

# Searching for *Fermi* GRB optical counterparts with the prototype Gravitational-wave Optical Transient Observer (GOTO)

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## ABSTRACT

The typical detection rate of  $\sim 1$  gamma-ray burst (GRB) per day by the *Fermi* Gamma-ray Burst Monitor (GBM) provides a valuable opportunity to further our understanding of GRB physics. However, the large uncertainty of the *Fermi* localization typically prevents rapid identification of multiwavelength counterparts. We report the follow-up of 93 *Fermi* GRBs with the Gravitational-wave Optical Transient Observer (GOTO) prototype on La Palma. We selected 53 events (based on favourable observing conditions) for detailed analysis, and to demonstrate our strategy of searching for optical counterparts. We apply a filtering process consisting of both automated and manual steps to 60 085 candidates initially, rejecting all but 29, arising from 15 events. With  $\approx 3$  GRB afterglows expected to be detectable with GOTO from our sample, most of the candidates are unlikely to be related to the GRBs. Since we did not have multiple observations for those candidates, we cannot confidently confirm the association between the transients and the GRBs. Our results show that GOTO can effectively search for GRB optical counterparts thanks to its large field of view of  $\approx 40$  deg<sup>2</sup> and its depth of  $\approx 20$  mag. We also detail several methods to improve our overall performance for future follow-up programmes of *Fermi* GRBs.

**Key words:** methods: observational – gamma-ray bursts.

## 1 INTRODUCTION

It is generally believed that gamma-ray bursts (GRBs) come from core-collapse supernovae (SNe; MacFadyen & Woosley 1999; MacFadyen, Woosley & Heger 2001; Woosley & Bloom 2006; Woosley 2011) or compact binary coalescence (CBC; Eichler et al. 1989; Paczynski 1991; Kochanek & Piran 1993; Bartos, Brady & Márka 2013; Rosswog, Piran & Nakar 2013). They are divided into two classes called long and short GRBs based on the duration of the

GRB with the boundary historically set at  $\sim 2$  s (Kouveliotou et al. 1993; Sakamoto et al. 2008, 2011; Zhang et al. 2012; Qin et al. 2013). Although GRBs have been studied for more than 50 yr since the first discovery on 1967 July 2 (Klebesadel, Strong & Olson 1973), it was the simultaneous detection of GW170817 and GRB170817A (Abbott et al. 2017a,b; Goldstein et al. 2017; LIGO Scientific Collaboration & Virgo Collaboration 2017; Savchenko et al. 2017) that the progenitor origin of short GRBs as CBC sources was confirmed.

The Burst and Transient Source Experiment (Goldstein et al. 2013) launched in 1991 established the first step in GRB research by showing the cosmological origin of GRBs. In order to explain the isotropic equivalent luminosity of GRBs, which could be up to  $\sim 10^{54}$  erg (Frail et al. 2001), the fireball model was introduced

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(Goodman 1986; Paczynski 1986; Shemi & Piran 1990; Rees & Meszaros 1992; Piran, Shemi & Narayan 1993; Vietri 1997; Wijers, Rees & Meszaros 1997; Mészáros et al. 2002). This model suggests that due to relativistic outflows, the gamma-ray emission is highly beamed with the beaming angle of  $\theta_b \sim 1/\Gamma$ , where  $\Gamma$  is the bulk Lorentz factor of the outflow (Rhoads 1999; Sari, Piran & Halpern 1999). The effect of relativistic beaming can significantly reduce the energy such that it can explain the energy scale of the GRB. The fireball model also invokes internal and external shocks to explain the prompt gamma-ray emission and the afterglow (Blandford & McKee 1976; Rees & Meszaros 1994; Beloborodov 2000).

In 1997, BeppoSAX (Boella et al. 1997) first enabled multiwavelength observations of GRB afterglows (Costa et al. 1997; Frail et al. 1997; van Paradijs et al. 1997). The Wide Field Cameras on board BeppoSAX can track Gamma-ray Burst Monitor (GBM) triggers to locate the GRBs with a precision of  $\approx 3$  arcmin (Jager et al. 1997). The detection of X-ray counterparts can help identify the optical counterpart, which can then be used to identify the host galaxy and constrain the luminosity of the GRBs with its redshift (Metzger et al. 1997). BeppoSAX also helped GRB980425 to be identified with SN1998bw (Galama et al. 1998), which suggested that the origin of long GRBs is core-collapse SNe (MacFadyen & Woosley 1999; MacFadyen et al. 2001).

Today the discovery of GRBs is mainly made through the *Swift* and *Fermi* satellites. There are two narrow-field instruments on board *Swift* – the X-Ray Telescope (Burrows et al. 2005) and the Ultraviolet and Optical Telescope (Roming et al. 2005) – which can search for the electromagnetic counterparts to the *Swift* Burst Alert Telescope (BAT; Barthelmy et al. 2005) triggers. However, the *Fermi* GBM (Meegan et al. 2009; von Kienlin et al. 2020) dominates the GRB detection rate, finding  $\sim 1$  per day, but with much larger location uncertainties compared to *Swift*. Other than *Fermi* GBM, *Fermi* also equipped the Large Area Telescope (LAT; Atwood et al. 2009) in order to detect high-energy gamma-ray emission from the GRB sources. The detection rate of *Fermi*-LAT is  $\sim 10$  per year (Kumar & Zhang 2015). As *Fermi* does not have either an X-ray or optical telescope on-board, and for GRBs for which there is no joint *Swift* detection, it is important to follow up the *Fermi* GRB detections in order to expand the sample of GRB afterglows available for GRB studies.

In general, there are two observing strategies that are used to find GRB optical afterglows: serendipitous and target-of-opportunity (ToO) follow-ups. For instance, Andreoni et al. (2021) demonstrate how to identify serendipitous fast transients, such as optical afterglow and kilonova, detected by *Zwicky Transient Facility* (ZTF) using ZTFReST, an automated filtering and follow-up infrastructure. The greatest science return of using this strategy is the proof of the existence of the orphan afterglow (Cenko et al. 2015; Coughlin et al. 2020; Ho et al. 2020). However, it is not an efficient way to searching for optical GRB afterglows as it is a blind search of the sky (Nakar, Piran & Granot 2002; Huang et al. 2020). On the other hand, ToO follow-up directly searches for the optical counterparts of the detected GRB by tiling skymaps generated by the *Fermi* or *Swift* instruments (Singer et al. 2013, 2015; Coughlin et al. 2019; Ahumada et al. 2021). However, the sizes of these skymaps can vary substantially, from square degrees to arcseconds, depending on the gamma-ray instrument. Targeting potential host galaxies and multicolour observations are usually used in ToO follow-ups, which help to improve search efficiency and identify the origin of the candidates. Other than ZTF, there are other observatories that actively follow up GRBs, such as Global MASTER-Net (Lipunov et al. 2012), Nordic Optical Telescope (Djupvik & Andersen

2010), and GROWTH-India Telescope, DDOTI (Watson et al. 2016).

The Gravitational-wave Optical Transient Observer (GOTO) is an optical robotic telescope that consisted of four 40 cm  $f/2.5$  unit telescopes (UTs) in its prototype design (GOTO-4) until 2019 November 1 and was upgraded to a full eight UTs (GOTO-8) in 2019 November (Dyer et al. 2020a; Steeghs et al., in preparation). The primary objective of the GOTO project is to detect the optical counterpart of gravitational-wave (GW) events detected by the LIGO–Virgo collaboration (LVC) network. GOTO followed up LVC triggers distributed during the third observing period, O3 (between 2019 April and 2020 March), and the details of GOTO’s O3 follow-up strategies and results are presented in Gompertz et al. (2020). In the absence of active GW triggers, GOTO also listens to automated alerts for GRBs from satellites such as *Fermi* and *Swift*. In the absence of any GRB or GW triggers, GOTO performs an all-sky survey on a routine basis. The pixel scale of GOTO is  $\approx 1.2$  arcsec pixel $^{-1}$ . GOTO can reach depths of  $\sim 21$  mag in the broad Baader  $L$  filter with a co-added set of  $4 \times 90$  s observations. Combined with a wide field of view (FoV) of  $\approx 40$  deg $^2$  in the GOTO-8 prototype design, GOTO has the capability for searching for *Fermi* GRB optical counterparts.

In this paper, we demonstrate how we perform automated follow-up and detail strategies for how we narrow down the number of potential candidates of a *Fermi* GRB optical afterglow. We also discuss the expected improvement in our follow-up strategy for the current GOTO-8 and for forthcoming upgrades to the network. In Section 2, we describe our current observing strategy and in Section 3 we indicate how we filter the GRB events automatically. The flow of manual filtering, which is different from the automated tasks from Section 3, in the final stages is outlined in Section 4. Our final results are described in Section 5. Finally, we discuss the planned expansion of the GOTO network and future follow-up strategy in Section 6.

## 2 GOTO OBSERVING STRATEGY FOR GRBS

GOTO is a fully autonomous telescope and is controlled by the GOTO Telescope Control System (G-TeCS; Dyer et al. 2018, 2020b). The G-TeCS software includes an alert monitor called the `sentinel`, which receives GW and GRB alerts released through the NASA Gamma-ray Coordinates Network (GCN) (Barthelmy et al. 1998). Once follow-up targets have been generated by GOTO `sentinel` for an event they are added to the GOTO observation data base from which the automated scheduler will select targets and instruct the telescope to move the mount to the requested position and take the requested exposures. The entire system is automated and results in fast reaction times on minute time-scales; if the sky position is visible and GOTO is able to observe then exposures can begin in  $\sim 1$ – $2$  min of the alert being received. The fast response of GOTO was successfully demonstrated by the follow-up observations of GW events detected in the LVC O3 run (Gompertz et al. 2020).

