Statistical Tools in Collider Experiments Multivariate analysis in high energy physics **Exercises**

Nicolas Chanon - ETH Zürich

February 9, 2012

Goal of these exercises: be able to estimate the sensitivity of a search for a small peak over a steeply falling background, using multivariate methods. The following problem is inspired by Higgs searches in $H \to \gamma \gamma$ channel at LHC.

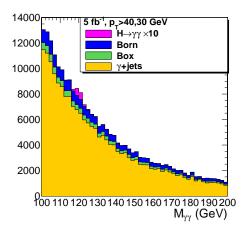


Figure 1: $\gamma\gamma$ invariant mass before photon identification (multijet not generated due to high number of events needed)

Samples provided :

Process	Number of events	weight (for $1 fb^{-1}$)
Pythia $gg \to H \to \gamma\gamma \ m_H = 120 \text{ GeV}$	100k	19.0*0.00223/100000.
Pythia Born	100k	$2.0^{*}22.37/100000.$
Pythia Box	100k	2.0*12.37/100000.
Pythia γ +jet	20M	1.3*19220./20000000.

1 Exercise 1 : TMVA basics

In this exercise, we will install root and start to use the TMVA package. To install ROOT (for these exercises, ROOT 5.28 was used), download the binaries from : http://root.cern.ch/drupal/content/downloading-root

Just unpack it and go in the bin directory. To setup ROOT, do

source thisroot.sh

Download the ROOT samples for the exercise from http://www.phys.ethz.ch/~pheno/Lectures2012_StatisticalTools/ Then go in the *tmva/test* directory and copy the samples there. To look at the samples, first launch root

root -l

Open the browser with $TBrowser \ b$ and open the Sample.root that you want.

You can check the diphoton invariant mass distribution just by running :

```
root -l PlotHggVariables.C
```

You can modify the selection in the macro $\mathit{PlotHggVariables.C}$ provided with the exercises.

Running the training of different classifiers can be done inside ROOT with:

.x TMVAClassification.C

In the file TMVAClassification.C, different classifiers can be tried (rectangular cuts, likelihood, MLP, BDT, SVM....). See the first lines of the code. You can choose what variable to use with *factory->AddVariable*. Lines with *factory->AddSpectator* can be commented Replace the relevant lines to use the samples provided for the exercises :

```
TFile *inputS = TFile::Open( "Signal.root" );
TFile *inputB = TFile::Open( "Background.root" );
// --- Register the training and test trees
TTree *signal = (TTree*)inputS->Get("Tree");
```

```
TTree *background = (TTree*)inputB->Get("Tree");
```

You can comment *factory->SetBackgroundWeightExpression("weight");* You can specify preselection cuts with

```
TCut mycuts = ""; // for example: TCut mycuts = "abs(var1)<0.5 && abs(var2-0.5)<1";
TCut mycutb = ""; // for example: TCut mycutb = "abs(var1)<0.5";</pre>
```

Once a classifier trained, results can be investigated by using the GUI: .x $TMV\!AGui.C$

To look at the background rejection versus signal efficiency plot, one can also directly do:

root -1 TMVA.root
TH1F* histo = _file0->Get("Method_MLP/MLP/MVA_MLP_rejBvsS");
histo->Draw();

Other things can be checked in TMVA.root, produced by the traning.

2 Exercise 2 : Training a MVA for photon identification

- Using Born sample as signal and γ+jet as background, train a MLP and a BDT for photon identification. Compare the results with rectangular cuts method. The photon identification variables are : brem, r9, sumiso03, sumiso04. You can use the trailing photon (γ+jet trailing photon is a jet most of the time)
- How are correlated the variables ? What if applying a preselection on the isolation variables ? Any remark on the *brem* variable ?
- How is improved the performance by vetoing events where the trailing photon is in fact a neutral meson ? (pdgId! = 22) How is changed the MVA output ?
- By the N-1 procedure described in the lecture, find what is the optimal set of variables
- Show if the classifiers are overtrained or not.
- Optimize the architecture of the classifiers
- What is the best background rejection for 90% signal efficiency ?

3 Exercise 3 : Application of the multivariate methods in the analysis

(Needs some knowledge about ROOT)

 Modifying TMVAClassificationApplication. C, produce the samples for gg → H → γγ, Born, Box and γ+jet including the output of the classifier for both of the photons

- Using the weights for the events in the 4 samples, plot the invariant mass of the diphoton system (you can use the macro *PlotHggVariable.C* provided with the exercises). Compute the significance for Higgs boson observation (can be taken as S/\sqrt{B} computed in a ±5 GeV window around the Higgs boson mass) as a function of the cut value on the classifier output. Find the value of the cut maximizing the significance.
- Show the invariant mass distribution before and after applying this cut. What is the purity of each process before and after ?

4 To go further...

- Train a regression to improve the resolution on the photon energy, using η , r9, brem, p_T variables. Apply it to determine the new photon energy and recompute the 4-momentum.
- Train a kinematic classifier with and without invariant mass. Compare the results.
- You can also include as input to this classifier the output of the photon identification.
- Use RooStat package (https://twiki.cern.ch/twiki/bin/view/RooStats/ WebHome) to compute the exclusion limits and p-values.