

Identification of gamma-ray burst precursors in Fermi-GBM bursts

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We present an analysis of more than 11 years of Fermi-GBM data in which 217 gamma-ray bursts (GRBs) are found for which their main burst is preceded by a precursor flash. We find that short GRBs (<2 s) are ~ 10 times less likely to produce a precursor than long GRBs. The quiescent time profile, given by the time between the precursor and the main burst, is well described by a double Gaussian distribution, indicating that the observed precursors have two distinct physical progenitors. The light curves of the identified precursor GRBs are publicly available in an online catalog (https://icecube.wisc.edu/~grbweb_public/Precursors.html).

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I. INTRODUCTION

Gamma-ray bursts (GRBs) are cataclysmic transient cosmic events characterized by the emission of one or multiple flashes of gamma radiation. They are the most powerful outbursts of electromagnetic radiation in our universe and a possible source of (ultra) high-energy cosmic rays [1–4]. The duration of GRBs can be described using a bimodal distribution [5], indicating the existence of two progenitor source classes. In general, bursts lasting longer than 2 s are related to the collapse of a supermassive star, as confirmed by the observation of type-Ic supernovae in coincidence with long GRBs [6–8]. Short bursts, lasting less than 2 s, are believed to occur when two coorbiting neutron stars collide. Evidence for this model was recently obtained by the detection of gravitational waves from a binary neutron star merger followed by a short GRB [9,10].

The main outburst of gamma radiation, called the prompt phase, is followed by an afterglow stage in which the ejected matter collides with the surrounding medium. Thanks to multiwavelength observations, ranging from x-ray to radio, the physical processes related to this afterglow emission are well understood [4]. Apart from the prompt and afterglow phases, there is a third emission phase, called the GRB precursor. Precursors are typically defined as relatively dim gamma-ray flashes that occur before the prompt emission. Previous studies [11–24] found that precursor flashes occur in a subset of both long and short GRBs. The fraction of bursts in which a precursor is observed strongly depends on the method and criteria used to define a precursor and typically ranges from 3% to 20%.

Numerous models have been proposed to explain precursor flashes and typically apply to a specific class of GRB progenitors. In the case of short GRBs caused by the

merger of a binary neutron star system, the interaction between the magnetic fields of the neutron stars could induce a Poynting flux prior to the GRB [25,26]. Alternatively, tidal forces could induce resonance modes that lead to a failure of the neutron star crust, potentially releasing enough energy to be observed as a precursor [27–29]. In the case of long GRBs, precursors could be related to the emission of an early weak jet. Such jets are predicted in two-step engines, where the star first collapses to a proto-neutron star or spinar, before collapsing to a black hole [30,31]. Multiple jets could also result from an effective turn-off of the central engine, related to sudden changes in the accretion rate [18,32]. In this case, the precursor and prompt emission would be caused by the same physical processes and thus have similar observational properties. Finally, precursors could also be related to the transition of the GRB ejecta into the optically thin phase [33–37]. The precursor is then typically expected to have a thermal spectrum, as was for instance observed for the first identified precursor [11]. Currently, there is no consensus on the origin of precursor flashes and, most likely, more than one model will be needed to explain all observed precursors. Given that precursors only occur in a subset of all GRBs, an extensive study is thus required to uncover their physical origins.

We performed an automated search that identifies precursor flashes observed by the Fermi-GBM detector. Out of a total sample of 2364 GRBs, 244 precursors were identified originating from 217 GRBs, of which 139 are newly identified GRBs with precursor emission.

In this paper, we present the details of our selection and show that short GRBs are ~ 10 times less likely to produce a precursor than long GRBs. We performed an analysis on the quiescent time profile, given by the time between the precursor flash and the main burst. The increased statistics from our search allowed us to identify a novel feature in the

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quiescent time profile, which is well described by a double Gaussian distribution, indicating that the observed precursors have two distinct physical progenitors. To allow for follow-up studies, searching for coincidences with other astrophysical messengers, such as neutrinos and gravitational waves, the obtained results for each individual GRB are presented in the Appendix A and have been made available via an online tool [38].

II. DATA

The Fermi Gamma-ray Space Telescope is currently the most efficient GRB detection satellite in orbit. Its two main instruments are the Large Area Telescope (LAT) and the Gamma-ray Burst Monitor (GBM). Whereas LAT has a sky coverage of 20%, GBM continuously observes the full region of the sky not occulted by Earth. On average, the GBM and LAT detect 240 and 18 GRBs per year, respectively [39–41]. In this study we analyzed 2684 GRBs, using all GBM recorded bursts up to the year 2020.

The GBM telescope is composed of 12 sodium iodide (NaI) and two bismuth germanate (BGO) detectors. Trigger and localization information is provided by the NaI detectors, which are sensitive to gamma rays of 8 keV to 1 MeV. The BGO detectors, which will not be used in this analysis, are sensitive from 200 keV to 40 MeV and serve to cover the energy gap with the LAT [40].

The GBM burst data was obtained from the Fermi Science Support Center [42] and provides the raw photon counts as a function of time and energy for each of the 14 detectors. Time-tagged event (TTE) data provides the highest temporal resolution of $2 \mu\text{s}$. Since August 2010, TTE data is available over a time window $[t_{\text{tr}} - 135\text{s}, t_{\text{tr}} + 300\text{s}]$, where t_{tr} is the GBM detector trigger time. Before August 2010, TTE data is only available starting 30 s before t_{tr} , but again up to 300 s after t_{tr} . CTIME data is provided over a 2000 s time window centered around t_{tr} , but has a coarser nominal resolution of 0.256 s. To allow the detection of very short emission periods, we have used TTE data whenever available. CTIME data was used to extend the examined time window to 1000 s before and after the trigger time.

III. METHOD

For every burst, we select the GBM NaI detectors that were triggered by the GRB. If more than three detectors were triggered, only the three triggered detectors which were pointing closest to the burst location are used. The data analysis is two-fold. An initial analysis on raw time data is performed to characterize the background, allowing to capture global fluctuations. Subsequently, a Bayesian block (BB) algorithm [43] is used to select the physical signal regions.

Our analysis aims at identifying all emission periods in which gamma-ray activity is observed from the detected

GRBs. This is achieved by constructing background subtracted light curves. For more than 90% of the identified bursts, a stable background fit is first obtained between 1000 s and 800 s before t_{tr} , marking the start time of the analysis interval. The end time of the analysis interval is set 50 s past the end of the Fermi T90 interval, defined as the central time window that contains 90% of the fluence of the GRB. If the Fermi T90 exceeds 250 s, we set the end time $0.2 \cdot \text{T90}$ instead of 50 s beyond the T90 interval. One final consideration is that a minority of all bursts have one or more gaps in their light curves due to missing data. For those bursts, we only examine the continuous data taking period containing t_{tr} . This choice is motivated by the observation that for $<1\%$ of all bursts, additional data is available at earlier times.

We automated the selection of background time intervals in which no increased gamma-ray activity is observed. Our selection is therefore fully reproducible and based on physically motivated parameters. Background times are selected based on the requirement that the rate does not undergo a sudden increase. For this purpose, we use an algorithm similar to the Fermi-GBM online trigger [40], which compares the observed rate to a prediction based on a fit to the rate at earlier times. The rate in the identified background intervals is then extrapolated to intermediate regions. As such, we obtain a fully data driven estimate for the background rate over the full light curve. A more detailed description of this method and a motivation for the use of Poisson statistics are provided in Appendixes B and C, respectively.

