Design for Fast Optogenetic Screen In Mammalian Cells

For Next Gen Ca^{2+} Sensors

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Abstract

Genetically encoded fluorescent biosensors are proving to be powerful tools in neuroscience. The GCaMP6 Ca\textsuperscript{2+} sensor is widely used, and there are now many proof-of-principle versions for many second messengers that show promise. Improving these has been challenging because testing them involves a low throughput, labor-intensive processes. Our goal was to create a live cell system that uses a simple, reproducible, optogenetic process for testing prototypes of genetically encoded biosensors.

Blue light was used to activate an adenylyl cyclase enzyme from the soil bacterium *Beggiatoa* (bPAC) that increases intracellular cAMP (Stierl et al. 2011) as detected by the red sensor R-CaDDis. In turn the cAMP opened a cAMP gated channel (olfactory cyclic nucleotide gated channel, CNG, or the hyperpolarization-activated cyclic nucleotide gated channel, HCN2). This produced slow Ca\textsuperscript{2+} transients as detected by R-GECO1.2. To speed these transients up, we added the inwardly rectifying potassium channel 2.1, Kir2.1, and the bacterial voltage gated sodium channel (NAVROSD). This is a modular system in which the kinds of channels, and their relative amounts, can be tuned to produce the cellular behavior crucial for screening a particular biosensor in an automated format.

Introduction

Why are Ca\textsuperscript{2+} sensors important?

The calcium ion is an important second messenger in cellular processes ranging from neuronal function and muscle contractility, to fertilization and embryogenesis (Bear, Connors,
and Paradiso 2007). In the mammalian neuron, Ca\textsuperscript{2+} plays a role in many signaling pathways, ultimately controlling the release of neurotransmitters from neurons. Genetically encoded Ca\textsuperscript{2+} indicators contain fluorescent proteins that change fluorescence intensity as a function of Ca\textsuperscript{2+} concentration (Kerruth et al. 2019; Huang et al. 2019). They are minimally invasive, and since they are genetically encoded their expression can be exquisitely targeted to specific cell types and tissues (Agetsuma, Matsuda, and Nagai 2017; Bruton, Cheng, and Westerblad 2020; Kim et al. 2019). The fluorescence of these biosensors is typically captured with wide field microscopy and standard CCD cameras. However, there are many new imaging approaches to capture the wealth of information they provide. Two photon microscopes make it possible to image micro domains of Ca\textsuperscript{2+} signals within neurons in thick living tissue (T.-W. Chen et al. 2013; Peron et al. 2015). Light sheet microscopy makes it possible to now capture entire larval zebrafish brain activity at single cell resolution at 2 Hz (Ahrens et al. 2013). Recently, Ca\textsuperscript{2+} sensors are being coupled with functional magnetic resonance imaging (fMRI) (Barandov et al. 2019; Ghosh et al. 2018).

There are now several Ca\textsuperscript{2+} sensors that can sense Ca\textsuperscript{2+} transients in cells with millisecond response times and that are bright enough to image in living tissue (Ouzounov et al. 2017). Most of these sensors have blue excitation and green emission peaks. Currently, the sensors with the fastest kinetic response time to Ca\textsuperscript{2+} flux are the Ultrafast GCaMP6f, jRCaMP and jRGECO1a (Helassa et al. 2016; Kerruth et al. 2019). The recent jGCaMP7 sensor makes it possible to capture individual action potentials and can track large neuronal populations when using two photon or wide field imaging (Dana et al. 2019).

