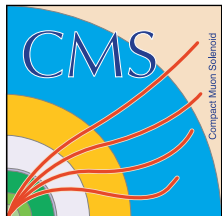


IPNL-PKU Collaboration on ttH at CMS

FCPPL - 22/05/2018, Marseille

Nicolas Chanon - IPNL, CNRS/IN2P3



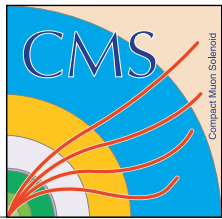
IPNL-PKU collaboration

IPNL-PKU is former IPHC-PKU collaboration

- The collaboration started in 2014, and was **formalised in 2015 within the FCPPL framework**
- **In 2016 and 2017: Ms. Jing Li visited IPHC for 3 months twice** (supported by FCPPL, IPHC, PKU)
- End of 2017 and **beginning of 2018, Junho Lee visited IPNL 1 week twice**, with support from FCPPL

Goals of the collaboration

- **New analysis techniques for CMS ttH multilepton analysis**
- Development of the **Matrix Element Method (MEM)**
- **Combination of MEM with multivariate techniques (NN, BDT)**
- Developments of new tools with **Deep Learning**

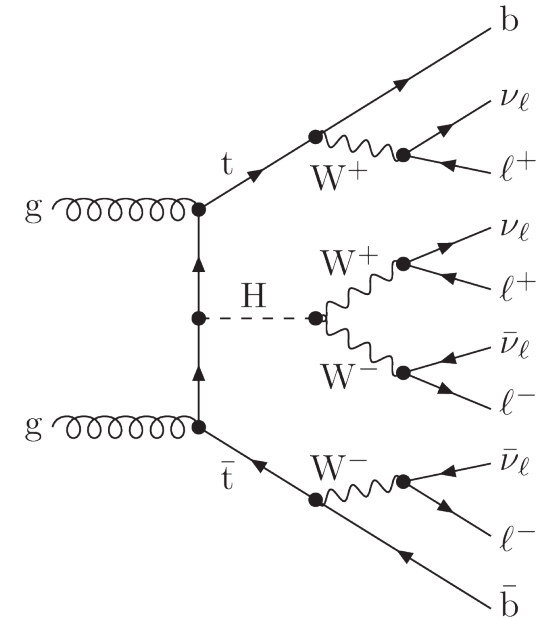


$t\bar{t}H$ multilepton

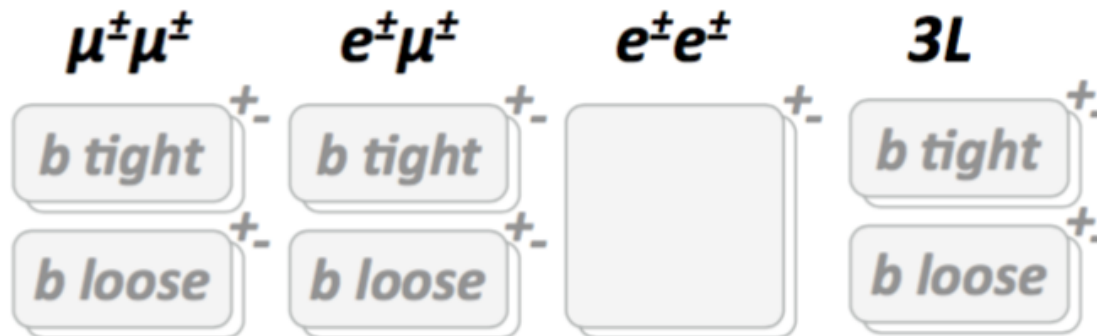
CMS HIG-17-004

Targeting final states with electrons and muons

- **2 same sign leptons:** ≥ 4 jets, ≥ 1 b-tag (same-sign required to reduce Drell-Yan and $t\bar{t}Z$)
- **3 leptons:** ≥ 2 jets, ≥ 1 b-tag
- **4 leptons:** same as 3 leptons, veto $H \rightarrow 4\ell$ (dedicated $H \rightarrow ZZ$)
- Veto presence of τ_h (dedicated $H \rightarrow \tau\tau$ analysis, see later)



Analysis categories:



Analysis sensitivity:

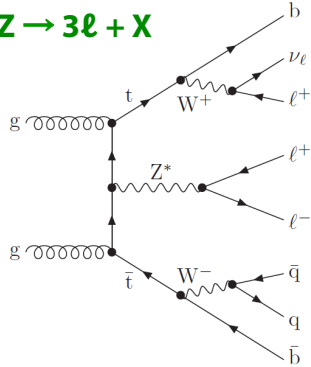
- 2ℓ ss and 3ℓ categories: Train **2 kinematic BDTs, against $t\bar{t}b$ and $t\bar{t}W/Z$**
- Map 2D BDTs into 1D discriminant (group into bins with similar s/b)



