



#### Application of Classification Trees to Fermi-LAT Data

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- The Fermi Large Area Telescope (LAT) is an e<sup>+</sup>e<sup>-</sup> 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)

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- Calorimeter (CAL)

#### Tracker (TKR)

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

**Anti-coincidence Detector (ACD)** - Segmented tiles of scintillating fiber. - Charged particle detection. gamma ray

#### **Calorimeter (CAL)**

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

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#### **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 10<sup>4</sup>.
- Separating and quantifying a small signal from a large background is a task ripe for machine learning algorithms...





**Attributes:** 

Shape

Texture

## **Classification Trees (CTs)**

Machine learning algorithms use known training data to create criteria that can used to classify new data sets.

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- **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

Train:

**Signal** 

Apply:

Background

P(sig)?











#### **Using Classification Trees**



# 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.



• Start with simple cuts

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- 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.



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- 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



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- 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

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- 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.







- The problem that the Fermi-LAT faces is difficult
  - Cosmic rays out number photons by 10<sup>4</sup> 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.







- W. Atwood. 2011. "The Utilization of Classifications in High Energy Astrophysics Experiments." Private Communication.
- TMVA Users Guide. 2009. <u>http://tmva.sourceforge.net/docu/TMVAUsersGuide.pdf</u>
- R. Brun and F. Rademakers. 1997. Nucl. Inst. & Meth. in Phys. Res. A 389, 81-86.
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**Toolkit for Multivariate Analyses** 





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- 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.

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