Improved Correlation between the Variability and Peak Luminosity of Gamma-Ray Bursts

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ABSTRACT
A new procedure for smoothing a gamma-ray burst (GRB) lightcurve and calculating its variability is presented. Applying the procedure to a sample of 25 long GRBs, we have obtained a very tight correlation between the variability and the peak luminosity. The only significant outlier in the sample is GRB 030329. With this outlier excluded, the data scatter is reduced by a factor of $\sim 3$ compared to that of Guidorzi et al. (2005), measured by the deviation of fit. Possible causes for the outlier are discussed.

Key words: gamma-rays: bursts – methods: data analysis.

1 INTRODUCTION
Despite many exciting progresses in observations, the nature of Gamma-Ray Bursts (GRBs) remains to be a big puzzle [see van Paradijs et al. (2000) and Piran (2004) for recent reviews]. In such a situation, it is extremely important to identify some good correlations between the apparent (easy to measure or calculate) and intrinsic properties of GRBs. Several such correlations have indeed been found. For example, an anti-correlation between the the peak luminosity and spectral lag of GRBs has been found by Norris et al. (2000), and a correlation between the peak luminosity and the variability of GRB lightcurves has been found by Fenimore & Ramirez-Ruiz (2000, hereafter FR00) and Reichart et al. (2001, hereafter R01). A correlation between the total isotropic energy and the peak energy of the spectrum (Amati et al. 2002), or the collimation-corrected total energy and the peak energy of the spectrum (Ghirlanda et al. 2004), has also been discovered.

Recently, Guidorzi et al. (2005, hereafter G05) tested the correlation between the variability and peak luminosity of GRBs, using an expanded sample of 32 GRBs with measured redshifts. The definitions of the variability and the peak luminosity are the same, but the size of the GRB sample of G05 is about three times bigger than that of R01. The existence of a correlation was confirmed, but the scatter in the correlation is significantly larger than that found by R01 (see, however, Reichart & Nysewander 2005). Although the issue is in debate (Guidorzi 2005; Reichart & Nysewander 2005, Reichart 2005), it is clear that with the definition of variability given by R01, the correlation between the variability and the peak luminosity is not tight.

In this paper, we present a new definition for the variability of GRB lightcurves. We then apply it to a sample of 25 long duration GRBs with measured redshifts, whose data are publicly available. We show that, with the new definition of the variability, the correlation between the variability and the peak luminosity of GRBs is significantly improved: the data scatter is considerably reduced.

2 NEW DEFINITION OF THE VARIABILITY
To measure the variability of a lightcurve, first we must define a reference lightcurve. The reference lightcurve should be sufficiently smoother than the original raw lightcurve. Since an ultimate model for GRBs does not exist yet, there is no first principle guiding us in choosing a reference lightcurve. What people usually do is to smooth the raw lightcurve with a linear “box car” filter (or moving window), which smoothes the lightcurve with linear average (FR00; R01).

Here we use a Savitzky-Golay filter (Press et al. 2002) to smooth a lightcurve. The Savitzky-Golay filter is a more general and more powerful approach for smoothing noisy data than the linear box-car filter. The basic idea of Savitzky-Golay filtering is to approximate the underlying function (i.e., the reference lightcurve) within the moving window by a polynomial of higher order. An advantage of the Savitzky-Golay filter to the linear filter is that the former preserves high moments while the latter does not.

A Savitzky-Golay filter is specified by three numbers: the order of the polynomial ($m$), the number of points used to the left of a data point ($n_L$), and the number of points used to the right of a data point ($n_R$). To apply the Savitzky-Golay filter, the data must be binned with constant spacing. For more details see Press et al. (2002).

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We use a third order Savitzky-Golay filter. That is, we set $m = 3$. The width of the filter, i.e. $n_p \equiv n_l + n_R + 1$, is determined with the approach of R01: $n_p$ is set to be equal to the number of data points corresponding to a timescale $T_f$—the time spanned by the brightest 100% of the total counts above the background (for details see R01). It turns out that $f = 0.5$ most suits our purpose (rather than the $f = 0.45$ used by R01 and G05; see Section 5). So, throughout the paper we use $n_p$ determined by $T_{0.5}$.