There is still a need for better Ca\textsuperscript{2+} sensors that are red (Zhao et al. 2011), or near IR (Qian et al. 2019), because these longer wavelengths enable investigators to image deeper
into thick tissues that absorb and scatter light in the blue-green range. A greater maximal change in fluorescence for these sensors is needed, since a better signal to noise ratio can help increase imaging speeds and increase the field of view. Brighter sensors can be detectable with lower expression levels. This could be important since overexpression can lead to epileptiform activity (Dana et al. 2019). Sensors built with calcium binding properties such as calmodulin present specific problems in vivo such as buffering the internal Ca$$^{2+}$$ concentration, and activation of kinases and phosphatases, and modulate ion channels (Mank and Griesbeck 2008). Further, creating sensors with linear response functions is needed, since the nonlinearities in GCaMP has led to decreased ability to accurately detect spikes (Deneux et al. 2016). Finally, jGCaMP7 and G-GECO1 sensors are imaged using green light. The short wavelengths of light needed to image these sensors can heat and damage the brain, which decreases the duration of imaging possible of the tissue (Podgorski and Ranganathan 2016).

One of the most effective approaches to expanding the fluorescent biosensor toolbox involves screening thousands of prototypes and then evolving for optimal function (Ibraheem et al. 2011; Park et al. 2013; Ding et al. 2015; Alford et al. 2012). For ease of use, this can be done in *E. coli* colonies or lysates where it is possible to screen >10,000 colonies per round of evolution (Zhao et al. 2011); (Zhao et al. 2014). A drawback of this approach is that a sensor that works in *E. coli* may not work as well in mammalian cells. A more labor intensive, lower throughput, approach is to screen in mammalian neurons. For example, in the evolution of GCaMP6 to GCaMP7, 446 variants were screened, similarly R-Geco to J-RGeco1 required screening 689 variants, and RCaMP1 to JRCaMP1 screened 934 variants (Dana et al. 2016, 2019).
What if we could screen thousands of variants in mammalian cells for sensitivity, speed and dynamic range? In theory Ca\(^{2+}\) sensors could be screened in excitilbe human embryonic kidney cells (HEK-293 cells). The Cohen lab has pioneered the use of excitilbe HEK-293 cells to screen voltage sensors. They combined the inwardly rectifying potassium channel (Kir2.1) and voltage gated sodium channel (NAV1.3) and field stimulation to produce rhythmic depolarization and hyperpolarization in HEK cells (Park et al. 2013; Zhang, Reichert, and Cohen 2016),(Shah et al. 2004). They then created a stable cell line that expresses Kir2.1, Nav1.7 and CheRiff, a channelrhodopsin. This can be photoactivated thereby creating a system that produced oscillating membrane voltage (Zhang and Cohen 2017). Yet, there are inherent problems with stable cell lines, such as drift and silencing. In addition, the level of expression of each of the channels cannot be easily changed.

Similarly, another paper described a cell line that had pacemaker like activity. They stably expressed the HCN2 and Kir2.1 channels, and transiently expressed a voltage-gated channel Cav3.1 or Nav5.1 (K. Chen et al. 2018). HCN2 is a hyperpolarization activated, cyclic nucleotide gated ion channel (K. Chen et al. 2018). Nav1.5 is a voltage gated sodium channel responsible for action potentials in cardiomyocytes (K. Chen et al. 2018). Chen et al. found that cells expressing HCN2 and Kir2.1 had action potentials with depolarizations that last over \(~450\) ms, and a hyperpolarization phase that lasts \(~737\) ms (K. Chen et al. 2018). With either Cav3.1 or Nav5.1 expressed in conjunction with the Kir2.1 and HCN2, they found that there was an average of one action potential per second. This suggests that even without field stimulation, cells expressing these channels can create their own auto-depolarization leading to rhythmic oscillations in membrane potential. They also showed that the currents produced by the voltage gated channels influenced the speed at which the action potentials occur.
In theory, we could use similar channels to create cells that do not require field stimulation, and have faster action potentials and whole cell Ca$^{2+}$ transients. We wanted the system to be automatable, inexpensive, modular, and useful in many different cell types. Our goal was to create a screening platform for biosensors that is consistent, reliable, and capable of quickly screening thousands of prototypes.