$t\bar{t}H$ multilepton: backgrounds

CMS HIG-17-004

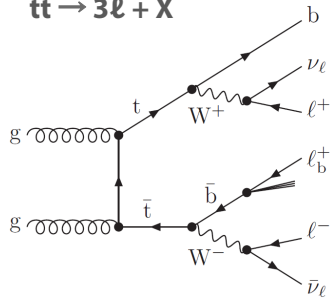
$ttZ \rightarrow 3\ell + X$



Irreducible: $tt+W/Z/\gamma^*$

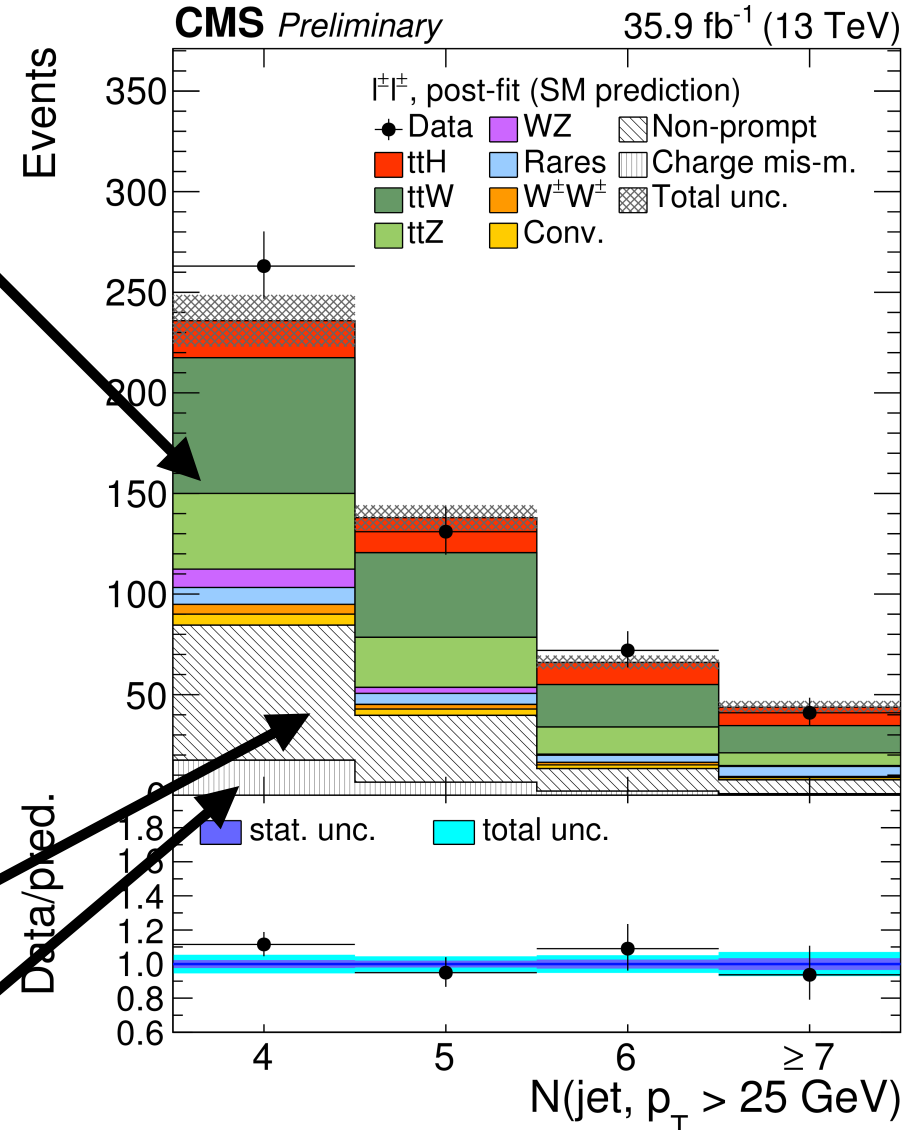
- from Monte Carlo,
- O(10%) uncertainty

