



UNIVERSITÀ DEGLI STUDI DI NAPOLI  
FEDERICO II

# ANALYSIS STATUS: DATASETS AND TRIGGER SELECTION

Workshop Roma-Napoli, Napoli, 18/12/2024

# Run 3 ATLAS dataset

- ATLAS data is collected in AOD (Analys Object Data), particularly in skimmed AOD (Derived AOD → DAOD)
  - DAOD\_PHYS is the unified data format, but it does not contain constituents information...

## ➤ Our new requested DAOD\_LLJ1 derivation:

- DAOD\_PHYS format as baseline, with addition of constituents for jets

## ➤ UFO collection for large-R jets, upgrade to previous TrackTopoClusters

- Possibility to add new features for ML with respect to LHC Olympics datasets
- Constituents taste, associated tracks information

## ➤ Trigger and event skimming applied

- Due to significant size

## ➤ Data

- 2022 and 2023 production;

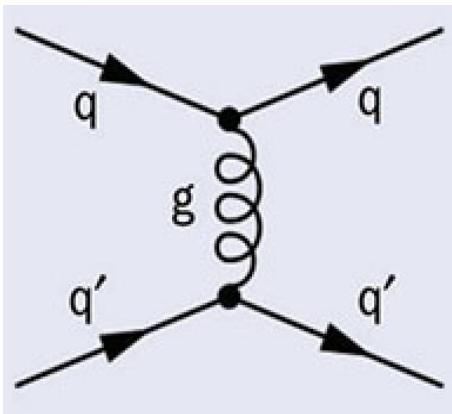
```
### 1 jet object skimming
sel_1jet_template = "((count (abs({0}eta) < 2.8 && {0}pt > 150*GeV && {0}m > 30*GeV) >= 1))"
topology_selection_1jet = "{0}").format(
    " || ".join([sel_1jet_template.format(j) for j in largeRJetsForSkimming])
)

### trigger skimming
TriggersList = [
    ### baseline run-2
    'HLT_j360_a10_lcw_sub_L1J100',
    'HLT_j420_a10_lcw_L1J100',
    'HLT_j460_a10t_lcw_jes_L1J100',
    ### new run-3
    'HLT_j460_a10sd_cssk_pf_jes_ftf_preselj225_L1J100',
    'HLT_j460_a10_lcw_subjes_L1J100',
    'HLT_j460_a10r_L1J100',
    ### new run-3 mass cut
    'HLT_j420_35smcINF_a10sd_cssk_pf_jes_ftf_preselj225_L1J100',
    'HLT_j420_35smcINF_a10t_lcw_jes_L1J100',
```

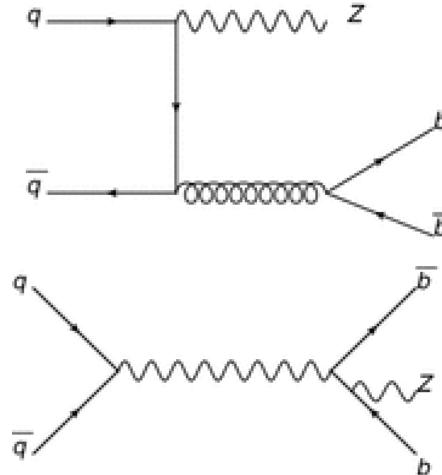
# Run3 MC production: background

- mc23d only, due to size constraints
  - Most dominant: QCD di-jet, divided in pT slices (JZ 0-9incl)
  - Can be used for machine learning validation or background estimation

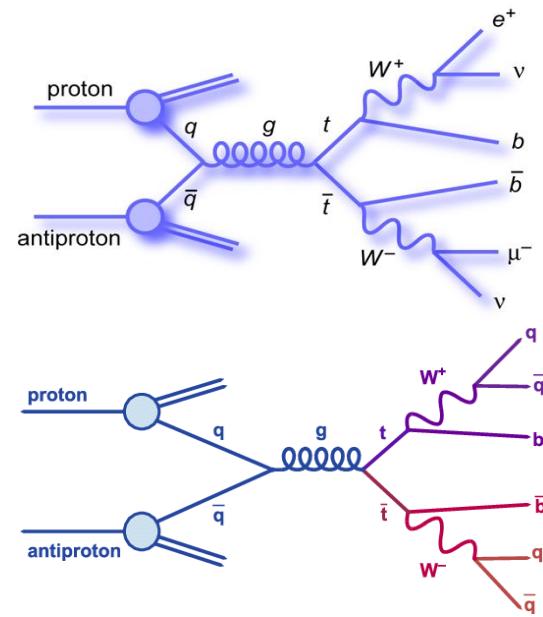
QCD di-jet (> 97%)



W/Z + jets (~2%)



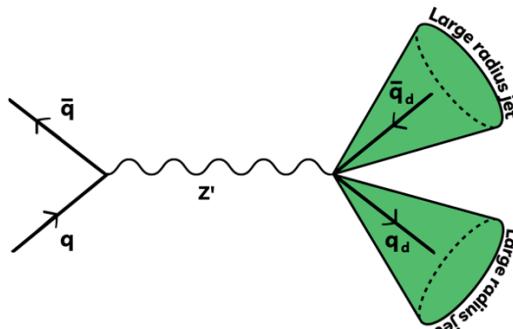
$t\bar{t}$  couples (~1%)



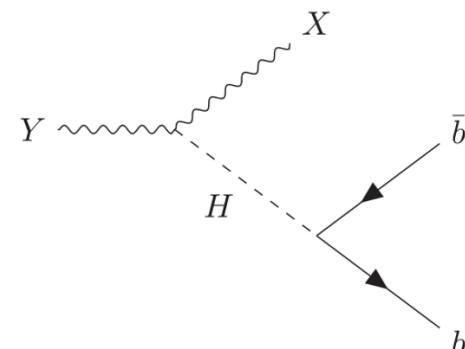
# Run3 MC production: signal

- mc23d only, due to size constraints
  - HVT signals;
  - Darkjets signals;
  - 3 prong signals;
- Produced for machine learning validation and results interpretation