Having characterized the background rate, we proceed by rebinning the data using the Bayesian block (BB) algorithm [43]. The BB algorithm was specifically designed to identify localized structures, such as bursts, in GRB light curves. It optimizes both the number of bins and the location of the bin edges by maximizing a fitness function. For every selected GBM detector, we construct a BB light curve. In addition, a single BB light curve based on a combination of the photon counts of the selected detectors is also constructed for every burst. These combined light curves contain the largest statistics and will thus serve as the basis for our selection. To illustrate the BB procedure, the light curve of GRB 190114C is displayed in Fig. 1.

IV. ANALYSIS

To quantify the physical signal, a background subtraction procedure is applied. The background rate is integrated over each bin to estimate the total number of background events N_b . Subtracting N_b from the observed count and dividing by the bin duration, we obtain an estimate for the signal rate r_s . A threshold condition $r_s > r_{\text{th}}$ is then imposed to tag those bins that potentially contain a physical signal. In addition, we impose the requirement that the excess is separately observed by two or more detectors.

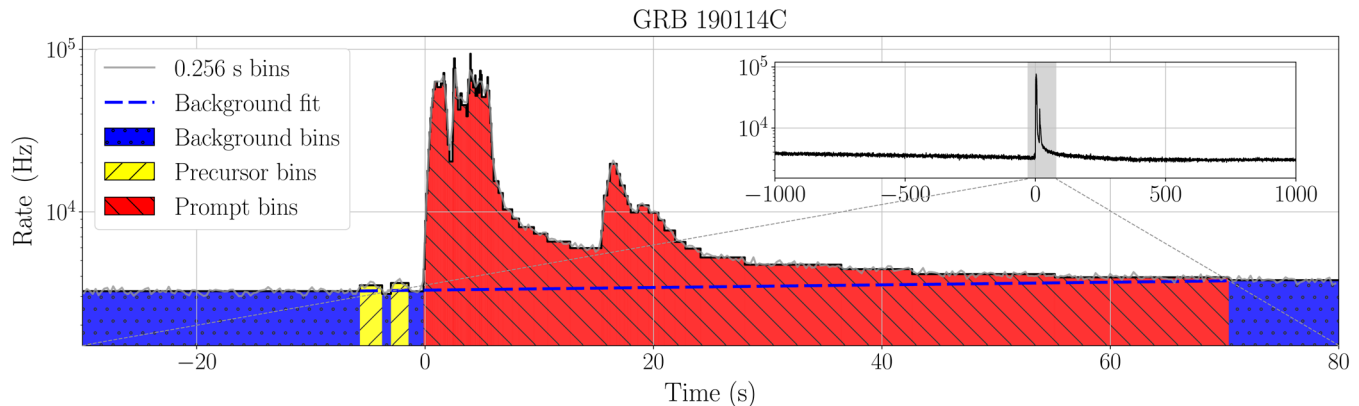


FIG. 1. Illustration of the Bayesian block light curve of GRB 190114C. Two dim precursors (yellow) are observed, starting 5.57 s and 2.85 s before the onset of the prompt emission (red). The total photon count in the precursor bins is 6839 and 5590, respectively, exceeding the expected background count by more than 6σ in both cases. The time range displayed in this figure corresponds to the grey shaded area in the inset image, which displays the full light curve.

The threshold rate r_{th} is based on the trade-off of minimizing the number of false positives, while maximizing the sensitivity of the search and set equal to $r_{\text{th}} = 30$ Hz, as motivated in Appendix D. By imposing a fixed threshold rate, we account for the uncertainty on the characterization of the background rate described in Appendix B. No additional criteria on the statistical excess of the rate are imposed, as the occurrence of a bin edge in the Bayesian block light curve already signifies that a statistically significant change in the rate has occurred.

We define an emission episode as a continuous period of increased emission in the background subtracted light curve. If a GRB has two or more emission episodes, we verify that the intermediate quiescent periods contain enough statistics to ensure that the rate has dropped back to the background level. Quiescent periods for which the Poisson uncertainty on the average background rate exceeds 5% are disregarded. For a typical burst [44], this corresponds to a lower limit on the allowed duration of the quiescent period of ~ 0.2 s. Given the data, we can hence not constrain models predicting precursors less than a few times 0.1 s before the start of the prompt emission. Particularly in the case of short GRBs, such short delay precursors have been proposed due to, e.g., tidal crust failure in binary neutron star mergers [27–29].

Having obtained our signal regions, we define the prompt signal phase as the emission episode with the largest photon fluence. If one or more emission episodes precede the prompt phase, we select them as GRB precursors. To verify that these early emission episodes are not caused by an unrelated weak transient at a different location, we compare the relative photon count ratios of the selected detectors. If the photons from the two emission episodes are coming from the same source and thus direction, their relative fraction should be the same between the different NaI detectors. In general, we find that the count ratios of precursors are consistent with those of the

prompt emission, except for five potential false candidates with deviating values, presented in Appendix E. These five edge cases are included in our analysis, but marked as being potentially unphysical in the precursor catalog. As a final check in our selection procedure, the light curves of all GRBs for which precursor emission is found are inspected by eye. This allows to verify that the identified emission was based on a reliable fit to the background rate and that the GBM detector was in a stable operation mode at the time of the GRB.

V. RESULTS

Applying our signal selection method on all 2684 bursts, we find 320 GRBs that were triggered by Fermi-GBM, but do not show a signal following our criteria. In the following, we therefore restrict ourselves to the 2364 bursts for which a signal is found.

Our analysis identified 244 precursor emission episodes spread over 217 GRBs. We thus find that 9% of all GRBs have one or more precursors. Any given burst is observed to have at most 3 precursors. The number of bursts having 1, 2 and 3 precursors corresponds to 192, 23 and 2, respectively. Based on the number of signals observed in a background control region and considering the combined time preceding the prompt emission of all GRBs, equal to 2.1×10^6 s, we estimate that the number of false positives in our analysis is 36.1 ± 8.8 , roughly 15% of the full sample. A complete catalog containing the start time, the duration, and the time separation of the precursors with respect to the prompt phase is given in Appendix A or can be accessed via the online tool [38].

Short GRBs. The selected GRBs can be subdivided based on the duration of the burst. While 14% of the 2364 examined bursts are short GRBs, only four (1.8%) of the 217 bursts with precursors are short GRBs [45]. For each of these 4 bursts, we observe that the precursors occur within

2 s before the prompt emission. All 4 short GRBs have a precursor that is shorter in duration than the prompt phase and their quiescent times are consistent with one another up to a factor ~ 3 . While limited in statistics, we note that the time intervals between the onset of the precursor and prompt emission are smaller than the 1.7 s time gap separating the gravitational waves and the gamma rays that were observed from GRB 170817A [10]. These short timescales are consistent with the predictions of binary neutron star models [25,28,29] and in line with the results from previous studies [12,28].

