

Study of the Higgs boson in the WW channel

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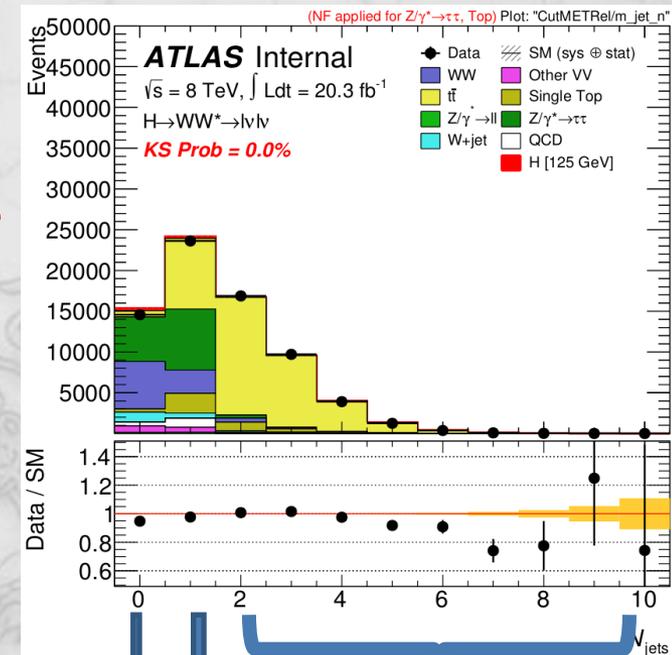


Outline

- ◆ Rate analysis in HWW:
 - ◆ top background study in 0jet channel
- ◆ Spin analysis in HWW
 - ◆ MVA optimization in 1jet channel
 - ◆ top background study 1jet channel
- ◆ Muon fake rates study
 - ◆ Fake muon suppression using BDT

Rate Analysis in HWW Channel

- ◆ HWW channel: Higgs $\rightarrow WW^{(*)} \rightarrow l\nu l\nu$
- ◆ HWW channel is divided into **3 sub-channels** according to the **jet multiplicity in the final states** (right plot):
 - ◆ 0jet, 1jet and VBF channels
- ◆ Rate analysis is to **measure H production rate** and test if it agrees with **SM prediction** by a likelihood fit
- ◆ Main backgrounds: **di-boson**, **top**, **Z+jets** and **W+jets**



0jet
 1jet
 ≥ 2 jets(VBF)

Top Background in 0jet Channel of Rate Analysis

- ◆ Top process is one of the main backgrounds in 0jet channel

Process	Signal(MC)	Di-boson(MC)	Top(MC)	Z+jets(MC)	W+jets&QCD(data-driven)	data
0jet channel	303	7961	1216	6627	1682	14340

- ◆ **It is important to normalize its MC prediction with data-driven method**
 - ◆ In principle, there are always 2 bjets in top final states. But we only count jet whose p_T is above 25 GeV, if both bjets are below this threshold, top event fall into 0jet channel.
- ◆ It includes several sub-processes:
 - ◆ ttbar and Wtop(main)
 - ◆ Single top of s/t channel
- ◆ **There are several data-driven methods to determine the normalization of top quark background:**
 - ◆ **Jet Veto Survival Probability(JVSP) method(baseline)**
 - ◆ **Template method and its simplified version**
 - ◆ **In-situ b-tagging efficiency based method(IBEb)**
- ◆ These methods, together with their performance in HWW+0jet analysis will be discussed and compared in the following