When not following-up alerts GOTO carries out an all-sky survey based on a fixed grid of tiles, which builds up an archive of reference images covering all points on the sky. When a GW or GRB alert is received its probability region is mapped on to the same grid as the survey, which allows any new images to be compared to the reference templates taken at the same position (see Section 3). Mapping the event skymap on to the survey grid is done using the `goto-tile` PYTHON package (Dyer 2020).

For *Fermi* GRB events, a two-dimensional Gaussian skymap is independently created by `goto-tile`, centred at the location

reported by a GCN. Since *Fermi* GBM shows a systematic error of  $\approx 3.6$  deg (at 68 per cent confidence level; Connaughton et al. 2015), we use the quadrature sum of the systematic error and the statistical error (which is included in the GCN) as the 68 per cent confidence region of the skymap. The measurements of the GRB location reported on GCN are only useful for generating our own Gaussian skymap as there are delays of  $\sim 10$  min between receiving the initial alert and the generation of the official *Fermi* skymap. We will ultimately replace the artificial Gaussian skymap with the official *Fermi* skymap once it has been received (see Section 6.1 for more details about the difference between our Gaussian skymap and the *Fermi* skymap).

Once the skymap has been created by `goto-tile`, the contained probability within each GOTO tile is calculated. For this work, a large fraction of our observing period overlapped with LIGO–Virgo O3. In order to avoid observations taking excessive amounts of time away from completing our template set before O3 and the follow-up observations of GW events during O3, only the five tiles with the highest probability are added into the observation data base and only if they each include at least 5 per cent of the overall localization probability error region.

As well as determining which tiles to target for follow-up observations, the G-TeCS *sentinel* also defines the number of times each tile should be observed and the delay between observations. In order to confirm whether an observed transient source is associated with the GRB at least two observations are required within 1 d after the trigger, to confirm the fast-decay nature of the GRB optical afterglow (Kann et al. 2010, 2011). Therefore, for GRB observations the *sentinel* uses the strategy `TWO_FIRST_ONE_SECOND`, which requires two observations to be taken within the first day after the trigger followed by another single exposure the day after. It should be noted that G-TeCS uses a ‘just-in-time’ scheduling system, which while it is flexible enough to allow for very fast initial observations, it cannot guarantee the exact timings for subsequent observations. Even accounting for delays due to poor weather or if the Sun rises before a second observation can be taken, it is possible that a higher priority target can be added into the scheduler queue, taking precedence over the GRB target while both targets are visible.

Finally, when targets are added into the observations data base, the exposure sets are then similarly defined. For *Fermi* GRB sources a set of four, 90 s exposures are taken using GOTO’s wide *L* filter (400–700 nm). In this work, all GOTO magnitudes have been calibrated in the standard way as described in Dyer et al. (2020a).

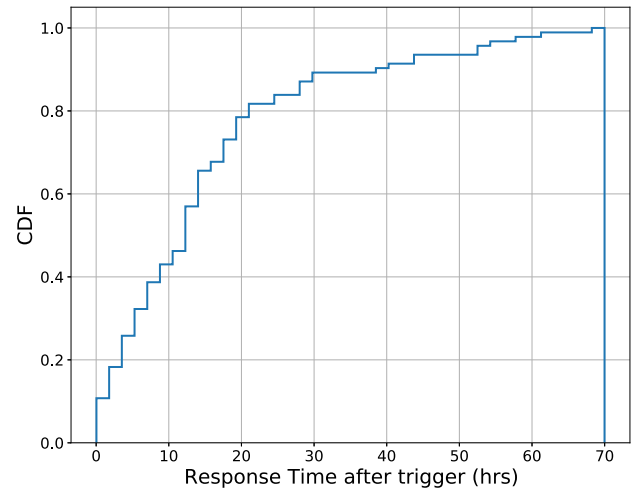
This work is an independent project of GOTO designed to test approaches for GRB afterglow searches, which does not represent the GOTO standard filtering procedures as described in Dyer et al. (2020a).

### 3 DATA PREPARATION

#### 3.1 Data selection

In this paper, we outline the results of the GRB follow-up from 2019 February 22 to 2020 June 7. We used the GOTO-4 prototype configuration until 2019 November 1. Beyond that date, we used the upgraded to GOTO-8 prototype configuration.

From 2019 February 22 to 2020 June 7, there were 508 *Fermi* GBM triggers reported on *Fermi* trigger GCN. Among those, there were 376 triggers located at declination greater than  $-30$  deg. Due to further restrictions of the GRB locations and timing and weather conditions at the site, GOTO observed a total of 93 of these alerts. Since one of the alerts (TriggerNum: 596808579) was later classified



**Figure 1.** Cumulative distribution of GOTO response time for all 92 observed *Fermi* GRB triggers. The response time has the range between 141 s and 70 h. The median response time is 12.9 h.

as a non-GRB trigger, we further reduce the observed targets to 92. The GOTO response times range between  $\approx 140$  s and  $\approx 3$  d, with the median response time  $\approx 13$  h (see Fig. 1). There are three reasons that can greatly delay the GRB follow-up observations. First, bad weather or poor site conditions are the most common issues preventing us from observing the event immediately after the trigger. Secondly, the event location is not always observable when triggered. Finally, since GW event triggers always have a higher follow-up priority, we do not interrupt any ongoing follow-up observation of the GW event due to the GRB trigger. Therefore, we postpone the GRB follow-up until the GW follow-up has finished.

Motivated by the nature of expected GRB afterglow decay time-scales (Kann et al. 2010), we focus only on events for which observing commenced within 16 h after the trigger. There are 58 out of 92 events fulfilling this criterion. The objective of this work is to perform a systematic search for optical counterparts to GRBs that were detected by solely by *Fermi* GBM. It also helps to inform our future GW search strategies. Therefore, we filter out five events that have joint-detection with *Swift*-BAT such that we have 53 *Fermi* events in total. The details of all events followed up with GOTO are shown in Table 3. It includes the event time and the location reported in GCN of each GRB. We also include the response time of our first follow-up observation after the trigger. The coverage percentages shown in Table 3 are calculated based on the official skymap provided by *Fermi*. The expected probability of detecting the optical afterglow for each GRB estimated based on equation (3) is also included in Table 3.

#### 3.2 Image processing

The raw images are processed with the GOTO standard pipeline (Dyer et al. 2020a), which includes bias correction, dark-frame subtraction, and co-addition of individual images. All images used to perform analyses in this paper are co-added median science images.

A stacked template image of four individual exposures is used to perform image subtraction. It acts as a reference image taken before the GRB trigger. The set of templates are updated regularly by tiling the sky on a fixed grid, as described in Section 2. We align the template to match with the science image by using python

package `spalipy`.<sup>1</sup> We use `hotpants` (Becker 2015) to perform image subtraction. Since not all fields have templates taken before the science images, we use the last observation prior to the GRB trigger as our reference image if the template does not exist. In the image subtraction step, we generate 1229 difference images in total for all 53 *Fermi* GRB events.

### 3.3 Filtering processes

After the difference imaging process, we gather all source detections in the difference images recovered by `sextractor` (Bertin & Arnouts 1996). Prior to any filtering, there are 15 049 101 detections in total among the 1229 difference images. For those GRBs with  $\approx 100$  per cent coverage, the set of images can cover more than the  $1\sigma$  region of the skymap. However, since the current observing strategy used is covering a fixed number of grids instead of covering certain proportion of the skymap, we account for all the detections in those difference images.

#### 3.3.1 Real–bogus classification

After the image processing, we apply the real–bogus classification on the difference images to separate the bogus detections and the real detections by using supervised machine learning technique (Mong et al. 2020; Killestein et al. 2021). We choose the decision boundary such that the upper limits of the false positive rate and the missed detection rate are 2 per cent and 4 per cent, respectively. After the real–bogus classification, there are 60 085 detections that are classified as ‘real’ on the difference images. All ‘real’ detections proceed to the next filtering step.

#### 3.3.2 Bad data pre-filtering

Any detection that satisfies one of the following conditions is classified as a bad detection or statistical noise;

- (i) the physical position lies within 50 pixels from the edges (the image size of GOTO is  $8176 \times 6132$  pixels with an angular resolution of  $\approx 1.2$  arcsec per pixel),
- (ii) the full width at half-maximum (FWHM) of the detection on the difference image is greater than the 95th percentile of the FWHM distribution of the science image detections,
- (iii) the signal-to-noise ratio (SNR)  $< 3\sigma$ ,
- (iv) the detection is fainter than the limiting magnitude of the science image,
- (v) the detection is brighter than 14 mag.

Of the 60 085 detections, 17 058 are classified as bad data or noise. Those detections are then rejected from our candidate list in order to reduce the time costs of cross-matching and human vetting process. Therefore, we further reduce the number of potential candidates down to 43 027 by filtering the set of ‘real’ detections using image-based characteristics.