Long bright GRBs. An important parameter that affects whether or not a precursor is observed, is the apparent brightness of the burst. If a burst is too dim, the even dimmer precursor will be indistinguishable over the background. Alternatively, if the prompt emission of a dim burst has multiple peaks, the first peak(s) might mistakenly be identified as a precursor. These effects are eliminated in some analyses [14,20,22,24], by only selecting sufficiently bright bursts. To study the impact of including dim bursts in our sample, the bottom half of Fig. 2 shows the fraction of GRBs for which a precursor is observed, when only including bursts whose per detector peak rate exceeds the threshold rate $r_{p,\min}$ displayed on the x -axis. Only long bursts are used, as the number of short GRBs with precursors is too limited to make a significant statement. An initial increase in the fraction of GRB precursors is observed for increasing threshold rates, which flattens out at 16% once the threshold rate exceeds 2×10^3 Hz. This effect indicates that the true fraction of long GRBs with precursors is underestimated when including dim bursts.

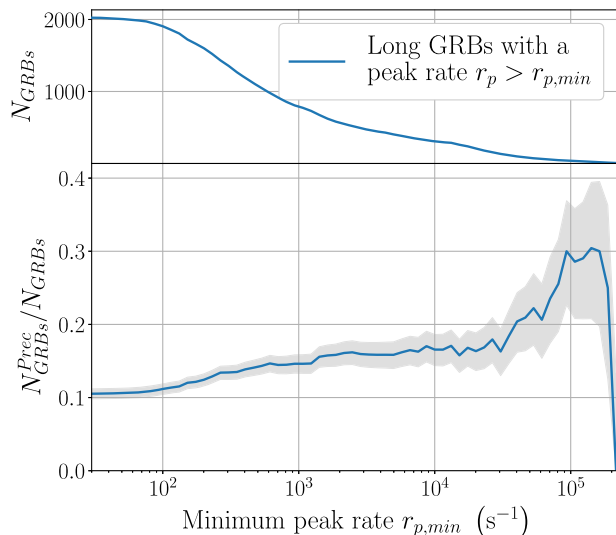


FIG. 2. Top: number of long GRBs whose peak rate exceeds the threshold value displayed on the x -axis. Bottom: fraction of those GRBs for which precursors are observed. When selecting increasingly bright bursts, we observe an increase in the fraction of GRBs with precursors. The shaded grey band shows the 1σ statistical uncertainty.

At rates exceeding 3×10^4 Hz, a second rise is observed. However, no significant statement can be made due to the limited number of increasingly bright bursts, as seen from the top half of Fig. 2.

Quiescent times. A quantity that can be used to relate the observed precursors to different theoretical models, is the duration of the quiescent interval separating two emission episodes. Figure 3 displays the full distribution of the quiescent time that follows the observed precursors. Two populations are observed, crossing over at $\Delta t_Q \sim 1.4$ s. Applying a two-component Gaussian likelihood fit, we find that the distributions peak at 0.55 s and 24 s and have a weight of 11% and 89%, respectively. To evaluate the goodness-of-fit, we calculated the likelihood of obtaining the observed bin counts given the best fit parameters. Comparing this value to that of 10^6 pseudoexperiments, randomly generated assuming Poisson statistics, we obtain a p -value $p = 0.36$, indicating that a two-component Gaussian fit provides a good description for the data. Performing a single-component Gaussian fit, we obtain a p -value of only $p = 8 \times 10^{-5}$, showing the data is incompatible with a single Gaussian fit. An apparent third component shows up in the last three bins. This contribution is however most likely not physical, as the expected number of false positives is proportional to the width of the bins, thus linearly increasing from left to right.

The leftmost component of the double Gaussian fit in Fig. 3 could have several origins [46]. A first contribution is found from the precursors of short GRBs, though they can only account for $\sim 15\%$ of the observed excess. A second contribution could come from bursts whose observed flux drops below the observable limit in between different peaks of the prompt phase, thus falsely identifying as precursor emission. Figure 4 illustrates that bursts with $\Delta t_Q < 1.4$ s are on average less bright than bursts with longer quiescent periods. To probe the effect of dim bursts on the bimodal feature, we repeated the analysis using only bright long

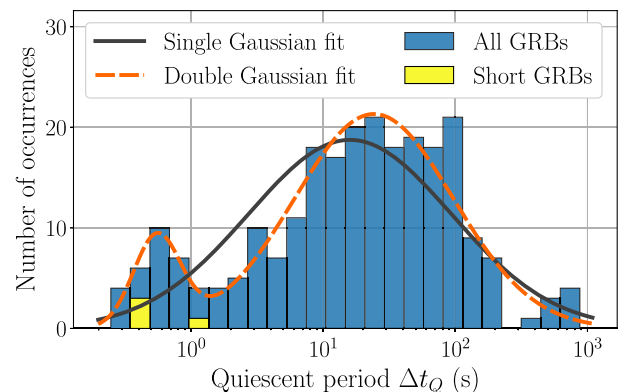


FIG. 3. Distribution of the quiescent time between two subsequent emission episodes. The data is found to be well described by the sum of two Gaussian functions. The 4 short GRBs are indicated in yellow.

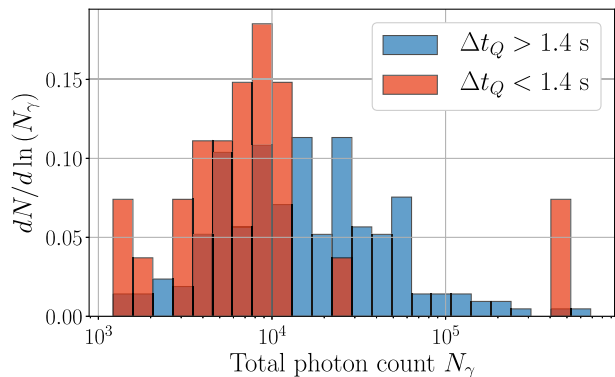


FIG. 4. Probability distribution of the total photon count of long GRBs in which we identified precursors. The normalization is taken such that the y-values sum to unity. Bursts for which the quiescent times in between emission episodes is less than 1.4 s are observed to be less bright on average.

GRBs, as detailed in Appendix F. For this subset of bursts, we found that the two-component Gaussian fit ($p = 0.57$) is still strongly preferred over the single Gaussian fit ($p = 0.076$), although the obtained p-value for the single component Gaussian fit increased due to the decrease of statistics.

The observation of two components in the quiescent time distribution might indicate different physical origins for the short delay precursors given by the first component and the long delay precursors given by the second component of the fit. Short delay precursors to long GRBs could correspond to those bursts for which the photospheric emission is observed [35]. In contrast, the longer quiescent times of tens to hundreds of seconds can be explained by models in which a jet is launched multiple times. The repeated launch of a jet has in some cases [18] been confirmed by identifying the separate afterglow of the precursor and prompt emission. To uncover the origin of the short delay precursors, follow-up studies looking further into the individual properties of these events are encouraged.