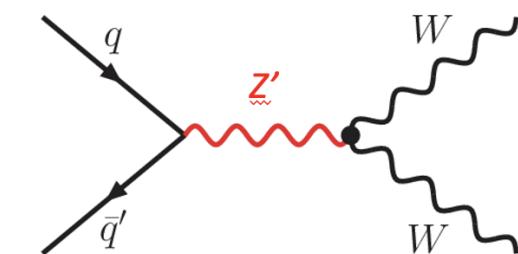
## ➤ DarkJets



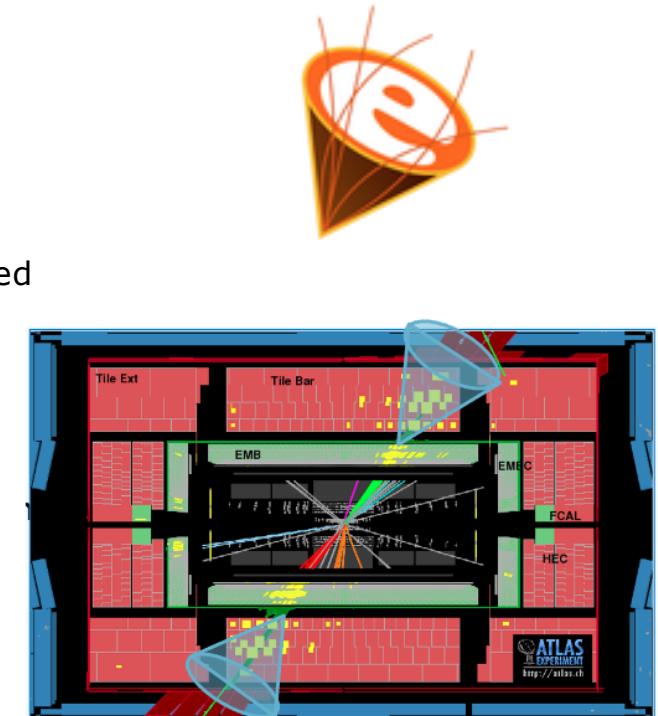
## ➤ YXH



## ➤ VVJJ



- Framework chosen to convert DAOD datasets (very hard to analyse) in ROOT ntuples (much easier to handle): [https://gitlab.cern.ch/atlas-roma1-napoli/easyjet/-/tree/rebase?ref\\_type=heads](https://gitlab.cern.ch/atlas-roma1-napoli/easyjet/-/tree/rebase?ref_type=heads)
- Preselection applied on jets during conversion:
  - $p_T > 200 \text{ GeV}$
  - $|\eta| < 2$
- Branches available in final ntuples:
  - Large-R jets and small-R jets 4 momenta
  - Large-R jets and muons trigger items
  - Trigger matching to large-R jets info for each trigger item included
  - Large-R jets constituents 4 momenta
  - Jets associated tracks 4 momenta,
  - Leptons (muons and electrons) 4 moments
- First production on CERN GRID available at:
  - [/eos/atlas/atlascerngroupdisk/phys-hmbs/mbl/AD\\_JJ/](https://eos/atlas/atlascerngroupdisk/phys-hmbs/mbl/AD_JJ/)
- Addition of new info after first production:
  - #jets, #constituents per jet, 4-momenta of leading dijet couples
  - PDGID of truth particles to be included in MC samples



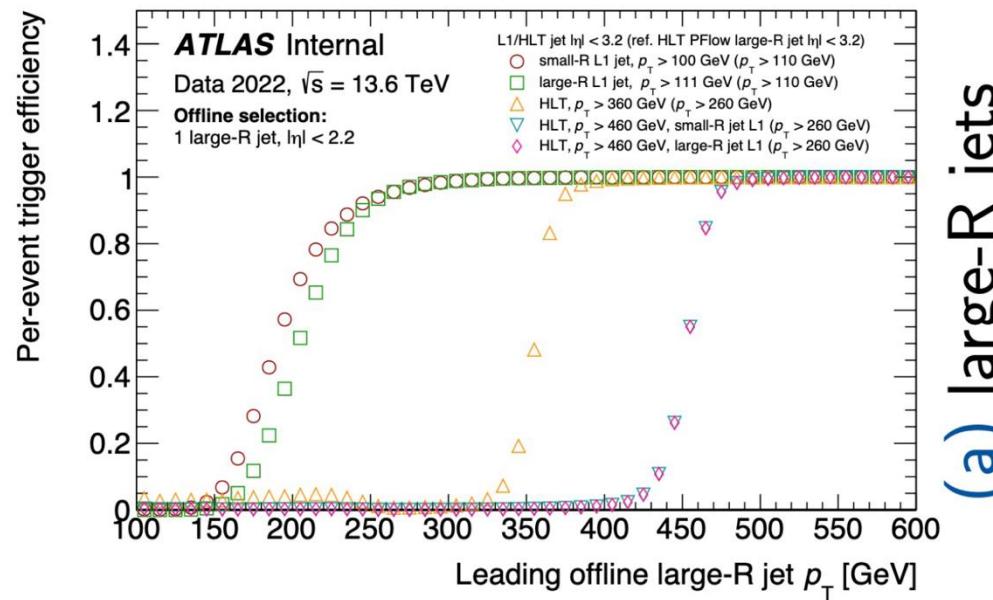
- Postprocessing step on Easyjet ntuples, necessary to include event weight for MC samples: <https://gitlab.cern.ch/atlas-roma1-napoli/adjf-fast-frames>
  - **Event weight** takes into account the actual number of events expected under certain conditions of **luminosity, cross section and pile up**
  - No need to generate all events in this way
- FF is very handy to combine minintuples of samples into one single ntuple, add selections and new variables and decide branches to keep
  - Selections for objects definition: leading jet  $m > 50$  GeV and subleading jet  $m > 50$  GeV
  - $n\text{Jets} > 2$
- Relatively fast, takes longer if constituents info is included
  - Size limit at 50 Gb for output, MC samples with constituents exceed this limit
- Produced ntuples are stored in mine and Graziella's eos workspace



# Trigger Efficiency Study

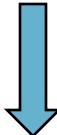
- Trigger study is important to build our signal region where we assure the full trigger efficiency
- Recommendations already exist, but we want to understand if it's possible to reduce pT threshold to help our model agnostic search
  - In principle, new physics could be hidden in jets below that threshold, so the more we can recover the better

Plots from ATL-COM-DAQ-2022-133



# An alternative trigger efficiency selection

- **Historic DBL trigger strategy:** trigger item passed  $\oplus$  leading jet  $p_T > 560 \text{ GeV} \oplus m_{JJ} > 1.3 \text{ TeV}$
- **Question: could it be lower to recover low- $p_T$  jets?**



Trigger matching of reconstructed objects implemented, can be used at our advantage!