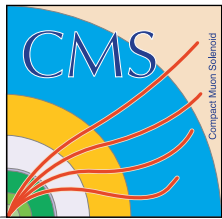
$tt \rightarrow 3\ell + X$



Reducible: mainly tt +jets,

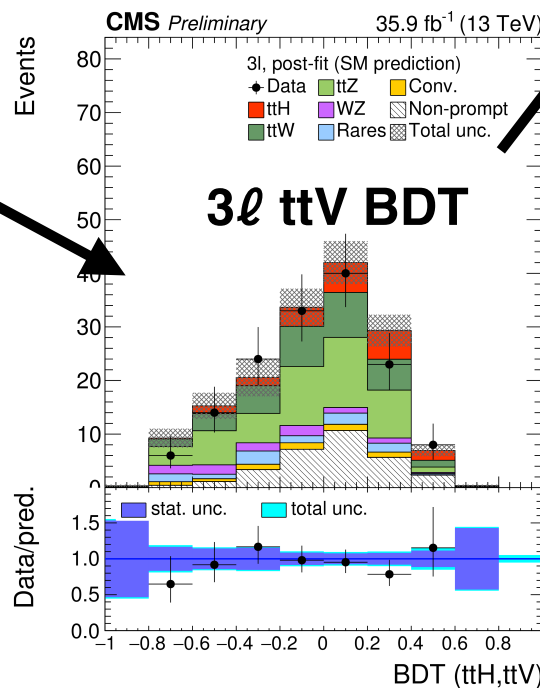
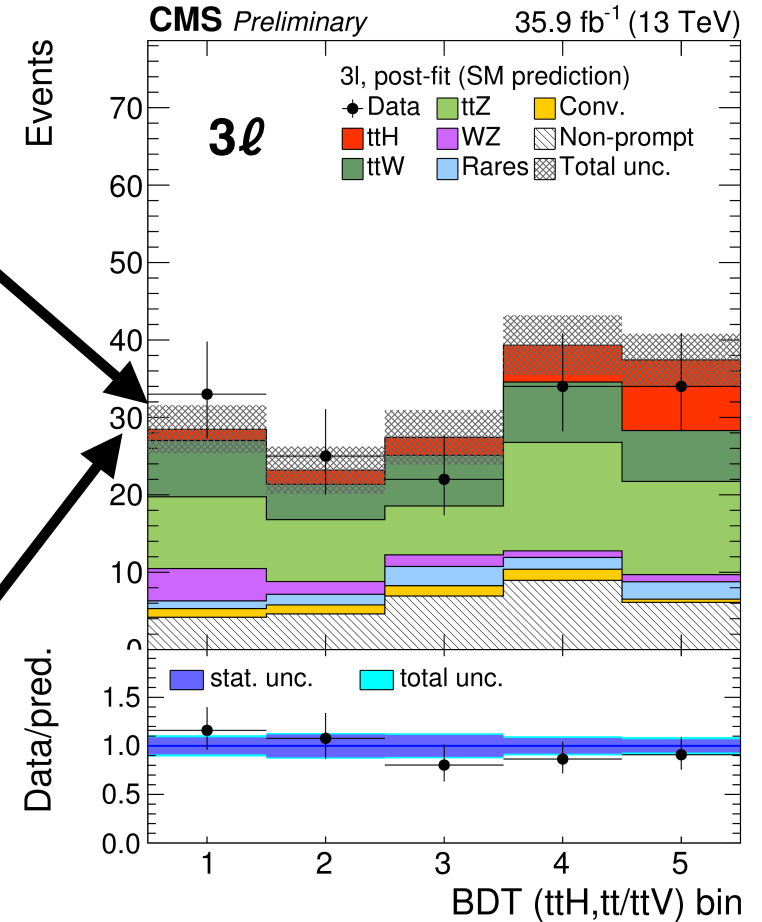
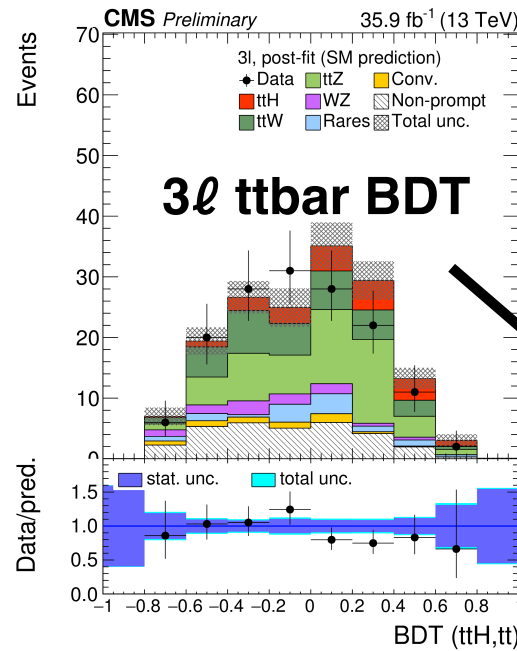
- shape obtained from data,
- O(30%) uncertainty
- **Jets faking leptons:** fake rate computed from QCD control region with loosened identification
- **Charge mis-assignment** (2ℓ ss only): flip rate from $Z \rightarrow \ell^\pm \ell^\pm$ data





$t\bar{t}H$ multilepton discriminants

CMS HIG-17-004



3ℓ MEM

CMS Preliminary 35.9 fb⁻¹ (13 TeV)

3ℓ, post-fit (SM prediction)

- Data
- ttZ
- Conv.
- ttH
- WZ
- Non-prompt
- ttW
- Rares
- Total unc.

stat. unc. total unc.