Quiescent times also provide an independent probe to investigate potential differences between the precursor and prompt emission. Previous studies generally found that precursor emission exhibits the same spectral properties as prompt emission [12,13,15,17,20]. In the case of long GRBs, this observation can be embedded in a model in which the precursor and prompt emission arise from the same physical mechanism and are caused by the accretion of different shells of matter falling onto the central engine [4,18,20]. As such, there would be no intrinsic physical difference between precursor and prompt emission. Hence, the distribution of the quiescent times between two precursors and between precursor and prompt emission should be identical. These two distributions are shown in Fig. 5. To quantify their resemblance, we use a two-sample Kolmogorov-Smirnov test. The resulting p -value of

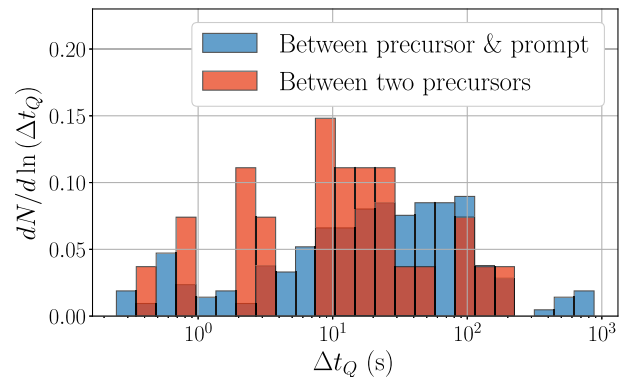


FIG. 5. Probability distribution of the quiescent time Δt_Q between two precursors (red) and between a precursor and the prompt emission (blue). The normalization is taken such that the y-values sum to unity.

$p = 0.030$, while not significant, indicates that there potentially could be a difference between the two samples.

Temporal correlations. A related study of quiescent times was performed in [22], where a strong linear correlation between the duration of the quiescent time Δt_Q and that of the subsequent emission episode Δt_{sub} was found. This correlation was explained using a model in which potential energy builds up during the quiescent interval and is released once a critical threshold is reached. However, due to lack of data, redshift effects, which could naturally induce such a correlation, were not considered. To probe possible redshift effects, we determined the correlation between Δt_Q and Δt_{sub} for the 21 bursts in our selection with known redshift z , and apply a correction for redshift. The obtained Pearson correlation factor is 34%. To determine the significance of this value, we composed a test statistic distribution by calculating the correlation coefficient between random combinations of the quiescent times and secondary emission episodes. Based on this distribution, we obtain a p -value of $p = 0.071$. No significant linear correlation is thus observed between the duration of the quiescent time following precursor episodes and the duration of the secondary emission episode. To test for a non-linear but monotonic correlation, we calculate the Spearman's rank coefficient ρ . Using the redshift corrected values of Δt_Q and Δt_{sub} , we obtain $\rho = 0.48$, corresponding to a p -value of $p = 0.020$. A slight tension is thus observed for the null hypothesis that there is no correlation.

We applied the same methods to look for correlations between the duration of the precursor and prompt emission. As before, only bursts with known distance are used, such that redshift effects can be corrected for. A linear correlation coefficient of 21% is recovered, corresponding to a p -value of $p = 0.16$. Computing the Spearman's rank coefficient to probe for non-linear correlations, we obtain a correlation coefficient of 35%, leading to a p -value of $p = 0.11$. Hence, no significant correlation is observed.

GRB 190114C. One object in our selection is of special interest, GRB 190114C, a particularly bright burst that occurred on the 14th of January 2019 [47]. GRB 190114C/bn190114873 is the first GRB from which TeV photons have been detected, as observed by the MAGIC telescope in La Palma [48]. Our analysis identified two faint precursors occurring 5.57 s and 2.85 s before the start of the prompt emission and lasting 1.94 s and 1.54 s, respectively. The detailed light curve of this burst is shown in Fig. 1.

VI. CONCLUSION

By applying a fully automated precursor search on the light curves of 2364 GRBs, we identified a total of 244 precursors spread over 217 bursts. Only four of those precursors occurred for short GRBs. We thus find that the fraction of long and short GRBs with one or more precursors equals 10.5% and 1.2%, respectively. All precursors for short GRBs occurred within 2 s before the start of the prompt emission. A notable long GRB for which we found two precursors is the extremely bright GRB 190114C. This burst was preceded by two dim precursors, indicating that gamma-ray production was already ongoing 5.6 s before the start of the prompt emission.

Apart from studying individual bursts, we also examined the quiescent time of all GRB precursors. A bi-modal distribution is observed, possibly indicating that precursors can have two types of progenitors. While we found no correlation between the duration of the precursor and prompt emission, we cannot exclude ($p = 0.020$) that there is a correlation between the quiescent time and the duration of the subsequent emission episode. Follow-up studies to further examine these claims are encouraged. To this end, and to allow other multimessenger correlation studies, we have included a full list detailing the emission times of the identified precursors in Appendix A and the online tool [38].

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APPENDIX A: PRECURSOR CATALOG

To enable follow up studies, we provide a complete list of the emission times of all precursors identified by our analysis. Table I provides the start time of the prompt emission in UTC, the start time of the precursor emission

TABLE I. Temporal properties of the identified precursors. For every GRB, we provide the start time of the prompt emission t_{prompt} , the start time of the precursor emission with respect to t_{prompt} and the duration of the precursor emission. The five potential false precursors with deviating count ratios are marked in *italic*. To access this table in a digital format, please visit https://icecube.wisc.edu/~grbweb_public/Precursors.html.

GRB	t_{prompt} (UTC)	$t_{\text{precursor}}$ (s)	Duration (s)
bn080723557	13:22:55.412	-34.284	28.319
bn080807993	23:50:44.177	-11.612	1.032
bn080816503	12:04:39.495	-21.823	1.823
bn080818579	13:54:44.589	-20.361	5.596
bn080830368	08:50:22.699	-8.559	5.112
bn081003644	15:27:27.738	-11.363	4.320
bn081121858	20:35:31.671	-8.498	7.855
bn090101758	18:13:07.574	-86.950	6.082
bn090113778	18:40:38.870	-0.475	0.150
bn090117335	08:02:26.183	-24.653	1.296
bn090131090	02:09:43.196	-22.324	12.445
bn090309767	18:25:41.699	-36.134	6.122
bn090326633	15:10:16.566	-583.057	0.256
bn090326633	15:10:16.566	-580.753	5.376
bn090419997	23:55:38.751	-37.348	23.251
bn090425377	09:04:14.740	-44.805	2.705
<i>bn090428441</i>	10:34:37.862	-26.762	18.048
bn090502777	18:40:11.917	-37.539	3.065
bn090510016	00:23:00.368	-0.420	0.024
bn090602564	13:32:22.296	-1.242	0.683
bn090610723	17:22:58.385	-90.937	6.686
bn090618353	08:29:16.651	-50.628	28.946
bn090720710	17:02:57.665	-0.776	0.264
bn090810659	15:50:40.542	-94.594	43.262
bn090811696	16:41:54.351	-4.958	1.583
bn090814950	22:48:30.233	-43.778	18.577
bn090815946	22:44:41.956	-179.466	12.722
bn090820509	12:13:25.368	-8.951	4.124
bn090907017	00:24:10.767	-1.967	1.664
bn090929190	04:33:04.488	-0.571	0.122
bn091109895	21:28:49.421	-9.606	2.788
bn100116897	21:32:19.006	-83.382	6.319
bn100130729	17:30:19.867	-65.378	23.215
bn100204566	13:34:36.243	-16.948	15.677
bn100323542	13:01:32.005	-54.935	9.109
bn100326402	09:37:30.596	-55.808	32.512
bn100424876	21:03:54.875	-123.791	2.521
bn100517154	03:42:30.304	-22.365	1.362
bn100619015	00:22:24.001	-77.870	9.918
bn100625891	21:22:58.362	-15.645	4.029
bn100709602	14:28:25.731	-56.254	16.328
bn100718160	03:50:13.287	-25.036	6.090
bn100718160	03:50:13.287	-7.415	6.808
bn100730463	11:06:50.220	-41.808	12.805
bn100730463	11:06:50.220	-18.243	0.001
bn100827455	10:55:49.710	-0.442	0.079
bn100923844	20:15:31.462	-24.128	4.019
bn101030664	15:56:24.411	-69.697	31.744
bn101224578	13:53:30.861	-33.455	10.658