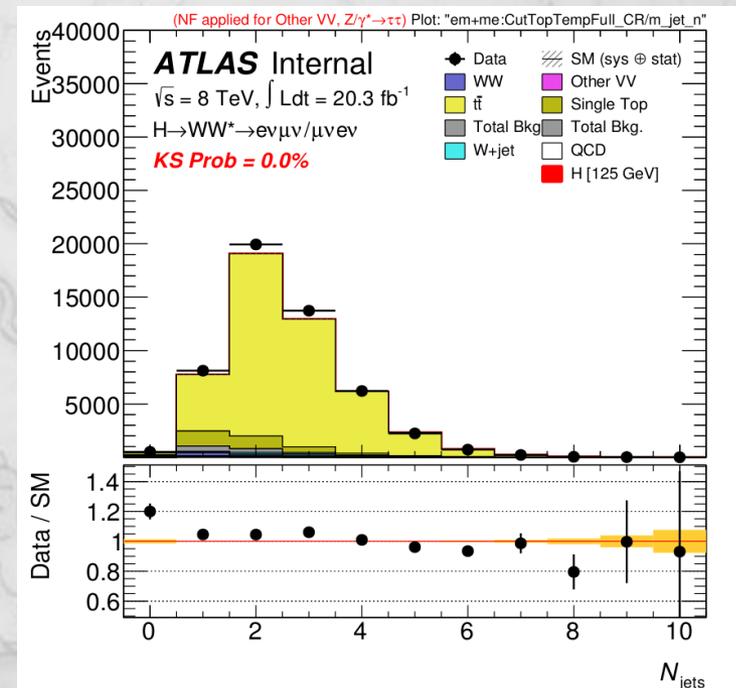
JVSP Method

- ◆ Based on the simple formula below:
 - ◆ $N_{0jet} = N_{all} \times P_{0jet}$ (1)
 - ◆ N_{all} is the number of top events with all possible jet multiplicity
 - ◆ P_{0jet} is the probability of top events having 0 jet
- ◆ Principle: using data to estimate quantities at the right side of formula (1)
- ◆ For N_{all} , we replace it with the prediction from (data – NonTop MC)
 - ◆ $N_{all} \rightarrow N_{all}^{estimated} = N_{all}^{data} - N_{all}^{NonTop,MC}$
- ◆ For P_{0jet} , we do a data-driven correction on MC predicted P_{0jet}^{MC}
 - ◆ $P_{0jet} \rightarrow P_{0jet}^{estimated} = P_{0jet}^{MC} \times \left(\frac{P_{1,CR}^{data}}{P_{1,CR}^{MC}}\right)^2$, $\frac{P_{1,CR}^{data}}{P_{1,CR}^{MC}}$ is a per-bjet correction (since there are 2, it is squared) calculated in a very pure top CR
- ◆ Finally, we get the data-driven version of formula (1):
 - ◆ $N_{0jet}^{estimated} = N_{all}^{estimated} \times P_{0jet}^{estimated}$

Template Method

- ◆ The goal is to get a **data-driven template of the jet multiplicity distribution** of top bkg in the signal region → its contribution in each sub-channel can be predicted just by **counting events in the corresponding bin**.
- ◆ First, a data-driven top jet multiplicity distribution is extracted in top CR by using data – NonTop MC (plot below)
 - ◆ $T_{top,estimated}^{CR} = T_{data}^{CR} - f \times T_{Nontop}^{CR}$
 - ◆ f is the normalization of non-top MC
- ◆ Then **extrapolate this template from CR to SR** by top MC predicted extrapolation factors:
 - ◆ $T_{top,estimated}^{SR} = \frac{T_{top,MC}^{SR}}{T_{top,MC}^{CR}} \times T_{top,estimated}^{CR}$
- ◆ The normalization factor “ f ” is determined by a likelihood fit in SR using the above template.
- ◆ Finally, data-driven prediction of top events in 0jet channel is the 0jet bin of the template:

$$N_{top,0jet}^{estimated} = T_{top,estimated}^{SR}(0jet)$$



Simplified Template Method

- ◆ Instead of extracting template from top CR and doing a fit, we **draw normalization factor** for 0jet bin directly from **a smaller top CR**:
 - ◆ Defined inside 0jet sub-channel by doing a 20~25GeV btagging.
 - ◆ For 0jet channel, there is no jet with >25GeV, but 20~25GeV is possible.

