

The background model of *CELESTINA*

C. Delgado (CIEMAT, IAC)

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Abstract

We describe the background model implemented in *CELESTINA*.

1 Introduction

The current *CELESTINA* version provides two background estimation methods intended to produce skyplots. A first one is based in the projection of the events with a reconstructed incident direction located in the half of the camera where the anti-source is located in the other half of the camera. A second one is the so-called **background model** which does not require to know in advance the source and anti-source positions, but only provides a good estimate for point like sources. This later method is explained in this note, emphasizing the formal description of the first step of the method, which consists in a linear transformation as we show below.

2 Notation

In this note all the spatial coordinates are in the camera reference frame, where the coordinates of the camera center are given by 0, 0. The rate per unit area of the gamma source and the background at position x, y are denoted as $S(x, y)$ and $B(x, y)$ respectively. The total rate per unit area after all analysis cuts is denoted as $M(x, y) = S(x, y) + B(x, y)$. We use the notation $G(x, y|\sigma)$ for the value of a bi-dimensional gaussian distribution centered at 0, 0, of width σ evaluated at x, y . The PSF of the telescope is called σ_{PSF} . The unknown position of the gamma-ray source is x_S, y_S . We denote the convolution operation of a function $f(x, y)$ with a kernel $K(x, y)$ as

$$K(x, y) * f(x, y) = \int dx' dy' f(x', y') \times K(x' - x, y' - y) \quad (1)$$

3 Background estimate construction

In a first step, the background model estimate is given by

$$B^*(x, y) = 2 \times G(x, y|\sqrt{2}\sigma_B) * M(x, y) - G(x, y|\sigma_B) * M(x, y) \quad (2)$$

where σ_B is a model parameter and the superindice $*$ denotes estimate.

In a second step a further correction of the radial distortions on this estimate is performed by scaling it with a function depending of the distance to the camera center obtained by fitting

$$\Delta(r) = \frac{\int dx' dy' M(x', y') \delta(x'^2 + y'^2 - r^2)}{\int dx' dy' B^*(x', y') \delta(x'^2 + y'^2 - r^2)} \quad (3)$$

where δ is a Dirac delta distribution. The fit function is an expansion in a normalized radial basis function space with typically 10 dimensions, although for large statistics this number can be increase. The fit procedure is a robust Least Trimmed Squares fit using a minimal fraction of the 90% of the points where, which guarantees that the correction factor is not overestimated in the case a gamma-ray source is in the field of view. Figure 1 shows the typical fit to the function $\Delta(r)$ in a case without any signal, showing that it is close to 1 for $r < 2^\circ$, and the correction is maximum at camera center.

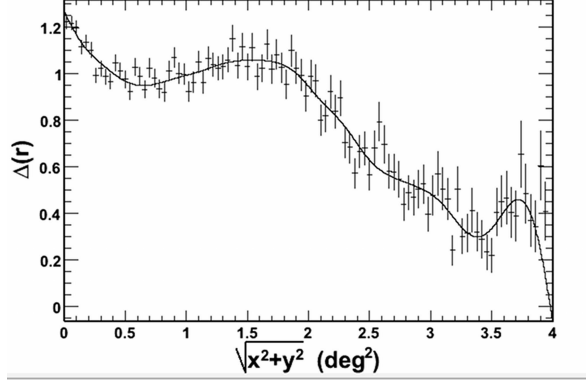


Figure 1: Fit of the correction factor $\Delta(r)$.

4 Formal justification

In this section we formally demonstrate that eq. 2 is a good estimate for the background. To this end we parameterize $B(x, y)$ as

$$B(x, y) = \sum_{i=1}^N b_i \times G(x - x_i, y - y_i | \sigma_i^B) \quad (4)$$

where N can be infinity and all $b_i > 0$. Notice that this parameterization is always possible since the set of all gaussians form an overcomplete basis. In addition we assume that the following approximation holds

$$S(x, y) = s \times G(x - x_S, y - y_S | \sigma_{PSF}) \quad (5)$$

where s is the number of gamma-rays detected from the source. This approximation assumes that the source is point-like and that no de-rotation is performed, which is the worst case for background estimation because the signal is concentrated in single point. Nevertheless all the results obtained under this approximation hold if $S(x, y)$ can be parameterized as $B(x, y)$ in equation 4 with the condition that the corresponding x_{S_i}, y_{S_i} have a zero-measure distribution due to the fact that the transformation given by 2 is a linear one.

With these assumption, the estimate given by equation 2 is obtained by sampling from the distribution given by

$$\begin{aligned} B^*(x, y) = & 2 \times G(\sqrt{2}\sigma_B) * (B(x, y) + S(x, y)) - G(\sigma_B) * (B(x, y) + S(x, y)) = \\ & \left(2 \sum_{i=1}^N b_i \times G\left(x - x_i, y - y_i | \sqrt{(\sigma_i^B)^2 + 2\sigma_B^2}\right) - b_i \times G\left(x - x_i, y - y_i | \sqrt{(\sigma_i^B)^2 + \sigma_B^2}\right) \right) + \\ & \left(2s \times G\left(x - x_S, y - y_S | \sqrt{\sigma_{PSF}^2 + 2\sigma_B^2}\right) - s \times G\left(x - x_S, y - y_S | \sqrt{\sigma_{PSF}^2 + \sigma_B^2}\right) \right) \end{aligned} \quad (6)$$