- 3ℓ vs $t\bar{t}W/Z$: Includes **Matrix Element Method** likelihood ratio of $t\bar{t}H$ vs $t\bar{t}W+t\bar{t}Z$

$\bar{t}tH$ multilepton : Matrix Element Method

Jing Li (PKU), NC

MEM weight

Integration with VEGAS

Matrix Element from Madgraph

$$w_{i,\alpha}(\Phi') = \frac{1}{\sigma_\alpha} \int d\Phi_\alpha \cdot \delta^4\left(p_1^\mu + p_2^\mu - \sum_{k \geq 2} p_k^\mu\right) \cdot \frac{f(x_1, \mu_F) f(x_2, \mu_F)}{x_1 x_2 s} \cdot \left| \mathcal{M}_\alpha(p_k^\mu) \right|^2 \cdot W(\Phi' | \Phi_\alpha)$$

Phase-space enforcing 4-momentum conservation

Parton distribution function from LHAPDF

Transfer functions from CMS simulation

Evaluate MEM weights under $\bar{t}tH$, $\bar{t}tW$, $\bar{t}tZ/\gamma^*$ hypotheses:

- Custom framework in C++
- **If jets are needed at ME level and are not reconstructed (“missing jets”):** included, as supplementary phase space to integrate
- MEM weight is the average weight of all possible **lepton, jets, b-jets permutations**



$t\bar{t}H$ multilepton : MEM in 3l category

CMS HIG-16-022, HIG-17-004

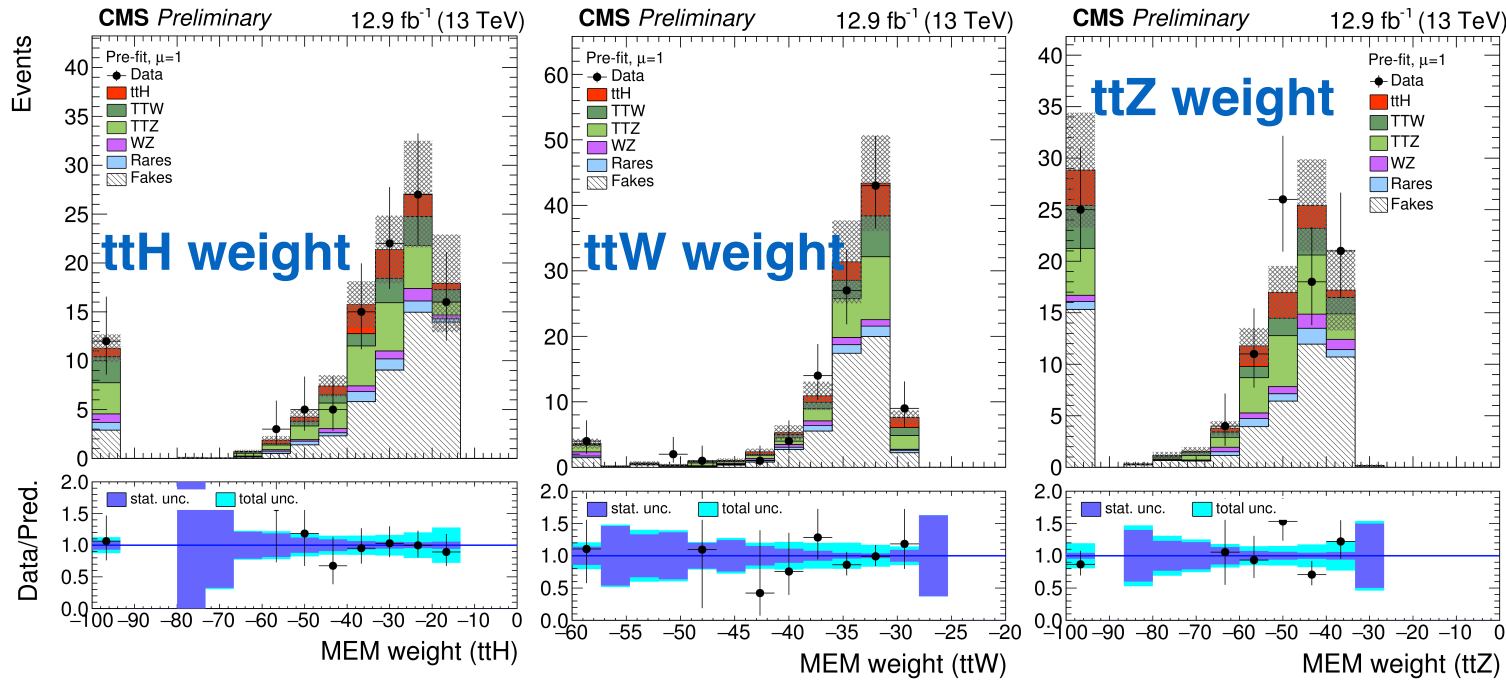
Jing Li (PKU), Xavier Coubez (formerly IPHC),
Daniel Bloch (IPHC), NC

HIG-16-022 (ICHEP 2016):

- improved discrimination by 10% in 3l category

- Include $\log(\text{weights})$ as input to a kinematic BDT trained against $t\bar{t}$

