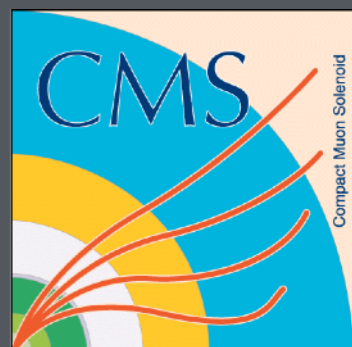
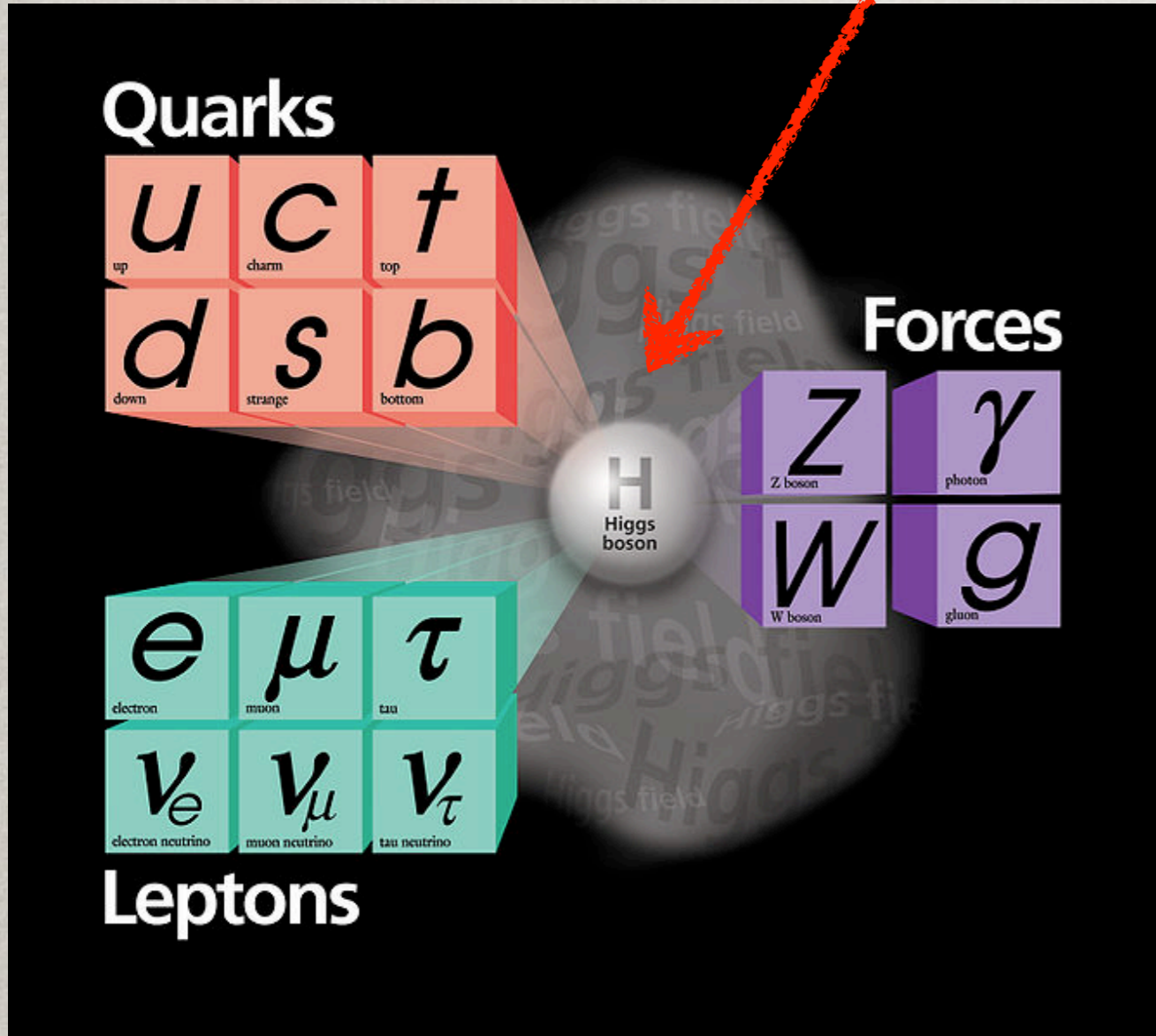


SEARCH FOR TTH(BB) @ CMS

W U M I N G L U O



WHAT IS IT?



I learned very early the difference between knowing the name of something and knowing something.

— Richard P. Feynman —

AZ QUOTES

WHAT IS IT REALLY?

Mr. Somebody

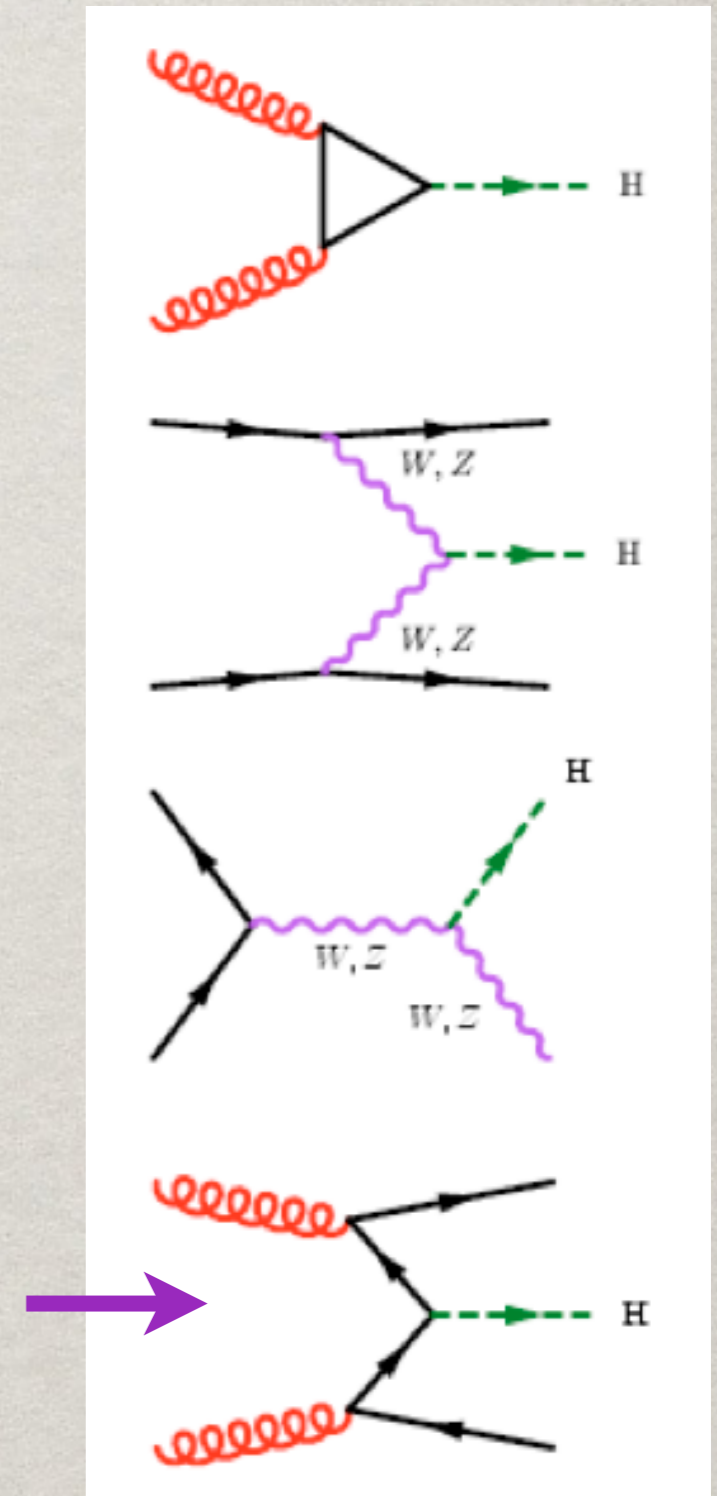
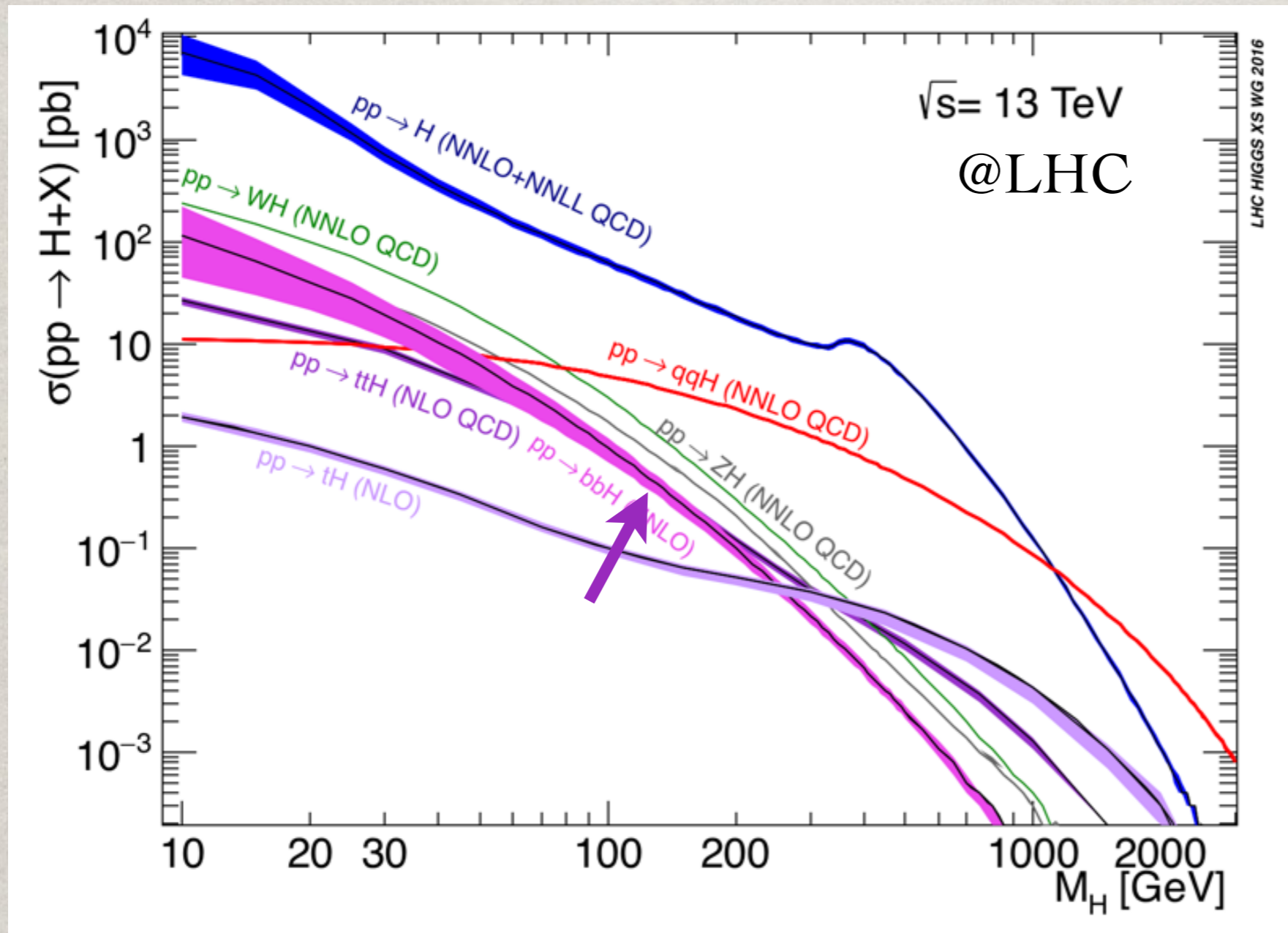


Mr. Nobody

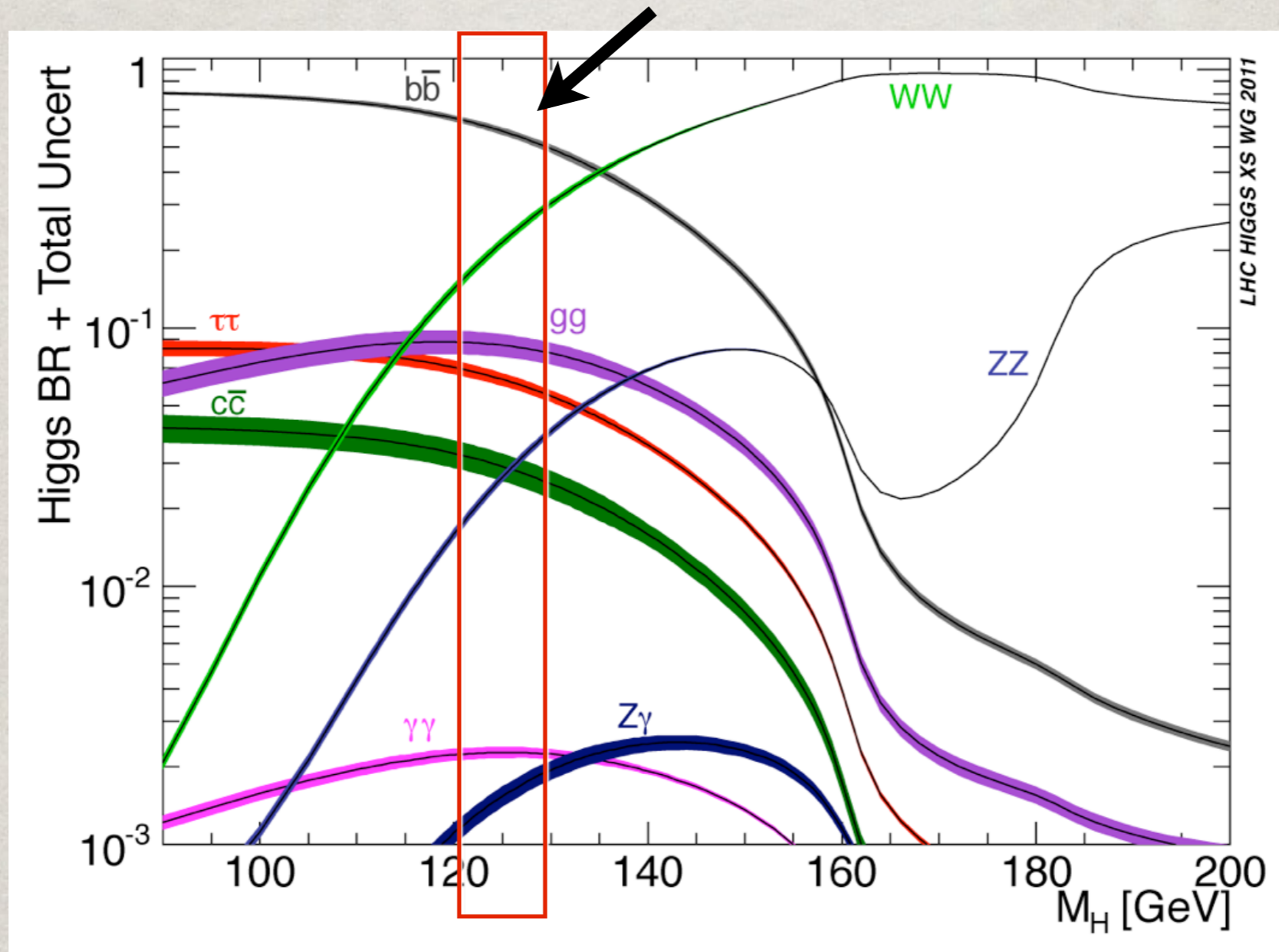


- ✿ Where do the masses of particles come from?
- ✿ Have been searching for the Higgs Boson for many years