Process	Signal(MC)	Di-boson(MC)	Top(MC)	Z+jets(MC)	W+jets&QCD(data-driven)	data
0jet & btagged	3	109	297	127	53	687

- ◆ We use the ratio $N = \frac{N_{data-nontop,MC}^{CR}}{N_{top,MC}^{CR}}$ to normalize MC predicted top events in 0jet bin $N_{top,0jet}^{MC} \rightarrow N_{top,0jet}^{estimated} = N_{top,SR}^{MC} \times N$

IBEB Method

- ◆ Based on the formula below:

- ◆
$$N_{top,0jet} = \frac{N_{top,tagged}}{\epsilon_{top,tag}} \quad (2)$$

- ◆ $N_{top,tagged}$ is the number of top events in 0jet channel being tagged to have bjet
- ◆ $\epsilon_{top,tag}$ is the top tag efficiency

- ◆ Same principle as JVSP method

- ◆ For $N_{top,tagged}$, we estimated it using (data – NonTop MC)

- ◆
$$N_{top,tagged} \rightarrow N_{top,tagged}^{estimated} = N_{top,tagged}^{data} - N_{top,tagged}^{NonTop,MC}$$

- ◆ For $\epsilon_{top,tag}$, we do a data-driven correction on MC predicted $\epsilon_{top,tag}^{MC}$

- ◆
$$\epsilon_{top,tag}^{MC} \rightarrow \epsilon_{top,tag}^{estimated} = \epsilon_{top,tag}^{MC} \times \frac{\epsilon_{top,tag,CR}^{data}}{\epsilon_{top,tag,CR}^{MC}}, \quad \frac{\epsilon_{top,tag,CR}^{data}}{\epsilon_{top,tag,CR}^{MC}} \text{ is a correction factor derived in top CR}$$

- ◆ Finally, we get the data-driven version of formula (2) :

- ◆
$$N_{top,0jet}^{estimated} = \frac{N_{top,tagged}^{estimated}}{\epsilon_{top,tag}^{estimated}}$$

Comparison between Each Method

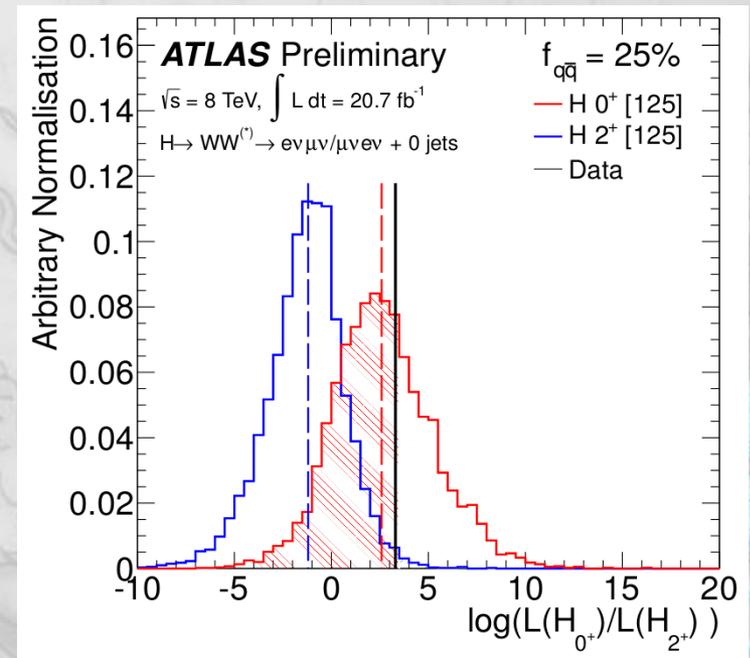
- ◆ Here are the results of each method and their corresponding sys errs

