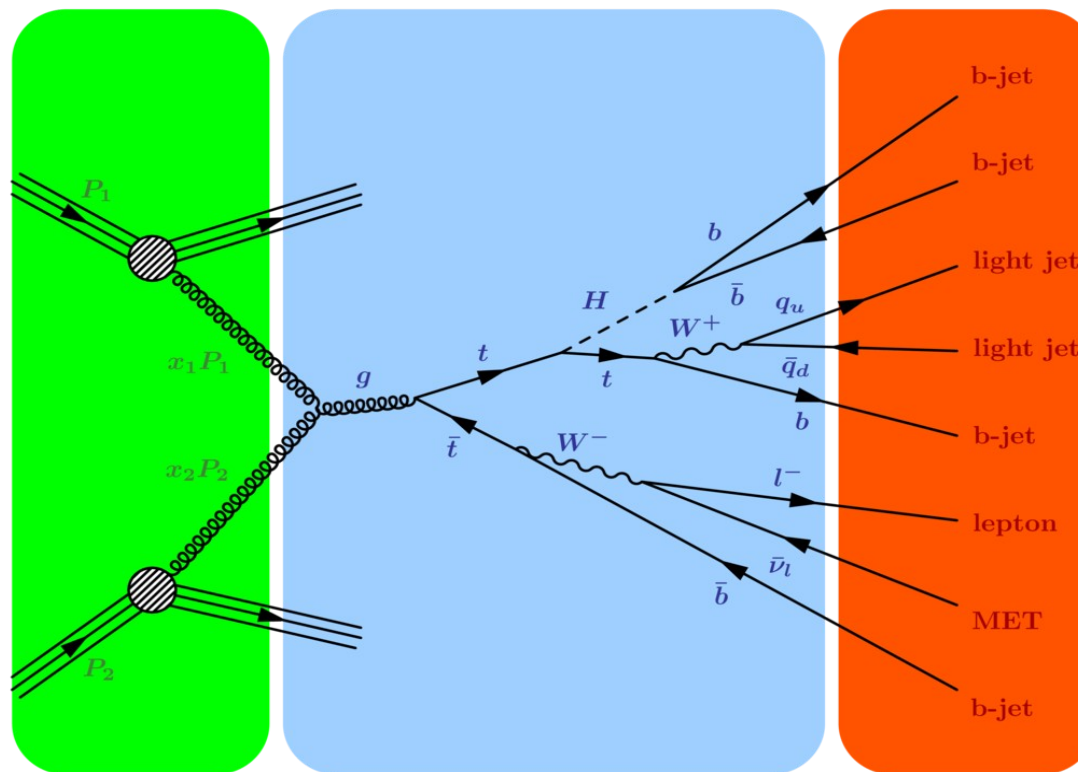


# MEM discussion: ATLAS $t\bar{t}(H \rightarrow b\bar{b})$ experience



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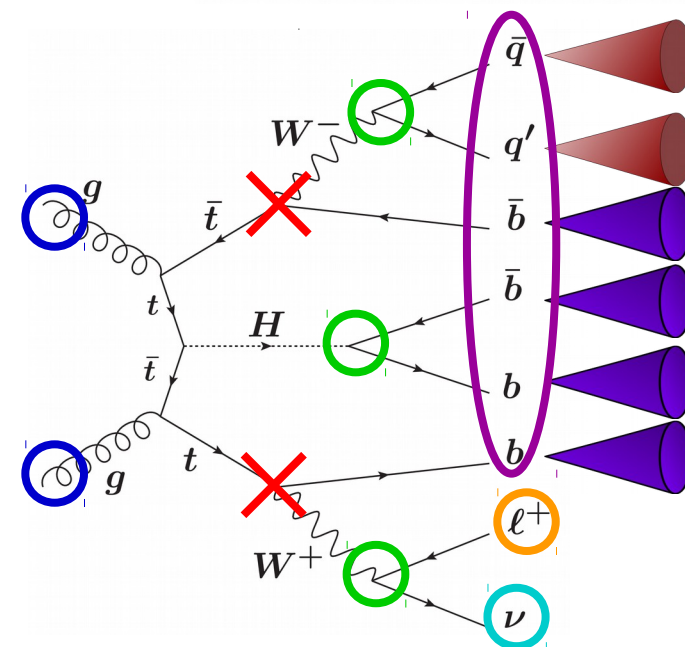
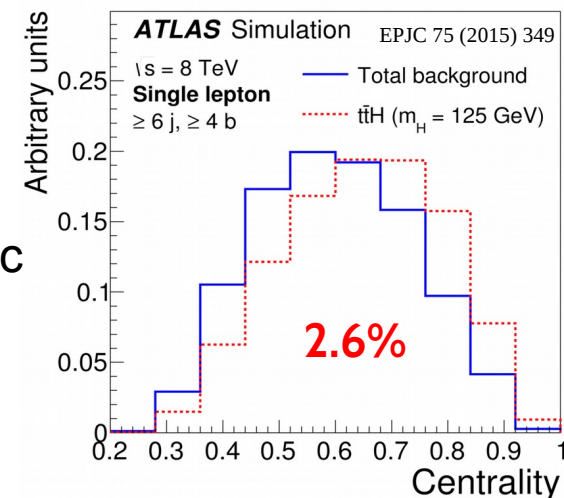
EPJC 75 (2015) 349

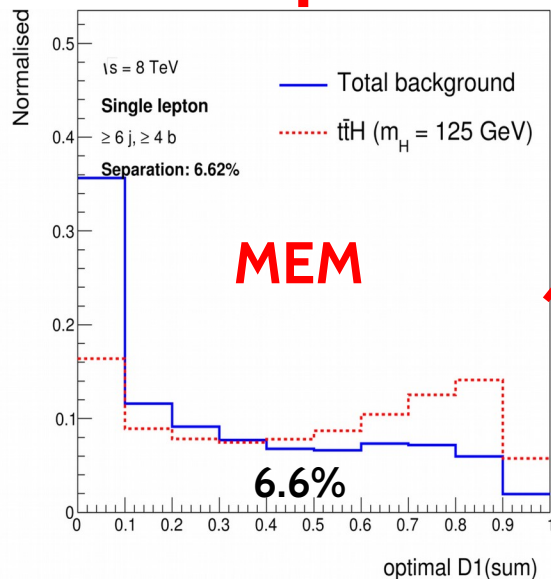
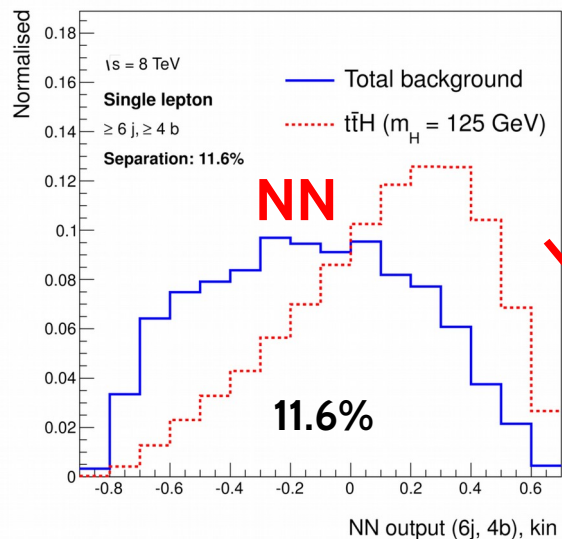
- The ideal world
  - Machine Learning (ML)
    - The underlying physics model can be learned using MC simulation
  - Matrix Element Method (MEM)
    - Describes perfectly the underlying physics of the observed data

→ **A perfect ML & MEM should give you the same answer**
- The real world
  - Machine Learning
    - MC simulations are not perfect, training statistic is limited
    - Input information blurred by resolution & acceptance affects, syst. uncertainties
  - Matrix Element Method
    - Computational constraints demand simplifications & approximation
    - Acceptance & combinatorial ambiguities largely affects input

