CalciumInsights: An Open-Source, Tissue-Agnostic Graphical Interface for High-Quality Analysis of Calcium Signals

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Abstract

Fluctuations and propagation of cytosolic calcium levels at both the cellular and tissue levels show complex patterns, referred to as calcium signatures, that regulate growth, organ development, damage responses, and survival. The quantitative analysis of calcium signatures at the cellular level is essential for identifying unique patterns that coordinate biological processes. However, a versatile framework applicable to multiple tissue types, allowing researchers to compare, measure, and validate diverse responses and recognize conserved patterns across model organisms, is missing. Here, we present a post-processing tool, CalciumInsights, which leverages the R packages Shiny and Golem. This tool has a graphical user interface and does not require software programming experience to perform

calcium signal analysis. The open-source software has a modular framework with standardized functionalities that can be tailored for various research approaches. CalciumInsights provides descriptive statistical analysis through various metrics extracted from dynamic calcium transients and oscillations, such as peak amplitude, area under the curve, frequency, among others. The tool was evaluated with fluorescence imaging data from three model organisms: *Danio rerio*, *Arabidopsis thaliana*, and *Drosophila melanogaster*, demonstrating its ability to analyze diverse biological responses and models. Finally, the open-source nature of CalciumInsights enables community-driven improvements and developments for enabling new applications.

Author Summary

This manuscript introduces CalciumInsights, an open-source tool for calcium signature analysis. Designed to be a versatile tool that works with various tissue types and biological systems, CalciumInsights has an easy-to-use graphical user interface. Our program simplifies metrics extraction while maintaining the quality of the analysis by integrating several algorithms. CalciumInsights stands out for its user-friendliness, ease of use, and robust data exploration features, such as tunable filters for improved accuracy. These features promote inclusivity and lower barriers to scientific research by making calcium signature analysis accessible to users of all programming skill levels.

Introduction

Cytosolic calcium is a crucial second messenger that facilitates communication within and between eukaryotic cells, coordinating activities across different cellular compartments (1,2). For example, these activities include but are not limited to the following biological processes: exocytosis, contraction, metabolism, transcription, fertilization, growth and development of multicellular organisms, immunity, and wound healing (3–6).

As a universal second messenger in almost all biological contexts, understanding the commonality of oscillation patterns or wave propagation across divergent systems is an essential step toward recognizing and defining the Rules of Life, which are fundamental principles governing biological processes and behaviors (4,7–10). Cellular and tissue

dynamics of cytosolic Ca^{2+} are complex, encompassing intracellular waves, spikes, oscillations, and coordinated cell-to-cell transmission (2,11,12). Diverse fluorescent, genetic, and imaging technologies are employed to acquire time-series data of Ca^{2+} fluctuations within cells and tissues. These tools include aequorin (13), fluorescent Ca^{2+} indicator dyes (14), genetically encoded calcium reporters or indicators (15,16), fluorescence and confocal microscopy (17), two-photon microscopy, electrophysiological techniques, optogenetic tools (18), and high-speed imaging systems.

By utilizing dedicated software and hardware, precise and reliable time-lapse series that accurately depict the dynamics of Ca^{2+} within target cells or tissues in multiple organisms can be generated. Determining the region of interest (ROI) where the Ca^{2+} signals occur is essential for a thorough quantitative analysis. Both manual and automated segmentations are frequently employed. Typically, the former is done with image analysis software such as Fiji (19). On the other hand, automated ROI segmentation can analyze the system's spatiotemporal properties and extract the ROI without human intervention, accelerating the process and reducing the potential human error (20,21). Following the segmentation of cells or identification of ROIs, fluctuations in fluorescence known as Ca^{2+} transients can be extracted for further analysis. Standard computational tools like MATLAB, Python, R, or Java (22–29) can be used to analyze Ca^{2+} transients; however, these programming platforms provide a barrier for many biological applications due to a steep learning curve, especially for researchers lacking the necessary computational background. As such, there is a compelling need for enhanced functionality in novel, readily available tools to extend the analytical potential of the previously outlined methods to a broader user base. There are proprietary software tools like CytoSolver (30), Clampfit (31), Igor Pro (32), CyteSeer (33), and HARVELE that facilitate the analysis of the Ca^{2+} transients (34–36). However, open-source techniques enhancing the development of tools that are as powerful but more widely accessible are still lacking (37). To address this deficiency, we developed CalciumInsights. This analysis tool is designed to be user-friendly, featuring an interactive graphical user interface (GUI) that eliminates the need for programming knowledge or the purchase of proprietary software with complex and unintuitive functions (38). CalciumInsights leverages the Golem R-package to create modular Shiny applications in R (39,40), enabling streamlined development and allowing users to incorporate custom-built

modules. This integration fosters collaboration within the open-source community, as users with advanced computational skills can tailor the application more precisely to their specific model system. These tailored applications can then be shared and utilized by others, expanding the utility and adaptability of CalciumInsights across diverse research contexts.

In this work, we used time-lapse fluorescence microscopy images from three common model organisms --*Danio rerio* (zebrafish), *Arabidopsis thaliana* (plant), and *Drosophila melanogaster* (fruit fly) -- to assess the efficacy and broad applicability of CalciumInsights. The Ca^{2+} traces from these images were analyzed and characterized using the metrics described in material and methods section. This versatility highlights CalciumInsights as an adaptable tool for analyzing data from different organisms across various imaging modalities and spatiotemporal resolutions, without bias toward any specific biological system or response.

Materials and Methods

Mathematical background

Ca^{2+} signals are essential in many scientific and medical settings because they provide vital insight into cellular activity involved in physiological processes. However, noise and artifacts often hinder the precise interpretation of these signals, complicating accurate analysis (41). Therefore, extracting meaningful information from these signals is essential for a comprehensive understanding of Ca^{2+} activity. To address this, we employ the Loess approach, a variation of *Locally Weighted Regression* (42,43). This method adjusts local polynomials to the data with weighted modifications to facilitate signal noise modeling and attenuation (44–46). The outcome is a smoothed representation of the original analog signal. The metrics of interest are then deduced from this smoothed signal. It is important to note that Loess parameters must be carefully chosen based on the characteristics of the experimental data and the specific metrics of interest.

