



Performance of reconstruction and identification of τ leptons in their hadronic decays in pp collisions at $\sqrt{s} = 13$ TeV at CMS

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Speaker:

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(on behalf of CMS collaboration)



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Introduction

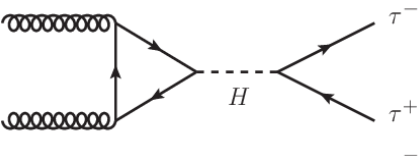
■ Features:

- hadronic decays ($m = 1.78 \text{ GeV}$, $\text{Br}=65\%$)
- large lifetime of ($2.9 \times 10^{-13} \text{ s}$)

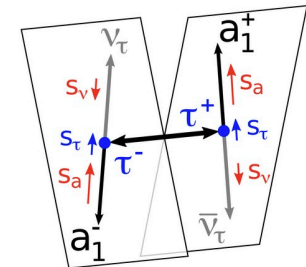
■ Motivation:

→ SM tests:

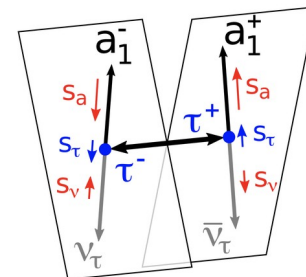
- » Higgs Yukawa couplings proportional to the mass
- » highest branching ratio due to the highest mass among leptons.



- Possible final state CP measurements in $H \rightarrow \tau\tau$
- Measurement of Higgs to fermion coupling via $H \rightarrow \tau\tau$



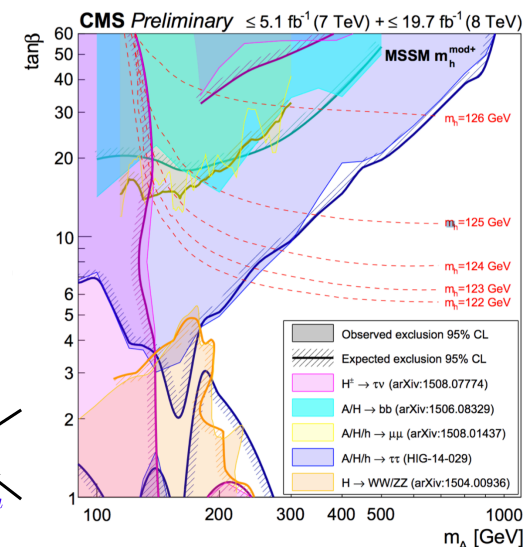
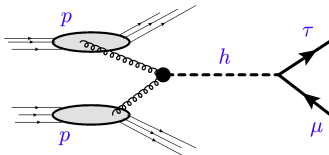
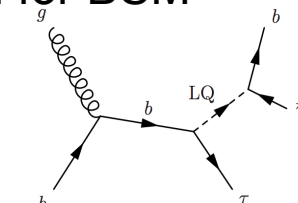
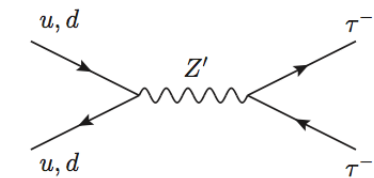
(a) Decay planes for $S = 1$ (scalar).



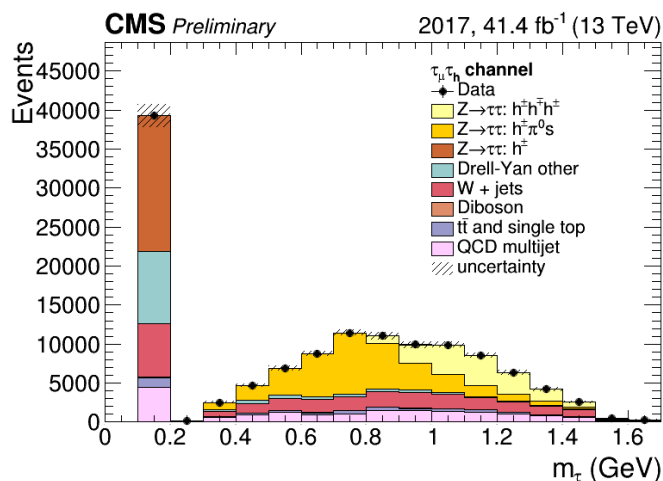
(b) Decay planes for $S = 0$ (pseudoscalar).

→ Often pure or enhanced channel for BSM searches:

- » Z', W' decays
- » Leptoquarks decays
- » Decays of MSSM Higgs boson at large $\tan(\beta)$
- » Lepton non-universality tests
- » CP violation in $B \rightarrow D^{**} \tau \nu$



- A "hadrons-plus-strips" (**HPS**) is based on reconstructed jets reprocessing:
 - Reassembles **jets(anti-kt, R=0.4)** to its constituents
 - » PF: electrons, photons, muons, hadrons
 - Reconstructs the π^0 (as "Strips"):
 - » 2016: $p_T > 0.5$ GeV ; 2017: $p_T > 1$ GeV
 - Reconstructs all possible decay modes (**DM**)
 - The **DM** with highest p_T and charge/strip multiplicity that pass the mass cut-window is assigned to each jet