**Tag&Probe method:** tag the **leading jet** if it matches the trigger item and use the **sub-leading  $p_T$**  to probe if it has passed the trigger

$$\text{efficiency}(\text{trig item}) = \frac{(\text{leading jet matching} == 1) \& \& (\text{subleading jet matching} == 1)}{\text{leading jet matching} == 1}$$

- Reason: Trigger skimming applied on our DxAOD, for this reason trigger efficiency study can't be made using the leading jet (i.e. bootstrap technique) without exploiting an auxiliary trigger item due to the bias
  - Efficiency curves are fitted with a sigmoid function  $\sigma(x)$
  - Used to extrapolate transverse momentum value at plateau

$$\sigma(x) = \frac{L}{1 + e^{-\frac{x-a}{b}}}$$

$$\begin{cases} L = \text{plateau} \\ a = \text{centre of sigmoid} \\ b = \text{interval of rise} \end{cases}$$

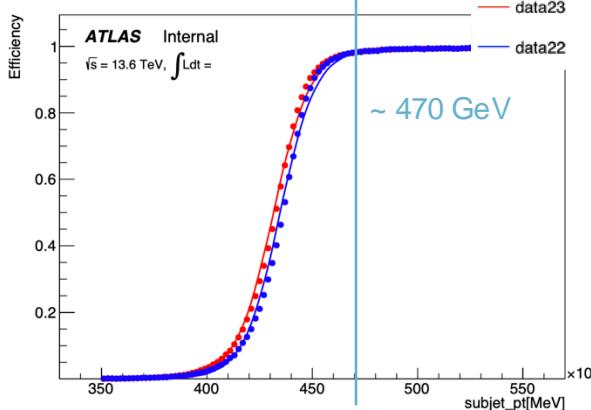


# It actually works!

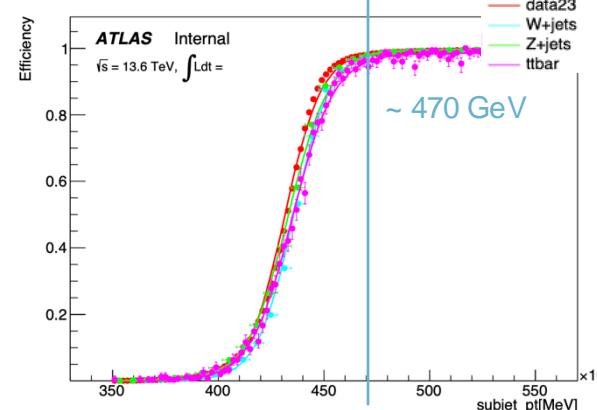
- Several trigger items studied
  - Best unprescaled trigger item: HLT\_j460\_a10\_sd\_cssk\_pf\_ftf\_preselj225\_L1J100

$$\sigma(x) = \frac{L}{1 + e^{-\frac{x-a}{b}}}$$

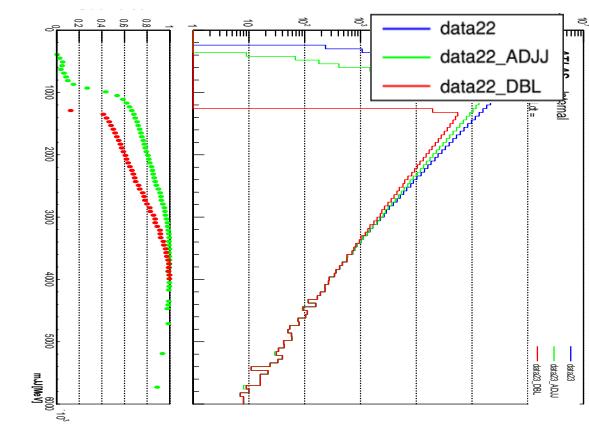
Sub-leading jet  $p_T$



Sub-leading jet  $p_T$



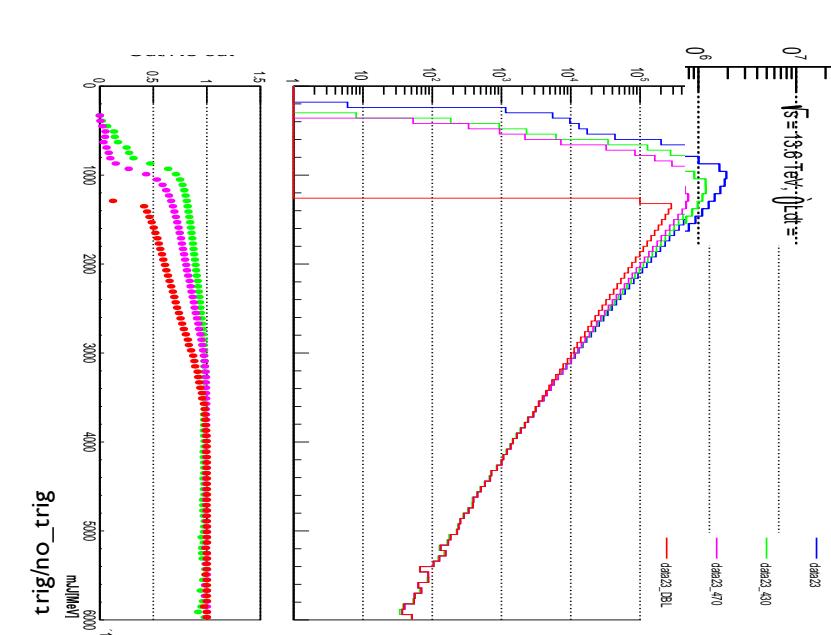
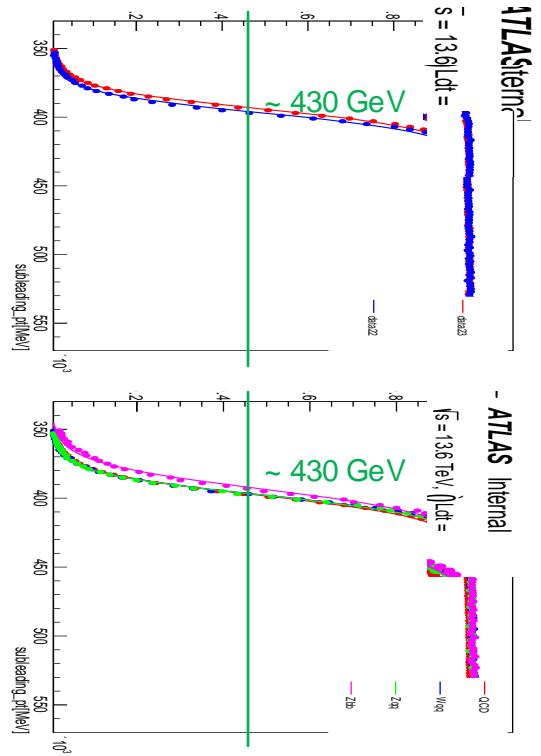
Invariant mass  $m_{JJ}$



