

Observations of Naturally Occurring Lightning with Event-Based Vision Sensors

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Abstract

Lightning is an impressive, widespread natural phenomenon, yet many questions about its physics remain unanswered due to its extreme speed, transience, and high energy. These intrinsic characteristics make optical observations capturing its formation, propagation, and discharge challenging with conventional optical cameras. Furthermore, optical sensors with extremely high-speed frame rates and high dynamic ranges are needed. While high speed cameras have been used to capture lightning, their lack of portability, high cost and high data storage requirements can limit lightning research. To address these challenges, the use of neuromorphic technologies, inspired by the sensing and data processing mechanisms of biological photoreceptors, offers a unique approach. Event-based vision sensors offer low latency, less power than a conventional camera, and have sensing capabilities that operate across a dynamic range of over 120dB. This paper demonstrates the effectiveness of Event-Based Vision Sensor in lightning research by presenting data collected during a full lightning storm and provides examples of how event-based data can be used to interpret various lightning features. We used a Prophesee Gen4 Event-Based Vision Sensor to record a thunderstorm over a fifty-minute span on 24 January 2023, from Western Sydney, New South Wales, Australia. During this observation, we recorded numerous Cloud-to-Ground and Cloud-to-Cloud lightning strikes. To assess the Event-Based Vision Sensor's effectiveness in capturing commonly observed lightning features, we used custom algorithms and in-house post-processing software was used to analyze and interpret the data. We conclude that the Event-Based Vision Sensor has the potential to improve high-speed imagery due to its lower cost, data output, and ease of deployment, ultimately establishing it as an excellent complementary tool for lightning observation.

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7 **Key Points:**

- 8 • Event-based vision sensors record changes in contrast in a scene at high speed and
9 low data rate, making them ideal for recording lightning
10 • Presented here are observations of lightning with event-based vision sensors, plus
11 data analysis techniques
12 • Several known features of lightning were clearly observed in the data presented
13 here

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Abstract

Lightning is an impressive, widespread natural phenomenon, yet many questions about its physics remain unanswered due to its extreme speed, transience, and high energy. These intrinsic characteristics make optical observations capturing its formation, propagation, and discharge challenging with conventional optical cameras. Furthermore, optical sensors with extremely high-speed frame rates and high dynamic ranges are needed. While high speed cameras have been used to capture lightning, their lack of portability, high cost and high data storage requirements can limit lightning research. To address these challenges, the use of neuromorphic technologies, inspired by the sensing and data processing mechanisms of biological photoreceptors, offers a unique approach. Event-based vision sensors offer low latency, less power than a conventional camera, and have sensing capabilities that operate across a dynamic range of over 120dB. This paper demonstrates the effectiveness of Event-Based Vision Sensor in lightning research by presenting data collected during a full lightning storm and provides examples of how event-based data can be used to interpret various lightning features. We used a Prophesee Gen4 Event-Based Vision Sensor to record a thunderstorm over a fifty-minute span on 24 January 2023, from Western Sydney, New South Wales, Australia. During this observation, we recorded numerous Cloud-to-Ground and Cloud-to-Cloud lightning strikes. To assess the Event-Based Vision Sensor’s effectiveness in capturing commonly observed lightning features, we used custom algorithms and in-house post-processing software was used to analyze and interpret the data. We conclude that the Event-Based Vision Sensor has the potential to improve high-speed imagery due to its lower cost, data output, and ease of deployment, ultimately establishing it as an excellent complementary tool for lightning observation.

Plain Language Summary

Event-based vision sensors (EBVS) are a unique type of optical sensor that complements conventional cameras to observe the features of lightning. Instead of capturing an image of lightning frame by frame, the EBVS detects changes in contrast within a scene. They offer high-speed observations but with improved contrast, which is especially useful for bright phenomena like lightning. The EBVS’s ability to only record changes in brightness significantly reduces the amount of data generated compared to typical high-speed cameras. As a result, they can capture an entire storm event. Here we show that EBVS has the potential to capture lightning images with features that were previously difficult to capture using regular high-speed cameras.

1 Introduction

There are many unanswered questions regarding the physics of lightning, such as how lightning is initiated within thunderclouds, the physical mechanisms governing the propagation of diverse types of lightning leaders, and what determines the direction in which a lightning leader travels (Dwyer & Uman, 2014). To extend our understanding of the physical processes of lightning, novel approaches to uncovering previously unresolved lightning features are required to address these outstanding questions. Existing observations use a variety of methods across different spectral ranges, including radio, visual, and x-ray spectral ranges (Biagi et al., 2010; Jiang et al., 2021; Wang et al., 2022; Rakov et al., 2022; Zeng et al., 2016; Walker & Christian, 2017, 2019; B. M. Hare et al., 2020). High-speed cameras are often used to capture features of lightning due to their rapid frame rate (Biagi et al., 2009, 2010; Saba et al., 2006; Hill et al., 2011; Campos et al., 2014; Qi et al., 2019; Wu et al., 2022; Petersen & Beasley, 2013; Qi et al., 2016). However, they have some drawbacks, including the high cost of the cameras themselves and the generation of large data outputs which make continuous recording over prolonged periods difficult (Campos et al., 2014; Saraiva et al., 2014; Qi et al., 2019). We demon-

64 strate the use of an alternative sensor with high temporal resolution and wide dynamic
 65 range to observe lightning, and its effectiveness in capturing features of lightning.

66 Event-based vision sensors (EBVS) are CMOS silicon-based imaging sensors, whose
 67 pixels contain special analog circuitry that allows them to operate asynchronously and
 68 independently from one another. Each pixel operates as a temporal contrast sensor, emit-
 69 ting data only in responses to changes in contrast, which usually results from changes
 70 in brightness in a scene (Gallego et al., 2020). Unlike conventional cameras, these sen-
 71 sors do not emit frames, nor do they make use of a constant sampling rate. As a result,
 72 these sensors have no exposure times or frame rates, instead each pixel outputs an “event”
 73 whenever the log intensity of the brightness change at the pixel exceeds certain thresh-
 74 olds (Gallego et al., 2020). These thresholds, or bias settings, are user-configurable chip-
 75 wide parameters that control the sensitivity of the sensor to both positive and negative
 76 changes in contrast. These values can be asymmetrical, allowing for greater sensitivity
 77 for contrast changes in one direction over the other.