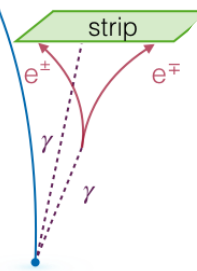
h^\pm

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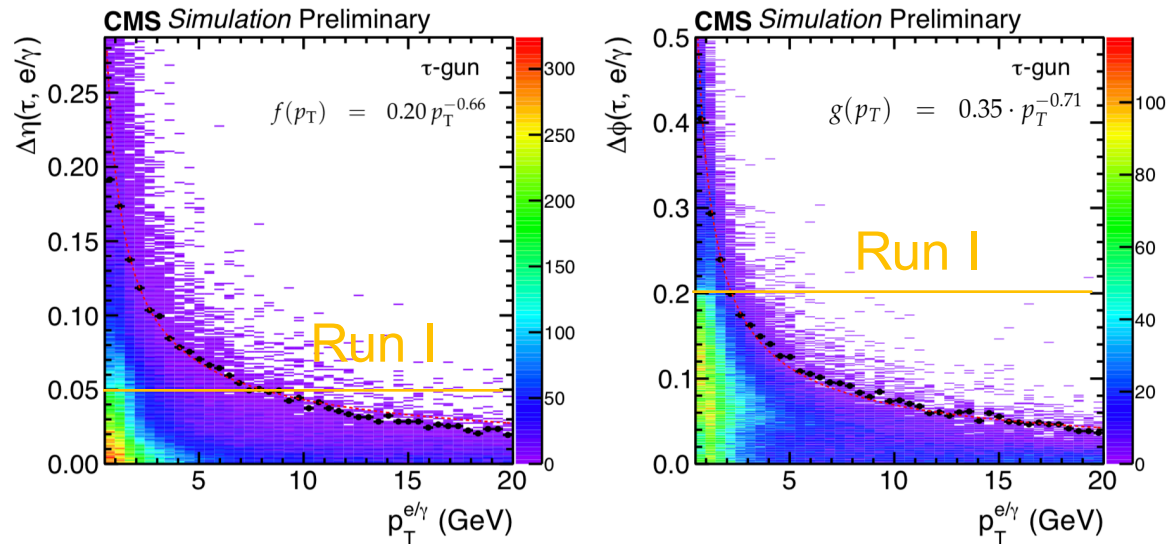
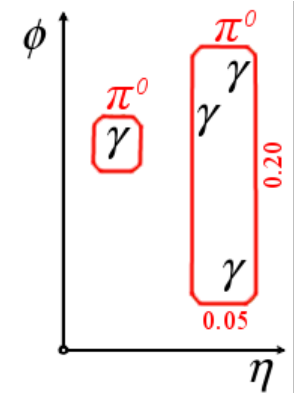
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Tau reconstruction and ID: Dynamic-strip reconstruction

1. e/γ with largest p_T (seed)
2. Merge p_T of the next largest e/γ within a strip
3. Strip position is recomputed
4. Until no e/γ to merge



Discrimination against jets

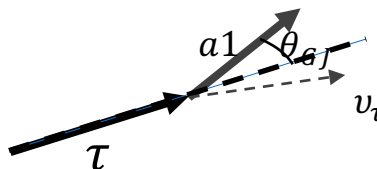
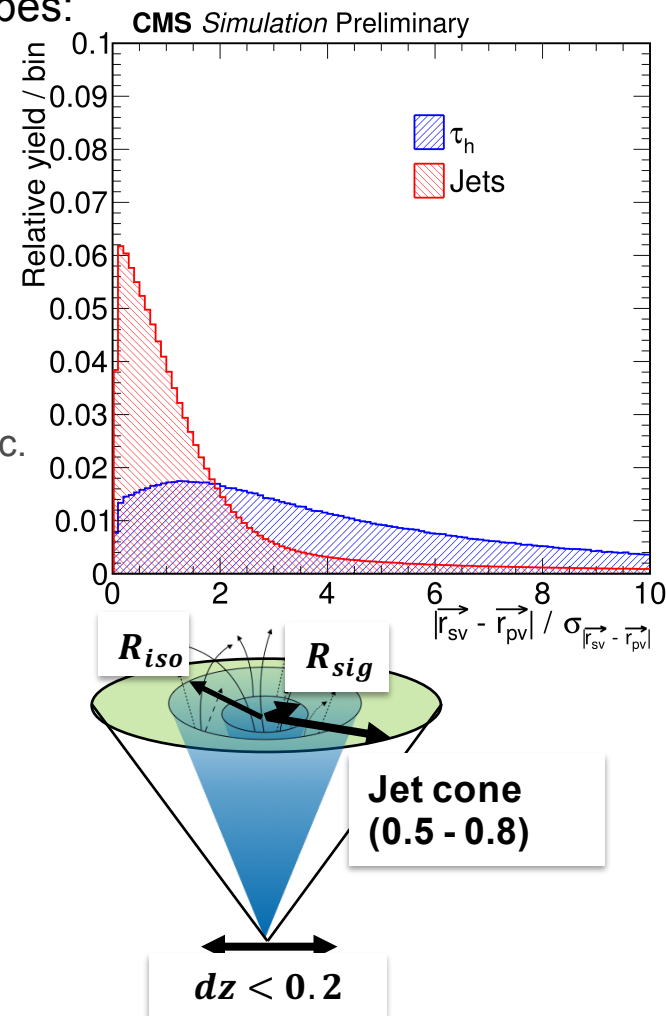
$$I_\tau = \sum p_T^{\text{charged}}(d_Z < 0.2 \text{ cm}) + \max\left(0, \sum p_T^\gamma - \Delta\beta \sum p_T^{\text{charged}}(d_Z > 0.2 \text{ cm})\right)$$

$$p_T^{\text{strip, outer}} = \sum p_T^{e/\gamma}(\Delta R > R_{\text{sig}}) < 0.10 \cdot p_T^\tau$$

