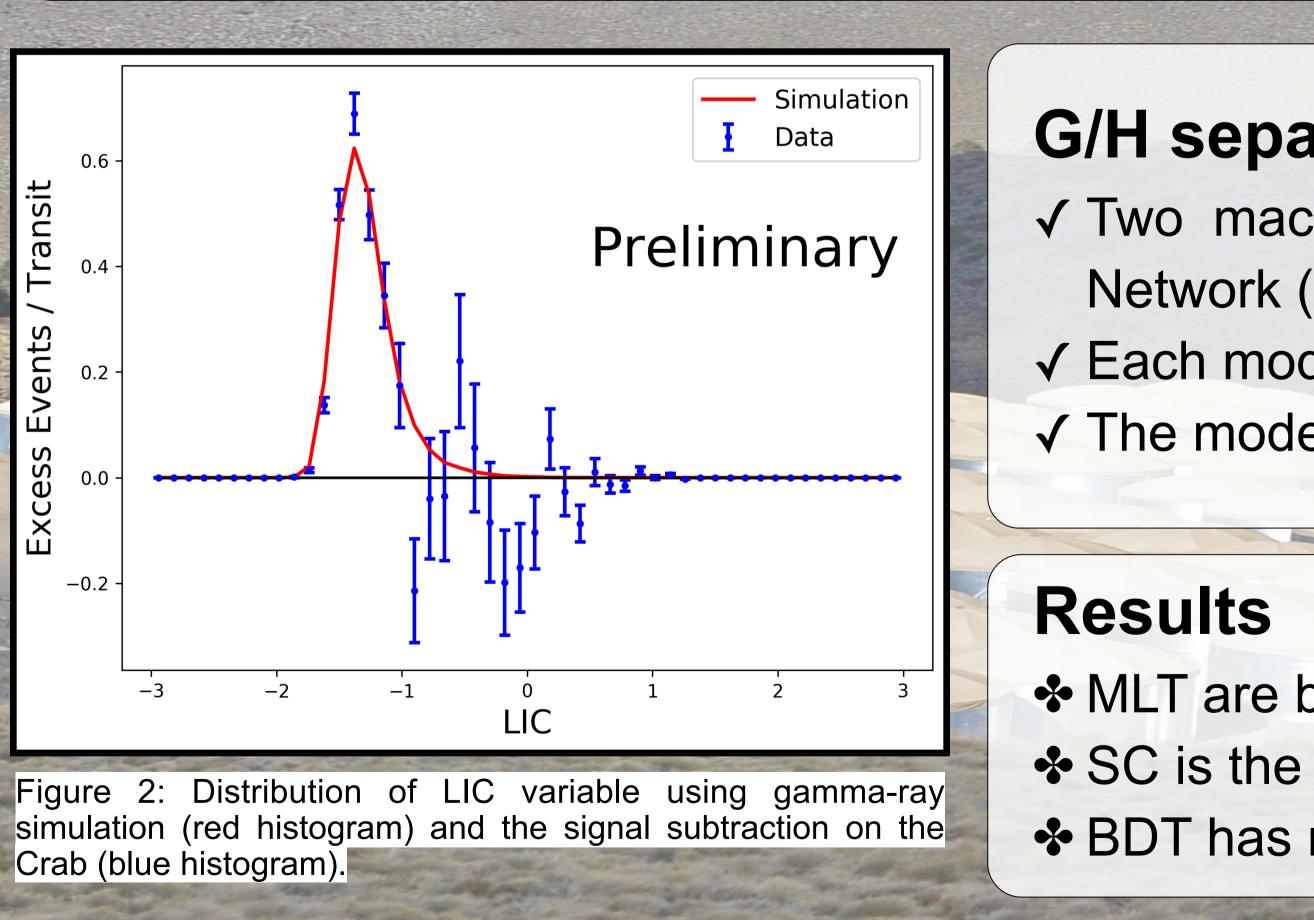
Background showers triggered by hadrons represent over 99.9% of all particles arriving at ground-based gamma-ray observatories. An important stage in the data analysis of these observatories, therefore, is the removal of hadron-triggered showers. Currently, the High-Altitude Water Cherenkov (HAWC) gamma-ray observatory employs an algorithm based on a single cut in two variables, unlike other ground-based gamma-ray observatories (e.g. H.E.S.S., VERITAS), which employ a large number of variables to separate the primary particles. In this work, we explore machine learning techniques (Boosted Decision Trees and Neural Networks) to identify the primary particles detected by HAWC. Our new gamma/hadron separation techniques were tested on data from the Crab nebula, the standard reference in Very High Energy astronomy, showing an improvement compared to the standard HAWC background rejection method.

HAWC

- * Ground-base gamma-ray detector,
- * Detect high energy sources,
- * Located in Mexico.