Results

Establishing an Optical Actuator

How can we actuate a biosynthetic cell line using light? To develop a optogenetically controlled screen we chose to use bPAC (blue photo activated cyclase), a blue light activated adenylyl cyclase from the soil bacteria *Beggiatoa* (Stierl et al. 2011). bPAC enzyme is activated with 480 nm light which then converts ATP into cyclic adenosine monophosphate (cAMP). R-caDDis is a fluorescent cAMP sensor that is excited with 561 nm light (Bernal Sierra et al. 2018). We began by examining whether blue light stimulus can produce a measurable change in cAMP levels in HEK-293 cells.

We transduced HEK-293 cells with bPAC (2x10$^{11}$ Vg/mL, viral titer) and R-caDDis (5 x 10$^{10}$ Vg/mL) (Fig. 1 A). The following day the cells were stimulated with 10 milliseconds of blue light and then images were collected continuously with 561 nm excitation that does not activate bPAC. Upon blue light photoactivation, the R-CaDDis red fluorescence increased. The quantification of the red fluorescence reveals that cAMP levels within the cell rise over the time frame of ~100 seconds (Fig. 1C). An obvious increase in red fluorescence can be seen upon blue light activation and is demonstrated in the images of HEK-293 cells taken pre and
10 seconds post blue light stimulation (Fig.1B). Control cells with no bPAC did not show an increase in red fluorescence. The system response is remarkably reproducible and the same cells can be repeatedly stimulated. Indeed the only limitation is that too much blue light given over a short period of time will create too much cAMP and saturate the sensor. The slow decrease in fluorescence of R-CaDDis is most likely due to the phosphodiesterases present in the cell which work to eliminate cAMP (Tewson et al. 2016).

**Figure 1 : 20ms of blue light is sufficient to activate bPAC.** A) HEK-293 cells were transduced with bPAC (the actuator) and red-caDDis, the red fluorescence cAMP sensor. B) The cells were briefly illuminated with 20ms of 480nm light. This activation of bPAC increases cAMP levels which increases R-cADDIs fluorescence. Brightness and contrast were the same for both images. C) 20 ms of blue light activates bPAC which produces bursts of cAMP over a period of 100 seconds. This is a stereotyped response that is repeatable with multiple doses of blue light.

**Dose Dependence**

To examine the dose-response relationship for blue light and cAMP production, the intensity of the blue light was systematically varied during stimulation while quantifying the r-CaDDis fluorescence. HEK-293 cells were transduced with bPAC and R-caDDis, and each well in a 96-well plate was given a 15 milliseconds dose of blue light. For each well, the power of the LED 480nm light was increased stepwise from 0% to 100% in increments of five using a THORLABS LED controller. At 100% power the light source measured 70mW/cm². Figure 2 reveals a near-linear dose response in which increasing levels of blue light produces increases in cAMP levels. Above 50% power to the LED, there was an increased variability in the
response that is most likely due to saturating the sensor with cAMP. It is possible that lowering the level of bPAC expression would make it possible to create a system with a greater range of tunability before saturating the sensor with cAMP. The fluorescence change, delta F, was used to determine the cAMP level and all points in the dose response are an average of multiple trials (Fig. 2).

**Figure 2: There is a linear dose/response relationship between the power of blue excitation and the fluorescence response of R-caDDis.** As the blue light intensity increases so does the cAMP production. The delta F is a measure between the base and max of the change in fluorescence (inset). Each data point represents an average of four trials.

**Coupling the Actuator with an ion Channel**

To couple cAMP production with depolarization, we expressed a cyclic nucleotide-gated ion (CNG or HCN2) channel. The CNG channel is normally gated by cyclic guanosine monophosphate (cGMP) (Fesenko, Kolesnikov, and Lyubarsky 1985). However, several mutations can be introduced (Rich et al. 2000) that render it sensitive to cAMP. In theory, this mutant channel should couple bPAC activation to a current that depolarizes the cell. In addition the HCN2 is cAMP gated and must be hyperpolarized to activate (K. Chen et al. 2018).