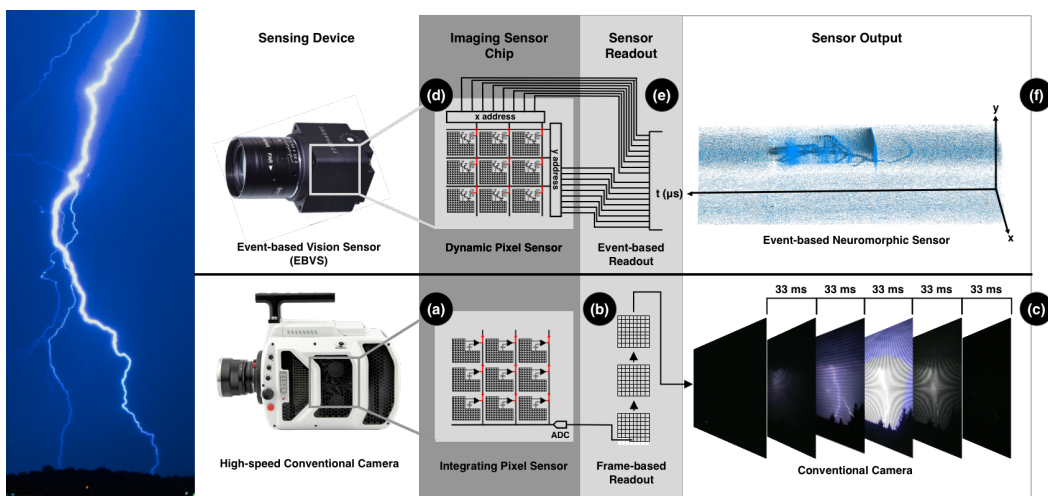


Figure 1: Diagram illustrating the differences between a conventional imaging sensor and an event-based vision sensor. The above figure shows the information flow through a representation of a conventional cameras (bottom row) and an neuromorphic event-based camera (top row). In a conventional camera, such as the high-speed camera, a large-format CMOS or CCD sensor is used to synchronously capture a visual representation of the scene. The pixels inside these cameras integrate the photocurrent and allow it to read out sequentially, as shown in (a). The output of these cameras is digitized using an ADC and assembled into conventional image frames, as shown in (b). These frame are generated at the frame rate at which the camera is sampled, resulting in discrete visual representations of the scene separated uniformly in time, as shown in (c). By contrast, the EBVS contains a sensor with specialized analog pixels that implement a temporal contrast detector, as shown in (d). These pixels do not output frames, but rather emit a stream of contrast change events only when they occur, each with microsecond timing. This is illustrated in (e), showing the continuous output of the sensor, rather than the assembly of discrete frames. The output of the EBVS, shown in (f) is fundamentally different from the frames generated by conventional camera. The output is a sparse spatiotemporal signal, providing many of the benefits of the high-speed camera by breaking the relationship between frame rate and the volume of data produced.

78 The pixels in an EBVS are fabricated using conventional silicon technologies, mak-
 79 ing their spectral response similar to conventional CMOS and CCD cameras. Unlike con-

80 conventional sensors, the photocurrent from the photoreceptor is not integrated but rather
 81 clamped using a set of transistors to produce a signal proportional to the log of the pho-
 82 tocurent. This signal is then passed through an temporal differencing detector which
 83 compares it to the signal last used to generate an event. This difference is finally com-
 84 pared, via two comparators, to the positive and negative threshold biases, which will sig-
 85 nal that an change event has occurred.

86 Instead of a conventional frame readout, the EBVS has an arbiter that is respon-
 87 sible for checking and retrieving events generated by each pixel. Once an event is detected
 88 at a pixel, the pixel itself is reset (causing the temporal differencing detector to change
 89 the stored threshold). The location and polarity of the change is then combined with an
 90 absolute timestamp to indicate the moment at which the event occurred. These four val-
 91 ues $[x, y, t, p]$ comprise an event from the sensor, in which x and y represent the spatial
 92 location of the pixel within the sensor array, p is a binary value indicating whether the
 93 event was a positive change in contrast (ON event) or a negative change in contrast (OFF
 94 event), and t is the timestamp at which the event occurred (usually provided in μ). This
 95 operation is described in detail in (Gallego et al., 2020).

96 In addition to the bias settings that control the sensitivity to contrast changes, the
 97 sensor also provides other bias settings that control the responsiveness, speed, bandwidth,
 98 and gain of the analog circuit within each pixel, allowing for the sensor to be tuned for
 99 specific tasks.

100 EVBS sensors produce data only in response to changes, necessitating the need for
 101 data-driven systems. In place of a constant frame rate, the pixels only produce data in
 102 response to changes. Therefore, it is possible to achieve the equivalent of a high-speed
 103 camera performance without the need to sample each pixel at extremely high speeds. There-
 104 fore, if an EBVS is stationary, it ideally produces no data output where there are no de-
 105 tectable changes in brightness in a scene, suppressing redundant information such as static
 106 and unchanging background features. As a result, the data rate becomes activity-driven
 107 and can capture high-speed phenomena with high temporal resolution without the need
 108 for external triggering.

109 The EBVS offers several advantages over conventional sensors. Data rate and down-
 110 stream computational load can be significantly reduced with the EBVS's highly efficient
 111 frame-free operation. Processing of event data is different to frame-based image process-
 112 ing as there are no regular sampling intervals, color channels, or light intensity informa-
 113 tion. Therefore, event data also requires specialized software to visualize the data for in-
 114 terpretation.

115 These features make the EBVS an excellent candidate for use in the study of light-
 116 ning. The purpose of this paper therefore, is:

- 117 1. to demonstrate the capabilities of EBVSs in capturing the features of lightning,
 118 and
- 119 2. to demonstrate methods of visualizing event data that may be useful in interpret-
 120 ing features of lightning.

121 We demonstrate the visualization of event data through event-rate plots, static plots
 122 with just ‘on’ and ‘off’ events, and plots with temporally colour coded information. Fur-
 123 ther, to show what can be achieved through analysis of event-based data for lightning,
 124 we use a published, event-based tracking algorithm to perform analysis on the data and
 125 to extract and model the relative speeds of lightning leaders. We also provide an overview
 126 of improvements to an EBVS observing system that may enhance future observations
 127 and results.

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1.1 Features of Lightning and Nomenclature

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The nomenclature of lightning features varies across the literature. Accordingly, in this paper, we will focus primarily on the most common and recent definitions previously established to describe the features we demonstrate.

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In this paper, lightning will be categorized as either cloud-to-ground (CG) or cloud-to-cloud (CC) strikes (Dwyer & Uman, 2014). During CG strikes, a downward lightning leader propagates from the cloud to the ground (Dwyer & Uman, 2014), following the electric field gradient through the conductive air (Dwyer & Uman, 2014; Ding & Rakov, 2022; Jiang et al., 2015). As the leader approaches the ground or nearby objects, it induces an upward streamer or leader of opposite polarity (Dwyer & Uman, 2014). While our focus is not on these upward streamers or leaders, it is speculated that the characteristics of the downward leader's propagation significantly influence upward leaders, and in turn, the attachment processes involved (Jiang et al., 2015). This necessitates further investigation of leader propagation.