trigger item passed  $\oplus$   
leading jet  $p_T > 560$  GeV  $\oplus$   
 $m_{JJ} > 1.3$  TeV

# OPTIMIZING TRIGGER STRATEGY

- Current trigger selection: trigger item passed  $\oplus$  trigger item matched with leading jet  $\oplus$  leading jet  $p_T > 470$  GeV
  - Trigger item: HLT\_j460\_a10\_sd\_cssk\_pf\_ftf\_preselj225\_L1J100
  - New checks on trigger items with mass cuts, 2 available
- Most promising one: [HLT\\_j420\\_35smcINF\\_a10sd\\_cssk\\_pf\\_jes\\_ftf\\_preselj225\\_L1J100](#)



**data23\_MBL:**  
trigger item passed  $\oplus$   
leading jet  $p_T > 560$  GeV  $\oplus$   
 $m_{JJ} > 1.3$  TeV

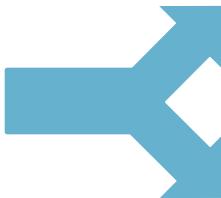
**data23\_470:**  
trigger item passed  $\oplus$   
leading jet matching  $\oplus$   
leading jet  $p_T > 470$  GeV

**data23\_430 (new proposed):**  
trigger item passed  $\oplus$   
leading jet matching  $\oplus$   
leading jet  $p_T > 430$  GeV

## Preselection = object definition + trigger selection

### Object definition

# jets > 2  $\oplus$   
leading jet  $m > 50 \text{ GeV} \oplus$   
subleading jet  $m > 50 \text{ GeV} \oplus$   
jets  $p_T > 200 \text{ GeV} \oplus$   
 $-2 < \text{jets } \eta < 2 \oplus$



### Trigger selection 470

trigger item passed  $\oplus$   
trigger item matched with leading jet  $\oplus$   
leading jet  $p_T > 470 \text{ GeV}$

### Trigger selection 430

trigger item passed  $\oplus$   
trigger item matched with leading jet  $\oplus$   
leading jet  $p_T > 470 \text{ GeV}$

- FastFrames ntuples produced with both trigger selections with constituents addition
  - 1M events max for QCD samples due to size



**BACKUP**

- Production of ntuples from our run 3 LLJ1 DxAOD based on EasyJet framework, large-R jets → Antikt10UFO jets.
- **News:**
  - Produced ntuple for data22, ~100k events.
  - Increased trigger list with new largeR-jet items, for both 2022 and 2023.
    - 2 items give problems with MC (#), can be turned off in their case.

**trigger list 2022**

```
'2022':
- 'HLT_j360_a10t_lcw_jes_L1J100'
- 'HLT_j420_a10sd_cssk_pf_jes_ftf_preselj225_L1J100'
- 'HLT_j460_a10sd_cssk_pf_jes_ftf_preselj225_L1J100'
- 'HLT_j460_a10t_lcw_jes_L1J100'
#- 'HLT_j460_a10r_L1J100'
#- 'HLT_j460_a10_lcw_subjes_L1J100'
- 'HLT_j420_35smcINF_a10sd_cssk_pf_jes_ftf_preselj225_L1J100'
- 'HLT_j420_35smcINF_a10t_lcw_jes_L1J100'
- 'HLT_mu24_ivarmedium_L1MU14FCH'
- 'HLT_mu50_L1MU14FCH'
- 'HLT_mu60_0eta105_msonly_L1MU14FCH'
- 'HLT_mu60_L1MU14FCH'
- 'HLT_mu80_msonly_3layersEC_L1MU14FCH'
```