WHERE DOES IT COME FROM: HIGGS PRODUCTION

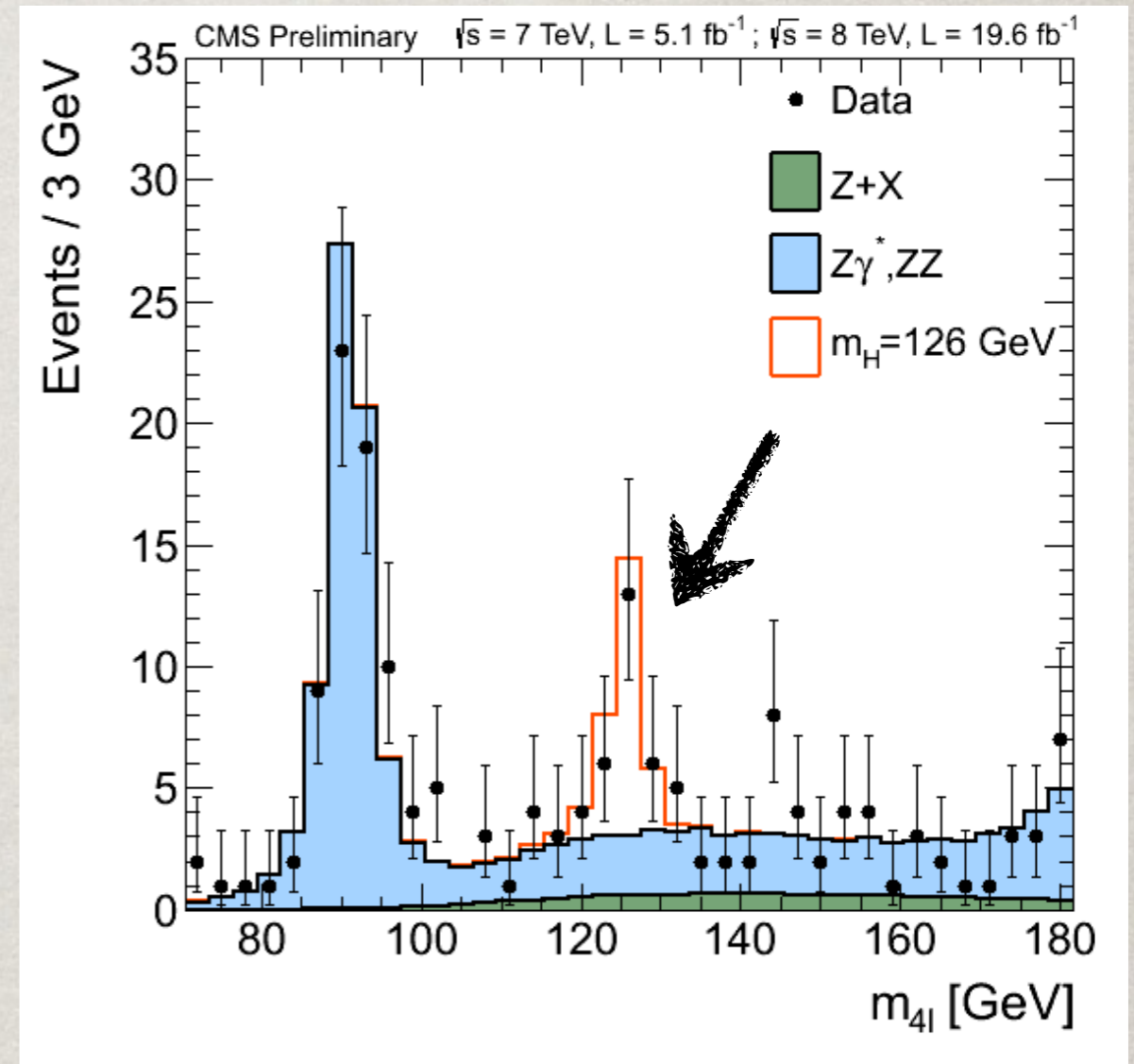
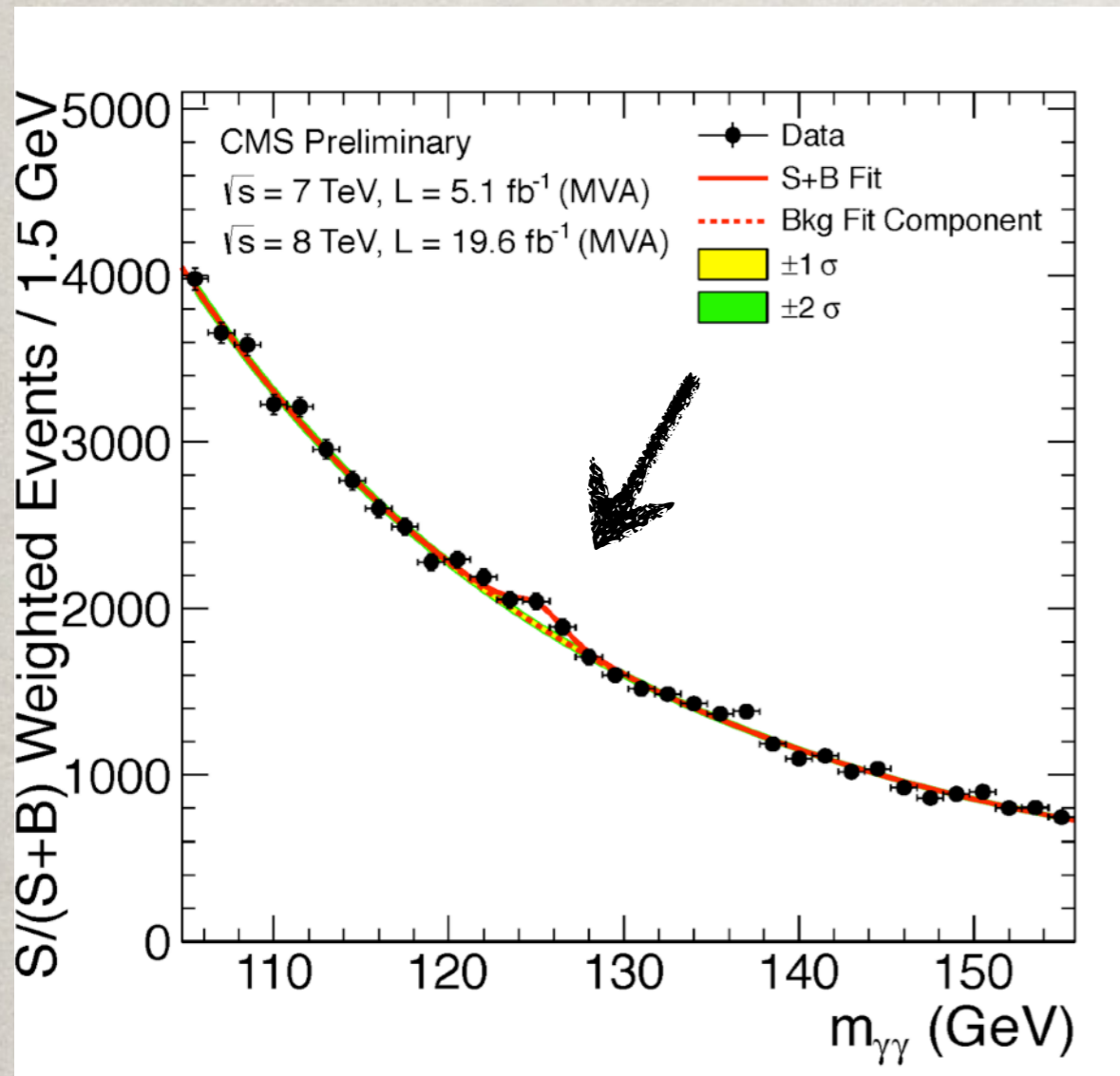


WHERE DOES IT GO: HIGGS DECAY



- ✿ At 125.7 GeV, $H \rightarrow bb$ dominates.
- ✿ Huge background makes direct search difficult.

HIGGS BOSON DISCOVERY

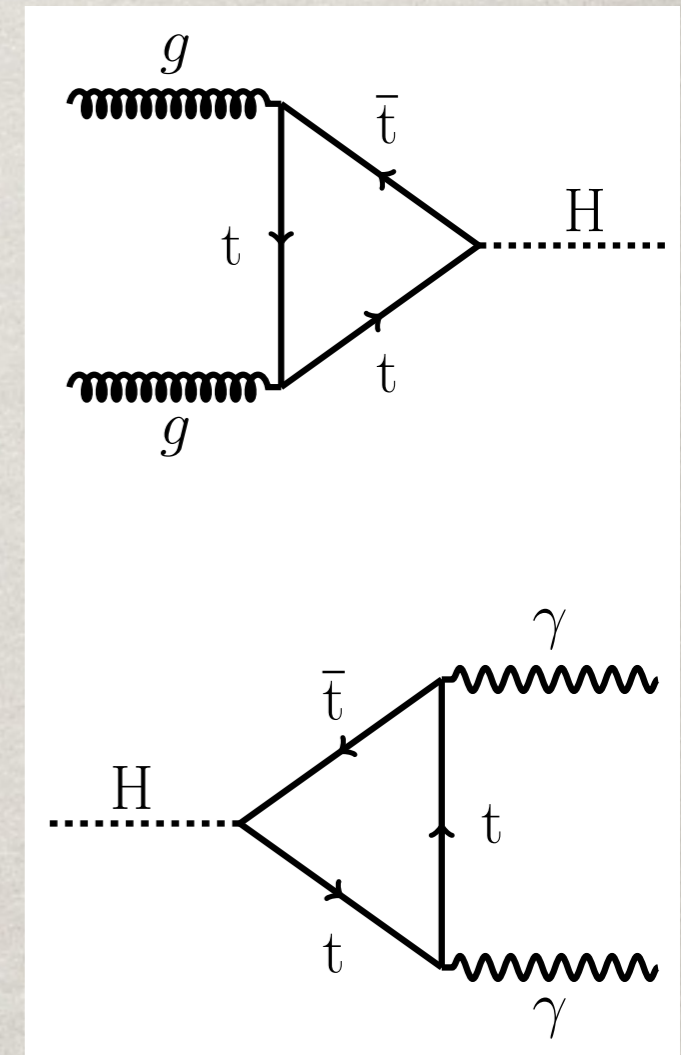


☀ We found “A” Higgs boson!

☀ Is it truly “The” SM Higgs Boson???

WHY TTH ?

- ☀ Complementary to other H searches
- ☀ Directly probes the top-Higgs Yukawa coupling (Y_t)
 - ☀ Key component to evaluate the consistency of the new boson with SM expectations
- ☀ It could be sensitive to Beyond SM physics



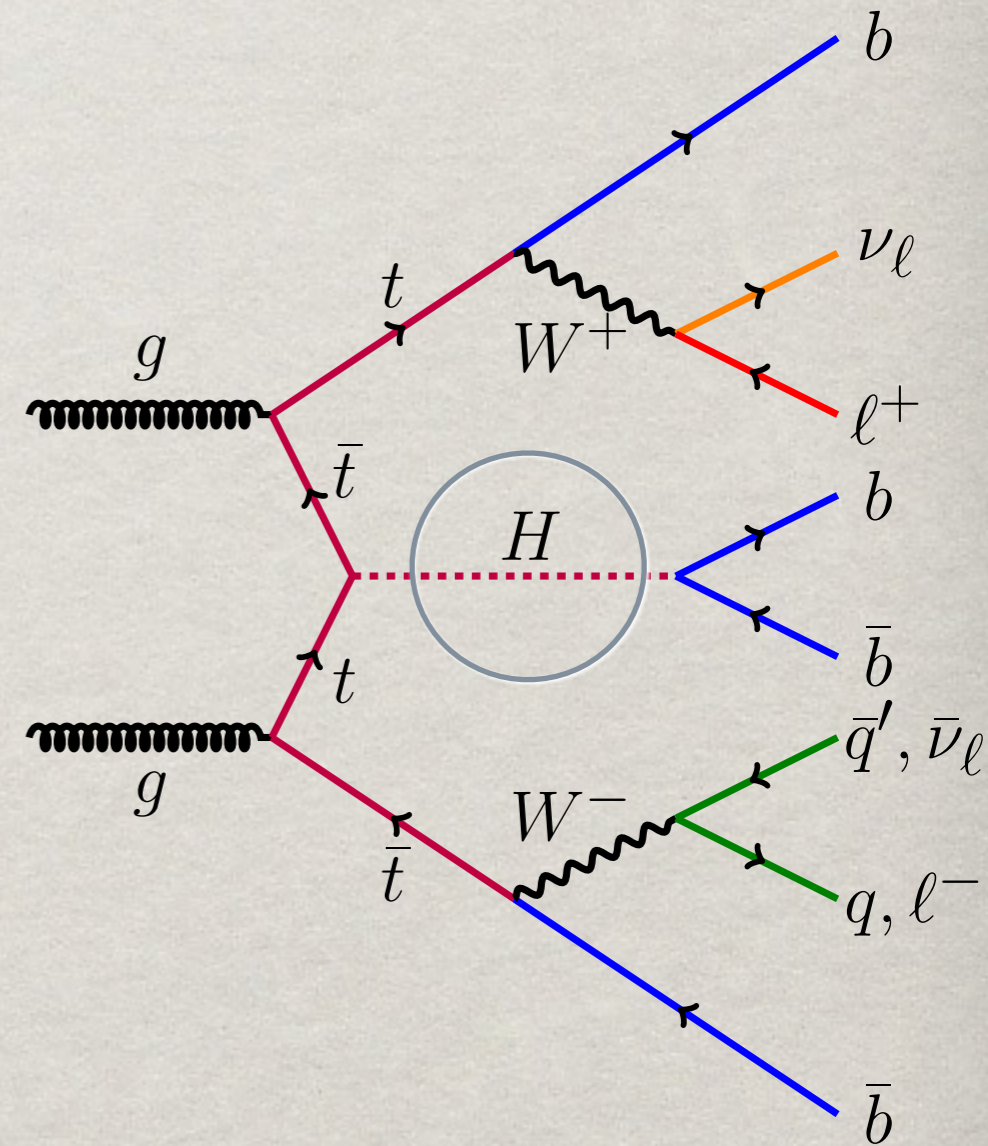
TTH @ CMS

	Decay Mode	No. of Institutes	Paper	Data(fb ⁻¹)
7TeV	bb	3	x	5
8TeV	bb/ ττ/WW/ZZ/γγ	3(bb) ≥4(non-bb)	1 bb* 1 combination	19.5
13TeV(2016)	bb/ ττ/WW/ZZ/γγ	11(bb) ≥7(non-bb)	1 bb 1 ττ/WW/ZZ 1 comb.	35.9

- ✻ 13TeV paper in preparation, huge ttH groups
- ✻ This talk will focus on the ttH (H→bb) analysis

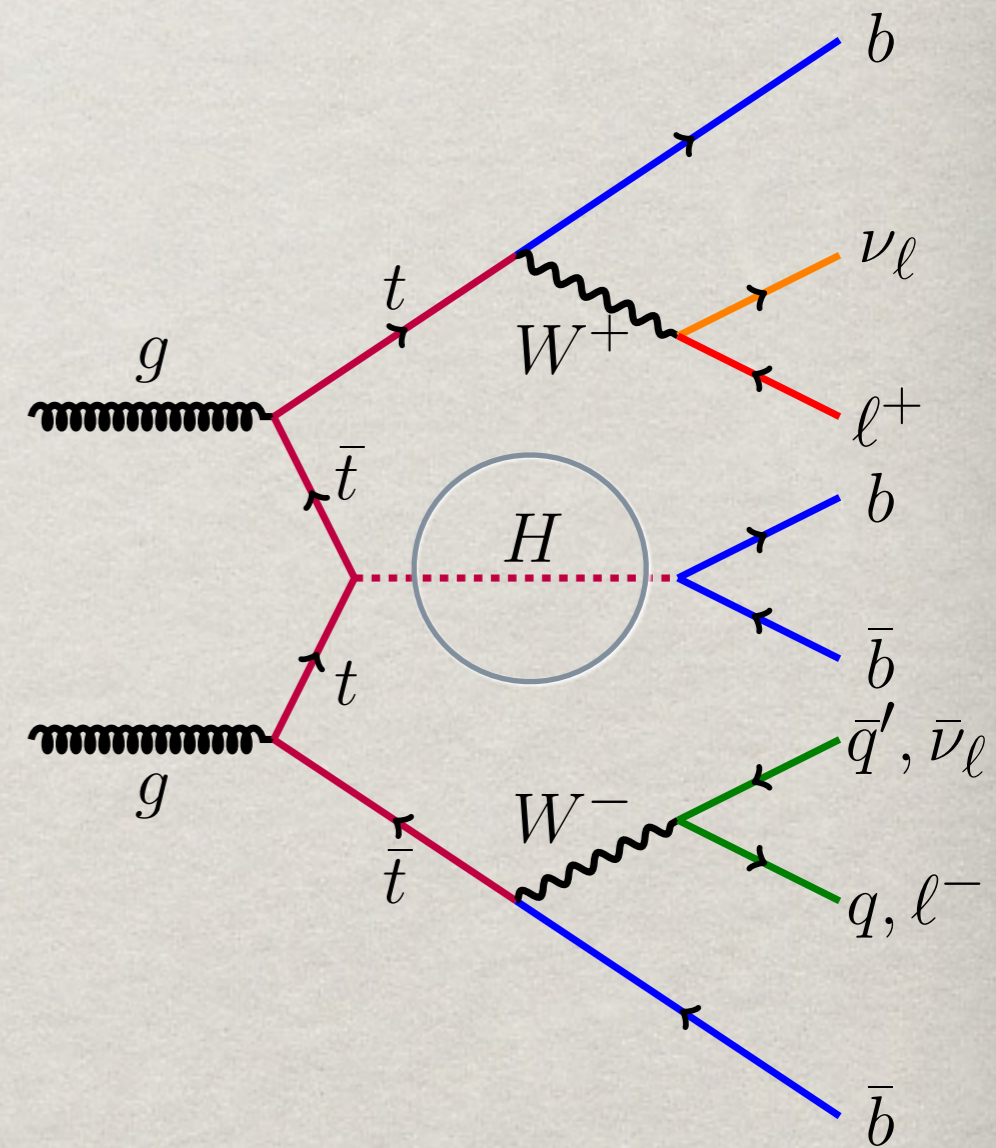
TTH(BB)

- ☀ Advantage:
 - ☀ Highest branching fraction
- ☀ Challenges:
 - ☀ Tiny production cross section
 - ☀ Higgs invariant mass hard to reconstruct
 - ☀ Difficult irreducible background $ttbb$



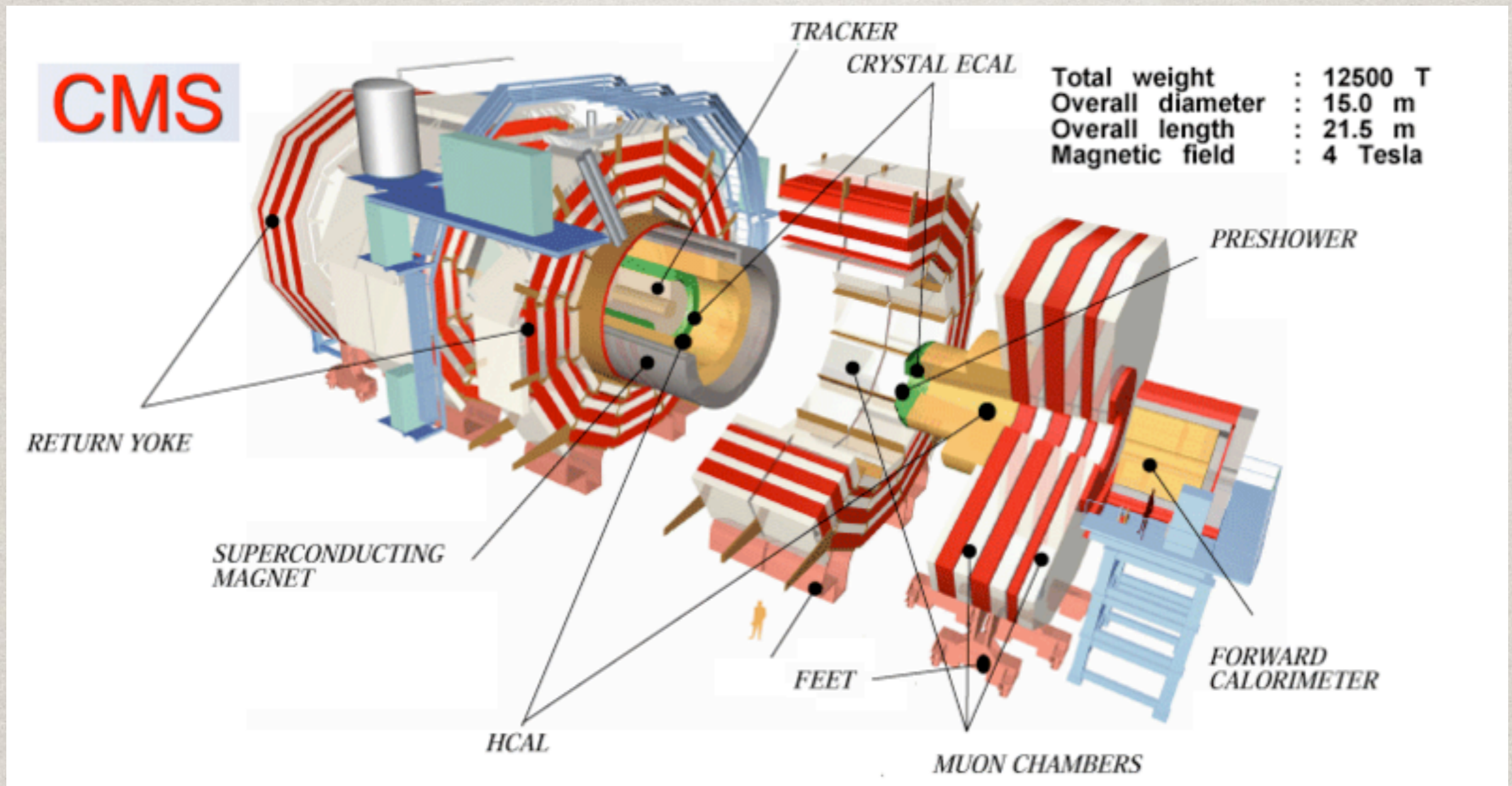
OVERVIEW

- ✿ Split channels by **top pair decay**
 - ✿ Lepton + jets (e, μ): LJ channel
 - ✿ Dilepton: DIL channel
- ✿ For each channel, separate events into categories
- ✿ Use MVA* to separate S/B and fit simultaneously all categories to data to extract signal