We expressed a CNG channel (1.26 x 10^11 Vg/mL), bPAC, and R-GECO1 in HEK cells (Fig. 3A). We triggered the activation of bPAC with blue light. Figure three illustrates that upon 20ms of blue light activation of the enzyme, the red fluorescence of R-GECO increased. The opening of the CNG channel increased Ca^{2+} levels in the cell which was quantified by an increase in red fluorescence levels of R-GECO1 [16] (Fig. 3). The Ca^{2+} levels (Fig 3. B) stayed
elevated for ~30-80 seconds before their levels began to fall. The control cells without bPAC showed no response (supplemental Fig. 2).

**Figure 3: Activation of bPAC leads to a Ca^{2+} transient when CNG or HCN2 is expressed.** A) Blue light stimulation of bPAC should produce cAMP that in turn opens the mutant CNG channel. B) The red trace is the R-GECO response to 20 ms of blue light activation in cells with both the CNG and bPAC. C) Blue light stimulation of bPAC should produce cAMP that opens the HCN2 channel. D) the trace is the R-GECO1 response to 20 ms of blue light activation in cells with both the HCN2 and bPAC.

**Additional Channels Can Tune the Speed and Frequency of Ca^{2+} Transients**

While the bPAC and CNG or HCN2 produced very reliable Ca^{2+} transients, these were long lived events with slow kinetics. In theory introducing voltage regulated potassium and sodium conductance could speed these transients up. The recent paper by Chen et al. found that varying the levels of Kir2.1, and Cav1.3 or Nav1.5 affected the membrane oscillations when recording membrane voltage using electrophysiology (K. Chen et al. 2018).

Wild type HEK-293 cells have a resting membrane potential of -25 mV but with Kir2.1 expressed their resting potential decreases to -70 mV (Nguyen, Kirkton, and Bursac 2016; Hille 2001; Kirkton and Bursac 2011; Yu et al. 2004). Kir2.1 is an inwardly rectifying potassium channel that upon hyperpolarization greater than -20 mV increases its conductance producing an outward current (Hille 2001).

A 2016 study by the Bursac lab (Nguyen, Kirkton, and Bursac 2016) describes the bacterial sodium channel, NavRosDg217a (NavD). It is characterized by a rapid activation state that begins to open when the cell depolarizes to -30 mV and a slow inactivation state
(hundreds of milliseconds) that starts during the depolarization (Nguyen, Kirkton, and Bursac 2016). The NavD channel is rapidly inactivated at +40mV, although, it must return to a hyperpolarized state below -40mV for it to move from an inactivated state to deactivated, at which point the channel can re-open (Nguyen, Kirkton, and Bursac 2016). Co-expressing the NavD channel with Kir2.1 can create an excitable system because the Kir2.1 hyperpolarizes the cell after each spontaneous depolarization of the sodium channel.

We choose the NavRosDg217A for two reasons. First, the human sodium channel has an enormous coding region that exceeds the carrying capacity of most viruses. Bacterial channel subunits are much smaller. Secondly, we wanted to create a system with long slow depolarizations such that membrane depolarization events could be captured with slower imaging speeds (10Hz). We transduced the bacterial channel NavRosDg217A (Nguyen, Kirkton, and Bursac 2016), Kir2.1, CNG, and R-GECO1 into HEK-293 cells. There is some degree of spontaneous activity, but when we activate bPAC with 20ms of blue light there were robust responses in all of the wells. There were obvious Ca$^{2+}$ transients that occurred in 90% of the cells, but the activity was largely uncoordinated. Each cell has its own “signature” change in Ca$^{2+}$ levels, as seen in the R-GECO1 fluorescence traces (Fig.4B). The fluorescence traces encode the change in the intracellular Ca$^{2+}$ concentration levels of the cell.

**Figure 4: The kir2.1 and voltage gated sodium channel is vital for creating faster Ca$^{2+}$ transients.** A) In theory introducing the NavD and Kir2.1 channel should create faster oscillating Ca$^{2+}$ transients in the cell. B) Each cell has a unique Ca$^{2+}$ transient response after blue light stimulation of bPAC.