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Negative downward leaders usually have a distinctive stepped pattern, leading to their classification as 'negative stepped leaders', which has been observed in both natural and triggered lightning strikes (B. M. Hare et al., 2020; Biagi et al., 2014; Wang et al., 2019). Many studies rely on triggered lightning techniques to understand the behavior of natural lightning as in these cases the location and time of the lightning leaders can be easily determined (Cai et al., 2022; Wang et al., 2022, 2019; Biagi et al., 2014; Zhang et al., 2014; Gamerota et al., 2014). As the downward negative stepped leaders propagate, they typically branch out across the sky. CC lightning, whilst not as well-documented, is also seen to have negative stepped leaders that follow the same behaviors. As the negative leader branches extend in a step-like manner, each step generates a travelling wave of moving positive charge from the branch tip along the channel (Ding & Rakov, 2022). While negative stepped leaders are extensively documented, the exact nature and mechanism of the phenomenon have yet to be fully understood (Lowke & Szili, 2022; Wang et al., 2022; Ding & Rakov, 2022). Moreover, a negative stepped leader with pronounced branching creates a complex and continuously changing electric field pattern, leading to the formation of complex branching structures (Ding & Rakov, 2022). Branches at higher altitudes may decay faster than those formed later, as they may lose connection with the main channel (Ding & Rakov, 2022).

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After the attachment process, the lightning leader forms a highly conductive hot plasma channel through which the return stroke traces (Liang et al., 2014). In CG lightning, the return stroke is an extremely rapid and bright current wave typically described as travelling up the main plasma channel, without branching as it propagates, even if the downward leader had extensive branching (Liang et al., 2014). The speed of the return stroke cannot be directly measured, although its current wave has been estimated to be between 1/3 and 2/3 of the speed of light (Liang et al., 2014; Zhou et al., 2019). Additionally, the optical wave has been found to exhibit varying speeds throughout its propagation, making capturing the return stroke challenging (Liang et al., 2014; Zhou et al., 2019).

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Following the initial return stroke that neutralizes charges within the channel, secondary leaders known as 'dart leaders' have been observed travelling through the dormant remnant path, initiating a secondary return stroke (Thiemann & Gasiewski, 2014; Gamerota et al., 2014; Cai et al., 2022; Wang et al., 2022). Despite extensive research, the mechanisms by which the dormant plasma channel induces secondary leaders remain a mystery (B. M. Hare et al., 2023). Nevertheless, it is generally accepted that remnants of the hot plasma channel provide a conductive path for the dart leader to follow (Thiemann & Gasiewski, 2014; Gamerota et al., 2014; Cai et al., 2022). A successful dart leader may then induce a secondary return stroke, leading to the occurrence of multiple return strokes

(Gamerota et al., 2014), a process which may continue until no new dart leader is initiated (Dwyer & Uman, 2014; Yuan et al., 2020; Shi et al., 2019; Filik et al., 2021).

Multiple return strokes have been observed in both positive and negative lightning discharges, occurring in both natural and triggered lightning events (Yuan et al., 2020; Shi et al., 2019; Filik et al., 2021). We will refer to an unsuccessful lightning strike as a leader that does not result in a return stroke. If the dart leader remains in the cloud and is unsuccessful in initiating a return stroke, a different terminology is typically applied (Jensen et al., 2020). There is variation in terminology surrounding dart leaders that fail to reach the ground; some papers refer to them as ‘K leaders’, while more recent publications use the term ‘recoil leaders’ (Jensen et al., 2020; Jiang et al., 2021; Huang et al., 2021; Cruz et al., 2022; Liu et al., 2021; B. M. Hare et al., 2023). Therefore, in this context, we will use the term ‘recoil leaders’ to encompass dart leaders that are unsuccessful, while acknowledging that their phenomenon has been observed to align with dart leaders (Jensen et al., 2020).

2 Materials and Methods

2.1 Equipment and Data Collection

On 24th of January 2023 at 19:27 AEST (08:27 UTC), a thunderstorm was observed from a Western Sydney University campus in Werrington, New South Wales, Australia. A Prophesee GEN4 EBVS was used, which has 1280 x 720 pixels, with a pixel pitch of 5 microns. Further information on the specifications of this camera can be obtained from Prophesee on request at their discretion. A commercial off-the-shelf Fuji 12mm telephoto lens was used for data collection. The EBVS was controlled with a laptop via USB and software developed by ICNS. The EBVS was placed next to a window pointed westerly towards the incoming storm and left to record until it passed over the top of the building, approximately 50 minutes later. A total of 22.95GB of data was collected.

2.2 Data Analysis

2.2.1 Event-Rate Plots and Rainbow Plots

Various software was written in the programming language Python3 to assist in the visualization of the lightning captured in our observations. The software takes the output of the EBVS and plots it on an x,y plane over time. Typically, there is one color for “on” events and one color for “off” events, which in this paper are yellow and blue, respectively.

Presented in this work are ‘event-rate plots’, which show the number of events per second over time. While searching through data for lightning, one can easily detect a significant increase in events per second, which, in this case, is an indication that lightning occurred. Further, one may ascertain some arbitrary or relative measure of intensity of features such as return strokes.

It is possible to visualize event data in a manner similar to a conventional image frame by binning all events that occur over a fixed time interval and accumulating them from each pixel. This forms a 2D histogram that counts the number of events per pixel and produces an output that resembles a conventional 2D image frame. Colored plots were produced, which visualize the strike as it would appear in a stack of frame-based images, but show events as colors. The program uses color gradients to discern between the first and last events that appear in the scene. The in-house software developed to create figures of lightning uses both on and off events to produce these colored images, which we will henceforth refer to as “rainbow plots”. We present event-rate plots and rainbow plots.

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2.2.2 Tracking Lightning Leaders

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Given the data presented in this paper was only captured with a single EBVS, absolute speeds of leaders cannot be determined by tracking. However, we can measure the 2D projected speed on the image plane in pixels per unit time and infer relative speeds of leaders in relation to each other. Here we demonstrate a simple tracking algorithm that automatically analyses the motion of lightning phenomena to estimate their relative speeds and the motion of surrounding phenomena. The spatio-temporal nature of event-based data is very well suited to performing high-speed tracking. Specifically, the high temporal resolution of events results in highly frequent position updates for trackable targets. For the purposes of this paper, we define tracking as the challenging process of estimating the state of an unknown and time varying number of targets among noise and clutter based on measurements from a sensing system.