We then define $n_l = \lfloor (n_p - 1)/2 \rfloor$, and $n_R = n_p - n_l - 1$. If $n_p$ is odd, we have $n_R = n_l$, i.e. the filter is symmetric about the point to smooth. If $n_p$ is even, we have $n_R = n_l + 1$, the filter is asymmetric. In this case, we smooth the lightcurve twice: first we use the $n_l$ and $n_R$ defined above, then we switch $n_l$ and $n_R$. The results are then averaged.

Suppose we have obtained a reference lightcurve by applying the Savitzky-Golay filter to the raw lightcurve. Let us denote the count of the raw data in the $i$-th time bin by $C_i$, the count given by the reference lightcurve by $Y_i$. The total squared deviation of the raw lightcurve from the reference lightcurve is then $\sum_{i=1}^{N_{\text{bin}}} (C_i - Y_i)^2$, where $N_{\text{bin}}$ is the total number of bins to be summed. We obtain the intrinsic squared deviation by subtracting the Poisson noise $N_{\text{Poisson}}$

$$\Delta C^2 = W \sum_{i=1}^{N_{\text{bin}}} (C_i - Y_i)^2 - N_{\text{Poisson}},$$

(1)

where $W \equiv n_p/(n_p - m - 1)$ is a statistical weight accounting for the fact that among the $n_p$ data points only $n_p - m - 1$ are statistically independent. The inclusion of $W$ allows us to apply the variability definition to any lightcurve with $n_p > m + 1$.

The summation in equation (1) is from time $t_1$ (corresponding to $i = 1$) to time $t_2$ (corresponding to $i = N_{\text{bin}}$), enclosing a major part of the lightcurve. Following FR00 and R01, we define $t_1$ to be the start of $T_{90}$, $t_2$ to be the end of $T_{90}$, where $T_{90}$ is the time during which the cumulative counts of the GRB increase from 5% to 95% above background [Kouveliotou et al.1993].

The Poisson noise is calculated by

$$N_{\text{Poisson}} = \sum_{i=1}^{N_{\text{bin}}} [C_i + (\xi - 1)C_{bg,i}],$$

(2)

where $C_{bg,i}$ is the background, the factor $\xi$ is the ratio of the background fluctuation to the Poisson noise of the background (given by the reduced $\chi^2$ of the background fit). For GRBs detected by Swift we found that $\xi$ is often significantly larger than unity, indicating that the background fluctuation is quite non-Poissonian.

The variability of the lightcurve is defined by the normalized squared deviation. We find that the following definition leads to the tightest correlation between the variability and the peak luminosity

$$V = \frac{\Delta C^2}{(N_{\text{bin}} - 1) C_{\max}^2},$$

(3)

where $\Delta C^2/(N_{\text{bin}} - 1)$ is the average of the squared deviation, $C_{\max}$ is the net peak count (i.e., the background is subtracted).

Our variability is defined in the observer’s frame, so that the information of GRB redshift is not needed. This not only makes the computation of variability simple, but also eliminates an uncertainty arising from the assumption about the dependence of lightcurve variability on photon energy. The effect of GRB redshift was considered by R01 and FR00 who defined their variabilities in the GRB frame, but the dependence of variability on redshift turned out to be very weak.

As we applied our smoothing procedure to the GRBs in the sample described in the next section, we found that the tightest correlation between the variability and the peak luminosity is obtained if we iteratively apply the Savitzky-Golay filter $N_{\text{iter}}$ times, where $N_{\text{iter}}$ is the integer closest to $T_{90}/T_f$ (i.e., $N_{\text{iter}}$ is roughly the number of moving windows contained in $T_{90}$).

In summary, our definition of variability differs from that of FR00 and R01 in the following aspects: (1) Our variability is defined in the observer’s frame, while the variabilities of FR00 and R01 are defined in the GRB’s frame. (2) We normalize the average of the squared deviation by the squared peak count (the same as FR00), while R01 normalize the total squared deviation by the sum of squared counts. (3) FR00 and R01 use a linear box-car filter, while we use a nonlinear Savitzky-Golay filter.