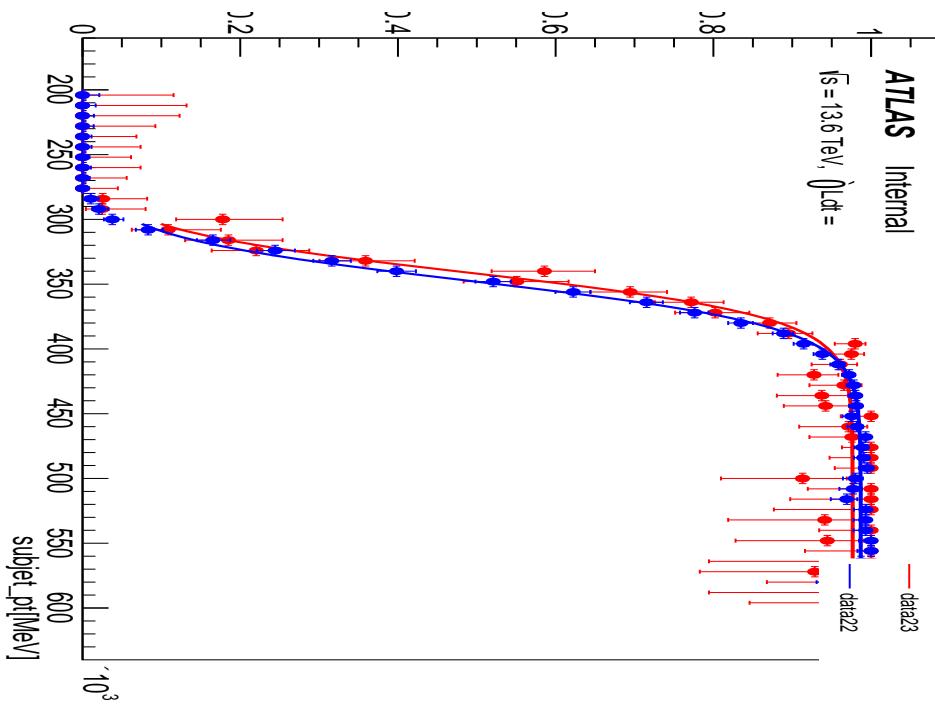
**trigger list 2023**

```
'2023':
- 'HLT_j360_a10t_lcw_jes_L1J100'
- 'HLT_j420_a10sd_cssk_pf_jes_ftf_preselj225_L1J100'
- 'HLT_j460_a10sd_cssk_pf_jes_ftf_preselj225_L1J100'
- 'HLT_j460_a10t_lcw_jes_L1J100'
#- 'HLT_j460_a10r_L1J100'
#- 'HLT_j460_a10_lcw_subjes_L1J100'
- 'HLT_j420_35smcINF_a10sd_cssk_pf_jes_ftf_preselj225_L1J100'
- 'HLT_j420_35smcINF_a10t_lcw_jes_L1J100'
- 'HLT_mu24_ivarmedium_L1MU14FCH'
- 'HLT_mu50_L1MU14FCH'
- 'HLT_mu60_0eta105_msonly_L1MU14FCH'
- 'HLT_mu60_L1MU14FCH'
- 'HLT_mu80_msonly_3layersEC_L1MU14FCH'
```

Unprescaled run 3 triggers twiki [here](#)

# Trigger study on data

- Trigger item: HLT\_j360\_a10t\_lcw\_jes\_L1J100
  - 2023 and 2022 data
- Trigger Item is prescaled for 2023 data



$$\text{fit function (sigmoid)} = \frac{p_2}{1 + e^{-\frac{x - p_1}{p_0}}}$$

- fit and extrapolation values

**data23**

p0	=	17241.7	+/-	1047.67
p1	=	341034	+/-	1508.44
p2	=	0.976619	+/-	0.00605004

Found cut off at 0.97 efficiency: 427025 MeV

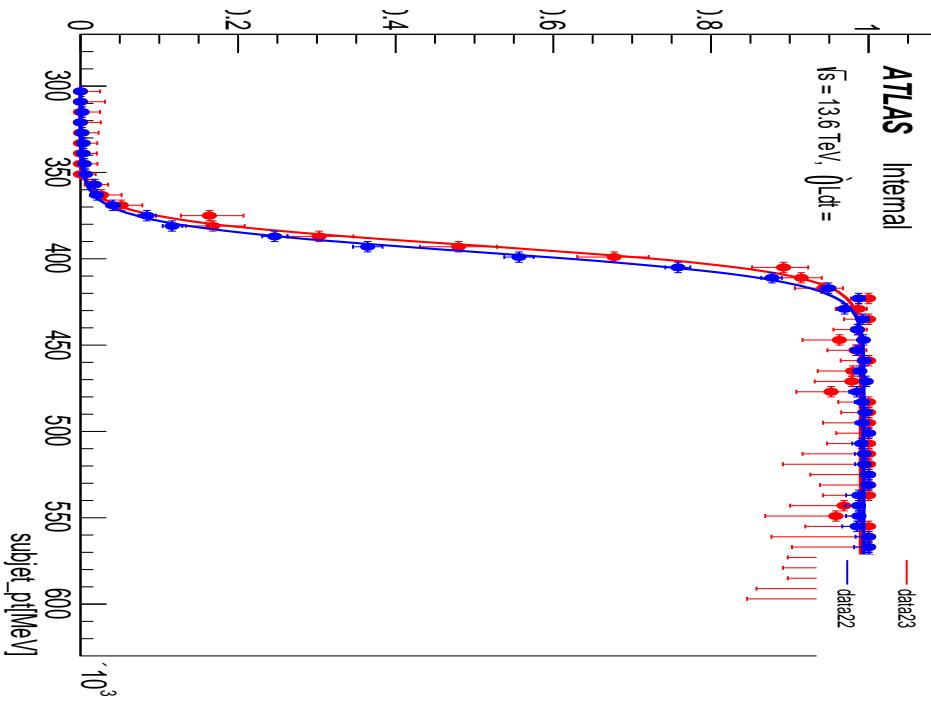
**data22**

p0	=	17694.9	+/-	382.35
p1	=	347636	+/-	580.226
p2	=	0.986821	+/-	0.00191717

Found cut off at 0.98 efficiency: 435537 MeV

# Trigger study on data

- Trigger item: HLT\_j420\_a10sd\_cssk\_pf\_jes\_ftf\_preselj225\_L1J100
  - 2023 and 2022 data
- This one is also prescaled for 2023, not for 2022



