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Arecibo observations of a burst storm from FRB 20121102A in 2016

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ABSTRACT

FRB 20121102A is the first known fast radio burst (FRB) from which repeat bursts were detected, and one of the best-studied FRB sources in the literature. Here we report on the analysis of 478 bursts (333 previously unreported) from FRB 20121102A using the 305-m Arecibo telescope – detected during approximately 59 hours of observations between December 2015 and October 2016. The majority of bursts are from a burst storm around September 2016. This is the earliest available sample of a large number of FRB 20121102A bursts, and it thus provides an anchor point for long-term studies of the source's evolving properties. We observe that the bursts separate into two groups in the width-bandwidth-energy parameter space, which we refer to as the low-energy bursts (LEBs) and high-energy bursts (HEBs). The LEBs are typically longer duration and narrower bandwidth than the HEBs, reminiscent of the spectro-temporal differences observed between the bursts of repeating and non-repeating FRBs. We fit the cumulative burst rate-energy distribution with a broken power law and find that it flattens out toward higher energies. The sample shows a diverse zoo of burst morphologies. Notably, burst emission seems to be more common at the top than the bottom of our 1150–1730 MHz observing band. We also observe that bursts from the same day appear to be more similar to each other than to those of other days, but this observation requires confirmation. The wait times and burst rates that we measure are consistent with previous studies. We discuss these results, primarily in the context of magnetar models.

Key words: methods: data analysis – (transients:) fast radio bursts.

1 INTRODUCTION

Fast radio bursts (FRBs) are flashes of radio emission with exceptionally high isotropic equivalent luminosity that usually last no longer than a few tens of milliseconds and originate at extragalactic distances (Lorimer et al. 2007; Thornton et al. 2013; Cordes & Chatterjee 2019; Petroff, Hessels & Lorimer 2019, 2021). Hundreds of FRBs have been detected to date (see the Transient Name Server¹ or Petroff et al. 2016; The CHIME/FRB Collaboration 2021), but only approximately 20 of these sources are known to repeat (e.g. Spitler et al. 2016; Shannon et al. 2018; Kumar et al. 2019; CHIME/FRB Collaboration 2019a, b; Fonseca et al. 2020). Over a dozen FRBs (both repeating as well as apparently non-repeating) have been localized to a variety of host galaxies (Bhandari et al. 2021), but the emission mechanism and physical origin remain enigmatic. While the repeaters rule out cataclysmic models for at least some fraction of the FRB population, the models that have been proposed for FRBs span a wide range of astrophysical scenarios (for a catalogue of models see Platts et al. 2019). Recently, there has been the noteworthy detection of an extremely bright radio burst from the Galactic magnetar SGR J1935+2154 (Bochenek et al. 2020; CHIME/FRB Collaboration 2020b), which has strengthened the case for magnetars as the engines of FRBs. Even so, the discovery of a nearby repeating FRB in a globular cluster shows that not all FRBs can be young magnetars formed via core collapse (Kirsten et al. 2022) – though they may be, in some cases, young magnetars formed via accretion-induced collapse or binary merger.

It remains unclear whether the repeating and apparently nonrepeating FRBs come from physically distinct source types. Recently, it has been shown that the burst properties of repeating and nonrepeating FRBs are statistically different, with repeaters on average exhibiting larger temporal widths and narrower bandwidths (Pleunis et al. 2021a; The CHIME/FRB Collaboration 2021). Additionally, many repeater bursts display complex time-frequency downwarddrifting structure (the 'sad-trombone' effect; Hessels et al. 2019). Another commonality that seems to be emerging among repeating sources, is near 100 per cent linear polarization, a very low circular

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polarization fraction and polarization position angles (PPAs) that remain approximately constant between and during bursts (Michilli et al. 2018b; Nimmo et al. 2021). While there have been a small number of detections of circular polarization in repeater bursts (Hilmarsson et al. 2021a; Kumar et al. 2022) and swinging PPAs (Luo et al. 2020), the non-repeating FRB population shows more diverse polarimetric properties (e.g. Day et al. 2020). These differences in the burst properties of repeaters and apparent non-repeaters could indicate different emission mechanisms, different progenitors, and/or different local environments.

FRB 20121102A is the first FRB source from which multiple bursts were detected (Spitler et al. 2014, 2016). This repeating nature allowed for follow-up observations that localized the source to near a star-forming region in a dwarf galaxy at redshift z = 0.193(Bassa et al. 2017; Chatterjee et al. 2017; Tendulkar et al. 2017) and, moreover, a milliarcsecond association of the bursts with a persistent compact radio source (Chatterjee et al. 2017; Marcote et al. 2017). The most recent dispersion measure (DM) values of the bursts from FRB 20121102A are $563.5 \pm 0.8 \text{ pc cm}^{-3}$ on MJD 57836 (Platts et al. 2021) and 565.8 \pm 0.8 pc cm⁻³ for the period between MJD 58724 and MJD 58776 (Li et al. 2021). Li et al. (2021) also estimated that this DM is increasing by approximately $1 \text{ pc cm}^{-3} \text{ yr}^{-1}$. Moreover, there is an evidence that this secular DM increase is non-linear (Jahns et al. 2022,; Seymour et al., in preparation). Michilli et al. (2018b) showed that FRB 20121102A has an extremely large and variable rotation measure (RM $\sim 10^5$ rad m⁻²), demonstrating that the bursts originate from a source embedded in an extreme and dynamic magneto-ionic environment. In contrast to the DM, the RM of the bursts is decreasing by an average of ~ 15 per cent per year (Hilmarsson et al. 2021b). Plavin et al. (2022) showed that the source depolarizes towards lower frequencies, possibly because of minor non-uniformities in the Faraday screen (they find that the Faraday width of the burst environment is only approximately 0.1 per cent of the total Faraday rotation).

FRB 20121102A is in many ways the prototypical repeating FRB. As with other repeaters, the burst profile and spectrum vary significantly from burst to burst. Furthermore, the burst activity of FRB 20121102A appears to be frequency dependent, an aspect that has not been well studied in most other sources (though see, e.g. Pleunis et al. 2021b). While thousands of bursts were detected in less than two months in the 1-2 GHz range (Li et al. 2021), only one burst has been detected in the 400-800 MHz range (more specifically at ~600 MHz; Josephy et al. 2019) despite years of nearly daily monitoring by the Canadian Hydrogen Intensity Mapping Experiment Fast Radio Burst Project (CHIME/FRB). On the other extreme, the highest frequency at which bursts have been detected from FRB 20121102A (and from any FRB for that matter) is ~8 GHz (Gajjar et al. 2018). From a temporal point of view, the bursts from FRB 20121102A are clustered in time (Oppermann, Yu & Pen 2018), but show no evidence of a rotational period underlying their times-of-arrival (Zhang et al. 2018; Aggarwal et al. 2021). Analysis of burst data over five years has, however, resulted in an activity period of ~160 d and a duty cycle of ~55 per cent (Rajwade et al. 2020; Cruces et al. 2021). FRB 20121102A is one of only two known FRB sources with a long-term periodicity: The other being FRB 20180916B, which has a well-established activity period of 16.35 d with an activity window of \sim 5 d. (Chime/Frb Collaboration 2020a).

Various models have been proposed to explain the repeating nature and periodic activity of some FRBs, typically by either invoking precession of a neutron star (NS; e.g. Levin, Beloborodov & Bransgrove 2020; Sob'yanin 2020; Yang & Zou 2020; Zanazzi & Lai 2020), the rotational period of a NS (e.g. Beniamini, Wadiasingh & Metzger 2020, although such slow-rotating magnetars have not yet been observed) or the orbital period of a NS in a binary system with another astrophysical object (e.g. Gu, Yi & Liu 2020; Lyutikov, Barkov & Giannios 2020; Du et al. 2021; Sridhar et al. 2021; Wada, Ioka & Zhang 2021; Kuerban et al. 2022). The literature is also divided on the location of the emission region, assuming a NS origin: i.e. whether this is within the magnetosphere of the NS (e.g. Kumar, Lu & Bhattacharya 2017), or much further out beyond the magnetosphere as the result of relativistic shocks (e.g. Metzger, Margalit & Sironi 2019).