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In this paper, we approached the problem of tracking lightning leaders using the first-principles Global Nearest Neighbour (GNN) tracker (Blackman & Popoli, 1999) as a simple and well-established tracking algorithm. We also use a ‘tracking-by-detection’ framework to sequentially detect then track lightning leaders to estimate their motion. Detection is performed using DBSCAN clustering (Ester et al., 1996), where trackable lightning leaders are assumed to be correlated with spatio-temporal consistent and coherent features in the event-stream. This detection via clustering is performed synchronously on a ‘center-surround spiking filtered’ window of events in 50 μ s intervals. Figure 2 provides an overview of the processing pipeline of the tracker. A detailed description of how the tracking system works is given in the Supporting Information.

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A track measurement is any observational information of a target, which can range from raw sensor outputs of a given target characteristic such as position. In tracking-by-detection, the output of a detector such as a clustering or filtering algorithm is also considered a track measurement. To estimate the state of a target, track measurements are associated to known or potential targets, which are used to infer desired target state qualities including position, speed and acceleration using Bayesian techniques such as Kalman Filtering (Kalman, 1960), Particle Filtering (Del Moral, 1997) or a general Markov-Bayes recursion (Ralph et al., 2022). Tracking algorithms generally use a track manager to maintain unique track identifiers and the life cycle of tracks. The track life cycle begins when a new measurement is received and an unconfirmed track is initialized, progressing to track confirmation when the track has high confidence (i.e. it has produced many position measurements) and track pruning when the target has disappeared or left the field of interest. The core distinguishing factor of most tracking algorithms is their approach to the data association, where measurements are assigned to the tracks which they most likely represent.

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3 Results

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The EBVS recorded a lightning storm continuously for approximately 50 minutes, resulting in 12.3GB of uncompressed data being recorded. 33 CG and 16 CC strikes were clearly visible in the data. Other “flashes” were observed, but any structure within those flashes was obscured by cloud, or the strike occurred out of frame. Leader propagation was shown in all observations of CG and CC strikes captured throughout the observing campaign. Distinct leader propagation and return stroke initiation was present in several strikes. Dart leaders were also observed, some of which initiated multiple return strokes within a lightning flash. Given the number of strikes captured, we have chosen some outstanding examples of both CG and CC strikes with distinctive features to present here, including an example of a negative stepped branched leader with a single return stroke, multiple return stroke and no return stroke. Event-rate plots and rainbow plots were produced for these examples, followed by a demonstration of the use of the tracking algorithm for one strike.

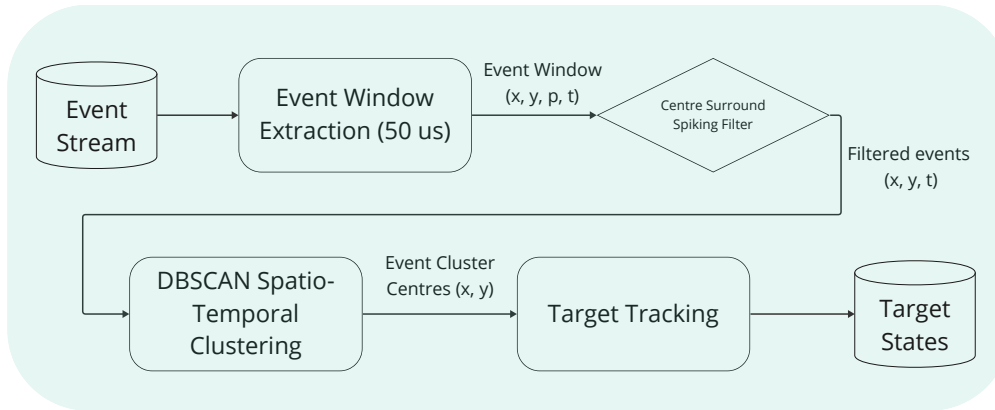


Figure 2: Abstracted overview of the proposed lightning feature tracking algorithm pipeline from raw event-stream data to estimated target position and velocity states.

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3.1 Event-Rate Plots

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To demonstrate event-rate plots, 3 different sequences of lightning flashes captured by the EBVS are provided. During the thunderstorm, at 19:46:31, the EBVS recorded a negative stepped leader propagating and branching across the sky in a CG flash (Figure 3). Later, at 19:49:47, a CG lightning flash, that was initiated by a negative stepped leader and had multiple return strokes, was recorded (Figure 4). The EBVS then captured, at 19:51:59, a small sequence of discharge events that reionize channels from an unsuccessful branched negative stepped leader (Figure 5). Rainbow plots are provided for two of the flashes as visual aids to complement the event-rate plots.

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At around 19:46:31, a CG negative branched stepped leader was captured, figure 3. This leader initiated a single return stroke with no apparent reionization of the plasma channel. Corresponding to the single return stroke, the event-rate plot of this flash, figure 3a, has one peak. This peak has a magnitude of around 10^9 events, where the initial background noise of the camera averages at 10^6 events. Figure 3a shows a gradual increase in events corresponding to leader branching, followed by a rapid spike during the return stroke, and finally, a gradual decrease. The propagation of the leader and its branches before the return stroke is visually represented with a rainbow plot in Figure 3b. Although event rates do increase with leader propagation and branching, there remains a clear distinction between the leader and its branches, and the return stroke, as evidenced by the magnitude change in event rates.