$$\text{fit function (sigmoid)} = \frac{p_2}{1 + e^{\frac{x - p_1}{p_0}}}$$

- fit and extrapolation values

**data23**

p0	=	7705.95	+/-	378.696
p1	=	392137	+/-	637.458
p2	=	0.990094	+/-	0.00280787

Found cut off at 0.99 efficiency: 463507 MeV

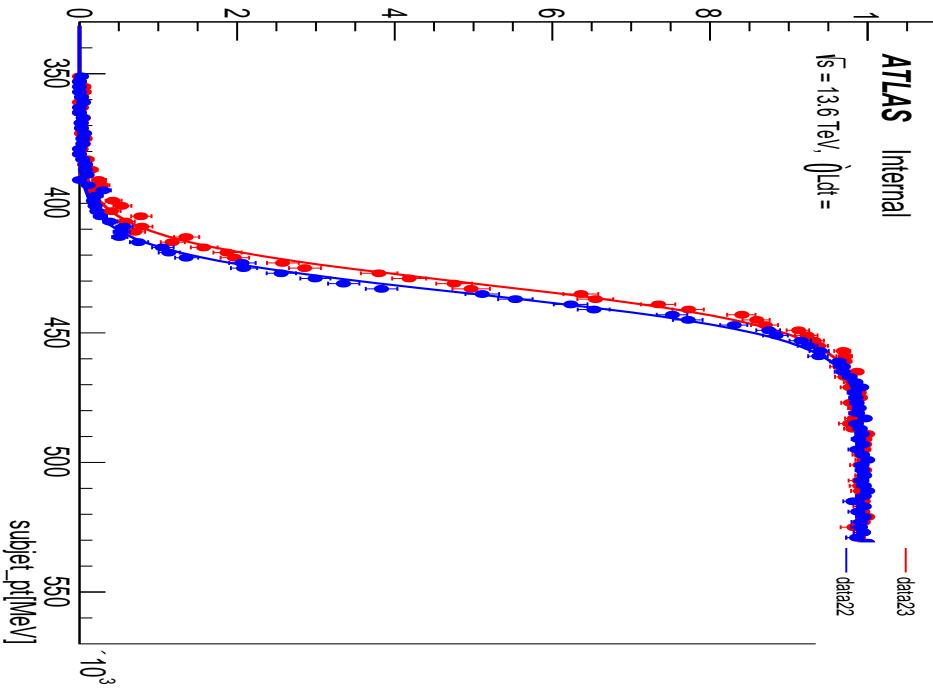
**data22**

p0	=	8120.13	+/-	159.395
p1	=	395867	+/-	265.546
p2	=	0.993706	+/-	0.000985083

Found cut off at 0.99 efficiency: 441240 MeV

# Trigger study on data

- Trigger item: HLT\_j460\_a10sd\_cssk\_pf\_jes\_ftf\_preselj225\_L1J100
  - 2023 and 2022 data
- Good candidate for data 2023