**Optimizing the Activity**
Our goal was to create a highly reproducible system that can be optimized. We propose that different levels of the Kir2.1 and NavD would lead to varying responses. We systematically transduced cells with varying ratios of Kir2.1 and NavD while holding the levels of bPAC and CNG channels constant. The different ratios of the channels lead to different responses. To quantify the Ca\(^{2+}\) imaging data we analyzed data using a custom made Matlab Program (Fig. 5A). We wanted to break down each Ca\(^{2+}\) transient into components that could be analyzed and looked at in multiple ways. For each Ca\(^{2+}\) transient we examined the prominence of the peak, the full width of the peak at half maximum (FWHM), and the inter-peak interval.

**Figure 5: Optimizing the ratios of voltage gated channels in the unstable system illustrated in figure 4.** A) Diagram of the Matlab analysis of the Ca\(^{2+}\) transients. Peak intensity, full width of the peak at half maximum (FWHM), and inter-peak intervals were analyzed. B) Varying the ratios of Kir2.1 to NavD produce varying Ca\(^{2+}\) transient event widths. The mean peak (FWHM) was analyzed with the standard deviation(SD) of the mean reported. C) The mean intervals between peaks correlate with the varied levels of Kir2.1 to NavD. SD of mean shown. D) Peak intensity histograms of Ca\(^{2+}\) transient events in cells transduced with a Kir2.1 to NavD ratio of 14:1.

The Peak Finder in Matlab marks the peaks by their location in time and their maximum fluorescence intensity. The peak intensity was then measured from the base fluorescence to the peak that was pre-determined using the Matlab Peak Finder (Fig. 5A). The peak intensity of the events does not appear to be random. Histograms of peak intensity (Fig. 5D) show what appear to be units of activity as though the Ca\(^{2+}\) response is integrating one, two, or three events.
In addition, we examined the width of the peak which was measured as the full width of the peak at half maximum (Fig. 5A). The peak (FWHM) of the Ca\textsuperscript{2+} event is the length of time that the Ca\textsuperscript{2+} concentration stayed elevated for each defined peak. The fastest event occurred for 120ms, to ensure that each event was captured we sampled with continuous 40ms exposures. The fastest average event (Peak FWHM) for a Ca\textsuperscript{2+} transient occurred in cells with a Kir2.1 to NavRosD viral infectious unit ratio of 22:1 and generally lasted for ~400ms. Next, we measured the inter-peak interval, which is the difference between two consecutive peaks. The inter-peak interval is a measure of how frequently Ca\textsuperscript{2+} transients occur in a given system after a previous response. Again, the cells that exhibited the shortest average inter-peak interval were cells that had the Kir2.1:NavD ratio of 22:1, although there was a great deal of cell to cell variability. A single burst of blue light produced activity for an average of two-three minutes.

Optimizing Activity with HCN2

The behavior of HCN2 was quite different than CNG. Figure six depicts the experimental setup, with R-GECO1, bPAC, Kir2.1, and NavD introduced by viral transduction followed by transient transfection of the HCN2 plasmid. The following day images were collected with 561 nm excitation light for several minutes after a blue light activation pulse of 20ms.

**Figure 6: Replacing the CNG channel with the pacemaker HCN2 channel** increases the Ca\textsuperscript{2+} transient duration and the time between events. A) The HCN2 channel can be exchanged for CNG. B) The transient events lasted over several seconds. C) The
Inter-Peak Interval time increased in concordance to increased Ca\textsuperscript{2+} event widths. The standard deviation of the mean is shown.

We found that the HCN2 produced longer inter-peak intervals, and longer Ca\textsuperscript{2+} transients, than cells expressing the CNG channel (Fig. 6). We found that the fastest events occurred in cells with a viral unit ratio of Kir2.1 to NavD of 14:1 with an average of ~3 seconds (Fig. 6b). Compared to cells expressing CNG, the HCN2 cells in general had Ca\textsuperscript{2+} transients that lasted around seven seconds with the time between transients ranging between 30 to 60 seconds. Importantly, the HCN2 cells show continued Ca\textsuperscript{2+} transients that lasted for up to 14 minutes after a single stimulus of blue light.