## 3 DESCRIPTION OF THE SAMPLE

Our GRB sample contains 19 GRBs from the sample in G05, and 6 more GRBs detected by FREGATE/HETE-2 and BAT/Swift. So, the total number of GRBs in our sample is 25. They are listed in Table 1 with measured redshift, calculated isotropic-equivalent peak luminosity, calculated variability, and the number of iterations in applying the Savitzky-Golay filter. We have chosen to use the GRBs with data available publicly, which include GRBs detected by BATSE/CGRO, HETE-2, and BAT/Swift.\(^{1}\) To obtain a reliable calculation of variability, we have only selected GRBs with more than 30% of total counts above the 3-σ of background. As a result, those GRBs with too low signal-to-noise ratios are not included in our sample.

The 19 GRBs from G05 are: 970508, 971214, 980425, 980703, 990123, 990506, 990510, 991216, 000131, 010921, 020124, 020813, 030328, 030329, 041006, 050401, 050505, 050525, and 050603. Their peak luminosities are taken from the same paper. The 13 GRBs in G05 are: 030115a, 030528, 050408 (from HETE-2), 050802, 050803, and 050820a (from Swift). Their peak luminosities are calculated in exactly the same

\(^{1}\) Although the lightcurve data of some GRBs detected by Konus/Wind are available from [http://gcn.gsfc.nasa.gov/gcn/](http://gcn.gsfc.nasa.gov/gcn/), we have chosen to not use them since the data are highly incomplete in the part of pre- and post-burst.
Thus, for GRB 050820a, the variability and peak luminosity
were also measured. The variability of the event files available in the public data archive.

Although Golenetskii et al. (2003), Pal’shin et al. (2004), and Cummings et al. (2003) have reported the detection of a second, larger episode of emission from GRB 050820a (by both Konus/Wind and BAT/Swift), only the data for the first episode from Swift are available to us. Thus, for GRB 050820a, the variability and peak luminosity listed in Table 1 refer to its first episode (of duration ∼ 30 sec).

The time bin of each GRB lightcurve is as follow. For GRBs detected by BATSE, we used the spectral fit provided by Sakamoto et al. (2003) and the HETE website. For GRBs detected by Swift, peak spectra are extracted from their event files, then fitted with a power law which is sufficient for all cases (in the energy range 15-350 keV).

As we analyze the GRB lightcurves we use the total counts in the energy range specified as follow: 25-300 keV for GRBs detected by BATSE, 30-400 keV for those detected by HETE-2, and 25-350 keV for those detected by BAT/Swift. We have tried to make the range of energy for GRBs detected by different instruments have largest overlap.

Finally, we remark that for the GRBs detected by BATSE, the lightcurve data in the pre- and post-burst sections were divided equally into 64-ms time bins from their original 1.024-sec time bins. When we calculate the Poisson
Figure 1. Peak luminosity versus variability of 25 GRBs. The solid straight line is the least-$\chi^2$ linear fit to the data, with GRBs 980425, 030528, and 030329 excluded (see the text): $\log L = 3.25 \log V + 59.42$ ($L$ in units of erg/sec). The $\chi^2$/dof = 38.58/20 = 1.93. The two dashed lines mark the 1-$\sigma$ deviation of the fit. (Filled circles: GRBs detected by BATSE; Squares: GRBs detected by HETE-2; Triangles: GRBs detected by BAT/Swift.)

noise in those sections, we multiply the result by a factor of 0.0625 (= 0.064/1.024) to take into account the reduction in the Poisson noise arising from the change in time bin.

4 RESULTS

We have applied the smoothing procedure described in Section 2 to our GRB sample. The obtained variabilities and errors, as well as the number of iterations in applying the Savitzky-Golay filter, are listed in Table 1. The errors of variability are principally caused by photon noises. However, the uncertainties arising from changing $N_{\text{iter}}$ (see Section 2) by ±1 are also taken into account: we find the changes in the variability for $N_{\text{iter}} = 1$ and add in quadrature the maximum to the statistical error.

