# Studying jets faking photons

#### Giulia Maineri, Marcello Fanti, Silvia Resconi, Federica Piazza

University of Milan

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#### Why don't we use ABCD method?



Problem: we don't have loose photons

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**Problem:** We can't use loose photons, because our trigger selects only Tight Photons.

Solution: define Non-Isol Region with cuts on relative isolation variables, in tracker and calorimeter.

$$isol_{track}^{rel} = rac{p_T^{track20}}{p_T^{\gamma}}$$

$$isol_{calo}^{rel} = \frac{E_T^{calo40} - 2450}{p_T^{\gamma}}$$

- We first based this study on Monte Carlo simulations, including dijets and  $\gamma+{\rm jets.}$
- We want to find the best **Non-Isolated Region**, i.e. the region richest in jet faking photons.

#### This Non-Isol Region will be used to estimate fake factor f.

$$f = \Big(rac{N^{isol}_{j\gamma}}{N^{non-isol}_{j\gamma}}\Big)_{tight}$$

where

 $N_{j\gamma}^{isol}$  is the number of jet faking photons in the Isolated Region  $N_{j\gamma}^{non-isol}$  is the number of jet faking photons in the Non-Isolated Region

#### Purity

We also need to take care of true photons in the Non-Isolated Region: this fraction is given by **purity** P.

$$P = \left(\frac{N_{\gamma}^{non-isol}}{N_{\gamma}^{non-isol} + N_{j\gamma}^{non-isol}}\right)_{tight}$$

i.e. the fraction of true photons in the Non-Isolated Region.

f and P will be used to estimate the number of jet faking photons in the Signal Region.

In this study the quantity P' was analized instead of P:

$$P' = \frac{N_{\gamma}^{non-isol}}{N_{j\gamma}^{non-isol}}$$

This choice doesn't affect the study, because minimizing P' means minimizing P:

$$\frac{1}{P} = 1 + \frac{1}{P'}$$

#### Calo relative isolation vs track relative isolation



Figure 1: Jet faking photons distribution

Calo\_vs\_track\_trueph

Figure 2: True photons distribution

# Fake factor and purity

Calo relative isolation vs track relative isolation



Calo\_vs\_track\_trueph



Figure 3: Fake factors method sketch

We would like to have P' as small as possible, and at the same time f reasonably small: few fake photons in the Isolated Region and few true photons in the Non-Isolated Region.

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# Which photons are "true" and which are "fake"?

True/Fake	Name	MC Truth Classifier Type	
True	IsoPhotons	14	Fragmentation photons
Fake	NonIsoPhotons	15	
	UnknownPhotons	13	
	BkgPhotons	16	
	everything else that is not γ neither e		

**Fragmentation** photons are photons emerging from parton showers. The photons we are looking for, from  $H \rightarrow \gamma \gamma_d$ , are much more similar to **direct** photons.

#### Which photons are "true" and which are "fake"?



Fragmentation photons are in any case prompt photons so in inclusive photon analyses they are considered part of the signal " $a_{\text{photon}} = a_{\text{photon}} = a_{\text{photon}}$ 

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The first approach consisted in defining a figure of merit M to elect the best Non-isolated Region.



#### Search for a Non-Isolated Region



Figure 4: Search for a Non-Isolated Region ortogonal to the Isolated Region

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The best Non-Isolated Region is defined by:

$$isol_{calo}^{rel} = \frac{E_T^{calo40} - 2450}{p_T^{\gamma}} > 0.022$$
$$\forall \quad isol_{track}^{rel} = \frac{p_T^{track20}}{p_T^{\gamma}}$$

Fake factor and purity in this Non-Isolated Region are:

 $f = 0.83 \pm 0.10$ P = 0.0864761

#### Elected Non-Isolated Region

Calo\_vs\_track\_fakeph



Figure 5: Pseudo-purity P' in fine bins of relative isolation in tracker and calo

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#### Elected Non-Isolated Region

We then abandoned relative isolation in the tracker: it is not recommended as working point.

This doesn't affect our choice for the Non-Isolated Region.



Figure 6: Relative isolation in calo for tight photons MC.

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Isolation is typically **not** well modelled by MC for jet faking photons.

We need to check our results on data.

How to get jet faking photons in data?

We have different possibile ID selection:

- loose, if they satisfy loose criteria but not the tight ones;
- loose5, if they satisfy the loose criteria and pass tight cuts on all egamma shower shapes of HCAL and ECAL Middle layer;
- loose4, if they are loose5 and pass tight cuts on  $W_{stot}$  shower shape of the ECAL Strips.

