



Application of Classification Trees to Fermi-LAT Data

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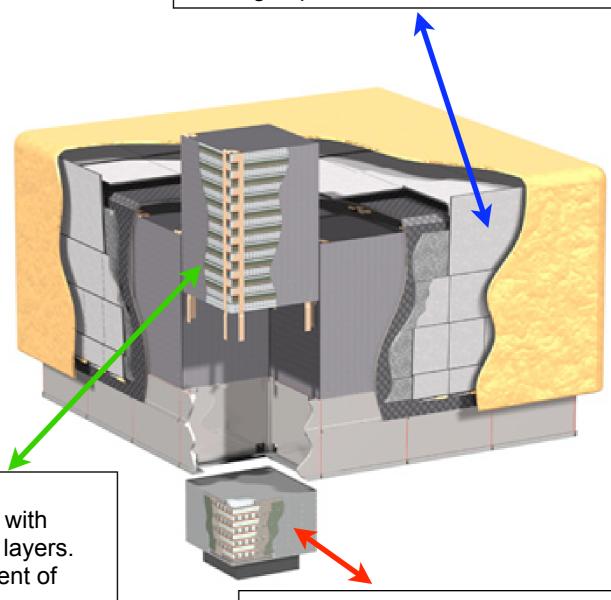
Fermi Large Area Telescope (LAT)



- The Fermi Large Area Telescope (LAT)
 is an e⁺e⁻ pair-conversion gamma-ray
 telescope.
- The LAT is sensitive to photons from 20 MeV to greater than 300 GeV and accepts events over a 2π sr field of view.
- Composed of three detector subsystems:
 - Anti-coincidence Detector (ACD)
 - Tracker (TKR)
 - Calorimeter (CAL)

Anti-coincidence Detector (ACD)

- Segmented tiles of scintillating fiber.
- Charged particle detection.



Tracker (TKR)

- Silicon strip sensor with tungsten conversion layers.
- Precise measurement of event direction.

Calorimeter (CAL)

- Hodoscopic array of CsI crystals.
- Event energy reconstruction.



Reconstruction of LAT Events



- Reconstruction algorithms knit together hundreds of readouts into a physical description of each event.
- We want to answer these questions about the primary particle in the event:
 - Where did the particle come from?
 - What energy did the particle have?
 - Was the the particle a photon?
- The last of these questions is very difficult:
 - As a pair-production telescope, the LAT tracks charged particles making it sensitive to cosmic-ray contamination.
 - At the LAT orbit cosmic rays can out number gamma-rays by more than a factor of 10⁴.
- Separating and quantifying a small signal from a large background is a task ripe for machine learning algorithms...

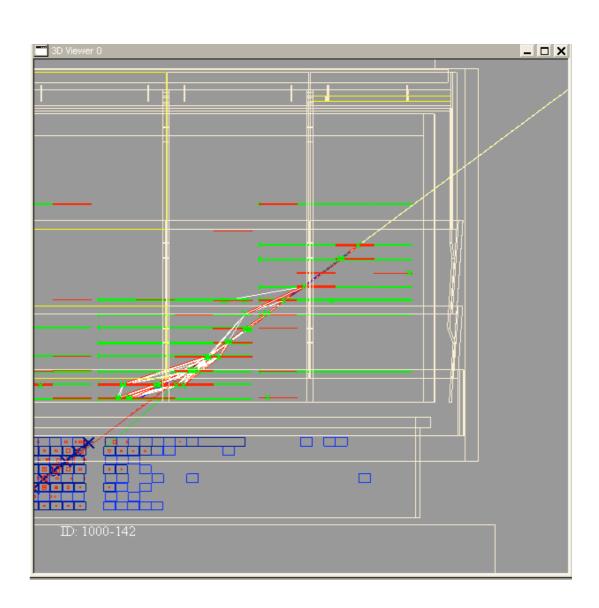


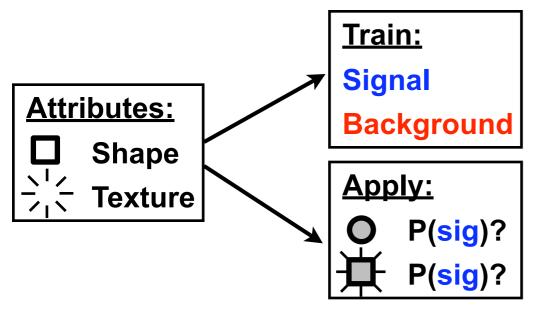
Figure - Reconstruction of an 8 GeV photon event in the LAT.



