Lecture 4

Analysis Necessities & Steps ...

Photon reconstruction Photon identification Photon isolation Primary vertex Energy calibration Background modeling

Event categories

Limits & signal strength



Energy Resolution



Energy Resolution

Test Beam Result Fractional Energy Resolution



Resolution @ 60 GeV: $\sigma_E \approx 0.014 \text{ [FWHM} = 3.3 \%]$

Event numbers and mass resolution for the H \rightarrow $\gamma\gamma$ ATLAS analysis ...

[Mass range: 100 - 160]

\sqrt{s}	7 Te	eV		8 TeV	
$\sigma \times B(H \to \gamma \gamma)$ [fb]		39		50	FWHM
Category	N _D	$N_{\rm S}$	N _D	$N_{\rm S}$	[GeV]
Unconv. central, low p_{Tt}	2054	10.5	2945	14.2	3.4
Unconv. central, high p_{Tt}	97	1.5	173	2.5	3.2
Unconv. rest, low p_{Tt}	7129	21.6	12136	30.9	3.7
Unconv. rest, high p_{Tt}	444	2.8	785	5.2	3.6
Conv. central, low p_{Tt}	1493	6.7	2015	8.9	3.9
Conv. central, high p_{Tt}	77	1.0	113	1.6	3.5
Conv. rest, low p_{Tt}	8313	21.1	11099	26.9	4.5
Conv. rest, high p_{Tt}	501	2.7	706	4.5	3.9
Conv. transition	3591	9.5	5140	12.8	6.1
2-jet	89	2.2	139	3.0	3.7
All categories (inclusive)	23788	79.6	35251	110.5	3.9

[ATLAS, Phys. Lett. B 716 (21012) 1]

Energy Calibration

Monte Carlo based calibration

Monte Carlo simulation tuned with Test Beam data.

Accurate description of materials is confirmed by measurements in data.

Energy scale corrections using $Z \rightarrow$ ee decay data ...

Energy scale correction applied to data ... Correction from a fit to the 2010 $Z \rightarrow$ ee data ... Extrapolation of energy scale correction from electron to photon is treated as uncertainty ... MC energy is smeared to match the energy resolution determined from data ...



Reconstructed Vertex Distribution



Energy Calibration



Excellent stability with time and pileup ! Di-photon mass resolution around 1 %

Di-Photon Vertex Selection



Likelihood combining calorimeter pointing, conversion vertex and track-based vertex selection used ...



Di-Photon Vertex Selection



Likelihood combining calorimeter pointing, conversion vertex and track-based vertex selection used ...



ATLAS Result Observation of a New Particle [$H \rightarrow \gamma \gamma$]

[Summer 2012]



ATLAS Result Observation of a New Particle [$H \rightarrow \gamma \gamma$]

[Spring 2013]



ATLAS-CONF-2013-012

Signal Model

Signal modeled using Crystal Ball function plus a broad Gaussian ...

[Width dominated by detector resolution; Gaussian account for poorly measured energy]

Taken from Monte Carlo ... [POWHEG and PYTHIA]



invariant mass of a simulated 120 GeV mass Higgs signal

Background Model

Background obtained from fit to observed di-photon invariant mass distribution ... [Exponential, 4th-order Bernstein polynomial, 4th order polynomial, exponential function of a 2nd-order polynomial]

Different parametrization chosen for different event categories ...

[Limit potential bias while keeping good statistical power]

Uncertainty estimated using Monte Carlo ...



Event Categorization

10 Categories

with different S/B and resolution increases expected signal sensitivity by 25% ...

make use of conversion status, |η|, p_{Tt} [≥ 60 GeV], 2 jets category

Unconverted central, low p_{Tt} Unconverted central, high p_{Tt} Unconverted rest, low p_{Tt} Unconverted rest, high p_{Tt} Converted central, low p_{Tt} Converted central, high p_{Tt} Converted rest, low p_{Tt} Converted rest, high p_{Tt} Converted transition region 2-jet category





Higgs $\rightarrow \gamma\gamma + 2$ jets

ATLAS EXPERIMENT

Run Number: 204769, Event Number: 24947130

Date: 2012-06-10 08:17:12 UTC

Mass Spectra for Different Categories

[2011 & 2012]



Mass Spectra for Different Categories

[2011 & 2012]



Observation or Fluctuation?



Signal Model Parameters

 $N \cdot \begin{cases} e^{-t^2/2} & \text{if } t > -\alpha \\ (\frac{n}{|\alpha|})^n \cdot e^{-|\alpha|^2/2} \cdot (\frac{n}{|\alpha|} - |\alpha| - t)^{-n} & \text{otherwise} \end{cases}$ with $t=(m_{\gamma\gamma}-m_H-\delta_{m_H})/\sigma_{CB}$

Category	σ_{CB}	FWHM	Observed	S	В
	[GeV]	[GeV]	$[N_{\rm evt}]$	$[N_{\rm evt}]$	$[N_{\rm evt}]$
Inclusive	1.63	3.87	3693	100.4	3635
Unconverted central, low p_{Tt}	1.45	3.42	235	13.0	215
Unconverted central, high p_{Tt}	1.37	3.23	15	2.3	14
Unconverted rest, low p_{Tt}	1.57	3.72	1131	28.3	1133
Unconverted rest, high p_{Tt}	1.51	3.55	75	4.8	68
Converted central, low p_{Tt}	1.67	3.94	208	8.2	193
Converted central, high p_{Tt}	1.50	3.54	13	1.5	10
Converted rest, low p_{Tt}	1.93	4.54	1350	24.6	1346
Converted rest, high p_{Tt}	1.68	3.96	69	4.1	72
Converted transition	2.65	6.24	880	11.7	845
2-jets	1.57	3.70	18	2.6	12

Estimating the Significance ...