#### 3.3.3 Catalogue cross-matching

Variable objects are commonly classified as real in the real–bogus classification (see Section 3.3.1). To effectively reject the variables and the known stellar objects from our candidates, we use PYTHON

**Table 1.** Catalogues used in cross-matching with our GRB-associated candidates.

Catalogue name	Criteria	Reference
<i>Gaia</i> -DR2	–	(1)
2MASS	–	(2)
AAVSO	–	(3)
APASS	–	(4)
PS1	$0 \geq \log  \mathcal{L}_{\text{ipsf}}  > -3$ $\text{iPSFMag} < 19.5$	(5)
UCAC4	Yale_FLAG = 0 LEDA_FLAG = 0 ExtCat_FLAG = 0 2MASSExt_FLAG = 0	(6)
IPHAS	mergedclass = $-1$ pstar $> 0.9$	(7)
SIMBAD	Non-Galaxy types	(8)

*Notes.* (1) Gaia Collaboration (2018); (2) Cutri et al. (2003); (3) Watson et al. (2017); (4) Henden et al. (2015); (5) Chambers et al. (2016); (6) Zacharias et al. (2013); (7) Barentsen et al. (2014); (8) Wenger et al. (2000).

package `CATSHTM` (Soumagnac & Ofek 2018) to perform cross-matching with other public domain data of stellar survey catalogues (see Table 1) including *Gaia*-DR2 (Gaia Collaboration 2018), 2MASS (Cutri et al. 2003), AAVSO (Watson, Henden & Price 2017), PS1 (Henden et al. 2015), UCAC4 (Chambers et al. 2016), IPHAS (Barentsen et al. 2014), and SIMBAD (Wenger et al. 2000).

GRB optical afterglows are often found close ( $\lesssim 10$  kpc) to their host galaxy (Bloom, Kulkarni & Djorgovski 2002; Fong & Berger 2013; Blanchard, Berger & Fong 2016). A search for the presence of any known galaxy within 1 arcmin of the detection can help us to prioritize those host-associated candidates. Candidates with multiple nearby galaxies are prioritized first as they have a higher chance of association with the GRB. With a matching radius of 10 kpc (95 percentile of the GRB physical offset distribution; Blanchard et al. 2016), 1 arcmin can cover down to  $z \sim 0.008$ , which is similar to the redshift of the second closest GRB, GRB980425 (Tinney et al. 1998). For those galaxies with known redshift,  $z$ , we can also estimate the isotropic equivalent energy  $E_{\text{iso}}$  of the GRBs.

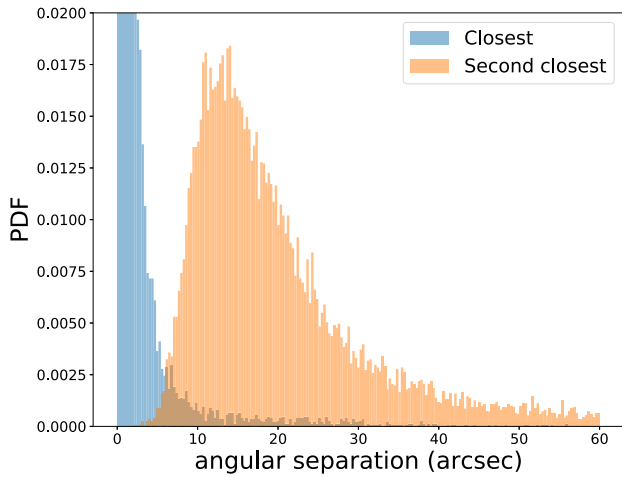
We use SIMBAD (Wenger et al. 2000) and the GLADE catalogue (Dálya et al. 2018) to perform galaxy cross-matching. SIMBAD provides reliable classification based on the physical nature of the objects. The GLADE catalogue is created by combining four galaxy catalogues, GWGC, 2MPZ, 2MASS XSC, and HyperLEDA, to achieve a high completeness up to  $\sim 40$  Mpc (Dálya et al. 2018). The completeness of the current galaxy catalogues is one of the main issue throughout GRB follow-up studies. However, it cannot be resolved easily. There are 2095 out of 43 027 candidates found next to a known galaxy within 1 arcmin. We label all those candidates with `near_galaxy=1`, otherwise `near_galaxy=0`.

To determine whether a real detection on the difference image is associated with a known stellar object from another catalogue, we set a decision boundary on the cross-matching offset. The average offset between the real detections on the difference images and the known sources from other catalogues is

$$\theta = \sqrt{\theta_{\text{ds}}^2 + \theta_{\text{sc}}^2}, \quad (1)$$

where  $\theta_{\text{ds}}$  is the astrometric offset between the difference image and the science image, and  $\theta_{\text{sc}}$  is the astrometric offset between the

<sup>1</sup><https://github.com/Lyalpha/spalipy>



**Figure 2.** Distribution of the angular separation between the detections on the difference images and the two closest detections on the science images. The *blue* histogram shows the distribution of the angular offsets of the closest sources between the science images and the difference images due to the image subtraction. The *orange* histogram shows the distribution of the angular separation from the second closest, but unrelated, source.

science image and the catalogue, respectively. Under the assumption that the astrometry from GOTO gives  $\theta_{sc} \approx 0$ , we have  $\theta \approx \theta_{ds}$ . Fig. 2 shows the distributions of  $\theta_{ds}$  and the angular separation between the detections on the difference images and the second closest detections on the science images, which they are presumably not associated with each other. We use the 95 percentile of  $\theta_{ds}$ , which is at  $\approx 6$  arcsec, to be our decision boundary for claiming the association between the detection on the difference image and the cross-matched source. However, from the *orange* histogram in Fig. 2, there are 0.7 per cent real detections having at least one more real detection within 6 arcsec. Therefore, this filtering step might cause  $\approx 0.7$  per cent false rejection rate.

We label all known stellar sources with `known_source=1`, otherwise `known_source=0`. We then only reject those candidates with `known_source=1` and `near_galaxy=0`. This helps us to filter out those known stellar objects that do not associate with known galaxy. With this filtering process, there are 12 568 candidates remain on our candidate list. Other than that, there could be a possibility that some unresolved galaxies, which are not recorded in SIMBAD and the GLADE, are misclassified as point sources in those stellar catalogues. We cross-match the candidates rejected in this step with NASA/IPAC Extragalactic Database<sup>2</sup> (NED) to obtain the misclassification rate as  $\approx 0.2$  per cent, which is estimated by the ratio between the number of the cross-matched galaxies in NED and the total number of the known stellar sources.

### 3.3.4 Minor planet checking

We pass our remaining candidates to cross-match with the Minor Planet Catalog (MPC) online.<sup>3</sup> Any candidates with a known cross-matched MP within 10 arcsec at the observing epoch are rejected. There are 10 126 detections remaining in our candidate list.

<sup>2</sup>The NASA/IPAC Extragalactic Database (NED) is operated by the Jet Propulsion Laboratory, California Institute of Technology, under contract with the National Aeronautics and Space Administration.

<sup>3</sup><https://minorplanetcenter.net/>

### 3.3.5 Multidetected filtering

To further verify whether a detection is associated with the GRB or not, we cross-match the detection with our source data base. Following the same procedure to perform cross-matching in Section 3.3.3, we associate the two cross-matched detections within 6 arcsec as the same detection.

Any candidates that have been recovered multiple times as real before the GRB trigger or far beyond the triggering time should be excluded as they are not associated with the GRB. Since GRB optical afterglows are fast transients following a power-law decay

$$F \propto t^{-\alpha}, \quad (2)$$

where  $F$  is the afterglow flux,  $t$  is the time after the trigger and  $\alpha$  is the power-law index with a typical value of  $\sim 1.2$ , we expect that they become too faint to be detected in 1–2 d after the trigger. However, as long GRBs are typically associated with SNe, they rise long after the optical afterglow has faded and show much longer decay time-scales in a few tens of days once the supernova has peaked in its emission (Woosley, Eastman & Schmidt 1999; Klose et al. 2019). We define a GRB-associated period as  $[t_0, t_0 + 100 \text{ d}]$ , where  $t_0$  is the GRB trigger time. Any candidates that have been recovered as real detections twice outside of the GRB-associated period are rejected. Finally, we have 6276 candidates left after all the automatic filtering processes.

## 3.4 Summary of automatic filtering

In the automated filtering processes, we have filtered out about 90 per cent of the real detections. The filtered detections include bad data (see Section 3.3.2 for the definitions of bad data), known sources, minor planets and GRB-unrelated sources.

## 4 MANUAL FILTERING

### 4.1 Human vetting

After automatically filtering about 90 per cent of the real detections, we start our vetting process by human inspection. Human vetting consists of three steps, the manual real–bogus classification, the selection of the potential GRB-related objects and the filtering of cosmic rays and bad data, which may have inadvertently passed the automated stages.

The first step could be subjective. We suggest that different people review each candidate multiple times to minimize subjective bias. In this work, we do not take the issue of misclassification by humans into consideration.

For the second step, we select only those detections that satisfy either one of the following criteria, there is a nearby galaxy within 10 arcsec or it does not exist on the template image. Since GRBs commonly show strong association with their hosts, any detection found next to known galaxies could be promising. The first criterion helps us to extract all detections with small angular offset from a known galaxy. For those GRB sources having angular offset  $> 10$  arcsec, they should be well separated from their host on the image, which should be included with the second criterion. The second criterion also includes any potential hostless GRB sources.

For the final step, we confirm that any candidate is not a cosmic ray by verifying the existence of the detections on all individual images that make up the median co-added science frame. For those detections that fall within the overlapping region on multiple UTs,

**Table 2.** Number of candidates filtered at each step of transient vetting.

Filtering step	Number of remaining candidates	
	Before	After
<b>Automated filtering</b>		
Real–bogus classification	15 049 101	60 085
Bad data pre-filtering	60 085	43 027
Catalogue cross-matching	43 027	12 568
Minor planet filtering	12 568	10 126
Multidetection filtering	10 126	6276
<b>Manual filtering</b>		
Human vetting	6276	116
Transient cross-matching	119	29

we also check to verify that the detections are co-located in the overlapping UTs.

Under the human vetting process, there are 116 candidates passing the above three steps. Among them, there are 55 candidates located within 1 arcmin from at least one known galaxy registered in GLADE or SIMBAD.