Definition of the Loess Method

The fundamental mathematical basis of Loess involves estimating a smoothed value $\hat{f}(x_0)$ at a point x_0 through a weighted combination of observed values y_i in its vicinity (42,43). Mathematically, this is expressed as:

$$\hat{f}(x_0) = \sum_{i=1}^{n} w_i(x_0) y_i, \qquad \text{(Error! Bookmark not defined.)}$$

where $w_i(x_0)$ are the weights assigned to each y_i around x_0, depending on the relative distance between x_0 and x_i, which is controlled by the hyperparameter $span \in [0,1]$. The span determines the neighborhood of points considered in the fitting process by setting the size of the window used for the local regression. The width of the window is determined so that the proportion of data points it includes, relative to the entire dataset, matches a specified span parameter. A small span gives more weight to nearby points, allowing for a more flexible and responsive adjustment to local variations in the Ca^{2+} signal. In contrast, a large span assigns a more uniform weight to a greater number of points, resulting in a smoother adjustment less sensitive to local variations. The weights are calculated using a tricubic or Gaussian weighting function (42,43), giving more importance to points closer to x_0 in the fitting process.

Metrics

Quantifiable measurements are required to evaluate and contrast Ca^{2+} activity from control and perturbed or pathological experimental conditions (47,48). The established technique for detecting this activity involves using fluorescent Ca^{2+} indicators, which display changes in fluorescence intensity to indicate variations in Ca^{2+} levels. Parameters such as Ca^{2+} signal amplitude, area under the curve, and transient frequency can provide insights into concentration-response relationships for different stimuli (49). Analyzing these types of metrics helps to interpret Ca^{2+} responses across different experimental conditions and organisms, allowing to find the difference and similarities in cellular response. In the following section, we outline the metrics that can be extracted from multiple transients using CalciumInsights and how they could be interpreted in a biological context (Fig 1).

Fig 1. Ca^{2+} transients metrics: Representative metrics that can be extracted from the Ca^{2+} transients using CalciumInsights: Peak, Rise Rate, Amplitude, Full Width at Half-Maximum (FWHM), Prominence, Full Width at Half-Prominence (FWHP), Peak Occurrence Time, Time Onset, Transient Occurrence Time, Peak Rise Time, Baseline, Area Under the Curve (AUC), Number of Peaks, and Frequency.

<u>Peak</u>

The peak is the maximum fluorescence ratio for the duration of a transient (50).

<u>Amplitude or Peak Height</u>

The peak height is the difference between the baseline and peak, i.e., resting state and the amplitude of the transient respectively, (50,51) constructed using the following formula:

$$amplitude = peak\ Ca^{2+}\ level - baseline\ Ca^{2+}.$$ (Error! Bookmark not defined.)

Full Width at Half-Maximum (FWHM)

Response duration is the time between half amplitude on the ascent and descent of the transient (48).

Prominence

Prominence refers to the distinctiveness of a peak in a Ca^{2+} signal. It measures how much a specific peak stands out from the surrounding fluctuations in Ca^{2+} concentration. Prominence considers both the height of the peak and its relative position compared to a neighboring transient event characterized by a temporary increase in Ca^{2+} concentration within a cell (47).

Full Width at Half-Prominence (FWHP)

The full width at half-prominence represents the width of a signal, typically a peak or an event, at the level where its prominence is equal to half of its maximum prominence. It measures the width of a feature in a signal at the point where its intensity has decreased by half from the peak value (47). This parameter is also referred to as transient duration 50 (TD50) (52).

Peak Occurrence Time

The peak occurrence time refers to the specific moment or point in time when a Ca^{2+} signal reaches its maximum amplitude or concentration during a transient event. This time provides information about when this increase reaches its zenith (53).

Time Onset

Time onset refers to the moment when the increase in Ca^{2+} concentration within a cell begins. It marks the initiation or starting point of the transient event (48,54).

Transient Occurrence Time

Transient occurrence time refers to the specific moment or time a transient event occurs. It signifies the initiation or onset of the transient event and is typically measured from the beginning of the recording or a specific reference point.

Peak Rise Time

Peak rise time is the time it takes a Ca^{2+} transient to peak from the baseline (55,56).

Baseline

The baseline is the fluorescence at the beginning of the transient (50); i.e., the initial level of Ca^{2+} before a signal occurs (56).

Area Under the Curve (AUC)

The area under the curve is a value obtained through numerical integration to quantify the amount of Ca^{2+} over a specific period (57).

Number of Peaks

The number of peaks denotes the number of Ca^{2+} transients or spikes in a signal. A peak is a pattern of abrupt change within the dataset, characterized by an ascent followed by a descent. This definition is implemented by analyzing the sign changes in the differences between adjacent data points in the dataset (58–60).

Frequency

Frequency represents how often an event repeats over time. It refers to the oscillatory feature of a Ca^{2+} signal, i.e., the number of peaks of Ca^{2+} divided by the total time of the signal (61).

Rise Rate

This metric is the maximum rate of change (first derivative) of the fluorescence during the transient (see inset in Fig 1). This expression can be used to quantify how quickly the average frequency is increasing in relation to time (62), and the maximal velocity is usually reported (63).

The specifications of how each parameter was calculated and implemented are provided in the user manual.

Filters

CalciumInsights provides filters that allow to fine-tune the peak detection algorithm. This helps the user to identify peaks that are relevant for biological system.

Minimum Peak Height

The minimum peak height is a predefined threshold that the transients must exceed to be considered significant. This threshold is established depending on the specific characteristics of the data and the analytical goals, and it helps distinguish relevant peaks from background noise or minor fluctuations.

Minimum Peak Distance

The minimum peak distance refers to the smallest allowable separation between consecutive transients. Specifying a minimum peak distance helps ensure that only distinct transients are considered. This parameter is essential for preventing the detection of closely spaced peaks as one single peak and avoiding false positives in peak identification.

Peak Ascent

The peak ascent refers to the minimum number of discrete time steps before reaching a peak (64).

Peak Descent

The peak descent refers to the minimum number of discrete time steps after the peak (64).

Minimum Width Threshold $FWHP_{min}$

The minimum width threshold introduces a minimum width parameter that allows the user to set a threshold for the minimum allowable FWHP.

Experimental Datasets

To assess the efficiency of CalciumInsights for analyzing Ca^{2+} traces collected from different organisms, we applied its workflow to extract all the metrics defined previously (Fig 1) from several sets of biological data. Data were derived from Ca^{2+} signal measurements in *Danio rerio, Arabidopsis thaliana,* and *Drosophila melanogaster,* and the experimental protocols can be found in the following section.