Screening Green Sensors Is Possible

Although there are several well engineered Ca\textsuperscript{2+} biosensors that are green, there is still room for improvement, and continued engineering on green Ca\textsuperscript{2+} sensors persists. Further, if it is possible to image both red and green Ca\textsuperscript{2+} sensors simultaneously, that can be used to benchmark the other in terms of speed, signal amplitude, and linearity. G-GECO1 is based on the GCaMP sensor and is based on the circularly permuted GFP with a fusion to the M13 and calmodulin at the N and C termini (Zhao et al. 2011). It is pH-sensitive and is relatively dim until it is excited, at which point it has a two fold increase in fluorescence compared to GCaMP3 (Zhao et al. 2011).

Imaging G-GECO while continuously activating bPAC was generally straightforward. We held the virus amount of Kir2.1 at 4x10\textsuperscript{14} VG/\mu L, and NavD at 1.7x10\textsuperscript{13} VG/\mu L, at a viral ratio of Kir2.1:NavD at 22:1. The CNG channel viral titer was held at 2.5x10\textsuperscript{14} VG/\mu L and we expressed G-GECO1 heterologously (Fig. 7A). We varied the viral titer of bPAC from .5 - 4 x
We found that decreasing the bPAC dilution to $1 \times 10^{16}$ VG/µL decreased the width of the Ca$^{2+}$ event, as well as decreased the time between events (Supplemental Fig.1). In principal an image splitter could be used to compare red sensors of unknown speed with a benchmarked green sensor. This experiment shows that the system can be continuously activated using very low power light.

**Figure 7: Green Ca$^{2+}$ indicators can be screened with low levels of blue light. A)**

The continuous blue light illumination needed to image green Ca$^{2+}$ sensors should continuously activate bPAC. However, by reducing the amount of bPAC expression we are able to create an unstable system. The viral titer ratio of Kir2.1:NavD are held constant (22:1) and the total viral amount of bPAC was systematically varied. B) Increased viral levels of bPAC lead to Ca$^{2+}$ transients with longer widths. C) Increasing bPAC expression lead to increased time between events. Standard deviation of the mean is shown.

**Discussion**

Our goal was to create a simple, modular, reproducible system to screen Ca$^{2+}$ sensors. The system can be optimized by varying the ratio of Kir2.1 to NavD channels, thereby changing the type of Ca$^{2+}$ transients that occur and the frequency at which they occur. Therefore, optimizing the ratio of the channels to the experimenters preference for cell type and sensor can empower an automated screen for Ca$^{2+}$ sensors with different dynamic properties.

The designing of the high throughput screen for Ca$^{2+}$ sensors can be modified. If a slow sensor is preferable, the HCN2 channel with a greater Kir2.1 to NavD ratio is desirable. Otherwise, to discover a fast sensor with increased linearity to Ca$^{2+}$ dynamics, the CNG
channel with a fast sodium channel would be favorable. For example, the Ca\textsuperscript{2+} transients width (length of time transient occurs) could potentially be greatly reduced by using a faster voltage gated sodium channel. In addition, several papers have shown that including the connexin-43 protein increases the gap junction coupling of the HEK-293 cells (Nguyen, Kirkton, and Bursac 2016; Kirkton and Bursac 2011; Fahrenbach, Mejia-Alvarez, and Banach 2007). We did not test how connexin-43 might affect this system but speculate that it could potentially stabilize the activity between the cells allowing for more consistent oscillations. If the throughput needs to be increased, the screening process could be broken into two steps. First, the screen could use a HCN2 channel (Fig.6A) and the cells could be imaged slowly without creating large data files. The best variants could then be screened in cells expressing the CNG channel and a fast sodium channel. These prototypes will be expressed in cells that also express a benchmark sensor such as GCAMP7f. Finally, the most promising variants would then be photo physically characterized (Molina et al. 2019; Barnett, Hughes, and Drobizhev 2017).