The peak luminosity versus variability is shown in Fig. 1. Clearly a very tight correlation between the two quantities exists, with only one prominent outlier: GRB 030329. Given the large error in its variability, it is unclear if GRB 980425—which is famous for its smallest redshift, lowest peak luminosity and least total energy, and association with SN 1998bw (Galama et al. 1998)—is off trend as in other kind of correlations (e.g., the total energy-peak energy correlation). In fact, the error in variability is larger than or about the same value as the variability itself for 980425 and 030528 (see Table 1).

We have made a least-$\chi^2$ linear fit to $\log L - \log V$, where $L$ is the peak luminosity in erg/sec. To take into account both the errors in $L$ and $V$, we have made use of the program fitexy in Press et al. (2002). The asymmetric errors in $\log L$ and $\log V$ are averaged. That is, we take

$$\sigma_{\log q} = \frac{1}{2} \left[ \log (q + \sigma_q) - \log (q - \sigma_q) \right],$$

where $q = L$ or $V$. Since the errors in $\log V$ are infinite for 980425 and 030528, these two GRBs do not contribute to the total $\chi^2$. GRB 030329 was not considered due to its large offset. With 980425, 030329, and 030528 excluded, the total number of GRBs in the final sample to fit is 22. We
have obtained the following result (the solid straight line in Fig. 1)
\[
\log L = a \log V + b ,
\]
where
\[
a = 3.25 \pm 0.26 , \quad b = 59.42 \pm 0.53 .
\]

The reduced chi-square \( \chi^2 \equiv \chi^2 / \text{dof} = 1.93 \), where the degree of freedom \( \text{dof} = 22 - 2 = 20 \). The smallness of \( \chi^2 \) indicates a very tight correlation between the peak luminosity and variability, for details see Section 5.

The two dashed lines in Fig. 1 mark the 1-σ width of the fit, measuring the scatter of data and defined by the deviation of fit (see Section 5). However, this should not be mixed with the error in the predicted log \( L \), which we discuss below.

In Fig. 2 we plot the confidence intervals in the \( a - b \) plane. The best fit values of \( a \) and \( b \), which correspond to the minimum of \( \chi^2 \), are indicated by the cross sign in the figure. The three contours enclose the regions containing, respectively, 68.3%, 90%, and 95.4% of normally distributed data. The highly elongated shape of the contours indicates that \( a \) and \( b \) are highly correlated.

For any measured value of \( \log V \), equation (5) predicts a value of \( \log L \). The error in the predicted \( \log L \) comes from two sources: the error in \( \log V \), and the uncertainties in \( a \) and \( b \)

\[
\sigma_{\log L} = \sqrt{a^2 \sigma_{\log V}^2 + \sigma_{ab}^2} .
\]

Generally, the errors are asymmetric in the upward and downward directions. Let us denote by \( \sigma_{ab,1} \) the upward error due to uncertainties in \( a \) and \( b \), by \( \sigma_{ab,2} \) the downward error. That is, the predicted log luminosity is \( \log L = \log L^{05} + \sigma_{ab,1} \), if \( \sigma_{\log V} = 0 \). In Fig. 3 we show the numerically calculated values of \( \sigma_{ab,1} \) (filled circles) and \( \sigma_{ab,2} \) (open circles) versus the predicted log luminosity. From Fig. 3 we find that \( \delta b \approx 2.05 \delta a \) when \( \delta a \) and \( \delta b \) are not close to zero, which leads to \( \sigma_{ab} \sim (\log V + 2.05) \delta a \). Thus, \( \sigma_{ab} \) is minimized at \( \log V \approx -2.05 \), i.e. at \( \log L \approx 52.8 \), consistent with the numerical results.

For the convenience of application of the \( \log L - \log V \) relation \( 5 \) to the prediction of GRB’s peak luminosity, we have attempted to approximate the numerical results for \( \sigma_{ab} \) by several analytic formulæ. The results are summarized in Appendix A, and represented by the solid lines in Fig. 3.