Danio rerio

The zebrafish experiment was conducted following internationally accepted standards. The Animal Care and Use Protocol was approved by The Purdue Animal Care and Use Committee (PACUC), adhering to the Guidelines for the Use of Zebrafish in the NIH Intramural Research Program (protocol number: 1401001018). Larvae were produced from *Tg(krt4:GAL4, UAS:GCaMP6f)* in a mixed AB/TL background that expresses the GCaMP6f (65) Ca^{2+} indicator in the outer epithelial layer (66). The larvae were raised in an E3 medium at 28°C until 3 days post fertilization (dpf) for experiments (67). Larvae were treated with 100 nM of MemGlow™ 560 live membrane dye (Cytoskeleton, Cat #: MG02-10) for 20 min, anesthetized with 164 mg/L tricaine methanesulfonate (MedChemExpress, Cat. No.: HY-W011777), then immobilized in a custom-designed microfluidics device (68). The caudal tail fin fold was centered in the frame and imaged for 60 min at 1 frame per min on a Nikon A1Rmp with a Plan Fluor 20x 0.5 NA air objective on green (488-nm excitation, 510-nm emission) and red (561-nm excitation, 610-nm emission) channels, with a Z-stack spanning the full thickness of the fin. At the 60-min mark, imaging was paused, and latrunculin A (Cayman, Item No. 10010630) was added to achieve a final concentration of 200 µM. Imaging was then resumed for an additional 90 min. Fiji (19) was used to extract Ca^{2+} trace values from a subset of visible epidermal cells. Individual cells were segmented by hand utilizing the polyline ROI selection tool, followed by measurement of mean intensity within the ROI. Each ROI was manually shifted throughout the video to account for sample movement within the frame. Fluorescence changes were calculated as $(F-F_0)/F_0$, where F_0 is the lowest mean fluorescence intensity demonstrated by that particular cell over the entire imaging period (Fig 5A, D).

Arabidopsis thaliana

Arabidopsis thaliana (Columbia-0 ecotype) expressing R-GECO1, and intensiometric Ca^{2+} reporter (69), was used in this work. Arabidopsis seeds were surface sterilized and stratified at 4°C for 3 days on half-strength Murashige and Skoog (MS) medium supplemented with 1% sucrose and 0.8% agar. Plants were grown under long-day lighting conditions (16 h light/8 h dark) at 21°C for 7 days. Before imagining, cotyledons were excised from seedlings

and soaked in water overnight. The cotyledons were then mounted in custom-made chambers for bottom imaging as described previously with modifications (69). Cotyledons were placed in water at the bottom of the chamber with a cotton pad on top as a spacer, and then thin agar blocks were placed over the cotton to stabilize it. Samples were allowed to rest in the chamber for at least 40 min before imaging.
Epidermal cells from the apical region of the abaxial side of cotyledons were imaged for R-GECO1 fluorescence. Spinning-disk confocal microscopy (SDCM) was performed using a Yokogawa scanner unit CSU-X1-A1 mounted on an Olympus IX-83 microscope, equipped with a 20X 0.4–numerical aperture (NA) Plan C Achromat Dry objective (Olympus) and an Andor iXon Ultra 897BV EMCCD camera (Andor Technology). R-GECO1 fluorescence was excited with the 561-nm laser line, and emission was collected through a 610/37-nm filter. Time-lapse image series were collected at a single focal plane with 5-s intervals for 481 frames (40 min). Image acquisition was paused after the 120th frame. A two-fold concentration of the immunogenic peptide flg22 (GenScript) was added to the imaging chamber to achieve a final concentration of 1 µM (Fig 5B, E). Fiji (19) was used to extract Ca^{2+} fluorescence intensity values from individual epidermal cells. The puzzle piece-shaped epidermal cells were manually traced using the freehand line tool, and the line selections with a pixel width of 2 were used as ROIs to extract the raw intensity values. The fluorescence changes were calculated as $(F-F_0)/F_0$ or $\Delta F/F_0$, where F_0 is the average intensity of a stable period in the first 120 frames after subtracting the background.

Drosophila melanogaster

Fly stocks and culture were grown and maintained at 25 °C with a twelve-hour light cycle. The recombined Ca^{2+} sensor line developed at the Zartman Lab (using the Bloomington Drosophila Stock Center fly lines #2575 and #42747) has the genotype nubbin-GAL4, UAS-GCaMP6f/CyO. Flies were grown in a 15:5 ratio of female to male and staged for 6 hours to ensure the collection of wandering third instar larvae. Dissected wing imaginal discs were cultured with organ culture media with 1 mM Yoda1 or insulin to stimulate Ca^{2+} dynamics. Dissected discs were mounted in polyethylene terephthalate laminate (PETL) microfluidic devices where cultured media was introduced for long-term imaging (70). The devices were then imaged on a confocal microscope, and culture media was introduced using syringe pumps (Harvard Apparatus). The organ culture media consists of Grace's insect medium supplemented with Bis-Tris, fetal bovine serum, penicillin and streptomycin, and 20-Hydroxyecdysone (71). Initial time points were taken as a control in Grace's media without any drug perturbation.

Imaging was acquired using a Nikon Eclipse Ti confocal microscope with a Yokogawa spinning disc. Image data were collected on an iXonEM+cooled CCD camera (Andor Technology, South Windsor, CT) using MetaMorph v7.7.9 software (Molecular Devices, Sunnyvale, CA). Discs were imaged throughout the entire depth with a step size of 1 µm, depending on sample thickness, with a 40x or 60x oil objective with 200-ms exposure time, 50 nW, and 488-nm laser exposure at 44% laser power. The imaging was performed from the apical to the basal surface so that periodical cells were imaged first, followed by the

columnar cells of the wing disc. Post-processing with Fiji (19) was used to adjust the brightness and contrast of the imaging data and create video files of raw data (Fig 5C, F). Fiji was used to extract Ca^{2+} fluorescence intensity values from individual epithelial cells of the wing disc. Each spiking cell was manually traced using the rounded square tool as ROIs to extract the raw intensity values. The fluorescence changes were calculated for all 360 frames using the Plot Z-axis Profile tool in Fiji.