By choosing a value of σ_B such that

$$(\sigma_i^B)^2 \gg \sigma_B^2 \quad (7)$$

for any i , and thus

$$\sqrt{(\sigma_i^B)^2 + \sigma_B^2} = \sigma_i^B \times \left(1 + O\left(\frac{\sigma_B^2}{(\sigma_i^B)^2}\right)\right) \quad (8)$$

In consequence we have that eq. 6 can be written as

$$\begin{aligned} B^*(x, y) \simeq & \underbrace{\sum_{i=1}^N b_i \times G(x - x_i, y - y_i | \sigma_i^B)}_{B(x, y)} + O\left(\frac{\sigma_B^2}{(\sigma_i^B)^2}\right) + \\ & \underbrace{+ 2s \times G\left(x - x_S, y - y_S | \sqrt{\sigma_{PSF}^2 + 2\sigma_B^2}\right) - s \times G\left(x - x_S, y - y_S | \sqrt{\sigma_{PSF}^2 + \sigma_B^2}\right)}_{S^*(x, y)} \end{aligned} \quad (9)$$

where we have defined the function S^* as the addition of the two last two terms on the right hand side of the equation, and the term $O(\epsilon)$ means a function of order ϵ .

With this, the estimate signal given eq. 9 is

$$M(x, y) - B^*(x, y) = S(x, y) - S^*(x, y) + O\left(\frac{\sigma_B^2}{(\sigma_i^B)^2}\right) \quad (10)$$

A plot of the function $S(x, y) - S^*(x, y)$ scaled such that $S(x_S, y_S) = 1$ is shown in figure 2. It shows that for σ_B large enough it is a good approximation for $S(x, y)$. Figure 3 shows these two functions for the case $\sigma_B = \sigma_{PSF}$, which is the one chosen for the MAGIC-I background model. Moreover, the application of the multiplicative correction factor defined in eq. 3, by computing $M(x, y) - \Delta(\sqrt{x^2 + y^2}) \times B^*(x, y)$ instead of eq. 10 guarantees a better approximation if no gamma-rays source is in the field of view by construction, and in the case of the presence of one it applies a further suppression factor to $S^*(x, y)$ due to the Least Trimmed Squares fit procedure used.

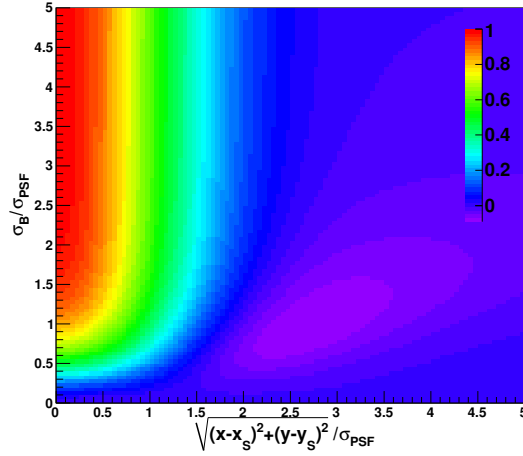


Figure 2: Plot of the function $(S(x, y) - S^*(x, y))2\pi/s$ as a function to the distance to x_S, y_S and the ratio σ_B/σ_{PSF} .

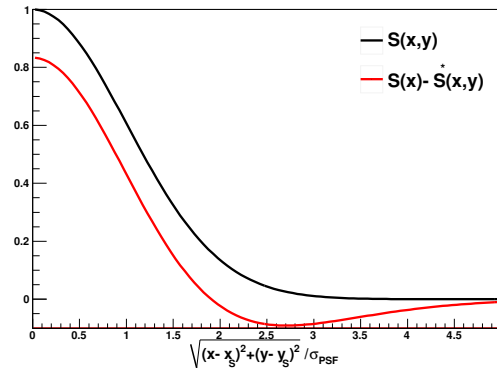


Figure 3: Plot of the functions $S^*(x, y) = (S(x, y) - S^*(x, y))2\pi/s$ and $S(x, y)2\pi/s$ as a function to the distance to x_S, y_S for the parameter condition $\sigma_B = \sigma_{PSF}$.

5 Conclusion

As a conclusion, we can say that $M(x, y) - \Delta(\sqrt{x^2 + y^2}) \times B^*(x, y)$ is an approximation for $S(x, y)$, that in the case of the parameter σ_B being equal to σ_{PSF} underestimates the true value by less than 20% of $S(x_S, y_S)$ close to the point x_S, y_S , and far from it it approximates to the true value $B(x, y)$, this approximation being better than that of the order of σ_B^2/σ_i^B due to the $\Delta(r)$ correction factor. This makes this background approximation useful to search for serendipitous sources in the field of view of MAGIC-I, although it is not well suited to provide background estimates useful for flux computations. It remains to show that the condition given by eq.

7 holds. We have checked this experimentally by explicitly searching for a parameterization as eq. 4, finding that taking $\sigma_i \simeq 0.3^\circ$ for all i such construction is possible giving good residuals. Taking into account that $\sigma_{PSF} \simeq 0.1^\circ$ above 150 GeV, the value of $\sigma_B = \sigma_{PSF}$ is a good compromise between the condition of eq. 7 and the necessity of taking a large enough value for the parameter as illustrated by figure 2.