$$\text{fit function (sigmoid)} = \frac{p_2}{1 + e^{-\frac{x - p_1}{p_0}}}$$

- **fit and extrapolation values**

**data23**

p0	=	8737.24	+/-	119.947
p1	=	430779	+/-	196.361
p2	=	0.993001	+/-	0.000923239

Found cut off at 0.99 efficiency: 481444 MeV

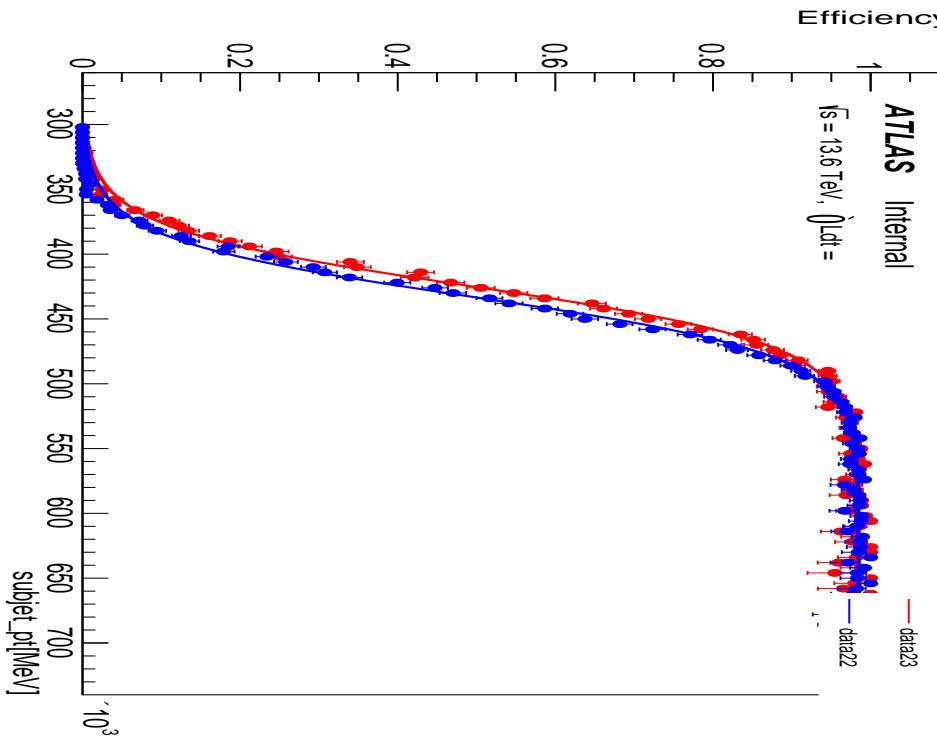
**data22**

p0	=	8290.88	+/-	105.462
p1	=	434886	+/-	176.388
p2	=	0.993693	+/-	0.000797063

Found cut off at 0.99 efficiency: 481243 MeV

# Trigger study on data

- Trigger item: HLT\_j460\_a10r\_L1J100
  - 2023 and 2022 data



$$\text{fit function (sigmoid)} = \frac{p_2}{1 + e^{-\frac{x - p_1}{p_0}}}$$

## ➤ fit and extrapolation values

data23

p0	=	22423.4	+/-	267.65
p1	=	424553	+/-	391.524
p2	=	0.981706	+/-	0.00166781

Found cut off at 0.97 efficiency: 523602 MeV

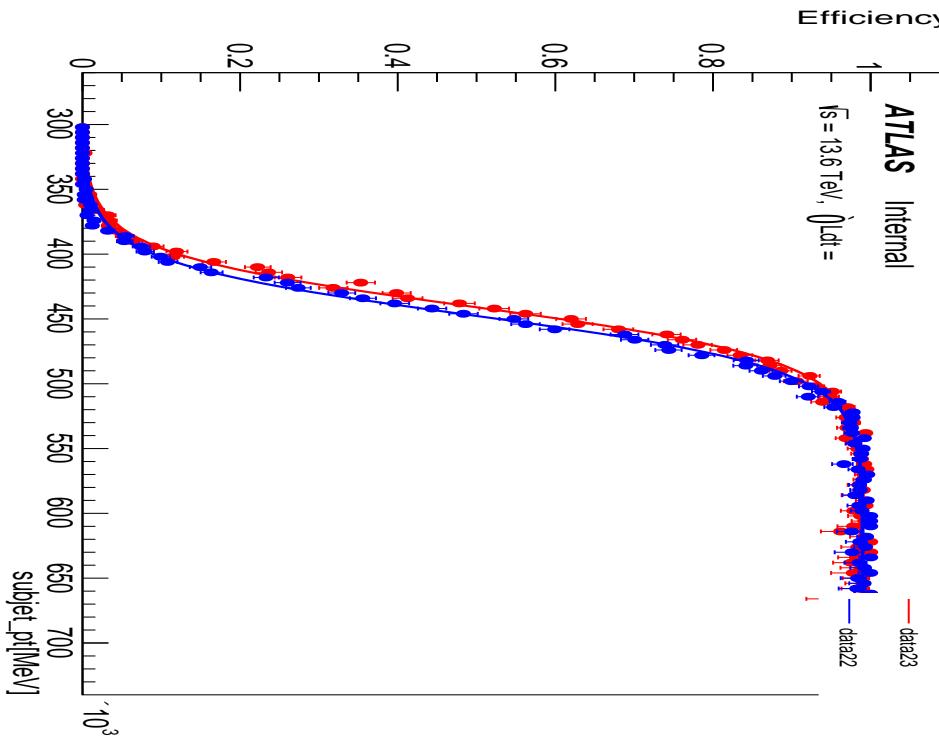
data22

p0	=	22383.8	+/-	245.315
p1	=	432760	+/-	358.861
p2	=	0.985058	+/-	0.00142138

Found cut off at 0.98 efficiency: 550647 MeV

# Trigger study on data

- Trigger item: HLT\_j460\_a10t\_lcw\_jes\_L1J100
  - 2023 and 2022 data



$$\text{fit function (sigmoid)} = \frac{p_2}{1 + e^{\frac{x - p_1}{p_0}}}$$

- fit and extrapolation values

data23

p0	=	20584.8	+/-	259.742
p1	=	440484	+/-	390.226
p2	=	0.987447	+/-	0.00150964

Found cut off at 0.98 efficiency: 540932 MeV

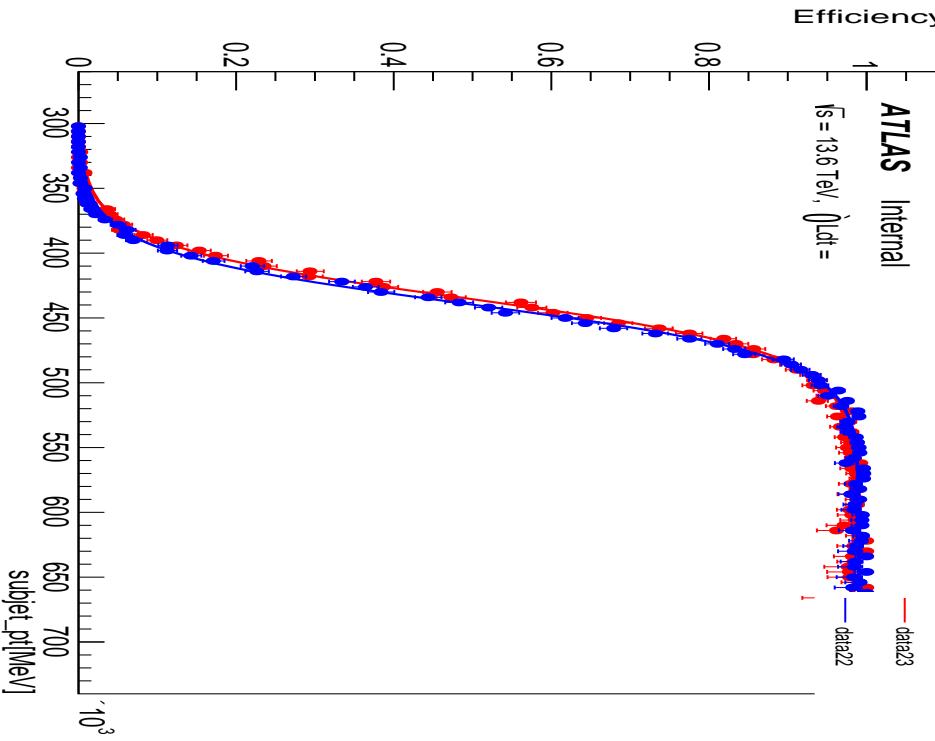
data22

p0	=	20580.3	+/-	239.521
p1	=	447298	+/-	363.356
p2	=	0.99001	+/-	0.00127226

Found cut off at 0.98 efficiency: 541637 MeV

# Trigger study on data

- Trigger item: HLT\_j460\_a10\_lcw\_subjes\_L1J100
  - 2023 and 2022 data



$$\text{fit function (sigmoid)} = \frac{p_2}{1 + e^{-\frac{x - p_1}{p_0}}}$$

- fit and extrapolation values

**data23**

p0	=	20702.3	+/-	264.489
p1	=	434846	+/-	390.602
p2	=	0.983288	+/-	0.00165146

Found cut off at 0.98 efficiency: 552791 MeV

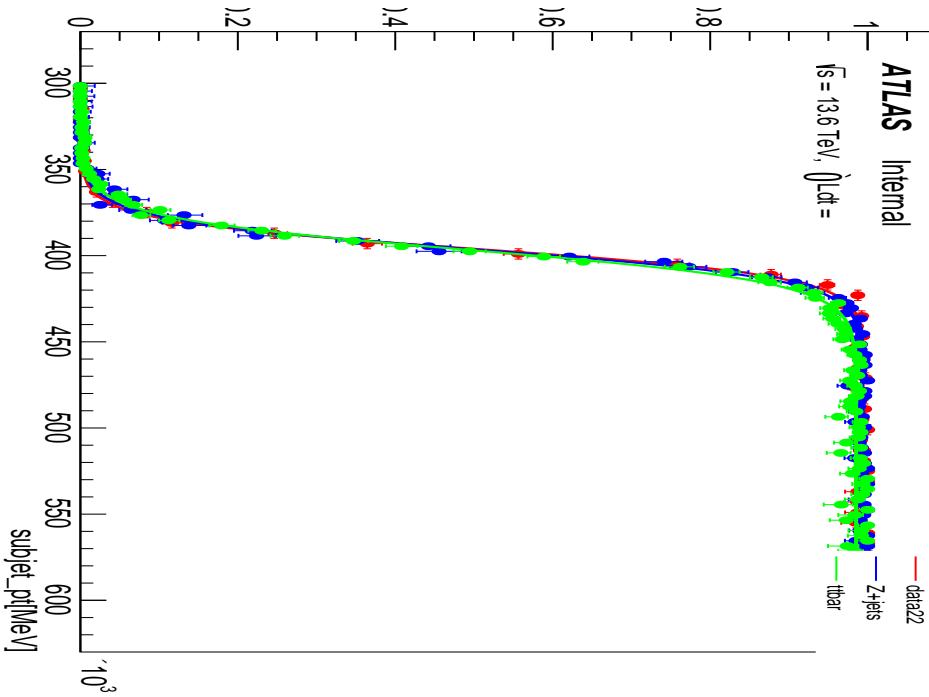