The spectro-temporal and polarimetric properties of the bursts, as well as their energetics and the evolution of these factors, are all valuable probes for constraining and differentiating between progenitor models. The short time intervals between bursts, as well as the high burst rate detected by FAST (Five hundred meter Aperture Spherical Telescope) favour those models where burst genesis occurs in the magnetosphere (Li et al. 2021). Burst structure down to $\sim 4 \,\mu s$ and ~60 ns is detected in FRB 20180916B and FRB 20200120E, respectively (Nimmo et al. 2021, 2022). Ignoring relativistic effects, these durations correspond to an emission region from $\sim 100 \,\mathrm{m}$ to a few km in size, which arguably supports magnetospheric models (Beniamini & Kumar 2020). The sad-trombone effect, on the other hand can be explained as a radius-to-frequency mapping effect in magnetospheric models (Lyutikov 2020), but is also a natural consequence of models invoking a synchrotron maser (e.g. Metzger et al. 2019; Margalit et al. 2020).

Through a monitoring proposal (P3054; PI: L. Spitler), the Arecibo Observatory's 305-m telescope has been used to detect over a thousand bursts from FRB 20121102A since its discovery (see also Jahns et al. 2022). In this paper, we present analysis of the earliest observed burst storm of FRB 20121102A. In Section 2, we describe the observations and our search analysis used to discover the bursts presented in this work. In Section 3, we present the search results and explain how the burst properties have been measured. We describe how we classify bursts as either 'low-energy bursts' or 'high-energy bursts', and evaluate the statistical significance of this classification. We also explain our procedures of fitting power laws to the energy distribution of the detected bursts. In Section 4, we investigate how the classification of low- and high-energy bursts relates to what is known from other FRBs. Furthermore, we discuss the results of the energy and spectro-temporal properties of this sample of bursts. Finally, in Section 5 we summarize our main findings.

2 OBSERVATIONS AND SEARCH FOR BURSTS

We observed FRB 20121102A with the 305-m William E. Gordon Telescope at the Arecibo Observatory using the L-wide receiver. In this paper, we report on 56 observations that sum to ~59 hr on source. These observations were conducted on different days, sometimes consecutive, between November 2015 and October 2016, and are summarized in Table 1. The data were recorded using the Puerto-Rican Ultimate Pulsar Processing Instrument (PUPPI) backend recorder,² which records from 980–1780 MHz and provides a time resolution of 10.24 μ s and 512 frequency channels of width 1.56 MHz. The L-wide receiver has a smaller observing frequency range of 1150–1730 MHz, and the data outside this frequency range are excluded from analysis. These data were coherently dedispersed within a channel to a DM of 557.0 pc cm⁻³, the best

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<sup>2</sup>http://www.naic.edu/puppi-observing/
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Table 1. Observation log.

Topocentric Start MJD	Duration (s)	Nr of Bursts	Burst Rate (hr^{-1})
57342.225104167	5753.9	0	0
57343.223344907	6747.7	0	0
57344.224456019	6421.4	0	0
57345.227222222	3000.1	0	0
57345.263599537	1300.3	0	0
57345.280393519	971.9	0	0
57352.220868056	2705.5	0	0
57364.196805556	245.4	1	14.7
57381.179918981	1534.4	0	0
57388.099490741	5858.8	0	0
57391.152835648	1200.8	0	0
57391 131307870	1700.2	0	0
57410 039421296	6485.3	0	0
57413 040474537	6058.0	0	0
57418 019942130	6362.4	0	0
57420 016736111	465.2	0	0
57420.023229167	524.2	0	0
57420.0292929107	606.5	0	0
57420.029490741	605.9	0	0
57420.030078241	508.4	0	0
57420.050083706	603.2	0	0
57420.058148148	602.0	0	0
57420.055280352	619.2	0	0
57428 086134250	6774.8	0	0
57/20 088370630	6/87 5	0	0
57504 810814815	625.5	0	0
57504 833035185	2175.1	0	0
57506 80/212063	4262.7	1	08
57510 801192130	3570 1	1	4.0
57511 836574074	292.8	-	4.0
57512 759571759	6709.4	0	0
57513 786701380	4125.2	1	0.87
57514 809259259	1945 5	0	0
57523 730046296	6659.5	0	0
57571 597083333	6831.0	0	0
57578 579027778	6661.3	0	0
57586 561643519	6351.8	0	0
57594 541250000	6232.9	0	0
57607 507083333	6326.2	10	57
57614 479687500	7179.5	48	24.1
57621 460729167	6818 5	0	0
57628 446423611	3615.3	44	43.8
57638 448877315	3831.5	39	36.6
57640 412835648	808 3	10	44 5
57640 435451389	307.4	1	11.5
57640 464143519	2277 3	10	15.8
57641 438032407	4059.9	34	30.1
57642,439664352	3683.7	8	7.8
57644 407719907	5967.1	56	33.8
57645.409861111	5545 3	76	49.3
57646 394166667	3751.5	24	23.0
57648.393437500	6036.4	29	17.3
57649.394710648	5672.7	26	16.5
57650,380810185	6595.3	15	8.2
57666.338912037	6723.1	41	22.0
57671.374085648	2501.7	0	0
Total	212386.1	478	_

DM measurement at the time of the observations, which effectively reduces intra-channel dispersive smearing to $<30 \,\mu s$ given a true DM of 560.5 pc cm⁻³ (Hessels et al. 2019).

Prior to searching for bursts, we downsampled the data using psrfits_subband,³ resulting in 64 channels of 12.5 MHz and a time resolution of $81.92 \,\mu$ s. Using PRESTO⁴ (Ransom 2001), the data were searched for bursts by first incoherently dedispersing the time series using prepsubband, for 77 equally spaced trial DMs in the range 554.90–570.10 pc cm⁻³ (steps of 0.2 pc cm⁻³). No radio frequency interference (RFI) masks were applied, in order to prevent masking very bright bursts. After this, the de-dispersed time series were searched with single_pulse_search.py, which implements matched boxcar filtering. Our search was most sensitive to bursts with boxcar widths between $81.92 \,\mu s$ and $24.58 \,m s$. We searched down to the events with a S/N of six, as provided by single_pulse_search.py. The detected events were grouped into burst candidates using SpS⁵ (Michilli et al. 2018a). Events within 20 ms and 1 pc cm⁻³ are grouped together at first. This chosen time window for grouping is motivated by the typical burst duration. Since each grouped burst is manually inspected over a larger time range, it is then possible to identify these bursts as multiple bursts, and/or to find additional weaker bursts that are closely spaced in time.

The candidates were then classified with FETCH⁶ (Fast Extragalactic Transient Candidate Hunter; Agarwal et al. 2020), a supervised deep-learning algorithm that uses deep neural networks to classify burst candidates. FETCH has been trained on simulated FRBs and real RFI from the Green Bank Observatory's (GBT) L-band data (comparable frequencies to the Arecibo data presented in this paper). There are eleven models that all perform well on GBT data. and have also proven very successful at detecting FRBs with S/N >10 in ASKAP (Australian Square Kilometre Array Pathfinder) and Parkes data. Since the Arecibo Observatory has additional disparate sources of RFI, and since we also probe to a lower S/N than the training data, we first evaluated the performance of the different models through manual inspection using previously searched data sets which contained 41 known bursts (Gourdji et al. 2019). We found that model H performed the best at identifying true bursts. We managed to recover all bursts except for B33 (which is missed by all available models) from Gourdii et al. (2019). In Jahns et al. (2022). where their data are very similar⁷ to those presented in this paper, they calculated the completeness of FETCH. When compared to a completely human-labeled classification, they also found that model H misses the least amount of bursts. In their analysis, they found that 90 per cent of the bursts that are missed have a S/N < 8 – the minimum S/N that FETCH has been trained on. They also find that, in their case, 34 per cent of the bursts with S/N < 8 deemed real (by eye) score a probability p < 0.5. They note that FETCH manages to recover 96 per cent of the bursts with $S/N \ge 8$ and mention that the ones that are missed, are strongly affected by (broad-band) RFI. All diagnostic plots of our candidates to which model H assigned a probability p >0.5 had to undergo two independent by-eye inspections to be deemed true astrophysical bursts.

³https://github.com/demorest/psrfits_utils

⁴https://www.cv.nrao.edu/~sransom/presto/

⁵https://github.com/danielemichilli/SpS

⁶https://github.com/devanshkv/fetch

⁷The same telescope, receiver, backend, and source but a factor of 8 higher frequency resolution.