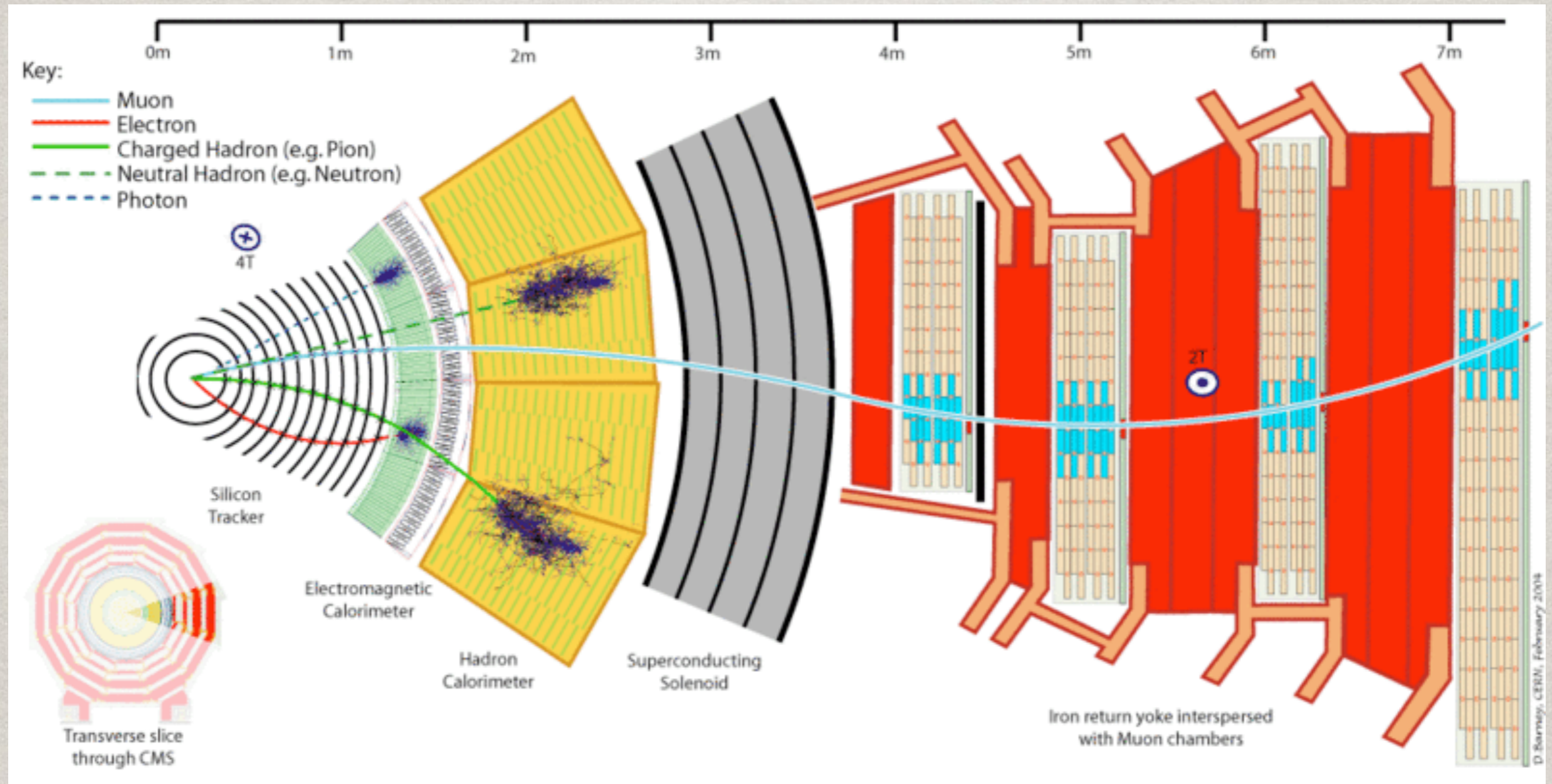
*MVA: Multi-Variate-Analysis method using various machine learning techniques

COMPACT MUON SOLENOID



Tracker, ECAL, HCAL, Magnet, Muon Chamber

DETECTOR SIGNATURES



Particle Flow: combine the info in different sub-detectors to identify particle types

SAMPLES AND SELECTION

DATA SAMPLES

- Full 2016 Data, $\sqrt{s} = 13\text{TeV}$
- The total integrated luminosity is: $L = 35.9/\text{fb}$

lepton+jets triggers

Dataset	Trigger Name
SingleMu	HLT_IsoMu22_v*
SingleMu	HLT_IsoTkMu22_v*
SingleEle	HLT_Ele27_eta2p1_WPTight_Gsf_v*

dilepton triggers

Channel	Trigger Name
$\mu^+\mu^-$	HLT_Mu17_TrkIsoVVL_TkMu8_TrkIsoVVL_v*
$\mu^+\mu^-$	HLT_Mu17_TrkIsoVVL_Mu8_TrkIsoVVL_v*
e^+e^-	HLT_Ele23_Ele12_CaloIdL_TrackIdL_IsoVL_DZ_v*
$\mu^\pm e^\mp$	HLT_Mu23_TrkIsoVVL_Ele12_CaloIdL_TrackIdL_IsoVL_v*
$\mu^\pm e^\mp$	HLT_Mu8_TrkIsoVVL_Ele23_CaloIdL_TrackIdL_IsoVL_v*

SIGNAL AND BACKGROUND

- ✱ Signal ttH samples
 - ✱ All Higgs and top decays allowed
 - ✱ $ttH(bb)$ sample and $ttH(\text{non-}bb)$ sample
- ✱ $t\bar{t} + \text{Jets}$ is the main background
 - ✱ Dedicated samples by $t\bar{t}$ decay mode
 - ✱ Separated by extra jet content: $tt+l\bar{l}/bb/b/B/cc$
- ✱ Other relevant bkg MC
 - ✱ $tt+Z/W$, $W\text{Jets}$, $Z\text{Jets}$, WW , WZ , ZZ , single top

*XS from CERN Yellow Report Page

SELECTION: LEPTON

Muons	Single Muon Channel ID	Leading ID Dilpeton	Sub-Leading Dilepton ID Veto ID for Single Muon
p_T [GeV] >	25	25	15
$ \eta $ <	2.1	2.4	2.4
ID	tight	tight	tight
$ISO_{\delta\beta}/p_T$ <	0.15	0.25	0.25

Electrons	Single Electron Channel ID	Leading ID Dilpeton	Sub-Leading Dilepton ID Veto ID for Single Electron
p_T [GeV] >	30	25	15
$ \eta $ <	2.1	2.4	2.4
ID	80% eff. non-trig. MVA ID	80% eff. non-trig. MVA ID	80% eff. non-trig. MVA ID
$ISO_{\rho A}/p_T$ <	0.15	0.15	0.15

$\mu^+ \mu^-$ and $e^+ e^-$ Channel:
$m_{\ell\ell} > 20$ GeV
$m_{\ell\ell} < 76$ GeV or $m_{\ell\ell} > 106$ GeV
MET > 40 GeV

- ✿ Lepton $P_t >$ trigger thresholds
- ✿ Tight ID and isolation to suppress multi-jet events
- ✿ Veto Z+jets events for DL

SELECTION: JETS

Jets	Single Lepton Channel Leading 2 Jets Dilepton	Dilepton Channel Subleading Jets Dilepton
Type	PFJets, CHS	PFJets, CHS
Algorithm	anti- k_T 0.4	anti- k_T
p_T [GeV] >	30	20
$ \eta <$	2.4	2.4
Lepton cleaning	Require $\Delta R(\ell, j) > 0.4$	Require $\Delta R(\ell, j) > 0.4$

- ✱ Jet multiplicity:
 - ✱ ≥ 4 jets in LJ channel
 - ✱ ≥ 2 jets in DL channel
- ✱ b-tags: jets originating from b quarks
 - ✱ use CSV(Combined Secondary Vertex) algorithm
 - ✱ identify as b-jets if passing Medium working point
 - ✱ 1(2) b-tags for DL(LJ) inclusive selection

EVENT CATEGORIZATION

- ✱ Different channels based on top pair decay
 - ✱ LJ channel and DIL channel
- ✱ For each channel, categorize events based on **number of jets** and **number of b-tags** (alternative scheme later*)

Lepton + Jets(LJ)			
	4jets	5jets	≥6jets
2tags	x	x	x
3tags	x	x	✓
≥4tags	✓	✓	✓

Dilepton(DIL)		
	3jets	≥4jets
2tags	x	x
3tags	x	✓
≥4tags	✓	✓

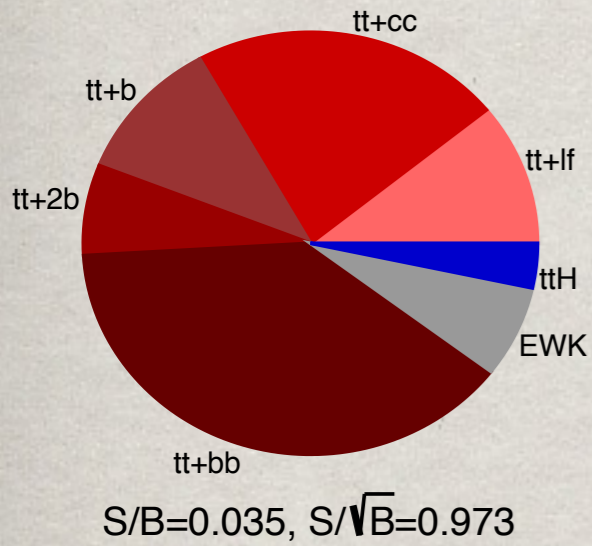
CATEGORIZATION

CMS

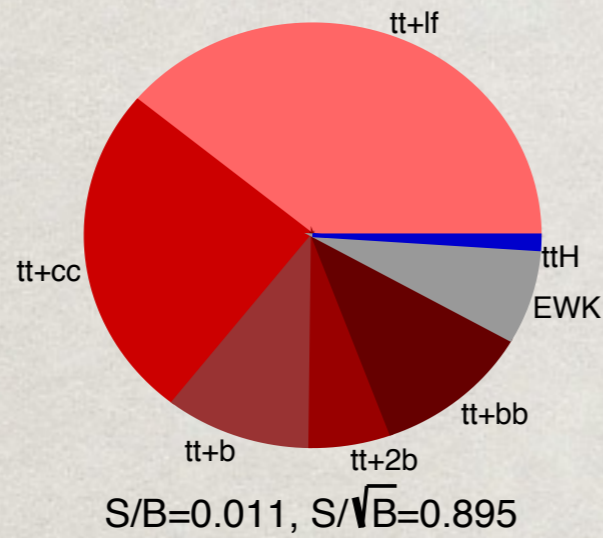
Simulation

Lepton+Jets Channel

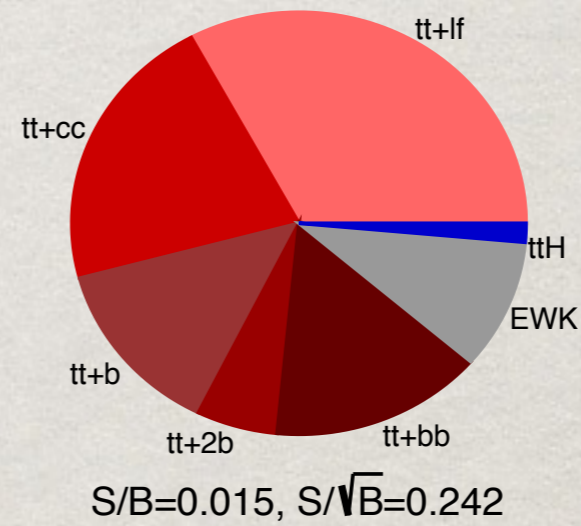
≥ 6 jet, ≥ 4 b-tags



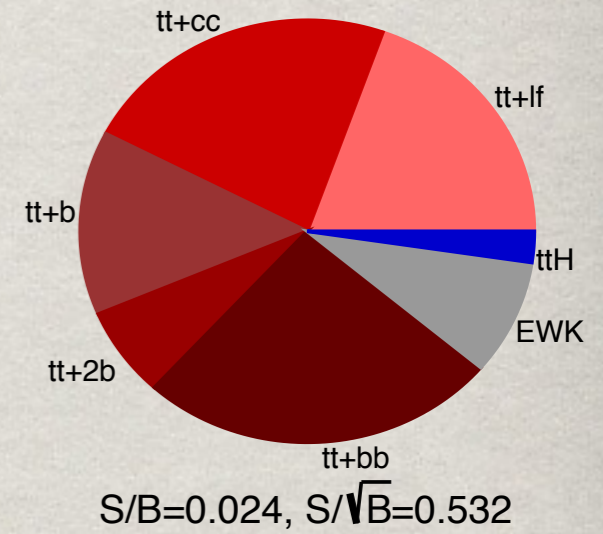
≥ 6 jets, 3 b-tags



4 jets, 4 b-tags



5 jets, ≥ 4 b-tags

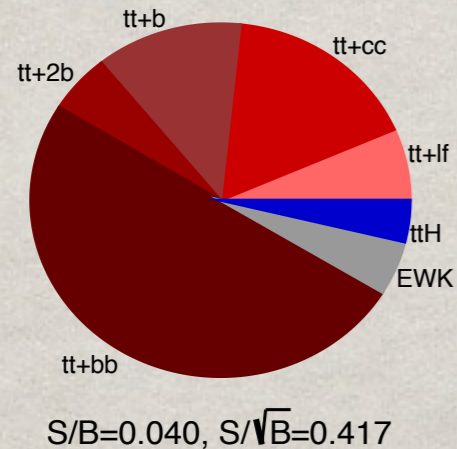


CMS

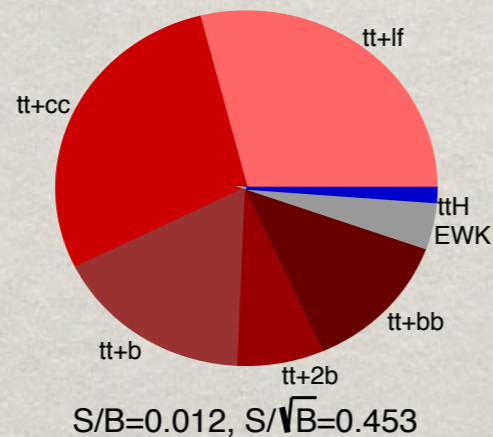
Simulation

Dilepton Channel

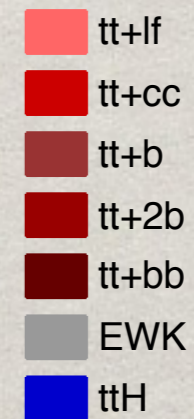
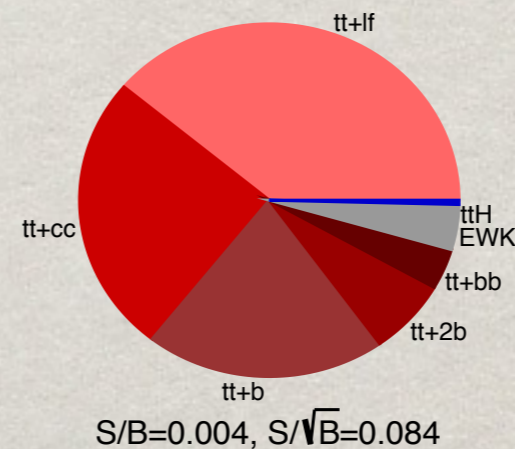
≥ 4 jets, ≥ 4 b-tags



≥ 4 jets, 3 b-tags



3 jets, 3 b-tags



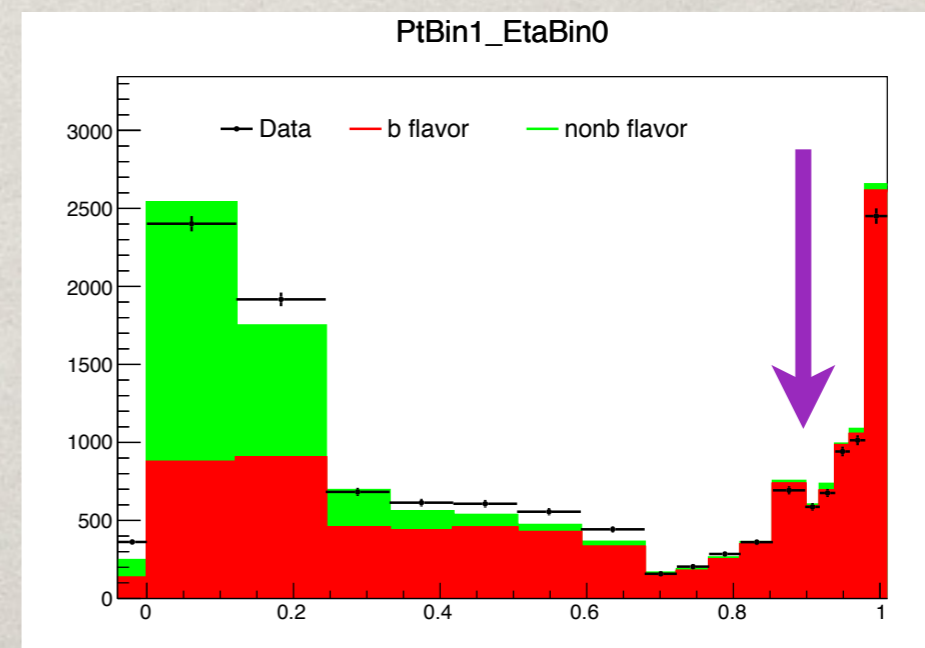
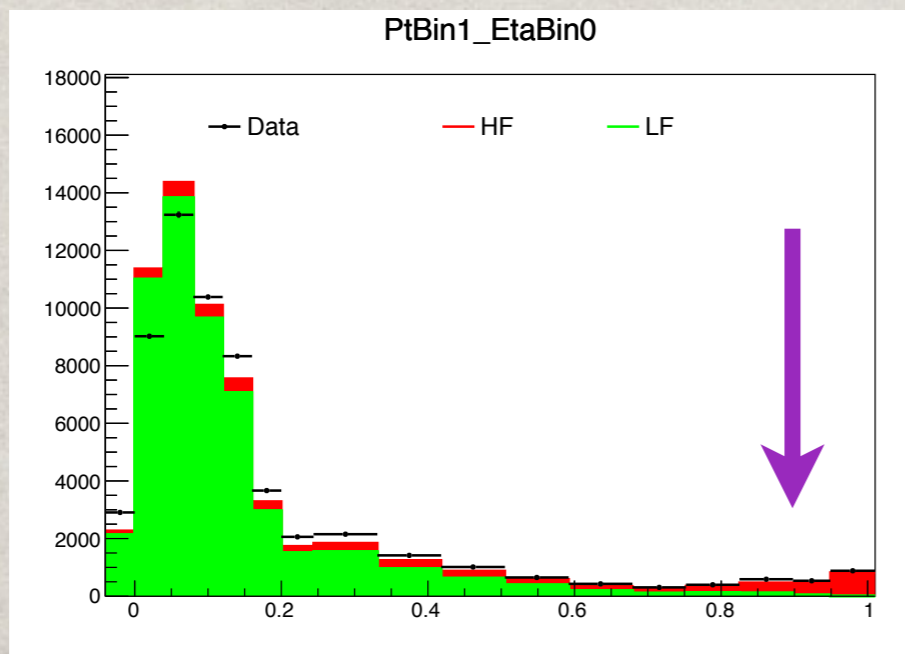
BACKGROUND MODELING