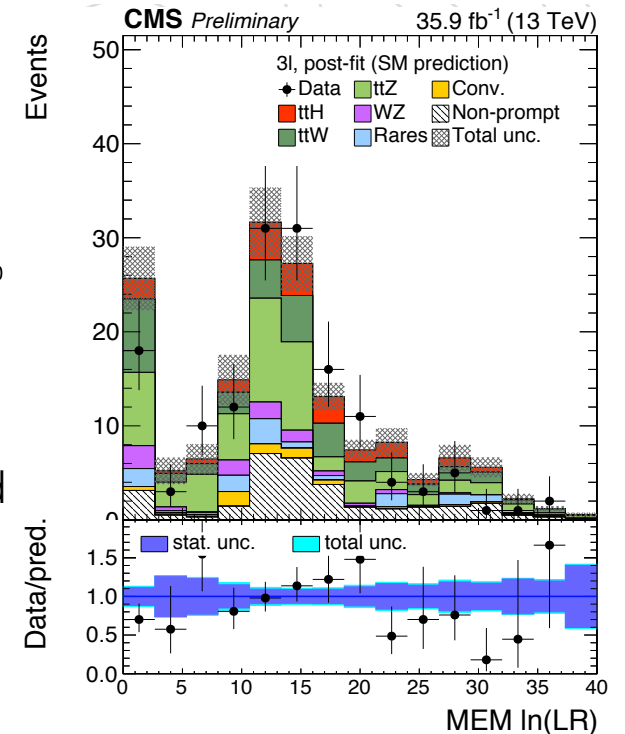
MEM weights under $t\bar{t}H$, $t\bar{t}W$, $t\bar{t}Z/\gamma^*$ hypotheses

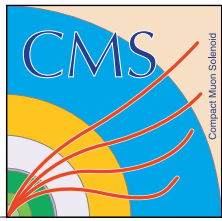


HIG-17-004 (Moriond 2017): include the likelihood of $t\bar{t}H$ vs $t\bar{t}V$ weights inside the $t\bar{t}W/Z$ BDT

Likelihood ratio of $t\bar{t}V$ vs $t\bar{t}H+t\bar{t}V$

