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A Nançay Radio Telescope study of the hyperactive repeating FRB 20220912A

David C. Konijn[®],^{1,2,3}* Danté M. Hewitt[®],³ Jason W. T. Hessels[®],^{2,3,4,5} Ismaël Cognard[®],^{6,7} Jeff Huang[®],³ Omar S. Ould-Boukattine[®],^{2,3} Pragya Chawla[®],² Kenzie Nimmo[®],⁸ Mark P. Snelders[®],^{2,3} Akshatha Gopinath[®],³ and Ninisha Manaswini[®],⁹

¹Kapteyn Astronomical Institute, University of Groningen, Kapteynborg 5419, NL-9747 AD Groningen, The Netherlands

³Anton Pannekoek Institute for Astronomy, University of Amsterdam, Science Park 904, NL-1098 XH Amsterdam, The Netherlands

⁴Trottier Space Institute, McGill University, 3550 rue University, Montréal QC H3A 2A7, Canada

⁵Department of Physics, McGill University, 3600 rue University, Montréal QC H3A 2T8, Canada

⁶Station de Radioastronomie de Nançay, Observatoire de Paris, PSL University, CNRS, Université d'Orléans, F-18330 Nançay, France

⁷Laboratoire de Physique et Chimie de l'Environnement et de l'Espace LPC2E UMR7328, Université d'Orléans, CNRS, F-45071 Orléans, France

⁸MIT Kavli Institute for Astrophysics and Space Research, Massachusetts Institute of Technology, 77 Massachusetts Ave, Cambridge, MA 02139, USA

⁹Max Planck Institute for Radio Astronomy, University of Bonn, Auf dem Hügel 69, D-53121 Bonn, Germany

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ABSTRACT

The repeating fast radio burst source FRB 20220912A was remarkably active in the weeks after its discovery. Here, we report 696 bursts detected with the Nançay Radio Telescope (NRT) as part of the Extragalactic Coherent Light from Astrophysical Transients (ÉCLAT) monitoring campaign. We present 68 observations, conducted from 2022 October to 2023 April, with a total duration of 61 h and an event rate peaking at 75^{+10}_{-9} bursts per hour above a fluence threshold of 0.59 Jy ms in the 1.2–1.7 GHz band. Most bursts in the sample occur towards the bottom of the observing band. They follow a bimodal wait-time distribution, with peaks at 33.4 ms and 67.0 s. We find a roughly constant dispersion measure (DM) over time (δ DM $\leq 2 \text{ pc cm}^{-3}$) when taking into account 'sad-trombone' drift, with a mean drift rate of -8.8 MHz ms^{-1} . None the less, we confirm small $\sim 0.3 \text{ pc cm}^{-3}$ DM variations using microshot structure, while finding that microstructure is rare in our sample–despite the 16 µs time resolution of the data. The cumulative spectral energy distribution shows more high-energy bursts ($E_{\nu} \geq 10^{31} \text{ erg Hz}^{-1}$) than would be expected from a simple power-law distribution, showing clustering. We discuss the various observational similarities that FRB 20220912A shares with other (hyper)active repeaters, which as a group are beginning to show a common set of phenomenological traits that provide multiple useful dimensions for their quantitative comparison and modelling.

Key words: radio continuum: transients - fast radio bursts.

1 INTRODUCTION

Fast radio bursts (FRBs) are short-duration extremely luminous pulses of coherent radio emission originating from unknown extragalactic sources (Lorimer et al. 2007; Thornton et al. 2013; Petroff, Hessels & Lorimer 2022). FRBs typically have durations ranging from microseconds (Snelders et al. 2023) to milliseconds, and have been observed at radio frequencies ranging from 110 MHz (Pleunis et al. 2021a) to 8 GHz (Gajjar et al. 2018). Thousands of sources of FRBs have been detected, with ~97 per cent observed as (apparently) one-off events (Chime/Frb Collaboration 2023). The remaining small fraction of FRBs are known to repeat (e.g. Spitler et al. 2016), often in a clustered manner (e.g. Li et al. 2019; Hewitt et al. 2022) and showing band-limited emission (Gourdji et al. 2019).

The bursts from these repeaters are, on average, longer in duration and narrower in bandwidth than those seen from apparently non-repeating FRBs (Pleunis et al. 2021b). Despite these spectrotemporal differences between the repeaters and non-repeaters, some factors point towards a single population of FRBs with a continuous distribution of repetition rates that are capable of producing bursts with different (observed) morphologies. Most importantly, the upper limits on repetition for the apparent non-repeating FRBs are not clearly distinct from the repetition rates observed in the vast majority of repeaters, though with with some notable exceptions (Chime/Frb Collaboration 2023). To then consolidate the implied relationship between repetition rate and burst duration, beaming could be invoked to explain temporal differences (Connor, Miller & Gardenier 2020; Kumar, Qu & Zhang 2024). Most notably, the high end of the burst energy distribution from the repeating FRB 20201124A resembles that of the non-repeating FRB population (Kirsten et al. 2024).

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²ASTRON, Netherlands Institute for Radio Astronomy, Oude Hoogeveensedijk 4, NL-7991 PD Dwingeloo, The Netherlands

^{*} E-mail: konijn@astron.nl

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The source(s) and emission mechanism(s) behind repeating and non-repeating FRBs remain unknown, but given the diversity observed in FRB host environments, burst properties and repetition rates, attributing all observed FRBs to a single progenitor is likely an oversimplification. Nevertheless, among the various models, magnetically powered neutron stars, commonly referred to as 'magnetars', are among the most promising candidates (e.g. Margalit, Berger & Metzger 2019; Wadiasingh & Timokhin 2019; Gourdji et al. 2020). The energetics (Margalit, Metzger & Sironi 2020a), time-scales (Camilo et al. 2008; Nimmo et al. 2021, 2022; Hewitt et al. 2023), and emission frequencies (e.g. Dai et al. 2019) of magnetars and FRBs suggest a connection between them. The most important observational connection to magnetar theories is the highluminosity radio burst that was detected from the Galactic magnetar SGR 1935+2154, which showed similarities to FRBs in terms of its duration, spectrum and luminosity (Bochenek et al. 2020; CHIME/FRB Collaboration 2020; Kirsten et al. 2021). However, the rate of SGR 1935+2154-like bursts alone is insufficient to account for the observed FRB rate and repetition fraction (Margalit et al. 2020b). These authors argue that an additional population of magnetars is required, with stronger magnetic fields than Galactic magnetars, but lower birth rates. Indeed, the discovery of an FRB source in a globular cluster (Kirsten et al. 2022a), further advocates for diverse formation channels, assuming magnetars as the progenitors of FRBs.

On 2022 September 12, FRB 20220912A was discovered by the CHIME/FRB Collaboration (McKinven & Chime/Frb Collaboration 2022). In only three days, nine bursts were detected, indicating an exceptionally high burst rate - especially considering that, after six years of monitoring by CHIME/FRB, the median number of bursts detected per repeater is three (Chime/Frb Collaboration 2023). Shortly after the discovery of FRB 20220912A, the source was localised by the Deep Synoptic Array (DSA-110) to its host galaxy PSO J347.2702+48.7066 at a redshift of z = 0.0771 (Ravi et al. 2023). Thereafter, the European VLBI Network (EVN) localized the source to milliarcsecond precision, with coordinates RA (J2000) $= 23^{h}09^{m}04.8989^{s} \pm 5$ mas Dec. (J2000) $= 48^{\circ}42'23.9078'' \pm$ 5 mas (Hewitt et al. 2024). These observations also ruled out the presence of a compact persistent radio source (PRS), as has been associated with some other active repeaters (Marcote et al. 2020: Niu et al. 2022; Bhandari et al. 2023). Multiple telescopes have detected more than a hundred bursts within a few hours of observation around \sim 1.4 GHz (e.g. Feng et al. 2022; Kirsten et al. 2022b; Zhang et al. 2023). Remarkably, at the peak of its activity, the burst rate of FRB 20220912A (above a fluence threshold of 100 Jy ms) constituted up to a few per cent of the all-sky rate of FRBs in general (Sheikh et al. 2024). This rate of high-fluence bursts is comparable to another hyperactive repeater, FRB 20201124A (Kirsten et al. 2024). Such high activity is short-lived, however, with considerable daily fluctuations (Zhang et al. 2023). In terms of spectral range, in addition to the detections at 1.4 GHz, FRB 20220912A has been detected at 300-400 MHz (Bhusare et al. 2022; Ould-Boukattine et al. 2022), and around 2 GHz (Perera et al. 2022; Rajwade et al. 2022), yet no bursts have been found in searches above 3 GHz or below 200 MHz (Rajwade et al. 2022; Sheikh et al. 2022; Kirsten et al. 2022b).

The activity of repeaters is often clustered in time (Oppermann, Yu & Pen 2018) and triggered by an unknown mechanism (peharps starquakes or magnetar flares; e.g. Wang et al. 2018), as there have been numerous reported 'burst storms' where a repeating FRB produces bursts at a rate of dozens to hundreds per hour (e.g. Li et al. 2021; Jahns et al. 2023; Nimmo et al. 2023). FRB 20220912A is the second FRB, after FRB 20201124A (Lanman et al. 2022; Xu et al.

2022), that has seemingly awoken from quiesence before entering a hyperactive state during which it exhibits prolonged activity for weeks to months, observable by numerous telescopes in multiple frequency bands. More recently, FRB 20240114A has been seen to exhibit similar behaviour (Shin & CHIME/FRB Collaboration 2024). CHIME/FRB monitored these sources for years without detecting any bursts, before their spontanteous flare ups. Roughly speaking, the discovery rate of hyperactive FRBs is currently on the order of \sim 1 per year. It is not known whether these intense periods of activity signal an awakening from quiescence, or the birth of a new source, but studying the evolution of the burst properties during these periods can inform theory.