Designing CalciumInsights

R is an open-source software and a widely used programming language for statistical computation and data analysis (72). Shiny is a framework for creating dynamic and interactive web applications that fully harnesses all features of R (40). Compared to other programming languages, it stands out for its significant contribution to developing numerous web applications (72). We used RStudio (1.3.1093) and R (v4.0.3) to develop a web application. This application has been modularized using the Golem package (39), enabling the bundling of two separate modules. Furthermore, it leverages the Shiny framework to provide a GUI. The first module aims to enhance Ca^{2+} signals by removing noise using the Loess algorithm. The second module focuses on data analysis, providing a comprehensive characterization of Ca^{2+} signals. We utilized several R functions including findPeaks (64), integrate (73), ggplot (74), and Loess (75). Additionally, we incorporated custom functions, such as Rise Rate, Full Width at Half-Maximum (FWHM), Prominence, FWHP, Time Onset, Transient Occurrence Time, Peak Rise Time, Baseline, and Frequency, which were designed to calculate specific metrics from Ca^{2+} signals.

Results

General Overview of CalciumInsights

The workflow of CalciumInsights focuses on simplicity without sacrificing analytical power (Fig 2). Input files accepted by CalciumInsights, include .csv or. tsv file format and must contain Ca^{2+} fluorescence intensity values (in raw or normalized form) for various annotated ROIs and timestamps for each frame. These datasets can be extracted from open-source image processing packages like Fiji. The analysis appears in the graphical interface as soon as the document is input into the workflow (Fig 2). As shown in Fig 3, CalciumInsights provides a wide range of metrics extracted from the transients, facilitating real-time visualization of the study. Please refer to the user manual in the supplementary material for detailed instructions on effectively utilizing the graphical interface.

Users can adjust the "span" parameter (also known as the "Smoothing Control" parameter) in CalciumInsights between values of 0 and 1 (Fig 3). For further information, please refer to the Methods section. CalciumInsights enables users to customize the analysis for any biological system they investigate by providing different filters.

Fig 2. CalciumInsights workflow. The input requires a data frame encoded in a .csv or .tsv file. Then, the user enters the iterative process of fine-tuning filters. After fine-tuning, metrics and visualizations can be downloaded for further analysis. See also Fig. S1 for a more in-depth representation.

Users can define and construct additional modules to add their metrics and visualizations to CalciumInsights using Golem. This enables the expansion of the tool's current features and increases the options for customized analysis. Please refer to the user manual in the Supplementary Material for a more detailed description of defining various metrics and visualizations and how the tool integrates them.

Fig 3. Menu option to fine-tune parameters and results of estimated metrics. The panel on the left provides options to import example data (I), import user's data (II), select the ROI to analyze (III), define Smoothness Control (IV), and adjust the filters for the Find Peaks function (V), define the baseline that represents the signal (VI), plot the area under the curve with the smoothed signal (VII) and plot the raw signal together with the smoothed signal (VIII). The controls define the page on the right. A summary of the ROI's metrics can be selected (IX) or the peaks of a specific ROI (X). In the latter, the metrics can be seen as the numerical values (XI) or in a plot (XII). A table is provided with the numerical value of the metrics of each transient (XIII) and the complete trace (XIV).

Linear regression is conducted to fully exploit the Loess smoothing, with the smoothed fluorescence data after passing through the Loess filter serving as the dependent variable and the raw fluorescence data as the independent variable (Fig 4). This example illustrates how well the original raw data computed using the least square approach fits into the smoothed data model.

Fig 4. Linear regression between raw data and smoothed data. A linear regression between raw and smoothed datasets is generated to assess the smoothing via Loess. A better fit is indicated when R^2 is closer to 1.

Testing

We assessed CalciumInsights using Ca^{2+} recordings obtained from the tissues and organisms depicted in Fig 5, which include the outer epithelial layer of the resting embryonic tail fin of *Danio rerio*, three days post-fertilization (Fig 5A), the epidermal cells from the cotyledon of *Arabidopsis thaliana* (Fig 5B), and an ex vivo wing imaginal disc from a third instar larva of *Drosophila melanogaster* (Fig 5C). Individual cells or regions exhibiting significant Ca^{2+} dynamics were identified as ROIs and manually annotated using Fiji (Fig 5).

Fig 5. Calcium imaging of different biological systems and manual segmentation. A) Ca^{2+} signal in the outer epithelial layer of the resting *Danio rerio* tail fin at 3 days post fertilization (dpf) after the application of 200 µM latrunculin A. B) Ca^{2+} imaging of the cotyledon epidermal cells of *Arabidopsis thaliana* expressing R-GECO1 after application of 1 µM of the immunogenic peptide, flg22. C) Ca^{2+} signal in an *ex vivo* wing imaginal disc from a third instar larva of *Drosophila melanogaster* after applying 1 mM Yoda1 or insulin. D) the outer epithelial layer of the resting *Danio rerio* tail fin at 3 dpf, E) the cotyledon epidermal cells of *Arabidopsis thaliana*, and F) an *ex vivo* wing imaginal disc from a third instar larva of *Drosophila melanogaster*. Image created using BioRender.com.

Representative Ca^{2+} traces from the dataset are visualized using datasets available in Supplementary File S1, corresponding to the embryonic tail fin of the zebrafish, the epidermal cells from the cotyledon of *Arabidopsis thaliana*, and the ex vivo wing imaginal discs from a third instar larva of *Drosophila melanogaster*, respectively (Fig 6A, 7A, 8A). Box plots represent the distribution of the datasets, including the median, quartiles, and

potential outliers. For each case, the application filters were carefully adjusted (Fig 6B, 7B, 8B).

Fig 6. Calcium traces and metrics of the epithelial layer of the resting *Danio rerio* embryonic tail fin. A) Representative Ca^{2+} Traces (one per trial to show the shape), B) Number of Peaks, C) Peak Occurrence Time, D) FWHP, E) FWHM, F) Frequency, G) Time Onset, H) Transient Occurrence Time, I) Baseline, J) Peak Rise Time, K) AUC, L) Amplitude, M) Rise Rate, and N) Prominence. Smoothness Control: 0.05, Peak Height (min): 500, FWHP (min): 40, Prominence (min): 70, and Baseline: Min. The number of ROIs per biological replicate is 5, and data from 3 biological replicates (S1, S2, S3) are shown.