Now we apply the above results to GRB 030329 to check how far it deviates from the \( \log L - \log V \) relation. From Table 1, for GRB 030329 we have \( V = 0.0070 \) and \( \sigma_V = 0.0010 \). Submitting \( \log V = -2.15 \) into equation \( 5 \), we obtain a predicted log luminosity for GRB 030329: \( \log L = 52.4 \). Since GRB 030329 lies below the straight line in Fig. 3 the relevant error in the predicted log luminosity is determined by \( \sigma_{ab,2} \) and \( \sigma_{\log V} \). We have \( \sigma_{ab,2} \approx 0.08 \). Then, by equation \( 5 \), the total error in the predicted luminosity is \( \sigma_{\log L} \approx 0.23 \). From Table 1, the measured luminosity of GRB 030329 is \( \log L \approx 50.8 \), with an upward measured error \( \sigma_{\log L} \approx 0.08 \) that is smaller than the error in the predicted log luminosity. So, GRB 030329 deviates from the correlation (the straight line in Fig. 1) by \( (52.4 - 50.8)/0.23 \approx 7 \sigma \).

5 COMPARISON TO GUIDORZI ET AL.’S RESULTS

To make a quantitative comparison of our results to that of G05, we need a parameter that measures the scatter of data points around the best fit model. The parameter should not be sensitive to the variation in data errors caused by different definitions of variability. Reichard (2001) has provided such a parameter based on the maximum-likelihood method, which
Definition of the variance of fit. The thick solid curve is the fit $y = f(x)$. The $y$-component of the variance of fit, $s_y^2$, is defined by equation (11). The two dashed curves, called “the 1-$\sigma$ width” of $y = f(x)$, are defined by $y = f(x) \pm s_y$. The $x$-component of the variance of fit, $s_x^2$, is related to $s_y^2$ by equation (12). The scatter of data points around $y = f(x)$ is measured by $s = s_x s_y / \sqrt{s_x^2 + s_y^2}$, the length of the straight line $OC$ that is perpendicular to the straight line $AB$. The straight line $AB$ locally approximates the curve $y = f(x)$ when $s_y$ is sufficiently small or the curvature of the curve is negligible.

is called the “sample variance”, or “slop”. However, here we choose to use the “variance of fit” defined with the approach of Bevington & Robinson (1992), which is much less sophisticated but enough for our current purpose.

Suppose a data set $\{x_i, y_i\}$, with errors $\{\sigma_{x,i}, \sigma_{y,i}\}$, is fitted by a model $y = f(x)$. The fit minimizes the $\chi^2$ defined by

$$\chi^2 \equiv \sum_{i=1}^{N} \frac{1}{\sigma_i^2} [y_i - f(x_i)]^2,$$

where $N$ is the total number of data points, and

$$\sigma_i^2 = \sigma_{x,i}^2 + \left(\frac{df}{dx}\right)^2_{x=x_i} \sigma_{y,i}^2.$$

The reduced $\chi^2$, i.e., the $\chi^2$ per degree of freedom, is

$$\chi^2_r = \chi^2 / \text{dof} = \chi^2 / (N - p),$$

where $p$ is the number of parameters in the function $f(x)$.

The variance of the fit, measured along the $y$-direction, is defined by Bevington & Robinson (1992)

$$s_y^2 = \frac{\langle \sigma_i^2 \rangle}{N - p} \sum_{i=1}^{N} \frac{1}{\sigma_i^2} [y_i - f(x_i)]^2,$$

where $\langle \sigma_i^2 \rangle$ is the weighted average of the individual data variance

$$\langle \sigma_i^2 \rangle \equiv \left[ \frac{1}{N} \sum_{i=1}^{N} \frac{1}{\sigma_i^2} \right]^{-1}.$$

The variance of fit measures the spread of data around the model. We denote the variance of fit by $s^2$ (then $s$ is the deviation of fit), to distinguish it from the data variance $\sigma^2$ which represents the error in the measurement of each data point.