Two out of our 56 data-sets have previously been analysed by both Gourdji et al. (2019) and Aggarwal et al. (2021). Aggarwal et al. (2021) also miss B33 from Gourdji et al. (2019), and speculate that it may be due to the narrow-bandedness and low S/N of the burst. Additionally, Hessels et al. (2019) have reported on 14 bright bursts that are also included in our analysis here. We emphasize that even despite the large number of bursts found, there are possibly still a small number of bursts being missed due to the range of our boxcar widths used in searching, S/N-limitations and the model predictions of FETCH. Our sensitivity to bursts, some factor shorter or longer than our minimum and maximum boxcar lengths decreases as a function of the square root of said factor. Furthermore, we do not implement any subbanded searches, which means we are less sensitive to very narrow-band bursts from FRB 20121102A (bursts that extend over less than approximately a third of the observing bandwidth).

3 RESULTS AND ANALYSIS

A total of 478 bursts were determined to be of astrophysical origin in these observations, most of which were low S/N (see Fig. B1). The majority (333) of these bursts have not been reported before. Due to the grouping of pulses within 20 ms into burst candidates, there were a small number of instances (\sim 7 per cent) where other nearby, much fainter bursts were also visible in addition to those pulses classified by FETCH. These bursts have been included in the total number of 478. Distinguishing between bursts and sub-bursts is non-trivial and becomes ambiguous at a certain level. In order to do so, we thus implemented the following definition: If a distinct Gaussian-like component in the frequency-averaged burst profile does not smear into other components but instead begins and ends at a baseline comparable to the noise, it is defined as a burst, unless the component exists as one of the downward-drifting components often seen from repeaters (the sad-trombone effect), in which case we define it as a sub-burst. Fig. 1 shows the dynamic spectra of a subsample of bursts that are either bright, short-duration or narrow-band, or have complex time-frequency structure. We do not see any clear upward drifting bursts (a 'happy trombone' effect) in this sample.

All bursts were dedispersed to a DM of 560.5 $pc cm^{-3}$, determined by Hessels et al. (2019), which maximized the band-integrated structure of the bursts and provided a smaller range (more than an order of magnitude) in the inferred DMs of the bursts than the DM that optimizes peak S/N. To read in the fits files, we make use of the PsrfitsFile.read_subint function from the psrfits.py script of PRESTO, which reads in sub-intervals from a fits file. When doing so, the scales, weights, and offsets had to be explicitly applied to prevent saturation of the brighter bursts as a result of previous downsampling. Given that the Arecibo observations in Hessels et al. (2019) are a sub-sample of the observations presented here, we choose this DM and assume that any change in the DM over the course of our observations is negligible. For each burst dynamic spectrum, the channels in which RFI were present were flagged and excluded from further processing. The dynamic spectra were then corrected for bandpass variations by subtracting the off-burst mean and dividing by the off-burst standard deviation, per channel. In Fig. 1, in addition to dynamic spectra, we also show a time averaged spectrum, a full-band averaged burst time profile, and the burst time profile obtained by only averaging over the frequencies where the burst is present (the 2σ region of a one-dimensional Gaussian fit to the burst spectrum). A one-dimensional Gaussian function is also fit to each full-band burst profile. The burst spectra are produced by time-averaging the dynamic spectrum over the 2σ region of this Gaussian fit. The 2σ regions of both Gaussian fits are indicated by

green (and blue) shaded regions. We define the temporal width of a burst as the FWHM of the band-limited burst profile, and bandwidth as the 2σ region of the fit to the spectrum. We calculated the fluence of a burst by summing over the 2σ time region of a burst and then multiplying each channel of the spectrum with a channel-specific scaling based on the radiometer equation. Thereafter, the mean of the on-pulse spectrum is multiplied with the sampling time. This is equivalent to taking the product of the sampling time, the normalized time-profile and the radiometer equation.

$$F = \sum S/N \frac{\text{SEFD}(\nu, \theta_z)}{\sqrt{N_p \nu_{\text{ch}} t_{\text{samp}}}} \times t_{\text{samp}}$$
(1)

where *S/N* is the signal-to-noise ratio per channel, SEFD(ν , θ_z) is the system equivalent flux density of Arecibo, which is dependent on the observing frequency (ν) and the zenith angle (θ_z), N_p is the number of polarizations (2 for the Arecibo L-wide receiver), ν_{ch} is the bandwidth of a single channel and t_{samp} the sampling time. In the literature, the gain (*G*) and system temperature (T_{sys}) of Arecibo have typically been approximated as 10.5 K Jy⁻¹ and 30 K, respectively. We have modelled the SEFD ($=T_{sys}/G$) as a function of observing frequency and zenith angle, by fitting an exponential function to system performance measurements taken with the *L*-band wide receiver over a few years.⁸ The sensitivity decreases as zenith angle increases and frequency decreases. This approach improves upon the standard convention of using one average value, as an SEFD ≈ 2.9 Jy (often used in literature) can underestimate the SEFD by ≤ 25 per cent on average.

Finally, the isotropic equivalent burst energy,⁹ E, is calculated as follows

$$E = \frac{4\pi F \Delta \nu D_{\rm L}^2}{1+z} \tag{2}$$

where z is the redshift, F is the fluence of the burst (calculated only over the time- and frequency-extent of the burst), Δv is the bandwidth of the burst and $D_{\rm L}$ is the luminosity distance to the host galaxy of FRB 20121102A (972 Mpc; Tendulkar et al. 2017).

The fluence of a burst can be calculated either over the entire observing frequency range (often the central observing frequency is used to offset the effect of varying instrumental bandwidths) or only over the extent of the burst. Common practice (e.g. Law et al. 2017) and recent recommendation (e.g. Aggarwal 2021) has preferred the latter. Importantly, however, when the isotropic equivalent energy is computed, one needs to consider the same frequency range as for the fluence estimation. Inconsistency in this matter may result in certain features of the burst energy distribution being diminished or accentuated.

3.1 Grouping bursts and comparing groups

In Fig. 2, the burst widths have been plotted against their isotropic equivalent energy, while also being coloured according to their observed bandwidth within the 1150–1730 MHz range we recorded. Our sample of bursts splits into two distinct groups in this three-dimensional parameter space. We refer to these two groups as the low-energy bursts (LEBs) and high-energy bursts (HEBs). We employed a Gaussian mixture models (GMM) clustering algorithm using the GaussianMixture.fit method in

⁸http://www.naic.edu/~astro/RXstatus/

⁹Even though the emission is expected to be beamed, we default to the isotropic equivalent energy because the beaming fraction is unknown.



Figure 1. Dynamic spectra of a sub-sample of the 478 detected bursts from FRB 20121102A. This sample contains bursts that are extremely bright, exhibit complex morphology, or inhabit the extremes of the observing frequency range. Table B1 is a sample of an online table which contains all 478 bursts and their properties. All bursts have been dedispersed to a DM of 560.5 pc cm⁻³. The horizontal white stripes are channels that have been excised in order to remove radio frequency interference. The top panel of each thumbnail shows two burst profiles: one from summing over the entire observing band (in grey) and a second from only summing over the spectral extent where the burst is detected (in black). The panels on the right show the on-pulse time-averaged spectra. The green shaded areas indicate the 2σ region of Gaussian fits to the burst profile and spectrum. In case where multiple bursts are present within a few tens of milliseconds, blue and darker green shaded areas indicate the 2σ region for these additional bursts. The burst number is displayed in the top right of each thumbnail. The dynamic spectra have been downsampled to a time and frequency resolution of 327.68 µs and 12.5 MHz, respectively. For visual purposes, the dynamic range of the colour scale is adjusted to saturate below the fifth percentile and above 95 per cent of the peak signal. Frequency and time axes are scaled equally to highlight the diversity in burst morphology. Three of the bursts displayed above were also presented in Hessels et al. (2019) at higher time resolution with 'AO' names: B125 (AO-02), B193 (AO-05), B209 (AO-06).

the sklearn.mixture package¹⁰ (Pedregosa et al. 2011) to determine which bursts belong to which group. GMM clustering algorithms assume that all data points are from a combination of a finite number of Gaussian distributions – this is not necessarily true for our data, but we only use this clustering method as a quantitative means of classifying bursts as LEBs or HEBs. We disregard bandwidth and only take into account the temporal width and isotropic equivalent energy parameters (both in logarithmic space) for the clustering. We did so, firstly, because the burst divide is most obvious in the width-energy plane. Secondly, the bandwidth is constrained by the maximum observing frequency range. While the bursts in our sample are also restricted to a range of temporal widths (due to boxcar lengths being searched) and energies (due to sensitivity limitations), the bandwidth is the only parameter where an abrupt cut-off is visible in its distribution. We use the isotropic equivalent energy density, which is dependent on the observed burst bandwidth) (and, importantly, differs from the intrinsic burst bandwidth). We have refrained from estimating the missing flux/energy outside the observing window since the spectrum of the more complex bursts are non-trivial to model and we believe a one-dimensional Gaussian function to be an oversimplification. If instead, we make use of spectral energy density (erg/Hz), we are able to remove the energy parameter's dependence on bandwidth. Fig. 2 has been replotted making use of spectral energy density rather than isotropic equivalent energy density in Fig. B2. While the separation between LEBs and