Cut-based

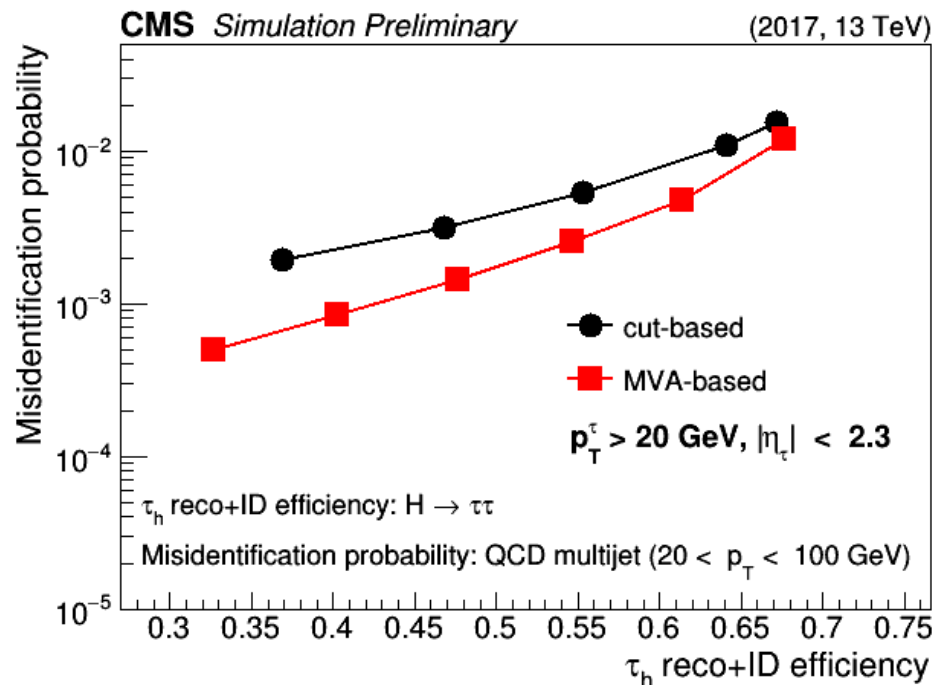
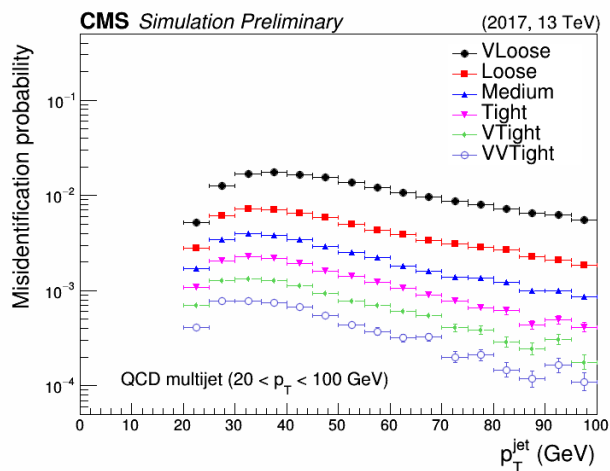
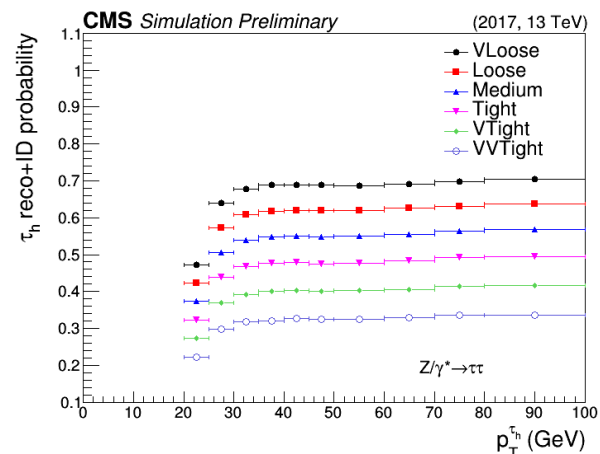
- MVA: BDT is seeded with 23 input variables of types:

- Isolation sums
- Distributions and multiplicities of particles inside/outside the cone:
 - » number of photons in tau, photons P_T outside signal cone etc.
- Lifetime-related:
 - » flight length (significance), secondary vertex, impact parameter vector etc.
- Tau quantities:
 - » p_T^τ , $|\eta_T^\tau|$, decay mode(DM), energy ratios etc.
- 2017 training strategy improvements:
 - » added Gottfried-Jackson angle for 3 prong decay mode
 - » p_T cut-off for photons is increased



Discrimination against jets

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The working points (WP) are chosen to have isolation efficiencies between 40%(VVTight) and 90%(VLoose), in steps of 10% and 95%(VVLoose)