In this paper, we report on 696 bursts that have been detected from the hyperactive repeating FRB 20220912A, using the Nançay Radio Telescope (NRT) as part of the Extragalactic Coherent Light from Astrophysical Transients (ÉCLAT) monitoring campaign. Three of these bursts were the subject of in-depth analysis in Hewitt et al. (2023), which revealed that some bursts from FRB 20220912A feature extremely bright, broad-band, clustered shots of emission with durations on the order of microseconds. These shots are potentially produced by a different emission mechanism than the broader burst components that show substantial drift not related to dispersion i.e. the time-frequency drift (Hessels et al. 2019) that is colloquially referred to as the 'sad-trombone' effect. Here, we have analysed the spectro-temporal properties and the energetics of all 696 bursts in our NRT data, drawing population-wide statistics for comparison with other FRB sources. Section 2 outlines the observational set-up and data acquisition, followed by a description of how we obtained the burst sample in Section 3. The burst property analyses are explained in Section 4. We discuss the results in Section 5. in the context of other FRB 20220912A observations, and finally present our main conclusions in Section 6.

2 OBSERVATIONS

We observed FRB 20220912A as part of the ÉCLAT (PI: D. M. Hewitt) FRB-monitoring campaign on the NRT. The NRT is a Kraustype transit telescope located in Nançay, France; it is operated by the Observatoire de Paris, and associated with the French Centre National de la Recherche Scientifique (CNRS). At 1.4 GHz, the NRT has a system temperature of $T_{\rm sys} \sim 35$ K and a gain of $G \sim 1.4$ K Jy⁻¹, making it effectively as sensitive as a radio dish with a diameter of 94 m. ÉCLAT was established at the start of 2022 and has since been monitoring about a dozen FRB sources with approximately weekly cadence to constrain their activity levels and map the evolution of burst properties. Between 2022 October 15 and 2023 April 20, we observed FRB 20220912A in total 68 times, each observation lasting \sim 1 h. We initially observed the source nearly daily, depending on the priority of other observations at the NRT. When it became clear that the burst rate dropped to one or no bursts per observation around the end of 2022 November, we decreased the cadence to approximately 2.5 observations per week. No observations occurred between 2022 December 22 and 2023 March 4 due to a holiday break followed by a telescope maintenance period. The observation log is shown in appendix Table A1.

Our observations used the NRT's low-frequency receiver (1.1-1.8 GHz total span) and the Nançay Ultimate Pulsar Processing Instrument (NUPPI; Desvignes et al. 2011). We recorded data at a central observing frequency of 1.484 GHz, with 512 MHz of bandwidth (from 1.228–1.740 GHz) split into 128 4-MHz channels, with 16 channels from each of the 8 NUPPI nodes. NUPPI provided a time resolution of 16 μ s and full Stokes information in a linear

basis. For pointing, we used the original DSA-110 localisation of FRB 20220912A: RA (J2000) = $23^{h}09^{m}05.49^{s}$ and Dec. (J2000) = $48^{\circ}42'25.6''$ (Ravi et al. 2023). This position is 6.1 arcsec offset from the EVN localization (Hewitt et al. 2024), but the true position is still well within the half-power beam width of the NRT, which is at least a few arcminutes at 1.4 GHz. The EVN localization was derived after the observations presented here.

To maintain the best possible time resolution, our observations were coherently dedispersed within each 4-MHz channel, to a dispersion measure (DM) of 219.46 pc cm⁻³ (McKinven & Chime/Frb Collaboration 2022). The maximum DM offset at the bottom of the observing band that corresponds to a temporal smearing equal to the time resolution of the NRT data (16 μ s) is 0.9 pc cm⁻³. In other words, any burst with a DM lower than 218.56 pc cm⁻³ or higher than 220.36 pc cm⁻³ will experience residual temporal smearing due to imperfect coherent dedispersion. Since the best-fitting DM falls within this range, we can be confident that the bursts do not experience significant intra-channel DM smearing.

3 SEARCH AND DISCOVERY OF BURSTS

In this work, we have expanded on the ÉCLAT pipeline presented by Hewitt et al. (2023), to find bursts from FRB 20220912A. In short, the pipeline combines the coherence products AA and BB from the eight recorded subbands to create Stokes I filterbanks with the full 512-MHz bandwidth, before flagging radio frequency interference (RFI), in frequency but not time, using the RFIFIND tool from the PRESTO pulsar software suite (Ransom 2011). Thereafter, a boxcar match-filter search is conducted for impulsive transient signals, between a DM of 200 and $250 \,\mathrm{pc}\,\mathrm{cm}^{-3}$, using the GPUaccelerated software HEIMDALL.¹HEIMDALL searches the full 512-MHz band, using a signal-to-noise (S/N) threshold that we set at seven. The resulting burst candidates are classified as being either astrophysical or RFI using the deep-learning models 'A' to 'H' from the machine-learning classifier FETCH (Fast Extragalactic Transient Candidate Hunter, Agarwal & Aggarwal 2020; Agarwal et al. 2020). Finally, all candidates assigned a > 0.5 probability of being astrophysical are manually inspected to confirm they are from FRB 20220912A.

In our pipeline, FETCH frequently assigned a > 0.5 probability of being an astrophysical transient to RFI and, less often, FRBs were misclassified as RFI. Konijn (2023) established that within a sample of 7100 HEIMDALL candidates containing 263 FRBs from FRB 20220912A, FETCH incorrectly classified 54 FRBs (though note that FETCH has not been retrained specifically on the NRT data). To mitigate this, we incorporatedFETCH into a new classification method which we term CATCH(Classification Algorithm and Transient Candidate Handler; Konijn 2023). CATCHclassifies FRB candidates based on the structure in the DM-time space. A detailed description of CATCH is presented in Appendix B.

HEIMDALL has identified a total of 34 546 potential FRB candidates, after which our CATCHclassification found 696 FRBs with a subband S/N >7.0. Here, the subband S/N is the S/N value calculated only in the exact frequency and time region of the burst. At the start of the observations, the event rate was ~60 burst per hour, which decreased and stabilized near the end of the observations to ~1.3 bursts per hour. A subset of the bursts is shown in Fig. 1. These bursts are not representative of the complete set, but were selected to showcase bursts with high S/N and detailed structures.

¹https://sourceforge.net/projects/heimdall-astro/

4 BURST PROPERTIES AND ANALYSIS

Table 1 summarizes a small sample of the burst properties; the full table is available online. In the following sub-sections, we describe how these properties were measured. While the ÉCLAT observations include full polarimetric information, we differ the polarimetric analysis of these data to a future paper.

4.1 Spectro-temporal properties

Our transient classification method, CATCH, determined the frequency extent $(\Delta \nu)$ and temporal width (τ_{burst}) of each burst by outlining a rectangular box around the burst region in the dynamic spectrum. The positioning of this box was verified manually for each FRB to ensure that the box encompasses all burst components, including any drifting features, but not any high-intensity RFI. We define the middle of the box in frequency extent as the central observed frequency (ν_{mid}) of a burst.

Fig. 2 shows the temporal width distribution for the 696 bursts we detected from FRB 20220912A. Here, the temporal width of the burst is the width of the box that was fitted to the burst, hence it encompasses the earliest and latest times of emission in the case of multicomponent bursts. The median temporal width is 7.68 ms and the distribution is slightly skewed towards longer durations.

Fig. 3 shows $\Delta \nu$ against ν_{mid} for each burst, coloured by fluence. Given that ν_{mid} represents the middle of the frequency extent where burst emission is visible within our observing band, bursts with larger frequency extents will naturally have ν_{mid} approaching the centre frequency of our observing band. The maximum allowed frequency extent as a function of central frequency is shown by dashed black lines that constrain the 2D distribution to a triangular region. The frequency extent distribution exhibits an artificial peak at the maximum value, which is the result of many bursts having a frequency extent larger than our observing band. The median frequency extent is 248 MHz, while the central frequency distribution demonstrates a preference for lower central frequencies. The top histogram shows that more bursts are present at lower frequencies.

Multiple bursts show intensity fluctuations in their time-integrated spectra, likely related to diffractive interstellar scintillation from the Milky Way interstellar medium. However, detailed study of scintillation and scattering are beyond the scope of this paper and may be addressed in future work.

4.2 Fluence and energetics

The S/N and fluence of a burst are calculated within the spectrotemporal box determined by CATCH. We note that for some bursts, the emission clearly extends beyond our observing band, but we refrain from extrapolating and estimating frequency extents and fluences outside of the observing band. In these cases, the quoted values are thus lower limits. We calculated the fluence of a burst, F (Jy ms), by integrating the flux with respect to time, using the radiometer equation (Lorimer & Kramer 2005):

$$F = S/N \times SEFD \times \sqrt{\frac{\tau_{burst}}{n_{pol}\Delta\nu}},$$
(1)

where SEFD is the system equivalent flux density (~25 Jy for the NRT) and n_{pol} is the number of polarizations recorded in the observation (2 in this case). We do not correct for the sensitivity difference introduced by the 6.1" offset between our pointing centre and the EVN localisation, as the NRT half-power beam width at 1.4 GHz is 4 (RA) × 22 (Dec.) arcmin. This offset constitutes only ~7 per cent of the half power beam width, which is a minor



Figure 1. A subset of the 696 NRT-detected bursts from FRB 20220912A. This sample consists of very bright bursts, some of which show multiple sub-bursts and/or microsecond-time-scale intensity fluctuations. Each thumbnail shows a dynamic spectrum (bottom-left panel; $\Delta t = 16 \mu \text{s}$ and $\Delta v = 4 \text{ MHz}$), frequency-integrated light curve (top panel; integrated over the entire frequency band shown in the dynamic spectrum), and time-integrated spectrum (bottom right panel; integrated over the entire time range shown in the dynamic spectrum). All bursts have been coherently dedispersed within each channel to DM = 219.46 pc cm⁻³ during recording. Additionally, each burst has been incoherently dedispersed between channels to the DM value (units pc cm⁻³) reported in the top-right of every thumbnail. Also noted in the top-right of each thumbnail is the burst ID, which corresponds to the entries in Tables 1 and 2. Horizontal white lines in the dynamic spectru represent channels removed due to the presence of RFI. For visual purposes, the colour maps of the dynamic spectrum range from the value 0.0 to the 99th percentile. The plotted S/N values for these bursts are lower than those indicated in Table 1 because these data have not been downsampled, and the S/N is calculated across the entire bandwidth. A detailed analysis of bursts B411, B491, and B582 was previously presented in Hewitt et al. (2023).