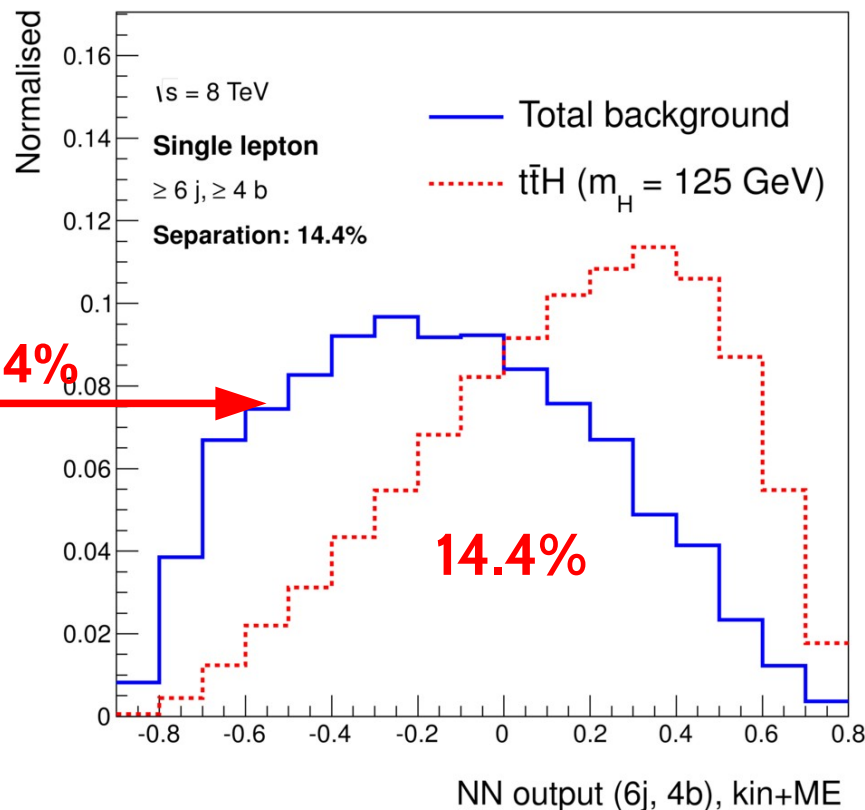
→ **Synergy of both methods still beneficial**

- Machine Learning (Neural Network)
  - Signal & main background ( $t\bar{t}b\bar{b}$ ) kinematically similar
  - Difficult to find good features, selection reduces statistic
  - > 300 input variables tested, saturation  $\sim 10$  variables
- Matrix Element Method
  - Even at LO computationally prohibitive
  - Many simplifications & approximations
    - Reduction of dimensionality (40  $\rightarrow$  6)
    - Optimization of ME calculation
    - Phase space reduction & alignment
    - Reduction of assignment permutations
  - Integration time reduction at precision costs
    - 24h  $\rightarrow$  **2 min per event**





## Most sensitive analysis region (6j, 4b)



Expected limit on signal strength: **3.1**  $\rightarrow$  **2.6 (16%)**



- Both MEM & ML output contains **valuable & unique information**
  - Loss of information is unavoidable in both methods
- Combination of both methods natural & most complete answer
  - **16% improvement** corresponds to a large amount of data ( $>10 \text{ fb}^{-1}@8\text{TeV} = \$\$$ )
- Disadvantage of combination ML+MEM
  - Need to run MEM on large MC statistic (6M nominal) required for training
  - Tiny signal  $\rightarrow$  many systematic variations of background to consider

$\rightarrow$  **62M total** number of events  $\rightarrow$  **2M CPUh (236 CPUy)** in 2 month real time
- Overcoming computational constraints
  - Further simplify model or dedicated MEM region (loss of information)
  - Highly parallelize CPU computation (challenging bookkeeping & inflexible)
  - Use GPUs instead of CPUs (limited by available GPUs)
  - Use of look-up tables (non-trivial interpolation and parametrization of PS)