The progress on improving Ca\textsuperscript{2+} and voltage sensors depends on the throughput of the screen (Park et al. 2013; Rodriguez et al. 2017). For example, the improvement of the voltage sensors has been very slow due to limitations in the throughput involved in whole cell patch clamp fluorometry (Storace et al. 2016; Park et al. 2013). The system described here will enable investigators to screen without drugs or field stimulation. Recording the activity in a well for a minute would eliminate false negatives and result in a time budget of roughly 100 minutes per 96 well plate. Alternatively, the FLIPR (Schroeder and Neagle 1996; Arkin et al. 2012) would make it possible to simultaneously screen 384 well plates which would move this screen into a high throughput realm.
Methods

Growth conditions for HEK-293 cells

A line of HEK-293 were cultured in DMEM, 10% FBS, and 1% Penicillin (100 U/mL). For experiments, HEK-293 cells were plated in 96 well plastic plates at 200μl per well (~3 x10^5 cells/mL).

Plasmids

The HCN2 plasmid was kindly donated by Joan Lemire at Tufts University from Dr. Michael Levin’s lab. The NavRosDg217 was reverse translated using a human codon preferences and synthesized as a Gblock by IDT (Coralville, Iowa). The cyclic nucleotide gated olfactory channel (Rattus norvegicus) has the following mutations, delta61–90/C460W/E583M and was synthesized as a Gblock by IDT. The rat channel can be found on NCBI at NP_037060.1 (Rich et al. 2001). pGEMTEZ-Kir2.1 was a gift from Richard Axel, Joseph Gogos & Ron Yu (Addgene plasmid #32641 ; http://n2t.net/addgene:32641 ; RRID:Addgene_32641)(Yu et al. 2004).

Baculovirus

The following channels, Kir2.1, CNG, and NavRosDg217 were placed into a BacMam plasmid and used to create high titer baculovirus (Montana Molecular, Bozeman, MT). The virus titers were: Kir2.1 titer 1.58x10^11 VG/mL, bPAC titer 1.98 x 10^11 VG/mL, CNG titer 1.26x10^11VG/mL, and NavRosDg217 titer 1.73 x 10^10VG/mL. The following bio-sensors
were obtained from Montana Molecular: The R-CaDDis (titer 5.06x10^{10} Vg/mL); R-GECO (1.32x10^{10} Vg/mL); and G-GECO (4.48x10^{10} Vg/mL).

**Transduction**

HEK-293 cells were prepared and transduced following manufacturer's recommended protocol (Montana Molecular). HEK-293 cells were plated in 96 well plastic plates at a monoconfluent layer of (3 x 10^5 cells/mL) and were imaged two days following transduction. The cells were washed with Tyrode’s solution once and then each well was filled with 100μl of Tyrode’s solution.

**Transfection of HCN2**

24 hours post transduction the HEK-293 cells were grown to ~80% confluency in 96 well plates and transfected using Mirus Bio TransIT®-293 transfection reagent using the manufacturer's recommendation. Cells were transfected with the HCN2 plasmid at 100ng/μl.

**An ASI modular microscope**

The experiments were conducted using wide field imaging on an ASI-XYZ stage fitted with a modular infinity microscope. The objective imaged onto the detector chip of a Hamamatsu ORCA-Flash 4.0 scientific-CMOS camera. The images were collected using either matlab scripts that controlled the camera, ASI stage, SH1-Thorlabs shutter, and ThorLabs DC4100 Four Channel-LED Driver, or by using μmanager (Edelstein et al. 2014). A custom illumination system was arranged. Briefly, a dichroic mirror (XX) was positioned at the entrance of the microscope to combine blue light LED illumination with yellow laser illumination. An SH1-Thorlabs shutter was used to create brief illumination from a blue
ThorLABS LED (488nm) for rapid stimulation of bPAC. At 100% the blue light illumination was 70mW/cm². The yellow illumination was provided with a Sapphire laser (561nm, 50mW, Coherent). The laser beam was steered with two mirrors (arranged in a periscope) into an entrance aperture of the ASI microscope. Before entering the microscope, the laser beam was diffused with a 50° diffuser (ED1 C50 MD, Thorlabs) placed in the focus of an f = 20 mm aspheric lens (ACL2520U, Thorlabs) that collimated the beam for further traveling into the microscope. The illumination intensity at 100% of the laser power at the sample was 10mW/cm².

