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Single Photon Kilohertz Frame Rate Imaging of Neural Activity

Tian Tian, Yifang Yuan, Srinjoy Mitra, Istvan Gyongy,* and Matthew F. Nolan*

Establishing the biological basis of cognition and its disorders will require high precision spatiotemporal measurements of neural activity. Recently developed genetically encoded voltage indicators (GEVIs) report both spiking and subthreshold activity of identified neurons. However, maximally capitalizing on the potential of GEVIs will require imaging at millisecond time scales, which remains challenging with standard camera systems. Here, application of single photon avalanche diode (SPAD) sensors is reported to image neural activity at kilohertz frame rates. SPADs are electronic devices that when activated by a single photon cause an avalanche of electrons and a large electric current. An array of SPAD sensors is used to image individual neurons expressing the GEVI Voltron-JF525-HTL. It is shown that subthreshold and spiking activity can be resolved with shot noise limited signals at frame rates of up to 10 kHz. SPAD imaging is able to reveal millisecond scale synchronization of neural activity in an ex vivo seizure model. SPAD sensors may have widespread applications for investigation of millisecond timescale neural dynamics.

reveal subthreshold activity and resolve spike timing with millisecond resolution from identified neuronal populations.^[1-3] These are substantial advantages over fluorescently encoded calcium indicators.^[4,5] However, fluorescence imaging of neural activity reported by GEVIs is challenging due to their millisecond dynamics and limited photon budget.^[6] An ideal imaging system would provide efficient shot-noise limited quantification of GEVI fluorescence at frame-rates above the Nyquist limit for sampling action potentials, which typically have durations on the order of a millisecond.

To date, electron multiplying CCDs (emCCD) and scientific CMOS (sCMOS) cameras have typically been employed to image GEVIs (Table S1, Supporting Information).^[1-3,7] These sensors face constraints in achieving high frame rates while minimizing the impact of inherent noise sources. Higher frame-rates are a

particular challenge as they demand a faster operation of the read-out electronics, which can conceal true signals through increased electronic readout noise.^[8] State-of-the-art emCCD cameras have low effective read noise and can achieve single-photon sensitivity.^[8] However, the limited frame-rate of the emCCD cameras (≈ 100 frames per second (fps)) does not allow kilohertz frame-rate imaging and the amplification process also introduces noise that effectively lowers the signal-to-noise ratio and their quantum efficiency.^[9] State-of-the-art sCMOS suffer from inherent dark noise and read noise,^[8] which makes it difficult at high frame rates to separate the weak signal generated by the incidence of a limited number of photons from spurious noise events.

The single photon avalanche diode (SPAD) is a photodiode that is reverse biased above its breakdown voltage, so that a single photon incidence at its photosensitive region creates an electron-hole pair that triggers an avalanche of secondary carriers and a large electric current.^[10,11] A SPAD is thus capable of single photon detection with the rising edge of the voltage pulse encoding the time of arrival of the photon. Local (in-pixel) circuitry then actively lowers the SPAD bias voltage to below the breakdown voltage to stop the avalanche and subsequently recharge the bias voltage to its initial value above the breakdown voltage, restoring the SPAD's sensitivity for the next photon detection event. This cycle typically takes on the order of tens of nanoseconds, enabling SPADs to perform ultrahigh frame-rate time-resolved

1. Introduction

A long-standing goal of neuroscience is to resolve the activity of ensembles of identified neurons with high spatiotemporal precision. Genetically encoded voltage indicators (GEVIs) can

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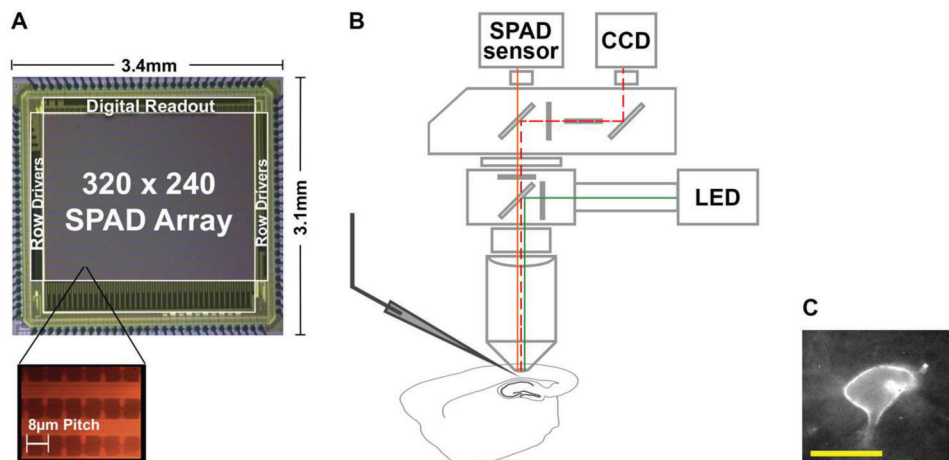


Figure 1. Photomicrograph of the SPAD image sensor and schematic representation of experiment setup. A) Photomicrograph of the SPAD image sensor and magnified image of the pixel array. B) Schematic of the experiment setup with colored lines indicating the simplified light path. The CCD camera was used for localization of neurons and positioning patch-clamp electrodes. Optical dynamics of the Voltron-JF525-HTL were recorded by the SPAD sensor. C) An example image captured by the SPAD array of a patch-clamped neuron expressing Voltron-JF525-HTL. The Voltron is mainly concentrated on the membrane of the soma. The distortion of the membrane caused by the pipette results in a local increase in fluorescence. Scale bar for neuron image: 10 μm .