(Table continued)

TABLE I. (Continued)

GRB	t_{prompt} (UTC)	$t_{\text{precursor}}$ (s)	Duration (s)
bn101227536	12:51:49.785	-3.895	3.646
bn110102788	18:55:41.740	-67.434	25.256
<i>bn110227229</i>	05:30:09.611	-111.145	21.120
bn110428338	08:07:18.821	-70.448	42.874
bn110428338	08:07:18.821	-18.748	13.398
bn110528624	14:59:12.297	-217.477	11.264
bn110528624	14:59:12.297	-35.653	13.654
bn110528624	14:59:12.297	-21.303	13.839
bn110725236	05:39:57.932	-16.720	7.619
bn110729142	03:30:47.288	-342.504	52.731
bn110729142	03:30:47.288	-185.188	51.556
bn110825102	02:26:58.702	-7.864	0.814
bn110903111	02:42:41.553	-187.466	22.062
bn110904124	02:58:55.085	-44.632	7.665
bn110909116	02:47:01.914	-4.433	1.670
bn110926107	02:34:30.183	-45.717	3.110
bn111010709	17:01:07.319	-34.749	31.018
bn111015427	10:15:22.011	-25.770	17.144
bn111228657	15:45:16.506	-46.111	10.496
bn111228657	15:45:16.506	-32.543	11.776
bn111230683	16:23:06.415	-11.301	4.631
bn111230819	19:39:41.521	-9.814	1.304
bn111230819	19:39:41.521	-8.120	4.234
bn120118709	17:00:24.779	-6.498	5.475
bn120308588	14:06:05.511	-21.363	3.092
bn120319983	23:35:18.709	-17.629	5.551
bn120412920	22:05:51.344	-71.057	5.502
bn120504945	22:40:07.713	-1.369	0.799
bn120513531	12:44:14.932	-15.008	1.330
bn120530121	02:54:31.969	-50.475	7.974
bn120611108	02:35:54.181	-8.321	6.602
bn120710100	02:25:09.865	-113.086	4.857
bn120711115	02:45:52.633	-61.735	4.838
bn120716712	17:08:00.170	-176.365	5.383
bn120819048	01:09:20.076	-60.316	7.618
bn120819048	01:09:20.076	-30.405	1.638
bn121005340	08:10:54.001	-101.730	38.794
bn121029350	08:24:27.774	-11.090	8.798
bn121031949	22:50:21.029	-191.769	38.495
bn121113544	13:03:25.589	-45.362	31.652
bn121125356	08:32:50.026	-29.374	20.325
bn121217313	07:29:53.089	-714.103	65.792
bn130104721	17:18:12.706	-5.969	3.898
bn130106995	23:52:56.117	-33.325	17.558
bn130208684	16:24:43.858	-21.975	5.099
bn130209961	23:03:46.502	-5.102	4.597
bn130219775	18:36:47.745	-56.310	20.260
bn130310840	20:09:45.591	-4.755	1.194
bn130318456	10:57:50.305	-82.735	6.897
bn130320560	13:29:06.051	-159.315	42.085
bn130404840	20:10:25.030	-21.354	8.355
bn130418844	20:16:08.506	-87.313	16.452
<i>bn130504314</i>	07:32:36.037	-32.672	0.464
bn130623130	03:07:03.470	-26.744	1.821
bn130720582	13:59:14.940	-146.139	115.366

(Table continued)

TABLE I. (Continued)

GRB	t_{prompt} (UTC)	$t_{\text{precursor}}$ (s)	Duration (s)
bn130813791	18:59:18.842	-5.810	1.680
bn130815660	15:51:22.993	-31.482	6.925
bn130818941	22:34:29.441	-70.463	8.706
bn130919173	04:09:40.924	-0.686	0.236
bn131014513	12:18:34.911	-20.917	2.089
bn131108024	00:34:43.981	-2.395	1.815
bn140104731	17:34:01.991	-120.439	66.204
bn140104731	17:34:01.991	-24.501	1.459
bn140108721	17:19:53.720	-71.900	11.570
bn140126815	19:33:40.215	-62.234	20.478
bn140126815	19:33:40.215	-24.368	14.110
bn140304849	20:25:37.760	-189.609	30.654
bn140329295	07:04:57.833	-19.534	0.630
bn140404030	00:43:22.825	-71.917	7.657
bn140512814	19:33:23.687	-98.421	11.788
bn140621827	19:50:14.988	-4.111	0.718
bn140628704	16:54:21.456	-66.005	4.910
bn140709051	01:13:51.906	-11.597	5.700
bn140714268	06:27:35.035	-109.468	27.544
bn140716436	10:29:26.513	-89.084	2.218
bn140818229	05:31:17.613	-69.604	10.233
bn140824606	14:34:24.964	-73.928	12.933
bn140825328	07:53:42.446	-59.289	11.821
bn140825328	07:53:42.446	-38.258	3.215
bn140917512	12:17:10.292	-4.434	3.940
bn141029134	03:14:24.675	-66.449	3.739
bn141029134	03:14:24.675	-41.574	6.940
bn141102536	12:51:40.471	-1.269	0.088
bn150126868	20:51:32.131	-55.037	13.019
bn150127398	09:32:49.909	-6.512	5.747
bn150226545	13:08:44.224	-202.152	1.028
bn150226545	13:08:44.224	-155.467	7.878
bn150226545	13:08:44.224	-41.188	16.158
bn150330828	19:53:59.254	-98.194	11.512
bn150416773	18:33:22.811	-824.534	42.496
bn150422703	16:52:31.997	-468.581	15.616
<i>bn150506398</i>	09:33:46.679	-116.285	27.791
bn150508945	22:40:36.620	-102.265	15.712
bn150512432	10:23:46.759	-86.467	43.029
bn150512432	10:23:46.759	-28.593	20.212
bn150522433	10:24:07.264	-19.511	7.822
bn150523396	09:30:14.993	-28.370	19.748
bn150627183	04:23:22.017	-458.665	3.072
bn150702998	23:56:45.108	-6.691	2.490
bn150703149	03:33:54.082	-13.280	0.008
bn150830128	03:04:38.646	-14.638	14.021
bn151027166	04:00:00.254	-96.360	40.571
bn151030999	23:59:47.634	-88.314	17.686
bn151211672	16:07:28.188	-151.405	26.022
bn160131174	04:12:52.609	-179.691	44.007
bn160201883	21:11:44.177	-1.590	0.968
bn160215773	18:36:08.605	-109.239	44.645
bn160219673	16:11:34.712	-110.393	12.546
bn160223072	01:45:54.364	-95.615	10.496
bn160225809	19:25:09.731	-48.115	23.215

