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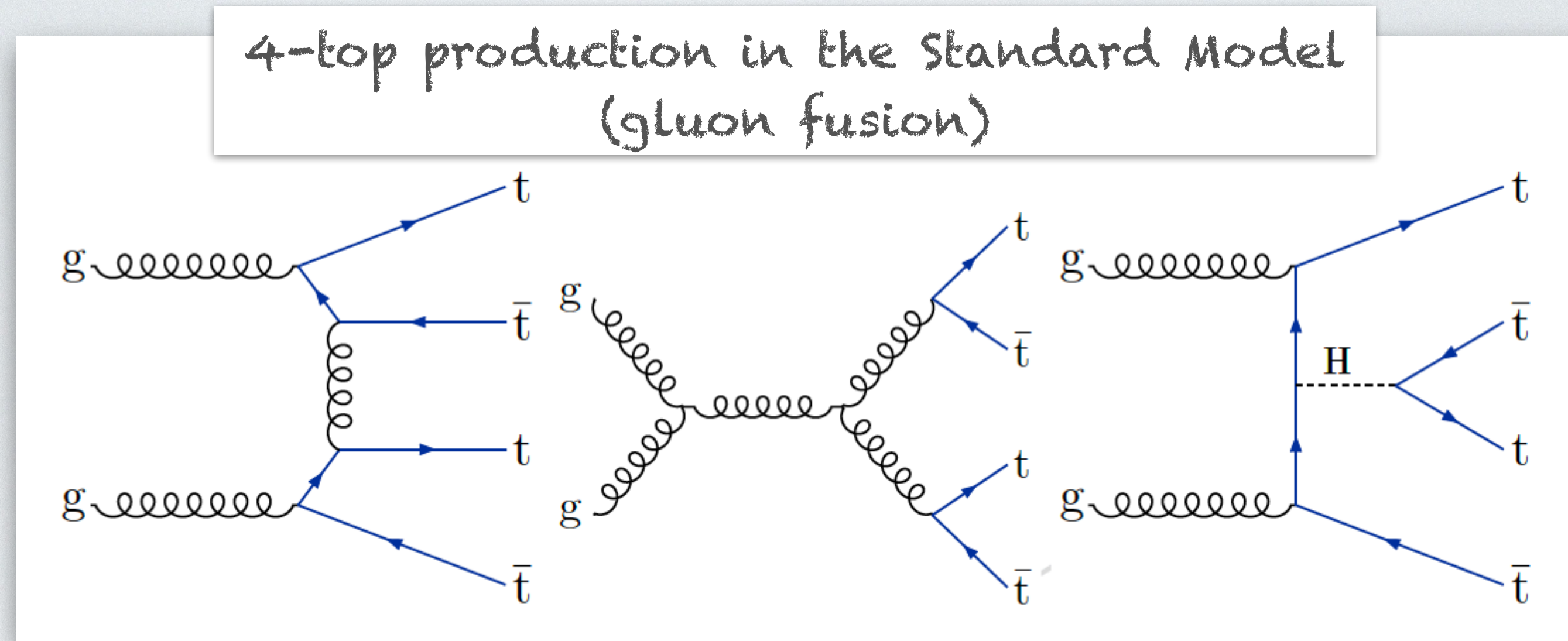
Fully-Hadronic Search for Standard Model Four-Top Production at CMS

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Four-Top Physics



- Cross-section is ~ 12 fb in the Standard Model
- Enhancement of cross section predicted by some BSM physics
- Promising probe of top-quark Yukawa coupling

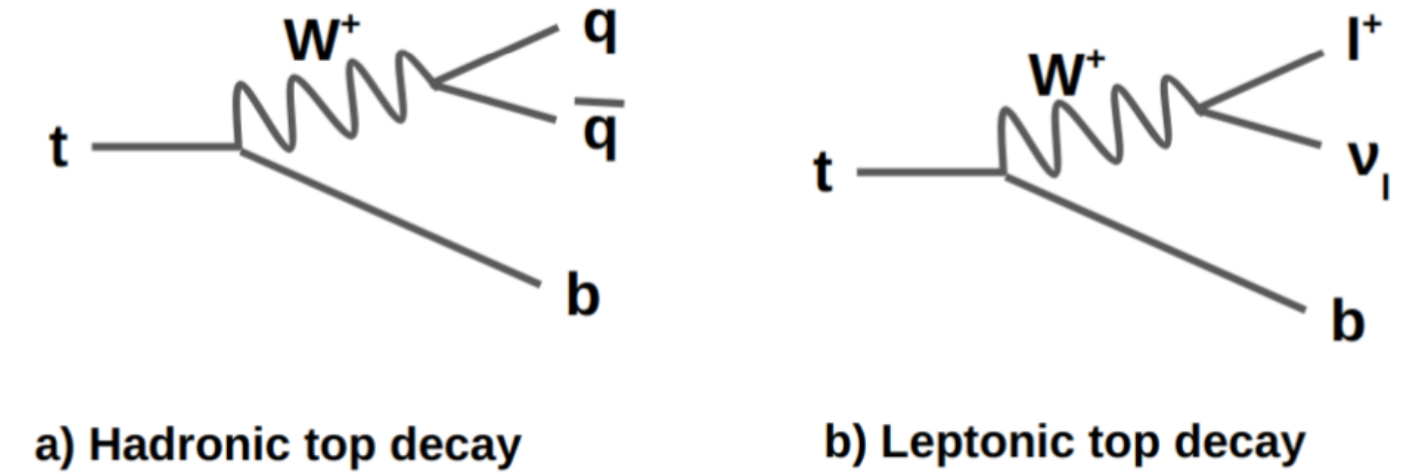
4-top Channels

- **All-Hadronic Channel**
 - Major background(s): QCD + hadronic TT
 - Same-Sign DiLepton + MultiLepton Channels
 - Run II result published ([arXiv:1908.06463](https://arxiv.org/abs/1908.06463))
 - Observed (expected) significance: **2.6(2.7) σ**
 - Single-Lepton + Opposite-Sign DiLepton Channels
 - Run II analysis ongoing
-
- ATLAS Run II Result submitted ([arXiv:2007.14858](https://arxiv.org/abs/2007.14858))
 - Same-Sign DiLepton + MultiLepton Channels ONLY
 - Observed (expected) significance: **4.3(2.4) σ**