imaging such as capturing light-in-flight.^[12] Because SPADs can detect the time at which a photon hits the sensor, their effective frame-rates are set by temporal binning and can be arbitrarily chosen after data acquisition according to the specific experimental demands so as to capture fast optical dynamics and maximize the available photon budget. SPAD-based image sensors have attained deep sub-electron read noise ($0.06 e^-$ to $0.17 e^-$, SPAD sensor used in this study).^[13] Therefore, at high frame rates the SPAD read noise is effectively negligible and images can be obtained at the shot-noise limit.^[13] As a result, in the low photon regimes encountered in high-speed imaging, the signal-to-noise ratio of a SPAD can exceed that of an sCMOS.^[14] Furthermore, in contrast to conventional emCCD and sCMOS cameras, temporal and/or spatial oversampling and binning can be carried out without any noise penalty.^[15] Although the application of SPADs to biological imaging has in the past been constrained by small array sizes, as well as large pixels with low fill factor, technological advances in these areas are making biological applications increasingly feasible.^[11]

Here, we evaluate the suitability of SPAD-based imaging for recording neuronal activity reported by GEVIs. We show that SPAD image sensors can resolve individual neuronal subthreshold and spiking activity reported with the GEVI Voltron-JF525-HTL with shot noise limited signals at frame rates of up to 10 kHz. They can also reveal spiking activities of individual neurons in neural ensembles during seizure-like events induced by 4-aminopyridine. SPAD sensors may have widespread applications for neural imaging at high frame rates.

2. Results

We utilized a 320×240 SPAD array image sensor (SPCImager, Table S1, Supporting Information) with $8 \mu\text{m}$ pixel pitch and 26.8% fill factor (FF) and a peak photon detection probability of 35% at 450 nm ^[13] (Figure 1A). When operating in binary mode each SPAD pixel produces a time-domain sequence of 0 s (no

photon detected) or 1 s (at least one photon detected) and the raw output of the SPAD at each exposure is the summed binary pixels in space, also known as a “bit plane”. A rolling shutter was used to enable back-to-back exposure at the maximum frame-rate (close to 10 kHz), so that the exposure time for each bit-plane is $\approx 100 \mu\text{s}$. The SPAD sensor was paired with an FPGA board (Opal Kelly XEM6310) that controls the acquisition of image data, relaying a continuous stream of bit-planes to a PC over a USB 3.0 link.^[15] Figure 1B shows a schematic of the experiment setup of the SPAD sensor incorporated with a microscope and electrophysiological recording set up. Figure 1C shows the temporally oversampled greyscale image of a neuron expressing the GEVI Voltron-JF525-HTL, obtained by binning 10 000 consecutive bit-planes over a 1-second long recording.^[13,16]

To test the suitability of the SPAD sensor for fluorescence voltage imaging, we used a viral approach to transduce cortical and hippocampal neurons with the GEVI Voltron-JF525-HTL^[1] and made whole-cell patch clamp recordings from GEVI-expressing neurons in ex vivo hippocampal and neocortical brain slices. We used the patch-clamp electrode to manipulate the membrane potential of the recorded neurons and simultaneously recorded the fluorescence changes of Voltron-JF525-HTL using the SPAD sensor. In sparsely labelled preparations,^[17] we observe fluorescence concentrated to the membrane of the soma of neurons expressing the soma-targeted Voltron-JF525-HTL (Figures 1C and A). For post-processing, the region of interest (ROI) was selected manually and a binary mask was applied to the ROI, allowing isolation of the fluorescence-positive pixels within the ROI and from which we obtained the time-dependent traces from each neuron (Figure 2). We took advantage of the negligible read noise and temporal oversampling of the SPAD sensor and adjusted the effective frame-rate by temporally binning consecutive bit-planes to optimize detection of subthreshold voltage changes or suprathreshold spikes (Figure 2; Figure S1, Supporting Information). Temporal binning was carried out by generating an average of summed photon counts within the ROI