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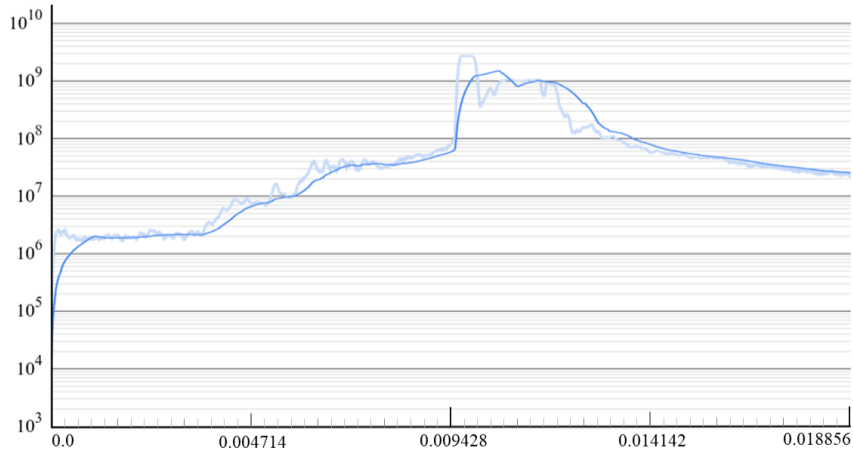
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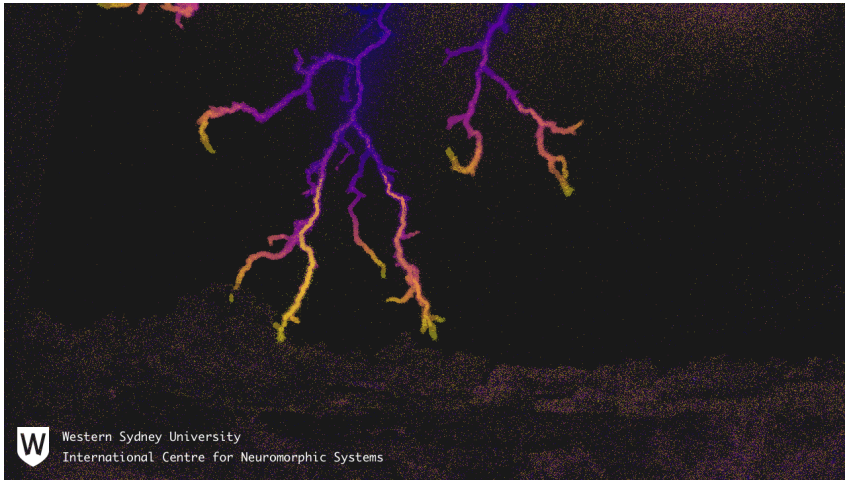
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A CG negative stepped lightning leader with multiple subsequent return strokes was captured at 19:49:47. The corresponding video of the lightning flash sequence, whilst not shown here, depicted several return strokes and dart leader through the same channel over 0.357 seconds. While it is difficult to visualize this without the video, it can be shown in the event-rate plot, 4. The main channel is momentarily dormant after the initial return stroke but becomes luminous again multiple times with varying event magnitude. This lightning flash closely reaches its peak magnitude of 10^9 events 3 times. Remaining peaks have magnitudes between 10^8 and 10^9 events per second. The series of return strokes of flash in figure 4 is temporally separated by two main sequences. The first sequence, following the initial negative stepped leader, has clear peaks that can be correlated to return strokes. A second sequence of peaks occurs after 0.285600 seconds when charge re-enters the main channel as an additional lightning flash is initiated outside the EBVS's frame. Furthermore, the less distinct peaks of the second sequence are attributed to an increase of noise from the secondary strike outside the camera's frame. Between



(a) Event over time (seconds)



(b) Rainbow plot with temporal range of 0.0 to 0.009412 seconds; right before the return stroke initiation.

Figure 3: Event-rate plot and rainbow plot of a lightning flash with a single return stroke

311 the two return stroke sequences, there are minor peaks at around 0.178500 seconds af-
 312 ter the flash initiation and when the first RS sequence died off. From their magnitude,
 313 and verification with visual techniques, we can see that these peaks relate to unsuccess-
 314 ful dart leaders in the main channel. Upon examining the total event-rate plot of the light-
 315 ning flash, Figure 4, we observe up to 10 peaks that correspond to RS. However, many
 316 of these return strokes were not as distinct when using our visual processing tools. More-
 317 over, whilst event-rate plots allow further investigation of lightning features, our visu-
 318 alization techniques require improvement to differentiate between dart leaders and re-
 319 turn strokes visually for particularly bright strikes that are clustered together.

320 Captured at around 19:51, lightning flash in figure 5 is a negative branched stepped
 321 leader that does not initiate a return stroke. The unsuccessful leader does not progress
 322 far from the cloud until its propagation halts. Following the initial propagation is a se-
 323 ries of luminous events traveling up and down the initial main propagation path. The
 324 reionization of the plasma channel indicates recoil activity, corresponding to local small
 325 peaks within the event-rate plot of magnitude below 10^7 . The event-rate magnitude from
 326 the background noise prior to the leader was between 10^5 and 10^6 . In both the event-

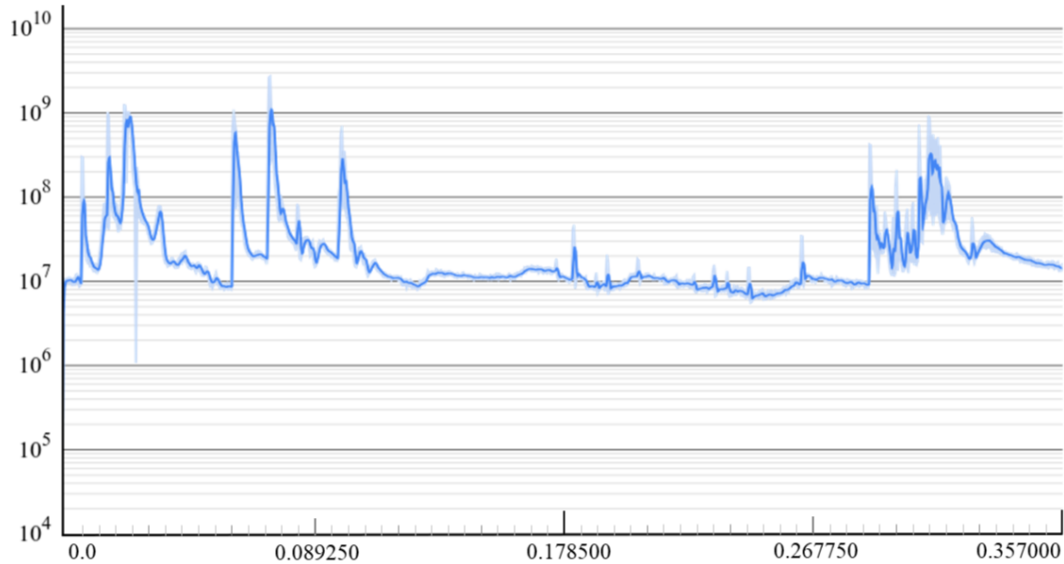


Figure 4: Events over time (seconds) of a lightning flash with several return strokes and dart leaders

327 based video and the event-rate plot of this lightning flash, there is no evidence of a re-
 328 turn stroke.

329 **3.2 Time-Dimension Plots**

330 Figure 6 features a CC lightning strike. Using a rainbow plot, we can select fea-
 331 tures of a strike, visualizing its start to end. Commencing from the top right of the plot
 332 (Figure 6b), an initial negative stepped leader propagates across the field of view, branch-
 333 ing as it progresses (Figure 6c). The negative leader is followed by a return stroke through
 334 the main channel, Figure 6d, and a second return stroke is initiated at a different main
 335 channel after the first, Figure 6e. Shortly after these events, a brief, intense, and irreg-
 336 ular bright flash occurs within a previously inactive branch channel (Figure 6f).

