

# ESTIMATION OF THE ERRORS INVOLVED IN THE INTENSITY MEASUREMENT OF LOW S/N RATIO EMISSION LINES

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We present a detailed analysis on the extent of the errors in the intensity measurements of low signal-to-noise narrow emission lines. Our first goal is to determine a model for the probability distribution function (p.d.f.) associated with the measured intensities of a line characterized by its signal-to-noise ratio. Our final purpose is to provide an error domain: - for the measured signal-to-noise ratio of an observed line; - for the ratio of two lines in terms of their signal-to-noise ratios, and eventually (with the knowledge of the noise energy) to get errors on the corresponding intensities and intensity ratios.

To reproduce a real emission line intensity measurement process, we have designed a program based on a Monte-Carlo simulation procedure. For the purposes of this simulation, the knowledge of the true line parameters (the Intensity - imposed by the true signal-to-noise  $(S/N)_{\text{true}}$ , the position, and width) and the noise characteristics were necessary. Each spectrum was modeled by the sum of an emission line plus a continuum of a known constant level, this sum being considered as the mean of a certain stochastic process. The line was a Gaussian profile. To this model was added a non-correlated noise ("white noise") following the Student distribution.

Our simulation was divided in two main parts: the first, called detection, tells if a line is suspected to be present in a specific segment of the spectra; the second tries to measure the detected line, that is, to give estimates of the intensity, position and width of the suspected line. Once a line is suspected in one of the samples, the program proceeds to a fitting algorithm. The line model is also a Gaussian, the parameters of which are known only to lie within reasonable bounds. The adjustment is made through a standard  $X^2$  minimizing routine. Once the minimum is attained a  $X^2$  rejection test is performed on the residuals of the fit. If the test is satisfied, a Gaussian line is detected.

We made at least 1000 simulations for each  $(S/N)_{\text{true}}$ . For each signal-to-noise  $(S/N)_{\text{true}}$ , we calculated the intensity of each line in the detected lines sample. We call it "observed" intensity and denote it by  $S_{\text{obs}}$ . These intensities are then normalized to the known true intensity,  $S_{\text{true}}$ , and we call this normalised intensity,  $i$ , i.e.,  $i = S_{\text{obs}}/S_{\text{true}}$ . Fits made for each  $(S/N)_{\text{true}}$  led us to consider the log-normal distribution as a possible model for  $S_{\text{obs}}/S_{\text{true}}$ , which proved to give very reasonable results. We computed, for each normalised sample, a few statistics, the "true" signal-to-noise ratio, the mean of the sample (normalised to the true intensity), its variance and the square-root of the mean quadratic error calculated relative to one (the true normalised intensity value). Our results show clearly that for  $(S/N)_{\text{true}}$  up to three, there is a strong bias ( $> 50\%$ ) towards overestimation of the values of the "observed" intensities.

In practice, the observer’s interest is to determine from the knowledge of  $(S/N)_{obs}$  a plausible range where  $(S/N)_{true}$  may be found. In other words, one would like to know what are the limits which contain  $S_{true}$  with probability of  $\gamma$ . These error bars can be easily calculated since we know for each value of  $(S/N)_{true}$  the distribution followed by  $(S/N)_{obs}$ . By a change of variable it is easy to derive the distribution followed by  $(S/N)_{obs}$ , which is also a log-normal. This parameter has the advantage of being a non-dimensional universal one. By integrating the  $(S/N)_{obs}$  p.d.f. corresponding to each  $(S/N)_{true}$  up to a confidence level  $\gamma$  we obtain the chart presented in Fig. 1, where the continuous thick line delimits the confidence intervals for  $\gamma = 0.683$  (or one  $\sigma$ ), and the thin line, the one for  $\gamma = 0.954$  (or two  $\sigma$ ). This chart was constructed horizontally, but must be used vertically with the following reasoning: an observer measures a line of  $(S/N)_{obs}$  and obtains vertically the range on the  $(S/N)_{true}$  values determined with confidence level  $\gamma$ . This chart shows clearly that, for example, at the one  $\sigma$  confidence level (or  $\gamma = 0.683$ ), a line measured with  $(S/N)_{obs} = 4$  is compatible with  $0 \leq (S/N)_{obs} \leq 5$ . One cannot then rule out the possibility that a noise fluctuation was mistaken for a line.

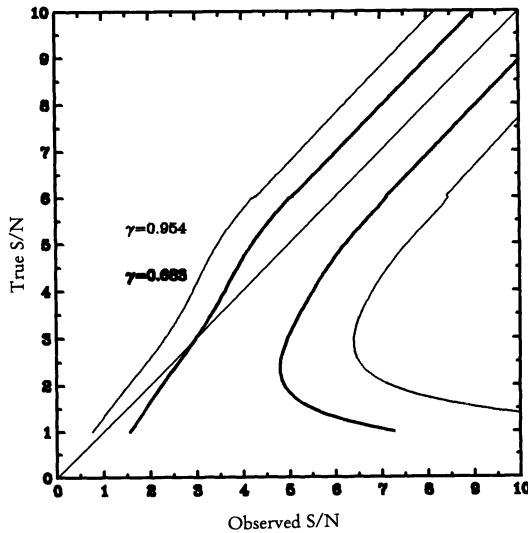


Fig. 1. Chart for the interval estimation of the  $(S/N)_{true}$ , given an observed value  $(S/N)_{obs}$ .

Furthermore, we calculated that a ratio of measured intensities or signal-to-noise ratios also follows a log-normal distribution and as such is subject to strong biases. Additionally, we determined the error domain on the ratio of a signal-to-noise ratios, which can be converted in an intensities ratio with the knowledge of the correspondent noise energy ratio.