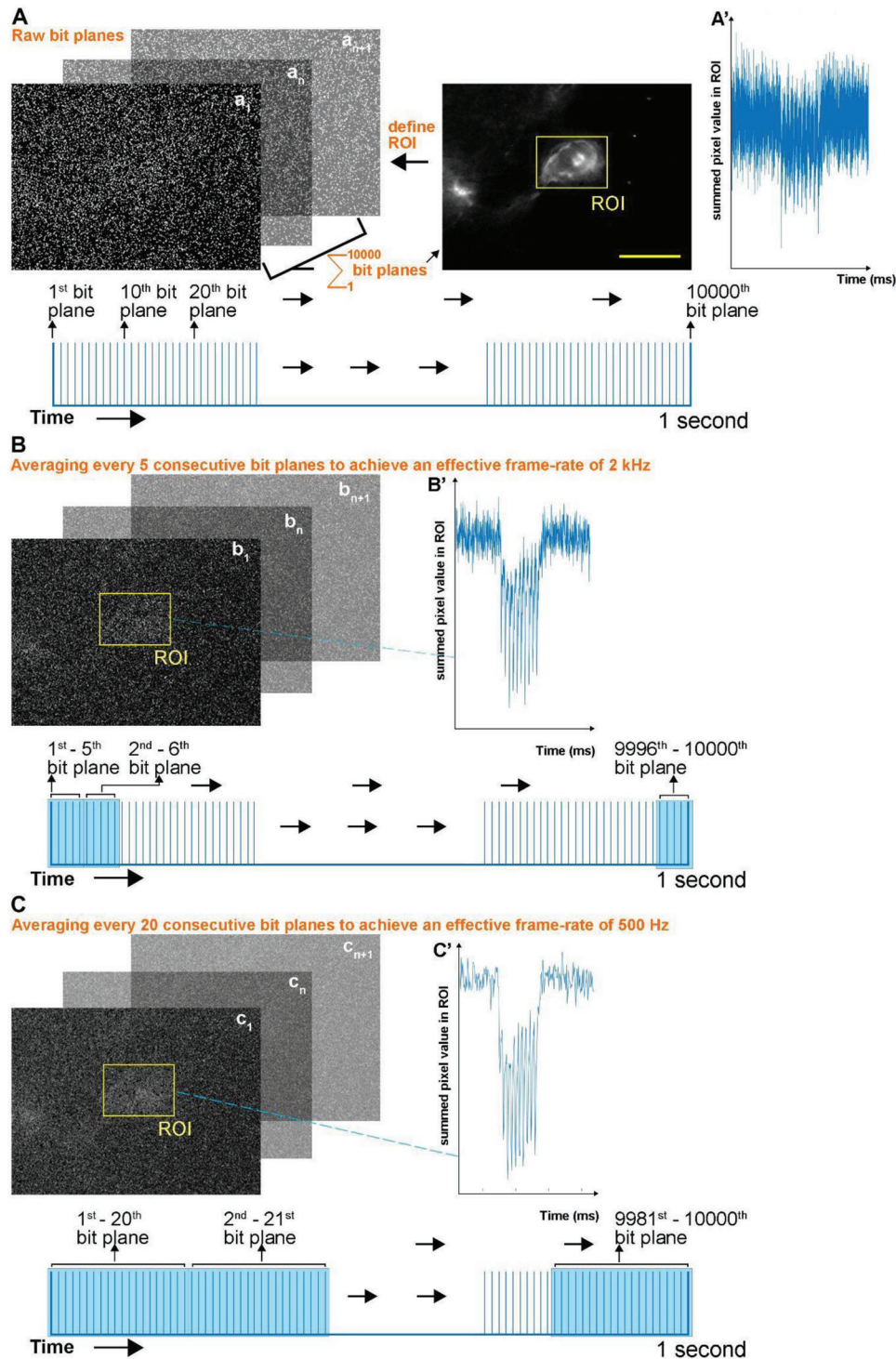


Figure 2. Strategy for temporal binning of bit planes. A) To define ROIs, 10 000 raw bit planes were collected from a 1-second-long recording and then summed to yield an image of the field of view. The summed photon counts within the ROI of each raw bit plane are plotted in (A') as a function of time during delivery of 10 spikes at 100 Hz. B) To perform temporal binning a set number of consecutive bit planes were averaged (in Example 5 consecutive bit planes were averaged, therefore achieving an effective frame rate of 2 kHz). Summed photon counts within the ROI of the averaged bit planes are plotted in B' as a function of time (corresponding to the unbinned data in A'). The images in the top left show the field of view after temporal binning of five consecutive bit planes was applied. C) As for (B) with temporal binning of 20 consecutive bit planes to achieve an effective frame rate of 500 Hz. Scale bar: 10 μm .