$$-\log \left(\frac{\sigma_{t\bar{t}Z} w_{t\bar{t}Z} + k \cdot \sigma_{t\bar{t}W} w_{t\bar{t}W}}{\sigma_{t\bar{t}H} w_{t\bar{t}H} + \sigma_{t\bar{t}Z} w_{t\bar{t}Z} + k \cdot \sigma_{t\bar{t}W} w_{t\bar{t}W}} \right)$$





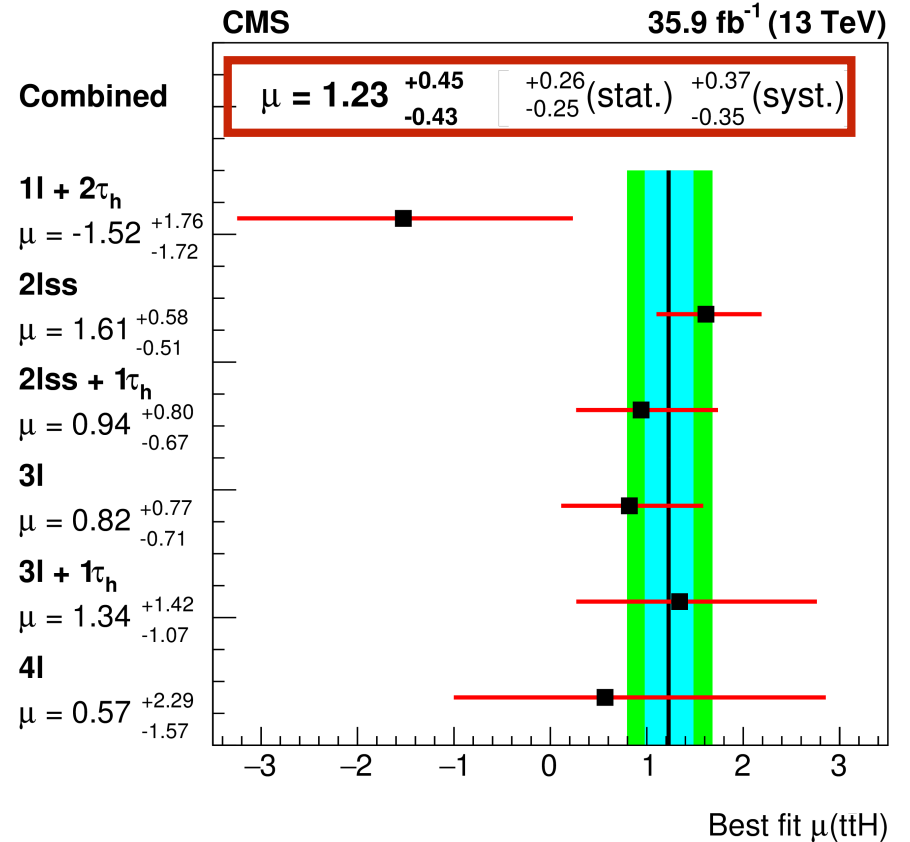
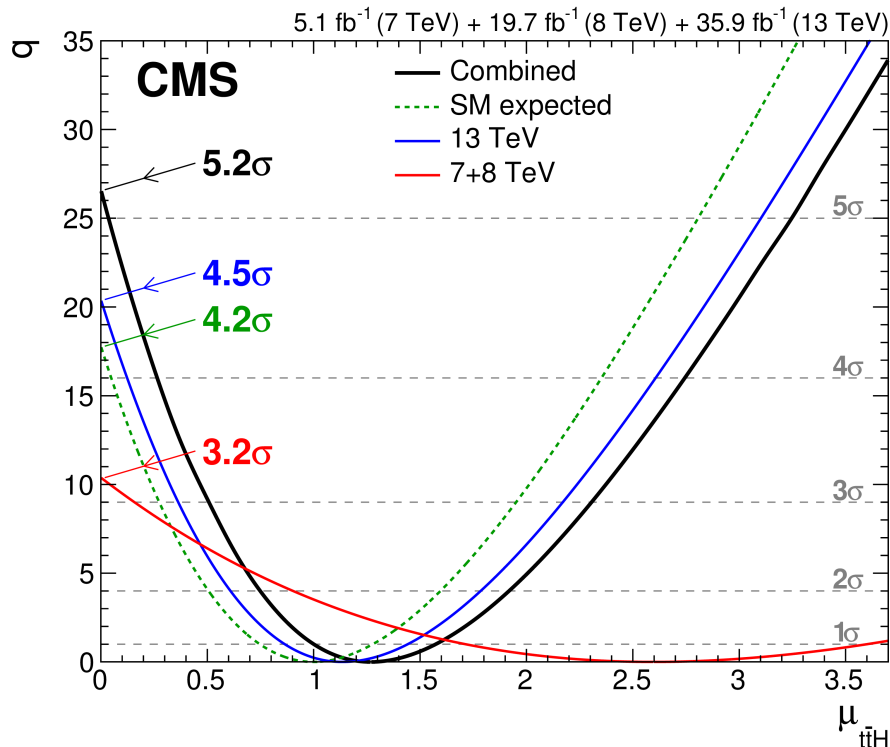
$t\bar{t}H$ multilepton results