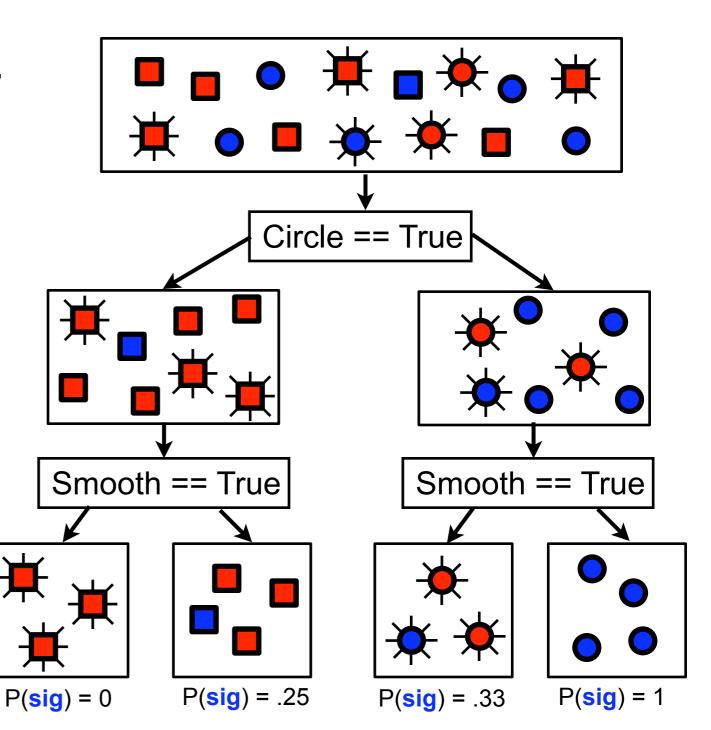
Classification Trees (CTs)



- Machine learning algorithms use known training data to create criteria that can used to classify new data sets.
- Classification Trees (CTs) distinguish signal from background with a set of binary cuts.
 - Maximize separation at each cut.
 - Terminate at a minimum number of events.
 - Find signal probability at end nodes



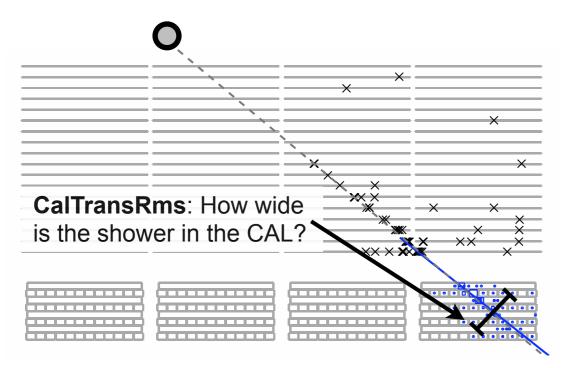
 In most cases, a "forest" of classification trees is used.



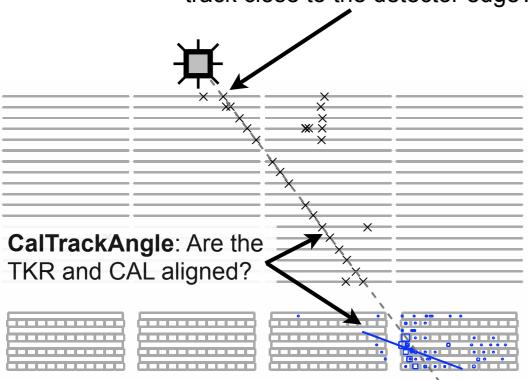


Using Classification Trees





Tkr1LATEdge: Is the head of the track close to the detector edge?