Method	JVSP	Template	Template(Simp)	IBEB
Stat	1.8%	7.6%	6.5%	6.6%
Exp.	4.8%	6.5%	6.5%	1.5%
Theo.	3.0%	3.5%	3.5%	1.8%
NonTop	2.7%	7.1%	7.1%	7.1%
Total sys err	6.3%	10.2%	10.2%	7.5%
Total sys + stat err	6.5%	12.8%	12.1%	10.0%
NF	1.11 ± 0.07	1.32 ± 0.17	1.38 ± 0.17	1.37 ± 0.14
Comment	Most precision Limited by exp err	Except for theo. err All others are large	Except for theo. err All others are large	Least affected by exp. and theo. err

- ◆ There results are preliminary (approximations are used)
- ◆ JVSP is the best (so it's our baseline method)

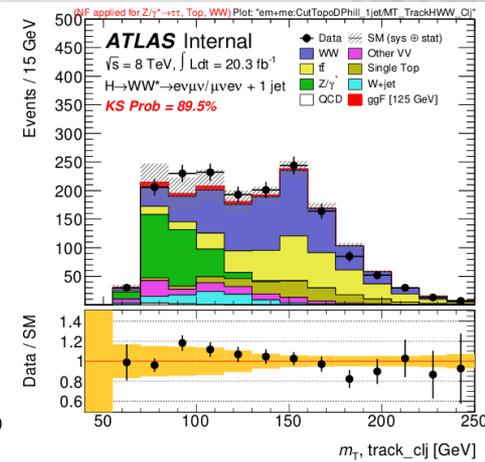
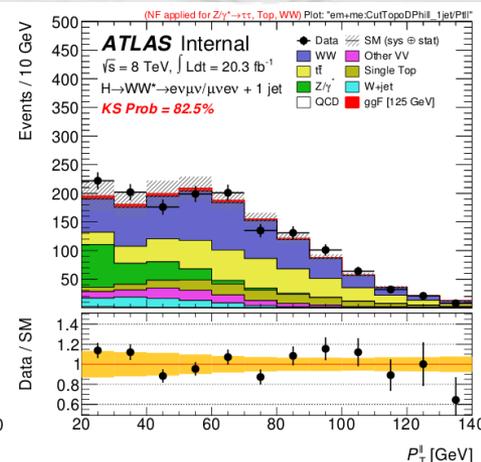
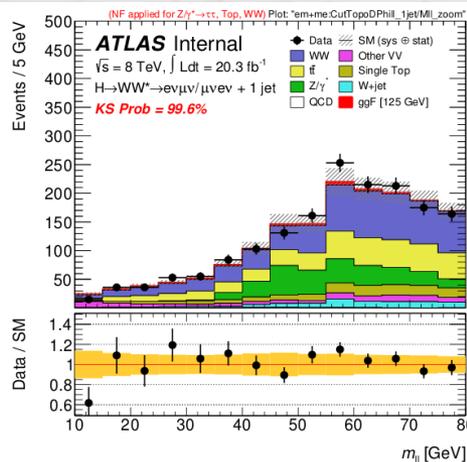
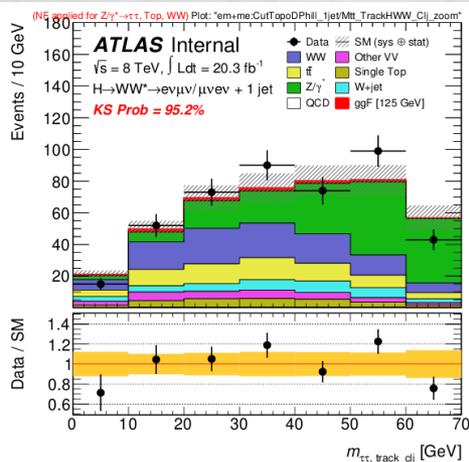
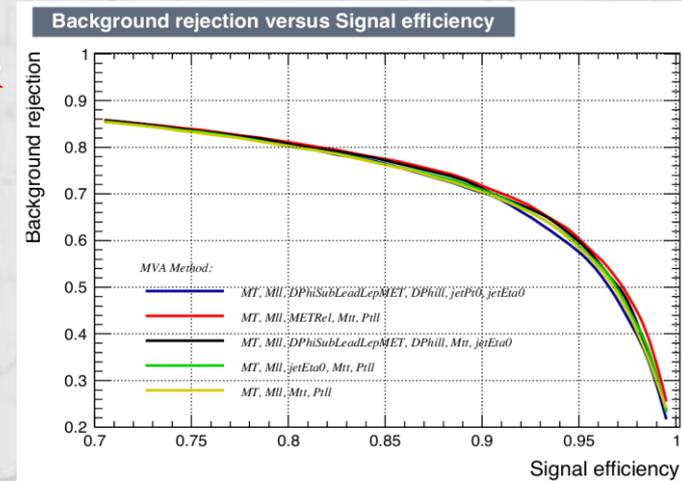
Spin Analysis in HWW Channel