arxiv:1803.05485, submitted to JHEP

Significance, combining $t\bar{t}H$ multilepton and tau analyses:
Observation: 3.2σ (2.8σ expected)

Our collaboration presented at

- Moriond QCD 2017 (NC)
- Poster at LHCP 2017 (Jing Li)

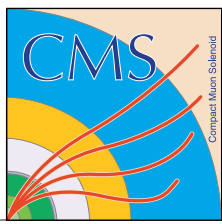


$t\bar{t}H$ observation

arxiv:1804.02610, accepted by Phys. Rev. Lett.

Full combination (Run 1, Run 2 $t\bar{t}H$ multilepton, τ_h+X , $\gamma\gamma$, ZZ and bb):

8 **5.2σ observed (4.2σ expected)**



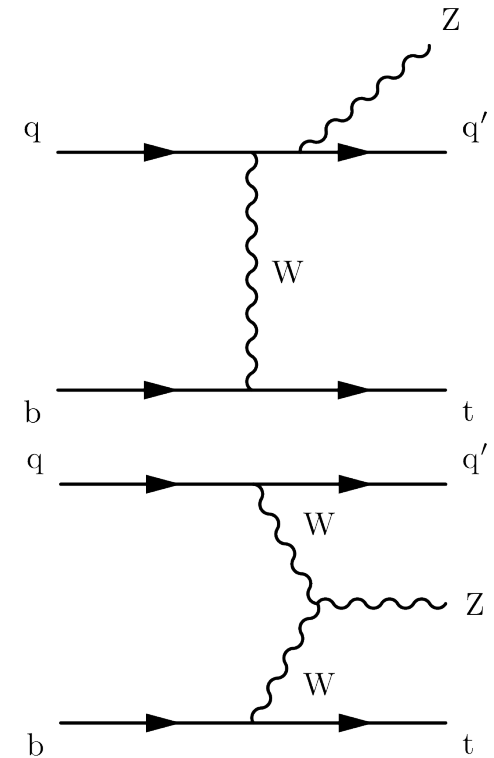
MEM in single top + Z

Phys. Lett. B 779 (2018) 358

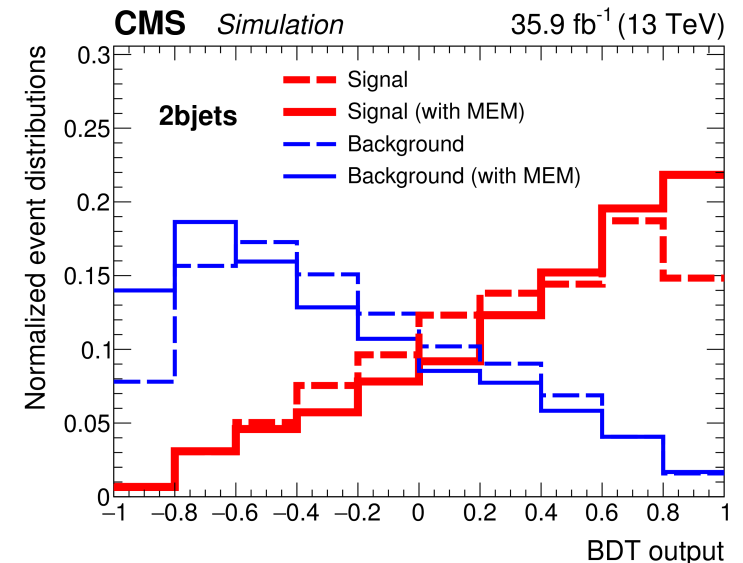
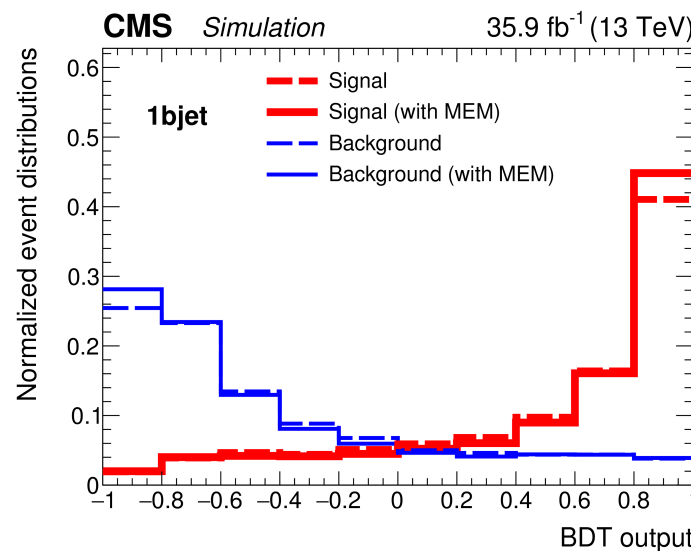
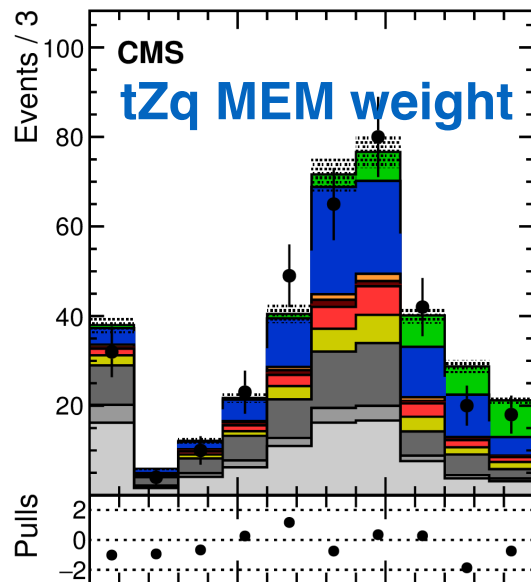
Nicolas Tonon (IPHC), Jeremy Andrea (IPHC), NC

As a spin-off, include MEM in CMS tZq analysis with 2016 data

- tZq is a rare process in the SM
- **Same MEM framework as developed for ttH multi lepton analysis**
- MEM is a powerful tool to use process kinematics: especially forward jet in tZq
- Include MEM weights and MEM as a kinematic fit
- **Observation 3.7σ (3.2σ expected)**



MEM improves the analysis significance by 20%

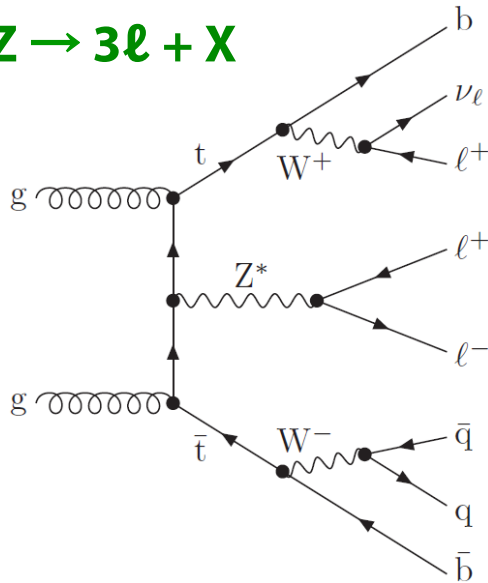


Going further with MEM (2017-)

(Preliminary)