#### 4.2 Transient cross-matching

In this section, we describe our final filtering process. We cross-match each potential candidate with the Lasair (Smith et al. 2019) and Transient Name Server<sup>4</sup> (TNS) candidate data bases using their API. Any candidates discovered before the GRB trigger epoch should not be associated with the event. Therefore, we filter out all candidates that have been discovered before the trigger. This step is separated from the catalogue cross-matching in Section 3.3.3 because `catsHTM` does not support the Lasair and the TNS data bases. Also, we do not loop our queries through their data bases in order to avoid overloading their servers.

In addition to the filtering according to the discovery epoch, we also filter our candidates by their object types if their classification has been reported by Lasair or TNS. The filtering process excludes the object types that are very unlikely to be associated with the GRB optical counterpart, such as variable stellar objects, Type-Ia SN, or AGNs. However, we have identified two candidates, ZTF18aaegvyd<sup>5</sup> and ZTF18acaujfk,<sup>6</sup> which show re-brightening within the first day after the trigger and do not have a confident source classifications. We decide to keep those two candidates in our candidate list for further photometric analysis. After the transient cross-matching, we obtain a final list of 29 potential candidates.

## 5 RESULTS

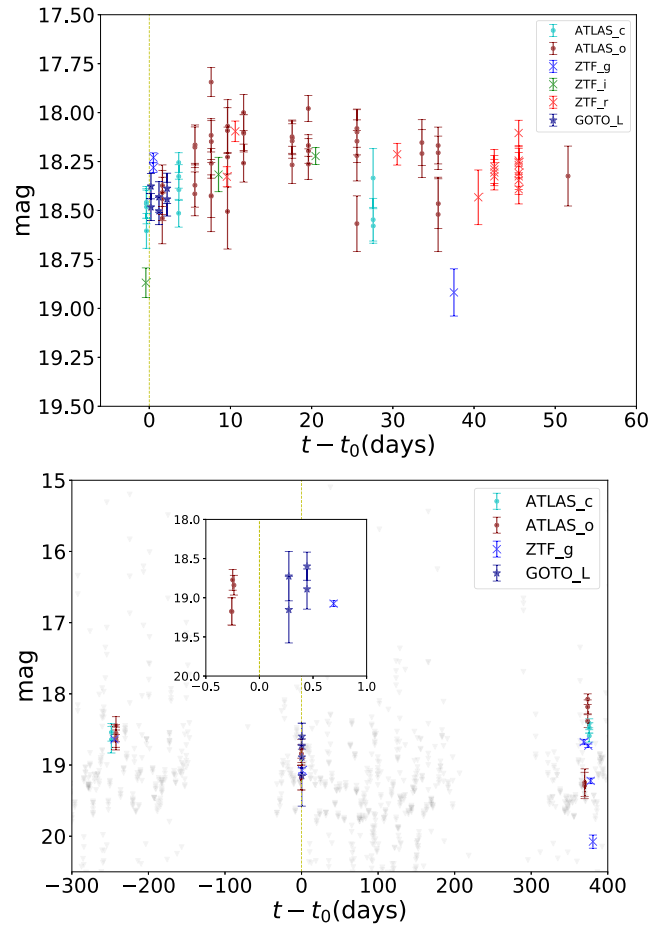
There are 29 final candidates that passed through all the filtering processes (see Table 2). In this section, we investigate the candidates primarily through photometric measurements derived from forced photometry.

We also estimate the expected number of GRB optical afterglows that could be detected for all those 53 GRB events.

<sup>4</sup><https://wis-tns.weizmann.ac.il>

<sup>5</sup><https://lasair.roe.ac.uk/object/ZTF18aaegvyd/>

<sup>6</sup><https://lasair.roe.ac.uk/object/ZTF18acaujfk/>



**Figure 3.** Light curves of ZTF18aaegvyd/SN2019env (top) and ZTF18acaujfk (bottom). The orange dashed line indicates the epoch of the GRB trigger. The grey inverted triangles indicate the limiting magnitudes of ATLAS observations. Both sources show re-brightening on the first day of the GRB trigger.

#### 5.1 Forced photometry analysis

We use the forced photometry service provided by ZTF and ATLAS<sup>7</sup> to inspect the light curves of all potential candidates. Any flux measurement with SNR greater than  $5\sigma$  is considered a detection. In order to avoid treating any subtraction residual on the difference image as real detections, we inspect all cut-out images for those detections with SNR  $> 5\sigma$ .

Among those 29 candidates, there are 2 candidates  $\geq 10$  arcsec away from known MPs, 4525 Johnbauer and 6384 Kervin. As we are unable to see any detection from both ZTF and ATLAS forced photometry data, we conclude that they are very likely minor planets. We generate the light curves for the remaining 27 candidates between 365 d before the trigger and 365 d after their trigger.

##### 5.1.1 ZTF18aaegvyd/SN2019env

As mentioned in Section 4.2, ZTF18aaegvyd re-brightened in the first day of the GRB triggers (see Fig. 3). ZTF18aaegvyd was spectro-

<sup>7</sup><https://fallingstar-data.com/forcedphot/>

scopically classified as a Type-II SN (Smith et al. 2020), SN2019env<sup>8</sup> in TNS, at  $z = 0.0235$  by the Spectral Energy Distribution Machine on the Palomar 60-in. telescope (P60). ATLAS reported the first detection of the SN about 9 h before the GRB was triggered. Since Type-II SN is not a typical origin of GRB (Galama et al. 1998; MacFadyen & Woosley 1999; MacFadyen et al. 2001), also with the pre-detection epoch that is inconsistent with the typical delay of the GRB-SN, we conclude that SN2019env was not associated with the GRB event, Fermi578679393.

### 5.1.2 ZTF18acaujfk

ZTF18acaujfk is another transient discovered by ZTF which re-brightened on the same day of the GRB trigger. However, the forced photometry data from ATLAS shows that the re-brightening had started within 1 d before the trigger time and the detection only lasted for one day (see Fig. 3). The brightness of the source seems to be stable during its active period. Due to the pre-detection from ATLAS, ZTF18acaujfk is unlikely associated with the GRB.

ZTF18acaujfk does not show any detections between 242 d before the trigger and 369 d after the trigger, except for the detection on the day of the trigger. The first detection was claimed by ZTF on 2018 October 9 with  $g = 18.3$ , which was 245 d earlier than the trigger time. However, according to the forced photometry data, it was first detected by ATLAS 33 d before the source was discovered by ZTF. More interestingly, ZTF claimed another SN detection at the same position 1 yr after the GRB trigger. Since the typical time-scale of SNe is much shorter than a year, if the classification performed by ZTF is correct (Smith et al. 2020), we expect that the source would be unrelated to the other events at the same position. However, as this classification has not been spectroscopically confirmed as a SN, we instead interpret this signal as being more consistent with a quasi-periodic variable source.

Cataclysmic variables (CVs) usually show a periodicity of several hours or repeated high-amplitude outbursts (Ritter & Kolb 2003; Kalomeni et al. 2016). Since ZTF18acaujfk was detected three times in around 2 yr, it could potentially be CV. For the galactic stellar flare, ZTF18acaujfk matches with the typical active period of a few hours. However, it should consistently be detected. We cannot confirm the transient type of ZTF18acaujfk solely by the available photometry due to its large variation of the decay time-scale from hours to days. In order to further verify the transient type of ZTF18acaujfk, we need to obtain more data by further observations and through spectroscopic classification.

### 5.1.3 Other GRB-unrelated transients

Among the remaining 25 candidates, 3 of them, which are labelled as known sources from catalogue cross-matching (see Section 3.3.3), are identified as variable stellar objects from the characteristic variability of their light curves (see Fig. 4). They initially passed our checks and are considered as candidates due to at least one known galaxy within 1 arcmin around the source.

We have also identified two unregistered variable sources. However, they show periodicities of a few hundreds of days with a smooth variability that do not look like typical variable stellar objects (Fig. 5). In addition, ATLAS shows that they are  $\sim 2\text{--}4$  mag brighter in  $o$ -filter detection than that in  $c$ -filter detection. It indicates that they are red