#### Loose photons



Figure 7: Variables describing shower shapes, energy ratios and width of the energy deposit

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#### Loose photons



Figure 8: Variables describing shower shapes, energy ratios and width of the energy deposit (loose5)

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Figure 9: Variables describing shower shapes, energy ratios and width of the energy deposit (loose4)

Larger statistic, lower true photons contamination



Lower statistic, larger true photons contamination

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#### Isolation comparison in MC: tight, loose, loose4, loose5

Isolation for tight, loose, loose4 and loose5 fake photons in MC is different



Figure 10: MC isolation for fake photons with different ID

## Isolation comparison (fake) $\gamma$ in DATA: loose, loose4, loose5

Isolation for loose, loose4 and loose5 photons in DATA is different



Figure 11: DATA isolation for (fake) photons with different ID

#### Isolation comparison for loose fake $\gamma$ : MC and data

MC and DATA isolations for a same ID are different



Figure 12: Calo isolation in MC and DATA for loose photons

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#### Isolation comparison for loose5 fake $\gamma$ : MC and data

MC and DATA isolations for a same ID are different



Figure 13: Calo isolation in MC and DATA for loose5 photons

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#### Isolation comparison for loose4 fake $\gamma$ : MC and data

MC and DATA isolations for a same ID are different



Figure 14: Calo isolation in MC and DATA for loose4 photons

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We would like to calculate fake factors using tight fake photons in data, to check our results obtained with MC.

**Problem**: In data, we cannot know which tight photons are true and which fake.

Solution: We can try to extrapolate the tight fake photons distribution in data from the loose photons distribution in data, assuming that:

- loose photons in data are mostly fake;
- the transformation that links tight and loose distributions in MC is "somehow" related to the one that links tight and loose in data.

#### Extrapolation L->T



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# Step 1: get L->T transformation from MC

Let's assume that tight  $x_T^{MC}$  and loose  $x_L^{MC}$  distributions in MC are linked by an affine transformation, the easiest transformation to reproduce mean and standard deviation of the start distribution:

$$x_T^{MC} = a + b x_L^{MC}$$

We want to find *a*, *b* such that:



# Step 1: get L->T transformation from MC

Let's assume:

- the scale factor *b* stays the same in MC and data;
- the offset *a* in data should depend on  $\sigma_L^{data}, \sigma_T^{data}$ , which is known, and on  $\mu_T^{data}$ , which is unknown. So we assume the shift of the average going from loose to tight is proportional to the rms in both data and MC.



It's all set, we can now apply the transformation to DATA:

$$x_T^{data} = \mu_L^{data} + \frac{\sigma_L^{data}}{\sigma_L^{MC}} (\mu_T^{MC} - \mu_L^{MC}) + \frac{\sigma_T^{MC}}{\sigma_L^{MC}} (x_L^{data} - \mu_L^{data})$$

 $\rightarrow$  We obtain tight fake photons distributions in DATA.

#### Step 3: obtained distributions



Figure 16: Distributions of tight fake photons (DATA) isolation, extrapolated from loose (DATA)

Problem: we need to introduce a trigger to be able to compare data and Monte Carlo. Solution: Ok, but which one?

#### Analysis Trigger

- $N_{\gamma^{tight}} = 1$
- $|\vec{p}_T^{\gamma}| > 50 \text{ GeV}$
- $|\vec{p}_T^{miss}| > 70 \, \text{GeV}$
- $m_T > 80 \, \text{GeV}$

# Leptonic TriggerMET Trigger• $N_{el} = 1$ or $N_{el} = 2$ ;• $|\vec{p}_T^{miss}| > 90 \text{ GeV}$

$$\frac{\mu_T^{data} - \mu_L^{data}}{\sigma_L^{data}} = \frac{\mu_T^{MC} - \mu_L^{MC}}{\sigma_L^{MC}}$$

Let's validate our hypothesis on loose5 photons, which we can compare in data and MC assuming they are mostly fake photons. We introduce a factor R; let's see when R is close to 1.

$$\frac{\mu_{L5}^{data} - \mu_{L}^{data}}{\sigma_{L}^{data}} = R \frac{\mu_{L5}^{MC} - \mu_{L}^{MC}}{\sigma_{L}^{MC}}$$

Pseudorapidity binning is chosen considering the detector geometry:

- etabin00 represents the inclusive region;
- etabin01: [0; 0.6], the upper limit  $\eta = 0.6$  is the point after which the material in front of ECAL increases a lot;
- etabin02: [0.6; 1.37], the upper limit is defined by the beginning of the crack region;
- etabin03: [1.37; 1.52], corresponds to the crack region;
- etabin04: [1.52; 1.81], the upper limit is the point where the presampler ends;
- etabin05: [1.81; 2.37].