Fig 7. Calcium traces and metrics for the cotyledon epidermal cells from *Arabidopsis thaliana*. A) Representative Ca^{2+} Traces (one per biological replicate to show the shape), B) Number of Peaks, C) Peak Occurrence Time, D) FWHP, E) FWHM, F) Frequency, G) Time Onset, H) Transient Occurrence Time, I) Baseline, J) Peak Rise Time, K) AUC, L) Amplitude, M) Rise Rate, and N) Prominence. Smoothness Control: 0.05, Peak Height (min): 0.5, FWHP (min): 40, Prominence (min): 0.5, and baseline: standard definition. The number of ROIs per biological replicate (cotyledon) is equal to 5, and there are 3 biological replicates (S1, S2, S3).

Fig 8. Calcium transients of *ex vivo* wing imaginal discs from a third instar larva of *Drosophila melanogaster*. A) Representative Ca^{2+} Traces (one per trial to show the shape), B) Number of Peaks, C) Peak Occurrence Time, D) FWHP, E) FWHM, F) Frequency, G) Time Onset, H) Transient Occurrence Time, I) Baseline, J) Peak Rise Time, K) AUC, L) Amplitude, M) Rise Rate, and N) Prominence. Smoothness Control: 0.04, Peak Height (min): 17000, Peak Ascend: 3, Peak Descend: 3, Min Peak Distance: 36, and baseline: standard definition. The number of ROIs per biological replicate is equal to 10, and 3 biological replicates (S1, S2, S3).

Discussion

Given that Ca^{2+} dynamics influence cell fate, organ development, and response to the environment, its intricate role in cellular and tissue processes has long been appreciated (76,77). This study addresses the critical need for a comprehensive post-processing tool to analyze Ca^{2+} traces in cells, tissues, or organs and quantifiably characterize Ca^{2+} dynamics under normal, pathological, or inducible conditions. There is a lack of available post-processing tools for this purpose, which prompted the creation of new solutions (22–26). In this work, we present a new tool for post-processing analysis of Ca^{2+} dynamics in tissues and individual cells (Fig 1, 2). Among the many features of this program is a set of crucial metrics specifically tailored for Ca^{2+} imaging data analysis. Due to its flexibility and adaptability, it can be utilized across various biological systems and conditions. Additionally, its open-source architecture enables users to incorporate additional metrics and analyses to enhance its capabilities (30–33,35,36). The flexibility provided by this tool ensures its ability to adapt to changing research needs. By emphasizing user-friendliness, expandability, and ease of use for researchers without programming experience and fostering a collaborative environment within the scientific community, CalciumInsights aims to empower researchers from diverse backgrounds by expanding access to Ca^{2+} analysis tools.

The way researchers present Ca^{2+} signaling data can vary according to their objectives and audience. For example, if the primary aim is to illustrate patterns or changes over time, researchers may choose to display Ca^{2+} traces alongside corresponding statistical analyses. Alternatively, parameters such as the proportion of cells exhibiting detectable signals, signal amplitude, and other relevant metrics can be emphasized for a more thorough

characterization of cellular or tissue responses to stimulation. This study presents changes in Ca^{2+} activity over time, along with multiple parameters from three different biological systems analyzed using CalciumInsights, offering a comprehensive characterization of cellular and tissue responses to stimulation (Fig 6-8). Subsequently, these metrics can facilitate establishing dose-response relationships for various stimuli.

In this study, Loess and Savitsky-Golay (Defined in the Supporting Information) smoothing were the two main techniques examined. Loess employs a non-parametric method to fit distinct polynomials to various data segments to capture intricate, nonlinear relationships. Unlike Savitzky-Golay, Loess does not assume that there is an underlying structure. Due to its weighted scheme, deliberately reducing the impact of isolated data points during the smoothing phase, Loess is also less susceptible to outliers. The ease of implementation also contributes to the popularity of Loess, with several software packages offering user-friendly tools for applying this method to different datasets (78). In this regard, Loess had a specific advantage over Savitzky-Golay, as the latter has the limitation of only smoothing or computing a derivative for a subset of the analyzed data; specifically, the data points at the beginning and end of a trace are not smoothed but instead removed. The explanation lies in the window's symmetrical requirement around the smoothing points. The polynomial used for regression around each point depends on a minimum window size and consumes too many points, rendering it impractical for most evaluated metrics (79). The only exception is the Rise Rate, which refers to the maximum derivative within the interval when a transient begins to rise until it reaches its corresponding peak. Savitzky-Golay was used to compute this specific metric after the original smoothing.

We evaluated whether the Loess smoothing technique impacts the calculation of the metrics by comparing the values extracted from the same trace with and without smoothing (Fig S5). The results showed no statistically significant differences (p-value > 0.05). In cases where the p-value was closer to 0.05, specifically for Peak Rise Time and Full Width at Half Peak (FWHP), these differences can be attributed to the metrics' dependence on the derivative of the trace (Table S1). Calculating this derivative using numerical methods is highly sensitive to noise, making these metrics more susceptible to noise-related variability than others.

The versatility and adaptability of CalciumInsights were demonstrated when applied to three different model organisms exhibiting Ca^{2+} dynamics (Fig 5). Our tool effectively deciphered the parameters of Ca^{2+} signals in various cellular structures, including the multi-edged polygonal cells of the zebrafish caudal fin, the puzzle-piece-shaped epidermal cells from a plant cotyledon, and the epithelial cells of the fruit fly (Fig 5). Various images taken at different time intervals for these biological samples were successfully analyzed, ranging from 54 to 361 frames with intervals of 5 to 60 seconds (SI Metadata). This highlights CalciumInsights' ability to measure and calculate a wide range of parameter values accurately. For instance, these three specimens' AUC values varied from 2000 to $4x10^6$ (Fig 6–8). Overall, our tool demonstrates the ability to analyze cellular Ca^{2+} signaling parameters regardless of the biological model of origin, even across different kingdoms.

The current version of CalciumInsights requires the user to generate time trace data from ROIs using a separate tool. This capability allows the analysis of various cell and tissue types, regardless of origin. The flexibility and extensibility of CalciumInsights also support the implementation of additional modules or metrics, enabling users to incorporate methods specific to their system.