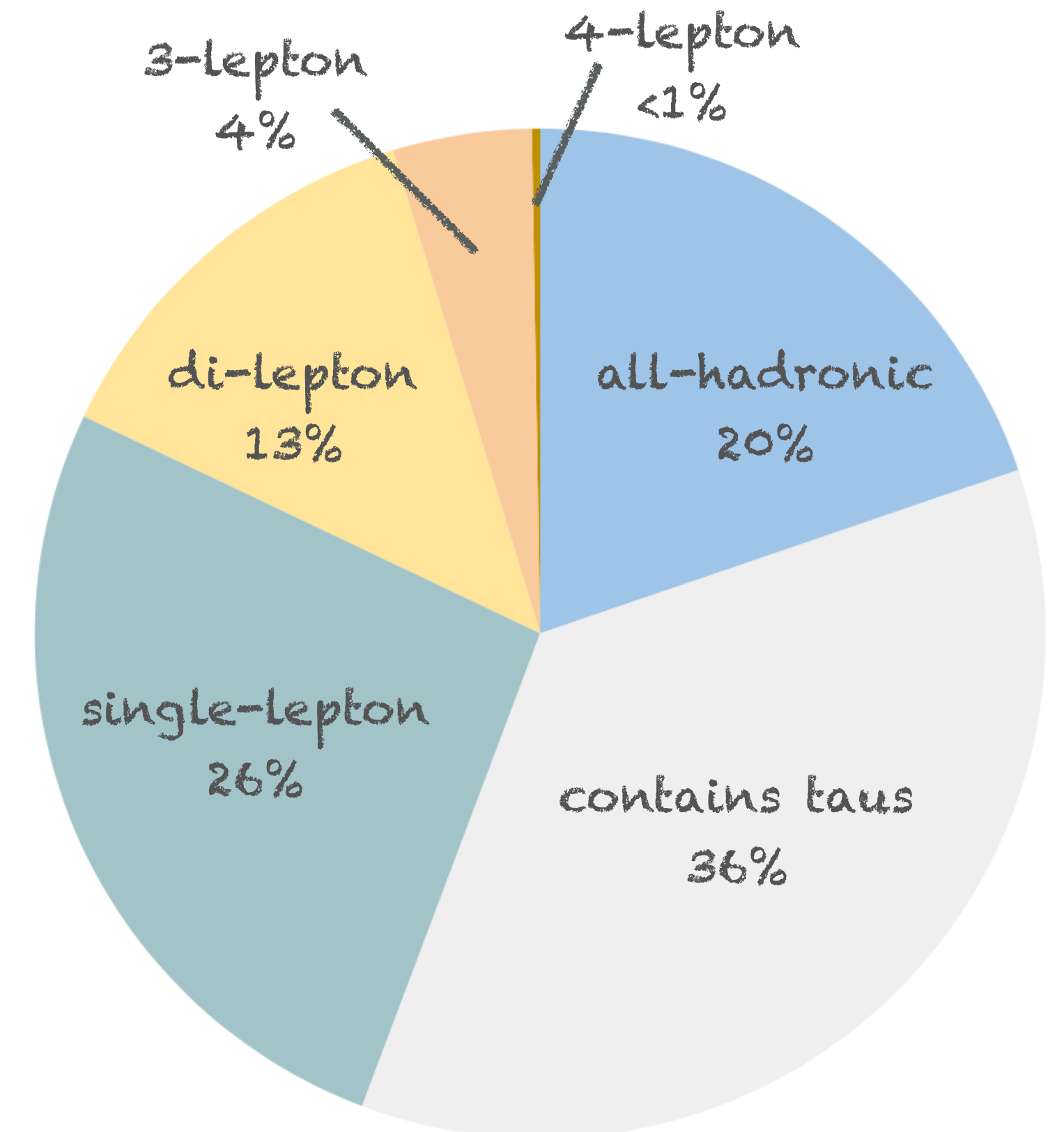
Combined
CMS Run II
Result

ATLAS
Run II
Result

Top Decays

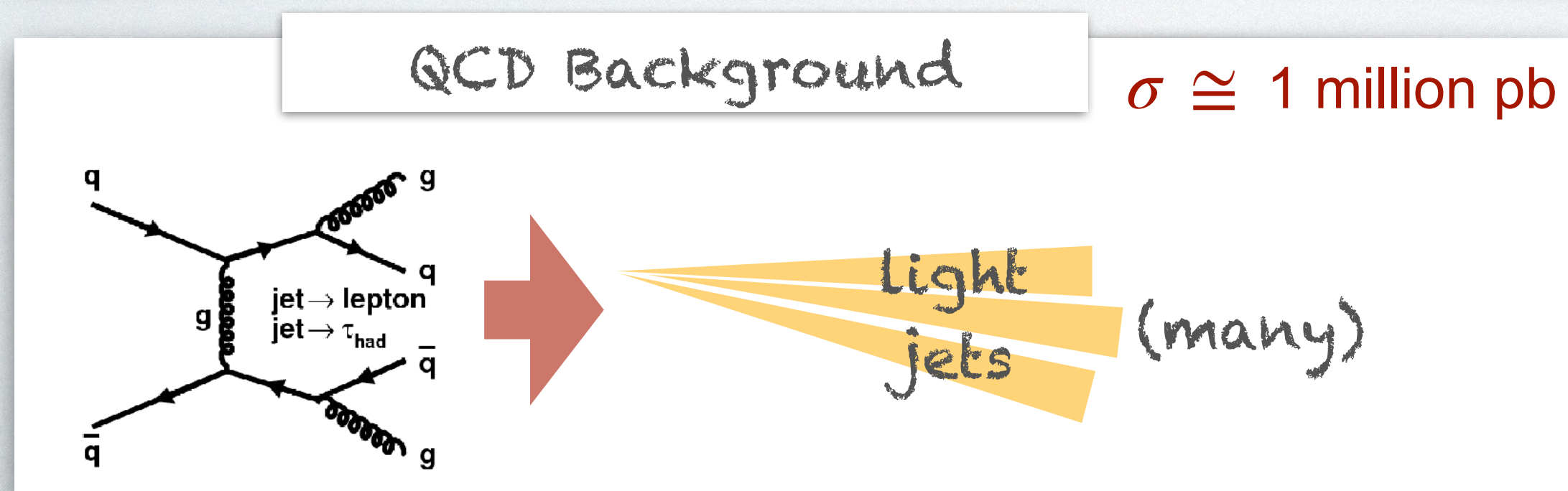
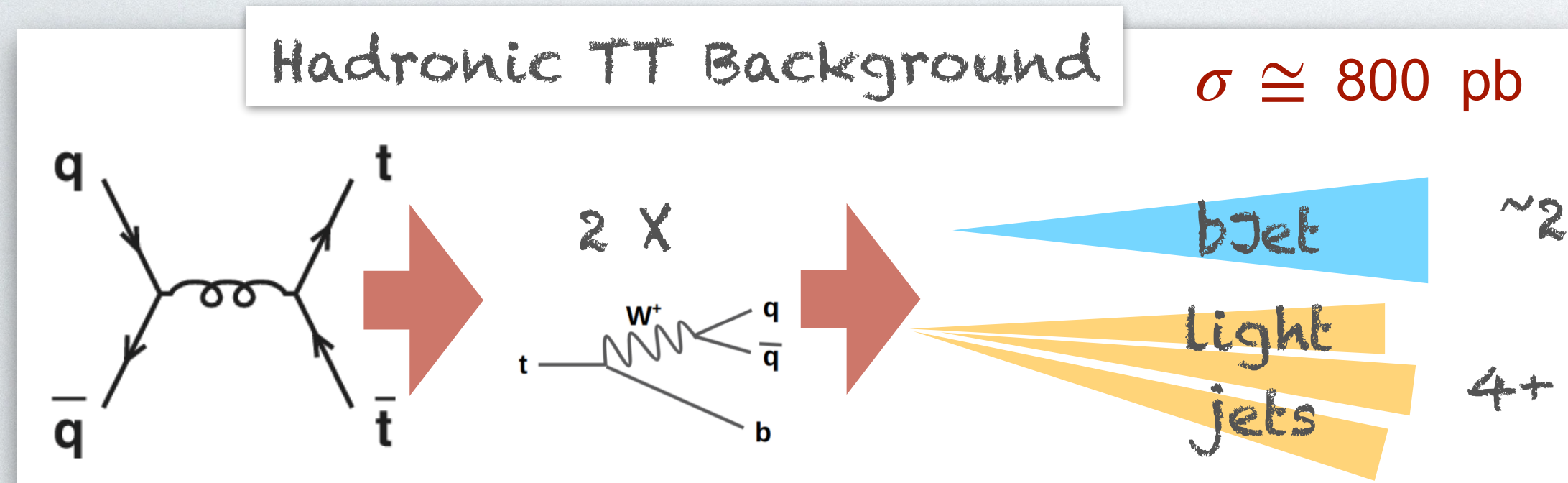
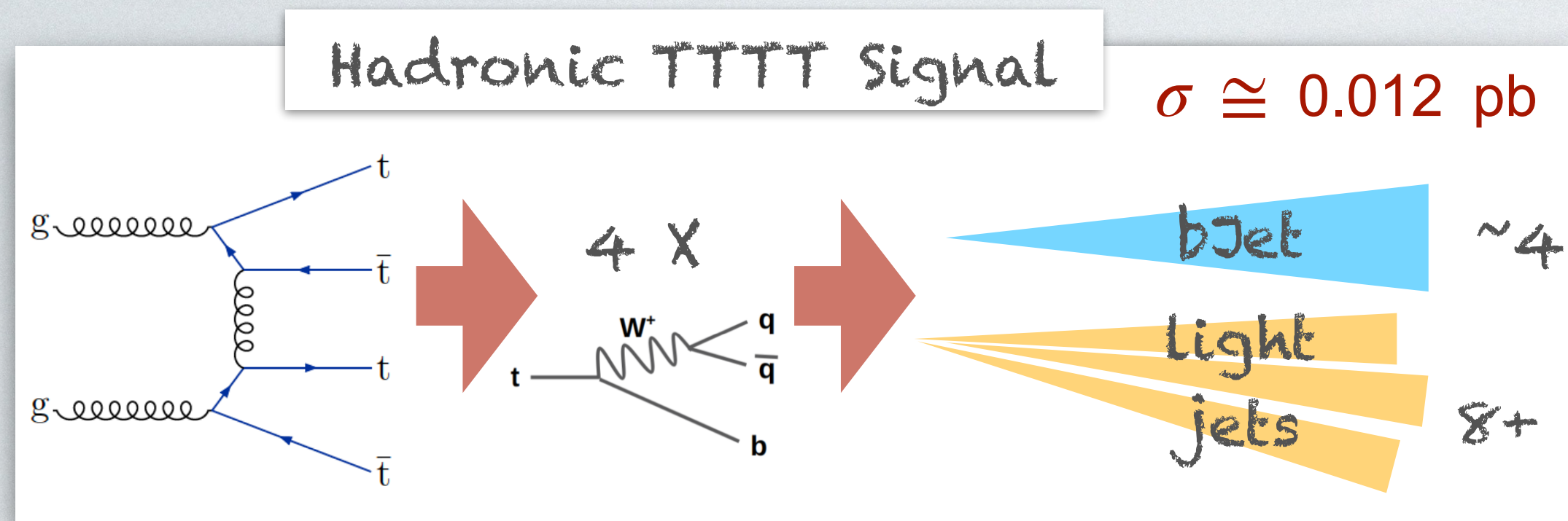