Transverse momentum binning is chosen as follows:

- ptbin00 represents the inclusive region;
- ptbin01: [25, 35]GeV;
- ptbin02: [35, 45]GeV;
- ptbin03: [45, 55]GeV;
- ptbin04: [55, 65]GeV;
- ptbin05: [65,75]GeV;
- ptbin06: [75, 100]GeV;
- ptbin07: [100, 150]GeV;
- ptbin08: [150, 250]GeV;

# Analysis trigger



R\_trigger\_analisi

#### Figure 17: R, analysis trigger

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Image: A matrix

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#### Leptonic trigger



R\_trigger\_lept

#### Figure 18: R, leptonic trigger

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# MET trigger



R\_trigger\_met

Figure 19: R, MET trigger

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#### Validation on loose5

Let's do the same validation using median and width, two indicators less sensitive to outliers.



where med is the median and w is calculated as:



# Analysis trigger



#### R\_median\_trigger\_analisi

Figure 20: R, analysis trigger

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# Leptonic trigger



R\_median\_trigger\_lept

#### Figure 21: R, leptonic trigger

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Image: A matrix

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# MET trigger



R\_median\_trigger\_met

Figure 22: R, MET trigger

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Problem: R is very unstable. It is not possible to perform the extrapolation in an exclusive regions in  $p_T$ ,  $\eta$ .

Solution: Let's be either inclusive in  $p_T$  or in  $\eta$ .

#### R comparison for different trigger, inclusive in $\eta$



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# R comparison for different trigger, median and width, inclusive in $\boldsymbol{\eta}$



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# $\eta$ distribution, MET trigger



# $\eta$ distribution, analysis trigger



#### R comparison for different trigger, inclusive in $p_T$



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# R comparison for different trigger, median and width, inclusive in $p_T$



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# $p_T$ distribution, analysis trigger



# $p_T$ distribution, MET trigger



We are left with these possibilities:

MET Trigger Mean: 0.94 Spread: 0.14 inclusive in  $\eta$ mean, sigma Analysis Trigger Mean: 0.87 Spread: 0.14 inclusive in  $\eta$ mean, sigma MET Trigger Median: 0.81 Width: 0.15 inclusive in  $\eta$ median, width

Analysis Trigger Median: 1.1 Width: 0.26 inclusive in η median, width

#### Analysis Trigger Mean: 0.94 Spread: 0.07 inclusive in *p*<sub>T</sub> mean, sigma

#### Analysis Trigger Median: 0.82

Width: 0.08 inclusive in  $p_T$  median, width

We are left with these possibilities:

MET Trigger Mean: 0.94 Spread: 0.14 inclusive in  $\eta$ mean, sigma Analysis Trigger Mean: 0.87 Spread: 0.14 inclusive in  $\eta$ mean, sigma

#### Analysis Trigger

Mean: 0.94 Spread: 0.07 inclusive in  $p_T$ mean, sigma MET Trigger Median: 0.81 Width: 0.15 inclusive in  $\eta$ median, width

#### Analysis Trigger Median: 1.1 Width: 0.26 inclusive in $\eta$ median, width

#### Analysis Trigger Median: 0.82 Width: 0.08 inclusive in $p_T$ median, width

#### Summary

- We presented this **new method** to estimate the jet faking photons background based on **extrapolation** of fake tight photons distributions in data from loose photons distributions.
- We tried to **validate** the method comparing loose5 distribution with the extrapolated one.
- Ratio R in exclusive regions in η, p<sub>T</sub> was found to suffer from fluctuations, hence the extrapolation should be done in a region inclusive in η or p<sub>T</sub> only;
- Different triggers were explored in order to have R ~ 1 and a small spread in either η or p<sub>T</sub>: the two best options is Analysis Trigger inclusive in p<sub>T</sub>, using as indicators median and width.

Next steps

- Extrapolate fake tight photons isolation distribution;
- Calculate fake factors;
- How to treat fake factors uncertainties: need to propagate mean and rms errors, envelope method?
- Calculate purity from the extrapolated distribution of tight fake photons in DATA: normalizing the tail and subtracting;