Tau ID scale factors

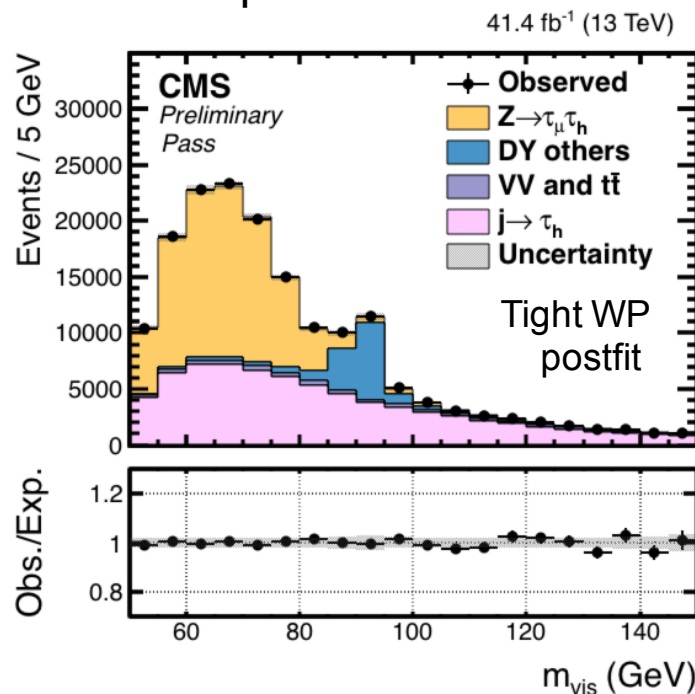
- Scale factors (SF) are measured using the Tag&Probe

- The most significant correction
→ <5% uncertainty

- For high tau p_T^τ the SF and it's uncertainty are extrapolated

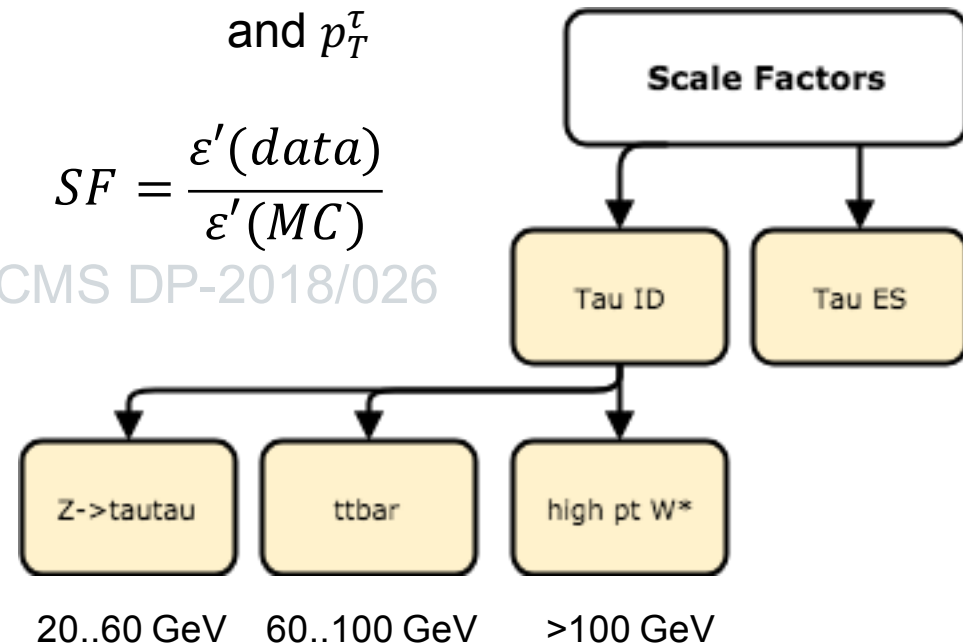
Example for $Z \rightarrow \mu\tau_h$:

- Tag: well identified μ
- Probe: reco tau with different isolation requirement
- Maximum-likelihood fit of **visible mass** performed in pass and control ($Z \rightarrow \mu\mu$) regions
- Backgrounds from MC but...
- W+Jets & QCD estimated from the data ($Z \rightarrow \mu\mu + jet \rightarrow \tau_h$) per DM and p_T^τ