4-top Branching Fractions



Analysis Strategy

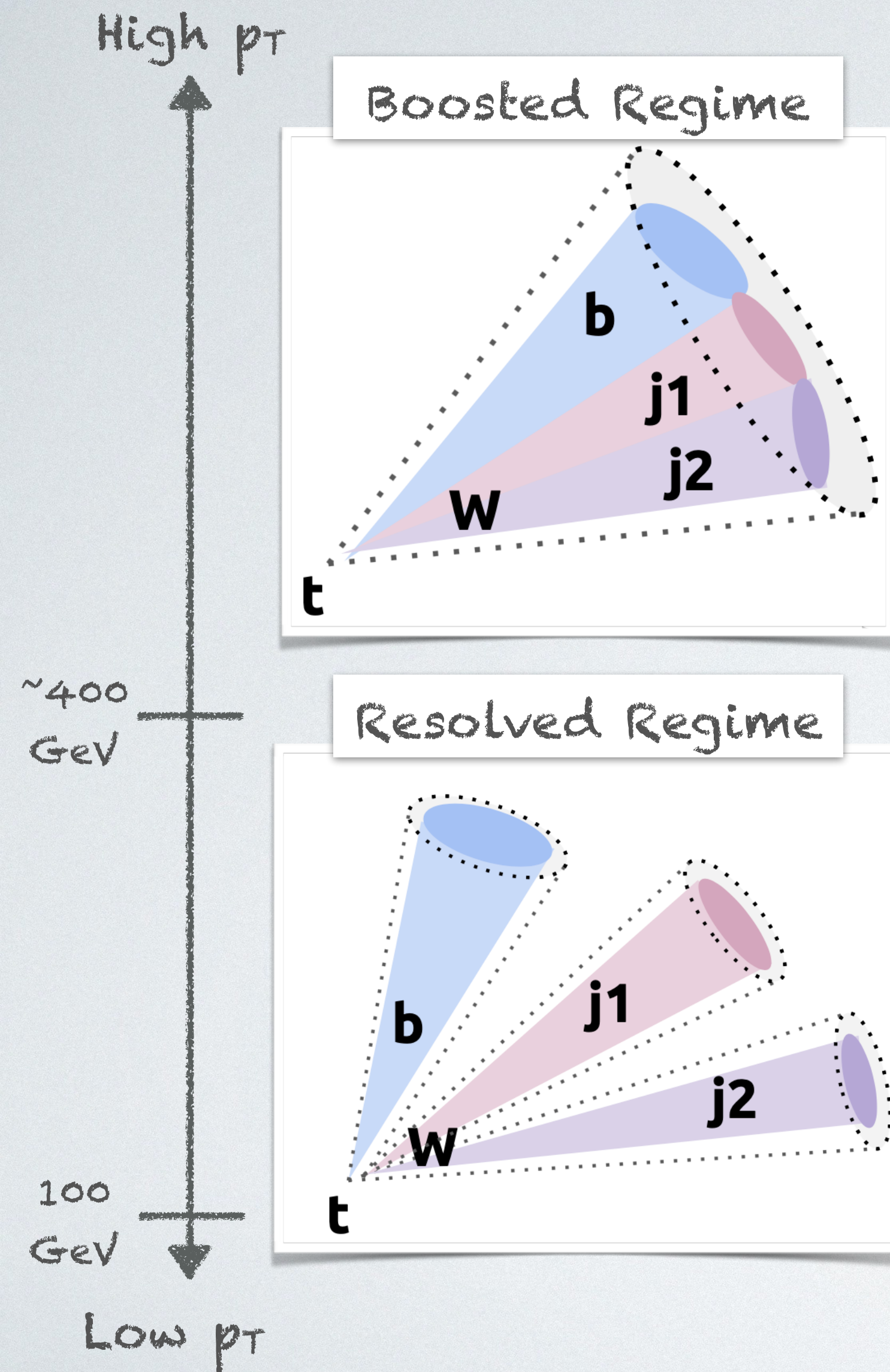
- In order to address QCD and TT backgrounds we need some special techniques...



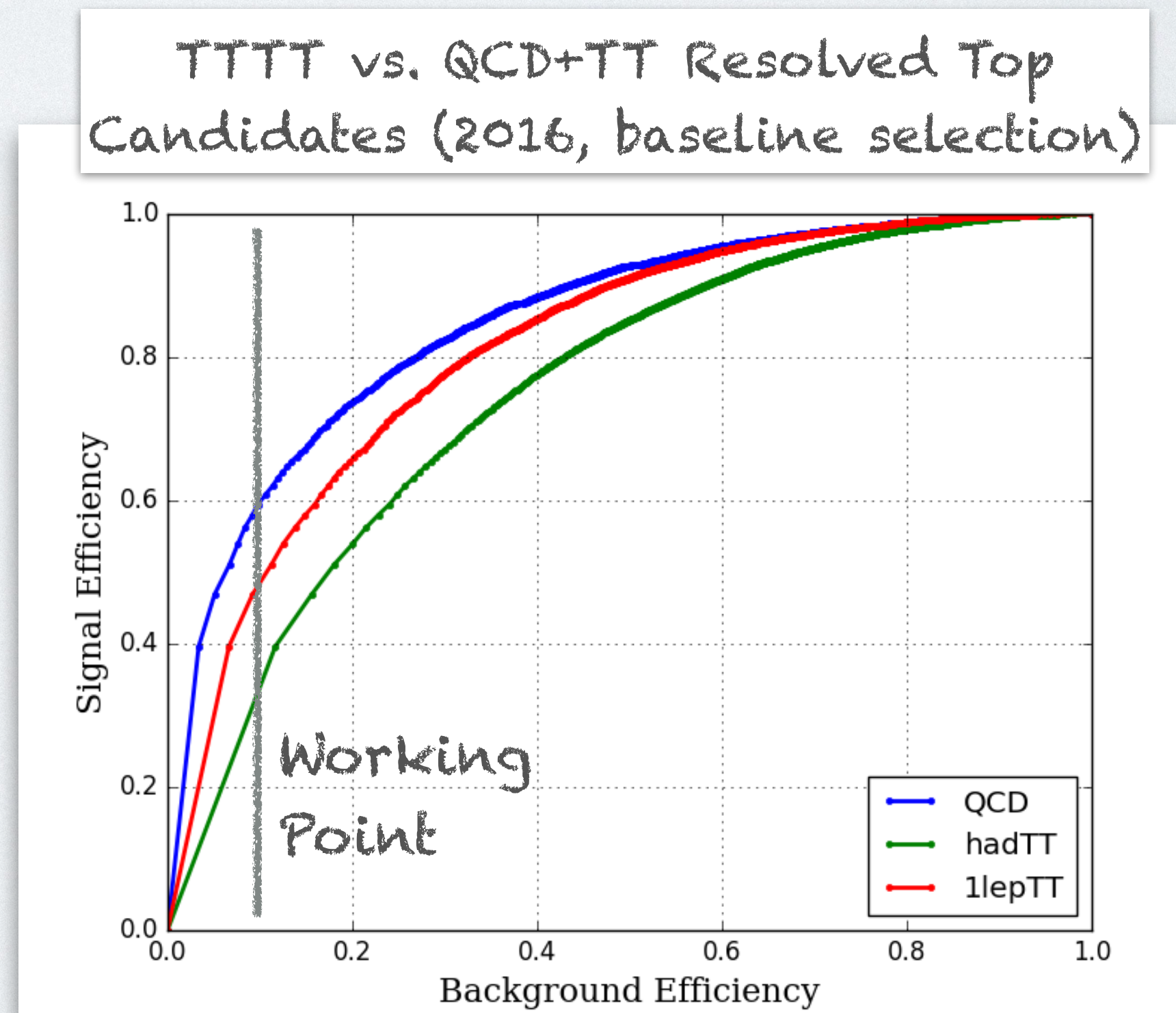
BASELINE:
 $N(\text{Jets} + \text{bJets}) \geq 9$, $N(\text{bJets}) \geq 3$, $HT \geq 700 \text{ GeV}$
 No leptons

- Overall Strategy**
- Start with data and apply **baseline**.
 - Tag **resolved and boosted tops**.
 - Evaluate **BDT** to discriminate 4-top signal vs. TT+QCD backgrounds.
 - Bin signal region in numbers of resolved and boosted tops. Weight bins by the **BDT shape**.
 - Estimate TT & QCD backgrounds in signal region from data in **control regions**.