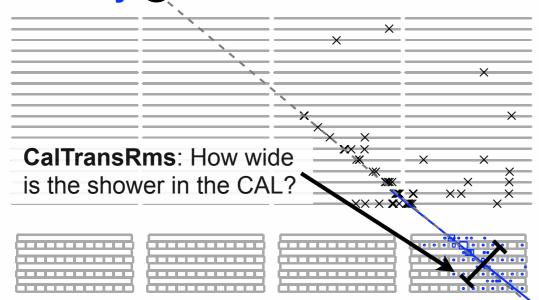
Which one is the gamma ray?



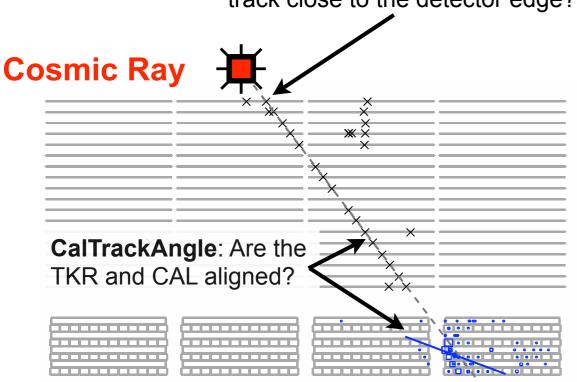
Using Classification Trees







Tkr1LATEdge: Is the head of the track close to the detector edge?



Energy Resolution Knob

- Goal: Distinguish events with well measured energy from those with poorly measured energy.
- Input: TKR and CAL variables as CT input.
- Output: A "knob" that can be adjusted to trade energy resolution for photon efficiency.

Image Resolution Knob

- Goal: Distinguish events lying within the angular containment radius from those outside.
- Input: Mostly TKR track variables.
- Output: A "knob" that can be adjusted to trade image resolution for photon efficiency.

Background Rejection Knob

- Goal: Distinguish photons from cosmic rays.
- Input: Variables from all subsystems.
- Output: A "knob" that can be adjusted to trade photon purity for photon efficiency.



More on Background Rejection



Start with simple cuts

- ACD: Seal gaps, corners and ribbons and reject events that point to hit tiles.
- TKR: Look for vertices (e+ e- pair)
- CAL: Look at transverse shower size.
- Train independent CTs for each subsystem.
- Global Background Rejection CT:
 - Input: Combined information from CAL and TKR (i.e. agreement in direction)
 - Input: The probabilities output from the individual subsystem CTs
 - Output: A single variable which can be used as a background rejection knob.
- Agreement between simulations and data is required. CTs are only as good as the simulated events they are trained on.

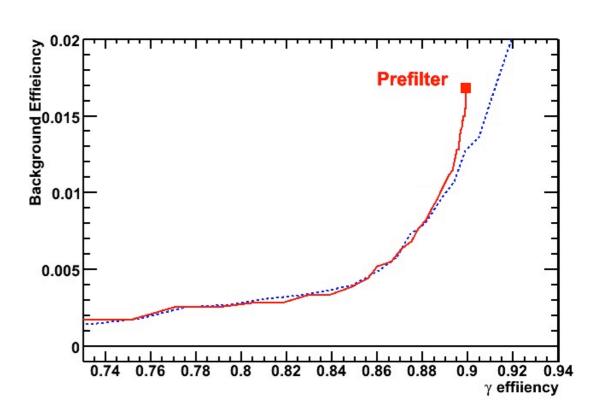


Figure: Example of the background contamination vs. signal efficiency knob derived from classification trees.



