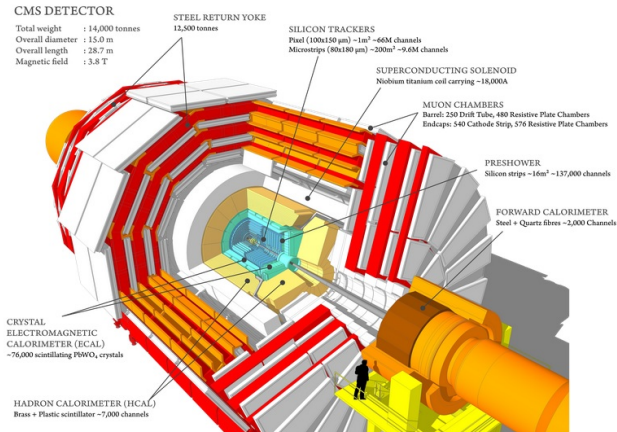


Presentation draft

Konstantinos Papadimos

The CMS Experiment overview

The CMS detector at the LHC



Coordinates at the CMS

Given the solenoid geometry of the CMS detector, it is more convenient to use a spherical type of coordinates(r, ϕ, θ).

$$\begin{aligned}p_x &= P_T \cos \phi \\p_y &= P_T \sin \phi \\p_z &= P_T \sinh \eta \\|\vec{P}| &= P_T \cosh \eta\end{aligned}\tag{1}$$

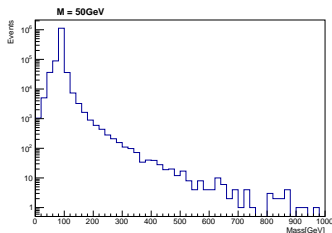
$\phi \in [0, 2\pi]$ the azimuthal angle, and $\eta \in [-\infty, +\infty]$ is defined as:

$$\eta \equiv -\ln \left[\tan \left(\frac{\theta}{2} \right) \right]\tag{2}$$

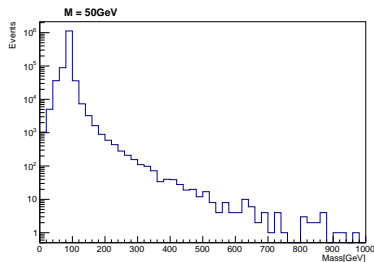
Decays & Resonances

Not every particle can be detected by the CMS detector(i.e neutrinos)

- Detectable Decay Products
→ Resonance



- Non Detectable Decay Products → Not a resonance



Calibration and energy scale uncertainties

- Calibration process adjusts energy scale and resolution to match well-known resonances (Z boson, J/psi meson) in data and simulation,
- Imperfect agreement due to subdetector complexities and nonlinear effects

How do analysis techniques respond to energy scale uncertainties ?

Our work will focus on the effects that energy scale uncertainties have, in a traditional fit-based analysis and a more modern Boosted Decision Tree-based analysis, using the generic diobject production process as the working example.

BDT 1: Supervised Learning

Supervised learning:

- The model is trained using training data
- The trained model is tested using testing data
- If we like the resulting model, we apply it!

but what is this model?

- A function that given the input features x , it returns the probability x being class A
- The goal of the training is to minimize the difference between the predicted output $y_i \in [0, 1]$ and the real output $\hat{y}_i = 0$ class B, or $\hat{y}_i = 1$ class A

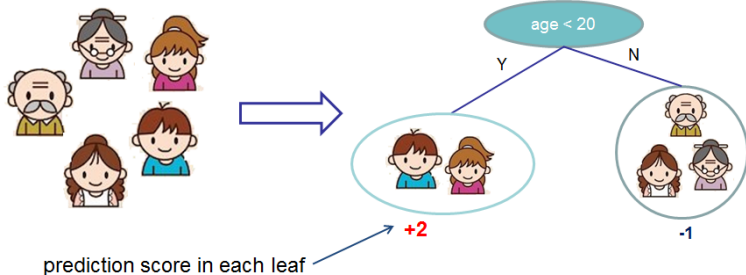
BDT 2a: Boosted decision trees

In this study the model of choice is Boosted Decision Trees(BDT).