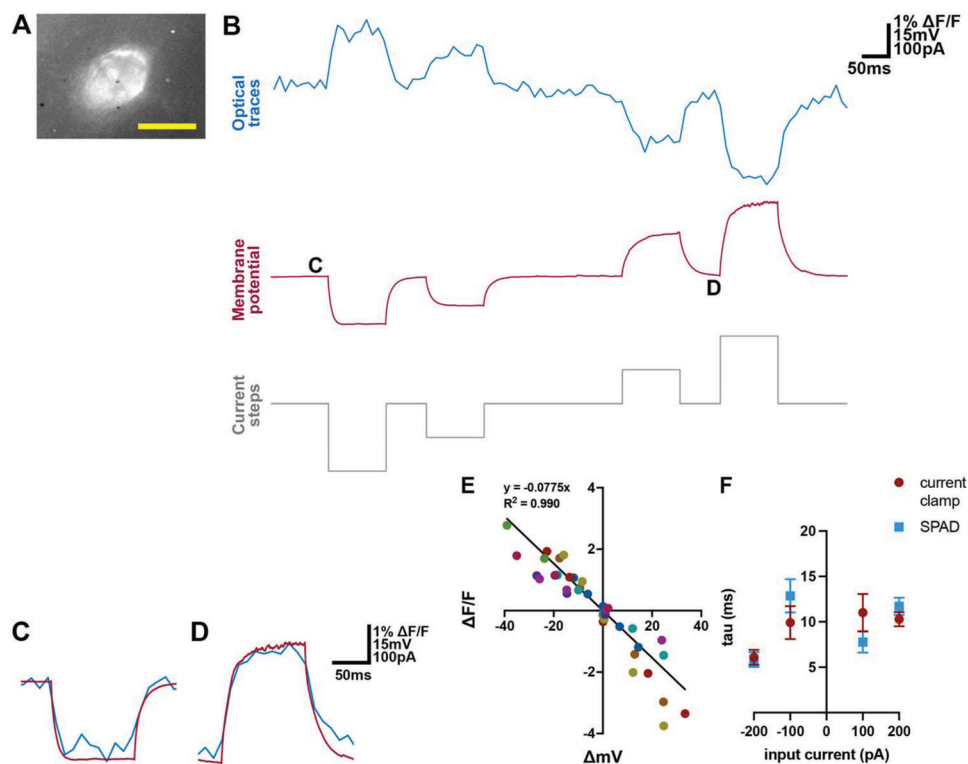


Figure 3. Detection of subthreshold membrane potential changes in the soma reported with Voltron-JF525-HTL. A) Image captured by the SPAD array of a patch-clamped neuron expressing Voltron-JF525-HTL, from which the traces in (B–D) was recorded. Scale bar for neuron image: 10 μm . B) Simultaneous optical (blue) and electrical (red) recordings of membrane potential changes in response to current steps (grey). The optical traces were captured by the SPAD sensor at a sampling rate of 9.9384 kHz, low pass filtered at 2 kHz, and temporally binned at 100 Hz. The electrical traces were sampled at 20 kHz. C,D) Segments from B on an expanded time base with the optical (blue) and electrical (red) traces superimposed. E) Change in Voltron signal ($\Delta F/F$) as a function of membrane potential change (ΔmV) in response to various current inputs ($n = 9$ cells). Data from each cell is marked with a different color. The line indicates the fit obtained with linear regression with intercept set at $x = 0$ and $y = 0$ ($p < 0.0001$). F) Time constants (mean \pm SEM) estimated from fitting responses to current steps plotted as a function of the current step amplitude. There was no significant difference between time constants estimated with electrical and optical methods ($p = 0.511$, $\chi(1)^2 = 0.4316$, likelihood ratio test, $n = 9$ cells.).

of consecutive bit-planes to reach an effective frame-rate suitable for the specific biological event (Figure 2B,C).

The SPAD image sensor captured slow changes (tens of milliseconds) in baseline membrane potentials reported with Voltron-JF525-HTL in acute brain slices, detecting subthreshold hyperpolarization and depolarization of the membrane potential in response to current and voltage stimuli (Figure 3; Figure S2, Supporting Information). Temporally binning more consecutive frames of the optical traces of subthreshold events significantly increased the SNR, but not the $\Delta F/F$ at each current or voltage step (Figure S1A–F, Supporting Information). Based on these considerations we chose an effective frame-rate of 100 Hz for analysis of subthreshold activity. We found that steady-state changes in baseline membrane potential (ΔV) were strongly correlated with changes in $\Delta F/F$ in both current clamp ($R^2 = 0.990 \pm 0.00200$, $p < 0.0001$, Figure 3E) and voltage clamp ($R^2 = 0.976 \pm 0.00355$, $p < 0.0001$, Figure S2C–E, Supporting Information) and the relationship between ΔV and $\Delta F/F$ appeared linear. The kinetics of membrane potential responses to current steps were also captured by the SPAD sensor (Figure 3C,D), with no significant difference in the rise and decay time con-

stant calculated from electrophysiological and optical recordings ($p = 0.511$, $\chi(1)^2 = 0.4316$, likelihood ratio test, $n = 9$ cells, Figure 3F).

We next tested if the SPAD sensor can capture individual action potentials. Although $\Delta F/F$ of spiking events decreased with temporal binning, the SNR increased significantly at lower effective frame rates (Figure S1G–L, Supporting Information). To capture the fast optical transient of Voltron during an action potential, we temporally binned ten consecutive SPAD bit-planes to reach an effective frame rate of 1 kHz. As shown in Figure 4A–D, the optical traces recorded by the SPAD sensor tracks single spikes in action potential trains evoked by 25 and 100 Hz current pulses (See Figure S3, Supporting Information). The relative fluorescence changes, quantified either by the SNR or the $\Delta F/F$ of single spikes, were comparable to the original report of Voltron-JF525-HTL (Figure 4E,F).^[1] Complex spikes evoked by a single pulse of current injection were also reliably reported by the SPAD signal (Figure S3B, Supporting Information). Occasionally, the expression of the soma-targeted Voltron-JF525-HTL can also be seen in subcellular structures close to the soma (Figure 4G). In these cases, the SPAD signal reported current-evoked