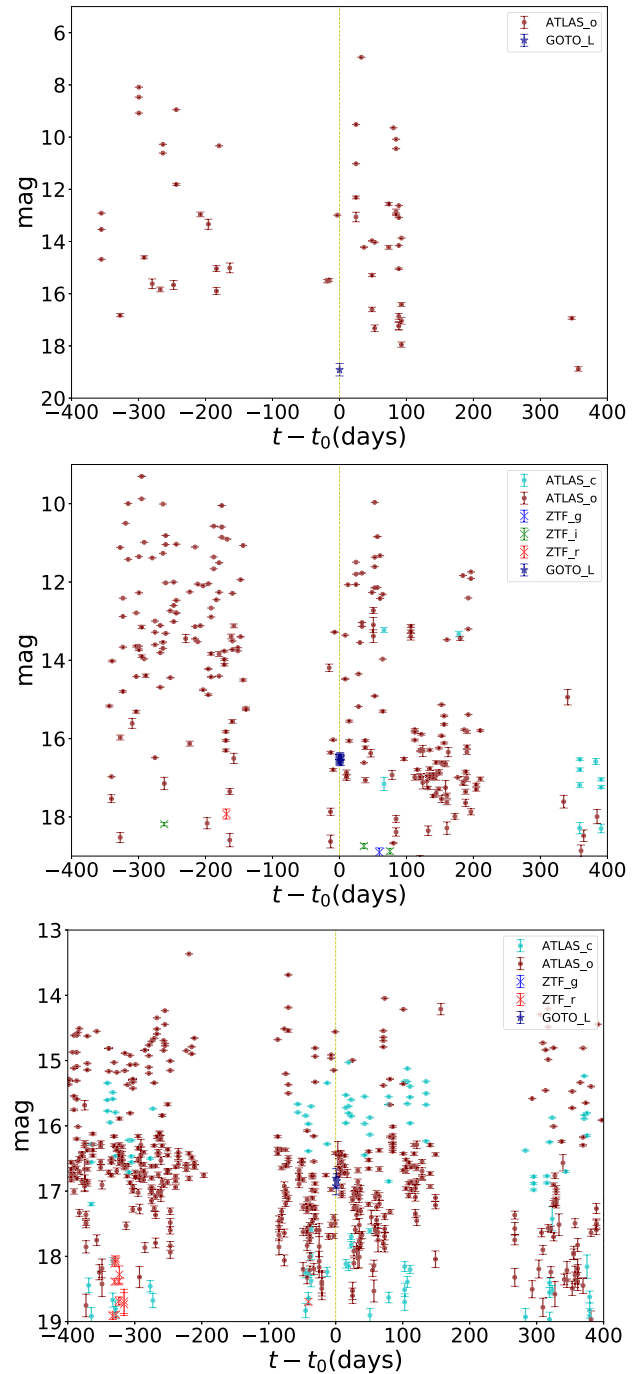


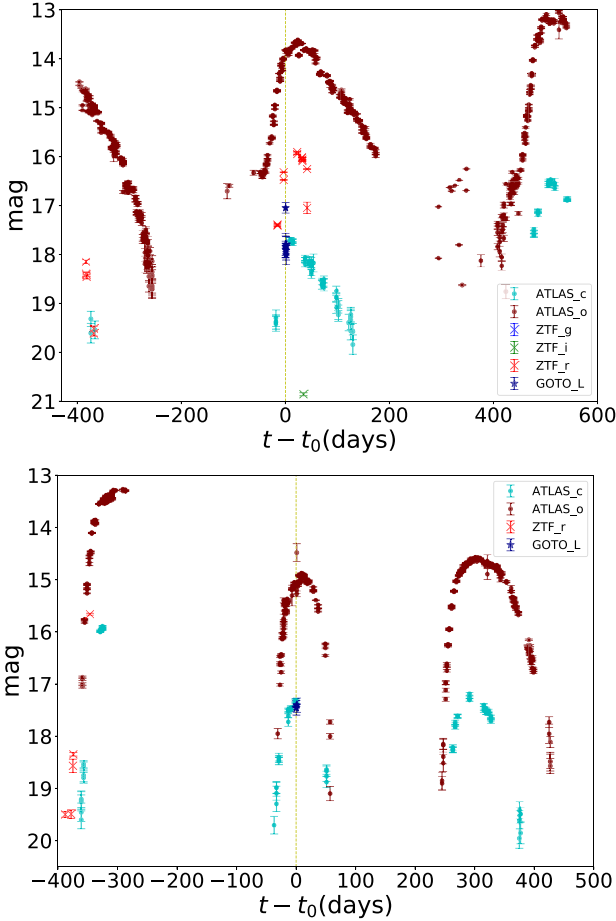
Figure 4. Light curves of the candidates that are identified as variable stars.

in colour, which match with the observational properties of Mira variables.

We found another two unknown transients, which are named as GOTO103202.04-120240.18 and GOTO062359.85-185857.69, which showed pre-detections before the triggers by ZTF and ATLAS, respectively (see Fig. 6).

GOTO103202.04-120240.18 was detected by all three instruments, GOTO, ZTF, and ATLAS. However, since the first detection obtained by ZTF was around 5 d before the GRB trigger, we conclude that GOTO103202.04-120240.18 was not related to the GRB. From the light curve generated using 2-yr forced photometry data, the

<sup>8</sup><https://www.wis-tns.org/object/2019env>



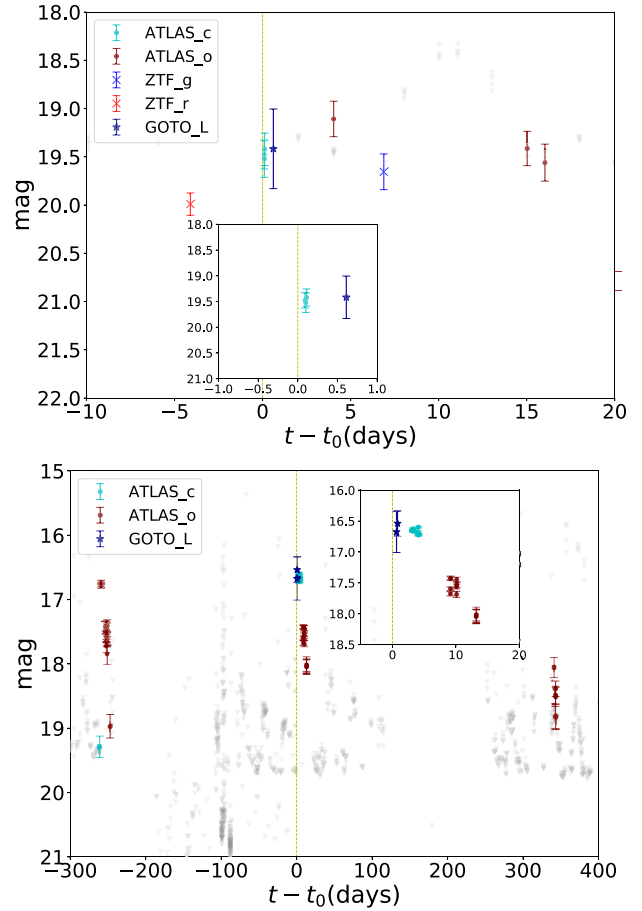
**Figure 5.** Light curves of two unregistered periodic transients. From their periodicities and colour, they are very likely to be Mira variables.

source only stayed active for  $\approx 20$  d around the GRB trigger. Also, it showed a relatively stable brightness over the entire observing period except for the first ZTF *r*-band detection. We can also see a faint extended source located at the position of GOTO103202.04-120240.18 on Pan-STARRS images. Due to the gap between the 6th and the 7th detections, and without any spectroscopic classification, we are unable to confirm the nature of GOTO103202.04-120240.18.

GOTO062359.85-185857.69 was detected by ATLAS starting  $\sim 250$  d before the GRB trigger. We do not find any point source appearing outside the active periods of the source as observed by ATLAS on both ZTF and GOTO images. However, we find a co-located source within archival Pan-STARRS images at the same position of GOTO062359.85-185857.69. The source can be seen in *g*, *r*, *i* and *z* bandpasses except *y* band, which indicates the blue colour of the source. Due to the multiple detections from ATLAS and the existence of the stellar source on Pan-STARRS images, it is unlikely to be associated with the GRB. Nevertheless, limiting magnitudes of ATLAS observations in Fig. 6 show that the source was being monitored around the period of the GRB trigger, which indicates that the re-activation of the source and the GRB triggering epoch are temporally coincident.

#### 5.1.4 Other potential candidates

For the remaining 18 candidates, we cannot draw any solid conclusion for them due to the lack of observational data. However, we can



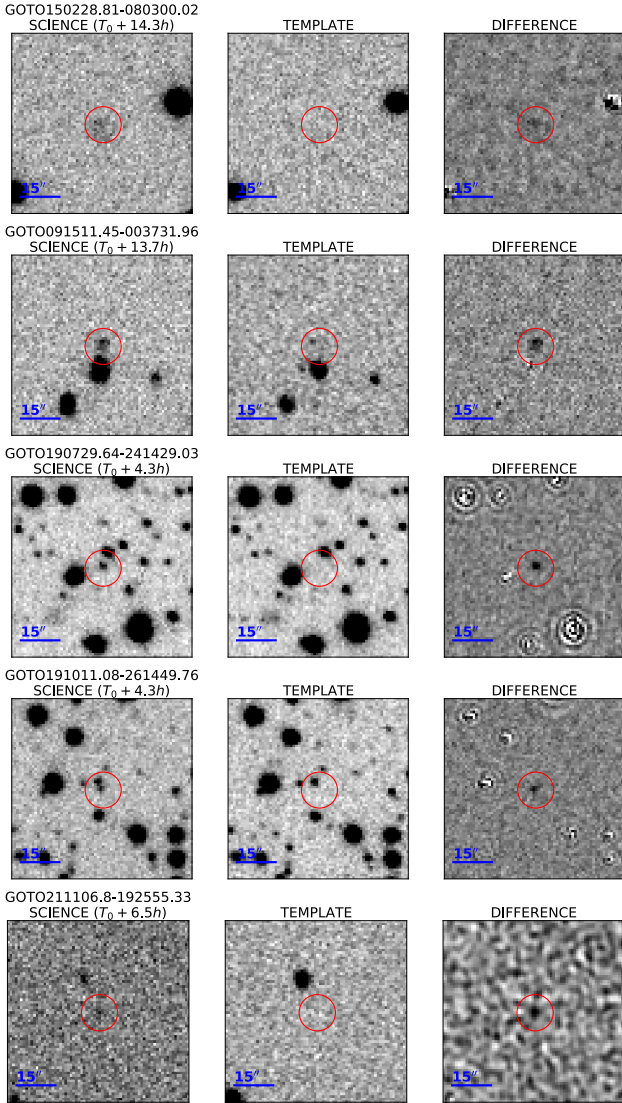
**Figure 6.** Light curves of GOTO103202.04-120240.18 (top) and GOTO062359.85-185857.69 (bottom). The orange dashed line indicates the epoch of the GRB trigger. The grey inverted triangles indicate the limiting magnitudes of ATLAS observations. Both of the sources were detected before the GRB trigger times, which imply that they are unlikely to be related to the GRB event.

still divide them into three groups according to similarities in their characteristics.