**data22**

p0	=	20236.8	+/-	226.398
p1	=	439348	+/-	343.221
p2	=	0.990331	+/-	0.00115468

Found cut off at 0.99 efficiency: 601296

# Trigger study on MC bkg

- Trigger item: HLT\_j420\_a10sd\_cssk\_pf\_jes\_ftf\_preselj225\_L1J100
  - Z + jet and ttbar fully hadronic
- Nice closure between curves and with data 2022



$$\text{fit function (sigmoid)} = \frac{p_2}{1 + e^{\frac{x - p_1}{p_0}}}$$

- fit and extrapolation values

Data 2022

p0	=	8120.13	+/-	159.395
p1	=	395867	+/-	265.546
p2	=	0.993706	+/-	0.000985083

Found cut off at 0.99 efficiency: 441240 MeV

Z + jet

p0	=	8756.98	+/-	170.396
p1	=	396203	+/-	281.215
p2	=	0.993021	+/-	0.000715131

Found cut off at 0.99 efficiency: 446924 MeV

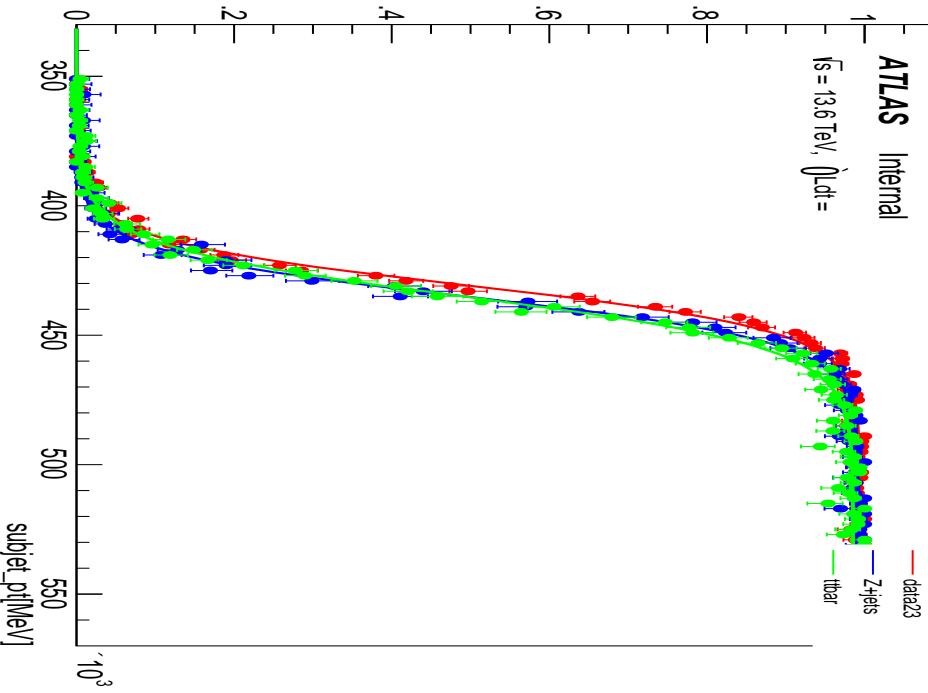
ttbar

p0	=	9527.8	+/-	137.616
p1	=	396632	+/-	226.778
p2	=	0.985301	+/-	0.0011313

Found cut off at 0.98 efficiency: 446364 MeV

# Trigger study on MC bkg

- Trigger item: HLT\_j460\_a10sd\_cssk\_pf\_jes\_ftf\_preselj225\_L1J100
  - Z + jet and ttbar fully hadronic
- Nice closure between curves and with data 2023



$$\text{fit function (sigmoid)} = \frac{p_2}{1 + e^{\frac{x - p_1}{p_0}}}$$

- fit and extrapolation values

Data 2023

p0	=	8737.24	+/-	119.947
p1	=	430779	+/-	196.361
p2	=	0.993001	+/-	0.000923239

Found cut off at 0.99 efficiency: 481444 MeV

Z + jet

p0	=	8704.49	+/-	184.957
p1	=	435131	+/-	294.569
p2	=	0.991279	+/-	0.00122803

Found cut off at 0.99 efficiency: 493027 MeV

ttbar

p0	=	9872.54	+/-	175.105
p1	=	435088	+/-	290.259
p2	=	0.985346	+/-	0.00170914

Found cut off at 0.98 efficiency: 486535 MeV

# ADJJ vs DBL

- **DBL**: trigger item passed  $\oplus$  leading jet  $p_T > 560 \text{ GeV} \oplus m_{JJ} > 1.3 \text{ TeV}$
- **ADJJ**: trigger item passed  $\oplus$  trigger item matched with leading jet  $\oplus$  leading jet  $p_T > 470 \text{ GeV}$