Figure 2. The 478 bursts have been clustered into two groups based on their width and energy as defined in Section 3.1, and these two clusters are depicted here as circles (the low-energy bursts, or LEBs) and triangles (the high-energy bursts, or HEBs). a: The width (FWHM) distributions of the bursts. The grey and black histograms correspond to the bursts indicated by circles (LEBs) and triangles (HEBs), respectively. b: The cumulative burst number-energy distribution. The data points are coloured according to the observed bandwidth of the bursts and correspond to the data points shown in panel c. c: The burst width plotted against the isotropic equivalent energy for all bursts. The detection thresholds for bursts with a S/N of six and various bandwidths (multiples of 100 MHz) are indicated by slanted dashed lines, which are also coloured according to observed bandwidth using the same scale as for the data points. d: Same as panel a, but for isotropic equivalent burst energy rather than temporal width.

HEBs diminishes, two groups are still visible. This metric is also not completely unbiased since, in some instances, bursts extend beyond the observable frequency range of the receiver. In such cases, we divide by an observed bandwidth that may be lower than the full spectral extent of the burst. We thus adhere to using isotropic equivalent energy, but note that the differences that we see between the HEBs and LEBs are much greater than what one would expect if arising solely from a change in bandwidth (on which the isotropic equivalent energy is dependent). In Fig. 2, the LEBs and HEBs have been plotted as circles and triangles, respectively. A total of 435 bursts were classified as LEBs and 43 as HEBs. Note that the LEBs typically have lower bandwidths than HEBs, and also larger temporal widths.

To determine whether the LEBs and HEBs can be drawn from the same parent population, we applied Kolmogorov–Smirnov (KS), Anderson–Darling (AD), and Epps–Singleton (ES) tests using the ks_2samp, anderson_ksamp, and epps_singleton_2samp methods from the scipy.stats package.¹¹ The critical value for the KS test is approximately $D = c(\alpha) \sqrt{\frac{n_1+n_2}{n_1n_2}} \approx 0.26$, where $c(\alpha) = 1.63$ is a coefficient corresponding to a significance level (α) of 0.01, $n_1 = 435$, and $n_2 = 43$ are the number of bursts classified as LEBs and HEBs, respectively. A KS-statistic, larger than this value, or, equivalently, a small *p*-value (<0.001), can be used to reject the hypothesis that the samples are drawn from the same parent population. For the ES-test, we followed the recommendation of Epps & Singleton (1986) and set the values where the empirical characteristic function will be evaluated to t = (0.4, 0.8). The results of all these statistical tests are tabulated in Table 2. All three tests concur that we can reject the hypothesis that the bandwidth, width, and fluence distributions between the two groups can be drawn from the same parent population.

To ensure that, the statistically significant dissimilarity we see between the LEBs and HEBs is not merely the effect of a few outliers, we made use of a quasi- bootstrapping resampling method, whereby we ran 1000 trials, randomly resampling 10 per cent of the bursts from both the LEBs and HEBs by drawing from the LEB and HEB distributions, respectively (with replacement). For each trial of each parameter (width, bandwidth, and fluence) we calculated the KS, AD,

¹¹https://docs.scipy.org/doc/scipy/reference/stats.html

 Table 2. The statistical test results from the comparison between LEB and HEB properties.

	KS statistic	<i>p</i> -value	Bootstrapping <i>p</i> -value
Width	0.47	3.0×10^{-8}	8.5×10^{-8}
Bandwidth	0.56	3.8×10^{-12}	1.3×10^{-10}
Fluence	0.63	2.0×10^{-15}	1.0×10^{-13}
	AD statistic	<i>p</i> -value	Bootstrapping p-value
Width	22.9	< 0.001	< 0.001
Bandwidth	50.5	< 0.001	< 0.001
Fluence	54.2	< 0.001	< 0.001
	ES statistic	<i>p</i> -value	Bootstrapping p-value
Width	63.3	5.9×10^{-13}	3.8×10^{-10}
Bandwidth	86.8	6.1×10^{-18}	1.4×10^{-12}
Fluence	39.9	$4.6 imes 10^{-8}$	1.7×10^{-6}

and ES statistic and accompanying *p*-value. The mean *p*-values of this investigation are also in Table 2, and confirm that the statistical significance of our results is not dominated by a small number of outliers.

3.2 Burst energetics

In order to estimate the completeness, limit of our observations, we assume a burst with a S/N of six that has a temporal width of 4.00 ms and a bandwidth of 275 MHz. These are the median values for width and bandwidth for all the bursts in our sample. Making use of equations (1) and (2), we can then estimate the fluence or energy threshold below which our burst sample is incomplete. This limit is ~0.057 Jy ms, or ~ 1.5×10^{37} erg. Assuming a worst case instead of using the median values quoted above, this limit is increased to ~ 5.5×10^{37} erg.

The cumulative energy distribution is usually approximated as a power-law $R(>E) \propto E^{\gamma}$, where R is the burst rate for bursts above some energy E and γ is the slope of the power law. Fitting for a single power law has proven to likely be an oversimplification when considering data over many orders of magnitude (Cruces et al. 2021). Moreover, Aggarwal (2021) have shown that the narrowbandedness of bursts can greatly affect the shape of the observed energy distribution. Because many bursts are cut-off at the top or bottom of the observing window, energy distributions typically do not represent the energy distribution of bursts but rather of the emission within the observing frequency range. One can estimate the missing flux by assuming the spectral shape of the burst (which vary a lot between bursts) or by only selecting the bursts that appear to be fully within the observing window. This, however, also introduces a bias as the brightest broad-band bursts will then be excluded. Approximating the energy distribution using slightly more complicated functions, for instance a broken power law (Aggarwal et al. 2021) or a combination of a log-normal and Lorentz function (Li et al. 2021) have yielded better results. As it has been discussed in Gourdji et al. (2019), correctly estimating the completeness threshold will also affect the interpretation of the results.

The rate-energy distribution of our sample of bursts can be seen in Figs 2b and 3. We used scipy.optimize.curvefit¹² to fit a power law and a broken power law to the energy distribution of our sample of bursts and subsets thereof. For the power-law fits, we excluded bursts that are below our calculated energy sensitivity threshold (i.e $E_{\rm iso}$ < 1.5 × 10³⁷ erg), which make up ~8 per cent of the total sample (when fitting the broken power law we make use of a very conservative sensitivity threshold of 5.5×10^{37} erg to ensure that we fit the kink in the energy distribution at higher energies and not the turnover due to loss of sensitivity at lower energies). These thresholds are indicated by red vertical lines in Fig. 3 and the turn-over at low energies due to our sensitivity loss is also apparent here. The best fitting was that of a broken power law to all the bursts above the conservative sensitivity threshold. This fit yielded a break energy of $E_{\text{break}} = 1.15 \pm 0.04 \times 10^{38} \text{ erg},$ with power-law slopes of -1.38 ± 0.01 where $E_{iso} < E_{break}$, and -1.04 ± 0.02 where $E_{iso} > E_{break}$ when considering all bursts above the sensitivity threshold. We then repeated the process, only selecting the bursts above our sensitivity threshold that are mostly within the band (the 2σ region of the Gaussian fit to the spectrum is within the observing range of 1150-1730 MHz) or which have a bandwidth larger than 500 MHz. This is done in an attempt to only account for those bursts for which we measure a significant fraction of the fluence. This yielded $E_{\text{break}} = 1.3 \pm 0.1 \times 10^{38}$ erg, with power-law slopes of -0.96 ± 0.02 where $E_{iso} < E_{break}$ and -0.68 ± 0.03 where $E_{\rm iso} > E_{\rm break}$. The results of these power-law fits (and power-law fits to the LEBs and HEBs, separately) are presented in Table B2. Furthermore, we estimated the power-law index using the maximum likelihood method for different completeness thresholds, as described by Crawford, Jauncey & Murdoch (1970), and James et al. (2019), but the power-law index does not to converge towards a single value (see the right-hand panel of Fig. 3). This suggests that the energy distribution is not well described by a single power law.