Top Tagging



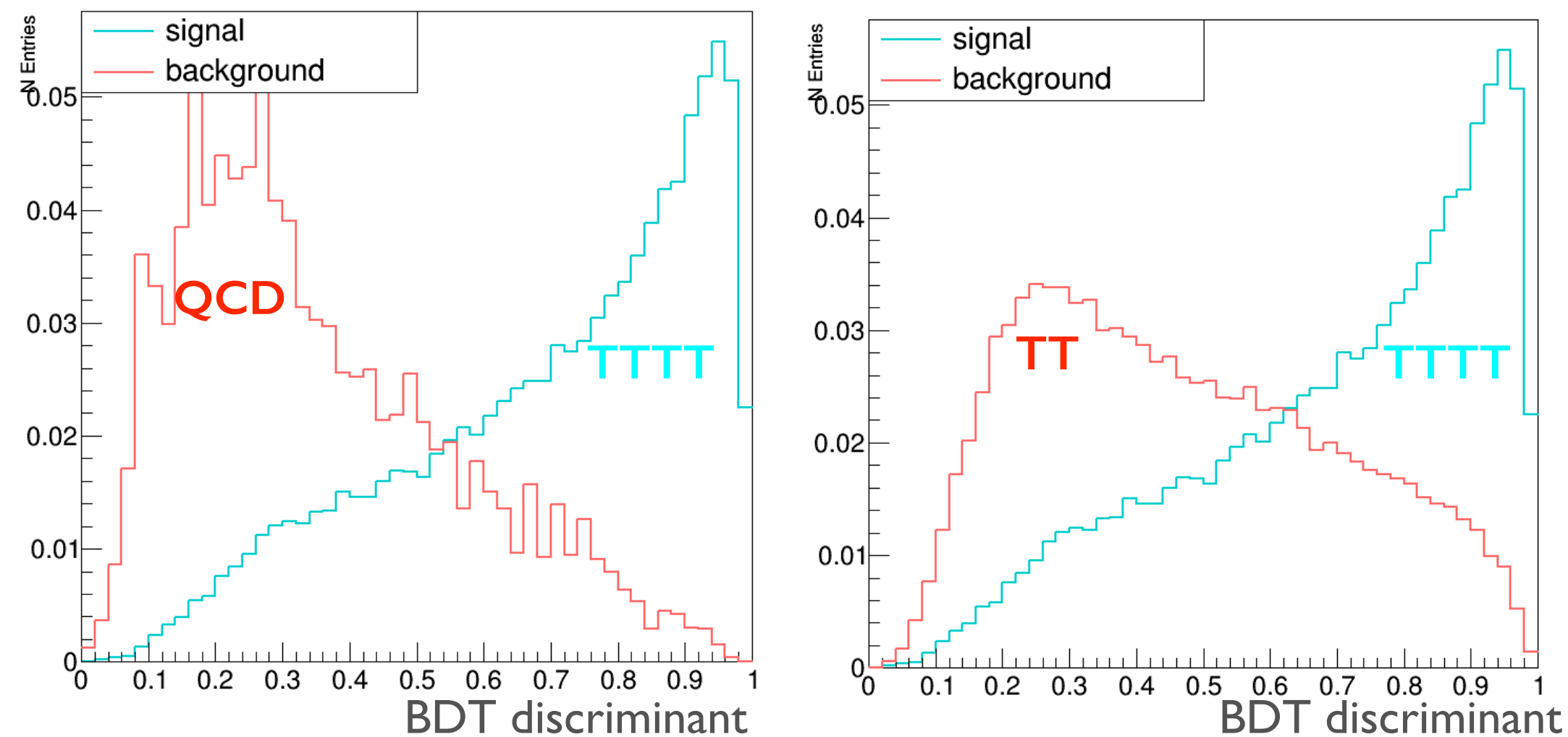
- Boosted top tagger: **DeepAK8** (CMS-JME-18-002)
 - Deep Neural Net (DNN) that inputs low level jet substructure and flavor information.
- Resolved top tagger: **Custom BDT built for this search**
 - Inputs Neural Net (NN) based b and c tagging information and jet kinematics



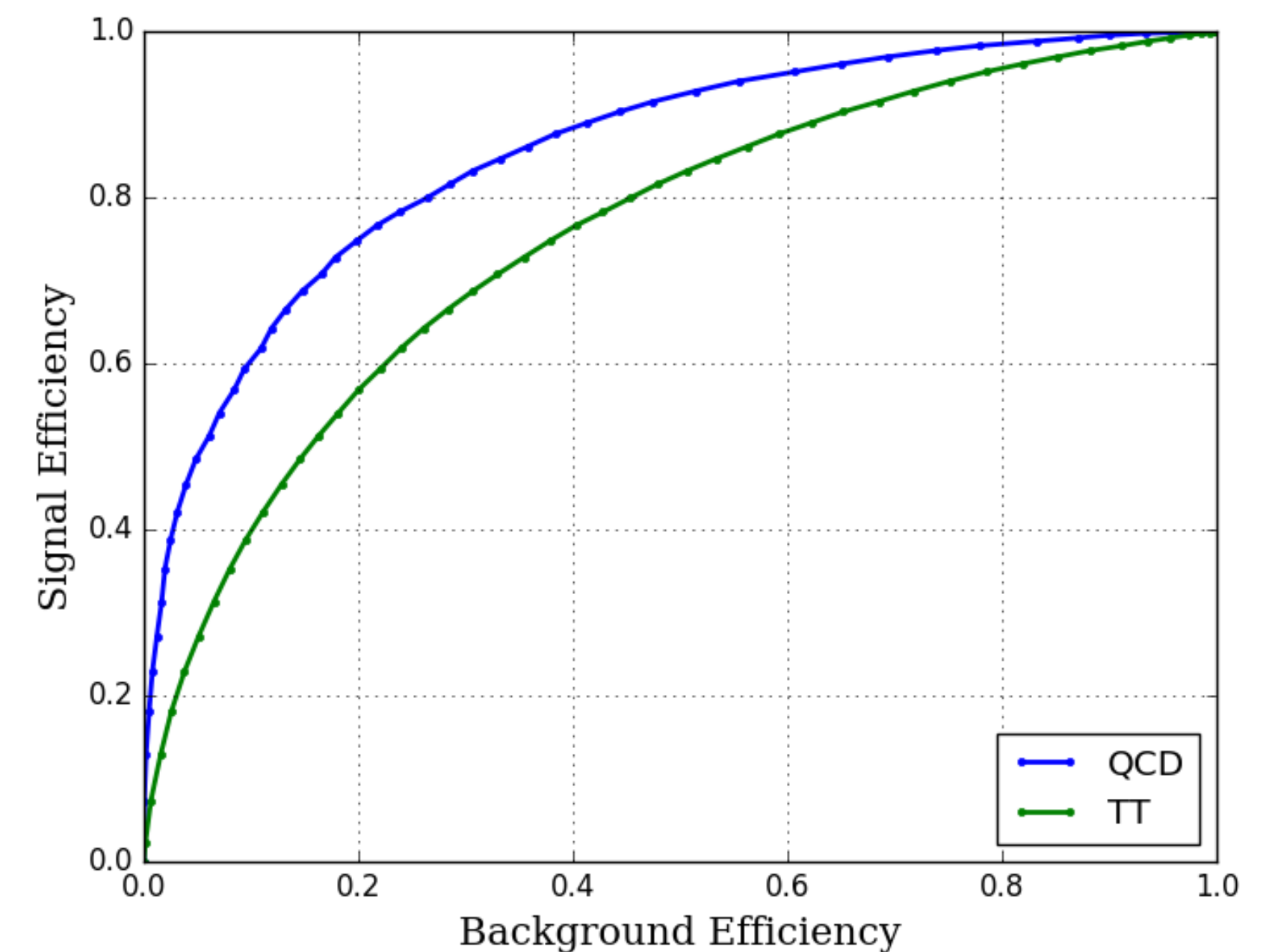
BDT Targeting QCD & TT Backgrounds

- Event-level BDT trained to discriminate between **TTTT signal** and **TT+QCD backgrounds**
- Inputs: jet and resolved top kinematics & angular variables
- Discriminant shape used to weight signal region bins

BDT Discriminant Shapes (normalized)