Further, programming expertise is required to integrate new metrics into the platform, which may limit accessibility for non-computational bench researchers. Striking a balance between advanced functionality and accessibility is a challenge in the continuous development of these technologies. To this end, we provide an instruction manual for utilizing our tool and an application integration template for adding new features and modules (see Supporting Information).

Optimization methods can be integrated to adjust the range of the Loess smoothing technique in future updates of CalciumInsights, functional decomposition, Fourier Transform analysis, and analysis of individual transients. As an enhancement, statistical tests such as functional ANOVA can also be added to CalciumInsight's workflow to enable statistical comparisons among diverse samples subjected to different stimuli. Thanks to the software's flexibility, all these improvements are feasible, with the limitation that the enhancements must be made exclusively in the R language (80).

In conclusion, this study introduces a tool for Ca^{2+} signature analysis in tissues and cells, facilitating comparisons among different biological systems through a unique integrated analysis platform. By developing an accessible and adaptable tool capable of evolving with changing perspectives in Ca^{2+} signal analysis, CalciumInsights will continue shedding light on the role of Ca^{2+} in vital processes and enable biological research breakthroughs.

Author Contributions

Conceptualization: ABT, CJS, CEIB, DG, DU, DSG, GIMC, GR, JZ, MCR, MSM, NCPR, QD, SRCB, SGT, WZ. Data curation: DSG, GIMC, MSM, NCPR, SRCB, ST, WZ. Formal analysis: CJS, CEIB, DG, DU, DSG, GIMC, GR, JZ, MCR, MSM, NCPR, QD, SRCB, SGT, WZ. Funding acquisition: ABT, CJS, CEIB, DU, JZ, MCR, QD. Investigation: CJS, CEIB, DG, DU, DSG, GIMC, GR, JZ, MCR, MSM, NCPR, QD, SRCB, SGT, WZ. Methodology: CJS, CEIB, GZ, DSG, GIMC, GR, JZ, MCR, MSM, NCPR, QD, SRCB, SGT, WZ. Project administration: DSG, GIMC, NCPR, SRCB. Resources: ABT, CJS, CEIB, DU, GR, JZ, MCR, QD. Software: DSG, GIMC, NCPR, SRCB, WZ. Supervision: ABT, CJS, CEIB, DU, GR, JZ, MCR, QD. Validation: CJS, CEIB, DG, DU, DSG, GIMC, GR, JZ, MCR, MSM, NCPR, QD, SRCB, SGT, WZ. Visualization: DG, DSG, GIMC, GR, JZ, MCR, MSM, NCPR, QD, SRCB, SGT, WZ. Writing Original Draft preparation: DG, DSG, GIMC, MSM, NCPR, SRCB, ST, WZ. Writing review-editing: CJS, CEIB, DG, DU, DSG, GIMC, GR, JZ, MCR, MSM, NCPR, QD, SRCB, SGT, WZ.

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Consent for publication

All authors have reviewed the manuscript and approved the final draft for publication.

Resource availability

Lead contact: Further information and requests for data may be directed to and will be fulfilled by Mauricio Cabrera (mauricio.cabrera1@upr.edu)

Code: All codes used are publicly available in GitHub at https://github.com/AOG-Lab/CalciumInsights