(Table continued)

TABLE I. (Continued)

GRB	t_{prompt} (UTC)	$t_{\text{precursor}}$ (s)	Duration (s)
bn160512199	04:45:57.662	-56.663	9.377
bn160519012	00:18:55.054	-83.260	3.345
bn160519012	00:18:55.054	-65.164	17.101
bn160523919	22:04:13.977	-38.410	5.424
bn160625945	22:43:14.090	-178.317	2.418
bn160724444	10:40:02.521	-7.324	1.790
bn160821857	20:36:22.642	-117.067	31.832
bn160825799	19:10:50.313	-1.449	0.599
<i>bn160908136</i>	03:16:48.679	-87.733	6.845
bn160912521	12:31:42.840	-57.422	36.635
bn160912521	12:31:42.840	-17.072	5.193
bn160919613	14:43:36.685	-24.729	0.498
bn160919613	14:43:36.685	-15.527	0.761
bn161105417	10:01:18.575	-30.217	12.749
bn161111197	04:44:50.633	-102.555	11.125
bn161117066	01:37:14.177	-103.474	77.027
bn161119633	15:11:02.131	-10.916	7.666
bn161129300	07:11:45.292	-5.373	0.040
bn170109137	03:21:41.186	-245.940	18.163
bn170109137	03:21:41.186	-217.040	6.377
bn170115662	15:54:01.580	-95.287	18.563
bn170209048	01:09:05.007	-28.188	8.228
bn170302719	17:15:41.992	-22.294	12.259
bn170323775	18:36:31.186	-12.963	12.697
bn170402961	23:03:40.777	-15.936	1.501
bn170402961	23:03:40.777	-12.442	0.230
bn170416583	14:00:34.758	-35.298	12.494
bn170514152	03:38:43.989	-5.895	0.678
bn170514180	04:19:54.177	-79.666	35.908
bn170830069	01:38:59.546	-19.395	5.987
bn170831179	04:18:03.061	-73.621	8.547
bn170831179	04:18:03.061	-43.400	6.309
bn170923188	04:31:15.015	-10.012	1.018
bn171004857	20:33:34.433	-2.263	1.378
bn171102107	02:34:03.231	-29.516	10.393
bn171112868	20:50:13.004	-198.952	8.192
bn171112868	20:50:13.004	-43.928	9.502
bn171120556	13:20:33.596	-31.460	4.221
bn171211844	20:17:18.932	-82.541	12.393
bn180124392	09:23:59.613	-4.987	0.611
bn180126095	02:16:29.991	-820.685	11.776
bn180307073	01:44:35.183	-39.275	23.342
bn180411519	12:28:28.650	-54.086	26.673
bn180416340	08:10:01.701	-36.541	10.291
bn180426549	13:10:59.907	-13.182	5.544
bn180618724	17:22:55.701	-61.611	26.238
bn180620354	08:29:22.735	-72.842	5.855
bn180710062	01:29:21.269	-49.933	13.542
bn180720598	14:21:26.039	-29.189	10.000
bn180728728	17:29:11.437	-15.219	10.040
bn180822423	10:08:32.522	-5.898	2.803
bn180822562	13:30:29.570	-128.070	7.513
bn180822562	13:30:29.570	-118.178	6.344
bn180906988	23:42:36.388	-2.471	1.039
bn180929453	10:52:35.121	-1.456	0.606

(Table continued)

TABLE I. (Continued)

GRB	t_{prompt} (UTC)	$t_{\text{precursor}}$ (s)	Duration (s)
bn181007385	09:14:19.608	-23.373	3.996
bn181008877	21:04:29.161	-131.183	27.879
bn181119606	14:32:19.202	-2.566	1.798
bn181122381	09:09:04.964	-1.937	0.299
bn181203880	21:06:37.705	-6.482	0.870
bn181222279	06:42:52.975	-79.631	40.808
bn190114873	20:57:02.490	-5.573	1.942
bn190114873	20:57:02.490	-2.854	1.537
bn190205938	22:31:11.876	-40.086	9.198
bn190228973	23:21:30.204	-15.148	7.989
bn190310398	09:33:20.756	-49.157	4.120
bn190315512	12:17:39.138	-366.193	6.912
bn190323879	21:05:17.600	-893.855	26.624
bn190324947	22:44:18.392	-17.146	2.474
bn190326314	07:32:13.823	-27.769	1.672
bn190326314	07:32:13.823	-18.099	2.115
bn190610750	18:00:04.042	-14.819	1.160
bn190611950	22:48:51.696	-62.594	20.082
bn190719624	15:00:01.045	-86.830	1.579
bn190806675	16:12:34.836	-1.664	1.188
bn190828542	12:59:58.210	-46.588	38.536
bn190829830	19:56:40.582	-47.965	5.565
bn190901890	21:21:37.555	-63.144	20.014
bn190930400	09:38:17.809	-162.830	40.308
bn191019970	23:18:48.942	-96.333	29.779
bn191026350	08:23:43.801	-5.943	4.110
bn191031025	00:39:28.692	-178.171	10.422
bn191101895	21:28:37.561	-44.664	1.903
bn191111364	08:44:52.025	-25.765	16.425
bn191225309	07:26:50.763	-94.689	2.024

with respect to the onset of the prompt emission, and the duration of the precursor emission. An electronic version of this table can be downloaded from https://icecube.wisc.edu/~grbweb_public/Precursors.html.

APPENDIX B: BACKGROUND CHARACTERIZATION

During normal operation, the Fermi telescope functions in a sky survey mode [49]. This implies that the orientation of the spacecraft continuously changes to allow the LAT telescope to monitor the entire sky. A downside to this mode of operation is that the background rates of the GBM detectors are changing with time. A linear approximation can still be used over periods of time less than ~ 100 s, as the period of the oscillatory motion of the spacecraft is on the order of 3 hours [40,49].

During previous searches, the time ranges used to estimate the background rate were generally set by hand [13–15]. Since we plan to examine a time range of 2000 s for over 2000 bursts, this would become a very demanding