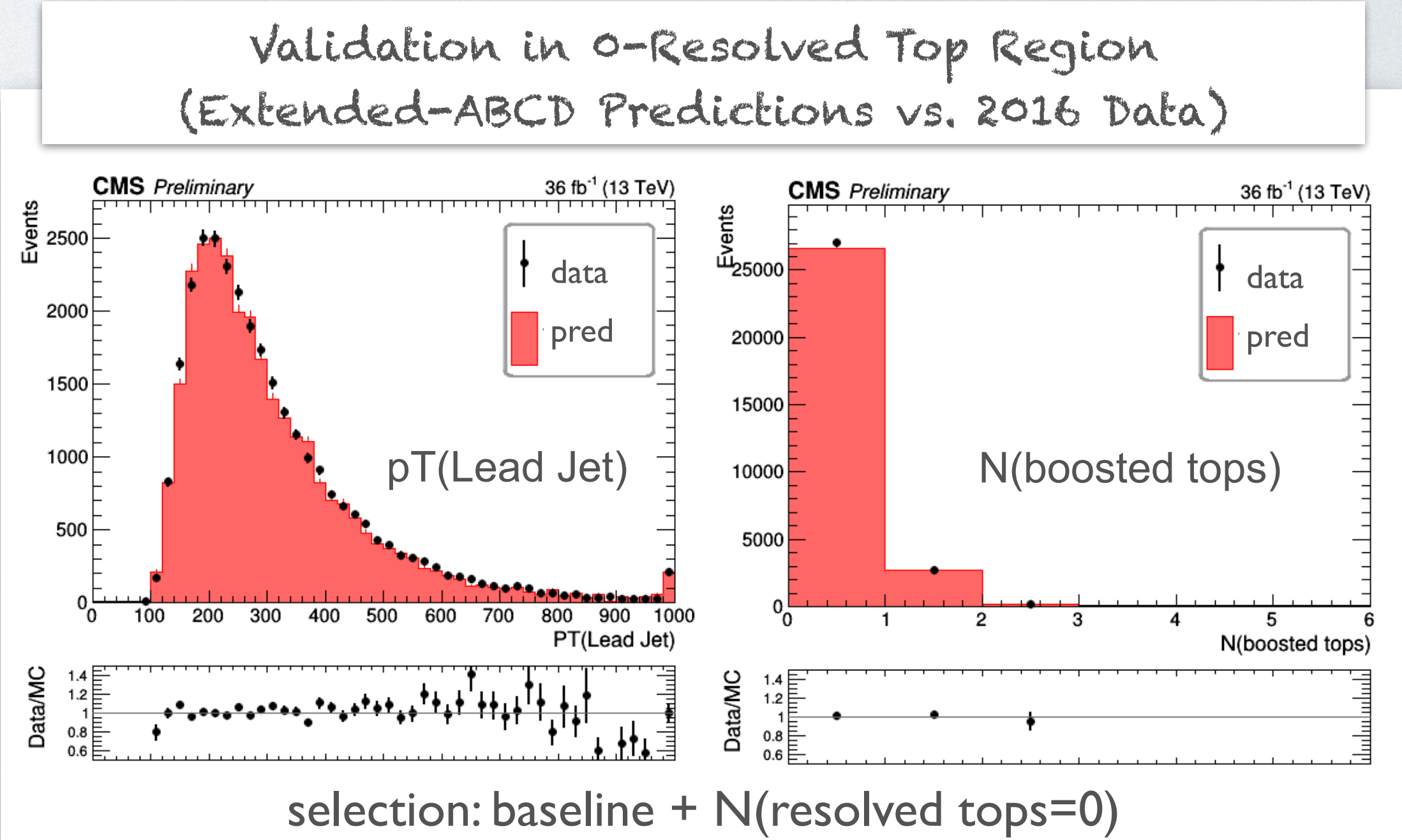
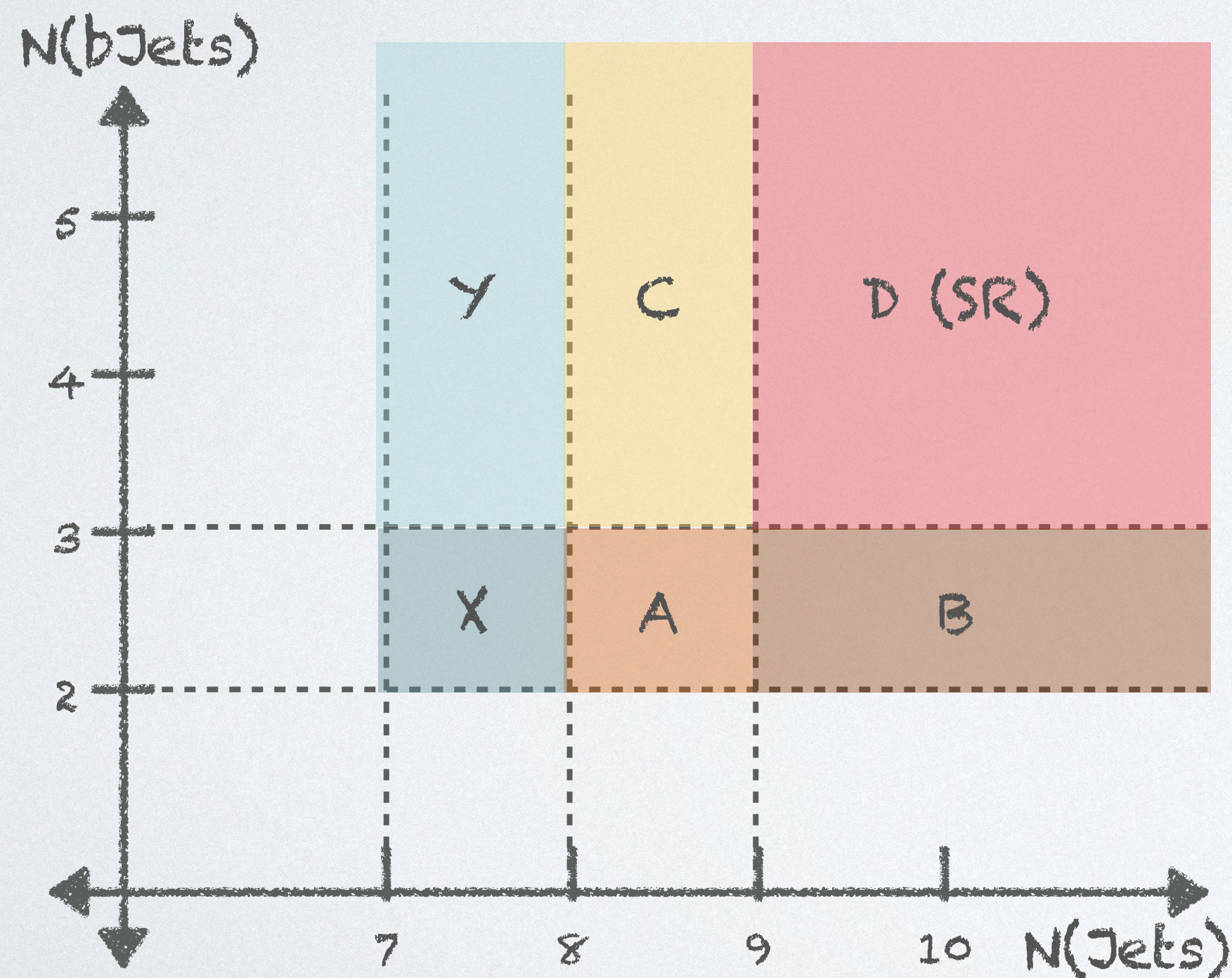
TTTT vs. TT & QCD
(baseline selection)



Data-Driven Background Estimates

- Extended ABCD method: predict QCD+TT background in signal region (SR) from data in control regions (CRs) ([arXiv:1906.10831](https://arxiv.org/abs/1906.10831))

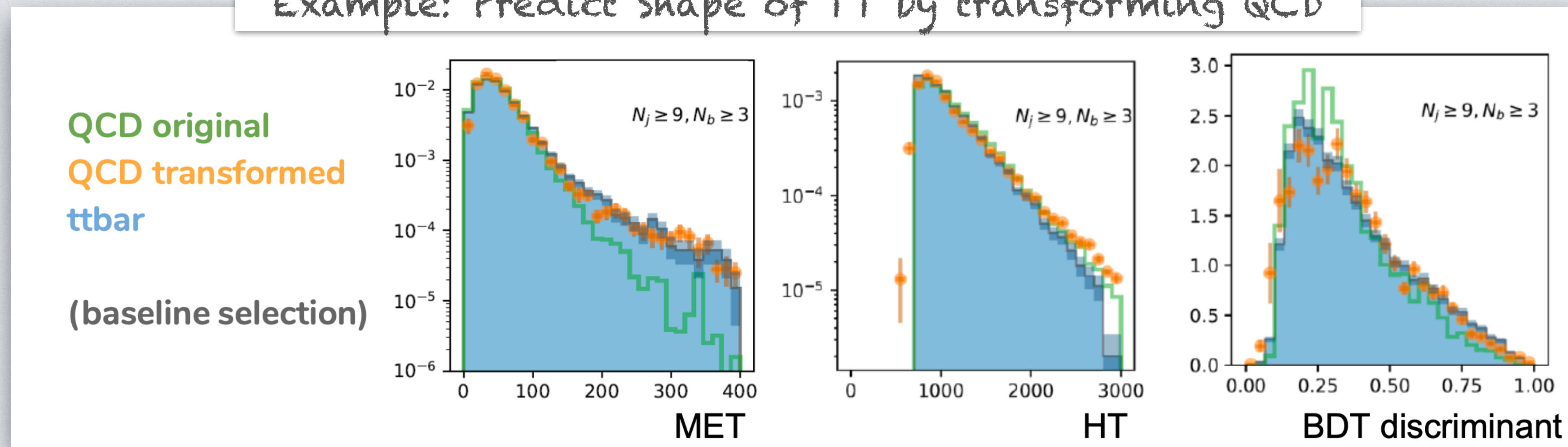
$$D \text{ (SR)} \approx \left(\frac{B \quad C}{A} \right) * \left(\frac{B \quad C}{A} \right) \left(\frac{B \quad y}{X} \right)^{-1}$$



Predicting BDT Shape from Data with Normalizing Flows

- **Problem:** Extended-ABCD method can fail if variables are correlated, distribution shape changes too fast, or low statistics in one phase space
 - BDT discriminant shape is a case of this!