- ◆ **Two spin models are considered: spin0, spin2**
- ◆ Method to test which model is preferable:
 - ◆ Different H spin -> different **topo structure of final state**(angles between objects)
 - ◆ Use quantities describing this structure to train BDT: **2 BDTs are trained**, one from spin0 vs All bkg, the other from spin2 vs All bkg.
 - ◆ **Construct a 2-D BDT and fit it to data.**
- ◆ An example fit shown at right plot.
- ◆ **Different training variables** are used in 0jet and 1jet channels
- ◆ **Only 0jet channel** was used previously



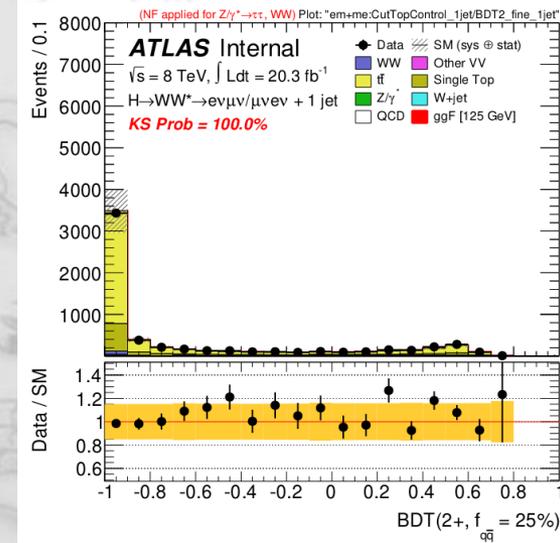
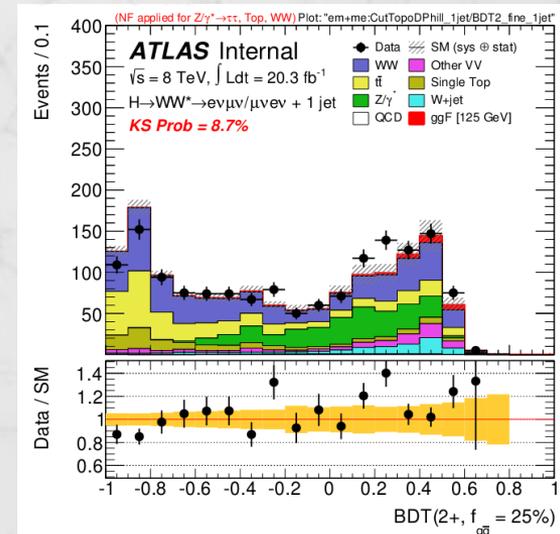
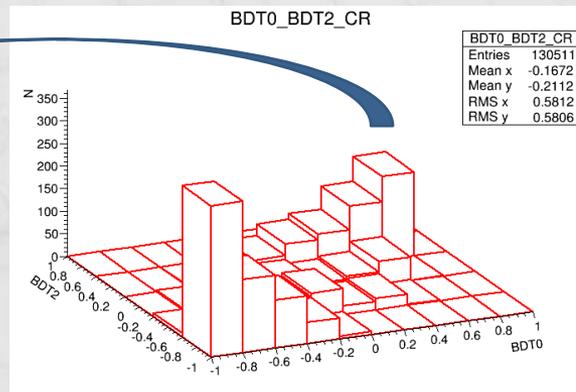
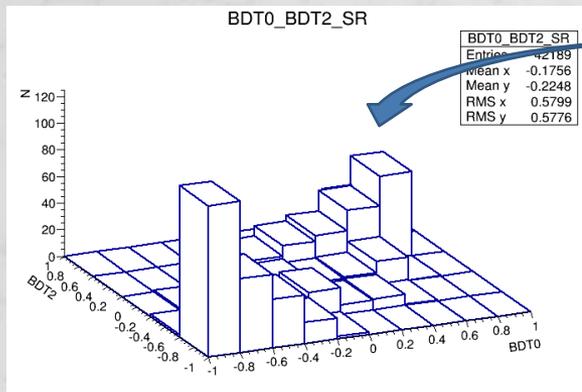
BDT optimization in 1jet Channel of Spin Analysis