TMine Analysis Tool



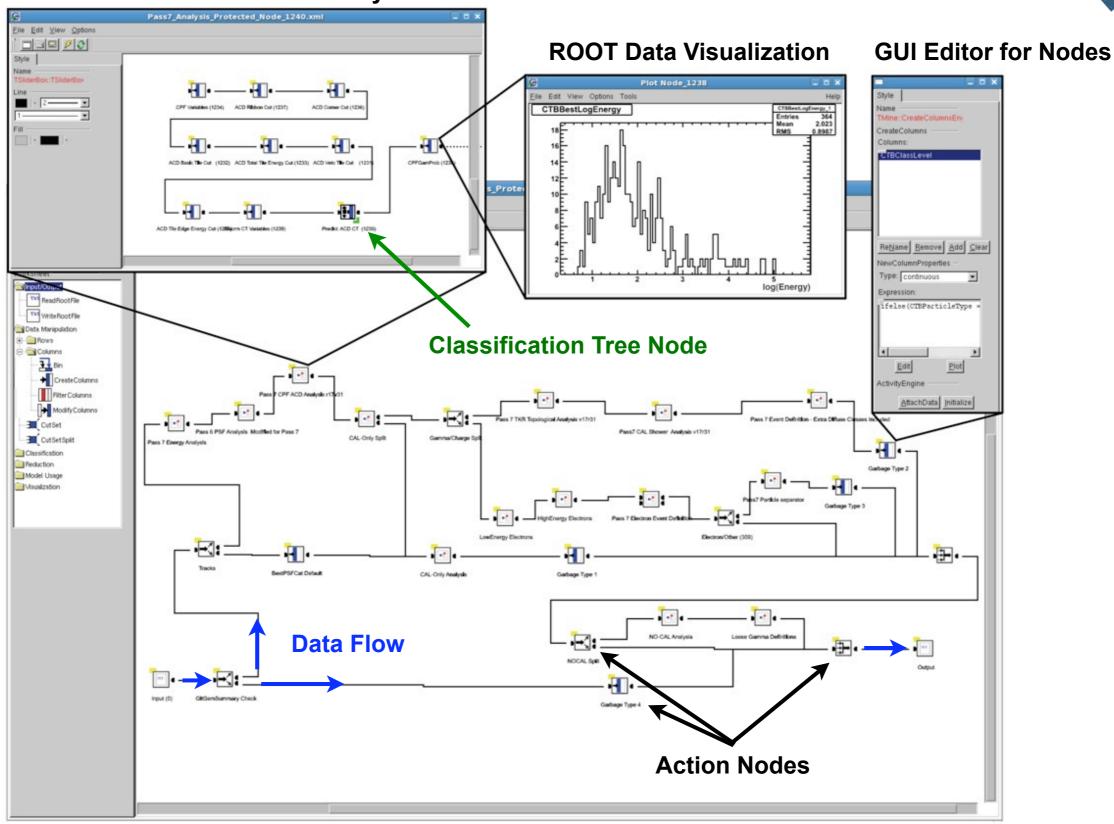
- Historically, the Fermi CTs have been generated with Insightful Miner
 - Proprietary software designed for the financial industry.
 - Uses a visual framework for designing complex analyses.
 - Not optimized for large physics data sets.
 - Does not interface well with the LAT data processing pipeline.
- TMine is a new open source alternative developed here at SLAC
 - Based on ROOT, the de-facto framework for high energy physics experiments.
 - Interfaces with the ROOT Toolkit for Multivariate Analysis (TMVA).
 - Handles larger data sets with increased speed and efficiency.
 - Integrates easily into the LAT pipeline.
- TMine goes beyond classification trees
 - Provides a GUI framework for designing complex analyses
 - Incorporates ROOT data visualization
 - Provides command line interface for batch processing.
- Rather than telling you about TMine, I'll show you...



The LAT Event Analysis with TMine



Subsections of the Analysis





Other Applications of TMine



Measurement of the Electron Spectrum

- The LAT team has published the spectrum of cosmic-ray electrons from 7 GeV to 1 TeV
- TMine was used to reprocess the large data set used for this analysis

Measurements of the Proton Spectrum

- Ongoing work here at SLAC
- TMine used to distinguish protons from leptons and heavy nuclei.