CORRECTIONS TO MC

- ✱ MC samples were created based on certain theory assumptions
 - ✱ Need corrections to match Data
- ✱ Pileup vertex reweighting
 - ✱ Reproduce number of PU interactions
- ✱ Jet energy calibration: jet energy resolution/scale
- ✱ Lepton data/MC scale factor
 - ✱ Based on lepton P_T and η
- ✱ **B-tag CSV reweighting**
 - ✱ Correct MC b-tag efficiency as a function of the CSV discriminator

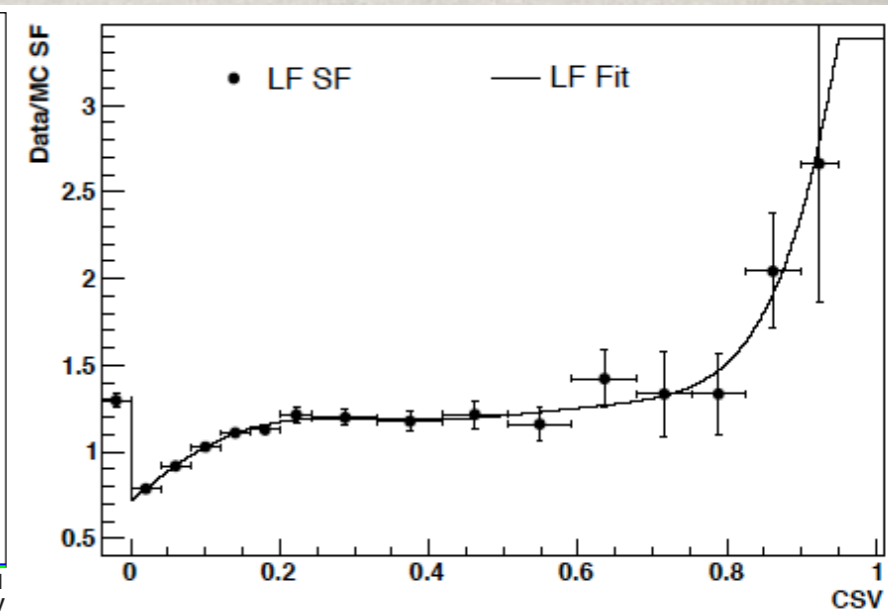
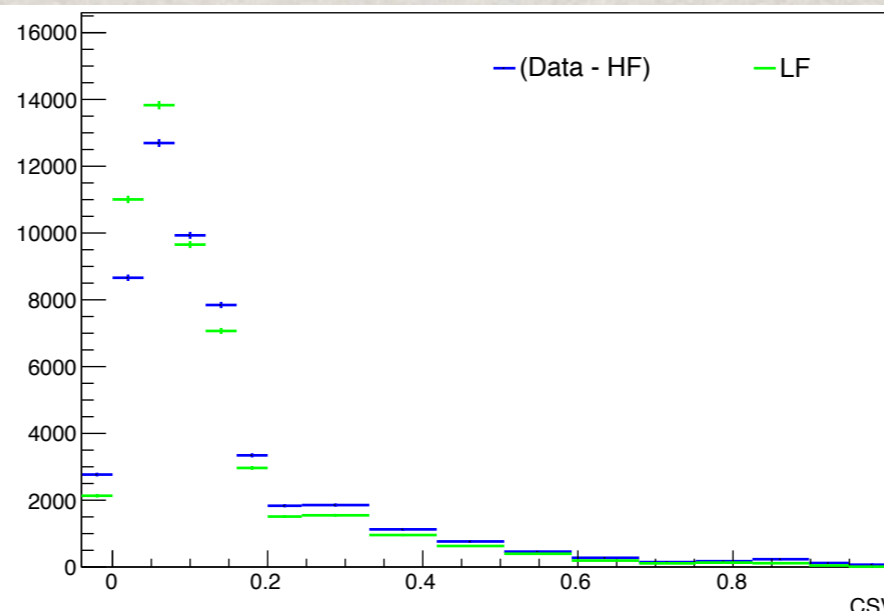
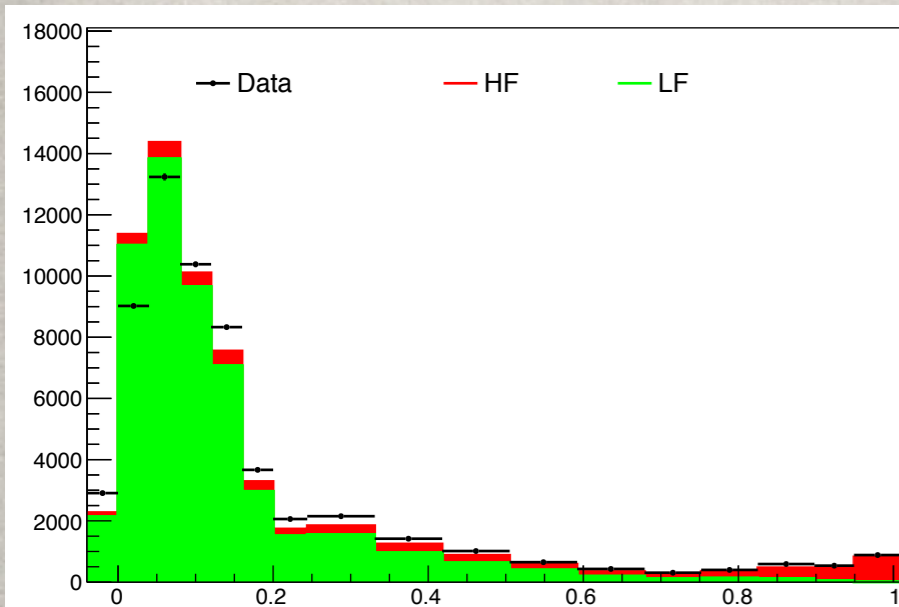
B-TAG CSV REWEIGHTING

- ✱ We need to model CSV b-tagging correctly:
 - ✱ Use CSV medium working point to identify b-tagged jets
 - ✱ Use CSV distributions to separate signal from background
- ✱ We require corrections to CSV shape bin-by-bin
- ✱ We derive a reweighting of CSV discriminant shape for both light and heavy flavor jets



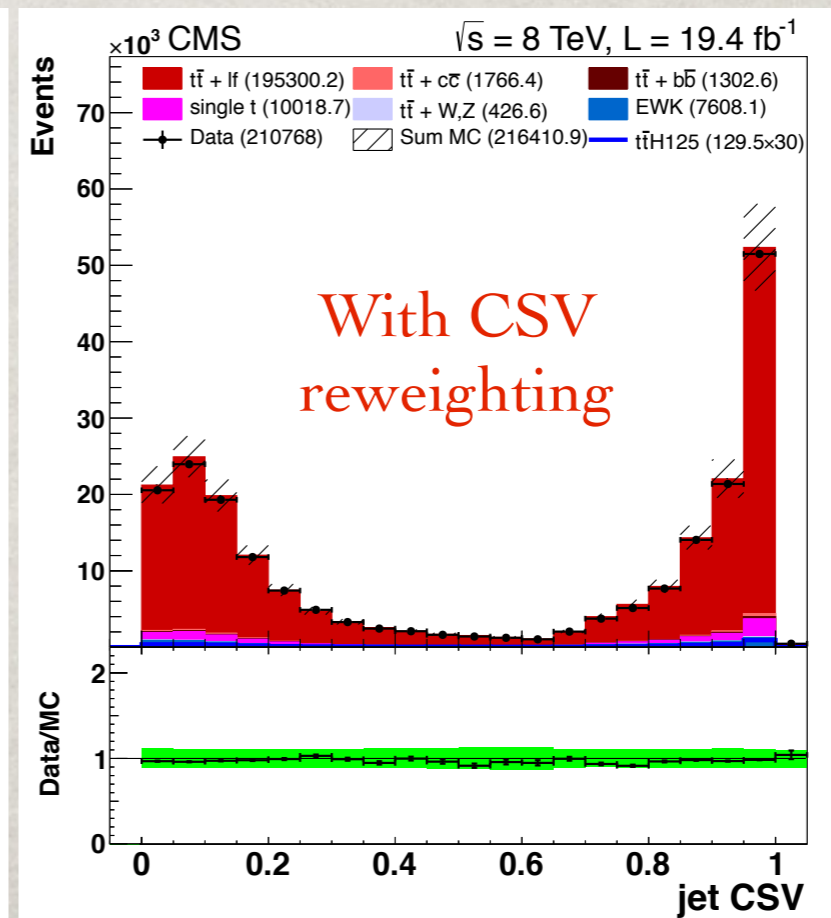
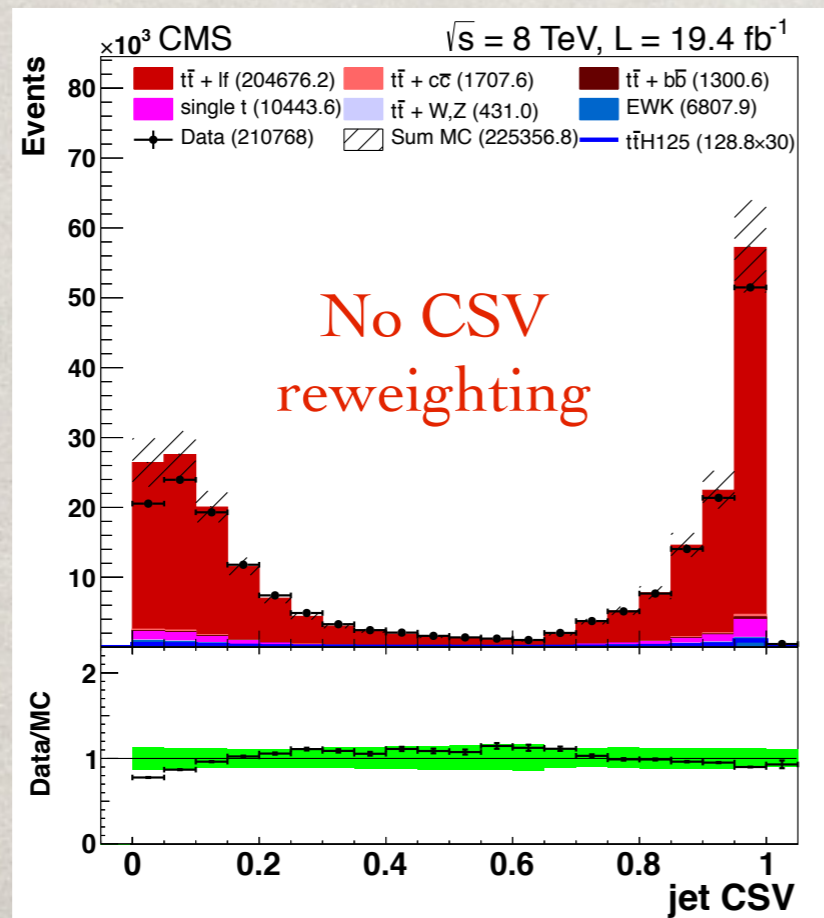
TAG AND PROBE

- ✱ Use Tag-and-Probe method to calculate CSV scale factors bin-by-bin from independent control samples not used in analysis
 - ✱ Heavy flavor SF: DIL ttbar enriched sample
 - ✱ Light flavor SF: DIL Z+jets enriched sample
- ✱ Require one jet to be (anti) tagged, account for LF(HF) contamination, correct probe jet CSV shape in MC to match data
- ✱ Scale Factor is defined as: $SF(CSV, p_t, \eta) = \frac{Data - MC_A}{MC_B}$ (A/B=HF/LF)



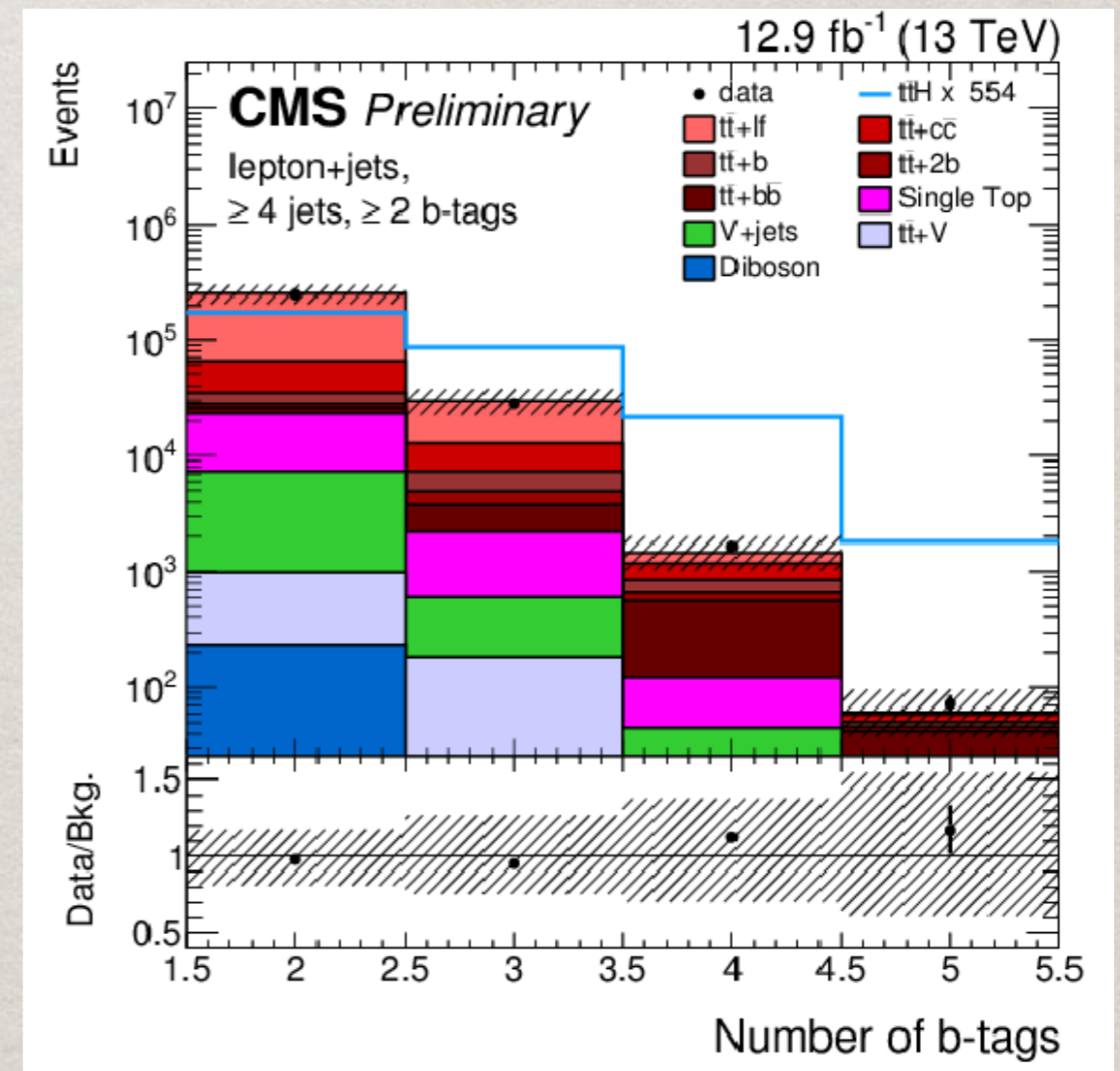
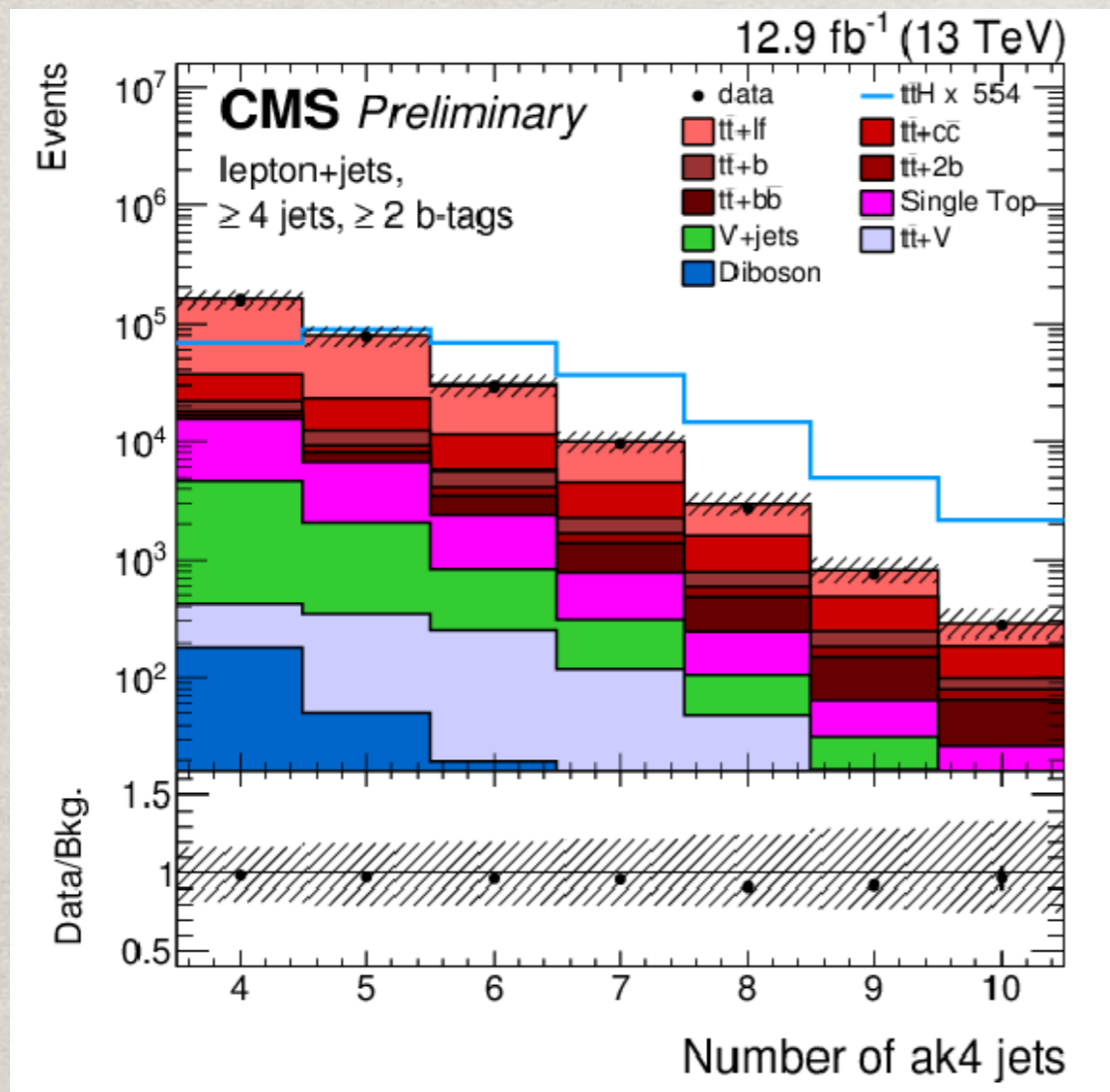
CSV SF