Table 1. Properties of our sample of 696 NRT-detected bursts from FRB 20220912A (full table available	online).
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Burst ID	Time of arrival ^a (MJD)	Detection DM^b (pc cm ⁻³)	Temporal width ^c (ms)	Frequency extent ^c (MHz)	Central observed frequency ^c (GHz)	S/N ^{c, d}	Fluence ^c (Jy ms)
B000	59867.90197943	222.94	5.63	140	1.462	26.8	3.06
B001	59867.90198050	222.94	9.12	132	1.322	16.1	2.90
B002	59867.90229435	226.96	5.89	120	1.286	22.1	4.02
B003	59867.90471852	218.94	6.66	216	1.538	24.1	2.83
B004	59867.90471883	218.94	8.32	200	1.398	23.2	2.83
B005	59867.90517875	226.96	5.38	348	1.414	14.1	1.18
-	-	-	-	-	_	_	_

^aQuoted to ~ 1 ms precision. Corrected to the Solar System Barycentre at infinite frequency using the per-burst determined DM, a dispersion constant of $1/(2.41 \times 10^4)$ MHz² pc⁻¹ cm³ s, and the EVN-derived FRB 20220912A position (Hewitt et al. 2024)². The times quoted are dynamical times (TDB). ^bAs determined by HEIMDALL. ^cAs defined CATCH (see Appendix B). ^dSubband S/N values calculated on data downsampled by a factor of eight.

difference. The uncertainty in the SEFD already assumes an error of approximately 20 per cent.

Fig. 4 shows the fluence distribution of our NRT-detected bursts for the completeness thresholds that depend on assumed burst width. We estimated these completeness thresholds using equation (1), assuming an S/N = 7, a frequency extent of 248 MHz (the median value from fitting our burst sample), and three different temporal widths: the first-quartile value from our burst temporal duration distribution (5.63 ms), the second quartile (or median; 7.68 ms), and the third quartile (10.79 ms). We then fit a lognormal distribution to the data above the three aforementioned completeness thresholds. In all cases, the reduced χ^2 values of 1.31, 1.57, and 1.03 do not exceed the critical χ^2 values of 1.58, 1.60, and 1.60 at 95 per cent significance, with the degrees of freedom equal to the data size subtracted by the number of free parameters. This confirms that the fluence distribution is well fit by a lognormal function regardless of the assumed completeness threshold.

²https://github.com/MSnelders/FRB-Burst-TOA



Figure 2. The temporal width distribution for our sample of 696 NRTdetected bursts from FRB 20220912A. The vertical dashed line indicates the median at 7.68 ms.



Figure 3. The frequency extent, Δv , of each detected burst from FRB 20220912A is plotted against its central observed frequency v_{mid} , and coloured by its fluence. The right and top panels show the distributions of Δv and v_{mid} , respectively, where the NRT bandpass has been added to the top histogram. The error bars represent the Poissonian uncertainty, with a 1σ confidence interval. The dashed black triangular shape in the middle panel represents the limits of our observing band (see the text for more details). The colour of the bins of the side histograms represents the median fluence of the bursts within each bin.

The main panel of Fig. 5 shows the bursts' spectral energy densities, E_{spectral} , which we calculated as follows (adapted from equation 10 in Macquart & Ekers 2018):

$$E_{\text{spectral}} = \frac{F \times 4\pi D_L^2}{(1+z)^{2-\alpha}} \times 10^{-23} \text{ erg Hz}^{-1}, \qquad (2)$$

where the luminosity distance D_L is denoted in cm, the fluence F is denoted in Jy s, and α , the spectral index of the flux density or fluence, is set to be zero to be consistent with the fluence calculation. We calculate the fluence based on the band-averaged time series where we have assumed $\alpha = 0$. The redshift z and luminosity distance D_L



Figure 4. The burst fluence distribution for completeness thresholds 0.59, 0.69, 0.82 Jy ms, calculated for three temporal widths: 5.63, 7.68, and 10.79 ms. The areas on the left indicate the incompleteness regions, whereas the coloured areas around the fit indicate the 1σ fit error of a log-normal function to the distributions. Fit residuals are shown in the smaller panels beneath each fluence distribution.

are taken to be 0.0771 and 1.11×10^{27} cm (362.4 Mpc), respectively (Ravi et al. 2023).

Fig. 5, Panel A shows the duration of each observation (median of 0.92 h) together with the cumulative burst count, illustrating the high cadence of our observations. The majority of the bursts that were detected at the beginning of the observational campaign have spectral energy densities in the $10^{29}-10^{30}$ erg Hz⁻¹ range. The distribution of spectral energies is shown in Panel C, where a lognormal function is fitted to the data. The reduced χ^2 value is 1.09, which does not exceed the critical value of 1.39, indicating a good fit. Panel D shows the cumulative distribution of spectral energies. It depicts the burst rate per hour with spectral energies larger than a specific threshold. The errors on the rate are Poissonian, and we assume a fractional error of 20 per cent for the derived fluence values. The apparent turnover visible at lower energies is a reflection of bursts being detected close to the completeness threshold of the telescope, where it is easier to miss some bursts.

We fit a power law to the distribution using the Python POWER-LAW package (Clauset, Shalizi & Newman 2009). In addition to performing power-law fits, this package identifies the initial turnover point in the cumulative distribution to determine the appropriate minimum spectral energy from which to start the fit. Here, the initial turnover point is at $10^{29.51}$ erg Hz⁻¹. In the case of a perfectly lognormal spectral energy distribution, the start of the power-law fit (i.e. the mimimum spectral energy being fit) should coincide with the completeness threshold. The three completeness thresholds for bursts with temporal widths of 5.63 7.68, and 10.79 ms deviate from the start of the power law by approximately $10^{0.61}$, $10^{0.54}$, and $10^{0.47}$ erg Hz⁻¹, respectively. This could, in principle, indicate that the spectral energy distribution flattens at lower energies. Note that, the completeness threshold can also deviate even more from its estimated value, in which case the lognormal distribution would not fit the data well.



Figure 5. Temporal evolution of the spectral energy distribution of FRB 20220912A bursts detected with the NRT. Slashes through Panels A and B indicate a break in the continuity of the plotted timeline. Panel A: The bars show the duration of each observation, while the solid line indicates the cumulative burst count with time. Panel B: The green data points display the spectral energy of each detected burst, while the purple data points show the bursts' median spectral energy on each observing day. The vertical dashed line indicates the time when CHIME/FRB reported the discovery of FRB 20220912A and when we started observing with NRT. Panel C: The spectral energy distribution, considering all the detected bursts, is shown in green with a lognormal fit and the 1σ fit error indicate by the purple line and shaded region. The mean lies at $10^{29.48}$ erg Hz⁻¹, with a standard deviation of $10^{0.33}$ erg Hz⁻¹. Panel D: The cumulative spectral energy distribution fitted with a power law. The horizontal dotted purple line indicates the initial turnover point found by the PYTHONPOWERLAW package. The horizontal green lines indicate the theoretical completeness thresholds for various assumed burst widths, assuming a minimal S/N= 7 detection, as in Fig. 4. The solid green line indicates the completeness threshold for assumed burst widths at 5.63 ms; the dash-dotted line at 7.68 ms; and the dashed line at 10.79 ms. The distribution is seen to flatten at higher energies, indicating that bursts with these energies happen more frequently than predicted by extrapolating a simple power law based on fitting lower-energy bursts. This effect is less obviously present in Panel C, as several one-burst bins are obscured by the fit line.

The fitted power law has a slope of $-1.67 \pm 0.01 \pm 0.06$, where the first error is determined as the quadrature sum of the error on the fit using the fractional 20 per cent fluence errors and the second error is determined by bootstrapping. The data follows the power law until spectral energies of $10^{30.5}$ erg Hz⁻¹, after which the distribution appears to flatten.

The isotropic-equivalent energy (in erg) is the total energy released, assuming that the energy is radiated isotropically. We assume isotropic emission due to the unknown beaming angle, despite the expectation that FRB emission is highly directional. The isotropic-equivalent energy can be determined by multiplying the spectral energy by the bandwidth of the burst in Hz. Fig. 6 shows the temporal widths of the bursts plotted against their isotropicequivalent energies. The colours of the histograms are based on the median of the bursts inside each bin. Bursts with higher isotropicequivalent energies and greater temporal widths tend to span larger bandwidths, whereas bursts with lower isotropic-equivalent energies tend to span smaller bandwidths.

4.3 Dispersion measure and time-frequency drift

Determining the DM of a burst is non-trivial, due to the complex spectro-temporal morphology seen in many FRBs, along with the lack of a priori knowledge about how the burst should look. There is no exact model for each individual FRB and effects such as scintilation, the sad-trombone effect (Hessels et al. 2019), and the residual intra-burst drift not related to dispersion seen in many bursts (e.g. Rajabi et al. 2020; Chamma et al. 2021; Jahns et al. 2023) make

Burst ID	Time of arrival ^{<i>a</i>} (MJD)	$DM^b (pc cm^{-3})$	DM error ^{c} (pc cm ^{-3})	Min. fluctuation frequency $(1 \times 10^{-1} \text{ ms}^{-1})$	Max. fluctuation frequency $(1 \times 10^{-1} \text{ ms}^{-1})$	Temporal width ^d (ms)	Spectral extent ^d (MHz)
B038	59870.88721446	219.51	0.06	60	175	8.37	380
B050	59871.87238284	219.37	0.10	50	120	1.42	348
B090	59873.88115071	219.58	0.20	10	51	5.44	128
B267	59877.86661735	219.47	0.05	120	244	5.14	444
B318	59879.83220702	219.45	0.23	21	57	3.49	404
B384	59881.86993869	219.36	0.01	850	2050	47.78	452
B411 ^e	59881.83351237	219.39	0.10	29	102	26.08	308
B491 ^e	59884.83816794	219.38	0.01	1000	3125	10.50	500
B510	59889.82448343	219.40	0.01	1000	3125	8.67	412
B528	59891.81058320	219.64	0.10	33	112	8.21	248
B578	59898.82970864	219.62	0.22	10	50	1.98	220
B582 ^e	59899.78833990	219.72	0.02	200	550	9.73	492
B602	59900.82699925	219.68	0.06	40	220	7.95	484
B664	59917.75117914	222.53	0.48	7	27	6.42	440
B669	59919.74566971	223.09	0.28	20	50	6.66	392

Table 2. Properties of the FRB 20220912A bursts with fluctuation frequencies $> 2.5 \text{ ms}^{-1}$.