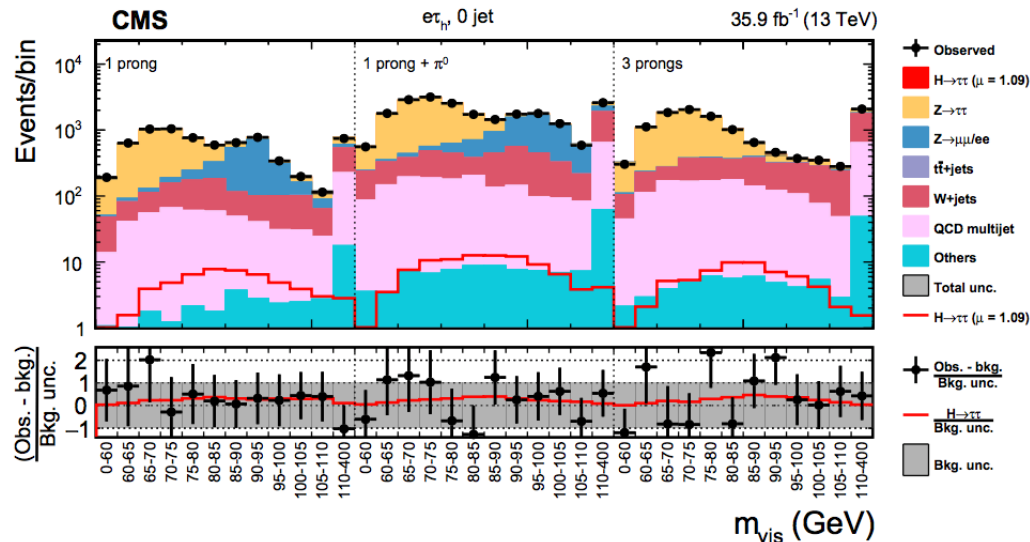


$$SF = \frac{\varepsilon'(data)}{\varepsilon'(MC)}$$

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Discrimination against electrons



Background events in final states involving taus can arise from $Z \rightarrow ee$ decays, where the electrons can be misidentified as taus

CMS-HIG-16-043

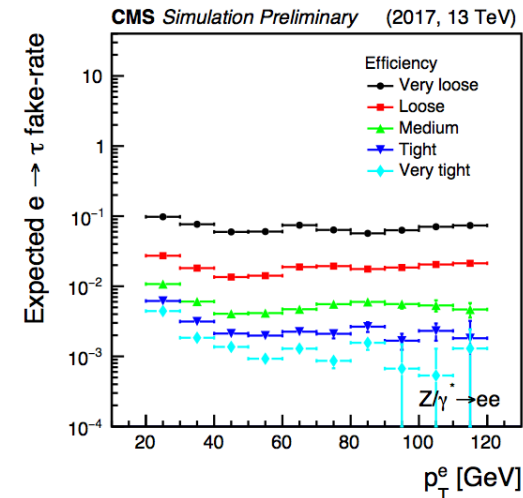
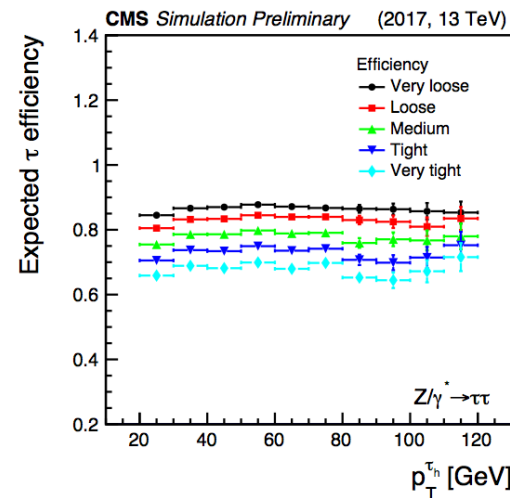
MVA BDT classifier

Input variables:

- Photon multiplicity per strip
- Shower-based variables
- Relation of strips and τ_h quantities