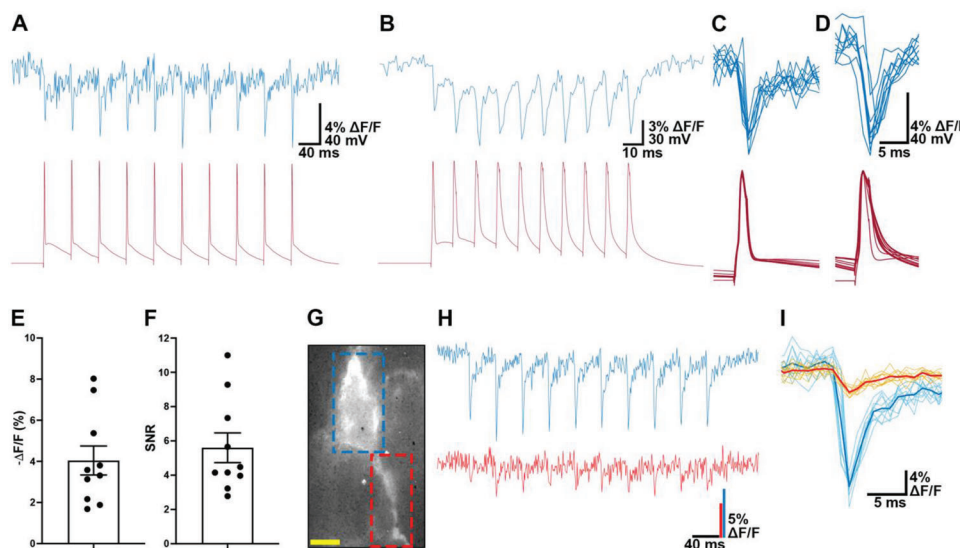


Figure 4. Detection of action potentials in the soma and subcellular structures reported with Voltron-JF525-HTL. Simultaneous optical (blue) and electrical (red) recordings of action potentials in response to ten current pulses at A) 25 Hz and B) 100 Hz. Pulse amplitude is 2 nA duration is 2 ms. The optical traces were originally captured by the SPAD image sensor at a sampling rate of 9.9384 kHz, then low pass filtered at 2 kHz and temporally binned at 2 kHz. The electrical traces were sampled at 20 kHz. C, D) Action potentials from (A, B) plotted on an expanded time base. E, F) Mean (\pm SEM) $\Delta F/F$ and SNR of peak fluorescence response in ex vivo brain slices. Each neuron fired 10 APs at 25 Hz, the output of each cell was the mean over these 10 spikes ($n = 10$ cells). SNR was calculated as $\Delta F / (\text{standard deviation of baseline fluorescence})$. G–I) Imaging fluorescence changes in a primary dendrite (red square in G) as well as the adjacent soma (ROI shown by the blue square in G) detects action potentials in both areas in responses to a train of current pulses (H). The amplitude of the action potential is smaller in the dendrite (I) (solid lines indicating average spike waveforms are overlaid on the individual responses). Scale bar for neuron image: 10 μm .

action potential trains in both the soma and in the subcellular structure (Figure 4H), with reduced peak amplitude in the subcellular structure (Figure 4I).

We next tested if the SPAD sensor can report activities of individual neurons in neural ensembles during epileptiform activity. We induced seizure-like events in hippocampal slices by application of 4-aminopyridine (4-AP). Interneurons in the striatum oriens layer of CA1 fired bursts of action potentials that were detected similarly with patch-clamp recordings and simultaneously with SPAD imaging of Voltron-JF525-HTL signals (Figure 5A, B; Figure S5, Supporting Information). The SPAD sensor reported action potentials and subthreshold events from multiple neurons during low frequency and bursting activity (Figure 5B–D; Figure S5A, B, D, E, Supporting Information). Peristimulus time histogram (PTSH) analysis of spiking activities of cell pairs showed evidence of millisecond scale synchronization (Figure 5E) and coordinated firing with lags of up to 100 ms (Figure 5F). Consistent with this, cell pairs showed strong subthreshold signal correlations reflecting synchronous depolarizations during epileptiform bursts of activity (Figure 5B, G).

3. Discussion

We demonstrate kilohertz frame-rate voltage imaging with subcellular resolution in ex vivo mouse brain slices using a SPAD imaging sensor. We recorded both supra- and subthreshold activity of neurons during electrically triggered responses and seizure-like events induced by 4-AP. The SPAD imaging sensors, we use here, can readily be applied to existing microscopes hence giving

access to voltage imaging in most in vitro, ex vivo, and head fixed settings.

The ability of SPADs to detect the time of photon arrival makes them well suited to imaging fast changes in neuronal membrane potential. The effective frame rate can be chosen arbitrarily in post-processing; higher frame-rates can be achieved by shorter binning intervals, although at the cost of decreased SNR. Given their low intrinsic noise at high frame rates the quality of the imaging signal is essentially shot noise limited. SPAD-based sensors also have the advantage that their operation does not require bulky cooling. This may make them well suited to development of miniaturized imaging systems, for example to monitor neural activity in freely moving animals.^[18]

The primary constraints for imaging neuronal membrane potential via SPAD sensors come from their quantum efficiency and from the properties of the GEVIs used. Our experiments establish that a SPAD sensor with an external quantum efficiency of $\approx 10\%$ is sufficient for imaging activity reported by GEVIs. Because the internal efficiency of our sensor is $\approx 40\%$ (at 480 nm), a potential way to improve the external efficiency would be through microlensing strategies that increase the fill factor. The native fill-factor of the current SPAD, at 26.8%, is relatively high for “planar” SPADs, but could be improved through microlensing. For example, we previously achieved a 2 \times increase in fill factor by integrating a custom microlens array onto the chip,^[14] while other SPADs have used a combination of microlenses and 3D stacking to optimize the fill factor.^[19] Recently developed SPADs^[20] featuring multibit pixels and efficiency $> 50\%$ will likely lead to further substantial additional improvements in signal to noise ratios, with greater proportions of detected photons reducing the

