



TMVA Tutorial

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for the ROOT-TMVA Team**

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Outline

- **Status and Overview**
- **New TMVA Features**
 - **External Interfaces**
 - **Deep Learning, Jupyter, Parallelization**
- **Future Plans and Outlook**
- **Summary**

TMVA

Toolkit for Multivariate Analysis:

- **HEP Machine Learning workhorse**
- **Part of ROOT**
- **In LHC experiments production**
- **Easy for beginners, powerful for expert**
- **New TMVA version released in the ROOT version 6.0.8**



New TMVA Features

Modularity, External Interfaces, Updated SVMs
Analyzer Tools: Variable Importance

Added beginning of 2016

Deep Learning CPU, GPU
Parallelization with multithreading and GPUs
Analyzer Tools: Cross-Validation, Hyper-Parameter Tuning

Added in TMVA ROOT 6.0.8

Regression Loss Functions
Jupyter: Interactive Training, Visualizations
Unsupervised Learning
Deep Autoencoders
Multi-processing, Spark parallelization

Upcoming

TMVA Interfaces

Interfaces to External ML Tools

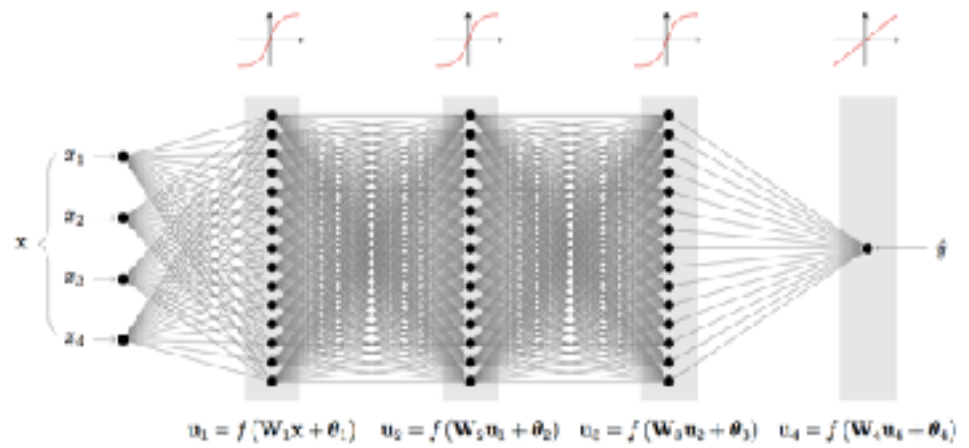
- **RMVA** interface to R
- **PyMVA** interface to scikit-learn
- **KMVA** interface to Keras
 - High-level interface to Theano,
TensorFlow deep-learning libraries