Notes.^{*a*}Quoted to ~ 1 ms precision. Corrected to the Solar System Barycentre at infinite frequency using the per-burst determined DM, a dispersion constant of $1/(2.41 \times 10^4)$ MHz² pc⁻¹ cm³ s, and the EVN-derived FRB 20220912A position (Hewitt et al. 2024)². The times quoted are dynamical times (TDB). ^{*b*}Computed using DM_phase within the spectro-temporal confines of the burst defined by CATCH. ^{*c*}See Appendix C.

^dThe temporal width and frequency extent of the burst defined by CATCH (see Appendix B).

^eThese three bursts are presented in detail in Hewitt et al. (2023). The DMs reported here are consistent with the DMs those authors determined, where they ensured that the extremely short time-scale broad-band structures (microshots) arrived at all observed frequencies at the same time.



Figure 6. The temporal widths of bursts plotted against their isotropicequivalent energy. The histograms are coloured based on the median value of the bursts within each bin.

it difficult and ultimately ambiguous to accurately determine the 'true' DM. In this paper, we determined the DMs using DM_phase (Seymour, Michilli & Pleunis 2019)³, which maximizes the coherent power of a burst across the bandwidth in order to determine a best DM. Using 32-bit, coherently dedispersed (DM = 219.46 pc cm⁻³) filterbank data, we estimated the DM for 15 bursts with subband S/N > 100 and visible (sub-)millisecond temporal structure. We have manually selected a suitable range of fluctuation frequencies to consider in DM_phase, excluding fluctuation frequencies above which there is no prominent emission (see Appendix C). The maximum fluctuation frequency can be inverted to obtain a time-scale of the shortest intensity fluctuations of the burst. We included bursts where the maximum fluctuation frequencies exceeded 2.5 ms⁻¹,



Figure 7. DM measurements for all bursts above aCATCH S/N of 100 and with the minimum fluctuation frequencies exceeding 2.5 ms^{-1} . The colour of each point indicates the subband S/N of the respective bursts. The dashed horizontal line, along with the shaded region, indicates the original DM and its associated error reported by McKinven & Chime/Frb Collaboration (2022). There appears to be a jump in the DM between MJDs 59 900 and 59917.

which ensures that the time-scales on which the intensity significantly changes are smaller than 4 ms. The selected bursts are: B038, B050, B090, B267, B318, B384, B411, B491, B510, B528, B578, B582, B602, B664, and B669. All are visible in Fig. 1. Our DM measurements are tabulated in Table 2.

In Fig. 7, the DM is plotted as a function of time. The DM appears constant over the majority of our campaign. However, towards the end of our campaign, we observe an apparent increase in the DM of 2.86 ± 0.54 pc cm⁻³ from Modified Julian Date (MJD) 59 900 to MJD 59917. We can determine the validity of this increase in DM by looking at the total drift rate of each bright (S/N > 30) burst.



Figure 8. The total drift of each burst with S/N > 30, where the DM has been fixed at 219.46 pc cm⁻³. The dashed grey line indicates no drift, whereas the black dotted line indicates the median drift of -8.8 MHz ms⁻¹. The histogram bars are coloured based on the median value of the bursts within each bin.

The total drift of a burst is mainly determined by three factors: the DM, the sad-trombone-induced drift between sub-bursts, and the intra-(sub-)burst drift. Assuming the influence of the sad-trombone drift and the intra-burst drift remains constant over time, any observed increase in the total drift over time could then be attributed to an increase in the DM. To determine the total drift, we follow the methodology presented in Gopinath et al. (2024), where we fix the DM to 219.46 pc cm⁻³ (the DM used for coherent dedispersion), and measured the burst drift for all bursts that have a S/N > 30. This entails fitting a 2D Gaussian function to the autocorrelation function (ACF) of the bursts, where the zero-lag noise spike has been masked. The drift rate is then essentially $cot(\theta)$, where θ is the angle of the 2D Gaussian, measured counterclockwise from the positive *y*-axis to the semi-major axis. The total drift is shown in Fig. 8.

We find a median drift rate of -8.8 MHz ms⁻¹. The drift rate appears to stay more or less constant throughout our campaign. As the burst rate (apparently) decreases, fewer measurements are available. We performed a Kolmogorov-Smirnov test using scipy.stats.kstest, comparing the drift measurements before and after MJD 60000, to see if there is a statistically significant decline in the magnitude of the drifts over time. We find p = 0.0986and K = 0.38, and thus cannot reject the null hypthesis that the samples are drawn from the same distribution. This lack of an increase in drift, given a fixed DM, also argues for a DM that is (roughly) constant over time. An increase in DM by $\sim 2 \,\mathrm{pc}\,\mathrm{cm}^{-3}$ would result in significant differences between the samples before and after MJD 60000. It is likely that we have absorbed some of the sad-trombone drift or intra-burst drift in bursts B664 and B669 when determining the DM - thus leading to an apparently higher DM when naively fitting those bursts. Alternatively, the higher DM values for B664 and B669 could arise from the bifurcating structure they exhibit, similar to that seen in FRB 20121102A bursts (Platts et al. 2021). Such a structure could introduce ambiguity in the DM measurement.

4.4 Wait-times

The time between successive bursts is referred to as the 'waittime'. We consider two bursts to be separate, rather than multiple components of a single burst, if the signal between the two highintensity regions returns to the level of the noise. In Fig. 9, we show the wait-time distribution for our sample of NRT-detected bursts. As for several other repeaters, e.g. FRB 20121102 (Li et al. 2021), FRB 20201124A (Xu et al. 2022), and FRB 20200120E



Figure 9. The distribution of the wait-times from FRB 20220912A plotted on a logarithmic scale. A double lognormal distribution has been fitted to the data, and the peaks of the wait-time distribution are at 33.40 ± 6.41 ms and 67.03 ± 2.61 s, respectively. Note that, the quoted error is the uncertainty on the location of the wait-time peak. The purple areas indicate the 1σ fit error and the bottom plot shows the fit residuals. The top plot shows the cumulative wait-time distribution of the second wait-time peak, to which a Weibull cumulative distribution function has been fitted. The total error on the shape parameter k is a combination of the error on the fit, and the error derived from bootstrapping 1000 times. The burst rate deviates from a Poisson distribution, as the shape parameter of the fitted Weibull CDF $k = 0.88 \pm 0.01 \pm 0.01$ is not equal to 1.0.

(Nimmo et al. 2023), the distribution is bimodal. We fit two lognormal functions to the peaks, yielding a reduced χ^2 value of 1.08, which is below the critical threshold of 1.36, indicating a good fit. To minimize the effect of outlier data points, and to more robustly estimate uncertainties, we have bootstrapped the data by randomly selecting 90 per cent of the data without replacement 1000 times, and only fitting that sample with a lognormal function. From our bootstrapping analysis, we find wait-time peaks at 33.40 ± 6.41 ms and 67.03 ± 2.61 s. Note that, the quoted error is the uncertainty on the location of the wait-time peak. The full width at half maximum of the shorter-time-scale (left) peak at 33.40 ms is 63.30 ms.

Also shown in Fig. 9 is the cumulative wait-time distribution for wait-times corresponding to the second peak at longer times. We can use this cumulative wait-time distribution to test if the burst rate is Poissonian, over the course of our observing campaign, or if there is clustering in time. If the bursts are distributed in time according to a Poisson process, then the distribution of intervals t_{wait} between subsequent bursts would have an exponential distribution. One generalization of an exponential distribution of intervals is the Weibull distribution

$$\mathcal{W}(t_{\text{wait}}|k,r) = k t_{\text{wait}}^{-1} [t_{\text{wait}} r \Gamma(1+1/k)]^k e^{-[t_{\text{wait}} r \Gamma(1+1/k)]^k}, \qquad (3)$$

where *r* is the burst rate, Γ is the gamma function, and *k* is the shape parameter (Papoulis & Unnikrishna Pillai 2002). The cumulative density function (CDF) of the Weibull distribution can be fitted to the cumulative wait-time distribution to test for possible burst clustering



Figure 10. The activity of FRB 20220912A during our observing campaign. The error bars on the burst rate represent the Poissonian uncertainty, with a 1σ confidence interval. Panel A shows the duration of each observation, T_{obs} , in hours. Panel B shows the average burst width per observation, where the errors indicate the root-mean-square deviation. Panel D stacks the frequency band of each burst per observation to illustrate the total number of bursts within each frequency channel.

(e.g. Oppermann et al. 2018). The Weibull CDF is defined as

$$P_{\text{Weibull}} = 1 - e^{-(t_{\text{wait}} r \, \Gamma(1+1/k))^k}.$$
(4)

This function is identical to the Poisson CDF when k = 1, while $k \neq 1$ implies that the bursts have a variable rate and cluster in time.

We find that the best-fitting Weibull burst rate is 31.7 ± 0.4 bursts/hour, with a shape parameter of $0.88 \pm 0.01 \pm 0.01$. The total error on the shape parameter is a combination of the error on the fit and the error derived from bootstrapping. As before, the data is bootstrapped 1000 times.

4.5 Burst rate

We mapped the burst activity and spectral evolution of FRB 20220912A bursts in Fig. 10. The main panel shows the average burst rate per hour, with 1 σ Poissonian errors, for each observation against the corresponding MJD. Shortly after the beginning of the observation campaign (MJD 59880), the burst rate peaked at 75^{+10}_{-9} bursts per hour, as seen by NRT. The burst rate was highly time-variable, with consecutive observations fluctuating by up to 90 per cent. In Panel A, we show the duration of each observation in hours. Despite our observations having more or less the same duration throughout the campaign, we witness a significant drop in the burst rate around MJD 59910. The burst rate rapidly peaked and then gradually diminished over a period of ~ 80 d, displaying variations in burst rate during this period.

Panel D of Fig. 10 is a density plot displaying a superposition of the frequency extent of each individual burst for every observation. When multiple bursts overlap, the coloured lines appear stronger. There is no obvious correlation with the burst rate and the location of the bursts in frequency space, within the observing bandwidth.