Example: Predict Shape of TT by transforming QCD



- **Solution: Normalizing Flows** ([arXiv:1804.00779](https://arxiv.org/abs/1804.00779))
- Adversarial neural net that learns transformation from one distribution to another
 - Predict BDT shape in SR by learning MC \rightarrow data shape differences in CR

$$\mathcal{T}(\vec{x} | \vec{x}_0; \vec{c}) f_{MC}(\vec{x}_0; \vec{c}) = f_{Data}(\vec{x}; \vec{c})$$

doesn't learn physics
("f"), only "T"

Expected Sensitivity

SR Top Bins

N(boosted tops) (N_{BT})	N(resolved tops) (N_{RT})		
		=1	≥ 2
	=0	$N_{RT} = 1$ $N_{BT} = 0$	$N_{RT} \geq 2$ $N_{BT} = 0$
	≥ 1	$N_{RT} = 1$ $N_{BT} \geq 1$	$N_{RT} \geq 2$ $N_{BT} \geq 1$

Can further split bins by HT, N(Jets), or other kinematic variables

Channel	Expected Run II Limit \times SM σ (139 fb ⁻¹)
All-Hadronic (preliminary)	~3.0
Single-Lepton *scaled from <u>2016</u> result	~3.7
Opposite-Sign DiLepton *scaled from <u>2016</u> result	~2.9
Same-Sign Di+Multi Lepton (<u>run II</u> result)	1.6

*re-optimized full Run 2 result should improve on this scaling by luminosity

- Split SR bins by number of boosted and resolved tops and weight by shape of BDT
- Sensitivity estimate based on MC study, considering statistical and rough estimate of systematic uncertainties.
- Adds sensitivity when combined with other channels

Outlook

- Multifaceted and novel approach to a challenging channel
- Contributing to combined CMS result by end of the year
 - **Very close to “evidence” of 4-top production in CMS!**
- BSM interpretations also of great interest

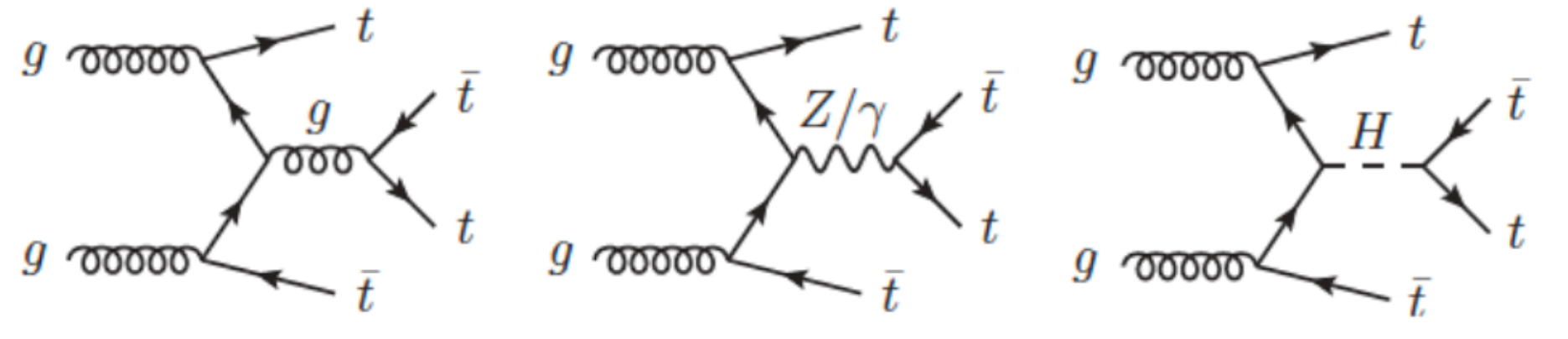




BACKUP

4-top Cross Section Phenomenology

$$\sigma(t\bar{t}t\bar{t}) = \sigma^{\text{SM}}(t\bar{t}t\bar{t})_{g+Z/\gamma} + \kappa_t^2 \sigma_{\text{int}}^{\text{SM}} + \kappa_t^4 \sigma^{\text{SM}}(t\bar{t}t\bar{t})_H$$



$$\begin{aligned} \sigma^{\text{SM}}(t\bar{t}t\bar{t})_{g+Z/\gamma} &\propto |\mathcal{M}_g + \mathcal{M}_{Z/\gamma}|^2, \\ \sigma^{\text{SM}}(t\bar{t}t\bar{t})_H &\propto |\mathcal{M}_H|^2, \\ \sigma^{\text{SM}}(t\bar{t}t\bar{t})_{\text{int}} &\propto \mathcal{M}_{g+Z/\gamma} \mathcal{M}_H^\dagger + \mathcal{M}_{g+Z/\gamma}^\dagger \mathcal{M}_H \end{aligned}$$

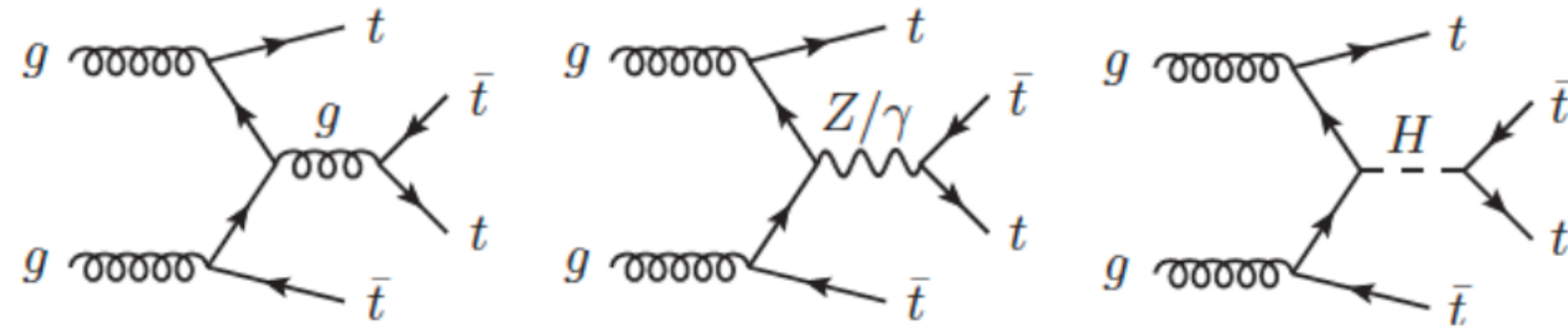
- $\kappa_t = y_T/y^{\text{SM}}_T$ where y_T is the top Yukawa coupling, and y^{SM}_T is its value in the SM.
- Strong, electroweak, and higgs terms contribute:
 - σ_{int} is interference between strong+EW and H terms
 - Higgs term proportional to yukawa coupling to 4th power:

$$\sigma(t\bar{t}t\bar{t})_H \propto \kappa_t^4 \sigma^{\text{SM}}(t\bar{t}t\bar{t})_H$$

- Sensitive to top yukawa coupling, but independent of Higgs decay
 - Deviations from standard model predictions ($\kappa_t \neq 1$) would be red flag for BSM physics
- More information: [arXiv:1602.01934](https://arxiv.org/abs/1602.01934) and [arXiv:1908.06463](https://arxiv.org/abs/1908.06463), Giovanni Zevi Della Porta's presentation

4-top Cross Section Phenomenology

$$\sigma(t\bar{t}t\bar{t}) = \sigma^{\text{SM}}(t\bar{t}t\bar{t})_{g+Z/\gamma} + \kappa_t^2 \sigma_{\text{int}}^{\text{SM}} + \kappa_t^4 \sigma^{\text{SM}}(t\bar{t}t\bar{t})_H$$



$$\begin{aligned}\sigma^{\text{SM}}(t\bar{t}t\bar{t})_{g+Z/\gamma} &\propto |\mathcal{M}_g + \mathcal{M}_{Z/\gamma}|^2, \\ \sigma^{\text{SM}}(t\bar{t}t\bar{t})_H &\propto |\mathcal{M}_H|^2, \\ \sigma^{\text{SM}}(t\bar{t}t\bar{t})_{\text{int}} &\propto \mathcal{M}_{g+Z/\gamma} \mathcal{M}_H^\dagger + \mathcal{M}_{g+Z/\gamma}^\dagger \mathcal{M}_H\end{aligned}$$

Leading Order Cross
sections @ 13 TeV

$$\sigma^{\text{SM}}(t\bar{t}t\bar{t})_{g+Z/\gamma}: 9.997 \text{ fb},$$

$$\sigma^{\text{SM}}(t\bar{t}t\bar{t})_H: 1.168 \text{ fb},$$

$$\sigma^{\text{SM}}(t\bar{t}t\bar{t})_{\text{int}}: -1.547 \text{ fb},$$

$$\text{Total}: 9.6 \text{ fb}^*$$

*Note: cross section on slide 2 includes NLO calculations, which is why it doesn't match the "Total" value. Detailed info here: [arXiv:1711.02116](https://arxiv.org/abs/1711.02116)