The appearance of $\langle \sigma_i^2 \rangle$ in the numerator and $\sigma_y^2$ in the denominator in equation (11) results that the variance of fit is not sensitive to the variation in data variance.

By the definition of $\chi^2_r$, equation (11) can be recast into a simpler form

$$s_y^2 = (\sigma_y^2)\chi^2_r.$$

We define the variance of the fit measured along the $x$-direction by

$$s_x^2 = s_y^2 \left(\frac{df}{dx}\right)^2,$$

where $df/dx$ is the slope of the curve $y = f(x)$. Note, by definitions, $s_y$ is always a constant, but $s_x$ is a function of $x$ unless $f(x)$ is a linear function of $x$.

We measure the scatter of data points around the curve $y = f(x)$ by the variance of fit along a direction locally perpendicular to the curve, which is (see Fig. 4)

$$s^2 = \frac{s_y^2 s_x^2}{s_y^2 + s_x^2} = \frac{s_y^2}{1 + (df/dx)^2}.$$

For the problem in this paper, we have $x = \log V$, $y = \log L$, and $f(x) = ax + b$. Thus, $df/dx = a$ is a constant, and $s_x = s_y / a$, $\sigma_i^2 = \sigma_{y,i}^2 + a^2\sigma_{x,i}^2$. The deviation of fit is then

$$s = \frac{s_y}{\sqrt{1 + a^2}}.$$

With the above tools, we are ready to compare our results to that of G05 quantitatively. To make the comparison well defined, we analyze the overlapping sample which contains 19 GRBs (see Section 3). In Fig. 5 we plot the variability with our definition versus that with R01’s definition (the latter was copied from G05). The two variabilities are highly correlated, but the scatter is large. This large scatter is necessary for obtaining a better correlation between the peak luminosity and the variability of GRBs.

The largest scatter in Fig. 5 occurs at GRB 980425, for which our background fit leads to $T_{90} = 60.29$ sec, in contrast to the 34.88 sec reported in the BATSE web page. If we adopted $T_{90} = 34.88$ and $\xi = 1$ ($\xi$ is the ratio of the background fluctuation to the Poisson noise, see Section 2), we would obtain $V = 0.00053 \pm 0.00057$ for 980425. This value of $V$ is only slightly larger than that listed in Table 1, so 980425 remains to have the largest scatter in Fig. 5. With our definition of variability, GRB 980425 is much less offset from the variability-luminosity correlation.

Figure 5 also shows that the values of the variability with our definition are systematically smaller than that with R01’s definition, as indicated by the straight line defined by “our variability = R01’s variability divided by 15”. This is mainly caused by the different normalization in the two definitions. In addition, our variability generally has a larger
error bar than that of R01, caused by the fact that we normalize our variability by peak count, which suffers larger photon noise than total counts; and we have considered the error arising from changing $N_{\text{iter}}$ by $\pm 1$.

Now, we make a linear fit to $\log L - \log V$ for the overlapping sample, with our definition of variability. For the reasons explained in Section 4, we exclude GRBs 980425 and 030329 during the fit. Thus, the total number of GRBs to fit is $N = 17$. We obtain $\log L = 3.10 \log V + 59.13$, and $\chi^2/\text{dof} = 2.43$ where $\text{dof} = 15$. The weighted average of the individual data variance is $\langle \sigma_i^2 \rangle = 0.02715$. Thus, for the fit with our definition of variability, we have $s_y = 0.257$, $s_x = 0.0829$, and $s = 0.079$.

Next, we make a linear fit to $\log L - \log V$ for the overlapping sample, with R01’s definition of variability. Again, GRBs 980425 and 030329 are excluded, to make the GRB members in the sample remain the same. We obtain $\log L = 1.77 \log V + 54.06$, and $\chi^2/\text{dof} = 19.89$. The weighted average of the individual data variance is $\langle \sigma_i^2 \rangle = 0.009240$. Thus, for the fit with R01’s definition of variability, we have $s_y = 0.429$, $s_x = 0.2417$, and $s = 0.211$.