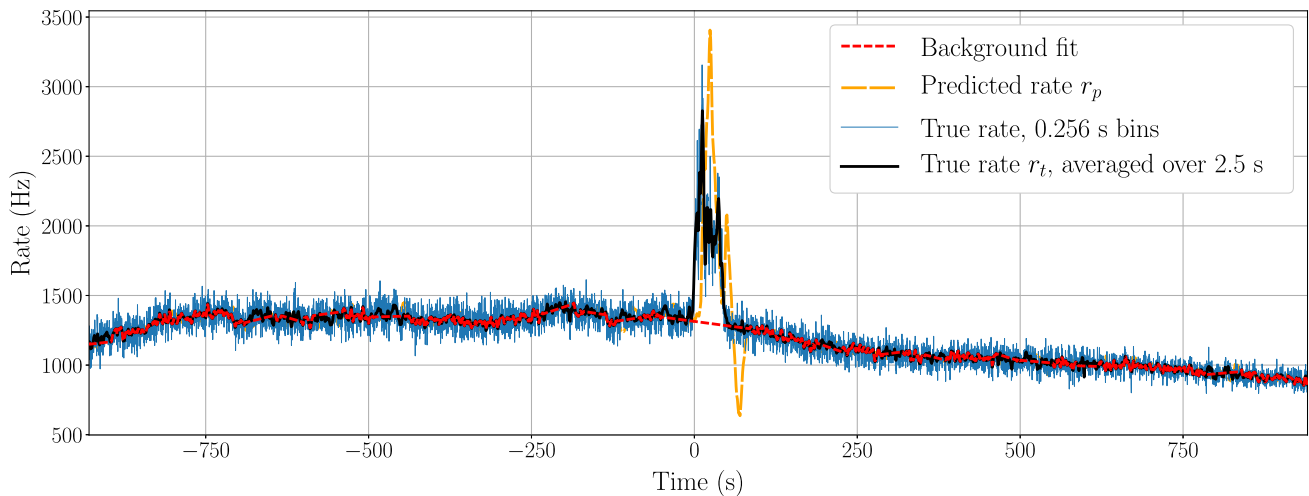


FIG. 6. Characterization of the rate of GRB trigger bn150422703 and the GBM detector labeled nb. The true rate averaged over 2.5 s (black) is compared to a prediction based on prior data points (orange). Time intervals in which these two distributions match are used to characterize the background rate (red).

endeavor. Therefore, we automated the selection of time intervals in which no increased gamma-ray activity is observed. This method has the added advantage that the selection is fully reproducible and based on physically motivated parameters.

Hence in this section, we will focus on the selection of good background intervals only. The tagging of reliable background time intervals is illustrated in Fig. 6 and based on the assumption that the observed rate can be predicted using the rate at earlier times. To predict the background rate at an arbitrary time t_1 , we perform a linear fit to the data in the time interval $[t_1 - 30 \text{ s}, t_1 - 10 \text{ s}]$. By extrapolating the fit to time t_1 , we obtain a prediction r_p for the background rate at time t_1 . This prediction is then compared to the true rate r_t found at time t_1 , averaged over 2.5 s. As long as the true rate is within a 3σ Poisson upper-fluctuation of the predicted background rate, i.e.,

$$r_t < r_p + 3 \cdot \sqrt{\frac{r_p}{2.5 \text{ s}}}, \quad (\text{B1})$$

the time t_1 is tagged as background. The next point in time $t_2 = (t_1 + 1 \text{ s})$ is then subjected to the same procedure, until a time t_n is found for which Eq. (B1) no longer holds. Knowing that we have arrived at a possible non-background region, we immediately advance 25 s. This is done to overshoot the nonstable background period with possible excess emission. We then proceed by verifying if the RMS of $r_p - r_t$, averaged over a 10 s period centered around $t_n + 25 \text{ s}$ is within 1.5σ of the Poisson expectation to verify background stability. If the RMS exceeds 1.5σ , $t_n + 25 \text{ s}$ is labeled as non-background. If, on the other hand, the RMS is sufficiently low, a new background interval is started at $t_n + 25 \text{ s}$. We then proceed with the procedure outlined above.

Using this method, background regions are identified in each of the light curves. The background rate is then set equal to the true rate, averaged over 2.5 s, in these background intervals. In intermediate possible regions of interest, a linear interpolation is used based on the last and first point of the adjacent background intervals. Figure 6 displays a visualization of this procedure for GRB trigger bn150422703.

APPENDIX C: POISSON FLUCTUATIONS

The method used to characterize the background rate builds on the assumption that the observed photon count follows Poisson statistics. To validate this assumption, we verify that the fluctuation in the number of observed photons describes a Poisson distribution. Using TTE data up to 30 s before the first signal is observed, we perform 10.000 trials, counting the number of photons observed during a time window $\Delta t = 50\lambda$, typically $\sim 0.05 \text{ s}$. Here, λ corresponds to the observed rate averaged over a 10 s period. Figure 7 displays the resulting distribution and the Poisson probability function expected from theory. The close match between the data and the theoretical expectation demonstrates that, on sufficiently small timescales, the variation in the observed photon count is Poissonian.

As a second check, we verify that the time delay δt between two subsequently observed photons follows an exponential distribution. To this end, we apply the same procedure as before, taking 10.000 trials from background control regions with average rate λ . Each trial uses 1 s of data, meaning we obtain a total of $\sim 10^7$ values for δt . As δt follows an exponential distribution, the variable $w = \exp(-\lambda\delta t)$ is uniformly distributed, where $w \in [0, 1]$. Figure 8 shows the distribution of w , combining the data from all 10.000 trials. While some deviations are observed

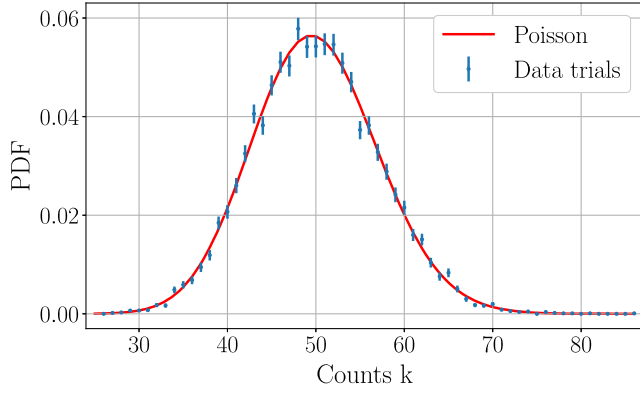


FIG. 7. Probability distribution of the observed number of counts for 10,000 trials, compared to the theoretical Poisson distribution. The close agreement confirms that, on small time-scales, the variation in the observed rate is Poissonian.

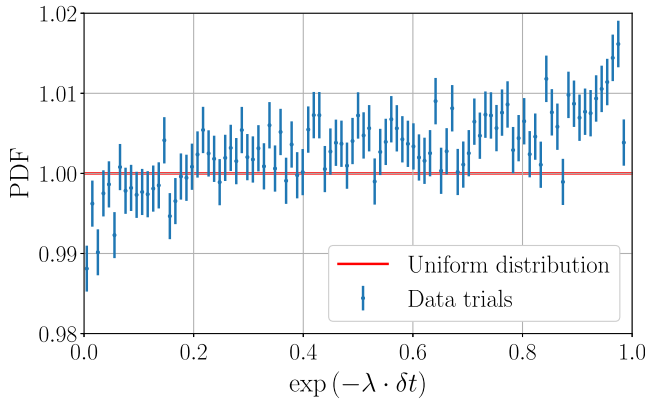


FIG. 8. Probability distribution of $w = \exp(-\lambda \delta t)$, where δt is the time delay between two subsequently observed photons and λ is the average rate. The rightmost bin is not displayed as it has a y -value of only 0.7 due to detector dead time.

due to detector effects, the observed distribution matches the theoretical expectation within the 1% level. The increase observed as $w \rightarrow 1$ is due to the pulse pile-up in the detector [40].

APPENDIX D: OPTIMIZATION OF THE THRESHOLD RATE

To select bins that may contain a physical signal, we require that the background subtracted rate r_s exceeds a threshold rate r_{th} . The Bayesian block algorithm already ensures that a statistically significant change of the rate is observed between adjacent bins. Hence, by imposing a fixed threshold rate, we mainly aim to account for the uncertainty that stems from the characterization of the background rate, as described in Appendix B.