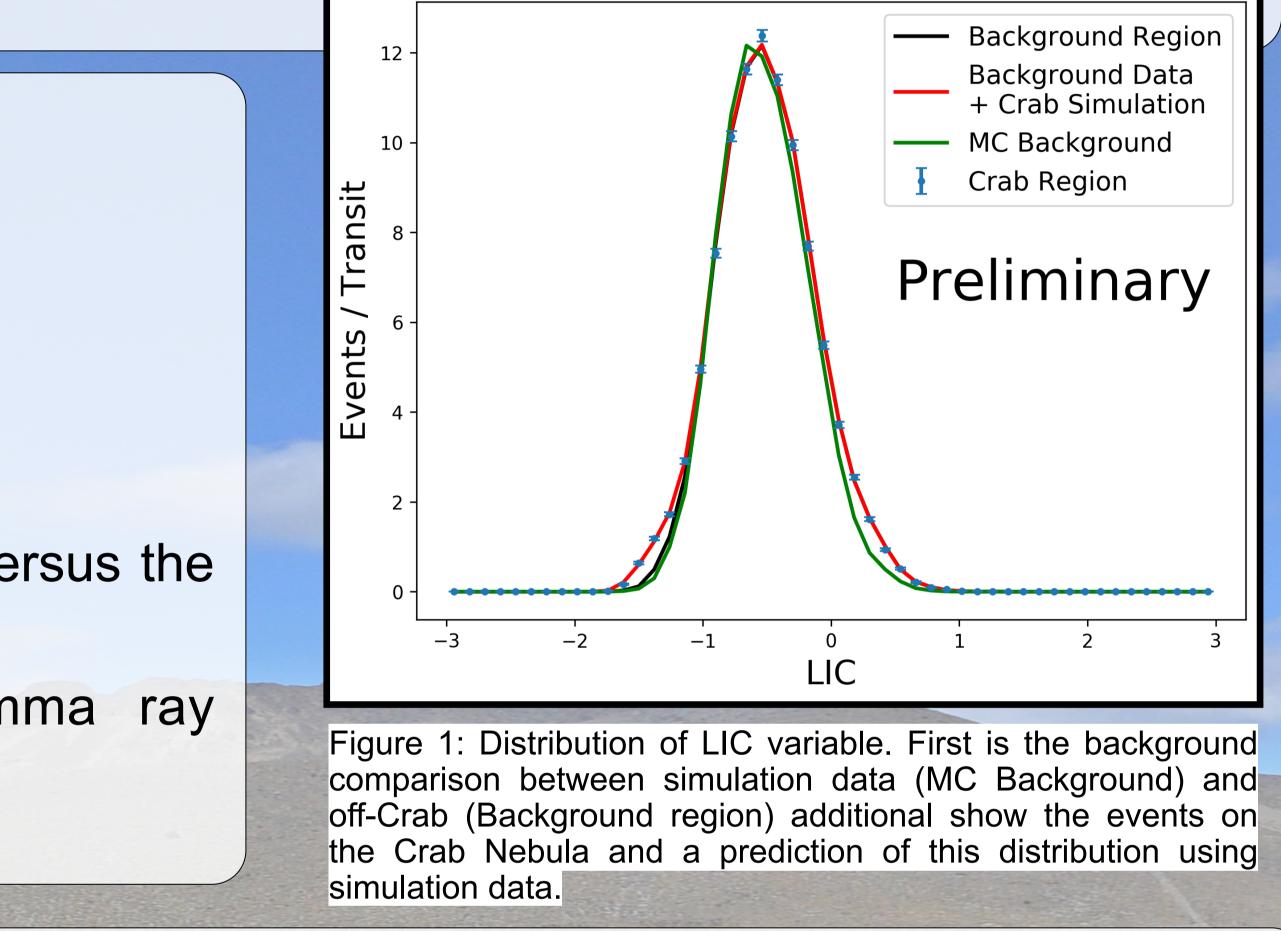
Data/simulation comparison

- * Figure 1 shows the background data near Crab versus the eight nuclei particles simulated for LIC variable.
- * Figure 2 shows the comparison of the gamma ray simulation versus Crab Nebula data.



Acknowledgments

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G/H separation using MLT

√ Two machine learning technique (MLT) were trained Neural Network (NNModel) and Boosted Decision Tree (BDT). ✓ Each model were fed with 7 input variable.

 \checkmark The models were training and testing with simulation.

MLT are better than SC for identifying the gamma events (Figure 3), SC is the best model to recognize the hadron events (Figure 4), ✤ BDT has reported be the best model using the Crab data. (Figure 5 & 6).