- ◆ To include 1jet bin as well, **need optimization**
- ◆ Tried various combination of input variables and get several sets having similar performance(right plot) → **choose the simplest one**, which are shown below:



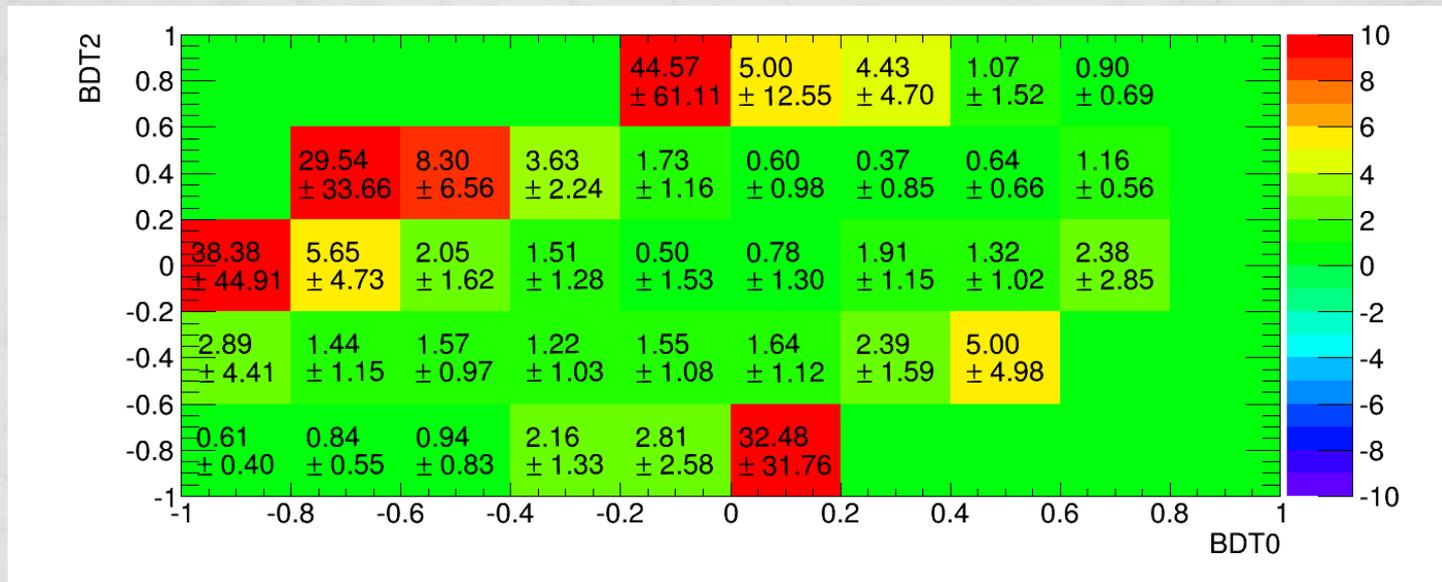
Top Background in 1jet Channel of Spin Analysis