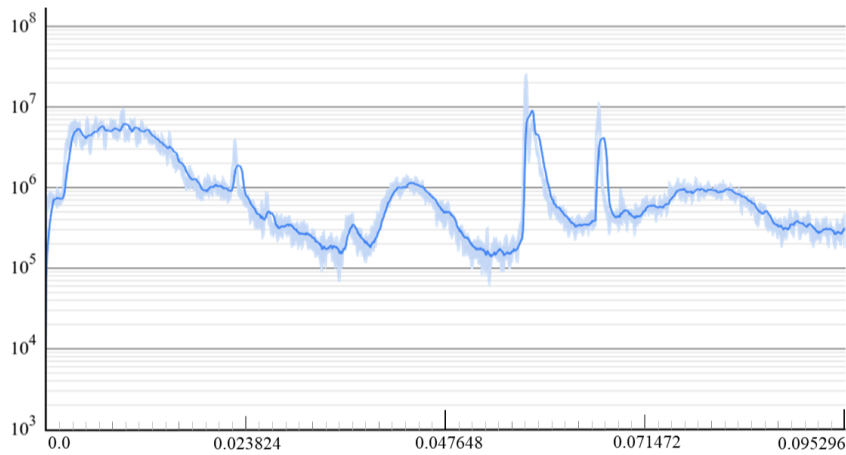
337 **3.3 Tracking Results**

338 Figure 7 demonstrates the tracking of lightning leaders in the same strike as shown
 339 in Figure 6a. DBSCAN is shown here to successfully detect lightning leaders as promi-
 340 nent event clusters in the presence of background noise. The weighted centroid of these
 341 clusters was used as position measurements for the GNN tracker. Using this tracking-
 342 by-detection approach while observing from a single fixed perspective produces seem-
 343 ingly relative position and speed estimates for lightning leaders in a 2D plane. The tracker
 344 assumes a point target model which asserts that targets have no spatial extent, which
 345 causes spatially extended phenomena such as return strokes to be poorly tracked as no
 346 discernible single point target can be detected by DBSCAN for tracking.

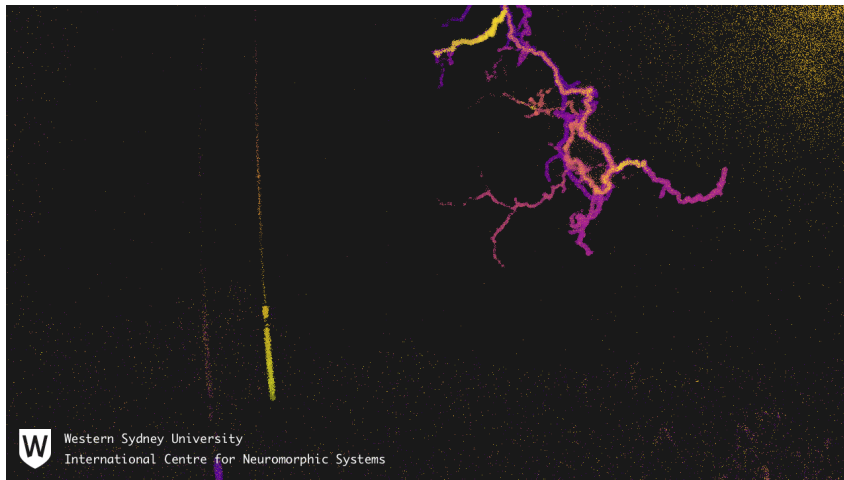
347 **4 Conclusions**

348 **4.1 Interpretation of Results**

349 The features captured by the EBVS as presented here largely align with previously
 350 published observations with conventional sensors, thereby validating the use of the EBVS



(a) Events over time (seconds)



(b) Rainbow plot with temporal range of 0.0 to 0.072310 seconds

Figure 5: Event-rate plot and rainbow plot of a lightning flash with no return stroke but several dart leaders. Note the yellow streak on the left of the image is rain.

351 as a complementary tool for lightning analysis and may provide improved observations
 352 on some features of lightning. The benefits of the EBVS, demonstrated in this paper,
 353 include the ability to record a storm from beginning to end, capturing orders of mag-
 354 nitude less data than a comparable high-speed camera, high temporal resolution, porta-
 355 bility, and alternative methods of presenting and analyzing the data. The high tempo-
 356 ral resolution combined with the sensor's high dynamic range provides opportunities to
 357 visualize features of lightning previously unresolved and provide complementary data to
 358 high-speed camera observations.

359 The data analysis techniques available to use on event-based data may also pro-
 360 vide new opportunities in capturing features of lightning. The colored rainbow plots demon-
 361 strate the ability to add a temporal aspect to static images by adding a color spectrum,
 362 an advantage for displaying a lightning strike where it is not possible to publish a video
 363 of the strike. Event-rate plots are useful in demonstrating the comparative relative in-
 364 tensities of leaders and are particularly useful in identifying return strokes or discharge
 365 activity, aiding in data processing, and capturing elusive information.