5 DISCUSSION

5.1 Burst morphology

Most of the bursts we detected are narrow-band, as is often the case for repeating FRBs (e.g. Chime/Frb Collaboration et al. 2023). A few bright bursts in our sample show 'microshots', extremely shorttime-scale, broader-band structures. These events are rare ($\leq 1 \%$ of the bursts in our sample show this), and we have only identified them in high-S/N bursts that also show millisecond-scale sub-bursts as well. The best examples in our sample were previously discussed in detail by Hewitt et al. (2023), who also showed that the occurrence of microshots is clustered in two high-fluence bursts (B411 and B491 in our naming convention). Such microshots have previously been observed superimposed on millisecond-duration bursts from repeaters FRB 20180916B (Nimmo et al. 2021) and FRB 20200120E (Nimmo et al. 2023). In the case of FRB 20121102A, isolated microshots were seen at high frequencies (3.9-9.3 GHz; Snelders et al. 2023). Overall, microshots appear to be a rare phenomenon in repeating FRBs, compared to the more typical millisecond-duration bursts, and they may have a different emission mechanism that is triggered by the same underlying event that creates the bursts in general (Hewitt et al. 2023).

The NRT bandpass is calculated using the NRT noise diode scans, and is visible in the top histogram of Fig. 3. It shows greater sensitivity at lower frequencies, potentially (partially) explaining the higher occurrence of bursts at lower frequencies. However, these lower frequencies are also more susceptible to RFI, significantly influencing the bandpass. Therefore, the bandpass may also somewhat reflect the higher levels of RFI, instead of the innate higher sensitivity. The apparent preference for burst emission at lower frequencies in the NRT observing band agrees with observations of FRB 20220912A by the Five-hundred-meter Aperture Spherical Telescope (FAST; Zhang et al. 2023) and the Allen Telescope Array (ATA; Sheikh et al. 2024). The top histogram of Fig. 3 also indicates that bursts at these lower frequencies were generally more energetic. Observations by FAST and ATA were conducted with an observing band that stretches to slightly lower frequencies (1000-1500 MHz and 900-2244 MHz for FAST and ATA, respectively) than our NRT data. The central frequency of the bursts was found to be drifting down by 6.12 \pm 0.76 MHz per day (Sheikh et al. 2024). Furthermore, the majority of burst emission seems to occur just below the NRT observing band which, in combination with the decreasing central frequencies of the bursts, gives rise to a more rapid drop in burst rate observed by NRT (around MJD 59900), compared to FAST and ATA.

5.2 Energetics

Plotted in Panel D of Fig. 5 is the cumulative spectral energy distribution, where the three representative completeness thresholds deviate from the start of the power law by approximately $10^{0.61}$, $10^{0.54}$, and $10^{0.47}$ erg Hz⁻¹. This could indicate that the spectral energy distribution flattens at lower energies. Zhang et al. (2023) observe a bimodality of the energy distribution for FRB 20220912A, and other repeating FRBs, like FRB 20121102A and FRB 20201124A, also show this bimodal energy distribution (e.g. Aggarwal et al. 2021; Li et al. 2021; Jahns et al. 2023). Due to our limited sensitivity compared with FAST, we are likely only sampling the tail of the distribution at lower energies, which is why the distribution also seems to flatten at lower energies and which is likely why the completeness thresholds and start of the power-law fit are misaligned. We do not attempt to model a low-energy turnover in the burst rate.

The cumulative spectral energy distribution also flattens at higher energies, showing that higher-energy FRBs occur more often than would be expected based on an extrapolation from a simple power law. This effect has been seen in other repeaters such as FRB 20121102A (Hewitt et al. 2022) and FRB 20201124A (Kirsten et al. 2024), and will be discussed in more detail for FRB 20220912A by Ould-Boukattine et al. (submitted), who incorporate additional data from several other telescopes as well as more advanced modelling. The flattening for FRB 20220912A has also been reported by Pelliciari et al. (2024), who compare various other literature observations of this source.

5.3 Wait-times

Fig. 9 shows the burst wait-time distribution. Similar to many other repeaters (e.g. Li et al. 2021; Hewitt et al. 2022; Nimmo et al. 2023), the distribution seen from FRB 20220912A is bimodal. We find that the longer-time-scale (right) peak of the wait-time distribution lies at 67.03 s, which is at longer time-scales than the equivalent wait-time peak seen from FAST observations of FRB 20220912A (18.05 s;

Zhang et al. 2023). The fluctuating burst rate, 20-fold sensitivity difference between the NRT and FAST, and the fact that FAST observations are typically half as long on average, all influence the position of this peak. Aggarwal et al. (2021), e.g. have shown how this right peak shifts to shorter time-scales as more bursts are detected in a constant-length observation.

The shorter-time-scale (left) peak in the wait-time distribution (33.40 ms) is comparable not only to what has been found by other telescopes observing this source, e.g. 50.98 ms (Zhang et al. 2023), but also to other sources such as FRB 20201124A (51 ms in Zhang et al. 2022 and 39 ms in Xu et al. 2022), and FRB 20121102A (24 ms in Hewitt et al. 2022). Note that, we classify candidates as separate bursts if the intensity between the peaks returns to the noise level. Other authors have used different methods to distinguish between a single multicomponent burst and multiple bursts, which could result in different peaks in the wait-time distribution (e.g. see Jahns et al. 2023, for a discussion and distinction). The increased likelihood of detecting a burst very shortly (< 100 ms) after another burst, appears to be universal among the most active repeaters, with the exception of FRB 20200120E, the globular cluster repeater. FRB 20200120E also shows a potential bimodality in its wait-time distribution, but rather than at tens of milliseconds, the short-wait-time peak is at 0.94 s (Nimmo et al. 2023). Future studies should aim to constrain whether the short-wait-time peak is constant with time for a particular source, and how the time-scale compares between sources. Additionally, the existing data sets need to be reanalysed with consistent assumptions above what constitutes a sub-burst versus a separate burst.

The time-scales of the short-wait-time peaks seen from repeaters might be associated with the physical size of the region where bursts are generated, or perhaps the time required for perturbations to propagate through this region (Nimmo et al. 2023; Totani & Tsuzuki 2023). Such a bimodality in the wait-time distribution has also been seen in short X-ray bursts from magnetars (e.g. Huppenkothen et al. 2015). While Huppenkothen et al. (2015) suggest that these wait-times align with both repeated crust failure and magnetospheric reconnection models, Wadiasingh & Chirenti (2020) propose that the shorter time-scales observed in the wait-time distributions of magnetars are more likely associated with crustal magnetar oscillations. Another alternative that might explain the presence of the short-wait-time peak is the occurrence of aftershocks following a starquake within a neutron star crust (Totani & Tsuzuki 2023). If consecutive bursts were indeed the result of aftershocks, one might expect a trend where fluence of the first burst is consistently higher; however, this relation has not been found in this work (see Appendix D), or in previous studies (e.g. Gourdji et al. 2019).

5.4 Burst rate

High variability in the burst rate, as illustrated for FRB 20220912A in Fig. 10, has also been seen in other repeating FRBs, namely: FRB 20121102, FRB 20200120E, FRB 20201124A, and FRB 20240114A (e.g. Li et al. 2021; Lanman et al. 2022; Nimmo et al. 2023; Shin & CHIME/FRB Collaboration 2024). This variability is not only seen in other FRBs but also in magnetars and pulsars that emit giant pulses. Magnetars can exhibit outbursts during which they emit numerous short X-ray bursts within a short time frame, typically ranging from tens to hundreds of bursts per hour (Gavriil, Kaspi & Woods 2004; Israel et al. 2008; van der Horst et al. 2012). An excellent example of this high variability was seen in 2020 from the magnetar, SGR 1935+2154, when it generated a millisecond-duration radio burst with a luminosity comparable to the lowest-luminosity FRBs (Bochenek et al. 2020; CHIME/FRB Collaboration et al. 2020). During its



(a) Dynamic spectrum of burst B664 dedispersed to a DM of 219.46 pc cm⁻³ and to a DM of 222.539 pc cm⁻³.



(b) Dynamic spectrum of burst B669 dedispersed to a DM of 219.46 pc cm⁻³ and to a DM of 223.090 pc cm⁻³.

Figure 11. Dynamic spectra of bursts B664 and B669, dedispersed to a DM of 219.46 $pc cm^{-3}$ and to the DM optimized for maximizing coherent power in each burst. The black dotted line marks the onset of each burst. Visual inspection suggests consistency with a DM of 219.46 $pc cm^{-3}$ for both bursts.

outburst phase, the X-ray burst rate was approximately 720 bursts per hour (Younes et al. 2020), but rapidly declined to approximately 29 bursts per hour within a mere 3-h period. In the context of giant pulse emitters, significant variations in the rate of giant pulses have also been observed across different observing epochs. For instance, in the case of the Crab pulsar (PSR B0531+21), the rate of giant pulses with high fluence (> 130 Jy ms) has been found to vary by up to a factor of 5 between consecutive observing days (Bera & Chengalur 2019).

Motivated by the lack of clear bimodality in the repetition rate distribution of FRB sources (repeaters versus non-repeaters; Chime/Frb Collaboration et al. 2023), and the potential of a correlation between repetition rate and burst (temporal) width (Connor et al. 2020), we searched for a correlation between the burst rate and burst width in our sample. Such a correlation across a population of sources could be attributed to differences in the beaming angle, whereby FRBs with a broader beaming angle would exhibit higher observed repetition rates. The top panel of Fig. 10 shows the median burst width per observation against the burst rate of that observation. We see no correlation between the median burst width and burst rate on a per-observation basis, concluding that for our sample of bursts from FRB 20220912A the burst width is independent of activity rate. While this correlation does not exist for our sample of bursts from FRB 20220912A, it might still exist across a larger sample of different sources and we encourage future works to investigate this.

The total wait-time distribution, considered over the entire observing campaign, shows clustering, but individual observations on hourlong time-scales appear Poissonian (see Appendix E). To investigate further, we applied an identical Weibull analysis to the wait-time distribution of the FAST data presented in Zhang et al. (2023). These observations have higher burst rates, and have shorter observation durations (~30 min). Using the total set of observations, we obtained a shape parameter of $k = 0.93 \pm 0.003 \pm 0.01$. Despite FAST's increased sensitivity, higher burst rate, and the shorter duration of the observations, it also demonstrates clustering in the wait-times.