4 DISCUSSION

4.1 Energy distribution

The energy distribution of bursts from FRB 20121102A has been studied by various other authors (Law et al. 2017; Gourdji et al. 2019; Oostrum et al. 2020; Aggarwal et al. 2021; Cruces et al. 2021; Li et al. 2021). These authors have found different power-law slopes over different energy regimes (summarized in Table 3). Making use of the maximum likelihood method described in Section 3.2, these authors find power-law slopes between -1.1 and -1.8. Notably Law et al. (2017) find a less steep average power-law index of ~ -0.7 , irrespective of the burst frequencies (1.4 and 3.0 GHz bands) and burst rate, but only have a very small sample of bursts. The values we obtain here, for a broken power-law fit, are roughly consistent with those reported by Jahns et al. (2022), who found power-law indices of -0.85 and -1.47 above and below a break energy of 1.6×10^{38} , respectively, but slightly flatter than what Aggarwal et al. (2021) find $(-1.8 \pm 0.2 > E \approx 2 \times 10^{37} \text{ erg})$ by fitting a broken power law. The power-law index appears to be steepening from higher to lower energies, before decreasing again as the incompleteness regime is entered. We also observe that the power-law index is potentially changing from day to day, but lack the number of bursts required to concretely quantify this. This behaviour is also seen in Arecibo observations of FRB 20121102A a few years later in 2019 (Jahns et al. 2022). More observations, over longer periods, are required to reveal how the time-dependent nature of the energy distribution - as seen in Li et al. (2021) – affects the slope of the energy distribution.

Highly magnetized neutron stars are promising candidates for comparison of burst energy distributions, by reason of the multitude of theories which advocate for them as the progenitors of FRBs. The power-law indices fit to the energy distributions that we and

¹²https://docs.scipy.org/doc/scipy/reference/generated/scipy.optimize.curv e_fit.html



Figure 3. Left: The cumulative energy distribution overplotted with a broken power-law fit that excludes bursts below the conservative sensitivity threshold (5.5×10^{37} erg). The vertical red line indicates the sensitivity threshold assuming a burst with a S/N of six and the mean burst properties of this sample, while the lighter red line indicates the conservative sensitivity threshold. The black dashed line illustrates the broken power-law fit to the data, with the black vertical line indicating the break of the power-law fit. Right: the power-law index as a function of threshold energy calculated using a maximum-likelihood estimation. The horizontal dashed lines indicate the power-law indices obtained from fitting a broken power law to the data with a least squares method. Vertical lines are the same as for the left-hand panel.



Figure 4. Width-fluence distribution for our Arecibo-detected bursts in blue, as well as for bursts from FAST (Li et al. 2021), Effelsberg (Cruces et al. 2021), Lovell (Rajwade et al. 2020), and Apertif (Oostrum et al. 2020) in black, green, yellow, and grey, respectively. The diagonal lines represent the sensitivity threshold of the respective instruments. Side histograms in the left-hand and top panel show the distribution of the burst width and fluence, respectively. These histograms have been normalized by the number of bursts reported in each of the respective studies. We caution against over-interpretation of the values being compared here, since the definition of 'width' and method of calculating the fluence are somewhat inconsistent between samples.

other authors have found for FRB 20121102A, are consistent with voltage scaling models made for magnetars (Wadiasingh et al. 2020). There are a few pulsars, like the Crab pulsar (PSR B0531+21), that are known to display giant pulses (GPs; Heiles & Campbell 1970; Staelin 1970). GPs are typically very short (sometimes shorter than a ns) and 2–4 orders of magnitude brighter than normal pulses (e.g Hankins et al. 2003; Hankins & Eilek 2007). The pulse-energy

distribution of the GPs from the Crab pulsar obey a power law (e.g. Argyle & Gower 1972; Lundgren et al. 1995; Popov & Stappers 2007; Karuppusamy, Stappers & van Straten 2010), although the power-law index is typically much steeper ($-\alpha \approx 2-3$) than for FRB 20121102A. It is also worth noting that the V-shape seen in Figs 2 and 4 is similar to what is seen for GPs from the Crab pulsar, where the narrower GPs tend to be brighter (Popov & Stappers 2007;

Table 3. Of	bservation pro	perties of FRB	20121102A	bursts at <i>L</i> -band.
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Telescope	SEFD (Jy)	Observing frequency range (MHz)	Observing time (hrs)	Energy range (erg)	Power-law index
Apertif ^a	700	1130-1760	~130	$\sim 7 \times 10^{38}$ -6 $\times 10^{39}$	-1.7 ± 0.6
Arecibo (This work)	3.5	1150-1730	\sim 57	$\sim 4 \times 10^{36} - 8 \times 10^{39}$	See Table B2
Arecibo ^b	2.9	1150-1730	~ 3	$\sim 7 \times 10^{36} - 2 \times 10^{38}$	-1.8 ± 0.3
Effelsberg ^c	17	1210-1510	~ 128	$\sim 1 \times 10^{38} 1 \times 10^{39}$	-1.1 ± 0.1
FAST ^d	1.25	1050-1450	~ 60	$\sim 4 \times 10^{36} - 8 \times 10^{39}$	-0.61 ± 0.04
	_	-	_	$\sim 3 \times 10^{38}$ -8 $\times 10^{39}$	-1.37 ± 0.18
Lovell ^e	38	1400-1800	~ 198	-	-

Note.^{*a*} Oostrum et al. (2020) ^{*b*} Gourdji et al. (2019) ^{*c*} Cruces et al. (2021) ^{*d*} Li et al. (2021), in this case the power-law fit is to the isotropic equivalent energy distribution rather than the cumulative-rate energy distribution. ^{*e*} Rajwade et al. (2020)

Karuppusamy et al. 2010), however, the time-scales are much shorter for the Crab pulsar.

MeerKAT observations of another young pulsar, PSR J0540-6919, yielded hundreds of GPs with a flux distribution that can be fit by a power law with a very steep slope of -2.75. Interestingly, there are also deviations from the power-law slope that are seen at all observing epochs (Geyer et al. 2021), similar to the deviations we observe in our power-law fits in this work. Only a small fraction of the GPs (<3 per cent) were detected in a fraction of the band, however, implying that these kinks are not the result of narrow-bandedness.

4.2 High- and low-energy bursts

The classification of the bursts into two groups (LEBs and HEBs), as discussed in Section 3.1, also appears to be consistent with what we can infer from the shape of the burst energy distribution. The break in the broken power-law fit occurs at, approximately, the same energy where we find the transition between LEBs and HEBs (see Fig. 3). Li et al. (2021) have reported that the energy distribution of FRB 20121102A can be modelled by a combination of a log-normal and Cauchy distribution. The break in our power-law fit occurs at approximately the same energy as where these two functions in their fit merge ($\sim 10^{38}$ erg).

The two potential groups of bursts we have identified are unlikely to inhabit mutually exclusive areas of the parameter space. Instead, the burst rate of high-fluence bursts from both groups is lower than that of the low-fluence bursts: with sufficient observation time and activity from FRB 20121102A, the LEBs and HEBs will partially overlap in their distributions. This can be seen in Fig. 4 where we plot the fluence against the width of our sample of bursts, as well as bursts from Rajwade et al. (2020), Oostrum et al. (2020), Cruces et al. (2021), and Li et al. (2021). The details of these other observations are summarized in Table 3. The forward-slanted V-shape seen in Fig. 2 can also be seen here, although less distinctly, and the group of LEBs is broader and extends down to lower fluences due to the sensitivity of the FAST observations. We observed for a total of \sim 59 hr, \sim 39 hr of which fall within the proposed activity period of FRB 20121102A (Rajwade et al. 2020; Cruces et al. 2021). In comparison, Cruces et al. (2021) observed \sim 128 hr, \sim 75 hr within the activity window, and Li et al. (2021) observed \sim 60 hr, all within the active period. Given that these authors spent more time observing during the active period than we did, this illustrates how, given sufficient observation time, rarer high-fluence LEBs (and HEBs) will be detected and the two groups will appear to merge in the fluence-width parameter space. We caution against over-interpreting the overlap of bursts in this figure, as widths and fluences are not always defined or calculated

in a consistent manner.¹³ Furthermore, the energy distribution of the bursts from FRB 20121102A has been seen to change over the course of one activity cycle (Li et al. 2021). Moreover, in the case of FRB 20180916B, the burst activity has been shown to drastically change from one cycle to the next (Pleunis et al. 2021b). If the same mechanisms are at play in FRB 20121102A, this further complicates the comparison of observations during different activity windows. Finally, we note that the classification of the lowest-energy bursts with the smallest temporal widths is particularly ambiguous, and can be influenced by the energy-metric used, since the LEBs and HEBs overlap in this part of the parameter space.