Jing Li (PKU), NC

$ttZ \rightarrow 3\ell + X$

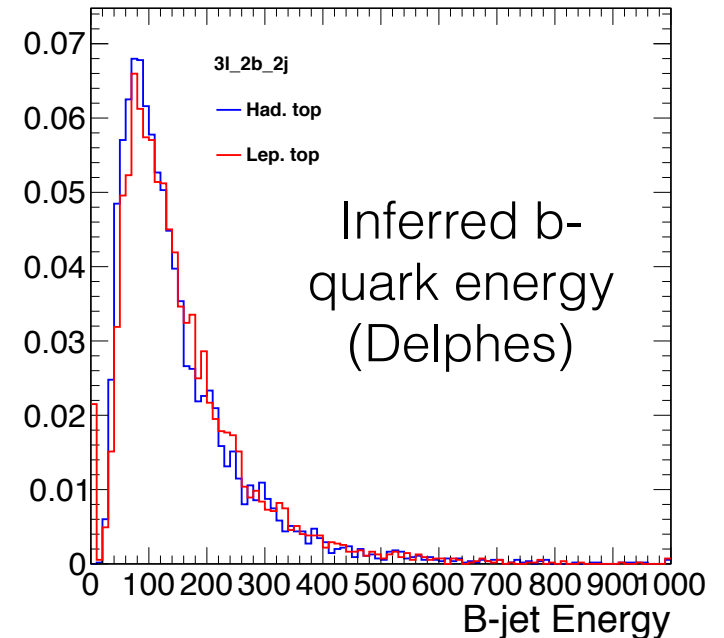
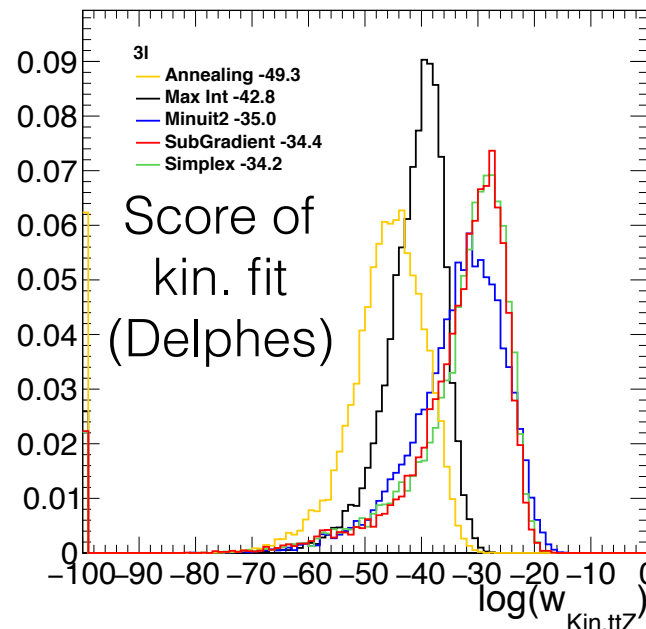


Studying MEM as kinematic fit

- Given reconstructed particles in the event, a MEM score can be attributed to the **most probable kinematic configuration**
- Also obtain unreconstructed quantities : neutrino momenta, etc.

- Studies on behaviour of MEM function
- Explore several minimisation algorithms
- **Target: phenomenology publication**

ttZ in 3 ℓ category, Delphes simulation



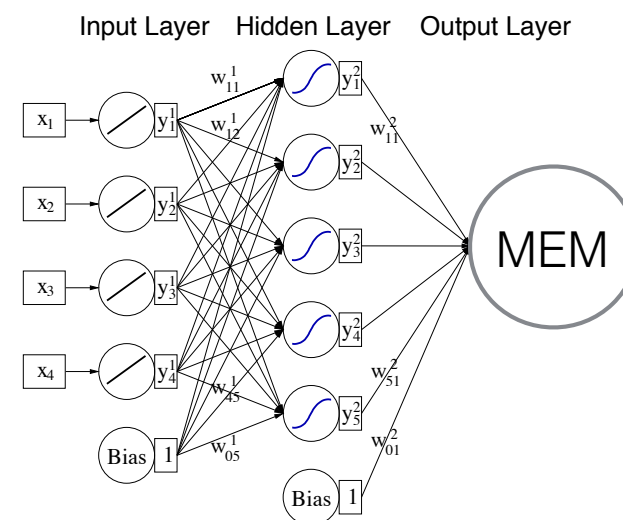
MEM and Deep Learning

(Exploratory work)

Jing Li (PKU), Junho Lee (PKU), Qiang Li (PKU), NC

Working on new hybridisations of MEM and multivariate techniques

- Infer unknown parton-level quantities needed to compute the MEM, using neural network
- **First studies performed** with custom code (Jing Li): showed the need for more powerful tools
- Moving to **modern Deep Learning tools** : Keras + Tensorflow (Junho Lee)
- As a first step, working on a regression of MEM : can be useful to accelerate evaluation of MEM in CMS analysis





Perspectives

- Involvement in **CMS ttH multi lepton analysis** with 2017 data and planned contribution to Run 2 legacy paper
- Working on a **phenomenology publication** to expose new analysis techniques, with MEM and Deep Learning
- Possible involvement in **CMS tHq analysis** with 2017 data
- Consider later possible contributions to VBS WW analysis

- FCPPL support in 2018 will allow a **3 months stay of Junho Lee at IPNL in Autumn**
- PKU, IPHC and IPNL are **grateful for continued FCPPL support** since the start of our collaboration

Back-up slides