- Apply the CSV Scale Factors(SFs) based on jet flavor:
 - for b jets, assign heavy flavor SF
 - for light jets, assign light flavor SF
 - for c jets, no correction
- Final scale factor for each event is: $SF_{total} = SF_{jet1} * SF_{jet2} * \dots$
- Significant improvement for CSV shapes



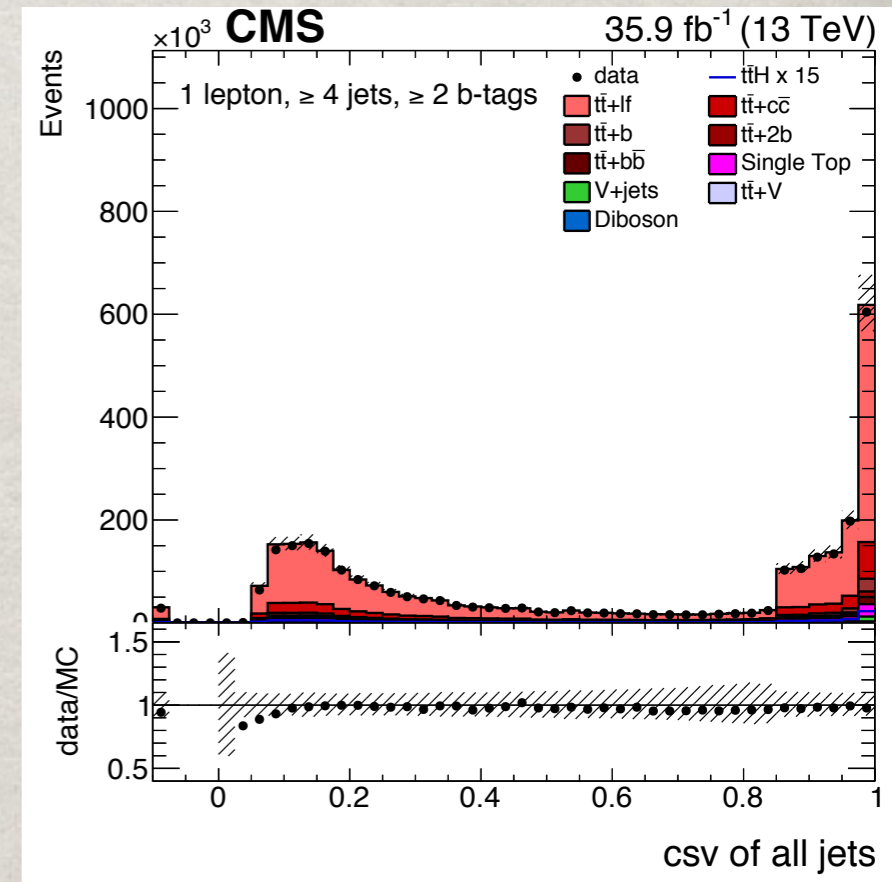
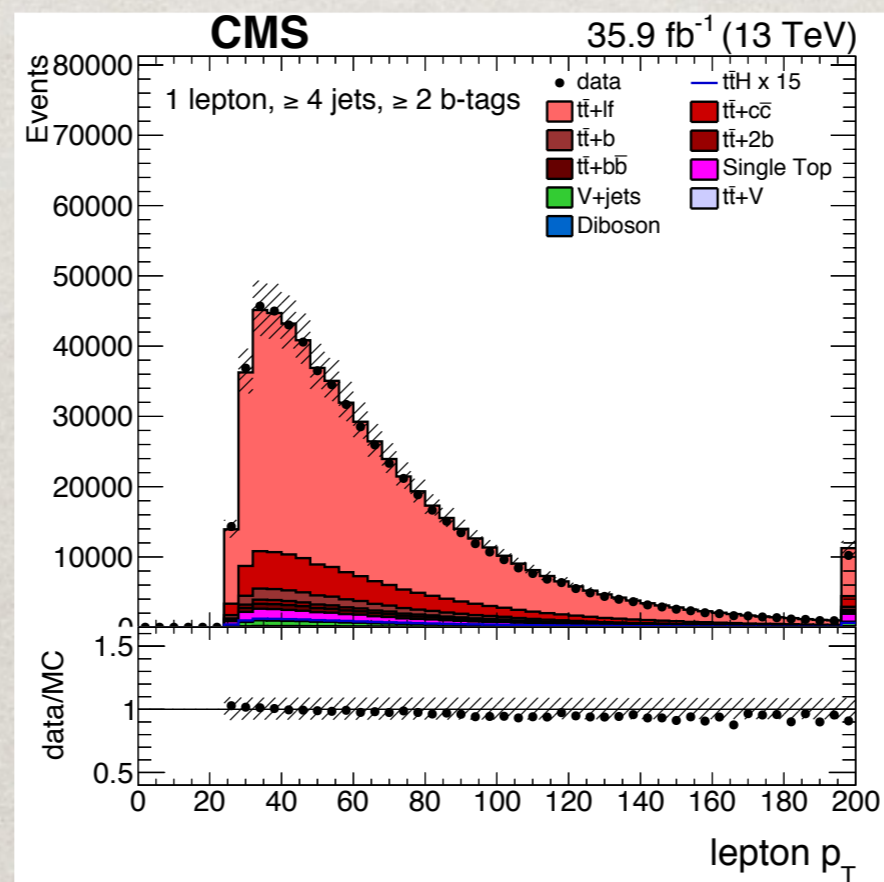
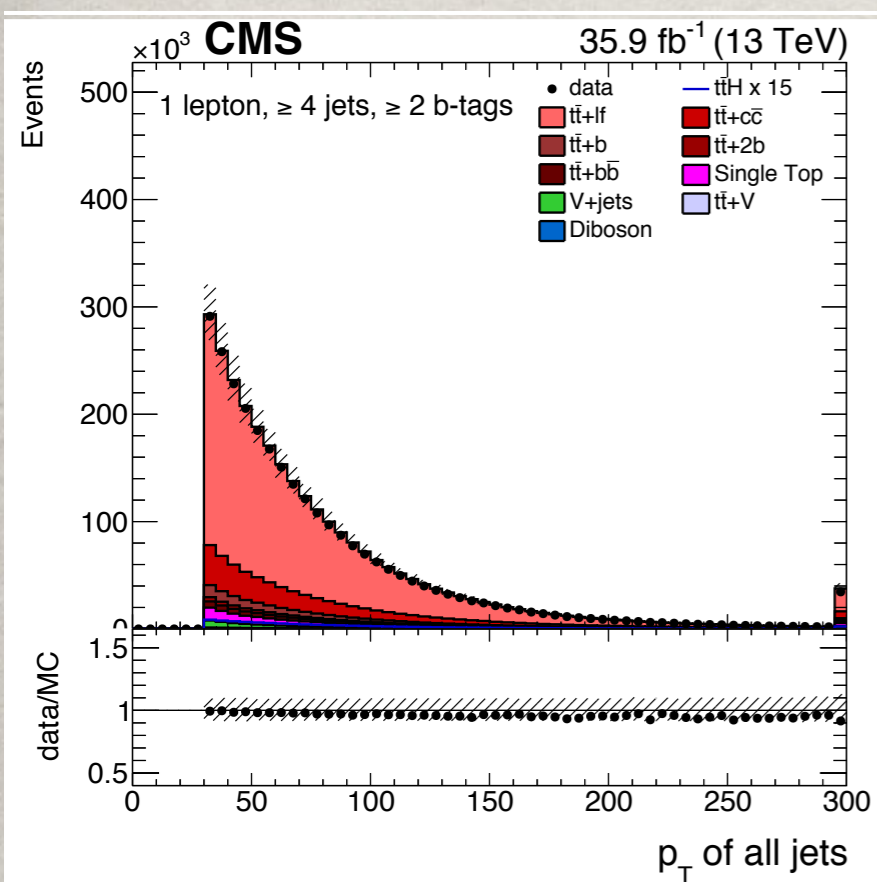
DATA/MC AGREEMENT: LJ

- ☼ All corrections to MC applied
- ☼ Good agreement between Data and MC



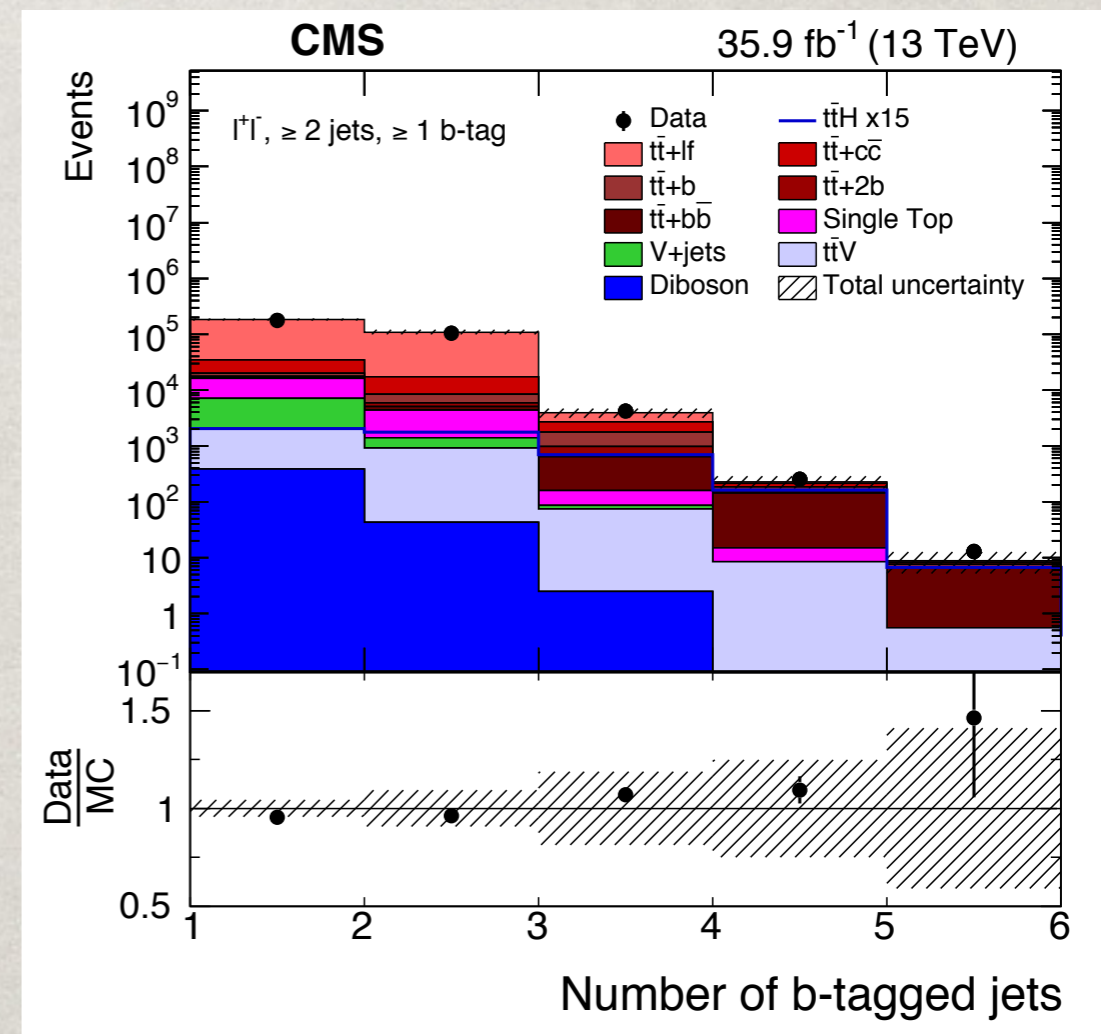
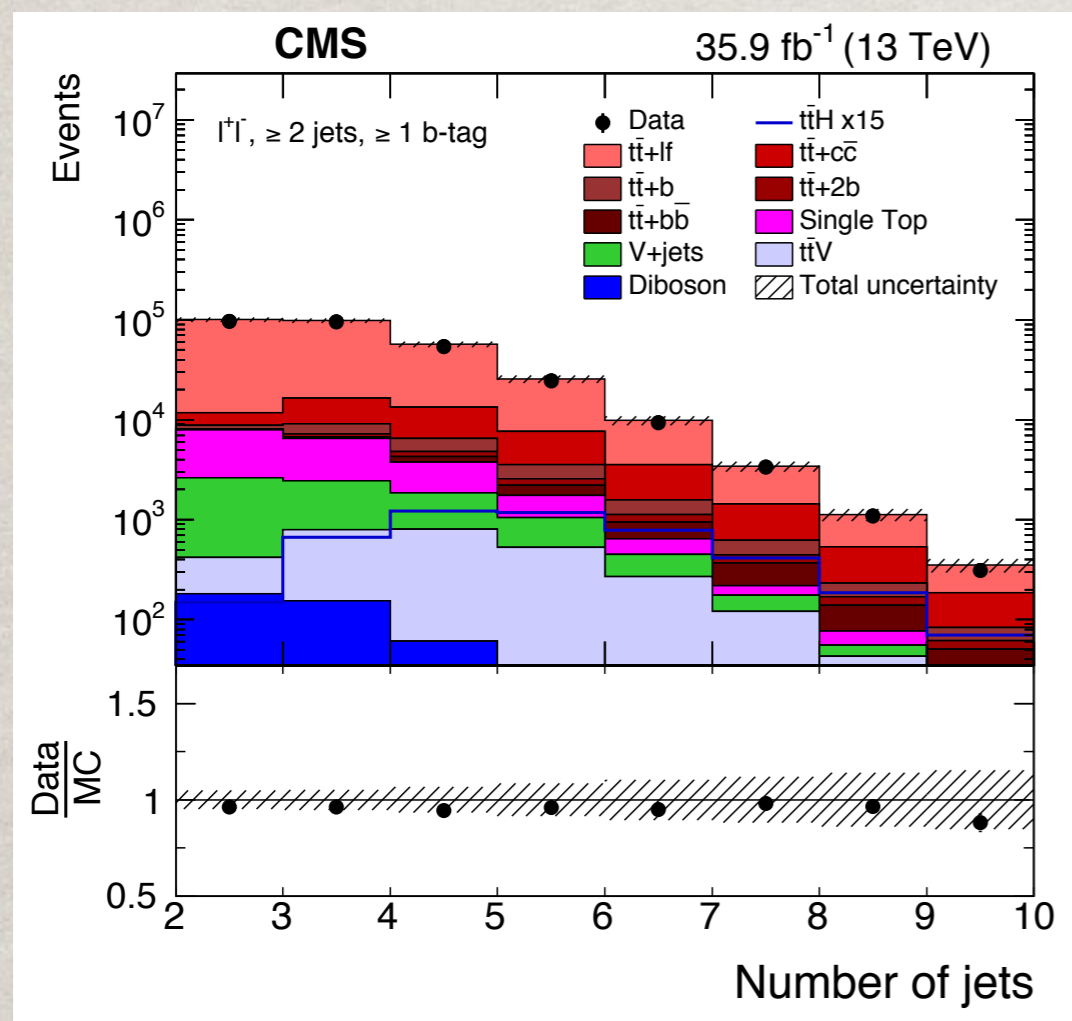
DATA/MC AGREEMENT: LJ