There are six candidates that fall into group 1: GOTO105852.27+491055.26, GOTO175424.66+694237.24, GOTO175722.6+692716.7, GOTO175823.15+694250.58, GOTO180217.26+695300.86, and GOTO190520.52+630533.28. All of these candidates are within 2 arcsec from the centre of a galaxy. They were not detected by either ZTF or ATLAS. As they are found with low spatial offsets from the galaxies, the resulting sources on their subtracted images do not show clear point-like structures. Therefore, at least another observation within the first day after the trigger can help us to confirm the detection and constrain the decay properties of the afterglow.

There are seven candidates classified as group 2: GOTO064003.86+294644.32, GOTO051809.33+222414.61, GOTO190511.9-260027.86, GOTO190642.62-211323.51, GOTO191337.34-244759.56, GOTO192152.79-193955.67, and GOTO190259.23-243739.33. Both ZTF and ATLAS have never shown any historical photometric detections of these seven candidates. Only a single detection or multiple detections taken at the same time were obtained by GOTO for each of them. Therefore, we cannot strongly constrain their light curves, including rise time and decay rate. However, we can see a faint point-like object near



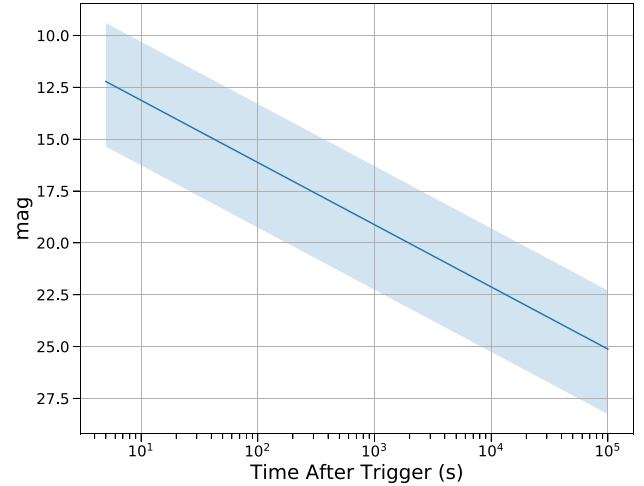


**Figure 7.**  $70 \times 70$  pixel cut-out thumbnails of the group 3 candidates. The left, the middle and the right thumbnails are from the science image, the reference image and the difference image, respectively.

each of the candidate positions on the Pan-STARRS images. Given that they are stellar objects, these candidates could be M-dwarf flares or GRB optical afterglows that are located right next to a stellar object.

Group 3 involves five candidates: GOTO150228.81-080300.02, GOTO091511.45-003731.96, GOTO190729.64-241429.03, GOTO191011.08-261449.76, and GOTO211106.8-192555.33. Group 3 candidates and group 2 candidates share the same features, except that we do not find any nearby objects on the Pan-STARRS images. This difference makes the group 3 candidates the most interesting as they are unlikely to be stellar objects (see their cut-out thumbnails in Fig. 7). However, there is still a possibility that they are unregistered MPs. To confirm whether they are GRB optical afterglows, we need extra observations to constraint their decay properties.

To conclude, all candidates mentioned in this section could potentially be the GRB optical afterglow, or other transient types. However, we do not have sufficient data to further confirm their origins.



**Figure 8.** Simulated light curve of GRB optical afterglow following power-law decay  $t^{-1.2}$ . The shaded region indicates the  $1\sigma$  confidence level.

In our future follow-ups and based on our results, all candidates that pass all of the filtering processes are deemed likely to be associated with the GRB event and deserve extra follow-up. To do this, we need to improve the latency at all stages of our vetting procedures in order to trigger spectroscopy with other facilities nearby. The GRB association can further be confirmed if we obtain a power-law spectrum. Also, extra observations on an identified host galaxy can help to constrain physical properties of the system, such as the star formation rate, metallicity, age, redshift, and spatial offsets between the host galaxy and the transient. These observables can help us to identify and improve the confidence of the progenitor origin of the candidate.

## 5.2 Expected number of candidates

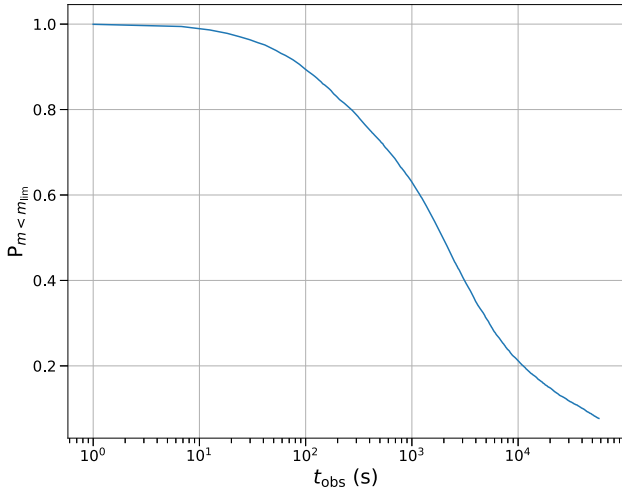
The probability of detecting an optical afterglow to a GRB event by GOTO can be estimated by using

$$P_{\text{OA}} = P_{\text{cov}} \cdot (1 - P_{\text{dark}}) \cdot P_{m < m_{\text{lim}}}(t_{\text{obs}}, \theta_{\text{obs}}). \quad (3)$$

The probability  $P_{\text{cov}}$  of covering the GRB position is simply the total coverage of the official *Fermi* healpix skymap. The term  $P_{\text{dark}}$  is the ‘dark’ GRB rate, which is assumed to be the lower limit of 0.4 in our estimation (Greiner et al. 2011).  $P_{m < m_{\text{lim}}}(t_{\text{obs}})$  is the probability that the optical afterglow is bright enough to be detected with the an observing angle  $\theta_{\text{obs}}$  at  $t_{\text{obs}}$ . The observing conditions will affect the image quality and the limiting magnitude of our observations, which could affect the probability of detecting a transient. However, in our probability estimation, we do not take the observing conditions into account.

We simulate 14 302 afterglow light curves with a power-law temporal decay  $F \propto t^{-1.2}$  using different combination of redshift  $z$  and the corrected magnitude  $m_0(z = 1, t = 86 \text{ s})$ . The samples of  $z$  and  $m_0$  are drawn from the distributions presented in Zitouni et al. (2018) and Kann et al. (2010). The simulated light curve with  $1\sigma$  confidence region is shown in Fig. 8. We can obtain  $P_{m < m_{\text{lim}}}(t_{\text{obs}})$  using the simulated light curve (see Fig. 9). Here, we assume that  $\theta_{\text{obs}} \approx 0$ . With the equation (3), the  $P_{\text{OA}}$  for each GRB event is shown in Table 3.

The total expected number of optical afterglows is estimated as the sum of  $P_{\text{OA}}$  for all events. For our sample of 53 events, we expect



**Figure 9.** Probability  $P_{m < m_{\text{lim}}}$  of detecting an optical afterglow brighter than  $m = 20$  as a function of observing time  $t_{\text{obs}}$ , assuming that the GRB is not dark and that GOTO covers the GRB location at  $t_{\text{obs}}$ .

to detect  $\lesssim 3$  afterglows as the upper limit since we do not take the weather conditions into account.

## 6 DISCUSSION AND IMPROVEMENTS

In equation (3), the factors  $P_{\text{dark}}$  and  $P_{m < m_{\text{lim}}}$  depend on the observational properties of the GRB and the limitation of the telescope, which cannot be improved manually. However, the coverage  $P_{\text{cov}}$  can be improved by optimizing the observing strategy and with the addition of more telescopes.

### 6.1 Gaussian skymap against official skymap

The skymaps used for tiling in our current GRB follow-ups are created independently using a two-dimensional normal distribution. The distribution is centred at the GRB position reported by the *Fermi* GBM with the confidence region of the quadrature sum between the statistical error and the systematic error (3.61 deg; Connaughton et al. 2015).

The advantage of using manually created skymaps is to minimize the follow-up response time. The final version of the official skymaps published by *Fermi* has a median delay time of  $\sim 600$  s. Since the GRB optical afterglow has a fast temporal decay (see the simulated light curve in Fig. 8), the early post-trigger phase is the essential period for detecting the afterglow emission. If we start follow-up observations after the *Fermi* skymap is released, and assuming that we can cover the correct position within the first 10 min, then  $\sim 30$  per cent of the GRBs would become too faint to be detected under GOTO limiting magnitude of 20 (see Fig. 9).

On the other hand, the official skymap generated by *Fermi* GBM provides a more precise localization for the GRB detection. Our generated Gaussian skymaps do not always cover the highest confidence region of the official skymap. The similarity of the two skymaps can be measured by using the overlapping index (Pastore & Calcagnì 2019)

$$\eta(f, g) = \oint \min[f(\mathbf{r}), g(\mathbf{r})] d\Omega, \quad (4)$$

where  $f$  and  $g$  are the probability distributions of the two skymaps. The index  $\eta$  closer to 1 implies higher similarity between two skymaps. Fig. 10 shows that the median value of  $\eta$  is 0.75 for all 53 events, which implies that the Gaussian and the official skymaps usually share reasonable amount of overlapping portion. We also show two skymap tiling examples with  $\eta = 0.87$  (Trigger ID: 593536021) and  $\eta = 0.53$  (Trigger ID: 578679393) in Fig. 11 to illustrate how  $\eta$  affects the tiling of the skymap.