- It classifies data using decision tree models

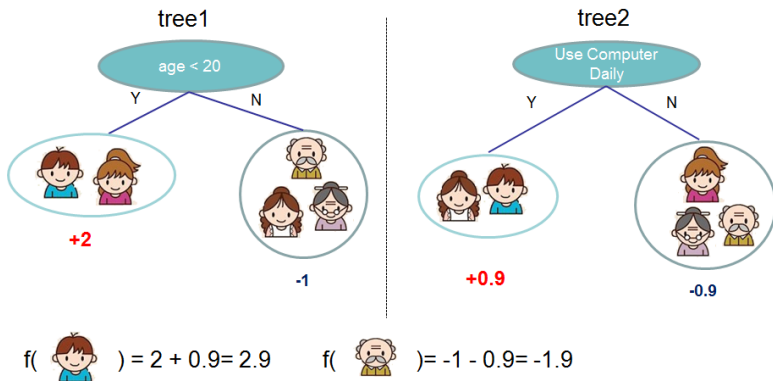
Input: age, gender, occupation, ...

Like the computer game X



BDT 2b: Boosted Decision Trees

Usually only one tree is not power full enough \rightarrow Use more trees in additive manner(Boosting)



BDT 3a: Signal from Background Separation

In our case:

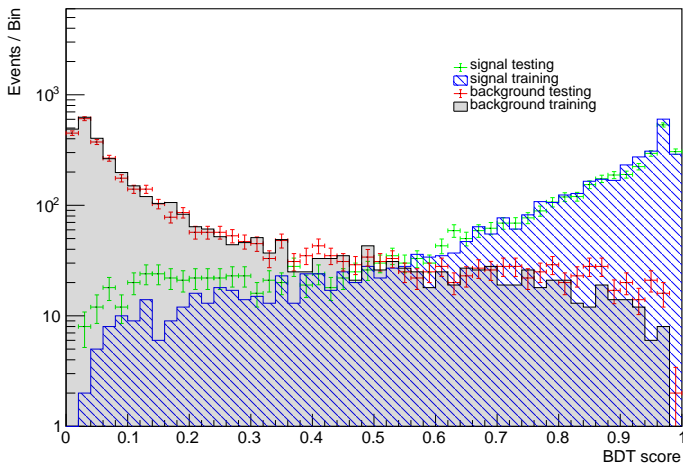
- Signal: a resonant decay $Y \rightarrow xx$
- Background: a non resonant process

How to separate them?

- Plot the number of Signal and Background events per BDT score \rightarrow BDT histogram

BDT 3b: Signal from background separation

Where should we place the cut in order to accept most of the signal while rejecting most of the background?



Explain fit based signal from backgroun separation

Statistical interpretation of results

Talk about significance.

Energy scale uncertainties

How we implemented the smearing in our data set. How do we proceed from that, how many smearing cases.

BDT approach 1

Train Testing application set. Summarize the number of events.
Explain that in order to compare apples to apples, we will be
analyzing the application set from now on.

Application summarize the results

Fit based approach 1

Show the mass spectrum that will be fitted

Fit based approach 2

discuss bkg fit is kept constant throughout the analysis. discuss signal fitting, show the plots(I will probably need more than one slide) at this part talk about the fact that after 20% the fit based technique fails.

Fit based approach 2

Present the significances.

Results 1

Compare the BDT and Flt in terms of significance and robustness. Comment that even though fit based achieves a higher significance in the 0 smearing case, it is not as robust as bdt, it completely fails at extreme cases of smearing,. BDT is more robust

Results 2

Try to explain that bdt uses not only energy related features (Pts) but also geometrical ones, which do not get affected by smearing. Therefore, more stability to smearing. Nevertheless robustness does not mean greater classification "power" (how many events got classified correctly and how many didn't) → Outlooks for better training methods in order to increase classification power.

and therefore, the invariant mass calculation from the detected particles of such events will not result in a peak at the mass spectrum(Non resonant proces). Even though in decays where the poducts are detectable particles, the invariant mass calculation leads to a peak in the mass spectrum(resonant decays). In the present work we are interested in the later.