The scatter of data points around the fitted $\log L - \log V$ is measured by $s$. Thus, according to the above numbers, for the overlapping sample the scatter in our $\log L - \log V$ relation is smaller than that of G05 by a factor of $0.211/0.079 = 2.7$, although the weighted average of the individual data error ($\equiv \langle \sigma_i^2 \rangle^{1/2}$) with our definition of variability is larger than that with R01’s definition of variability by a factor of $(0.02715/0.009240)^{1/2} = 1.7$.

The above results, for the overlapping sample excluding 980425 and 030329, are summarized in Fig. 6, where filled circles and thick lines represent the results with our data, open circles and thin lines represent the results with G05’s data. The luminosities in the two set of data are the same, so each GRB is represented by a horizontal line in Fig. 6.

**Figure 5.** The variability with our definition versus the variability with Reichart et al.’s definition for 19 GRBs. The straight line is defined by the equation “our variability = Reichart et al.’s variability divided by 15.”

**Figure 6.** Comparison of our results (filled circles and thick lines) to that of G05 (open circles and thin lines). Straight lines represent the least-$\chi^2$ fit and the 1-$\sigma$ width. (Our variability has been multiplied by a factor 15.)
However, the variabilities in the two set of data differ, so we see the offset along the horizontal direction of each pair of filled circle-open circle. The solid straight lines are least-$\chi^2$ fits to each data set, and the dashed straight lines mark the “1-$\sigma$ width” of the fits. We have set the unitary length of the horizontal and vertical axes to be the same, so that for each data set $s$ is just the half-width of the region bounded by the two dashed lines that you see. The figure clearly shows that the scatter of our data set is much narrower than that of G05’s data set.

With the above defined scatter parameter $s$, we can also check how sensitive our results are to the parameters in our definition of variability. As mentioned in Section 2, we have chosen $T_f/T_{0.5}$ and $N_{\text{iter}}$ to be the integer closest to the ratio $T_{0.5}/T_f$. With this choice, for the GRB sample described in Section 3 (excluding 980425, 030329, and 030528), we obtained $\chi^2/\text{dof} = 1.93$ and $s = 0.078$ (Fig. 4) where the two dashed lines are obtained by shifting the solid line upward/downward by $s_y = 0.265$). If we followed R01 and G04 to choose $T_f/T_{0.5}$, we would obtain $\chi^2/\text{dof} = 2.53$ and $s = 0.087$ in the linear fit to $\log L - \log V$. A slightly larger value of $s$ indicates that the correlation is slightly looser.

To test how sensitive the results are to the choice of $N_{\text{iter}}$, let us redefine $N_{\text{iter}}$ by the integer closest to $T_{0.5}/\alpha T_{0.5}$, where $\alpha$ is a parameter that is set to be the same for all GRBs in the sample. When $\alpha = 1$, we return to the definition that we have adopted. We now let $\alpha$ vary from 0.2 to 2, and calculate the corresponding $s$ in the least-$\chi^2$ fit to $\log L - \log V$. The calculated $s$ versus $\alpha$ are presented in Fig. 4, which shows that the results are stable to the choice of $N_{\text{iter}}$.

That is, the variance of fit does not change drastically with a small change in $\alpha$. The figure also shows that $\alpha = 1$ is around the minimum of $s(\alpha)$.

### 6 CONCLUSIONS AND DISCUSSION

We have presented a new definition for the variability of GRB lightcurves. The new variability is defined in the observer’s frame, and the lightcurve is smoothed with the Savitzky-Golay filter. The former makes it easy to apply the procedure to all GRBs with well measured lightcurves (i.e., with a significant signal-to-noise ratio, and a reasonable size of time bin—neither too large to erase the variability on short timescales, nor too small to introduce too large Poisson noises), without referring to the redshift information.

We have applied our smoothing procedure to a sample of 25 long duration GRBs with measured redshifts and publicly available data. A very tight correlation between the variability and the peak luminosity is found, with only one prominent outlier: GRB 030329. Excluding 030329, 980425, and 030528 (the latter two are due to their too large errors in variability), a linear fit to the log of variability and the log of peak luminosity is obtained (equations 5 and 6), with $\chi^2/\text{dof} = 1.93$. The smallness of the reduced $\chi^2$ indicates that the data scatter is very small (Fig. 1 and Fig. 5).