- ☼ All corrections to MC applied
- ☼ Good agreement between Data and MC



DATA/MC AGREEMENT: DIL

- ☼ All corrections to MC applied
- ☼ Good agreement between Data and MC

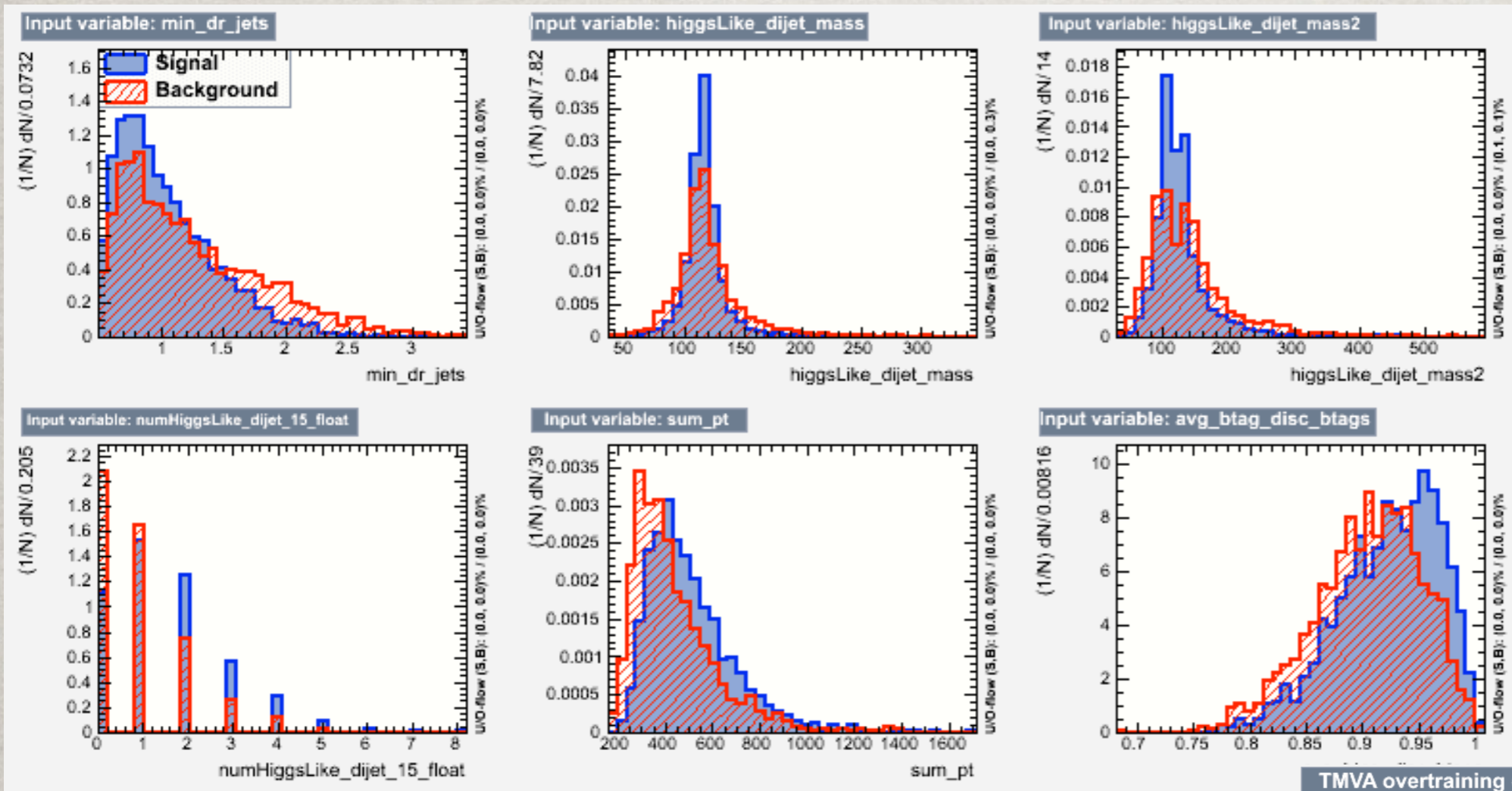


SIGNAL EXTRACTION

MULTIVARIATE ANALYSIS

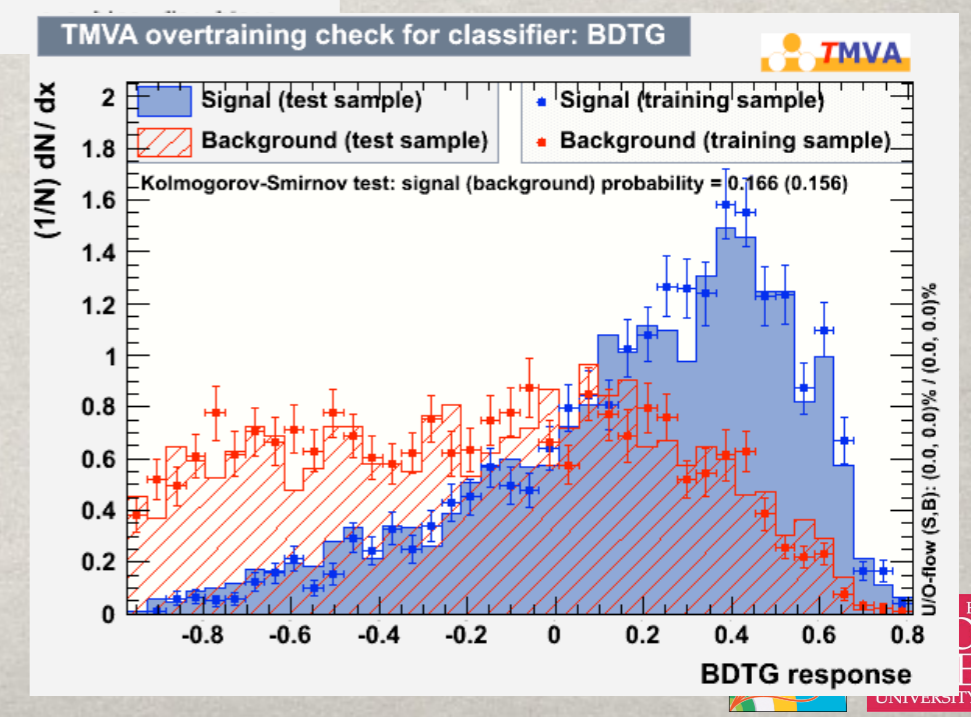
- ✱ Can't use Higgs invariant mass as discriminant like other analyses ($H \rightarrow \gamma\gamma$ or $H \rightarrow ZZ \rightarrow 4l$).
- ✱ Background very similar to Signal
- ✱ *Multi-Variate Analysis:*
 - ✱ Combine several variables' discriminating power
 - ✱ Use Boosted Decision Tree (BDT) or Deep Neural Network (DNN)
- ✱ Train separate BDT/DNN for each category
- ✱ Fit BDT/DNN discriminators from all categories simultaneously to extract signal

BDT EXAMPLE



input variables

output discriminator



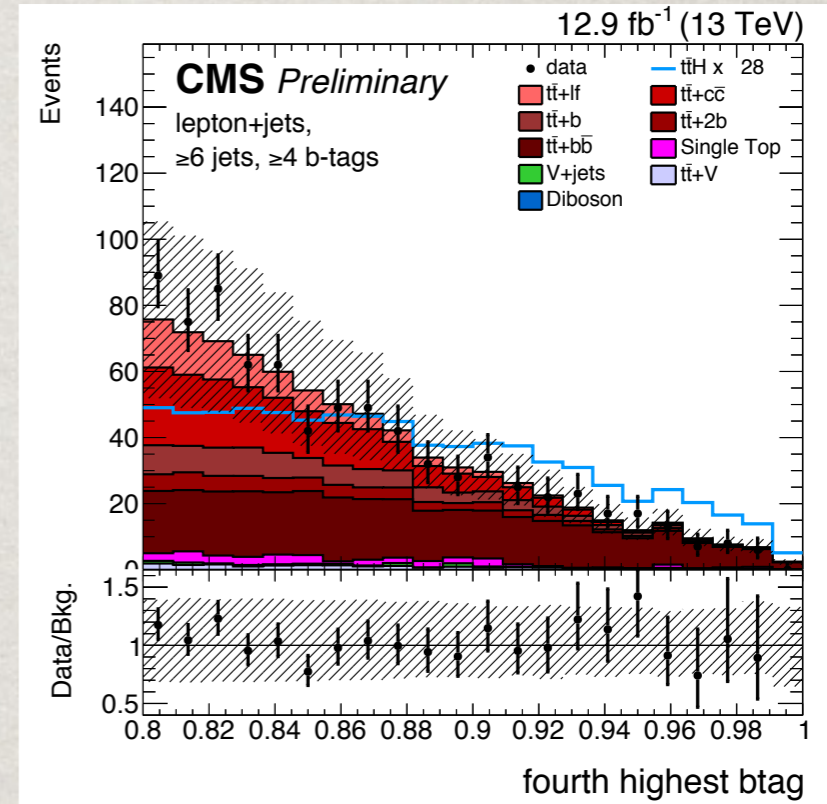
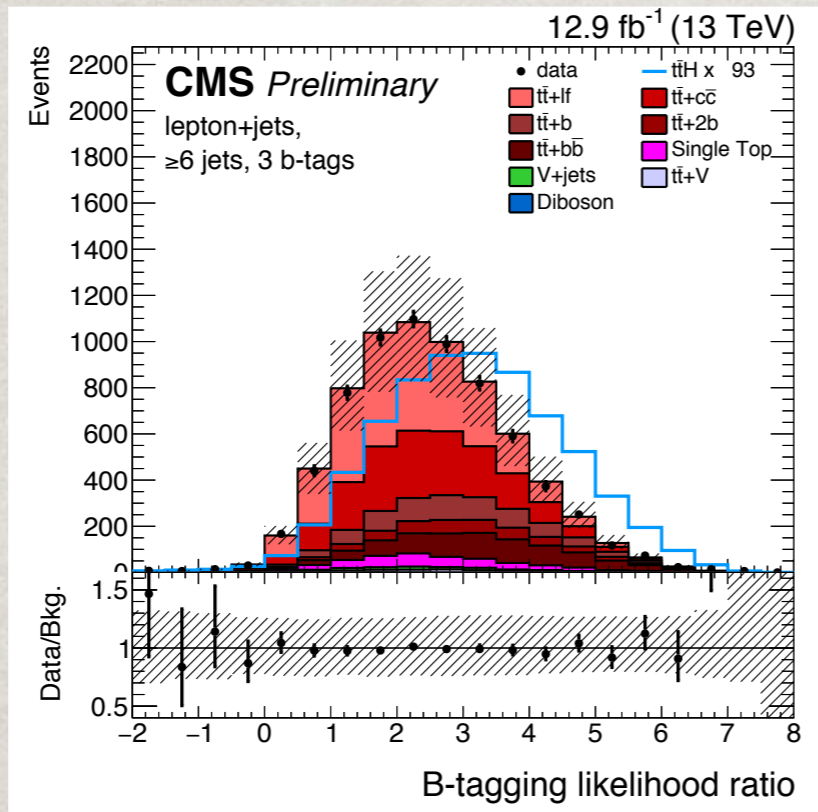
BDT

- ✱ BDT training is a **machine learning** process
 - ✱ **Aim to classify/separate signal or background**
- ✱ For the BDT training in each jet/tag category
 - ✱ ttH as signal(S) and ttjets as background(B)
- ✱ Input variables: CSV tagging, invariant mass, angular correlations, event shapes, and jet Pt variables

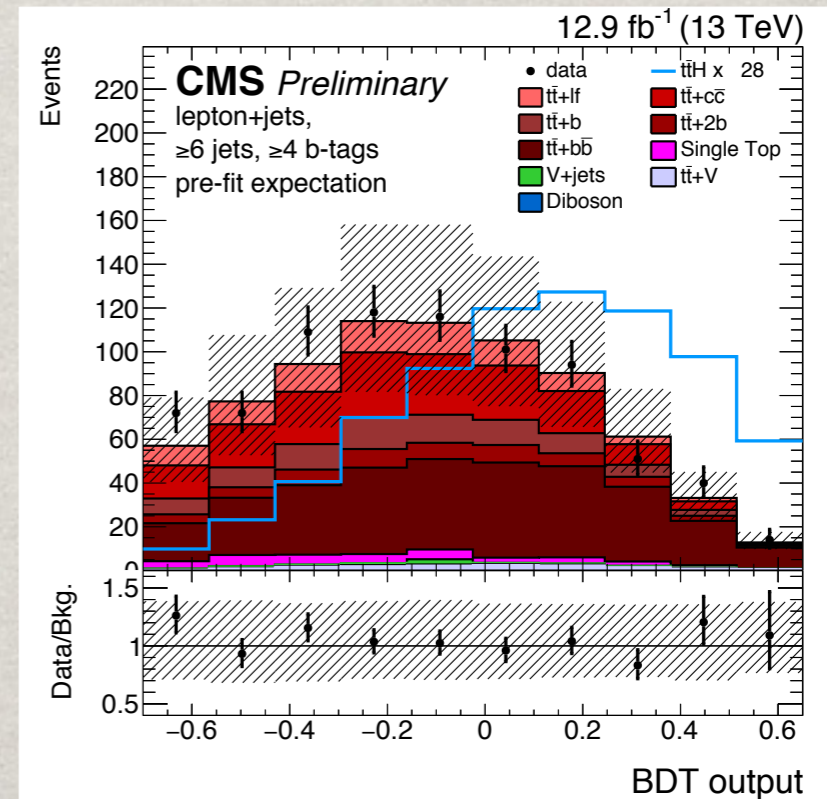
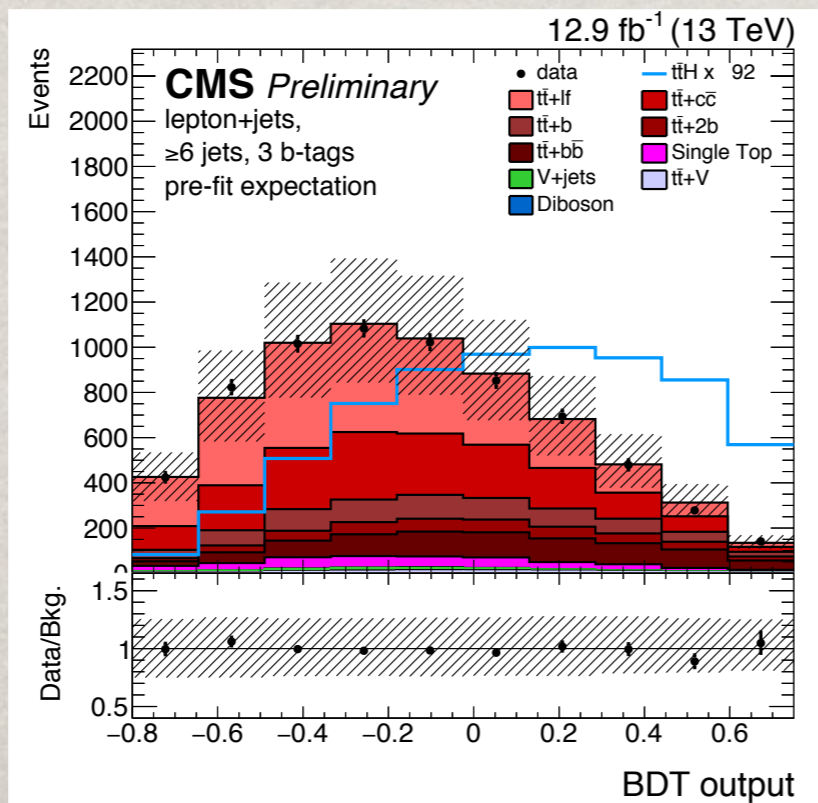


BDT INPUT/OUTPUT

Input variables

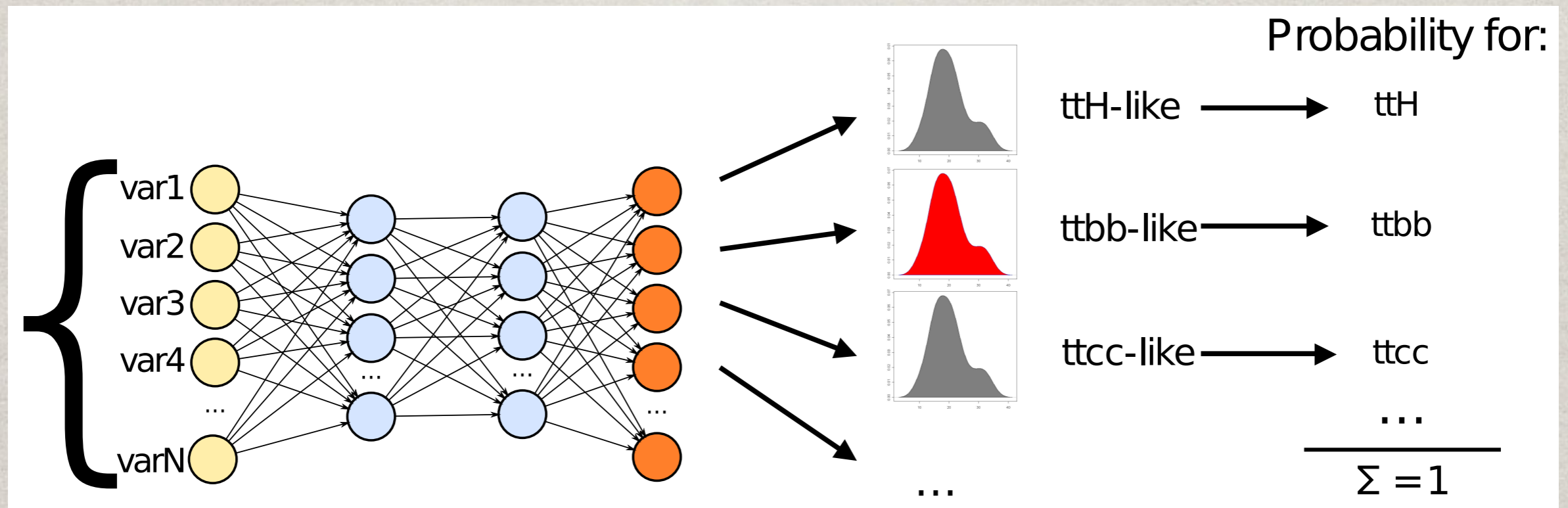


Output variable



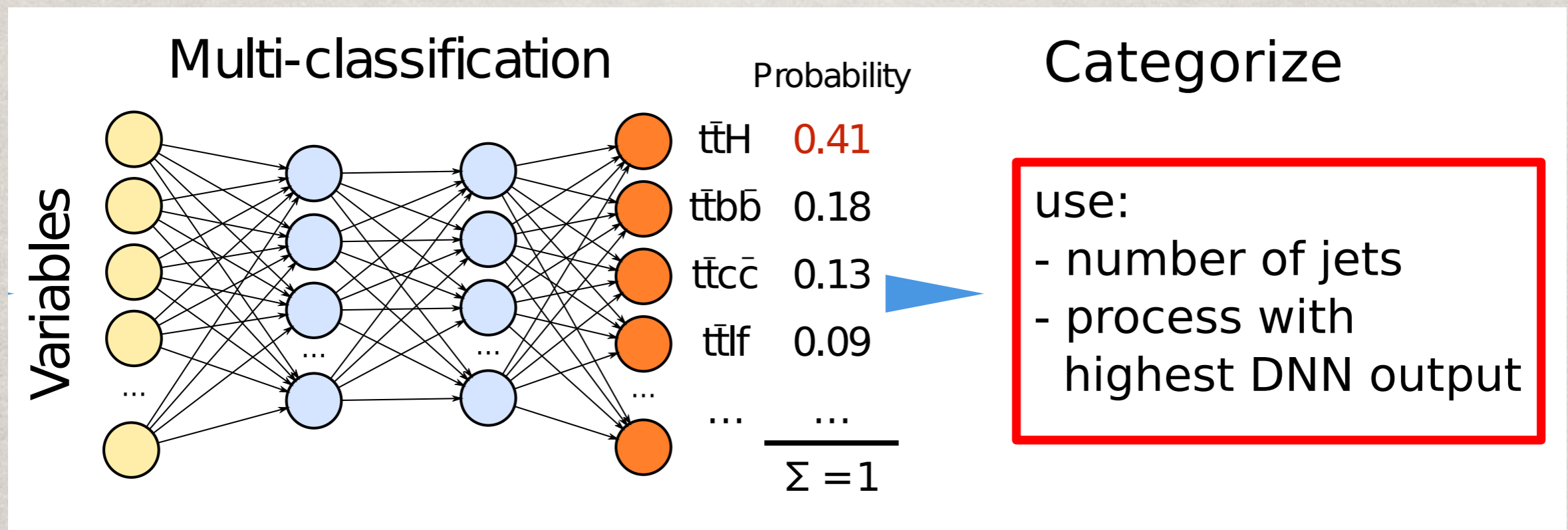
DEEP NEURAL NETWORKS

- ☼ Deep Neural Networks (DNNs) is very good at multi-classification:
 - ☼ split dominant bkg ttjets into all components
 - ☼ BDT only separates ttH from INCLUSIVE ttjets