In a comparison of burst fluences and luminosities, Shannon et al. (2018) found that FRBs detected by Parkes were on average of lower fluence, but higher DM than those detected by ASKAP. These different distributions arise from different telescope sensitivities and instantaneous field-of-view. Nonetheless, the Parkes and ASKAP FRBs have comparable luminosities (using DM as a proxy for distance). Additionally, the energies of bursts originally detected from FRB 20121102A were relatively low, compared to ASKAP and Parkes one-offs. This may lead one to speculate that the one-offs are in fact the most energetic bursts detected from repeaters. These FRBs are all plotted in an adapted version of the fluence-dispersion plot in Fig. 5 (Shannon et al. 2018; Petroff et al. 2019). The properties of bursts from the two groups we identify in Section 3.1, in fact correspond to some of the differences observed between one-off and repeating FRBs from the CHIME/FRB catalogue (The CHIME/FRB Collaboration 2021), possibly further supporting this notion: HEBs and CHIME/FRB one-offs are typically broader band and shorter duration than LEBs and CHIME/FRB repeater bursts.

The analogy is, however, not perfect. The CHIME/FRB Collaboration (2021) have found no evidence to suggest that the fluence distributions of the bursts from the one-offs and repeaters are different (both CHIME/FRB catalogue one-offs and repeaters are also plotted in Fig. 5). Neverthelesss, these fluence distributions might not fully be representative of the underlying populations, as it can also be seen that the majority of current instruments lack the sensitivity to probe burst fluences below approximately 1 Jy ms. Should the known repeating FRBs be LEB equivalents and the observed one-offs be HEB equivalents, one would not necessarily expect them to have different fluence distributions. Rather, the burst fluence distributions of known repeaters would extend into the territory of one-offs. The brightest bursts from FRB 20121102A are comparable in luminosity to the one-offs detected by ASKAP and Parkes, and the faintest

¹³Oostrum et al. (2020) and Li et al. (2021) quoted boxcar widths, while Rajwade et al. (2020), Cruces et al. (2021) and this work define the width as the FWHM of a Gaussian fit.



Figure 5. Fluence plotted against excess dispersion measure (corrected for the inferred Galactic foreground contribution) for the sample of bursts from FRB 20121102A detected in this paper (where the black vertical line indicates the range of fluences), as well as various other one-off and repeating FRBs. The different colours represent FRBs detected by different telescopes. Yellow bars indicate the range of fluences detected for CHIME/FRB repeaters. Lines of constant luminosity, assuming a DM contribution from the host galaxy $DM_{host} = 50 \text{ cm}^{-3} \text{ pc}$, are overplotted. Figure adapted from Shannon et al. (2018); Petroff et al. (2019).

bursts extend beyond the sensitivity threshold of all but FAST (and, previously, that of Arecibo).

Another potential caveat arises when considering a sample of 41 low-energy bursts from FRB 20121102A (almost all of which are also presented here) presented by Gourdji et al. (2019), who showed that the narrow-bandedness of fainter bursts might be explained by receiver sensitivity that is insufficient to detect fainter subcomponents. This argument suggests that the LEBs are merely the 'tips of the iceberg', but on the other hand HEBs possess smaller temporal widths in general than LEBs, contradicting this notion.

The strongest potential counter-argument comes from the accumulating evidence in the literature that supports polarimetric differences between repeaters and one-offs. If all FRBs are repeaters and known repeaters are akin to LEBs while known one-offs are the HEBs, the polarimetric properties of LEBs and HEBs are expected to be different. We do not perform any polarimetric studies as the frequency resolution of our data is insufficient to resolve the high RM of FRB 20121102A at our observing frequency, but others have found no polarization from any FRB 20121102A bursts at ~1.4 GHz (Li et al. 2021). Polarization is, however, detected at \sim 4.5 GHz (Michilli et al. 2018b), so the lack of polarization at \sim 1.4 GHz could potentially be due to the extreme local environment of the source imparting depolarization at lower frequencies (Plavin et al. 2022). Multi-path propagation in a scintillation screen between the FRB source and observer can cause linear depolarization, as well as impart significant circular polarization (Beniamini, Kumar & Narayan 2022).

4.3 Burst diversity and the average burst spectrum

Already at the time of FRB 20121102A's discovery as the first known repeating FRB, Spitler et al. (2016) remarked on the high variability between burst spectra. Many others also have noted the diversity in both the time and frequency properties of the bursts. That being said, the burst morphology of some of the more complex bursts is remarkably similar: e.g. the precursor bursts observed by Caleb et al. (2020) or the similarity between bursts AO-02 and GB-01 in Hessels et al. (2019). Whether this is a physically meaningful



Figure 6. Bandwidth–width distribution of bursts from our sample on days where more than 40 bursts were detected. Different days are indicated by different colours (see the MJD in the legend). Side histograms in the left-hand and top panels show the observed bandwidth and width distributions, respectively.

observation or mere coincidence remains to be seen. This diversity in the time and frequency properties of the bursts is showcased in Fig. 1. The narrowest and widest bursts we detected are 0.5 (B205) and 26.0 ms (B107), respectively. The bursts also extend over different frequency ranges, with some being very narrowband (as narrow as \sim 65 MHz), while others extend over the whole observing bandwidth of 580 MHz. We found no evidence for periodic variation in any of the burst properties (see Appendix Figs B3-B6). We do, however, observe that the bursts from the same epoch are, on average, more similar to each other than to bursts from other epochs. The irregular burst rate makes this effect especially difficult to quantify. In Fig. 6, the observed bandwidth of bursts are plotted against temporal widths for days where more than 40 bursts were detected. One can see, e.g. that a larger fraction of the bursts from MJD 57614 have larger bandwidths and that the bursts are of shorter duration compared to MJD 57628. For MJD 57614 and MJD 57628 the median bandwidths and widths are 292 MHz and 1.95 ms and 223 MHz and 5.08 ms, respectively. This might suggest an emission region that is evolving over the course of days, but stable over a period of ~hours. However, the statistical significance of these apparent similarities of burst properties within an observing epoch requires confirmation.

Fig. 7 shows two different methods for determining the average spectrum of the bursts in our sample. In the middle panel, the mean spectrum of the 435 LEBs and 43 HEBs detected from FRB 20121102A have been plotted as black and grey histograms, respectively. These S/N-weighted average spectra are obtained by simply summing the spectra together of all the bursts in a group and then dividing by the number of bursts in that group. This plot shows that, on average, we observed factors of approximately five more burst emission at higher observing frequencies than at lower observing frequencies, for both the LEBs and HEBs. All spectra have been corrected for bandpass variations and the effect of decreasing sensitivity at lower frequencies has already been accounted for by modelling the SEFD of Arecibo as a function of frequency and zenith angle (see Section 3).

The top panel in Fig. 7 shows the SEFD as a function of frequency for various zenith angles ranging from 14 degrees in dark blue to 19 degrees in yellow (the range covered by FRB 20121102A as it is transiting at Arecibo). Throughout the observation some channels were flagged more than once due the presence of RFI (manifesting as sharp dips in the average spectra in the middle panel). This decrease



Figure 7. Top panel: The observational limitations: the red and pink histogram shows how many times each frequency channel was flagged due to the presence of RFI for the HEBs and LEBs, respectively. This is given as a percentage of the total number of bursts on the right hand y-axis. The SEFDs for zenith angles from 14 to 19 degrees (the range covered by FRB 20121102A as it is transiting at Arecibo), in steps of 1 degree, are shown on the left hand y-axis. Middle panel: The spectra obtained from averaging all the spectra for the HEBs (in black) and LEBs (in grey). Bottom panel: The percentage of bursts for which emission was present in a given channel for the HEBs in black and LEBs in grey.

in flux density that we observe at lower frequencies is thus not an effect of RFI excision.