$$\text{LO strong} + \text{LO weak} + \text{LO Higgs} = 9.6 \text{ fb}$$

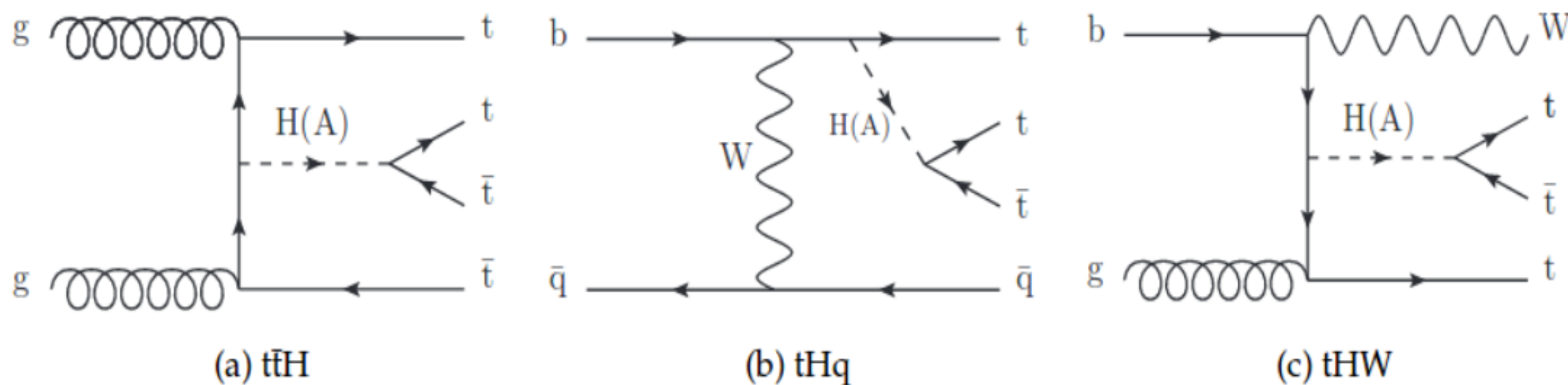
$$\text{LO+NLO strong} + \text{LO weak} + \text{LO Higgs} = 9.2 \text{ fb}$$

$$\text{LO+NLO strong} + \text{LO+NLO weak} + \text{LO Higgs} = 12.0 \text{ fb}$$

2-Higgs-Doublet-Models (2HDMs)

- Simple extension of the Standard Model (SM) with 2 doublets of Higgs-like bosons rather than one
- A heavy scalar or psuedo-scalar boson decaying to tops is produced in association with 1 or 2 tops
- Signal is 4 top quarks (ttH/A) or 3 top quarks (tHq , tHW)
- Leads to enhanced $tttt$ cross section signature
- Key ingredient in minimal supersymmetric standard model (MSSM)
- More information: [arXiv:1106.0034](https://arxiv.org/abs/1106.0034)

Top-associated Heavy Scalar Production Modes



'The' Higgs?

h

H

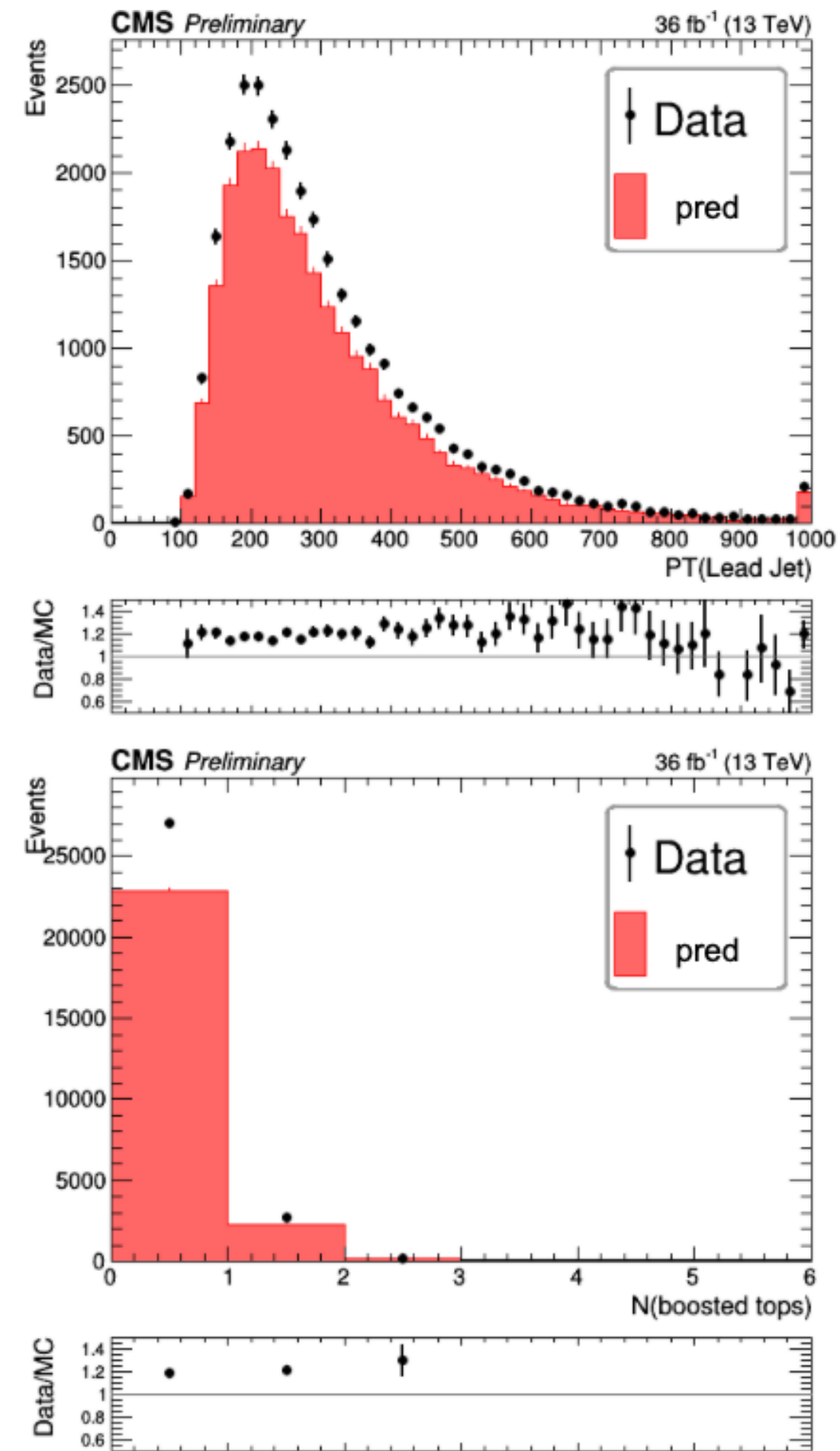
A

H^\pm

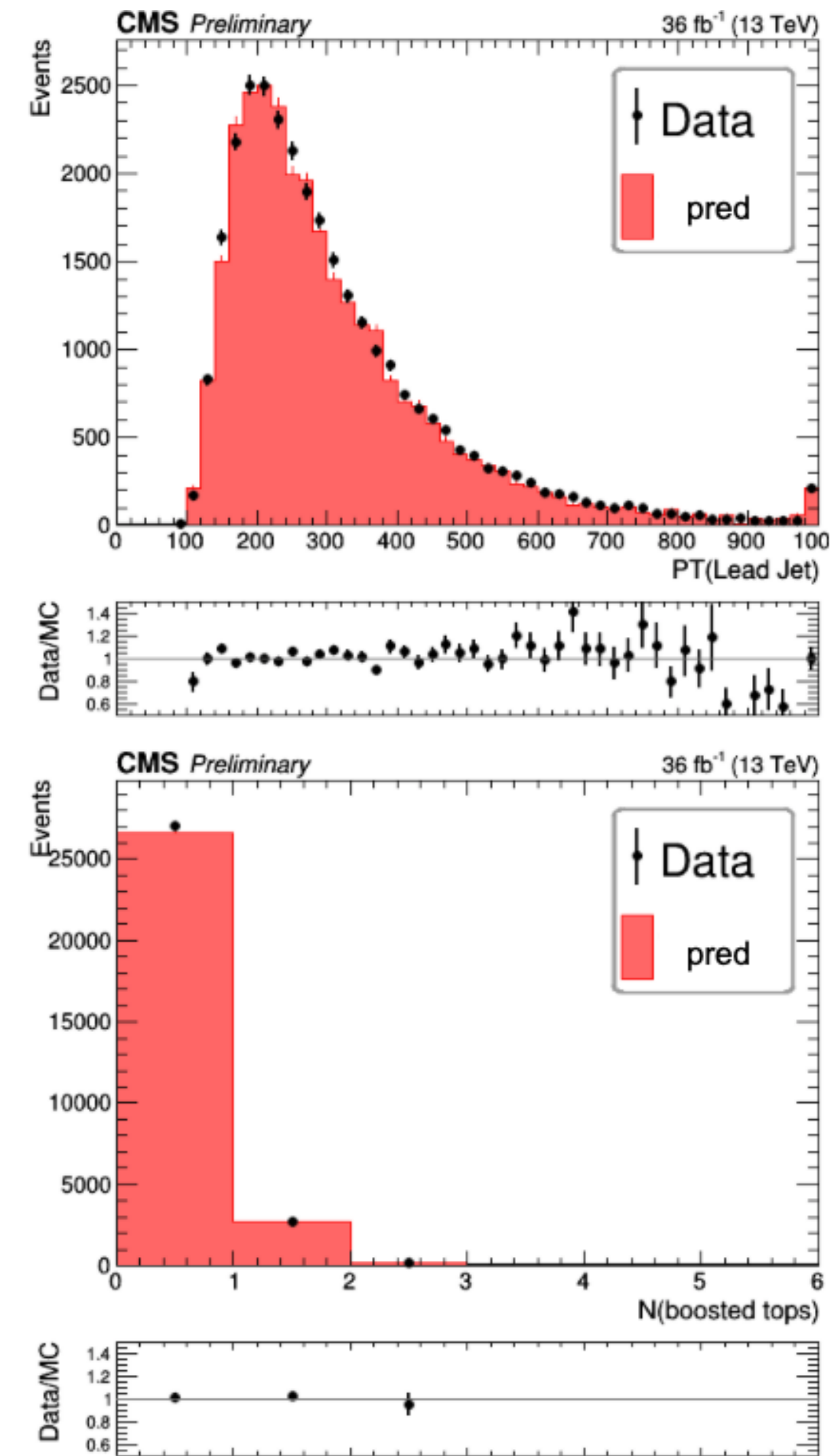
4 Higgs: One doublet of neutral bosons (including our 125 GeV Higgs) and 1 doublet of a psuedoscalar and a heavy charged Higgs