The threshold rate r_{th} is based on the trade-off of minimizing the number of false positives, while maximizing the sensitivity of the search. Ideally, every event triggering the GBM detector should also be selected by

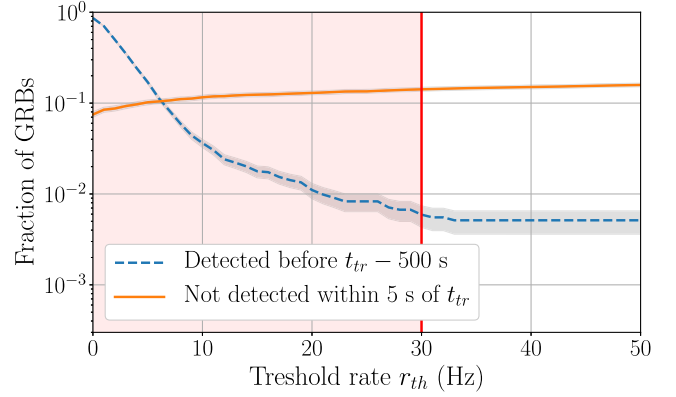


FIG. 9. Relative number of GRBs that are not detected within 5 s of the trigger time t_{tr} (full orange line) as a function of the threshold rate r_{th} . The same relation is shown for bursts that are detected more than 500 s before t_{tr} (blue dashed line), where few to no precursors are expected [16]. The shaded grey bands show the 1σ statistical uncertainty for both curves. In our analysis, the threshold rate is set equal to 30 Hz, indicated by the vertical line.

our analysis. To estimate the loss of sensitivity as a function of r_{th} , we therefore consider the fraction of GRBs in which, following our selection criteria, no excess is observed within 5 s of the GBM trigger time t_{tr} . This quantity is shown as the full orange line in Fig. 9 and shows a slow but steady increase as a function of r_{th} .

To estimate the false positive rate, we consider the number of GRBs which, following our criteria, yielded an emission episode in the period from 1000 s to 500 s before the Fermi-GBM trigger time t_{tr} . This time window is based on a previous study presented in [16], which, using a sample of 956 Fermi-GBM observed bursts, found only a single burst in which a precursor event occurred more than 500 s before t_{tr} . For our analysis, the fraction of GRBs which yielded a signal in this time range is displayed by the dashed blue line in Fig. 9 as a function of the threshold rate r_{th} . A plateau is reached at $r_{th} > 30$ Hz. We therefore set $r_{th} = 30$ Hz, as this corresponds to the minimal value for which the estimated false positive rate approaches the plateau at $\sim 0.5\%$, leading to an expected false positive rate $r_f = 1.7 \times 10^{-5}$ Hz. On rare occasions, physical precursors have been observed more than 500 s before the prompt emission. One example is the case of GRB 091024 [18]. As such, this approach is expected to result in a conservative estimate of the true false positive rate.

APPENDIX E: RELATIVE COUNT RATIOS

Apart from physical GRB precursors, the possibility exists that an emission episode preceding the prompt phase is caused by an unrelated astrophysical transient. To test this hypothesis, we verify that the sky location of the two emission episodes are consistent with one another. Defining the relative count ratio r to be the fraction of the total counts N observed by a given detector α , we compare the value

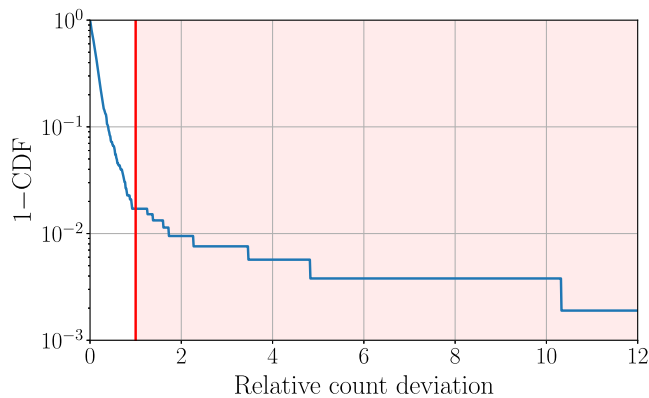


FIG. 10. Distribution of the relative deviation of the count ratios, as defined in Eq. (E1). Five of the eight emission episodes that contribute to the extended tail cannot be confirmed to have a count ratio consistent with that of the prompt emission.

of r_α between the precursor and prompt emission episode for each selected detector. Figure 10 displays the combined distribution of

$$\left| \frac{r_{\alpha,\text{precursor}} - r_{\alpha,\text{prompt}}}{r_{\alpha,\text{prompt}}} \right| \quad (\text{E1})$$

for all identified precursors. An initial steep decline is observed, indicating good agreement between the count ratio of the precursor and prompt emission. An extended tail however shows up at relative deviations larger than one. For the eight bursts contributing to this tail, we inspected the light curves by eye. In the case of bn110428338, bn180307073, and bn180618724, the large excess could be resolved by improving the characterization of the background rate. This leaves five emission episodes, namely the precursor of bn090428441, bn110227229, bn130504314, bn150506398, and bn160908136, for which we cannot confirm that the location of the precursor is consistent with that of the prompt emission based on the relative count ratios.

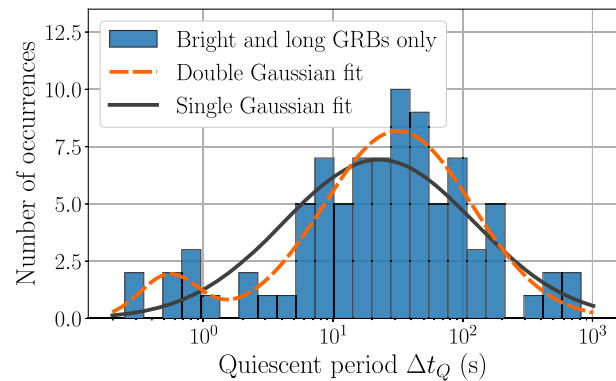


FIG. 11. Distribution of the quiescent time between two subsequent emission episodes of bright and long GRBs. While statistics are significantly reduced, we find that a double component Gaussian fit is still preferred over a single Gaussian distribution.

APPENDIX F: QUIESCENT TIME DISTRIBUTION

When examining the quiescent times between emission episodes, we found that the resulting distribution is well described by a two-component Gaussian fit. The inclusion of dim bursts in our sample could, however, lead to an artificial excess at short quiescent times, as the first peak of the prompt emission might be mistaken for a precursor. To probe this effect, we have repeated our analysis using only long bursts with a peak rate in excess of 3×10^3 Hz. Figure 11 shows the resulting distribution. The two Gaussians now peak at 0.54 s and 32 s and have a weight of 8% and 92%, respectively. Based on the goodness-of-fit p -value, we find that the double component Gaussian distribution ($p = 0.57$) is still preferred over the single-component Gaussian distribution ($p = 0.076$). As expected, the disagreement between the data and single Gaussian fit is less significant than when using the full precursor sample due to the reduced sample size.

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