- ◆ Top is one of the main bkg in 1jet channel(right upper plot)
- ◆ We need to **constrain top bkg's shape** in SR
- This can be done by **extrapolating the shape from CR to SR**
- The CR is defined by having non-zero bjet(right bottom plot)
- ◆ Then we need to **evaluate sys errs for the extrapolation factor** $\alpha = N_{SR}/N_{CR}$ in each bin. (graphics below)



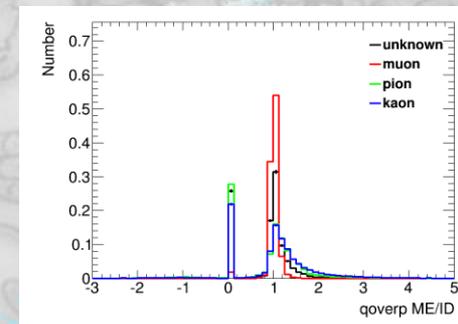
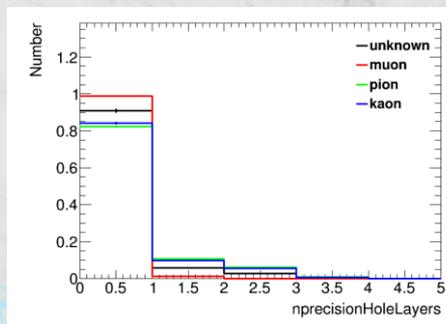
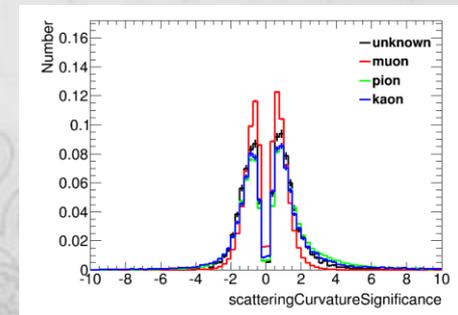
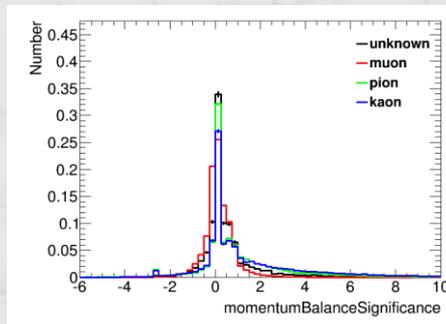
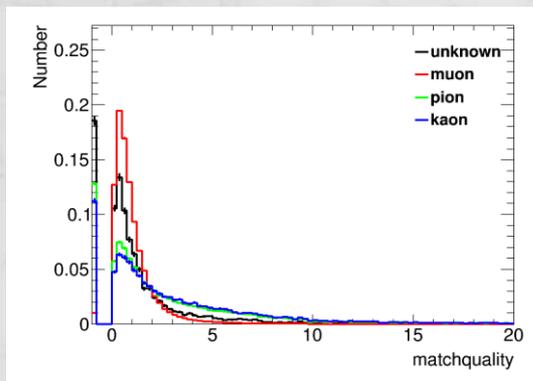
Theoretical Uncertainties of Top Shape Extrapolation

- ◆ Theoretical uncertainties for the extrapolation are studied **using the same group of samples in top 0jet** case (privately produced, large stat $\sim 10000 fb^{-1}$)
- ◆ Take the QCD scale uncertainty as an example here
- ◆ Large stat sample \rightarrow stat uncertainty $\sim 1\%$
- ◆ Stat distribute around diagonal \rightarrow errs on boarder bins can't be estimated precisely \rightarrow but they contributes little to the fit