The vast majority of bursts occur at the top of our observing band. We define the spectral extent of a burst as the 2σ region of a onedimensional Gaussian fit to the time averaged spectrum of each burst. In the bottom panel of Fig. 7, we stacked the spectral extents of all the bursts to produce an average 'spectrum' not weighted by signal intensity. This histogram thus indicates, given a burst detection, at which frequencies the burst emission is most likely visible. Here one can see that fewer bursts are detected at lower frequencies. For the LEBs, the shape is distinctly Gaussian-like and peaks at around 1600 MHz while dropping off quickly below 1400 MHz. Similarly, Gourdji et al. (2019) noted a 'dearth' of bursts below 1350 MHz. The difference in the bandwidths between the LEBs and HEBs noted before is also apparent here. The HEBs are typically broader band. We found that the peak of the summed spectral extent remains more or less constant across our observations (see Appendix Figs B3-B6). We suspect that most, if not all, of the bursts detected at our highest observing frequencies extend beyond the observing band. The average stacked spectral extents presented here are thus not a representation of the average burst, but rather the frequencydependent burst activity within our observing window.

Gajjar et al. (2018) plotted the peak flux density of a number of bursts in the literature as a function of frequency over a much larger frequency range of \sim 1–8 GHz, and found a relatively flat spectral index with no evidence of a spectral turnover. Both Gajjar et al. (2018), as well as Zhang et al. (2018), do however find that the bursts they detected between 4 and 8 GHz seem to have preferred frequencies. This effect is, at least partly, due to Galactic scintillation. Our observations, made at lower frequencies where the Galactic scintillation bandwidth is much narrower (Hessels et al. 2019), show that the bursts from FRB 20121102A occur at preferential frequencies (that likely evolve with time, see Jahns et al. (2022)).

4.4 Activity window

In Fig. 8, we illustrate the long-term monitoring of FRB 20121102A, since its discovery, at \sim 1.4 GHz with the Arecibo 305-m telescope. We also highlight the activity cycles proposed by Rajwade et al. (2020) and Cruces et al. (2021). Rajwade et al. (2020) proposed a period of 157 d with a 56 per cent duty cycle, whereas the period from Cruces et al. (2021) is slightly different: $161 \pm 5 d$ with a duty cycle of 54 per cent. After its discovery in 2012 (Spitler et al. 2014), FRB 20121102A was monitored again in 2013, and later in 2015 which resulted in the discovery of the first repeat bursts detected from an FRB (Spitler et al. 2016). Long-term Arecibo monitoring commenced around December 2015, but temporarily ceased early in 2016. Observations again resumed towards the end of this first activity window of 2016, during which a few bursts were detected. The vast majority of the bursts presented in this paper come from the burst storm detected during the second active period in 2016. During this period and after confirmation of burst detections, we increased the frequency of our observations to approximately daily cadence starting from around ~MJD 57640. The increased cadence, and consequently increased number of burst found, is evident in Fig. 8. Now, aware of the activity window of FRB 20121102A, the lack of detections early in the project can easily be explained and supports the case for an activity window of ~ 161 d. We do, however, note that a subset of the bursts presented here - and also in Gourdji et al. (2019), Aggarwal et al. (2021), and Hessels et al. (2019) – were used together with other data to determine this activity window. Continued monitoring of FRB 20121102A in the coming years is necessary to refine this activity window and to determine its long-term stability or variability.

Any progenitor model for FRB 20121102A naturally needs to be able to explain this fairly long period activity window and large duty cycle. Xu et al. (2021) have shown that a supernova fallback disc with the most reasonable values for a disc wind can slow down isolated neutron stars to spin periods of hundreds of hours. A period similar to that of FRB 20121102A can be obtained by invoking an extremely large disc wind, but this challenges what may be physically possible around a magnetar. Wada et al. (2021) have shown that the large activity window of periodic FRBs can be explained by incorporating eccentricity in the binary comb model of Ioka & Zhang (2020). Furthermore, specifically for FRB 20121102A, and assuming that its persistent radio counterpart is in fact powered by a disc wind, a supermassive or intermediate-mass black hole companion is preferred over a massive stellar companion.

4.5 Wait times

Despite its long-term periodic activity, FRB 20121102A has shown no short time-scale periodicity in the arrival times of the bursts like that seen from pulsars or FRB 20191221A (The CHIME/FRB Collaboration 2022),¹⁴ despite various searches using a range of techniques (Zhang et al. 2018; Aggarwal et al. 2021; Cruces et al. 2021). Other authors have found that the wait times between bursts form a bi-modal distribution with peaks at milli- and decaseconds (Gourdji et al. 2019; Katz 2019; Aggarwal et al. 2021; Li et al. 2021). The wait-time distribution between bursts is dependent on the definition of a burst, and on sub-second scales will look especially different depending on whether the wait times between sub-bursts (~ ms) are included or not. We have calculated the wait times for

¹⁴This FRB is a single 3 s event with sub-second periodicity between subbursts.



Figure 8. Timeline of Arecibo monitoring of FRB 20121102A, showing all the FRB 20121102A bursts detected with Arecibo at \sim 1.4 GHz up until the end of 2016. MJD is indicated on the bottom axes and calendar date on the top axes. The blue star indicates the date of the first burst ever to be detected from FRB 20121102A (Spitler et al. 2014). Blue histogram bars indicate the number of bursts detected in different observations (left axes). Note the first ten repeat bursts in mid-2015 that were presented in Spitler et al. (2016) and the first burst that was detected with PUPPI at the end of 2015 and presented in Scholz et al. (2016). The cumulative time spent observing is shown in black (right axes), and is reset to zero on MJD 57342, which is when the observations started that we report on in this paper. The tentative activity windows proposed by Rajwade et al. (2020) and Cruces et al. (2021) are shown as yellow and green shaded areas, respectively. The bottom panel is a zoom-in of the top panel, at the dates indicated by the thick horizontal red bar. Bursts from MJD 57644 and MJD 57645 were also presented in Gourdji et al. (2019) and Aggarwal et al. (2021).

our entire sample of bursts (and not between sub-bursts) and also see two humps in the distribution. We fit the two humps with a log-normal function, by first excluding wait times <1 s and then >1 s. The log-normal functions peak at ~24 ms and ~95 s, which is comparable to what has been found by other authors (Zhang et al. 2018; Aggarwal et al. 2021). The wait time distribution and log-normal fits are presented in Fig. 9. The durations of our observations are tabulated in Table 1. The median observation duration is ~3800 s and implies that we are not sensitive to wait times longer than a few kiloseconds. Additionally, we also calculated the wait times separately for the HEBs and LEBs. The HEBs show no sign of clustering in time (see also Appendix Figs B3– B6) and they typically have longer wait times, consistent with their lower rate.

The peak at longer time-scales is consistent with what is expected from a Poissonian distribution. Aggarwal et al. (2021) have shown how this decaseconds peak shifts to shorter time-scales exponentially as more bursts are detected in a constant-length observation. Our sub-second peak, in combination with the absence of wait times around one second, potentially describes the typical event duration. While we do not intentionally measure the wait times between sub-bursts, at least some fraction of the sub-second wait times are likely the wait times between non-subsequent sub-bursts, where the faintest sub-bursts are not being detected. A 'sub-burst' peak is very evident in Li et al. (2021) at \sim 3 ms, but not here. In their wait time distribution, a second sub-second peak can be seen (but is not explicitly mentioned) corresponding to the one we observe. The sub-second peak we detect is also consistent with those found by Gourdji et al. (2019), Aggarwal et al. (2021), and Li et al. (2021), and supports the claims made by the latter authors that challenge the feasibility of certain coherent emission models that require orderly magnetic field lines.

This bi-modality in the wait time distribution is also seen in magnetars (e.g. Huppenkothen et al. 2015). In the specific case of SGR J1550-5418, the wait times are between the spikes (akin to subbursts) seen in the magnetar X-ray bursts and the peak at lower wait times is around 50 ms. While Huppenkothen et al. (2015) note that these observed wait times are consistent with both repeated crust failure as well as magnetospheric reconnection models, Wadiasingh & Chirenti (2020) argue that the shorter time-scale observed in the wait time distribution of magnetars is likely associated with magnetar oscillations. It is also worth mentioning that SGR 1935+2154,



Figure 9. The wait time distribution for all detected bursts is shown by the blue outline. The separate wait time distributions for the LEBs and HEBs are indicated by solid grey and black histograms, respectively. Log-normal functions are fit to the full sample distribution above and below 1 s, and peak at ~24 ms, and ~95 s, as indicated by the dotted blue lines. Our sensitivity to longer wait times is degraded by the length of our observations. The median length of our observations is indicated by the red vertical line. For comparison, the peaks of the waiting times identified by Li et al. (2021) are indicated by yellow dashed vertical lines at 3.4 ± 1.0 ms, 70 ± 12 s, and 220 ± 100 s. The latter being the peak for the higher energy bursts ($E > 3 \times 10^{38}$ erg) they observe.

the only Galactic magnetar known to show an FRB-like burst, has been shown to exhibit (X-ray) burst rates and a fluence power-law index consistent with that observed from FRB 20121102A, although the wait times between bursts are $\sim 2 \text{ s}$ (Younes et al. 2020). We encourage better quantification of the wait time distribution for other repeating FRBs, as well as searches for periodicities in the wait time distributions.