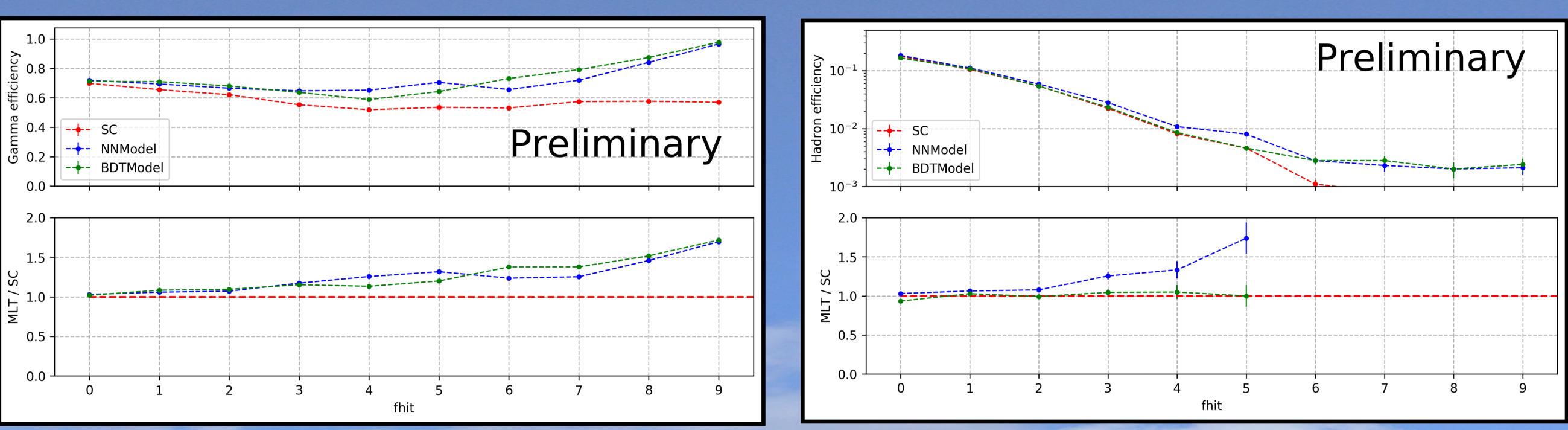


Figure 3: Top panel shows the gamma efficiency for three G/H separation method. The bottom shows the comparison between MLT and SC.

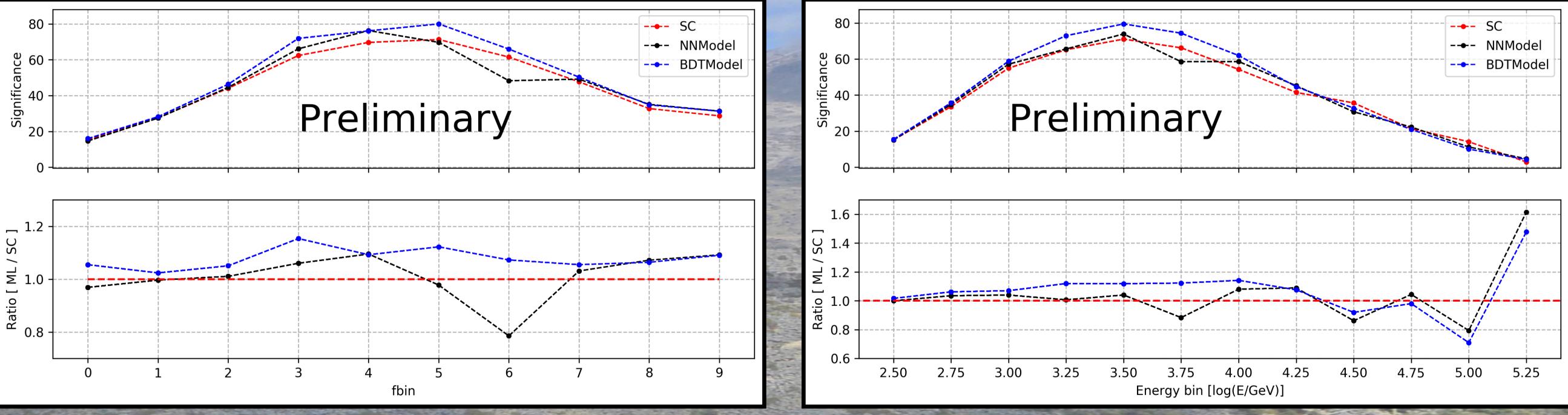


Figure 5: Top panel show the significance on the Crab using the G/H Figure 6: Same as to Figure 5 but here report for each energy bin. separation model as a function of fbin

Conclusion and discussion

Two MLTs were trained using seven input parameters. Both tested with simulation and Crab Nebula data. → Using simulation, both MLTs report higher performance than SC. However, using real data, only the BDT shows that improvement.

References



Figure 4: Same as to Figure 3 but here shows the hadron efficiency.

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