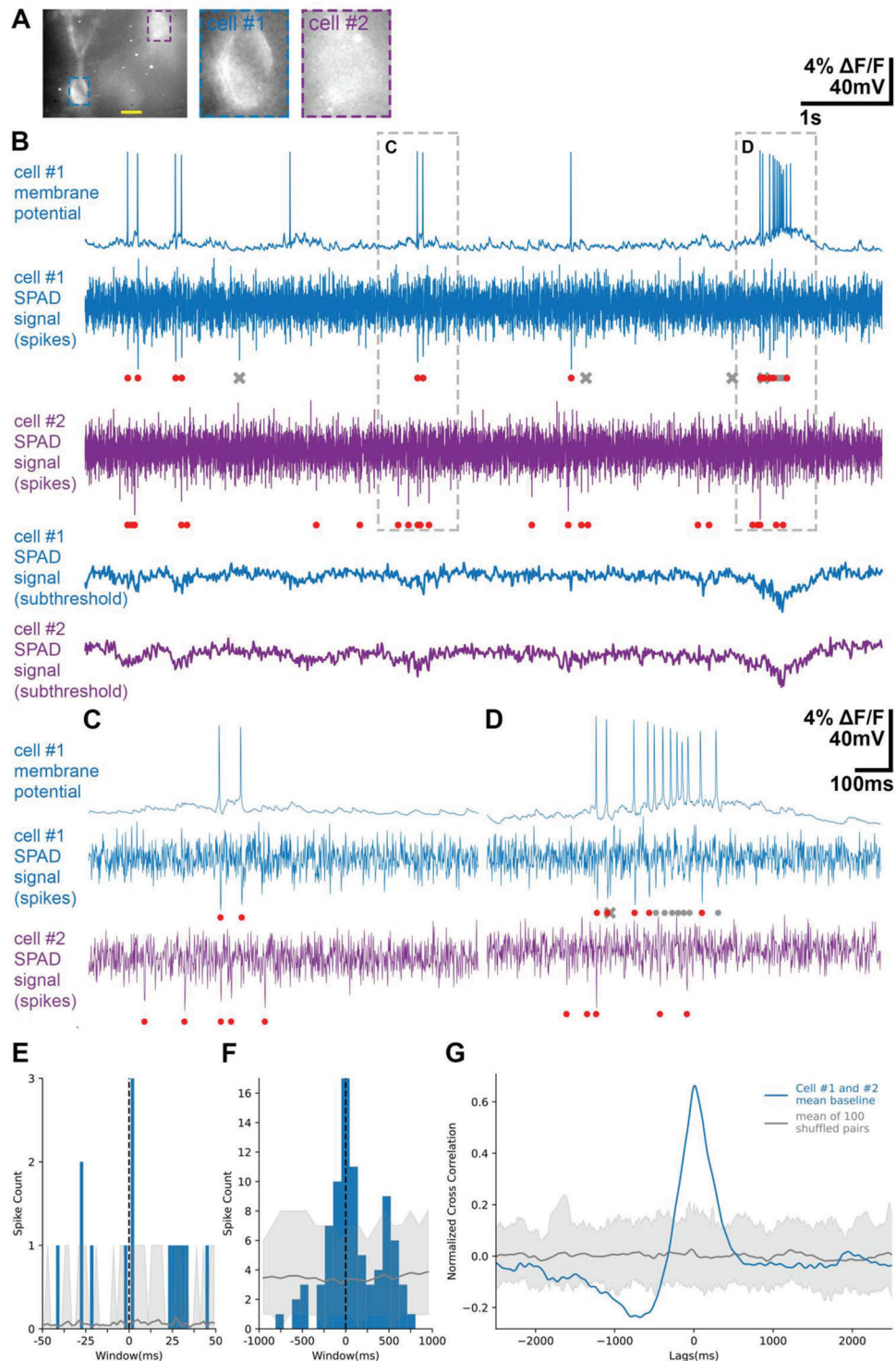


Figure 5. Coordination of spiking activity during seizure-like events. A) A field of view of recorded neurons in the striatum oriens layer of CA1 expressing Voltron-JF525-HTL, taken with a SPAD image sensor using a 20 \times water immersion objective. Scale bar: 10 μm . B) Activity of the neurons in (A). The membrane potential changes of Cell #1 recorded with the patch-clamp electrode is shown above the optical traces captured by the SPAD image sensor for Cell #1 (blue) and Cell #2 (purple) and processed to isolate spikes and subthreshold activity (see Figure S5, Supporting Information). C, D) Dotted-line areas in (B) are expanded, showing low frequency firing (C) and bursts (D). When the optical trace of cell #1 was compared with the ground-truth electrical trace, the correctly detected spikes were marked with a red dot, the false negative spikes were marked with a grey dot, the false positive spikes were marked with a grey cross. E, F) Peristimulus time histogram (PSTH) of spiking activity between Cell #1 and Cell #2 from (B). With the spikes of Cell #1 being set as reference events, the sum of spike counts of Cell #2 was plotted with 2 ms bins (E), and the sum of spike counts of Cell #2 was plotted with 100 ms bins (F). Blue trace: PSTH of the two traces from the cell pair. Grey line: mean PSTH of shuffled data. Grey shaded area: 5–95% percentile

shot noise. For example, while the SPAD we used here was designed for acquisition of binary frames to allow for compact pixels built into a relatively large array, introduction of multi-bit photon counting will increase the dynamic range of the sensor, potentially enabling lower magnification objectives to be used, and more neurons to be captured in the field of view. Further improvement in brightness and sensitivity of voltage indicators^[21–23] could further enhance detection of voltage signals in subcellular compartments and more densely-labelled samples.