Deep Learning

New Deep-Learning Library in TMVA

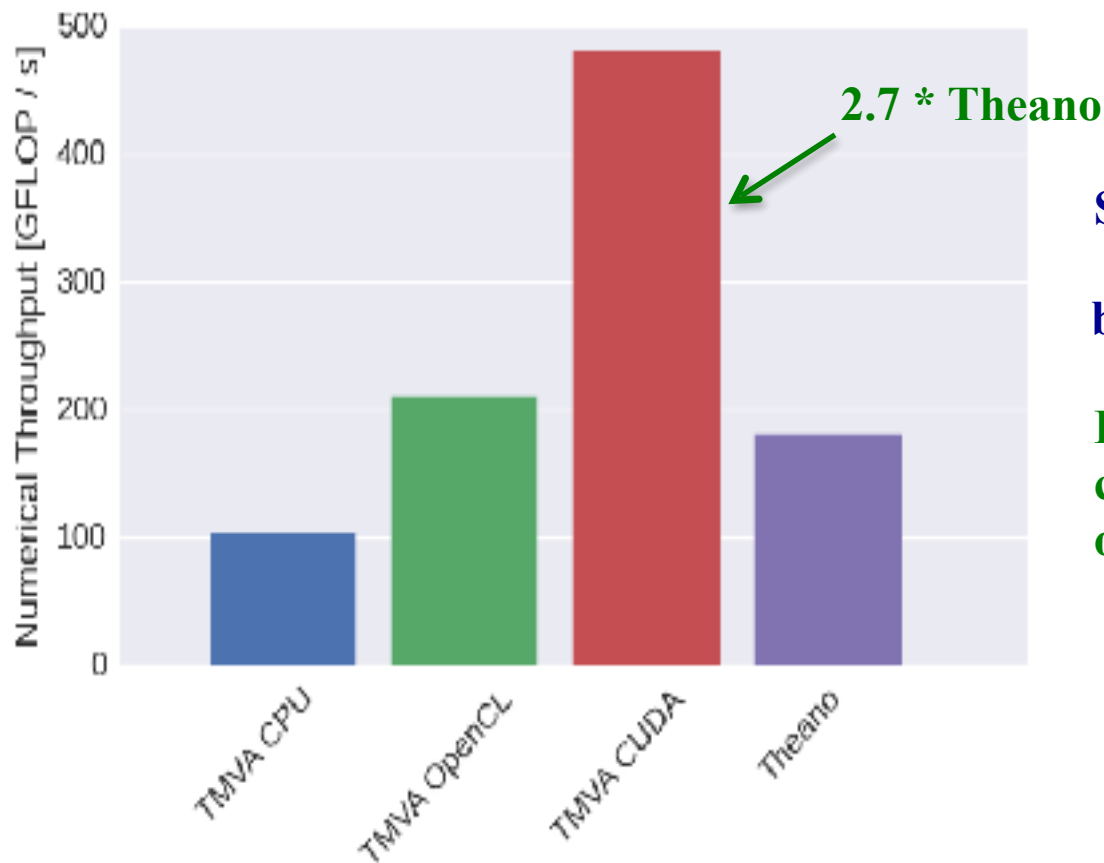
- **GPU support**
 - CUDA
 - OpenCL



- **Excellent performance and high numerical throughput**

Deep Learning

Throughput Comparison



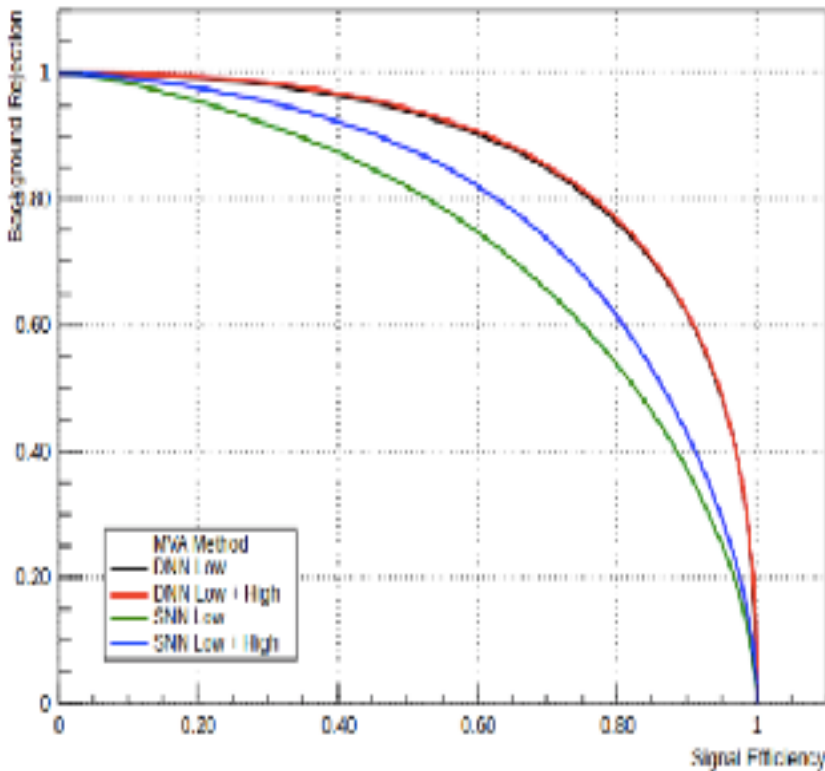
Single precision

batch size = 1024

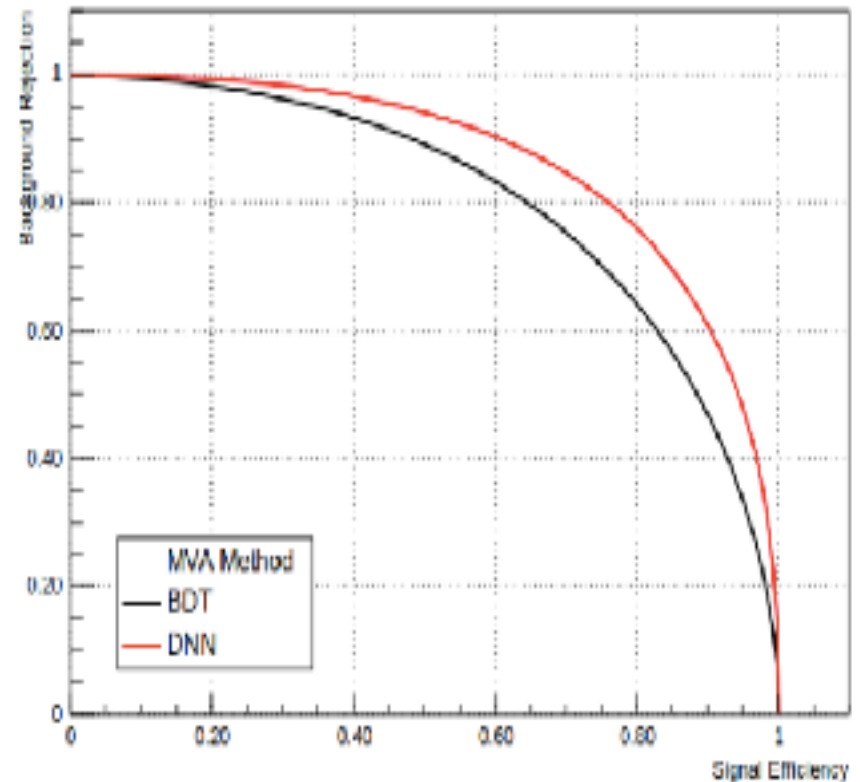