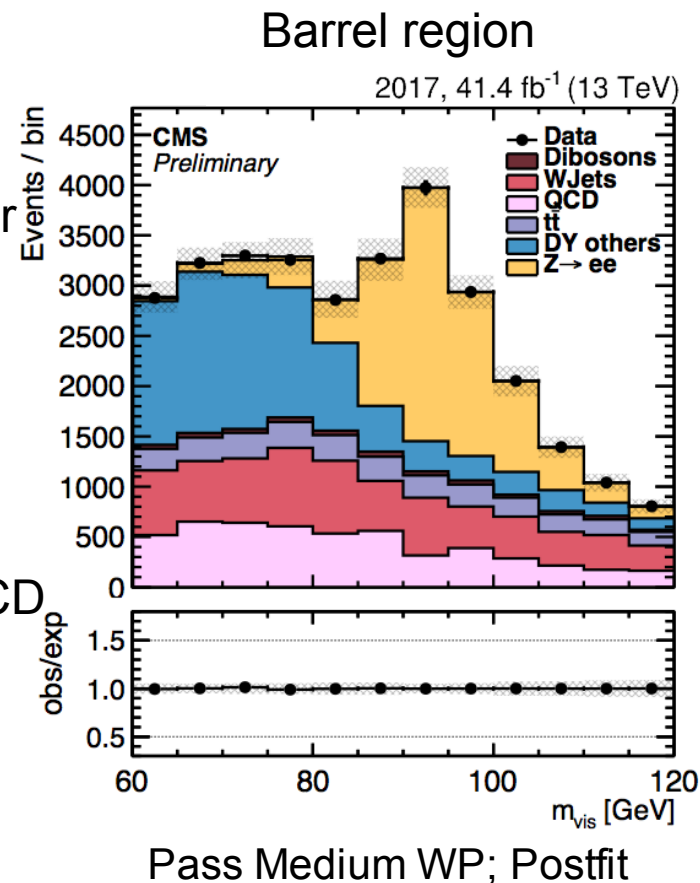
75% Efficiency

- <1% misID



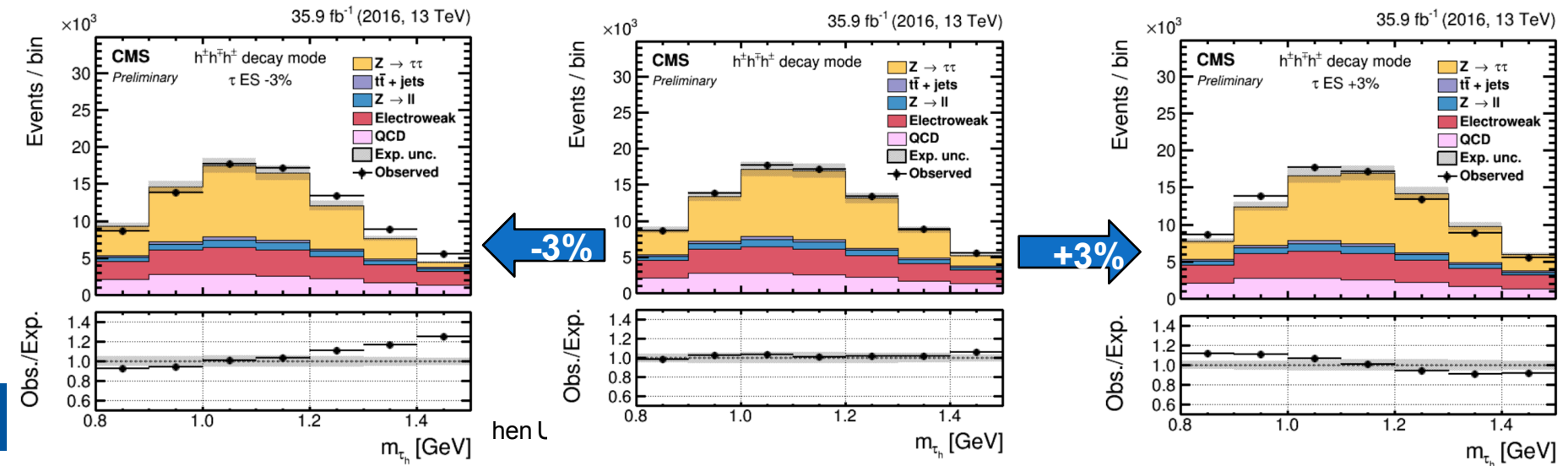
- $Z \rightarrow ee$
- SFs measured separately in barrel or endcap for different WP of anti-e discriminator using Tag&Probe
- Maximum-likelihood fit of **visible mass** performed separately in pass, fail:
 - probe(loose τ_h) is matched to generated electron
 - Backgrounds from MC but Wjets and QCD
- The MisID fake rate was shown to be compatible in Barrel and Forward regions (within 10%) and be strongly dependent on the chosen WP
 - 5% to 40% uncertainty

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- Sample: $Z \rightarrow \tau\tau \rightarrow \mu\tau_h/e\tau_h$
 - Both approaches are compatible across all decay modes (separately and combined)
- Shape templates are produced by shifting τ_h energy
- Results are compatible with unity within the uncertainty between different DM

$$ES = \frac{p_{preco}^{\tau,vis}}{p_{gen}^{\tau,vis}}$$



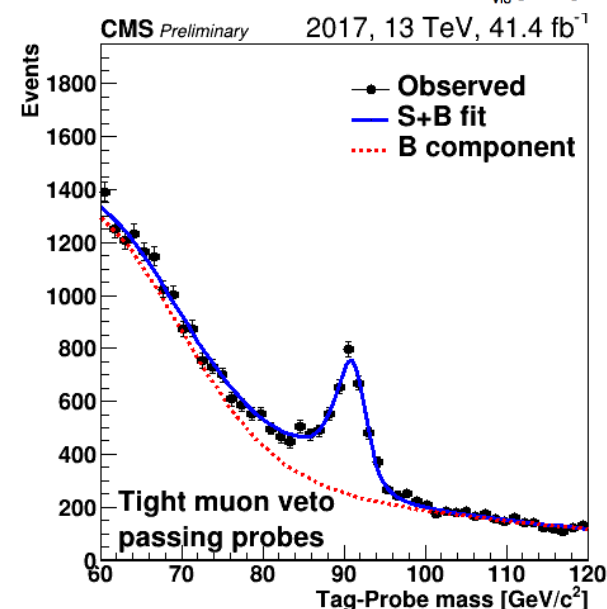
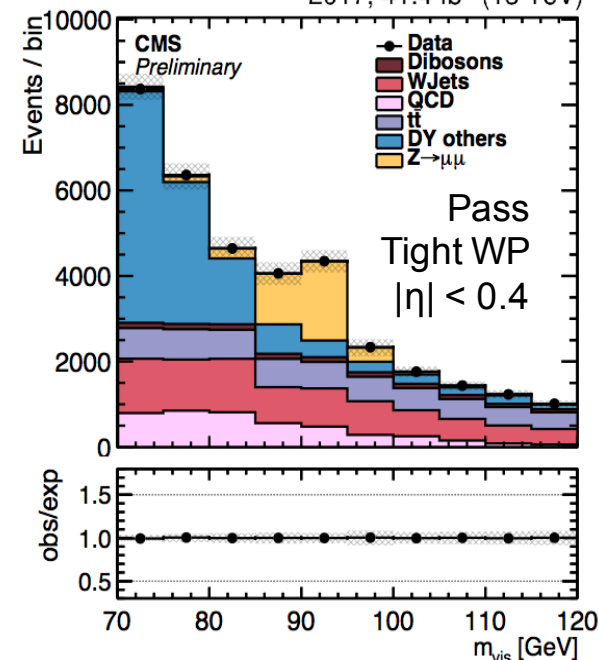
Fake rates measurements: $\mu \rightarrow \tau_h$