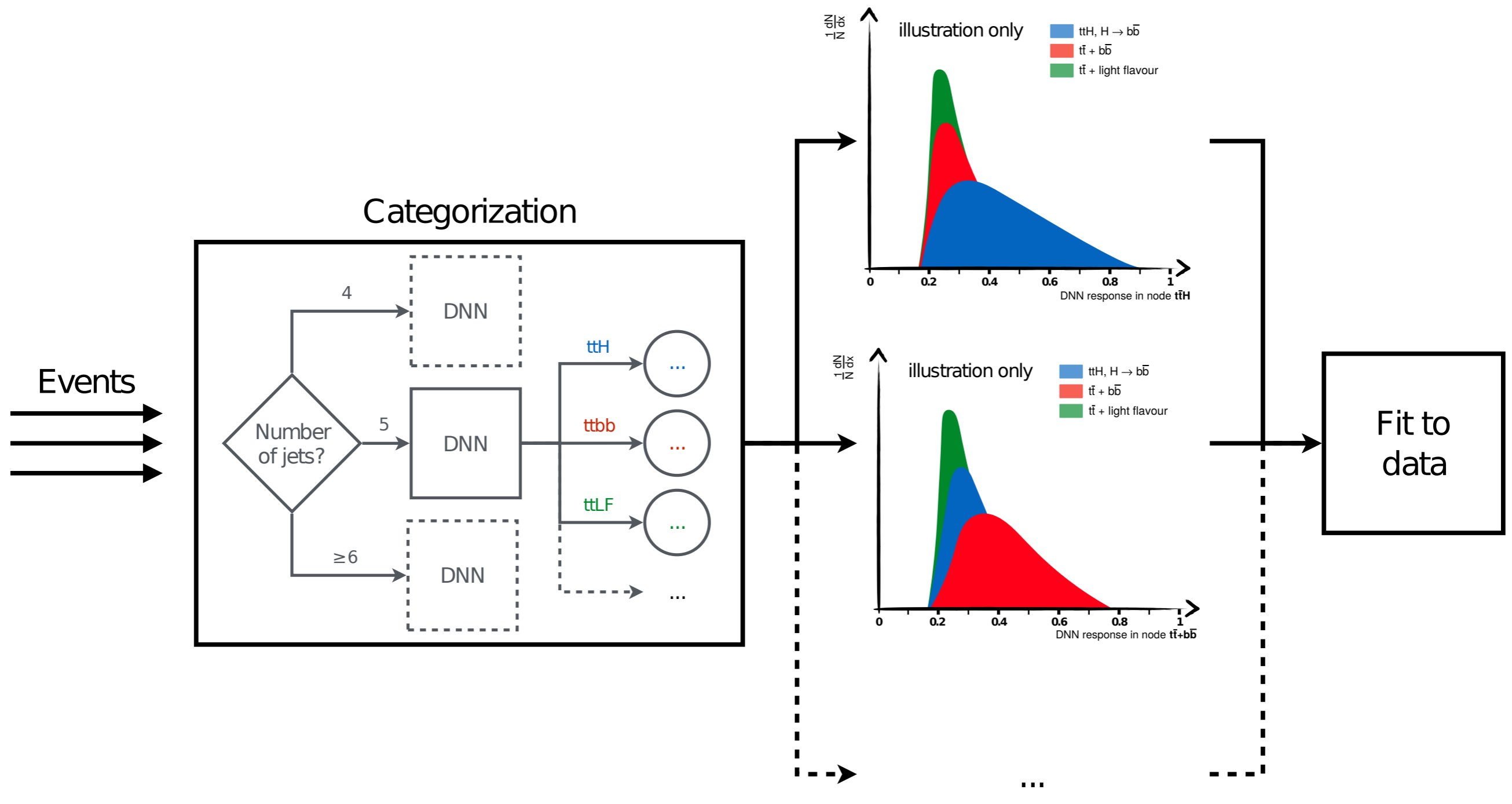


DNN

- Use DNN for alternative categorization, instead of the nJet-nTag scheme
- Use DNN output as final fit discriminant



DNN WORKFLOW



TTBAR+HF

MATRIX ELEMENT METHOD

- ✿ Irreducible background ttbb
- ✿ Use Matrix Element Method(MEM) to further distinguish ttbb from ttH

Numerical integration
Momentum conservation
Resolution function (allow ISR)

$$w(\vec{y}|\mathcal{H}) = \sum_{i=1}^{N_a} \int \frac{dx_a dx_b}{2x_a x_b s} \int \prod_{k=1}^8 \left(\frac{d^3 \vec{p}_k}{(2\pi)^3 2E_k} \right) (2\pi)^4 \delta^{(E,z)} \left(p_a + p_b - \sum_{k=1}^8 p_k \right) \mathcal{R}^{(x,y)} \left(\vec{p}_T, \sum_{k=1}^8 p_k \right) \times$$

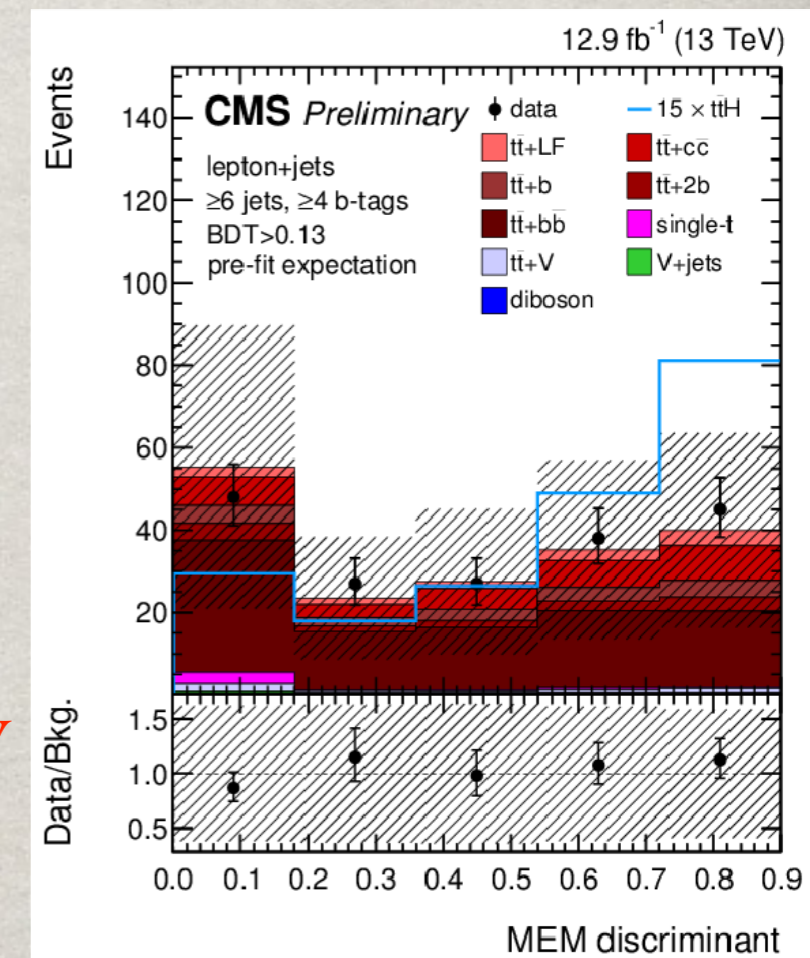
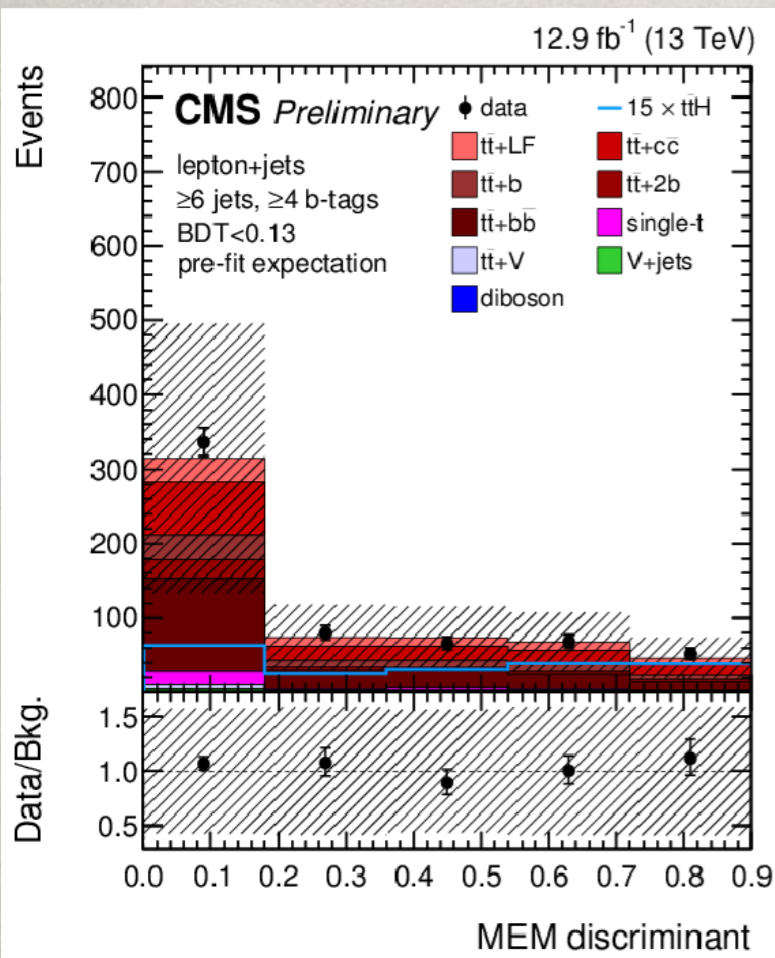
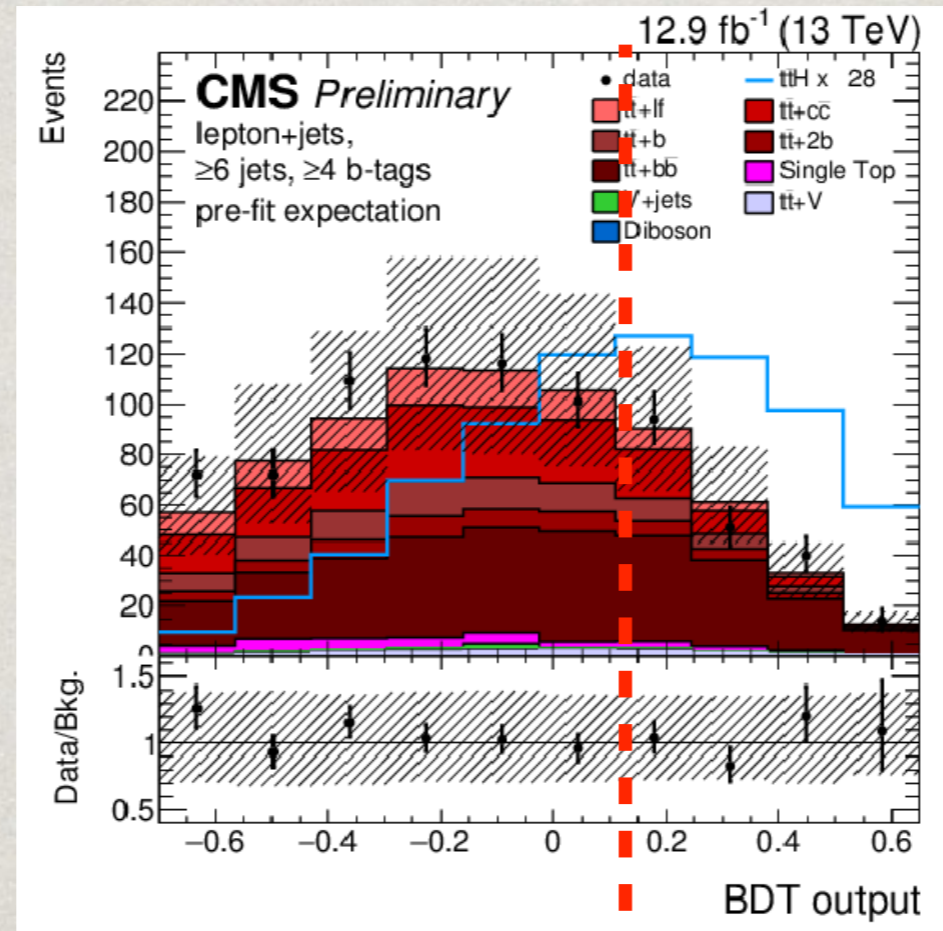
$g(x_a, \mu_F) g(x_b, \mu_F) |\mathcal{M}_{\mathcal{H}}(p_a, p_b, p_1, \dots, p_8)|^2 W(\vec{y}, \vec{p}),$

Parton density functions
LO Scattering amplitude (Open Loops)
Detector transfer function

$$P_{s/b} = \frac{w(\vec{y}|\bar{t}\bar{t}H)}{w(\vec{y}|\bar{t}\bar{t}H) + k_{s/b} w(\vec{y}|\bar{t}\bar{t}+b\bar{b})}$$

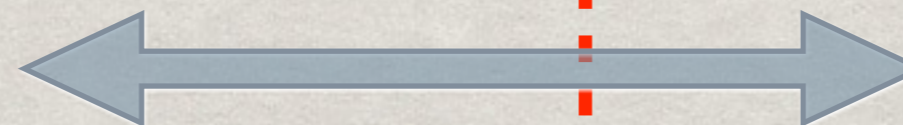
BDT/MEM 2D APPROACH

split category at the median of ttH BDT output



low purity

high purity



ADDITIONAL $tt+HF$ UNCERTAINTY

- ✱ Contribution from $tt+HF$ very similar to signal
 - ✱ uncertainty on rate/shape has a big impact on our search
- ✱ Due to lack of more **accurate higher order theory predictions**, we obtained $tt+HF$ estimate and uncertainty based on the inclusive $t\bar{t}$ sample
- ✱ On top of other uncertainty, assign an extra 50% rate uncertainty for $tt+bb/b/B/cc$ independently