Excellent throughput
compared to Theano
on same GPU

Deep Learning

Background Rejection vs. Signal Efficiency



Background Rejection vs. Signal Efficiency



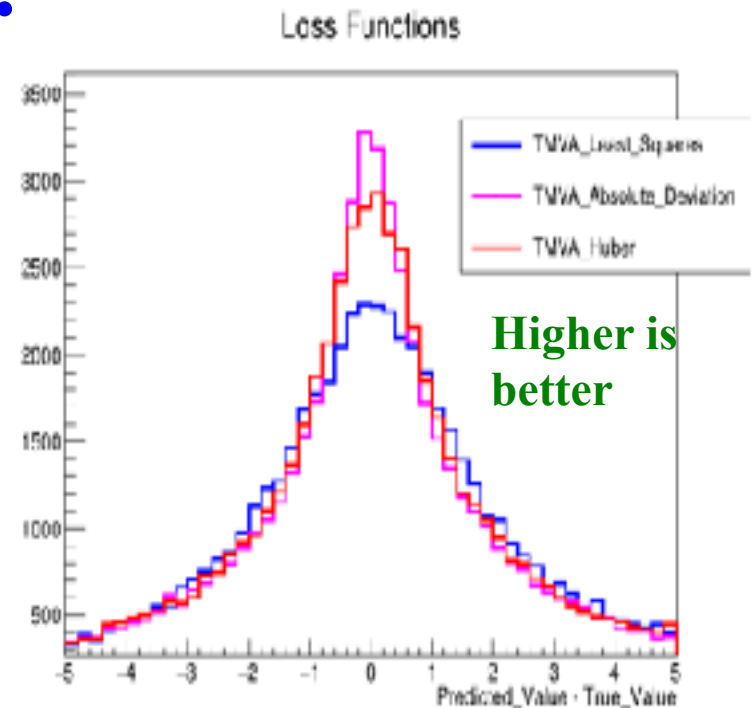
ROC Performance: significant improvements compared to shallow networks and boosted decision trees

Regression

New Regression Features:

Loss functions:

- Huber (default)
- Least Squares
- Absolute Deviation
- Custom Function



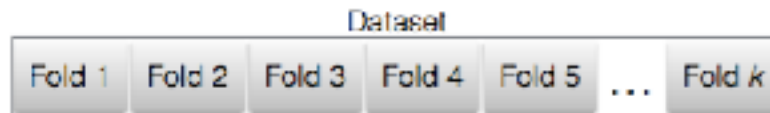
Important for regression performance

Cross Validation

New features:

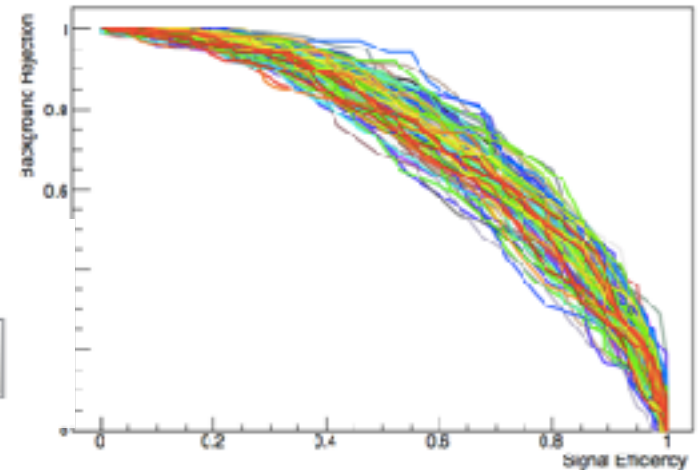
- **k-fold cross-validation**

k-fold cross-validation:



- **Hyper-parameter tuning**

– Find optimized parameters (SVM, BDT)



TMVA Tutorial

- **Run tutorial on notebook**
 - use **SWAN**
 - go to swan.cern.ch
 - **or running local notebooks**
 - `root —notebook`



If you don't have CERN account for using SWAN please contact me
Some temporary account can be made available

Starting SWAN

make sure to have selected
the development version

SWAN Customisation

Specify the parameters that will be used to contextualise the container which is created for you. See [the online SWAN guide](#) for more details.

[Software stack more...](#)

Development Bleeding Edge (might be unstable)

[Platform more...](#)

x86_64-elc6-gcc49-opt

[Environment script more...](#)

e.g. \$CERNBOX_HOME/MySWAN/myascript.sh

[Number of cores more...](#)

1

Start my Session

click here to start

Starting a Terminal in SWAN

After login cernbox home directory will be visible



Start a terminal window



Getting the Notebooks

- Clone the git repository
 - `git clone https://github.com/lmoneta/tmva-tutorial.git`
(link available also in Indico)



```
bash-3.2$ git clone https://github.com/lmoneta/tmva-tutorial.git
Cloning into 'tmva-tutorial'...
remote: Counting objects: 15, done.
remote: Compressing objects: 100% (11/11), done.
remote: Total 15 (delta 2), reused 15 (delta 2), pack-reused 0
Unpacking objects: 100% (15/15), done.
bash-3.2$
```



TMVA Contributors

- **Sergei Gleyzer** Analyzer Tools, Algorithm Development
 - **Lorenzo Moneta** Multi-threading, Multi-processing
 - **Omar Zapata Mesa** PyMVA, RMVA, Modularity, Parallelization
 - **Peter Speckmeyer** Deep-Learning CPU
 - **Simon Pfreundschuh** Deep-Learning CPU and GPU
 - **Adrian Bevan, Tom Stevenson** SVMs, Cross-Validation, Hyperparameter Tuning
 - **Attila Bagoly** Jupyter Integration, Visualization, Output
 - **Albulena Saliji** TMVA Output Transformation
 - **Stefan Wunsch** KERAS Interfance
 - **Pourya Vakilipourtakalou** Cross-Validation, Parallelization
 - **Abhinav Moudhil** Pre-processing, Deep Autoencoders
 - **Georgios Douzas** Spark, Cross-Validation, Hyperparameter Tuning
 - **Paul Seyfert** Performance optimization of MLP
 - **Andrew Carnes** Regression, Loss Functions, BDT Parallelization
- Continued invaluable contributions from Andreas Hoecker, Helge Voss, Eckhard von Thorne, Jörg Stelzer, and key support from CERN EP-SFT Group**



More Information



Websites: <http://root.cern.ch>
<http://iml.cern.ch>
<http://oproject.org>