2017, 41.4 fb⁻¹ (13 TeV)

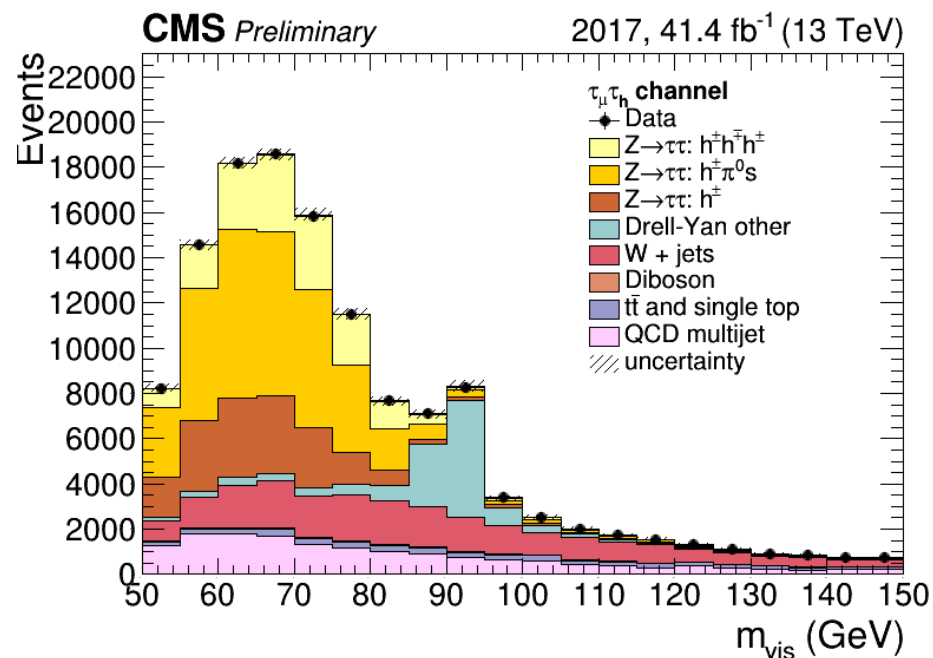
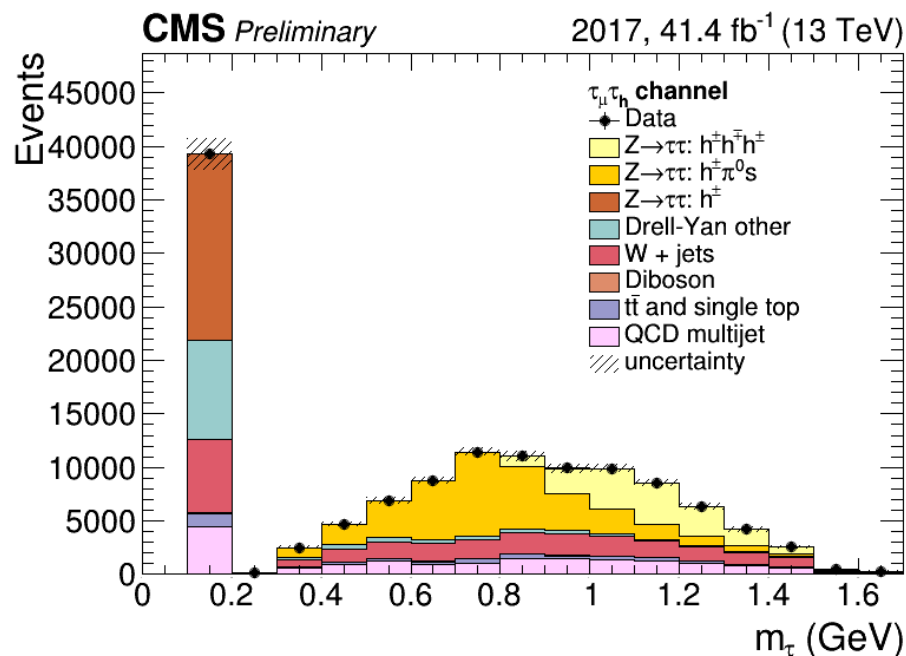
Cut-based anti-muon discriminator:

- Veto candidates matching segments in the muon detector
- ~100% efficiency

- $Z/\gamma^* \rightarrow \mu\mu$
- Measured in η bins per WP of anti- μ discriminator
- Tag&Probe:
 - Classical: CMS DP-2018/026
 - From ML fit of the tag and probe
 - Close to 1 in central region and up to 2 in forward
 - <10% uncertainty
 - Analytical:
 - the normalizations of signal ($Z/\gamma^* \rightarrow \mu\mu$) and background ($Z/\gamma^* \rightarrow \tau\tau$) component is done using the analytic functional forms fit to the data distributions.
 - compatible results with two methods



- After calibrations, we observe a very good agreement between data and simulation, proving an excellent understanding of the detector and the reconstruction!
- The various decay modes of the tau leptons are well resolved and identified