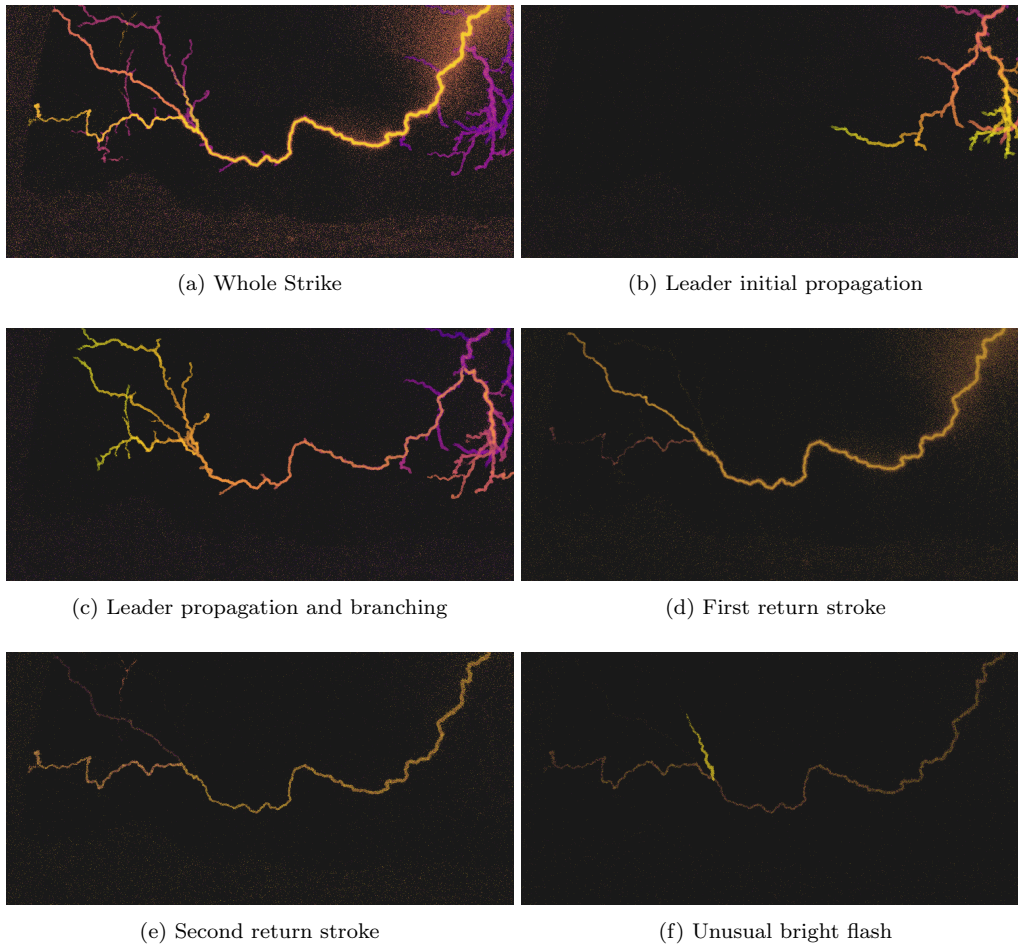
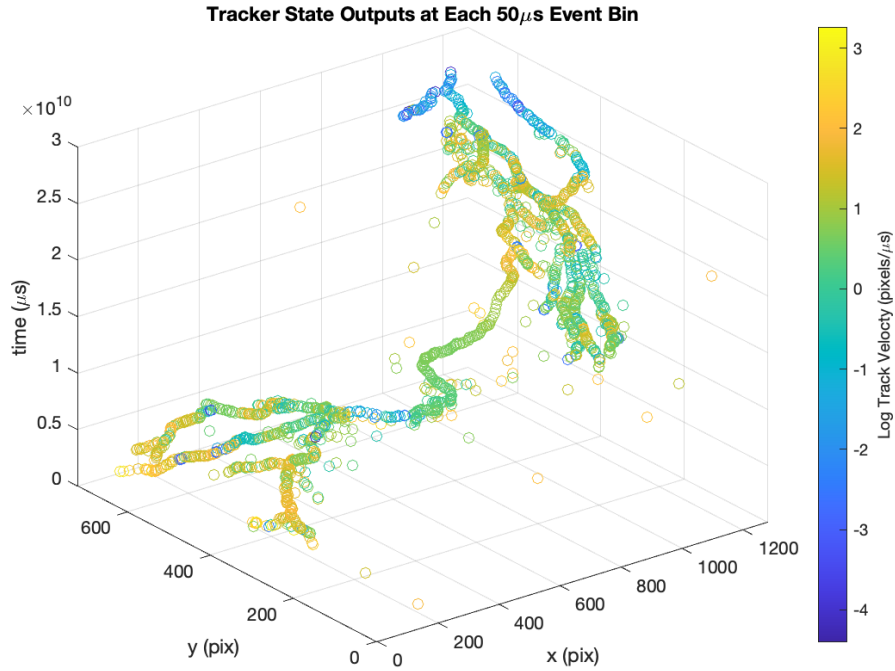


Figure 6: Rainbow plot of a lightning flash, consisting of a negative stepped leader with two main channels that the return strokes pass through and an unusual bright flash through a decayed branch channel after the return strokes.

366 The results of the tracking algorithm presented illustrate the potential for measur-
 367 ing the speed of lightning leaders. These results show that the high temporal resolution
 368 of the EBVS can be leveraged to produce similarly high temporal resolution state estimates
 369 of specific lightning phenomena. Multiple time synchronized EBVSs are required to track
 370 leaders and other phenomena in 3D via depth estimation.

371 Despite the comprehensive documentation of negative leaders, numerous charac-
 372 teristics remain poorly understood (Dwyer & Uman, 2014). Various theories have emerged
 373 regarding leader mechanisms and their subsequent features. In a recent paper, Wang et
 374 al. (2022) defined six types of lightning leaders that can initiate a return stroke (RS) or
 375 a continuing current (CC) process in negative cloud-to-ground (CG) lightning flashes.
 376 Our focus, aligning with most captured lightning strikes, centers on downward negative
 377 stepped leaders and downward negative dart leaders. The EBVS provided a clear visu-
 378 alization of full leader propagation, capturing the entire sequence from the initial branch-
 379 ing within the cloud to just before the return stroke (for example, Figure 3b). As fur-
 380 ther branching occurs, it is understood that highly structured and rapidly changing elec-
 381 tric field patterns form, leading to complex interactions between the branches (Ding &
 382 Rakov, 2022). As the negative leader propagates downward, we observe a luminous tip
 383 with surges of luminosity along the extending branch. Whilst we speculate that this phe-



(a)

Figure 7: Trajectory estimates of all confirmed tracks produced by the proposed GNN Tracking algorithm for observation 2023-01-24T19-51-46 in the spatial x, y pixel dimension and time t , colored by log pixel velocity in pixels per μs .

384 nomenon aligns with Ding and Rakov (2022) description of a negative branched leader
 385 generating traveling waves, moving positive charge from the branch tip up along the chan-
 386 nel, it requires further investigation in future research.

387 Return strokes are the result of a conductive hot plasma channel being formed dur-
 388 ing the attachment process (Liang et al., 2014). This plasma channel is surrounded by
 389 a cold plasma charged corona sheath with low conductivity, whose charge is subsequently
 390 neutralized when a current wave travels upward from the ground (the negative return
 391 stroke) and emits intense optical radiation (Liang et al., 2014). The extremely bright re-
 392 turn stroke propagates upward along the main channel at velocity around one-third the
 393 speed of light (Liang et al., 2014; Zhou et al., 2019; Dwyer & Uman, 2014). As the EBVS
 394 makes use of independent and asynchronous analogue pixels, these sensors can leverage
 395 their high dynamic range (both within each pixel, and across the array of pixels) to cap-
 396 ture the intense intense optical radiation was clearly detected in our results. The event-
 397 rate plot shows clear spikes corresponding with the lightning leaders initiating return strokes.
 398 This proved useful, particularly since observing the complete return stroke for each light-
 399 ning strike in the sensor footage posed challenges due to the extremely rapid and lumi-
 400 nous nature of the return strokes. Additionally, the path of the extremely bright return
 401 stroke in some instances could be traced using the off-events that remained after, which
 402 can be visualised in post-processing of the data. Furthermore, in certain instances, the
 403 trajectory of exceptionally intense return stroke could be discerned by following the resid-
 404 ual off-events. In these off-events, there are instances where the outward shock wave that
 405 occurs immediately after the return stroke is clearly visible.