Naive approach:

N_S	≈ 200	[120 - 130 GeV]
NΒ	≈ 60000	[100 - 160 GeV]

$$S = 2$$
 [= 200/ $\sqrt{10000}$]

as N_{B} needs to be corrected to 10 GeV range



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Use of extra information by performing a fit to the background ...

and optimizing by channel categorization ...

Needs procedure to combine ...

Background Model Systematics

Category	Parametrization	Uncertainty [N _{evt}]	
		$\sqrt{s} = 7 \text{ TeV}$	$\sqrt{s} = 8 \text{ TeV}$
Inclusive	4th order pol.	7.3	10.6
Unconverted central, low p_{Tt}	Exp. of 2nd order pol.	2.1	3.0
Unconverted central, high p_{Tt}	Exponential	0.2	0.3
Unconverted rest, low p_{Tt}	4th order pol.	2.2	3.3
Unconverted rest, high p_{Tt}	Exponential	0.5	0.8
Converted central, low p_{Tt}	Exp. of 2nd order pol.	1.6	2.3
Converted central, high p_{Tt}	Exponential	0.3	0.4
Converted rest, low p_{Tt}	4th order pol.	4.6	6.8
Converted rest, high p_{Tt}	Exponential	0.5	0.7
Converted transition	Exp. of 2nd order pol.	3.2	4.6
2-jets	Exponential	0.4	0.6

Discovery ...

A deviation from the expectation, i.e. the background only hypothesis ...

p-value: probability that a result is as or less compatible with expectation ...

Control region α (or size α) defines the significance level ...

If $p < \alpha$ hypothesis is rejected ...

Exclusion: $\alpha = 5\%$ [typical] i.e. exclusion with 95% CL ...

Test statistics Q for e.g. background only hypothesis

Discovery: $\alpha = 2.87 \times 10^{-7}$ [corresponds to 5 σ]



[e.g. Dulat et al., arXiv:1204.3851v2]

$X = \{X_i\}$	Set of measurements, e.g. number of observed event in specific phase space region
$g(X_i \mu, u)$	Probability density of X _i with model parameters µ with nuisance parameters v = {v _j }

Probability to measure the values X in an single experiment:

$$P = \prod_{i=1}^{N} P(X_i) = \prod_{i=1}^{N} g(X_i|\mu,\nu) dX_i = L(X|\mu,\nu) \prod_{i=1}^{N} dX_i,$$

with

$$P(X_i)$$
: probability to find the measurement within $X_i \pm dX_i$
 $L(X|\mu,\nu) = \prod_{i=1}^N g(X_i|\mu,\nu)$: likelihood, or joint probability density

independent input, e.g. lumi

[e.g. Dulat et al., arXiv:1204.3851v2]

Furthermore:

$$g(X_i \mid \mu, \nu) = e^{-\lambda_i} \lambda_i^{X_i} / X_i!$$

$$\lambda_i = \lambda_i(\mu, \nu)$$
$$\lambda(\mu, \nu) = \{\lambda_i(\mu, \nu)\}$$

 $\pi(\tilde{\nu}|\nu)$

Poisson distribution [as X_i very low compared to total event number]

Expectation value Depends on parameters μ and ν ...

Likelihood for nuisance parameters [includes uncertainties on values of $v = \{v_j\}$]

Thus:

$$\mathcal{L}(X|\mu,\nu) = \prod_{i} \frac{e^{-\lambda_i(\mu,\nu)}\lambda_i^{X_i}(\mu,\nu)}{X_i!} \times \prod_{j} \pi_j(\tilde{\nu}_j|\nu_j).$$

[e.g. Dulat et al., arXiv:1204.3851v2]

Likelihood ratio:

$$LR = \frac{\mathcal{L}(X|\mu,\nu)}{\mathcal{L}(X|\mu',\nu')}$$
Quantifies agreement of X
with prediction $\lambda(\mu,\nu)$ relativ to $\lambda(\mu',\nu')$

Profiled log-likelihood ratio:

Get's large if X disagrees with prediction ...

$$q_{\mu}(X) = \begin{cases} -2\log \frac{\mathcal{L}(X|\mu, \hat{\nu}_{\mu})}{\mathcal{L}(X|\mu', \hat{\nu}_{\mu'})} \ , \ \mu \geq \mu' \geq 0\\ 0 \qquad \qquad , \ \text{else} \end{cases}$$

with

$$\mathcal{L}(X|\mu, \hat{\nu}_{\mu}) \ge \mathcal{L}(X|\mu, \nu) \quad \forall \quad \nu.$$
$$\mathcal{L}(X|\mu', \hat{\nu}_{\mu'}) \ge \mathcal{L}(X|\mu, \nu) \quad \forall \quad \mu, \nu.$$

Maximized $\boldsymbol{\mathcal{L}}$ for a fixed $\boldsymbol{\mu}$... Globally maximized $\boldsymbol{\mathcal{L}}$...









[e.g. Dulat et al., arXiv:1204.3851v2]

Log-likelihood ratio LEP CL_S Method ... 0.5 f(q) Penalization in case of small sensitivity ... f(qlb) $CL_S(\mu) \equiv \frac{CL_{S+B}(\mu)}{1 - CL_B(\mu)}$ f(qls+b 0.4 $\mathsf{q}_{_{\mathrm{obs}}}$ 0.3 p_b 0.2 i.e.: Penalization by dividing by 1-CLB 0.1 Wide separation: $1-CL_B \approx 1$ no penalty p_{s+b} Overlap: strong penalty. By this one prevents exclusion of 0 models for which there is low sensitivity ... -8 -6 -2 -10 -4 0 q

[CL_s for mu=1]



 $[CL_s \text{ for } mu=1]$