We note that, although the existence of a second, larger episode of emission from GRB 050820a has been claimed (Golenetskii et al. 2003; Pal’shin et al. 2003; Cummings et al. 2005) and we have only made use of the data for its first episode, GRB 050820a fits the relation perfectly well.

Our results are a significant improvement to that of G05 (as well as that of FR00 and R01), who used a different definition of variability and obtained $\chi^2/\text{dof} = 1167/30 = 38.9$. An analysis on the overlapping sample of 17 GRBs (after GRBs 980425 and 030329 being excluded) shows that the scatter in our data is smaller than that in G05’s data by a factor of 2.7, measured by the deviation of fit (Section 5).

The improvement to the correlation is caused by using a new smoothing procedure and a different normalization in the definition of our variability, not by neglecting the effect of GRB redshift. R01 have shown that the dependence of variability on the redshift is extremely weak. The fact that a very tight correlation between the peak luminosity and the variability is obtained without including the corrections to the variability from the GRB redshift also leads us to believe that such corrections are not necessary.

A remarkable feature of the correlation that we have found is that it does not rely on the corrections to the luminosity from the collimation of GRB jets. In other words, the peak luminosity used above is simply the isotropic-equivalent peak luminosity. If this correlation is confirmed by future bursts, it will provide a convenient calibration to GRB luminosity and distance.

GRB 030329, which is considered to be a firm case for the connection of GRBs with supernovae (Stanek et al. 2003; Hjorth et al. 2003), is offset from the correlation by 7 $\sigma$ (Fig. 1). Although the cause is not clear, we have noticed that 030329 and 050525 differ from other GRBs in the sample in the following way: their lightcurves consist of two distinctly separated pulses with each containing at least 30% of total counts. Motivated by this observation and the fact that GRB 050820a fits the relation well although only the first episode of its emission has been used, we have recalculated the variabilities of 030329 and 050525 by dividing the smoothing window by a factor of 2 (i.e., dividing the
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We obtained $V = 0.0022 \pm 0.0003$ for 030329, which well fits the solid straight line in Fig. 1. For 050525, we obtained $V = 0.0032 \pm 0.0007$, which fits the straight line equally well as the value listed in Table 1. Whether this treatment is correct must be tested when more GRBs with lightcurves similar to that of 030329 and 050525 are available.

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APPENDIX A: ANALYTIC APPROXIMATION TO THE ERROR IN THE LOG LUMINOSITY

The numerical results for the error in the log $L$ predicted by equation (5) due to the uncertainties in $a$ and $b$, $\sigma_{ab}$, can be approximated by the following formulae ($\sigma_{ab,1}$ for the upward error, $\sigma_{ab,2}$ for the downward error):

1) When $log L < 52$

$$\sigma_{ab,1} \approx 7.2017 - 0.16777x + 0.0006006x^2,$$  \hspace{1cm} (A1)

$$\sigma_{ab,2} \approx 8.0929 - 0.18000x + 0.0005134x^2,$$  \hspace{1cm} (A2)

where $x \equiv log L$, $L$ is in units of erg/sec.

2) When $52 \leq log L < 54$

$$\sigma_{ab,1} \approx 2274.034248 - 124.984716x + 2.286986x^2 - 0.013930421x^3,$$  \hspace{1cm} (A3)

$$\sigma_{ab,2} \approx 3028.169929 - 167.524125x + 3.087093x^2 - 0.018948436x^3.$$

(4) When $log L \geq 54$

$$\sigma_{ab,1} \approx -6.8502 + 0.12989x,$$  \hspace{1cm} (A5)

$$\sigma_{ab,2} \approx -5.7594 + 0.10933x.$$  \hspace{1cm} (A6)

The fractional errors in $\sigma_{ab}$ given by these formulae are always $< 5%$.

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2 Roughly speaking, this is equivalent to smooth each GRB pulse separately.