4. Experimental Section

SPAD Imaging Sensor: The sensor chip was fabricated in STMicroelectronics 130 nm imaging CMOS technology.^[13] A detailed description can be found in.^[13] Custom FPGA firmware was developed to allow the camera to be triggered directly from the electrophysiology software (Axograph 1.7.6) via the digitizer board (National Instrument, USB-6212). In all experiments, raw bit frames were captured from the SPAD image sensor at a sampling rate of 9.9384 kHz. The resolution of the imaging system incorporating the SPAD sensor was tested using a NBS 1963A Resolution Target (Thorlabs) (Figure S6, Supporting Information). Live mode images were captured with a digital CCD camera and the SPAD camera separately. The intensity profile of images was calculated in MATLAB.

Animals: All animal experiments were conducted in accordance with the UK Animal (Scientific Procedures) Act 1986 and were approved by University of Edinburgh's Animal Ethics Committee (PP5105514).

Injection of Viruses and Dye: Male and female wild-type C57BL/6J mice aged between 21 and 40 days were anesthetized with isoflurane, mounted in a stereotaxic frame, and a small craniotomy made above the target region. To achieve sparse labelling, a mixture of pENN-AAV-hSyn-Cre-WPRE-hGH (Addgene # 105553-AAV1) (titer $\geq 2 \times 10^9$ vg mL⁻¹) and pAAV-hsyn-flex-Voltron-ST (Addgene # 119036-AAV1) (titer $\geq 1 \times 10^{12}$ vg mL⁻¹) injected into the primary visual cortex (stereotaxic coordinates AP -3.8, ML 3.0, DV -0.3, and -0.6) and CA1 (AP -3.8, ML 3.0, DV -1.3). Injection coordinates were calculated relative to bregma and 200 nL of the virus mixture was injected at each coordinate. JF525-HTL (100 nMol, Lavis Lab, Janelia Research Campus, HHMI) was mixed with 20 μ L of DMSO (Sigma, D2650-5 \times 5 ML), 20 μ L of Pluronic F-127 (20% Solution in DMSO) (ThermoFischer Scientific P3000MP) and 80 μ L of 1x PBS and administered intravenously into the lateral tail vein of mice 24–72 h prior to experimentation.

Electrophysiology and Fluorescence Imaging in Mouse Brain Slices: Preparation of brain slices and electrophysiological recordings were carried out as described previously.^[24] Briefly, sagittal brain slices were prepared from male and female wild-type C57BL/6J mice aged between 60 and 100 days. Mice were sacrificed by cervical dislocation. The brains were quickly removed and placed in ice-cold (2–4 °C) cutting artificial cerebrospinal fluid (ACSF) (pH 7.4) containing (in mM): 86 NaCl, 1.2 NaH₂PO₄, 2.5 KCl, 0.5 CaCl₂ (1 M), 7 MgCl₂ (1 M), 25 NaHCO₃, 25 Glucose, 50 Sucrose, and aerated with 95% O₂, 5% CO₂. The injected hemispheres were mounted on a Vibratome (Leica VT 1200, Leica Microsystems) and cut at 400 μ m thickness, then transferred to standard ACSF containing (in mM) 124 NaCl, 1.2 NaH₂PO₄, 2.5 KCl, 2 CaCl₂, 1 MgCl₂, 25 NaHCO₃, 20 Glucose and incubated at 37 °C for 15 min. The slices were stored at room temperature in a submerged chamber constantly aerated with 95% O₂, 5% CO₂ for at least 1 h before being transferred to the recording chamber perfused with standard ACSF at a flow rate of 3 mL min⁻¹ at 33 °C

The brain slices were first visualized with a digital CCD camera (SciCam Pro, Scientifica) mounted on an upright microscope (BX51-WI, Olympus) using either a 40x water-immersion objective lens (1.0 N.A., LUMPLFLN 40XW Olympus) or a 20x water-immersion objective lens (1.0 N.A., XLUMPLFLN 20XW, Olympus). For epifluorescent imaging of JF525 dye, which is maximally activated at 525 nm and has emission maximal at 549 nm, a green LED (Thorlabs M530L4) was used with a dichroic filter set (Semrock FF520-Di02, FF01-500/24, and FF01-562/40). Recording pipettes were pulled from borosilicate capillary glass (Havard Apparatus, 30–0060) on a horizontal electrode puller (P-97, Sutter Instruments) to a tip resistance of 4–6 M Ω and filled with K-gluconate-based internal solution (in mM: 130 K Gluconate, 10 KCl, 10 HEPES, 2 MgCl₂ (1 M), 0.1 EGTA, 2 Na₂ATP, 0.3 Na₂GTP, 10 NaPhosphoCreatine, Biocytin 0.5%, pH 7.0–7.5, 290–300 mOsm). Recording pipettes were positioned with a micromanipulator (Sensapex). Patch-clamp recordings were performed with a Multiclamp 700B amplifier (Molecular Devices), with membrane potential sampled at 20 KHz, filtered at 10 KHz with the built-in 4-pole Bessel Filter, and digitized (National Instrument, USB-6212). The command waveforms of electrical stimuli were generated, and the electrophysiological recordings were acquired using Axograph 1.7.6. Axograph was also used to generate signals to synchronize electrophysiological recordings and the SPAD imaging.

When whole-cell configuration was achieved, the light path of the microscope was switched to focus into the SPAD image sensor. To record baseline membrane potential changes, stimulation protocols of current steps (100 ms baseline, -200–200 pA, 100 ms per step, 100 pA increment) in current clamp and voltage steps (100 ms baseline, -50–30 mV, 100 ms per step, 20 mV increment) in voltage clamp were applied. In voltage clamp, cells were held at -70 mV. To record action potentials, a train of 10 current pulses (2 nA, 2 ms) was given to the neurons, with the frequency ranging from 25 to 100 Hz.