5.5 DM and Drift

Fig .7 shows the DM of bursts with S/N > 100 and fluctuation frequencies exceeding 2.5 ms^{-1} , where we observed an apparent increase in the DM from MJD 59 900 to MJD 59920. However, this apparent increase in the DM was not corroborated when we

looked at the total drift rate of a set of bright (S/N > 30) bursts, shown in Fig. 8. We therefore conclude that we have incorporated the sad-trombone drift or intra-burst drift into the DM calculation when determining the DMs of bursts B664 and B669. Fig. 11 shows both B664 and B669 dedispersed to a DM of 219.46 pc cm⁻³ and to the DM optimized for maximising coherent power in each burst. Here, both bursts appear consistent with a DM of 219.46 pc cm⁻³. Although we do not find any large DM variations over time, (Hewitt et al. 2023) reported robust evidence for DM variations on the level of ~ 0.3 pc cm⁻³ on the time-scale of weeks – using broad-band microshots to unambiguously determine the DM accurately and precisely.

We find a mean drift rate for FRB 20220912A of -8.8 MHz ms⁻¹, which is a combination of the excess DM beyond 219.46 pc cm⁻³, the drift induced by the sad-trombone effect, and the intra-burst drift. Our mean drift is more than an order-of-magnitude smaller than those quoted for FRB 20121102A in Hessels et al. (2019) (~ -200 MHz ms⁻¹), where the sad trombone effect dominated the drift rate. Our mean drift rate is comparable with the drift rate from FRB 20180916B quoted in Pastor-Marazuela et al. (2021) at similar frequencies (-39 ± 7 MHz ms⁻¹). Note that, our quoted drift rate is likely smaller, due to the fact that we have fixed our DM to a constant value, and due to contamination between sub-burst drift and intraburst drift. Our drift rate is also comparable to FRB 20201124A, where Zhou et al. (2022) quotes a mean single component drift rate of -61 ± 9 MHz ms⁻¹, and a mean multi-component drift rate of -21 ± 4 MHz ms⁻¹.

6 CONCLUSIONS & FUTURE WORK

We performed a study of the repeating FRB 20220912A using the NRT, where a total of 61 h of observations were conducted over the course of 6 months. Employing the new CATCHpipeline, a total of 696 bursts were detected, with an observed event rate up to 75^{+10}_{-9} bursts per hour (above fluence 0.59 Jy ms), demonstrating the remarkably high activity of the source, which has also been noted by other studies (e.g. Feng et al. 2022; McKinven & Chime/Frb Collaboration 2022; Zhang et al. 2023). FRB 20220912A shows various features which appear to be common among (hyper)active repeaters:

(i) The wait-time distribution is bimodal, showing peaks at 33.4 ms and 67.0 s. The short-wait-time peak at 33.4 ms is roughly consistent between repeating FRB sources, to within a factor ~ 2 (e.g. Hewitt et al. 2022).

(ii) The cumulative spectral energy distribution flattens towards the highest energies ($E_{\nu} \gtrsim 10^{31} \text{ erg Hz}^{-1}$), as also seen for another hyperactive repeater (Kirsten et al. 2024). We will analyse and discuss this in more detail in an upcoming paper (Ould-Boukattine et al. tted).

(iii) The DM is roughly constant in time when taking into account the confusing factor of 'sad-trombone' time-frequency drift (δ DM $\lesssim 2 \,\mathrm{pc}\,\mathrm{cm}^{-3}$). None the less, we confirm the results of Hewitt et al. (2023), and find that small DM variations, on the order of $\sim 0.3 \,\mathrm{pc}\,\mathrm{cm}^{-3}$, are occurring.

(iv) The bursts show a mean 'sad trombone' drift rate of $-8.8 \text{ MHz} \text{ ms}^{-1}$ (for bursts with S/N > 30). This is comparable to some known repeaters, but much less compared to others measured in the same frequency range (e.g. Hessels et al. 2019).

(v) We observe that more bursts occur towards the bottom of the NRT observing band, at around 1350 MHz, consistent with quasicontemporaneous observations from FAST (Zhang et al. 2023) and ATA (Sheikh et al. 2024).

(vi) While the burst rate within \sim 1-h observations appears Poissonian, we find that the rate in the entire observing campaign is better modelled by a Weibull distribution, with shape parameter $k = 0.88 \pm 0.01 \pm 0.01$, showing clustering.

(vii) The 16- μ s time resolution of the data allows us to probe microshots in the bursts, but we find that these occur in only $\lesssim 1 \%$ of the bursts in our sample. The rare occurrence of microshots has also been seen in other repeaters (e.g. Nimmo et al. 2023).

The breadth of phenomena displayed by repeating FRBs provides multiple dimensions to quantitatively compare their properties and to constrain theoretical models of their nature. ÉCLAT is continuing to monitor repeating FRBs at 1218–1740 MHz (and sometimes also at > 2 GHz) with approximately weekly cadence, providing a valuable complement to CHIME/FRB's daily monitoring of these sources at 400 – 800 MHz.

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DATA AVAILABILITY

The relevant code and data products for this work will be uploaded on Zenodo at the time of publication.

REFERENCES

Agarwal D., Aggarwal K., 2020, devanshkv/fetch: Software release with the manuscript

- Agarwal D., Aggarwal K., Burke-Spolaor S., Lorimer D. R., Garver-Daniels N., 2020, MNRAS, 497, 1661
- Aggarwal K., Agarwal D., Lewis E. F., Anna-Thomas R., Tremblay J. C., Burke-Spolaor S., McLaughlin M. A., Lorimer D. R., 2021, ApJ, 922, 115
- Bera A., Chengalur J. N., 2019, MNRAS, 490, L12
- Bhandari S. et al., 2023, ApJ, 958, L19
- Bhusare Y. et al., 2022, Astron. Telegram, 15806, 1
- Bochenek C. D., Ravi V., Belov K. V., Hallinan G., Kocz J., Kulkarni S. R., McKenna D. L., 2020, Nature, 587, 59
- CHIME/FRB Collaboration, 2020, Nature, 587, 54
- Camilo F., Reynolds J., Johnston S., Halpern J. P., Ransom S. M., 2008, ApJ, 679, 681
- Chamma M. A., Rajabi F., Wyenberg C. M., Mathews A., Houde M., 2021, MNRAS, 507, 246
- Chime/Frb Collaboration, 2023, ApJ, 947, 83
- Clauset A., Shalizi C. R., Newman M. E. J., 2009, SIAM Rev., 51, 661
- Connor L., Miller M. C., Gardenier D. W., 2020, MNRAS, 497, 3076
- Cordes J. M., McLaughlin M. A., 2003, ApJ, 596, 1142
- Dai S. et al., 2019, ApJ, 874, L14
- Desvignes G., Barott W. C., Cognard I., Lespagnol P., Theureau G., 2011, in Burgay M., D'Amico N., Esposito P., Pellizzoni A., Possenti A., eds, AIP Conf. Ser. Vol. 1357, Radio Pulsars: An Astrophysical Key to Unlock the Secrets of the Universe. Am. Inst. Phys., New York, p. 349
- Feng Y. et al., 2022, Astron. Telegram, 15723, 1
- Gajjar V. et al., 2018, ApJ, 863, 2
- Gavriil F. P., Kaspi V. M., Woods P. M., 2004, ApJ, 607, 959
- Gopinath A. et al., 2024, MNRAS, 527, 9872
- Gourdji K., Michilli D., Spitler L. G., Hessels J. W. T., Seymour A., Cordes J. M., Chatterjee S., 2019, ApJ, 877, L19
- Gourdji K., Rowlinson A., Wijers R. A. M. J., Goldstein A., 2020, MNRAS, 497, 3131
- Hessels J. W. T. et al., 2019, ApJ, 876, L23
- Hewitt D. M. et al., 2022, MNRAS, 515, 3577
- Hewitt D. M. et al., 2023, MNRAS, 526, 2039
- Hewitt D. M. et al., 2024, MNRAS, 529, 1814
- Huppenkothen D. et al., 2015, ApJ, 810, 66
- Israel G. L. et al., 2008, ApJ, 685, 1114
- Jahns J. N. et al., 2023, MNRAS, 519, 666
- Kirsten F., Snelders M. P., Jenkins M., Nimmo K., van den Eijnden J., Hessels J. W. T., Gawroński M. P., Yang J., 2021, Nat. Astron., 5, 414
- Kirsten F. et al., 2022a, Nature, 602, 585
- Kirsten F. et al., 2022b, Astron. Telegram, 15727, 1
- Kirsten F. et al., 2024, Nat. Astron., 8, 337
- Konijn D., 2023, Uva/VU Master Thesis, 35
- Kumar P., Qu Y., Zhang B., 2024, preprint (arXiv:2406.01266)
- Lanman A. E. et al., 2022, ApJ, 927, 59
- Li B., Li L.-B., Zhang Z.-B., Geng J.-J., Song L.-M., Huang Y.-F., Yang Y.-P., 2019, preprint (arXiv:1901.03484)
- Li D. et al., 2021, Nature, 598, 267
- Lorimer D. R., Kramer M., 2005, Handbook of Pulsar Astonomy, Cambridge University Press, 176, 202, 264
- Lorimer D. R., Bailes M., McLaughlin M. A., Narkevic D. J., Crawford F., 2007, Science, 318, 777
- Macquart J. P., Ekers R., 2018, MNRAS, 480, 4211
- Marcote B. et al., 2020, Nature, 577, 190
- Margalit B., Berger E., Metzger B. D., 2019, ApJ, 886, 110
- Margalit B., Metzger B. D., Sironi L., 2020a, MNRAS, 494, 4627
- Margalit B., Beniamini P., Sridhar N., Metzger B. D., 2020b, ApJ, 899, L27
- McKinven R., Chime/Frb Collaboration, 2022, Astron. Telegram, 15679, 1
- Nimmo K. et al., 2021, Nat. Astron., 5, 594
- Nimmo K. et al., 2022, Nat. Astron., 6, 393
- Nimmo K. et al., 2023, MNRAS, 520, 2281
- Niu C. H. et al., 2022, Nature, 606, 873
- Oppermann N., Yu H.-R., Pen U.-L., 2018, MNRAS, 475, 5109
- Ould-Boukattine O. S. et al., 2022, Astron. Telegram, 15817, 1