Since we start our observations from the portion of the skymap with the highest probability, we have to confirm that the peaks between our Gaussian skymap and the official skymap do not deviate from each other by more than the size of a single tile (FoV  $\approx 20$  deg<sup>2</sup>). Fig. 12 shows that most of the offsets are consistent to within 3 deg. Therefore, the offset between the two skymaps is smaller than the size of a single sky grid.

Other than the offset, the Gaussian skymap should ideally be the same size as the official skymap. If we use the systematic error of 3.61 deg to create the intermediary Gaussian skymaps, most of will be smaller in extent than the official ones (see Fig. 13), which implies that we have underestimated the systematic error. In order to generate a Gaussian skymap with a similar size of the official one, we increase the systematic error to maximize the likelihood function

$$\log \mathcal{L}(\varepsilon_{\text{sys}}) = -\frac{1}{2} \sum_{i=1}^n \left[ \frac{x_i - y_i(\varepsilon_{\text{sys}})}{\sigma_i} \right]^2, \quad (5)$$

where  $x_i$  is the number of pixels on the healpix grid (Górski et al. 2005), which is a pixelization algorithm of 2-sphere with equal area for each individual grid, of order=NESTED as defined by the *Fermi* skymap and  $y_i(\varepsilon_{\text{sys}})$  is the number of healpix pixels on the Gaussian skymap generated with a systematic error  $\varepsilon_{\text{sys}}$ . We apply  $\sigma_i = \sqrt{x_i}$  in the likelihood. The optimal value of  $\varepsilon_{\text{sys}}$  is 5.6 deg. With the optimal  $\varepsilon_{\text{sys}}$ , the median value of  $\eta$  increases from 0.75 to 0.82 (see Fig. 10), which indicates that the overall similarity between the Gaussian skymap and the official skymap has been improved.

Despite the high  $\eta$ , the official skymap cannot be reproduced perfectly by a simple two-dimensional Gaussian distribution. To solve this problem, the manually created Gaussian skymap will be replaced by the official skymap once it is published, and the typical creation time of the official skymap is  $\approx 10$  mins. The observation schedule will then be updated in response.

### 6.2 Tiling strategy

The size of the official skymaps shows a wide range of variation depending on the uncertainty of the *Fermi* GBM localization. Even with a high visibility for some events (see Table 3), the coverages within 16 h are still low, which indicates that only covering the top five tiles is not enough in our follow-up observations. The median size of the official skymap at  $1\sigma$  confidence level is  $\approx 150$  deg<sup>2</sup>. With the GOTO prototype FoV of 20 deg<sup>2</sup>, we need at least 8 tiles to cover 68 per cent of the skymap.

The solution to this problem is straightforward. Instead of tiling a fixed number of tiles, we should tile a fraction, e.g. 68 per cent, of the skymap. With this strategy, we can guarantee to cover a reasonable portion of the skymap even with a large uncertainty. However, with this strategy, we require that a minimum fraction, say 68 per cent, of the skymap can be covered within a certain amount of time. The time constraint should not be set for too long due to the fast-decay nature of the afterglow. According to Fig. 9, we estimate that 80 per cent of the afterglows would fall below the limiting magnitude of GOTO within 2.8 h.

**Table 3.** *Fermi* GRB event table.

Trigger number	Event time (UTC)	RA (hh:mm:ss)	Dec. (dd:mm:ss)	Error (deg)	Response time (h)	Skymap coverage/visibility <sup>a</sup> (per cent)	Expected brobability <sup>b</sup> (per cent)
572876510	2019-02-26T12:21:46	14:57:43.2	-08:36:36	5.11	13.911	23.6/87.2	1.2
573284727	2019-03-03T05:45:22	19:55:48	+29:51:00	23.97	0.476	22.7/60.7	7.2
573604668	2019-03-06T22:37:43	15:24:16.8	-00:22:48	2.55	0.0392	87.3/100	45.2
574345067	2019-03-15T12:17:42	09:46:14.4	-11:09:00	2.11	8.146	71.8/100	5.2
574676902	2019-03-19T08:28:18	12:50:24	-04:48:36	3.26	14.373	100/100	5.0
575018216	2019-03-23T07:16:52	09:28:16.8	+02:50:24	2.07	13.478	60.9/100	3.2
576241792	2019-04-06T11:09:47	19:05:21.6	+61:30:00	7.09	14.557	32.5/96.5	1.6
576265958	2019-04-06T17:52:33	19:30:33.6	+26:47:24	5.46	9.077	19.3/100	1.3
578252995	2019-04-29T17:49:51	13:20:12	-08:00:00	8.00	3.121	3.7/99.8	0.4
578679393	2019-05-04T16:16:28	09:13:57.6	+28:41:24	9.61	5.045	17.9/99.8	1.7
578711654	2019-05-05T01:14:09	22:21:33.6	+42:10:48	9.54	2.391	10.5/82.8	1.4
578903308	2019-05-07T06:28:23	10:23:50.4	-12:48:00	4.82	14.583	53.0/99.6	2.6
578963794	2019-05-07T23:16:30	19:11:16.8	-22:49:12	1.19	3.567	54.2/93.8	5.9
579036175	2019-05-08T19:22:50	11:51:50.4	+23:31:48	2.12	1.688	17.1/100	2.8
579814215	2019-05-17T19:30:10	18:00:04.8	+25:46:12	1.20	3.250	77.7/100	9.1
580437952	2019-05-25T00:45:48	22:32:04.8	+05:27:00	4.04	2.712	12.7/86.5	1.6
580904353	2019-05-30T10:19:08	08:03:02.4	+35:30:00	10.20	11.028	25.3/30.8	1.5
581068049	2019-06-01T07:47:24	10:51:55.2	+54:35:24	8.28	13.581	25.3/99.8	1.3
581281470	2019-06-03T19:04:26	01:20:19.2	+40:54:36	5.86	9.579	13.0/30.8	0.9
581337762	2019-06-04T10:42:37	22:50:12	+46:22:12	1.00	14.974	51.5/54.5	2.5
581469752	2019-06-05T23:22:27	22:28:57.6	+04:47:24	6.76	3.691	9.5/100	1.0
581882394	2019-06-10T17:59:50	21:49:31.2	+42:25:12	1.21	6.603	50.2/100	4.0
581889628	2019-06-10T20:00:24	20:59:19.2	-15:55:48	11.23	5.798	8.1/92.7	0.7
582004649	2019-06-12T03:57:25	14:55:48	+62:06:00	9.30	0.190	10.1/61.3	4.1
582304592	2019-06-15T15:16:27	12:45:36	+49:22:48	2.32	6.748	47.0/100	3.7
582596766	2019-06-19T00:26:02	23:17:14.4	+12:51:36	2.05	1.916	50.4/100	7.8
582725415	2019-06-20T12:10:11	10:48:19.2	+30:28:48	1.16	9.348	100/100	6.7
584590606	2019-07-12T02:16:42	19:13:33.6	+56:09:00	7.59	0.186	13.3/100	5.5
585007213	2019-07-16T22:00:08	23:02:31.2	-00:49:48	6.43	2.811	0.02/100	0
585559462	2019-07-23T07:24:18	19:17:52.8	+25:13:12	10.47	13.988	5.9/100	0.3
585847498	2019-07-26T15:24:54	20:41:02.4	+34:17:24	1.17	5.944	100/100	8.5
592297741	2019-10-09T07:08:57	01:47:14.4	+65:43:48	1.41	12.773	97.6/100	5.4
593045905	2019-10-17T22:58:21	08:58:12	+15:19:48	12.39	5.196	0.7/99.8	0.1
593176520	2019-10-19T11:15:15	07:18:14.4	+62:05:24	3.28	13.831	19.6/93.2	1.0
593419810	2019-10-22T06:50:05	18:12:33.6	-23:06:00	3.18	12.870	21.1/22.1	1.1
593536021	2019-10-23T15:06:57	19:41:02.4	+06:10:48	1.82	4.573	73.0/100	7.2
593928606	2019-10-28T04:10:01	21:16:48	-11:17:24	8.75	15.459	20.0/99.6	0.9
593964489	2019-10-28T14:08:04	18:27:31.2	+69:59:24	7.67	6.0339	49.2/97.8	4.2
596387570	2019-11-25T15:12:46	23:34:09.6	+18:12:00	4.13	4.207	38.3/100	3.9
596786686	2019-11-30T06:04:41	23:17:36	+63:04:48	2.27	14.151	35.4/99.9	1.8
597955752	2019-12-13T18:49:08	22:04:14.4	-13:56:24	6.67	0.665	36.0/94.3	14.0
598951521	2019-12-25T07:25:17	06:21:57.6	-17:21:00	6.66	15.246	35.7/82.3	2.2
598988276	2019-12-25T17:37:52	09:43:12	-07:10:48	2.84	7.410	100/100	7.6
600448273	2020-01-11T15:11:08	06:57:57.6	+31:43:12	3.47	11.244	88.6/100	5.9
600525396	2020-01-12T12:36:31	10:00:31.2	+64:24:36	2.00	9.431	90.3/100	6.5
601677816	2020-01-25T20:43:31	00:29:48	+64:41:24	1.00	0.202	100/100	40.6
601841483	2020-01-27T18:11:19	05:03:33.6	+20:04:12	3.26	1.740	45.5/100	11.8
603142206	2020-02-11T19:30:01	07:24:33.6	+59:00:36	10.89	0.741	17.1/99.8	6.4
603849435	2020-02-19T23:57:10	17:37:55.2	+08:23:24	1.00	4.818	87.5/100	9.6
610450873	2020-05-06T09:41:09	12:43:40.8	+40:14:24	3.47	12.422	7.2/100	0.5
610800081	2020-05-10T10:41:17	10:20:19.2	-01:55:48	1.50	10.659	63.7/100	5.8
611434353	2020-05-17T18:52:28	07:40:48	+29:25:48	7.06	2.329	6.1/95.3	2.0
613212114	2020-06-07T08:41:50	11:26:00	+30:52:48	2.10	12.744	67.0/100	4.1