RESULTS

FINAL STRATEGY

Categorization

{ nJet-nTag
nJet-DNN_process

Discriminant

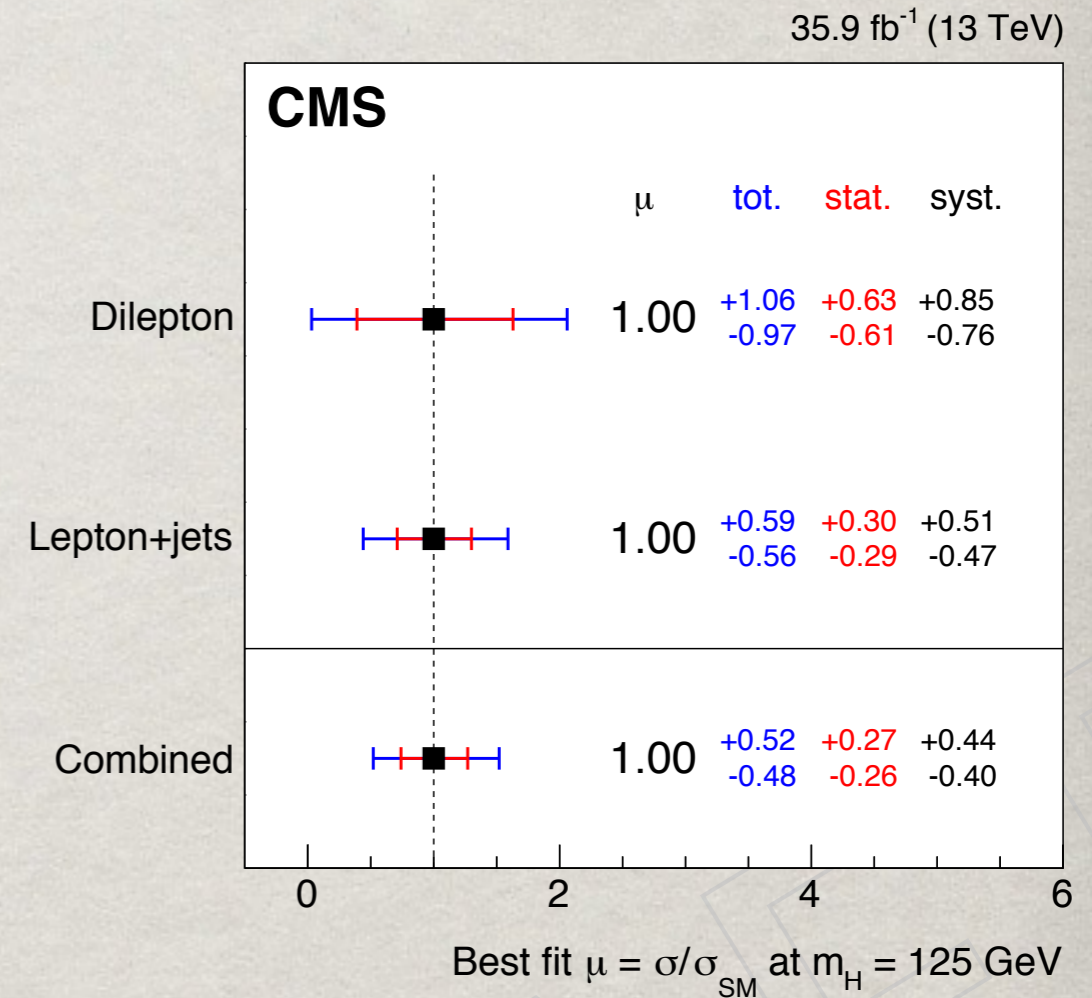
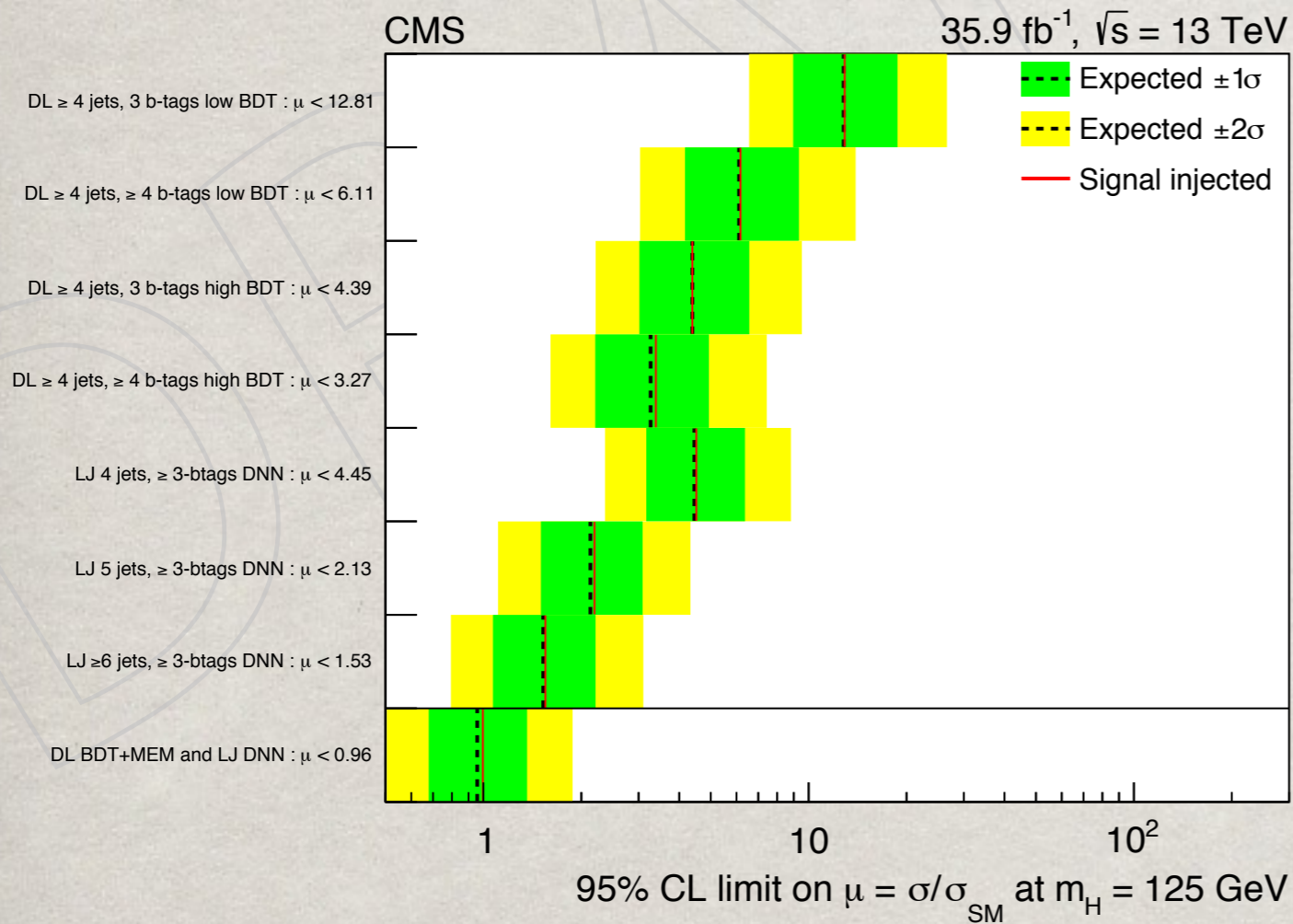
{ BDT
DNN
MEM

- ✱ Various strategies when combining the above steps
- ✱ Choose the one with the best sensitivity and robustness
- ✱ Use the other strategies as cross-checks

RESULTS SUMMARY

Channel	Analysis	Expected Limit	Best-fit μ	Significance
dilepton	MEM+BLR	$2.93^{+3.44}_{-1.46}$	$1.0^{+1.33}_{-1.50}$	
dilepton	2D BDT+MEM	$2.02^{+0.96}_{-0.61}$	$1.0^{+1.06}_{-0.97}$	1.02
dilepton	DNN	$2.59^{+1.12}_{-0.77}$	$1.0^{+1.34}_{-0.69}$	0.92
single lepton	MEM+BLR	$1.71^{+1.72}_{-0.96}$	$1.0^{+0.86}_{-0.90}$	
single lepton	2D BDT+MEM	$1.29^{+0.55}_{-0.36}$	$1.0^{+0.67}_{-0.66}$	1.49
single lepton	DNN	$1.14^{+0.48}_{-0.33}$	$1.0^{+0.61}_{-0.57}$	1.77
combined	MEM+BLR	$1.43^{+1.43}_{-0.68}$	$1.0^{+0.73}_{-0.76}$	1.38
combined	2D BDT+MEM	$1.07^{+0.45}_{-0.3}$	$1.0^{+0.57}_{-0.55}$	1.8
combined	DNN	$1.04^{+0.44}_{-0.3}$	$1.0^{+0.56}_{-0.53}$	1.99
single-lepton DNN + dilepton 2D BDT+MEM		$0.97^{+0.41}_{-0.28}$	$1.0^{+0.53}_{-0.5}$	2.04

LIMIT AND SIGNAL STRENGTH





Channel	Observed UL	Expected UL	Best-fit μ
Dilepton	X.X	$2.0^{+1.0}_{-0.6}$	$1.00^{+1.06}_{-0.97}$ (tot.) $^{+0.63}_{-0.61}$ (stat.) $^{+0.85}_{-0.76}$ (syst.)
Lepton+jets	X.X	$1.1^{+0.5}_{-0.3}$	$1.00^{+0.59}_{-0.56}$ (tot.) $^{+0.29}_{-0.29}$ (stat.) $^{+0.59}_{-0.56}$ (syst.)
Combined	X.X	$1.0^{+0.4}_{-0.3}$	$1.00^{+0.52}_{-0.48}$ (tot.) $^{+0.27}_{-0.26}$ (stat.) $^{+0.44}_{-0.40}$ (syst.)

Blinded

TTH@LHC

- Most recent $t\bar{t}H$ results by ATLAS and CMS (significance, $\mu_{t\bar{t}H} = \sigma_{t\bar{t}H}/\sigma_{SM}$)

	 36.1 fb ⁻¹	
Run-1 combination	JHEP 1608 (2016) 045 4.4σ (exp: 2.0σ) $\mu_{t\bar{t}H} = 2.3^{+0.7}_{-0.6}$	
$t\bar{t}H(b\bar{b})$	ATLAS-CONF-2017-076 1.4σ $\mu = 0.84^{+0.64}_{-0.61}$	CMS-PAS-HIG-16-038 (13 fb ⁻¹) $\mu_{t\bar{t}H} = -0.19 \pm 0.8$
$t\bar{t}H$ multilepton	ATLAS-CONF-2017-077 4.1σ $\mu = 1.6^{+0.5}_{-0.4}$	CMS-PAS-HIG-17-004 (ℓ only) 3.3σ (exp: 2.5σ) $\mu_{t\bar{t}H} = 1.5 \pm 0.5$ CMS-PAS-HIG-17-003 (τ_{had}) 1.4σ (exp: 1.8σ) $\mu_{t\bar{t}H} = 0.72^{+0.62}_{-0.53}$ arXiv:1706.09936
$t\bar{t}H(ZZ \rightarrow 4\ell)$	ATLAS-CONF-2017-043 $\mu_{t\bar{t}H} < 7.7$	$\mu_{t\bar{t}H} < 1.18$
$t\bar{t}H(\gamma\gamma)$	ATLAS-CONF-2017-045 1.8σ (exp: 1.0σ) $\mu_{t\bar{t}H} = 0.5 \pm 0.6$	CMS-PAS-HIG-16-040 3.3σ (exp: 1.5σ) $\mu_{t\bar{t}H} = 2.2^{+0.9}_{-0.8}$

SUMMARY

- ✱ $ttH(bb)$ directly probes directly top-Higgs coupling
- ✱ It also has a few challenges:
 - ✱ Small production XS: split events to channels/categories
 - ✱ Higgs invariant mass not applicable: use BDT/MEM to extract signal
 - ✱ Difficult $tt+HF$ bkg: MEM, extra uncertainty
- ✱ Latest results are approaching SM sensitivity
- ✱ Expecting ' 5σ ' significance for combined 13TeV ttH searches with both 2016 and 2017 data

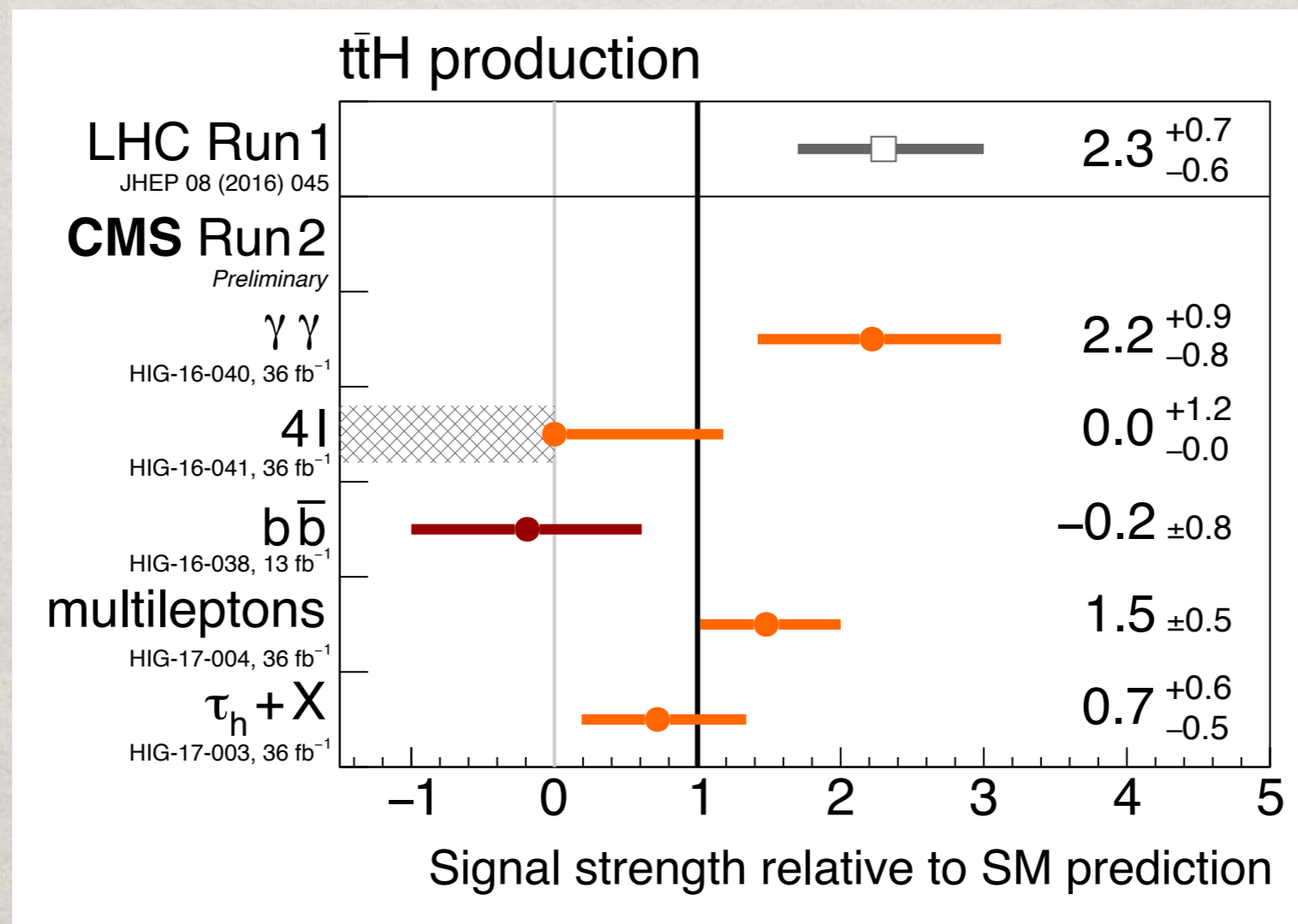
BACK UP

BDT TRAINING

- ✱ BDT training is a **machine learning** process
- ✱ **Aim to classify/separate signal or background**
- ✱ For the BDT training in each jet/tag category
 - ✱ ttH as signal(S) and ttjets as background(B)
 - ✱ events split in half: one for training and one for testing of overtraining

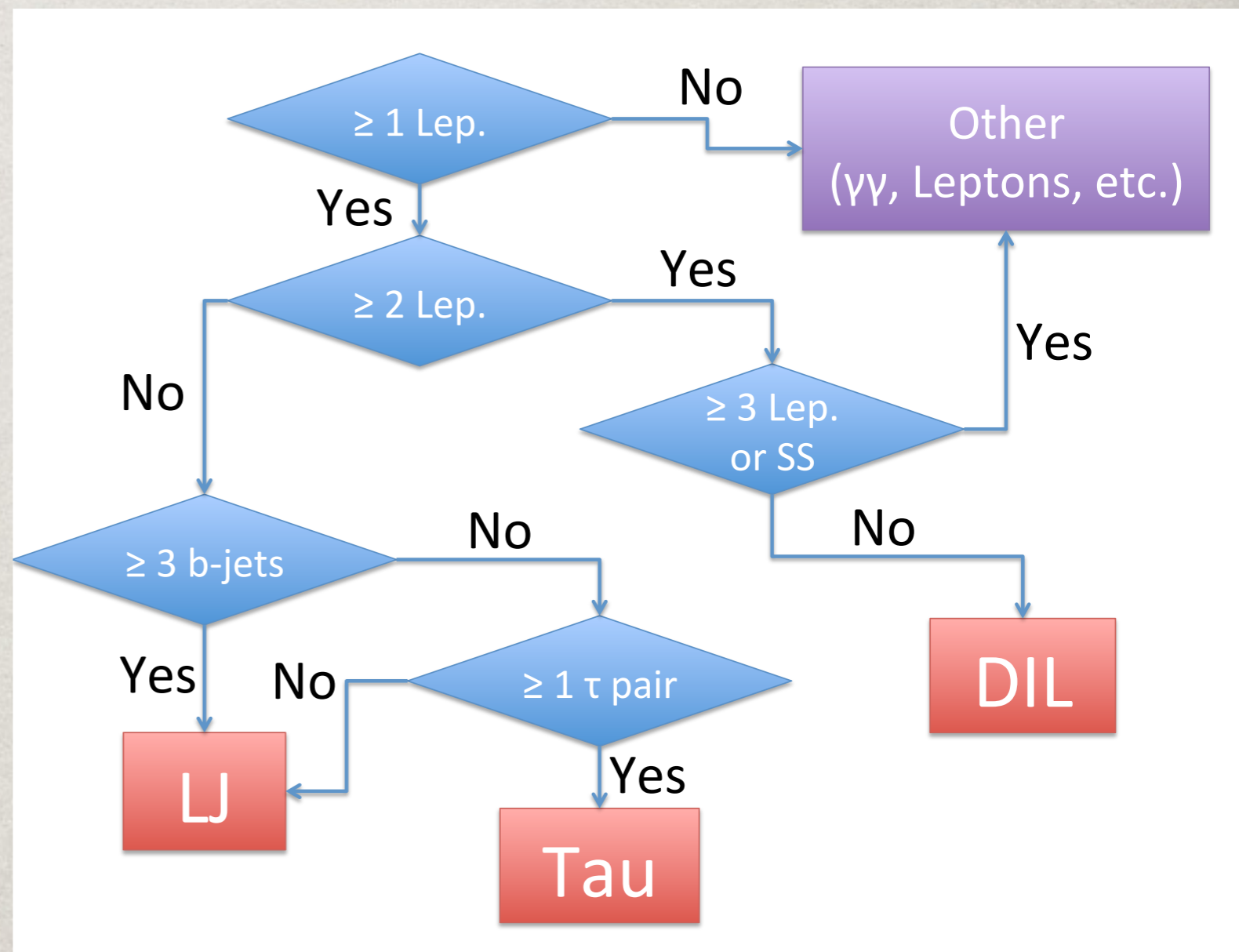
OUTLOOK

- ✿ Updated $t\bar{t}H(bb)$ results with 36/fb data come out soon
- ✿ Combined 13 TeV $t\bar{t}H$ searches (all decay modes) might yield interesting findings



SEARCH CHANNELS

- ✿ Different channels based on **top pair decay**(number of leptons) and **Higgs decay**(number of b-jets)



III: Matrix Element

Numerical
integration

Momentum conservation

Resolution function (allow ISR)

$$w(\vec{y}|\mathcal{H}) = \sum_{i=1}^{N_a} \int \frac{dx_a dx_b}{2x_a x_b s} \int \prod_{k=1}^8 \left(\frac{d^3 \vec{p}_k}{(2\pi)^3 2E_k} \right) (2\pi)^4 \delta^{(E,z)} \left(p_a + p_b - \sum_{k=1}^8 p_k \right) \mathcal{R}^{(x,y)} \left(\vec{p}_T, \sum_{k=1}^8 p_k \right) \times$$

$$g(x_a, \mu_F) g(x_b, \mu_F) |\mathcal{M}_{\mathcal{H}}(p_a, p_b, p_1, \dots, p_8)|^2 W(\vec{y}, \vec{p}),$$

Parton density functions

LO Scattering amplitude
(Open Loops)

Detector transfer function

$$P_{s/b} = \frac{w(\vec{y}|\text{ttH})}{w(\vec{y}|\text{ttH}) + k_{s/b} w(\vec{y}|\text{tt}+\text{bb})}$$

Construct per-event signal/background probability
using full kinematic information in an analytical approach

Ideal for final states with many reconstructed objects.

Built for ttH(bb) vs ttbb

SYSTEMATIC UNCERTAINTIES

- ✱ PileUp re-weighting: use $69.4 \text{ mb} \pm 7\%$
- ✱ Lepton SF
 - ✱ independently vary id and HLT efficiency
- ✱ **b-tag SF**
- ✱ top Pt reweighting
- ✱ JER and JES
 - ✱ JER has a negligible effect on shape or normalization
- ✱ Luminosity: 4.4%
- ✱ Cross section
 - ✱ Use CMS standard model cross section uncertainties
- ✱ **MC statistics**
- ✱ **Q^2 scale for MadGraph ttjets**
- ✱ **extra 50% rate uncertainty for tt+HF**
- ✱ Tau efficiency, Tau fake-rate and Tau energy scale

DATA/MC AGREEMENT: LJ

- ☼ All corrections to MC applied
- ☼ Good agreement between Data and MC