Ould-Boukattine O. S. et al., submitted, A probe of the maximum energy of fast radio bursts through a prolific repeating source Papoulis A., Unnikrishna Pillai S., 2002, Probability, random variables and stochastic processes. McGraw-Hill, Boston Pastor-Marazuela I. et al., 2021, Nature, 596, 505 Pelliciari D. et al., 2024, A&A, 690, A219, Perera B. et al., 2022, Astron. Telegram, 15734, 1 Petroff E., Hessels J. W. T., Lorimer D. R., 2022, A&A Rev., 30, 2 Platts E. et al., 2021, MNRAS, 505, 3041 Pleunis Z. et al., 2021a, ApJ, 911, L3 Pleunis Z. et al., 2021b, ApJ, 923, 1 Rajabi F., Chamma M. A., Wyenberg C. M., Mathews A., Houde M., 2020, MNRAS, 498, 4936 Rajwade K. et al., 2022, Astron. Telegram, 15791, 1 Ransom S., 2011, Astrophysics Source Code Library, record ascl:1107.017 Ravi V. et al., 2023, ApJ, 949, L3 Seymour A., Michilli D., Pleunis Z., 2019, Astrophysics Source Code Library, record ascl:1910.004 Sheikh S. et al., 2022, Astron Telegram, 15735, 1 Sheikh S. Z. et al., 2024, MNRAS, 527, 10425 Shin K., CHIME/FRB Collaboration, 2024, Astron. Telegram, 16420, 1 Snelders M. P. et al., 2023, Nat. Astron., 7, 1486 Spitler L. G. et al., 2016, Nature, 531, 202

Thornton D. et al., 2013, Science, 341, 53

Totani T., Tsuzuki Y., 2023, MNRAS, 526, 2795

Wadiasingh Z., Chirenti C., 2020, ApJ, 903, L38
Wadiasingh Z., Timokhin A., 2019, ApJ, 879, 4
Wang W., Luo R., Yue H., Chen X., Lee K., Xu R., 2018, ApJ, 852, 140
Xu H. et al., 2022, Nature, 609, 685
Younes G. et al., 2020, ApJ, 904, L21
Zhang Y.-K. et al., 2022, Res. Astron. Astrophys., 22, 124002
Zhang Y.-K. et al., 2023, ApJ, 955, 142
Zhou D. J. et al., 2022, Res. Astron. Astrophys., 22, 124001
van der Horst A. J. et al., 2012, ApJ, 749, 122

SUPPORTING INFORMATION

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APPENDIX A: OBSERVATION DURATIONS

The 68 total observations, each lasting approximately 1 h, are summarized in Table A1.

Table A1. NRT observation log for FRB 20220912A.

Observation ID	Start (MJD) ^a	Duration (s)	Observation ID	Start (MJD) ^a	Duration (s)
01	59867.90178	1298.7	O35	59906.76284	2425.7
O2	59869.87766	939.6	O36	59907.77528	2660.7
03	59870.86635	2004.7	O37	59909.75402	3952.2
O4	59871.86137	3649.0	O38	59911.74708	4103.1
O5	59873.85096	4138.0	O39	59912.73449	3417.0
O6	59874.84975	2429.7	O40	59915.73991	3792.0
O7	59875.85090	3578.9	O41	59917.73277	3980.9
O8	59876.84022	2172.6	O42	59919.72279	4285.1
O9	59877.84736	3451.9	O43	59921.72711	3451.9
O10	59878.83732	2560.8	O44	59923.71440	4139.1
011	59879.82461	4915.2	O45	59925.70626	4285.0
O12	59881.82641	4286.1	O46	59927.69795	4541.1
O13	59882.82632	2562.8	O47	59931.68961	4340.2
O14	59883.84423	2390.7	O48	59935.68565	3826.6
015	59884.82111	1884.6	O49	60007.49410	2948.7
O16	59884.84769	1849.9	O50	60011.47434	2559.7
O17	59885.81839	2559.8	O51	60012.47868	3451.9
O18	59887.80276	3418.0	O52	60014.47088	3649.0
O19	59888.81363	3792.0	O53	60016.46052	4138.1
O20	59889.80460	2848.8	O54	60019.46660	3652.0
O21	59890.80948	3648.1	O55	60020.44704	2846.8
O22	59891.80041	2706.8	O56	60021.45762	3199.8
O23	59892.80272	1748.7	O57	60023.45200	3182.9
O24	59892.82765	1696.8	O58	60026.44155	3335.9
O25	59893.79641	2560.8	O59	60030.41960	4284.0
O26	59894.78365	4915.3	O60	60033.41924	3274.9
O27	59896.78825	4136.1	O61	60036.41660	2678.8
O28	59897.78272	2897.0	O62	60040.40335	2950.0
O29	59898.78065	4272.2	O63	60042.40042	2671.4
O30	59899.78016	2559.8	O64	60044.39492	2678.8
O31	59900.77444	4336.1	O65	60047.38420	2950.0
O32	59901.76453	3417.0	O66	60049.38117	3199.9
O33	59903.76904	4178.0	O67	60051.37340	3314.9
O34	59904.76995	2288.6	O68	60054.36522	3314.8

Note.^aTopocentric at NRT.

APPENDIX B: THE CLASSIFICATION ALGORITHM AND TRANSIENT CANDIDATE HANDLER

Ideally, each FRB candidate from the search pipeline is validated by eye. However, conducting a manual review of tens of thousands of candidates, generated from tens of hours of observations, is not feasible or reproducible. In most FRB searches, convolutional neural network classifiers (e.g. FETCH) have been employed to greatly reduce the number of candidates requiring manual inspection. However, these classifiers are not perfect (especially when not retrained) and occasionally misclassify real FRBs.

In an attempt to address this issue, we developed a classifier algorithm called 'The Classification Algorithm and Transient Candidate Handler' (CATCH, for short). CATCHparses all candidates created by HEIMDALL that fulfill *any* of the following criteria: (i) FETCH's deeplearning models 'A' to 'H' indicate a probability larger than 0.5 that the candidate is astrophysical; (ii) candidates that have a subband S/N larger than 200; (iii) candidates identified based on their brightness distribution in the DM-time space through a method we will refer to as the 'bow tie' method (Konijn 2023).

B1 Identification of the bow tie feature

The bow tie method tries to classify candidates based on their brightness distribution in the DM-time spectrum, i.e. the frequency-integrated burst brightness as a function of DM and time. The DM-time spectrum is created by dedispersing the dynamic spectrum of a burst for a range of trial DM values, and then integrating across all frequencies. In this parameter space, an astrophysical transient with a large non-zero DM, gives rise to a symmetric 'bow tie'-like shape, as visible in Fig. B1.

Terrestrial RFI is not significantly dispersed (though it can be frequency-swept), effectively allowing for the identification of astrophysical candidates based on the presence of the bow tie shape in the DM-time spectrum. In an ideal case, the bow tie can be identified by comparing the intensity at the centre of the bow tie (at the true DM) with the intensity at the edges (at some other trial DM). However,



Figure B1. Frequency-integrated burst brightness as a function of DM and time for burst B049. The defining 'bow tie'-shaped feature of an impulsive and dispersive astrophysical signal is clearly visible. Fine 'spokes' are also visible in the bow tie, indicating time-frequency structure in the dynamic spectrum.



Figure B2. Dynamic spectrum of burst B049. A box has been placed surrounding the highest intensity areas, and is plotted as the dashed rectangle. The horizontal bands are the zapped frequency channels, which complicates the process of determining the frequency range of the box. Added to the top and the side of the dynamic spectrum are the one dimensional integrated spectra, where the dashed lines again indicate the location of the box.

this comparison is difficult because the bow tie can vary in size, intensity, opening angle, and pointing angle. To ensure consistency in our comparison, given the different spectro-temporal features of bursts, we first vertically align the bow tie, after which we estimate the size of the bow tie based on the duration and bandwidth of the burst in the dynamic spectrum (frequency-time space).

The bow tie can be vertically aligned by correcting for the temporal delay between the highest and lowest frequency of the burst caused by the dispersion. These frequencies can be obtained by placing a box around the burst in the dynamic spectrum, after masking those channels contaminated by RFI. The optimal location and size of this box are determined by maximising the summed intensity divided by the square root of its area. An example of an optimised box is shown in B2.

B2 Bow tie size

The next step involves determining the size of the bow tie, primarily influenced by the frequency extent and temporal width of the burst. Bursts with large frequency ranges or narrow temporal widths have their frequency-integrated intensity significantly affected by slight variations in DM, resulting in a smaller bow tie. We can calculate this change in intensity due to different DM values per burst and use it to characterize the size of the bow tie.

Due to dispersion, there is a frequency-dependent time-smearing effect (Cordes & McLaughlin 2003). This is equal to

$$\Delta t_{\rm DM} = 8.3 \cdot \mu s \cdot \rm DM \cdot \Delta \nu_{\rm MHz} \cdot \nu_{\rm GHz}^{-3}, \tag{B1}$$

for a frequency extent $\Delta \nu_{MHz}$, an observing frequency ν_{GHz} , and a dispersion measure DM. Using a DM value that is incorrect by an amount δ DM increases the temporal width of the pulse by

$$\Delta t_{\delta \rm DM} = \Delta t_{\rm DM} (\delta \rm DM/\rm DM). \tag{B2}$$

By combining these equations, an expression for the time smearing as a function of incorrect DM value can be found:

$$\Delta t_{\delta \text{DM}} = 8.3 \ \mu\text{s} \cdot \Delta \nu_{\text{MHz}} \cdot \nu_{\text{GHz}}^{-3} \cdot \delta \text{DM}. \tag{B3}$$





(a) The DM-time spectrum of an RFI candidate from FRB 20220912A detected by the NRT.

(b) The DM-time spectrum for burst B049.

Figure B3. Frequency-integrated brightness as a function of DM and time ('DM-time spectrum') are shown for an RFI candidate and for burst B049. In the centre of the plot, a box is placed at HEIMDALL's estimated DM value. Additionally, at the top and bottom of the plot, a set of boxes is positioned at the trial DM values where the burst would be smeared in time by a factor of 10. The dotted lines indicate the width of the region for which the intensity is calculated. The intensity of the RFI candidate does not decrease in the top and bottom regions compared to the central box. In contrast, for the FRB candidate, the intensity does decrease, signifying that this candidate is indeed an FRB.