4.6 Periodicity search

A Lomb–Scargle¹⁵ periodicity search was performed on the barycentric corrected burst times of arrival for every observation that contained at least 24 bursts. We tested 2×10^6 trial periods which were logarithmically spaced between 20 ms and 200 s. We find no significant periods that are consistent between multiple observations. Additionally, a fast folding algorithm (FFA) was applied to the observations that contain at least 40 bursts using RIPTIDE (Morello et al. 2020). As with the Lomb–Scargle search, periods between 20 ms and 200 s were searched. We again find no significant periods that are consistent between multiple observations. If the central engine of the FRBs is rapidly spinning, then the lack of detected periodicity in the burst arrival times suggests that the emitting region is not stable in location.

5 SUMMARY

In this paper, we have reported that the detection of 478 bursts (of which 333 are previously unreported) from FRB 20121102A that were obtained through monitoring with the Arecibo telescope at *L*-band, between December 2015 and October 2016. Our observations total is \sim 59 hr, and reflect the earliest large observing campaign of this iconic source. Our analysis is summarized as follows:

¹⁵https://docs.scipy.org/doc/scipy/reference/generated/scipy.signal.lombsc argle.html

(i) The bursts in our sample are not all identical, and divide into two groups in the width-bandwidth-energy parameter space. We refer to these two burst types as LEBs (low-energy bursts) and HEBs (high-energy bursts).

(ii) Statistical comparison tests concur that the temporal width, observed bandwidth and fluence distributions for the HEBs and LEBs cannot be drawn from the same parent population.

(iii) The HEBs are typically narrower and more broad-band than the LEBs, echoing the differences between one-off and repeating FRBs.

(iv) We fit the energy distribution of our sample with a broken power-law that has a break at $E_{\text{break}} = 1.15 \pm 0.04 \times 10^{38} \text{ erg}$, with power-law slopes of -1.38 ± 0.01 where $E_{\text{iso}} < E_{\text{break}}$, and -1.04 ± 0.02 where $E_{\text{iso}} > E_{\text{break}}$.

(v) We observed that bursts from the same epoch tend to be more similar to each other than to the bursts from other epochs, but note that this observation requires confirmation.

(vi) The majority of the bursts in our sample occur towards the top of our observing band (\sim 1730 MHz) and like previous authors we note a dearth of bursts detected below \sim 1350 MHz.

(vii) Our observations (including those with non-detections) support the case for a \sim 160 d activity period with a \sim 55 per cent duty cycle.

(viii) We find two log-normal peaks in the wait time distribution at \sim 24 ms and \sim 95 s, consistent with previous measurements and indicative of a characteristic event duration and Poissonian rate, respectively.

(ix) Despite using various methods, we find no short-term periodicity in the burst arrival times when searching between 20 ms and 200 s.

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Facility: Arecibo *Software:* ASTROPY, FETCH, DSPSR, PSRCHIVE, PRESTO, PSRFITS_UTILS, RIPTIDE

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DATA AVAILABILITY

The Arecibo data used in this work is available upon reasonable request.

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SUPPORTING INFORMATION

Supplementary data are available at MNRAS online.

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APPENDIX A: FIT FUNCTIONS

We define a broken power law as

$$N(>E) = \begin{cases} k(E/E_{\text{break}})^{\alpha_1}, & E < E_{\text{break}} \\ k(E/E_{\text{break}})^{\alpha_2}, & E \ge E_{\text{break}} \end{cases}$$
(A1)

where k is a scaling factor and α_1 and α_2 are the power-law indices below and above the break energy, E_{break} , respectively.

APPENDIX B: EXTRA MATERIAL



Figure B1. The S/N of all the bursts detected by PRESTO's single_pulse_search.py. This is the S/N of the boxcar with the highest S/N at the DM that maximizes S/N.

Table B1. Burst properties (full table available online).

Burst ID	TOA ^a	Width (ms)	Bandwidth (MHz)	$f_{\rm high}~({\rm MHz})$	$f_{\rm low}~({\rm MHz})$	Fluence ^b (Jy ms)	Energy ^b (erg)	Group
1	57364.204636306	0.68 ± 0.05	>546	>1730	1184	0.08	4.0×10^{37}	HEB
2	57506.802409024	4.7 ± 0.3	>381	>1730	1349	0.38	1.3×10^{38}	LEB
3	57510.807219384	2.27 ± 0.09	>546	>1730	1184	0.29	1.4×10^{38}	HEB
4	57510.819153558	2.2 ± 0.3	>305	>1730	1425	0.11	2.9×10^{37}	LEB
5	57510.826409121	3.4 ± 0.2	>419	>1730	1311	0.27	1.0×10^{38}	LEB
6	57510.838080436	2.0 ± 0.2	>267	>1730	1463	0.19	4.3×10^{37}	LEB
7	57513.786194552	6.9 ± 0.4	>508	>1730	1222	0.38	1.7×10^{38}	LEB
8	57607.531624205	2.7 ± 0.2	>331	>1730	1399	0.30	7.3×10^{37}	LEB
9	57607.532581875	1.1 ± 0.2	>217	>1730	1513	0.09	1.5×10^{37}	LEB
10	57607.532581931	0.8 ± 0.1	>217	>1730	1513	0.10	1.6×10^{37}	LEB

Note.^{*a*} These times are dynamical times (TDB), corrected to the Solar system barycenter to infinite frequency assuming a dispersion measure of 560.5 pc cm⁻³ and dispersion constant of $1/(2.41 \times 10^{-4})$ MHz² pc⁻¹ cm³ s.^{*b*} We estimate a conservative 25 per cent uncertainty on these measurements due to uncertainty in the SEFD of Arecibo.

Table B2.	Power-law f	fit parameters.
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Single power-law	α_1			R^2
All bursts	-0.87 ± 0.01	_	_	0.961
LEBs	-0.98 ± 0.01	-	-	0.943
HEBs	-0.46 ± 0.02	-	_	0.956
In-band bursts only	-0.72 ± 0.01	-	_	0.978
Broken power-law	α_1	$E_{\rm break}$	α2	R^2
All bursts	-1.38 ± 0.01	$1.15 \pm 0.04 \times 10^{38}$	-1.04 ± 0.02	0.999
LEBs	-1.94 ± 0.02	_	_	_
HEBs	-0.39 ± 0.02	$3.2 \pm 0.2 \times 10^{38}$	-0.89 ± 0.04	0.993
In-band bursts only	-0.96 ± 0.02	$1.3 \pm 0.1 \times 10^{38}$	-0.68 ± 0.03	0.995



Figure B2. Fig. 2 replotted with isotropic equivalent energy (erg) replaced by spectral energy density (erg/Hz). The classification of the lowest energy bursts with the smallest temporal widths becomes ambiguous, because the LEBs and HEBs overlap.



Figure B3. In Figs **B3**–**B6**, the burst properties have been illustrated for all days on which bursts were detected. In the top panel, the temporal width of the bursts is plotted as a function of the MJD. The range of the MJD *x*-axes are representative of the duration of each observation, and the scale changes between plots of different days. In the center bottom panel of a vertical line, coloured according to the calculated isotropic equivalent energy of the burst, shows the spectral extent of each burst as a function of MJD. A grey circle or black triangle, indicates whether a burst is classified as a LEB or HEB, respectively, and is plotted on top of the line at the frequency corresponding to the peak of the Gaussian fit to the spectrum. All the vertical lines are stacked upon each other and displayed in the grey histogram in the left-hand panel, illustrating the frequencies where a burst is most likely to be found during the observation. The overplotted red-dotted histogram indicates for how many bursts specific channels were flagged. Horizontal dashed-grey lines at the top and bottom of the observation of the observation is shown, which has been colour-coded according to energies. These colours also correspond to the data points in the bottom center panel. Here we show MJDs 57506, 57510, 57513, and 57607.



Figure B4. Same as B3, but for MJDs 57614, 57628, 57638, 57640, and 57641.



Figure B5. Same as B3, but for MJDs 57642, 57644, 57645, 57646, and 57648.



Figure B6. Same as B3, but for MJDs 57649, 57650, and 57666.

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