ABCD vs. Extended ABCD

ABCD method: $D \approx BC/A$



ext-ABCD method: $D \approx (BC/A)^2(A'/BC')$



Normalizing Flows

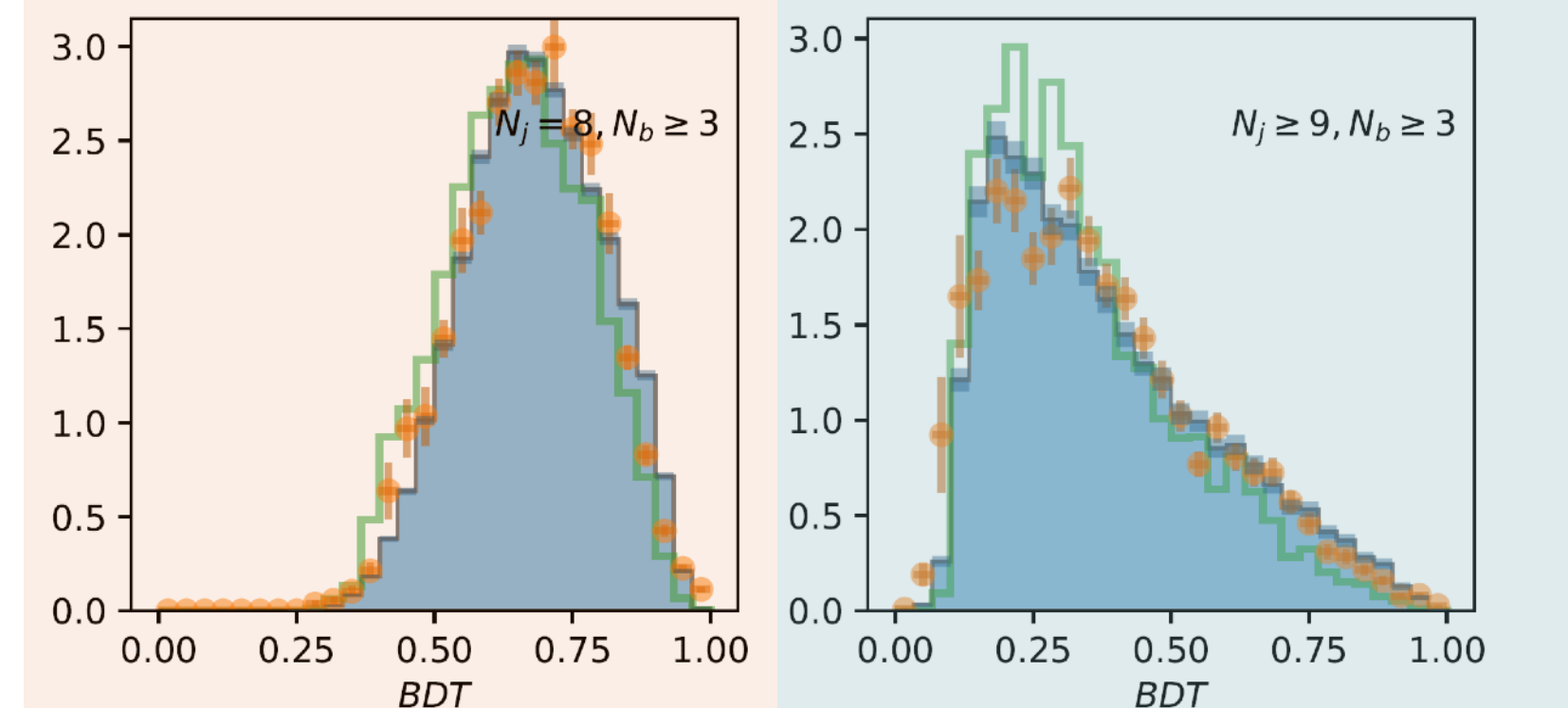
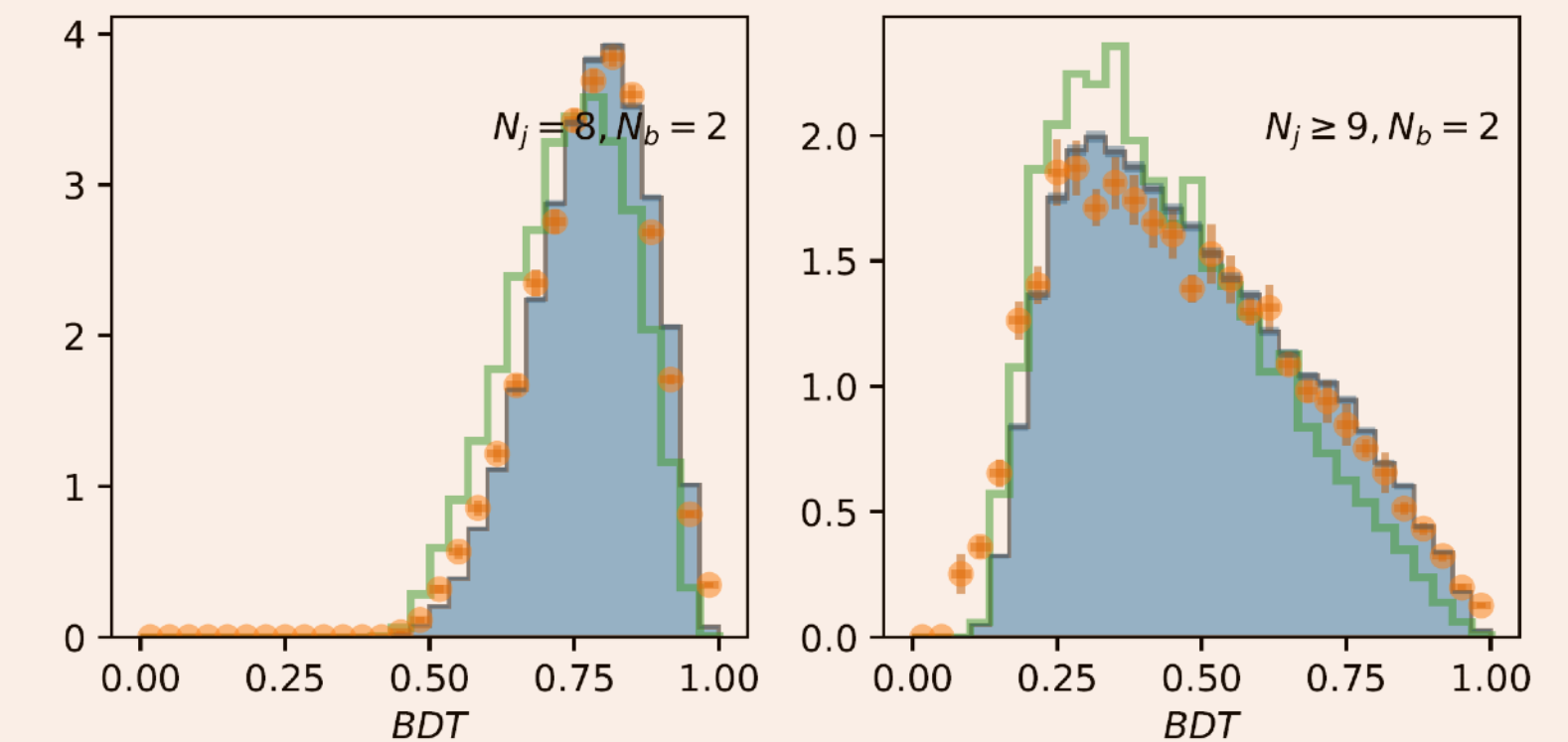
- More Details (arXiv:2008.03636v1)
- Through normalizing flows (NF), feature variables are transformed through multidimensional invertible bijections.
 - This transformation is implemented as a NN and learned during training.
 - We use the neural autoregressive flow (NAF), to construct a multidimensional invertible function
- More appealing and acceptable than GAN, since only transformation learned and it doesn't try to reproduce the underlying physics.
- Can predict different variables simultaneously

Example: Predict Shape of TT by transforming QCD

QCD original
QCD transformed
ttbar

(baseline selection)

TRAINING



PREDICTION