Appendix plan

- Signal samples covering large p-T contain:
 - H→tautau (NLO, Powheg v2)
 - Z'→ll, W'→lν (LO, Pythia8)
 - Matching: $\Delta R(reco, gen) = \sqrt{(\Delta\eta)^2 + (\Delta\phi)^2} < 0.3$
- Other samples:
 - Z/gamma→ll (LO, Madgraph_aMCatNLO)
 - W+jets(LO, Madgraph_aMCatNLO 2.2.2)
 - Single top (NLO, Powheg 2)
 - ttbar (NLO, Powheg 2)
 - Di-boson (NLO, Madgraph_aMCatNLO 2.2.2 or Powheg v2)
 - QCD (Pythia)
 - Highly-virtual W bosons (m>200 GeV, taunu, munu) (Pythia8)

↑

↓

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Normalised to NNLO

Normalised to NLO
- Hadronisation, Pile-up: Pythia8 + CUETP8M1
- PDF: NNPDF3.0

Tau reconstruction and ID: HPS:: dynamic-strip algorithm motivation

- Previously: $\Delta\eta \times \Delta\phi = 0.05 \times 0.20$
 - **nuclear interactions** of $p_{i\pm}$ with the tracker material, which create low p_T e/γ that may go outside the fixed strip
 - **conversions of photons** from π^0 decays to electron/positron pairs, and bremsstrahlung
- Idea:
 - ~~Increase size~~: more fakes; for boosted taus this size should be even smaller!
 - Strip size adjusted dynamically as a **function of the p_T of e/γ**

Classes of discriminating variables

- Isolation sums:

- $I_{\tau,ch}$, $I_{\tau,neut}$

- Distributions and multiplicities of particles inside/outside the cone:

- photons P_T outside signal cone, number of photons in tau etc.

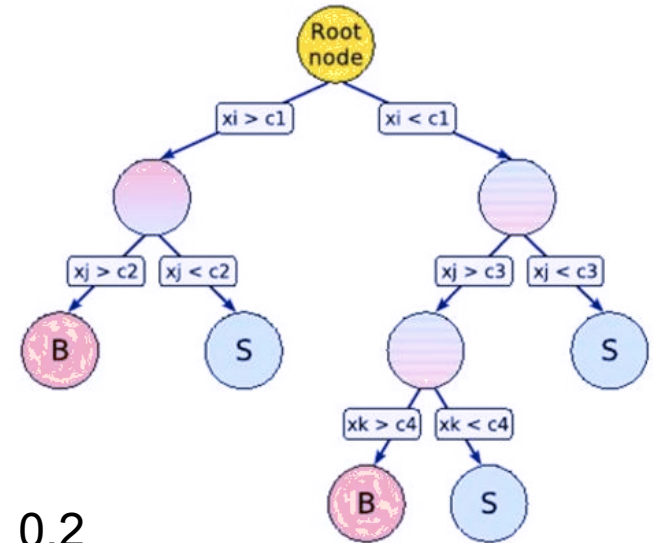
- Lifetime-related:

- SV, flight length, flight length significance, IP, etc.

- Tau quantities:

- $P_{T,\tau}$, $|\eta_{reco,\tau}|$, DM, Gottfried-Jackson-angle, energy ratio etc.

- Tool: TMVA
- Gradient Boost BDT
 - Classification & regression
 - 1000 trees in the forest
 - Boosting type for the trees in the forest: Gradient
 - Learning rate for GradBoost algorithm: 0.2
 - Use only a random subsample of all events for growing the trees in each iteration
 - Fraction of events to be used in each iteration: 0.5
 - Separation criterion for node splitting: GiniIndex
 - Number of grid points in variable range used in finding optimal cut in node splitting: 500
 - Maximum depth of cell tree: 5



Tau ID scale factors estimation using $Z \rightarrow \mu\tau_h$

- > events divided in two(anticorrelated) pass/fail region

$$\begin{array}{l}
 N_{pass} = \varepsilon \cdot N_0 \\
 N_{fail} = (1 - \varepsilon) \cdot N_0
 \end{array}
 \xrightarrow{\text{ML fit}}
 \begin{array}{l}
 N'_{pass} = \varepsilon' \cdot N_0 = r\varepsilon \cdot N_0 \\
 N'_{fail} = (1 - \varepsilon') \cdot N_0 = (1 - r\varepsilon) \cdot N_0
 \end{array}$$

with $r = \varepsilon' / \varepsilon$ as **parameter of interest(POI)** of the fit

so the yield in the pass/fail regions must scale as:

$$N'_{pass} / N_{pass} = (r\varepsilon \cdot N_0) / (\varepsilon \cdot N_0) = r$$

$$N'_{fail} / N_{fail} = (1 - r\varepsilon) / (1 - \varepsilon)$$

$$N_0 = N_{pass} + N_{fail}$$

matched $Z \rightarrow \mu\tau_h$ MC

$$\varepsilon = \frac{N_{pass}}{N_0} \text{ - prefit}$$

$$\varepsilon' \text{ - postfit}$$

$$r = \frac{\varepsilon}{\varepsilon'} \text{ - signal strength}$$

- > fit tag-probe pair visible mass, from 60 (70) GeV to 120 GeV

$$SF = \frac{\varepsilon'(data)}{\varepsilon'(MC)}$$