Analysis

Image data was stored in a Z-stack tiff file and loaded into the FIJI distribution of the ImageJ software (Schindelin et al. 2012). The background fluorescence was corrected using the photobleaching gui in FIJI. The cells were hand selected using a freehand ROI surrounding the cell of interest. The average pixel value within the ROI for each frame was then loaded into MatLab. Analysis was built in Matlab using the Find Peaks in the Signal Processing Toolbox. The find peaks returns a vector with the local maxima (peaks) from the ROI trace data. In addition, it returns the widths of the peak and the prominence of the peak. The inter-event interval is the vector difference between each consecutive peak.

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Figure one: Only 20ms of blue light is sufficient to activate bPAC. A) HEK-293 cells were transduced with bPAC (the actuator) and red-caDDis, the red fluorescence cAMP sensor. B) The cells were briefly illuminated with 20ms of 480nm light. This activation of bPAC increases cAMP levels which increases R-cADDis fluorescence. Brightness and contrast were the same for both images. C) 20 ms of blue light activates bPAC which produces bursts of cAMP over a period of 100 seconds. This is a stereotyped response that is repeatable with multiple doses of blue light.
Figure 2: There is a linear dose/response relationship between the power of blue excitation and the fluorescence response of R-caDDis. As the blue light power increases so does the cAMP production. The delta F is a measure between the base and max of the change in fluorescence (inset). Each data point represents an average of multiple trials.
Figure 3: Activation of bPAC leads to a Ca2+ transient when CNG or HCN2 is expressed. A) Blue light stimulation of bPAC should produce cAMP that in turn opens the mutant CNG channel. B) The red trace is the R-GECO response to 20 ms of blue light activation in cells with both the CNG and bPAC. C) Blue light stimulation of bPAC should produce cAMP that opens the HCN2 channel. D) the trace is the R-GECO1 response to 20 ms of blue light activation in cells with both the HCN2 and bPAC.
Figure Four: The voltage gated sodium channel is vital for creating an unstable system. A) In theory introducing the NavD channel should create oscillating calcium transients in the cell. B) Each cell has a unique calcium transient response after blue light stimulation of bPAC. C) A Z-projection of a time series shows the cumulative activity of cells over 60 seconds. Wells without bPAC show little to no activity.
Figure 5: Optimizing the ratios of voltage gated channels in the unstable system illustrated in figure 4. A) Diagram of the Matlab analysis of the Ca\(^{2+}\) transients. Prominence, width of half prominence, and inter-event intervals were analyzed. B) Varying the ratios of Kir2.1 to NavD produce varying calcium transient event widths. C) The mean inter-event Interval changes in concordance to the varied levels of Kir to NavD. D) Prominence histogram of Ca\(^{2+}\) transient events in cells transduced with a 3:2 ratio of Kir to NavD.
Figure 6: Replacing the CNG channel with the pacemaker HCN2 channel increases the Ca\(^{2+}\) transient duration and the time between events. A) The HCN2 channel can be exchanged for CNG. B) The transient events lasted over several seconds. C) The Inter-Event Interval time increased in concordance to increased calcium event widths.
Figure Seven: Green Ca$^{2+}$ indicators can screened with low levels of blue light. A) The continuous blue light illumination needed to image green Ca$^{2+}$ sensors should continuously activate bPAC. However, by reducing the amount of bPAC expression we are able to create an unstable system. The levels of Kir:NavD are held constant (2.5:1) and the levels of bPAC were systematically varied. B) Increased levels of bPAC lead to Ca$^{2+}$ transients with longer widths. C) Increasing bPAC leads to increased time between events.
Supplemental Data

Supplemental Figure 1: Inter-Event Intervals are linearly dependent on the event width.