Reordering terms results in an equation for δDM as a function of the temporal smearing:

$$\delta DM = \Delta t_{\delta DM} / (8.3 \ \mu s \cdot \Delta \nu_{MHz} \cdot \nu_{GHz}^{-3}). \tag{B4}$$

Since the total power of the burst remains constant, time smearing that will increase the width of the burst by a factor N will decrease the per-pixel intensity in the DM-time spectrum by a factor sqrt(N). By assuming $\Delta t_{\delta DM} = 10 \times t_{burst}$, we can compute the error associated with the DM required to achieve this temporal width. Thus, for each burst, we can calculate the number of DM trials where the bow tie width will be smeared by a factor of ten and the intensity reduced by a factor of $\sqrt{10}$. We define the size of the symmetrical bow tie as the number of DM units on both sides that reduce the intensity with a factor of $\sqrt{10}$.

To identify the bow tie, we compare the intensity in the centre of the bow tie with the intensity at the top and bottom of the bow tie. The width of the box is equal to the temporal width of the burst in the dynamic spectrum, and the height of the box is calculated using the above equations where $\Delta t_{\delta DM} = 3 \times t_{burst}$, to take into account the smearing of the burst. The amount of DM units between the centre of the box and the top and bottom has been calculated using $10 \times t_{burst}$. Here, the box is slid across the spectrum at the calculated trial DM using a sliding window of $10 \times t_{burst}$ ms, where the box with the highest intensity is selected for comparison with the box in the centre.

An example is shown in Figs B3(a) and B3(b) for an RFI candidate and for burst B049, respectively. For the RFI candidate, the intensity remains relatively consistent at both the top and bottom of the plot, whereas, for burst B049, there is a substantial decrease in intensity. The ratio between these intensities is a reliable indicator for distinguishing between FRBs and RFI, and we dub this the 'bow tie intensity ratio'. For the RFI candidate, the bow tie intensity ratio will never exceed 1.0. However, for burst B049, as the intensity decreases away from the centre, the bow tie intensity ratio will consistently be greater than one.

B3 Classification of burst candidates

To assess the validity of our algorithm, we must look at the sensitivity. In this context, 'sensitivity' is a classification metric that represents the number of true positives divided by the number of true positives, TP, plus the number of false negatives, FN:

Sensitivity =
$$\frac{\text{TP}}{\text{TP} + \text{FN}}$$
. (B5)

This metric penalizes false negative predictions, thus penalizing the misclassification of real FRBs. A perfect classifier would achieve a sensitivity of 1.0, as it would not misidentify any FRBs.

Sensitivity is unaffected by the number of false positives, so optimizing for sensitivity often results in a large number of potential FRB candidates that require manual examination. For instance, an algorithm can achieve a sensitivity of 1.0 by classifying all potential candidates as FRBs. We therefore try to optimize the sensitivity for our classifier while keeping the number of false positives below a manageable threshold. We have placed this threshold at 15 per cent of the total number of candidates.

CATCH classifies candidates based on the ratio between the average intensity in the central box and the highest average intensity in one of the boxes at the trial DM value that would smear the burst width by a factor of 10. To find this ratio, we have trained the algorithm using 7104 manually inspected burst candidates from eleven randomly selected datasets from FRB 20220912A. CATCHclassified these candidates based on whether their bow tie intensity ratios exceeded a specific threshold value. By analysing CATCH's performance at varying threshold levels, we can find the threshold at which the algorithm optimally performs. Fig. B4 shows the sensitivity of



Figure B4. The sensitivity of solely the bow tie classifier (at different bow tie ratio thresholds) as a function of the percentage of false positives. CATCH incorporates all candidates from the bow tie classifier, all candidates from FETCH, and all candidates that have a subband S/N larger than 200. The colour of the crosses indicates at what value the classification threshold has been placed. The dashed grey line represents the maximum fractional false positive limit, which has been set at 15 per cent of the total number of candidates.

this bow tie method plotted against the number of false positives, where the colour of the crosses indicates the thresholds. The best performance is achieved at a treshold level of 1.24.

Machine learning algorithms, like the conventionally used, FETCH, demonstrate a higher level of robustness when confronted with singular noisy frequency channels in the DM-time spectrum, but have difficulty with large RFI structures. To create a classifier that achieves the highest sensitivity possible, CATCH includes the candidates that FETCH identifies as FRBs with a probability higher than 50 per cent, all the candidates which have an S/N ratio above 200, and all candidates that are selected by the bow tie method. The result is a significant increase in sensitivity, without exceeding the 15 per cent limit for false positives. Fig. B5 shows the sensitivity of CATCH plotted against the number of false positives using varied classification ratios, like in Fig. B4. The sensitivity of FETCH has been plotted for comparison, visualized with crosses, which includes the static classification threshold across all models and randomly sampled classification thresholds for each model separately. At the aforementioned 15 per cent maximum false positive threshold, CATCH attains a sensitivity of 99.62 per cent on the test data set, surpassing the sensitivity of the conventionally used machine learning algorithm by ~ 29 times.

CATCH has proven valuable, finding bursts in the training data set with a sensitivity of 99.6 per cent, and an accuracy of 88.8 percent, while achieving a false positive total below the 15percnt imposed limit. In the 60.99 h of data obtained from the hyperactive FRB 20220912A, CATCH successfully detected 696 FRBs, surpassing the number that the standard method would have found by 26 per cent. Although CATCH displays promise, it is pertinent to acknowledge



Figure B5. The sensitivity of CATCH and FETCH plotted against the percentage of false positives. FETCH has been plotted as crosses, including the static classification threshold across all models and randomly sampled thresholds. CATCH has been plotted as circles. The dashed grey line marks the 15 per cent fractional limit on the total number of candidates. At the dashed grey line limit, CATCH vastly outperforms FETCH in terms of sensitivity.

the identified limitations and challenges. First, to achieve such high sensitivity, CATCH classifies a greater number of candidates as FRBs, leading to a higher number of false positives. This approach proves valuable when analysing data containing many FRBs, as it vastly reduces the likelihood of missing real FRBs. However, for real-time systems where the false positive rate *must* be low, or when monitoring sources that are not hyperactive, this method is not preferable. Secondly, when evaluating the types of bursts that conventional classification algorithms (such as FETCH) misclassify but CATCH correctly identifies, we find that lower S/N bursts are typically misclassified. However, the conventional classification algorithm also misclassified several brighter bursts (S/N ~ 20). A comprehensive discussion of these issues has been undertaken by Konijn (2023), and for brevity, we will refrain from further delving into the details here.

APPENDIX C: DETERMINING THE DISPERSION MEASURE UNCERTAINTY

We estimated the error associated with the DM by calculating the δ DM that corresponds to the inverse of the highest fluctuation frequency at which the burst still exhibits significant intensity in the power spectrum. Fig. C1 shows the coherent power of burst B586, plotted for various fluctuation frequencies against a range of DM values. At the DM of 219.717 pc cm⁻³, we can see that the coherent power extends to fluctuation frequencies of ~5.5 ms⁻¹. The inverted fluctuation frequency indicates the shortest intensity fluctuations of the burst (182 µs). Subsequently, we can calculate the number of DM units required to induce a variation on this time-scale, corresponding to the error on the DM.



Figure C1. The coherent power of burst B586, plotted for a range of fluctuation frequencies against a range of DM values. The higher intensity areas indicate vertically aligned structures across the frequency band. At the lower end of the fluctuation frequencies, the burst exhibits vertical structures across the entire range of DM values. However, at higher fluctuation frequencies, only the small-scale structures align, specifically when the DM is $219.717 \,\mathrm{pc} \,\mathrm{cm}^{-3}$, indicative of the true DM of the burst. There is no apparent coherent power above fluctuation frequencies of $5.5 \,\mathrm{ms}^{-1}$, which is the maximum fluctuation frequency of this burst.

APPENDIX D: CONSECUTIVE BURST FLUENCES

The ratios of fluences between two consecutive bursts that occurred within one second of each other are shown in Fig. D1. The median of the ratios is at 1.01 and shows that, similar to previous studies (e.g. Gourdji et al. 2019), there seems to be no relation between the fluences of consecutive bursts.



Figure D1. The ratio of fluences between bursts that occurred within one second of each other, plotted against the time between the bursts. The colour of each point represents the average fluence of the burst pair. The distributions of both the fluence ratio and the time between bursts are displayed at the top and right of the figure, respectively. The black dotted line indicates the median fluence ratio, which is 1.01. There is no evidence for a correlation in the fluences of consecutive bursts.

APPENDIX E: WAIT-TIMES OF INDIVIDUAL OBSERVATIONS

The wait-time distribution for the observation at MJD 59877 is plotted in Fig. E1. During a period of 0.96 h, a total of 72 FRBs were detected, indicating a burst rate of 75^{+10}_{-9} bursts per hour, where the errors are Poissonian. The best-fitting Weibull CDF indicates a burst rate of 76.7 ± 5.7 bursts per hour, which is consistent with the calculated burst rate. The shape parameter of the Weibull CDF is $1.08 \pm 0.10 \pm 0.04$, which is consistent with k = 1, indicating that the Poisson distribution hypothesis can not be rejected. Thus, for this observation, there is no significant evidence for clustering.

The wait-times for the observation on MJD 59879 are shown in Fig. E2. During a period of 1.37 h, a total of 82 FRBs were



Figure E1. The distribution of the wait-times from the observation on MJD 59877, with a Poisson burst rate of 75^{+10}_{-9} bursts per hour. The cumulative wait-time distribution is shown in the top plot and fitted with a Weibull CDF. The shape parameter $k = 1.08 \pm 0.10 \pm 0.04$ does not significantly deviate from k = 1, which indicates the Poisson distribution can not be rejected as a potential fit to the data. Thus, for this observation, there is no significant evidence for clustering.



Figure E2. The distribution of the wait-times from the observation on MJD 59879, with a Poisson burst rate of 60^{+7}_{-7} bursts per hour. The cumulative wait-time distribution is shown in the top plot and fitted with a Weibull CDF. The shape parameter $k = 1.16 \pm 0.11 \pm 0.05$ does not significantly deviate from k = 1, which indicates the Poisson distribution can not be rejected as an accurate fit to the data. Thus, for this observation, there is no significant evidence for clustering.

detected, giving rise to a Poisson burst rate of $60.06^{+7.35}_{-6.58}$ bursts per hour. The best-fitting Weibull CDF indicates a burst rate of 78.3 ± 5.2 bursts per hour, which is not consistent with the calculated burst rate. The shape parameter of the Weibull CDF is $1.16 \pm 0.11 \pm 0.05$, which is within errors consistent with k = 1,

again indicating that the Poisson distribution can not be rejected. Also for this observation, there is no significant evidence for clustering.

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