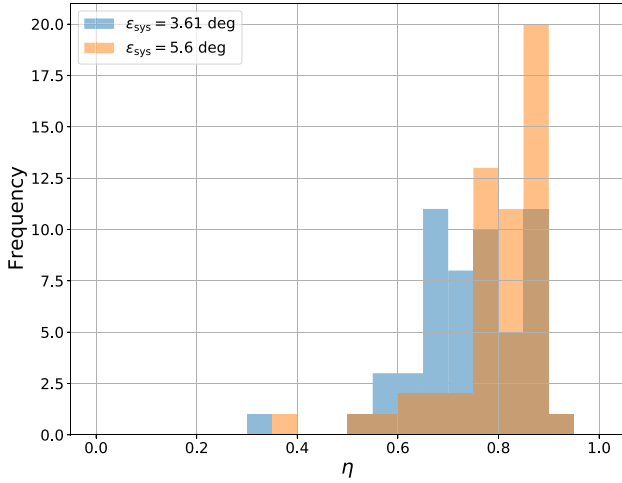
<sup>a</sup> Calculated with the time constraints of 16 h after the trigger. Official *Fermi* skymap is used.<sup>b</sup> The probability of detecting the GRB optical afterglow estimated based on equation (3).

### 6.3 Follow-up cadence

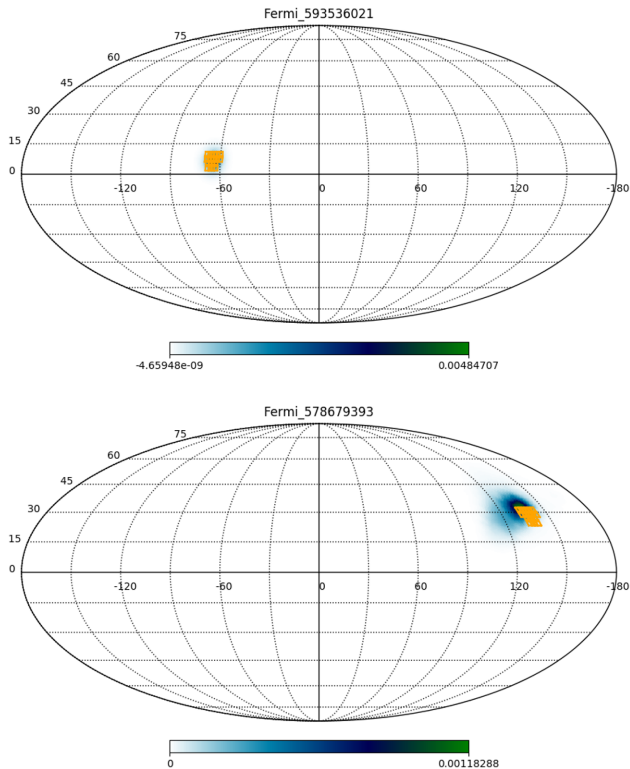
With the current configuration of GOTO, it takes  $4 \times 90$  s of exposure per each frame. Based on our five-tile observing strategy, it takes 30 min to complete the whole tiling process. Fig. 8 shows that GRB

optical afterglow can decay up to  $\sim 10$  mag within the first 30 min after the trigger.

In order to verify the fast-decay nature of GRB optical afterglows, if we can trigger the first follow-up observation within 30 min

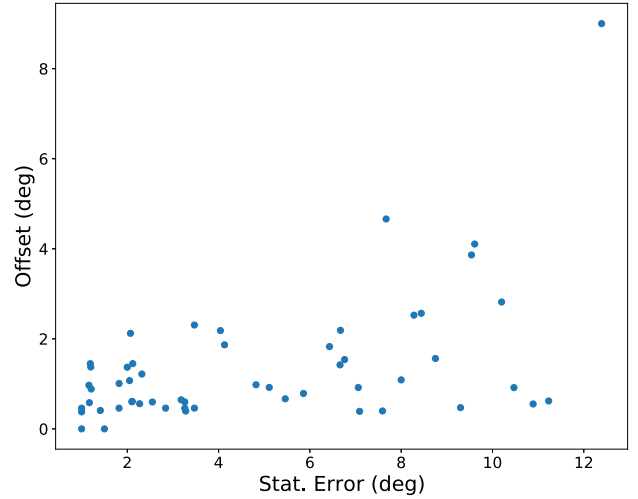


**Figure 10.** Histogram of the overlapping indices  $\eta$  for all 53 *Fermi* GBM GRB events. The orange and the blue histograms represent the skymap created by using the systematic errors of 3.61 and 5.60 deg, respectively. By optimizing for the systematic error, the median value of  $\eta$  has been improved from 0.75 to 0.82.

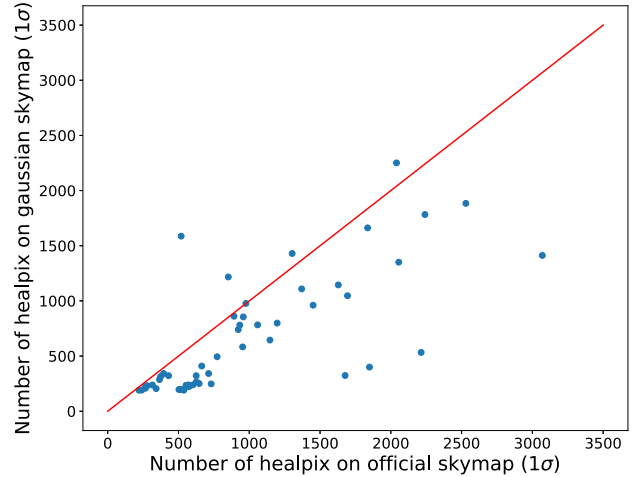


**Figure 11.** Examples of skymap tiling. The blue area represents the official skymap and the orange grids represent the observed tiles generated based on the Gaussian skymap. The top skymap shows a good tiling case. On the other hand, the observed tiles on the bottom skymap show a significant deviation from the centre of the skymap.

after the trigger, we should start our second follow-up observation immediately afterwards. Since the afterglow decay follows the power-law function, the magnitude should decay much slower on the linear time-scale after the first hour of the trigger. To estimate a



**Figure 12.** Offset between the peaks of the Gaussian skymap and the official skymap against the *Fermi* reported statistical error. The offsets mostly lie within 3 deg and show no correlation with the statistical error.



**Figure 13.** Correlation between the numbers of healpix pixels on the Gaussian skymap (with the systematic error of 3.61 deg) and the official skymap at  $1\sigma$  confidence level. The red line indicates the number of healpix pixels on both skymaps are equal. Most of the data points lie below the red line indicating the systematic error is underestimated.

proper cadence of the following observations, we use

$$\Delta m = -2.5 \log \left( \frac{t}{t_0} \right)^{-\alpha}, \quad (6)$$

where  $\Delta m$  is the magnitude difference between time  $t$  and  $t_0$ . By taking  $\Delta m = 1$  and  $\alpha = 1.2$ , we obtain  $t \approx 3.3t_0$ , which is our estimate of the cadence for the subsequent observation following the first observation taken at time  $t_0$ .

## 7 CONCLUSION

From 2019 February to 2020 June, the GOTO prototype followed up 93 GRBs detected by *Fermi* GBM. We conduct an archival search on 53 of the them followed up within 16 h and without *Swift* joint detections.

We have developed a process to effectively filter detections that are unlikely to be related to a GRB. We divide the process into two parts, automated and manual. The automated process involves real-bogus classification, bad data pre-filtering, catalogue cross-matching, minor planet checking, and multidetection filtering. These automated processes narrow down our candidate number from 60 085 to 6276. The manual filtering process involves human vetting, transient cross-matching, and forced photometry analysis. In the human vetting step, we finally obtain 119 transient-like objects, 55 of which have nearby galaxies. Transient cross-matching further helps us filter out some GRB-unrelated sources, ultimately resulting in 29 potential candidates.

We apply forced photometry on those 29 candidates using data obtained by GOTO, ZTF, and ATLAS. We analyse their light curves and find that 11 of them are unlikely to be related to any GRB. For the rest of the candidates, they can potentially be associated as a GRB afterglow. However, we do not have enough data to solidly constrain their transient types.

We expect to detect  $\approx 3$  GRB optical afterglows among those 53 events. The estimation is calculated by simulating the light curve of the afterglow. The skymap coverage of each event and the dark GRB rate are also considered in the calculation.

In order to improve our performance, we recommend using 5.6 deg as the systematic error to create a Gaussian skymap for tiling before the official skymap provided by *Fermi* is released. Once the official skymap is released, which should take  $\sim 10$  min after the trigger, we should use this updated skymap for tiling. For the tiling strategy, we should image at least  $1\sigma$  of the visible skymap to ensure that a reasonable portion of the skymap is covered. If the first follow-up observation can be triggered within 30 min after the GRB alert, we also recommend multiple observations. In order to verify the nature of fast decay, a second follow-up should be performed immediately following the first one.

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## DATA AVAILABILITY

Data files covering the system throughput and some of the software packages are available via public github repositories under <https://github.com/GOTO-OBS/>. Prototype data were mainly used for testing and commissioning and a full release of all data is not foreseen. Some data products will be available as part of planned GOTO public data releases.

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