Analysis of Unassociated Sources

- Ongoing work here at SLAC
- Could TMine help to identifying some unassociated LAT sources?
- More from Maria Elena Monzani

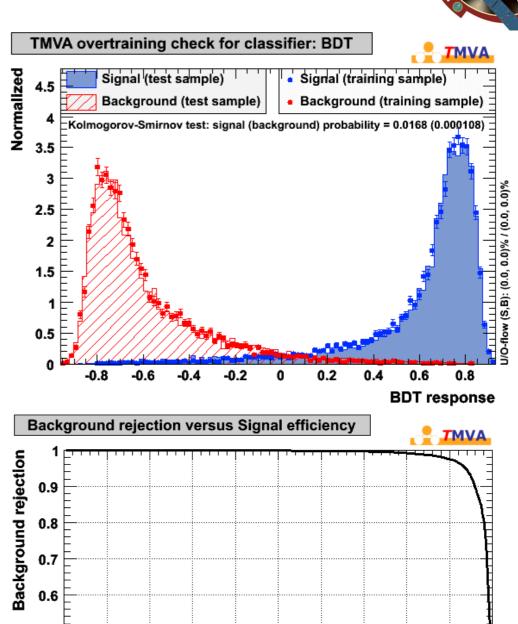


Figure - TMine implementation of the TMVA boosted decision trees to distinguish protons from electrons and positrons.

0.6

0.7

0.8

Signal efficiency

0.5

0.4

MVA Method:



Summary



- The problem that the Fermi-LAT faces is difficult
 - Cosmic rays out number photons by 10⁴ to 1.
 - Events must be characterized over a wide range in energy and angle.
- Classification trees are crucial for achieving mission science goals.
- TMine is a new tool for implementing LAT classification analyses
 - Open source and ROOT-based
 - Implements powerful TMVA package
 - Provides a GUI for designing complex analyses
- The future of TMine
 - Contribute to the ongoing effort to improve the LAT event analysis.
 - Help extend multivariate analyses to high level LAT science.



References



- W. Atwood. 2011. "The Utilization of Classifications in High Energy Astrophysics Experiments." Private Communication.
- TMVA Users Guide. 2009. http://tmva.sourceforge.net/docu/TMVAUsersGuide.pdf
- R. Brun and F. Rademakers. 1997. Nucl. Inst. & Meth. in Phys. Res. A 389, 81-86.
- A. Hoecker, et al. 2007. PoS ACAT 040
- M. Ackermann, et al. 2010. "Fermi LAT Observations of cosmic-ray electrons from 7 GeV to 1 TeV". Phys. Rev. D 82, 092003



Extra Slides





Toolkit for Multivariate Analyses



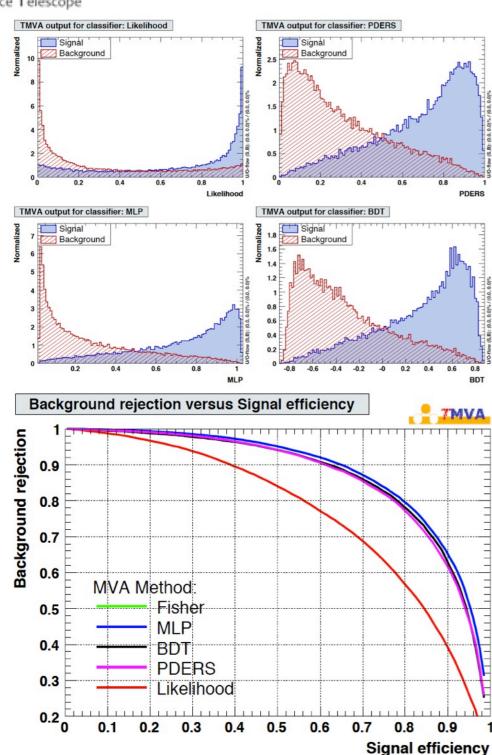


Figure: Toy comparison of some of the TMVA classifiers (TMVA Users Guide)



- TMVA is the multivariate engine under the TMine hood.
- ROOT-integrated environment for processing and evaluating multivariate classification and regression techniques.
- Developed at CERN for signal discrimination in large data sets.
- Comes standard in current ROOT distributions with multiple classification algorithms fully implemented.