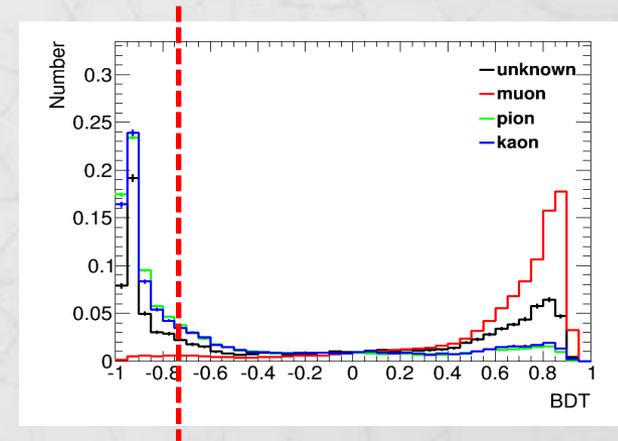
Other activities: Fake Muon Study

- ◆ This work could **contribute to HWW analysis indirectly**: since an quality-improved muon collections will reduce some main background, such as Wjets. Of course **other analysis will also benefit**(Bs to mumu rare decay search)
- ◆ The motivation is the existing working point for 3rd chain muon(the future muon collection that will be used) is **not tight enough**(too many fake muon), so we tried to train a BDT to suppress the fakes
- ◆ 5 variables that are chosen:



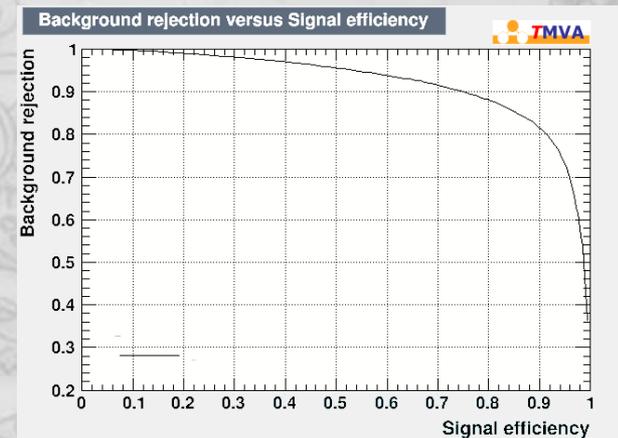
Expected Performance of the BDT

- ◆ Output BDT shapes and its ROC are shown in right plots.
- ◆ Choose the cut on BDT at which we'll have around the same true muon efficiency as the existing working point
- ◆ Comparison with the existing working point in terms of true muon efficiency / fake muon reduction



Effi/Redu.(%)	Muon	Pion	Kaon	Proton	Others	unknown
Medium+	97.87	-37.22	-37.02	-60.07	-55.74	-28.35
BDT	97.83	-59.60	-57.25	-80.78	-82.64	-37.12
Improvement	-0.04	60.12	54.66	34.49	48.25	30.92

- ◆ When having same true muon efficiency, BDT has almost 2 times stronger fake reduction than medium+ does
- ◆ Next step → validation of input variable shapes using data in Z- \rightarrow mumu control region
 - ◆ Some discrepancies are observed → to be understood and fixed.



Summary

- ◆ In HWW+0jet rate analysis, we normalized the top bkg using data-driven method(JVSP) and compared other optional methods with it:
 - ◆ JVSP method is the best.
- ◆ In HWW+1jet spin analysis, we optimized the BDT training and constraint top's shape as well as the theoretical uncertainties on this constraint.
- ◆ We have a promising fake muon suppression method which needs further study.

- ◆ We will continue the project with an extended term in 2014:
 - ◆ LAL: Z. Zhang, D. Rousseau
 - ◆ Nanjing U. / Shandong U. / USTC
 - ◆ S. Chen, L. Ma, Y. Zhu
 - ◆ Y. Li (Joint PhD student)