406 After the first return stroke, leaders may follow the main channel’s path, often called
 407 dart leaders. The mechanism by which dart leaders navigate through apparently dor-
 408 mant plasma channels is not well understood (Dwyer & Uman, 2014; B. M. Hare et al.,
 409 2023). It is understood that negative charge is propelled forward in the form of a corona
 410 streamer zone primarily guided by the existing warm channel, whilst simultaneously, pos-
 411 itive charge is introduced into the hot core behind the tip and travels backward toward
 412 the cloud charge source (Ding & Rakov, 2022). While discerning the dart leader’s travel
 413 direction at each instance captured by the EBVS is challenging, our recordings depict
 414 its travel both upward and downward along the main channel following the downward
 415 negative stepped lightning leader. However, in negative CG lightning flashes, most re-
 416 ported dart leaders propagate downward from the cloud to initiate subsequent return
 417 strokes in both downward and upward lightning flashes (Wang et al., 2022). Aligning
 418 with previous studies, the successive dart leaders observed by the EBVS were observed
 419 to have an increased velocity on average (Jensen et al., 2020). The dart leader, if suc-
 420 cessful, may trigger subsequent return strokes, leading to the occurrence of multiple re-
 421 turn strokes within a single lightning sequence. Previous studies have captured instances
 422 of multiple return strokes in various cases (Yuan et al., 2020; Shi et al., 2019; Filik et
 423 al., 2021). In our observations, we captured multiple natural negative return strokes that
 424 closely followed the dart leader. Throughout a sequence of multiple return strokes, there
 425 is a broad variation in intensity. However, the general trend indicates a decrease in in-
 426 tensity as successive return strokes traverse the same plasma channel. Despite some vari-
 427 ation, the successive return stroke never appears brighter than the initial return stroke
 428 initiated by the negative leader.

429 In certain instances, a dart leader may fail to successfully initiate a return stroke
 430 and instead remain within the cloud; here, they are identified as recoil leaders and these
 431 phenomena, like dart leaders, is very poorly understood (Jensen et al., 2020; Jiang et al.,
 432 2021; Huang et al., 2021; Cruz et al., 2022; Liu et al., 2021; B. M. Hare et al., 2023). In
 433 the data presented in this paper, the EBVS captured various instances of recoil leaders
 434 both inside and outside the cloud, where the leader is unable to reach the ground and
 435 generate a return stroke. Furthermore, we recorded negative lightning leaders that sim-
 436 ilarly that did not contact the ground or another cloud to facilitate the attachment pro-
 437 cess crucial for a return stroke, rendering them unsuccessful leaders. Despite lacking a
 438 return stroke, these unsuccessful leaders exhibited lightning activity like dart leaders and
 439 can be referred to as recoil leaders as they followed the path of the original leader and
 440 it’s branches. Typically, in such cases, the recoil leader would adhere to a main chan-
 441 nel, although activity could also be observed from nearby branches. Understanding of
 442 the underlying mechanisms of this process requires further work.

443 4.2 Limitations and Future Work

444 Obtaining and understanding quantitative information about lightning from the
 445 EBVS is presently challenging, as the use of EBVS in the context of studying lightning
 446 is novel, and the lack of publicly available research into the exact number of photons re-
 447 quired to activate a pixel. our current understanding of how the sensor precisely reacts
 448 to and interacts with individual photons is incomplete. Consequently, measuring phys-
 449 ical properties of lightning, such as luminance, is currently not possible with the current
 450 generation of sensors (Gallego et al., 2020). Nevertheless, qualitative information can still
 451 be obtained due to the sensor’s capabilities of capturing lightning features. The promis-
 452 ing results demonstrated in this paper highlight the capabilities of the current genera-
 453 tion of sensors. These have not been developed or optimized for lightning detection, and
 454 therefore the results demonstrated in this paper represent the baseline performance achiev-
 455 able with this technology.

456 Future observations will endeavour to use the EBVS in conjunction with other sen-
 457 sors, such as radio frequency detection will allow us to determine the distance to light-

458 ning, which in turn may assist in measuring the length and speed of leaders. Further-
 459 more, we are currently testing the EBVS ability to detect spectral lines. Some initial test-
 460 ing shows spectra can be detected when using grating filters. Further testing is required
 461 to determine how well the EBVS will see narrow band spectral lines, and if proven possi-
 462 ble, we intend to undertake observations attempting to detect spectral lines in light-
 463 ning using EBVS.

464 The ability to model lightning captured in the radio spectrum in multiple dimen-
 465 sions has recently been demonstrated by B. Hare et al. (2018) using low-frequency ra-
 466 dio observations. They demonstrate a similar result (albeit a different method) to the
 467 tracking lightning leaders through the sky as per the results presented in this paper. To
 468 expand the potential of our own tracking algorithms, observations of lightning with mul-
 469 tiple time-synchronized EBVSs would allow multi-dimensional tracking of leaders spa-
 470 tially, giving direction and speed of leaders, and increases potential to observe other fea-
 471 tures such as leader tortuosity and behavior between stepping leaders.

472 Future tracking using the EBVS with lightning data could also benefit from the
 473 use of more complex state estimation algorithms such as multi-hypothesis tracking (Blackman,
 474 2004) or probability hypothesis density filtering (Vo & Ma, 2006), which are more re-
 475 siliant to noise and measurement-to-target association ambiguity. Additionally, the point-
 476 target model assumption could be relaxed for more robust tracking of phenomena such
 477 as return strokes which have a large and non-uniform spatial extent.

478 There is also potential for the data presented in this work to be further analysed
 479 to investigate specific features. There are a multitude of features captured in the data
 480 used for this work that have the potential to be investigated further. Some of these fea-
 481 tures could be allocated to previously understood phenomena of lightning initiation, prop-
 482 agation and re-ignition. Examples of these include, but are not limited to, luminous events
 483 accumulating throughout the view of the EBVS before the initial propagation of the neg-
 484 ative leader, nature of negative stepped leader initiation and propagation, dart leader
 485 propagation and the succeeding return stroke and luminous events within the channel
 486 of a lightning discharge event after the discharge occurs.

487 The potential of the EBVS in furthering understanding of the physical processes
 488 of lightning is herein demonstrated. From the conclusions we discuss here, there is sig-
 489 nificant potential to further investigate a multitude of EBVS features in capturing light-
 490 ning, and in the analysis of the data itself. To fully exploit this technology, strong col-
 491 laborations between lightning researchers and neuromorphic researchers to ensure the
 492 maximum potential of the EBVS in the study of lightning is achieved.

493 Acronyms

494 **EBVS** Event Based Vision Sensor

495 **CG** Cloud-to-Ground lightning

496 **CC** Cloud-to-Cloud lightning

497 **DBSCAN** Density-Based Spatial Clustering of Applications with Noise

498 **GNN** Global Nearest Neighbour

499 **GNNT** Global Nearest Neighbour Tracker

500 Open Research Section

501 Raw data used in the analysis of this paper can be obtained in .csv form in the Zen-
 502 odo repository at <https://zenodo.org/doi/10.5281/zenodo.10656060> Registration to Zen-
 503 odo is required to download the data. The data is covered by the Creative Commons li-
 504 cence system.

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