To record neural activity during seizure-like events induced by 4-AP, one electrode was placed to record field potential whereas another electrode was used to achieve whole-cell configuration in a Voltron labelled neuron. When the whole-cell configuration was achieved, 200×10^{-6} M 4-AP was added to the ACSF to induce seizure-like activity. When changes in field potential were recorded, indicating the onset of the seizure-like events, the SPAD imaging sensor was then started to record for 100 s (Ten consecutive epochs of 10-second-long recordings).

Data Analysis: For composing images, post-processing algorithms were used to compensate for both the dark count rate of the pixels and the logarithmic response of the pixels.^[25] Regions of interest (ROI) were selected manually and a binary mask was applied to allow isolation of the fluorescence-positive pixels. For each bit plane counts were summed within the ROI and values for all bit planes concatenated to produce an optical trace. When generating time traces, only the logarithmic response of the pixels was corrected.^[26] For each optical trace, high-frequency noise was removed by applying a low-pass filter at 2 kHz. Temporal binning was carried out by averaging consecutive bit planes in time (Python 3.8), e.g., binning ten consecutive planes gives an effective sampling rate of 1 kHz.

For investigation of responses to subthreshold current steps and voltage steps the mean fluorescence intensity of the first 100 ms within the ROI was used as the baseline fluorescence (F). The ΔF was calculated as the mean fluorescence at the following time intervals of the optical trace: 120–180, 290–350, 460–520 (voltage step only), 630–690, and 800–860 ms minus the baseline fluorescence. The $\Delta F/F$ and SNR of each step were calculated as these values were divided by the baseline fluorescence or the standard deviation of the baseline fluorescence. For each trace, simple linear regression was applied with the intercept set at $x = 0$ and $y = 0$ (Excel Version 16.54). To calculate the time constant (τ) of each current step, the upward or downward deflection of 100 ms of electrophysiology and optical recordings after the onset of each current step were fitted with the

of shuffled data. Hundred randomly shuffled cell pairs were generated by shuffling the spike times of the original two traces. G) Cross-correlogram of subthreshold signals between the cells in (B). Blue trace: mean normalized cross correlation of ten 10-second-long optical recordings of subthreshold baseline fluctuations. Grey line: mean normalized cross correlation of shuffled data. Grey shaded area: 5–95% percentile of shuffled data. Hundred randomly shuffled cell pairs were generated by pairing 10-seconds recording traces from different cells in different brain slices.

nonlinear regression (curve fit) function in Prism (Version 9.1.2). The constant of the best-fitted nonlinear regression equation was the tau.

For the action potential dataset, each spike was identified using the data processing pipeline provided in,^[1] and the peak intensity was read accordingly. $\Delta F/F$ was calculated as the peak intensity of each spike minus the baseline fluorescence of the same optical trace, divided by the baseline fluorescence. SNR of each spike was calculated as the peak intensity of each spike minus the baseline fluorescence of the same optical trace, divided by the standard deviation of the baseline fluorescence. The $\Delta F/F$ and SNR of each trace was the mean of $\Delta F/F$ and SNR of all the spikes identified in that optical trace.

The optical traces from the 4-AP experiments were first low-pass filtered at 2 kHz to remove high-frequency noise. Subthreshold activities were revealed by temporally binning the low-pass filtered raw traces at 100 Hz and baseline drift was removed by applying the detrend command (no Type specified) in MATLAB (R2019b). To analyze the spiking events, the low-pass filtered raw traces were further low-pass filtered at 10 Hz to remove subthreshold events and baseline drift. The optical traces were then temporally binned at 1 kHz. Spike detection was carried out using the “get_spikes” function in the Voltage Imaging pipeline (https://github.com/ahrens-lab/VoltageImaging_pipeline).^[1] Three parameters (rolling window size, threshold sets for spike size and standard deviation) were optimized by comparing with the ground truth electrophysiology data to achieve minimal false positive spikes while preserving as many true positive spikes as possible. Peristimulus time histogram (PSTH, George Gerstein, U. of Pennsylvania, Neuroscience) was calculated in Python (Python 3.8) with optimized window sizes and bin sizes to show millisecond-level correlation (50 ms window, 2 ms bin) and second-level correlation (1 s window, 100 ms bin) separately.^[26] To compare with traces that have the same firing rate, 100 randomly shuffled cell pairs were generated by shuffling spike times of the original two traces. Cross-correlation of subthreshold activities was analyzed by normalizing the activity traces and using the “spicy.signal.correlate” function in Python with full lags. 100 randomly shuffled cell pairs were generated by pairing 10-seconds recording traces from different cells in different brain slices.

Statistical Analysis: All summary data were presented as mean \pm standard error of the mean (SEM) from at least three independent recordings. To compare the tau values measured by the optical and electrical traces, linear mixed effect models (LMEs) were fitted using the package lme4 v1.1-12^[27] in R v4.1.2 (R Core Team, 2014). Differences were considered significant when $p < 0.05$.

Supporting Information

Supporting Information is available from the Wiley Online Library or from the author.

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Conflict of Interest

The authors declare no conflict of interest.

Data Availability Statement

The data that support the findings of this study are available from the corresponding author upon reasonable request.

Keywords

genetically encoded voltage indicators (GEVIs), kilohertz frame rate, neural activity, shot noise, single photon avalanche diodes (SPADs), temporal binning, voltage imaging

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