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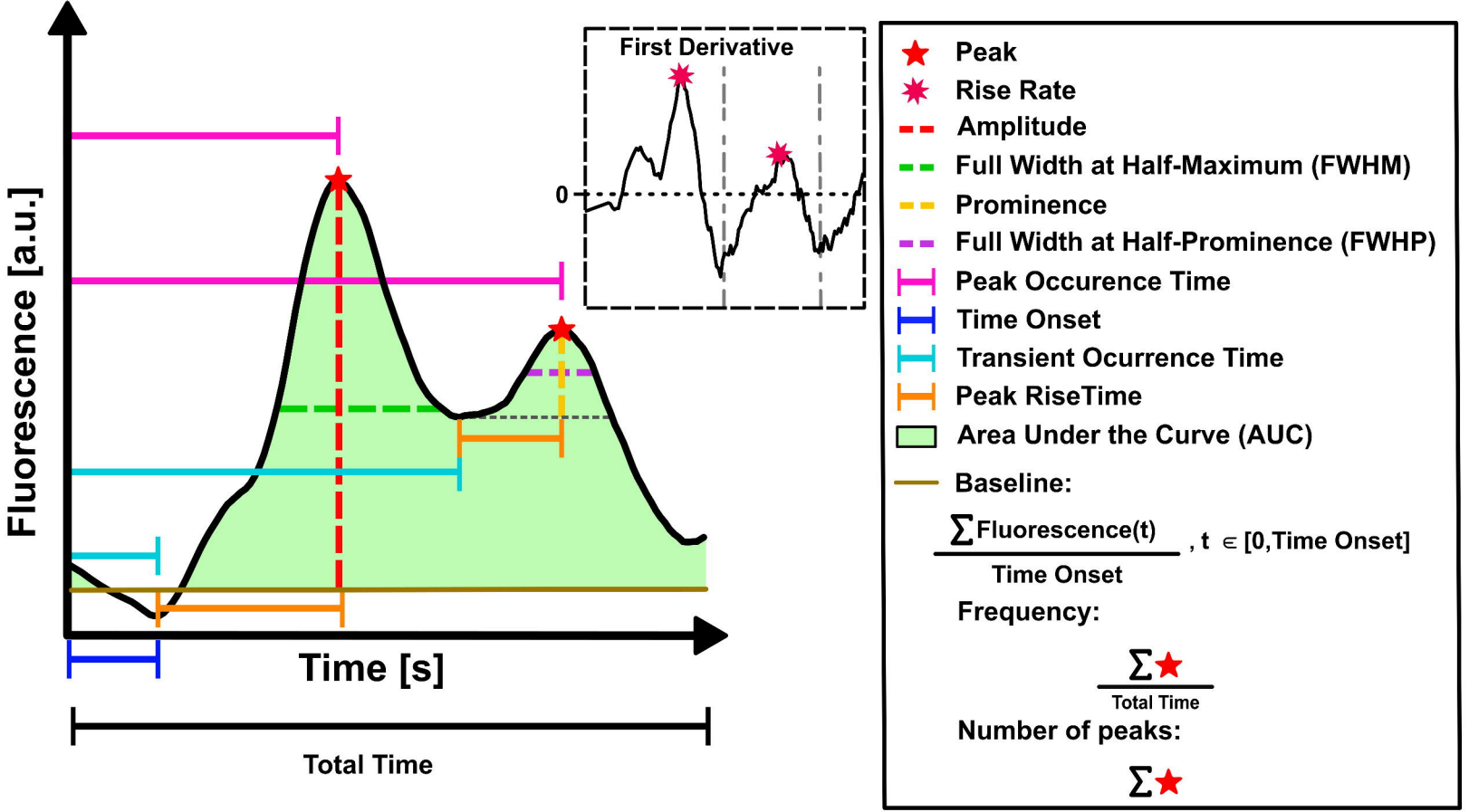
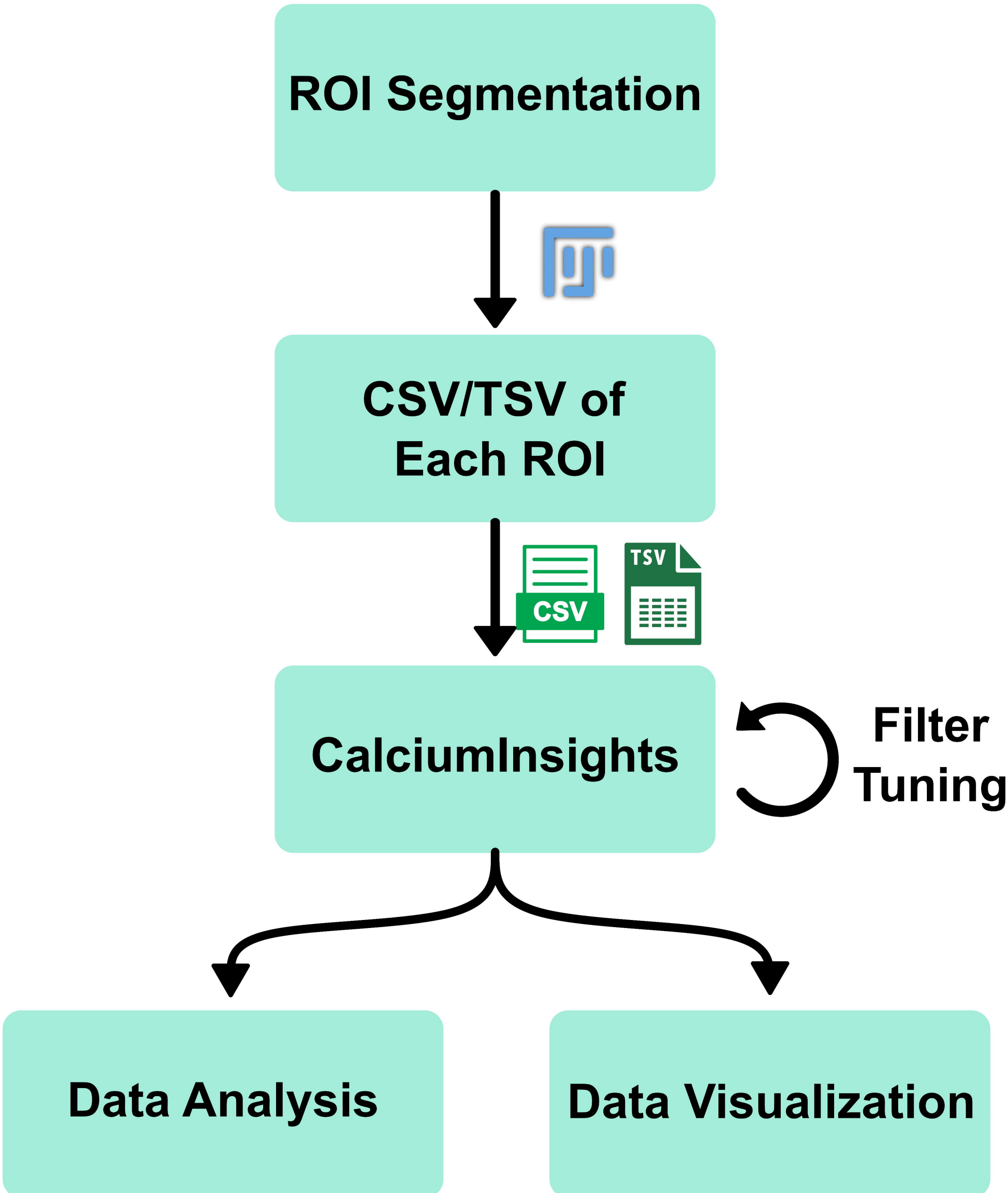

First Derivative
0
Fluorescence [a.u.]
Time [s]
Total Time
Peak
Rise Rate
Amplitude
Full Width at Half-Maximum (FWHM)
Prominence
Full Width at Half-Prominence (FWHP)
Peak Occurence Time
Time Onset
Transient Ocurrence Time
Peak RiseTime
Area Under the Curve (AUC)
Baseline:
$\frac{\sum \text{Fluorescence}(t)}{\text{Time Onset}}$, $t \in [0, \text{Time Onset}]$
Frequency:
$\frac{\sum \star}{\text{Total Time}}$
Number of peaks:
$\sum \star$


ROI Segmentation
CSV/TSV of Each ROI
CSV
TSV
CalciumInsights
Filter Tuning
Data Analysis
Data Visualization

CalciumInsights | Home | Descriptive Analysis | Inference

I) **Example Data**

Help

- (●) Yes
- ( ) No

II) **Dataset**

Browse... No file selected

III) **Region of Interest (ROI):**

1

IV) **Smoothness Control:**

0.05

V) **Find Peaks Function Arguments**

1.Peak Height (min): 0

2.Peak Ascent: 1

3.Peak Descent 1

4.Min Peak Distance: 1

5.FWHP (min) 0

6.Prominence (min) 0

VI) **Baseline:**

Reference Level 0

VII) **Area Under the Curve (AUC):**

- ( ) No
- (●) Yes

VIII) **Raw Data**

- (●) No
- ( ) Yes

IX) SummaryData | X) Peaks

XI) Metrics | XII) Metric plots

Show 10 entries | Search:

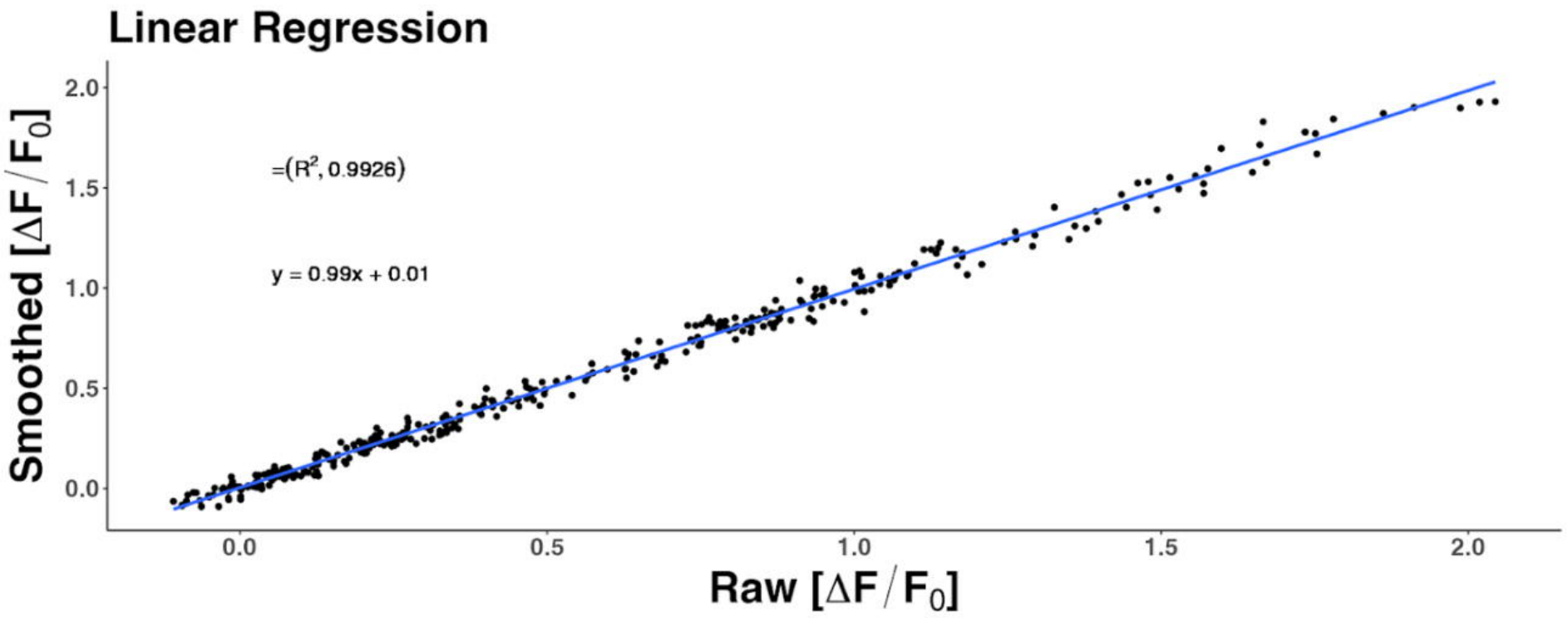
XIII) **Transient Metrics**

	Amplitude	Peak_Occurence_Time	Prominence	FWHP	FWHM	Peak_Rise_Time	Transient_Ocurrence_Time	Rise_Rate
1	0.445	170	0.534	759.56	529.45	70	100	0.073
2	1.93	285	1.507	120.405	146.23	100	185	0.113
3	0.852	405	0.003	5	444.315	5	400	-0.002
4	1.245	540	0.441	69.49	397.648	90	450	0.042
5	1.56	780	1.313	82.491	94.96	100	680	0.116
6	0.225	950	0.055	248.593	1066.636	35	915	0.011
7	1.067	1040	0.844	100.769	117.876	85	955	0.069
8	0.802	1355	0.724	123.211	131.724	120	1235	0.061
9	0.835	1395	0.034	20.348	127.182	35	1360	0.007
10	0.247	1520	0.022	14.075	299.52	20	1500	0.006

Showing 1 to 10 of 14 entries | Previous 1 2 Next

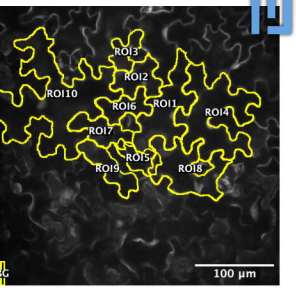
XIV) **Trace Metrics**

	Time_Onset	Frequency	Baseline	Number_of_Peaks	AUC	P_min	P_max
1	100	0.008	0	14	987.567	0	1770


Linear Regression
=(R², 0.9926)
y = 0.99x + 0.01
Smoothed [ΔF / F0]
Raw [ΔF / F0]
2.0
1.5
1.0
0.5
0.0
0.0
0.5
1.0
1.5
2.0

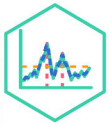
User Workflow

ROI Segmentation



CSV | TSV

Time	ROI1	ROI2
Time Measurements	Fluorescence Measurements	Fluorescence Measurements



Filter Tuning

Filters To apply
- Min Peak Amplitude
- Min Peak Distance
- Peak Ascent
- Peak Descent
- Min FWHM
- Max FWHM

Filtered Metrics

Metrics
- Number of Peaks
- Frequency
- Amplitude
- Prominence
- FWHM
- FWHP
- Peak Ocurrence Time
- Transient Ocurrence Time
- Peak Rise Time
- Time Onset
- Rise Rate
- Baseline
- AUC

Data Analysis

- ROI's Trace Analysis

Metrics for each Transient

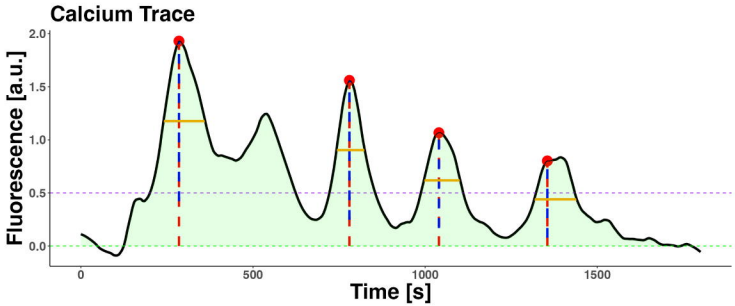
Transient	Metric 1	Metric 2
1		
2		
...		

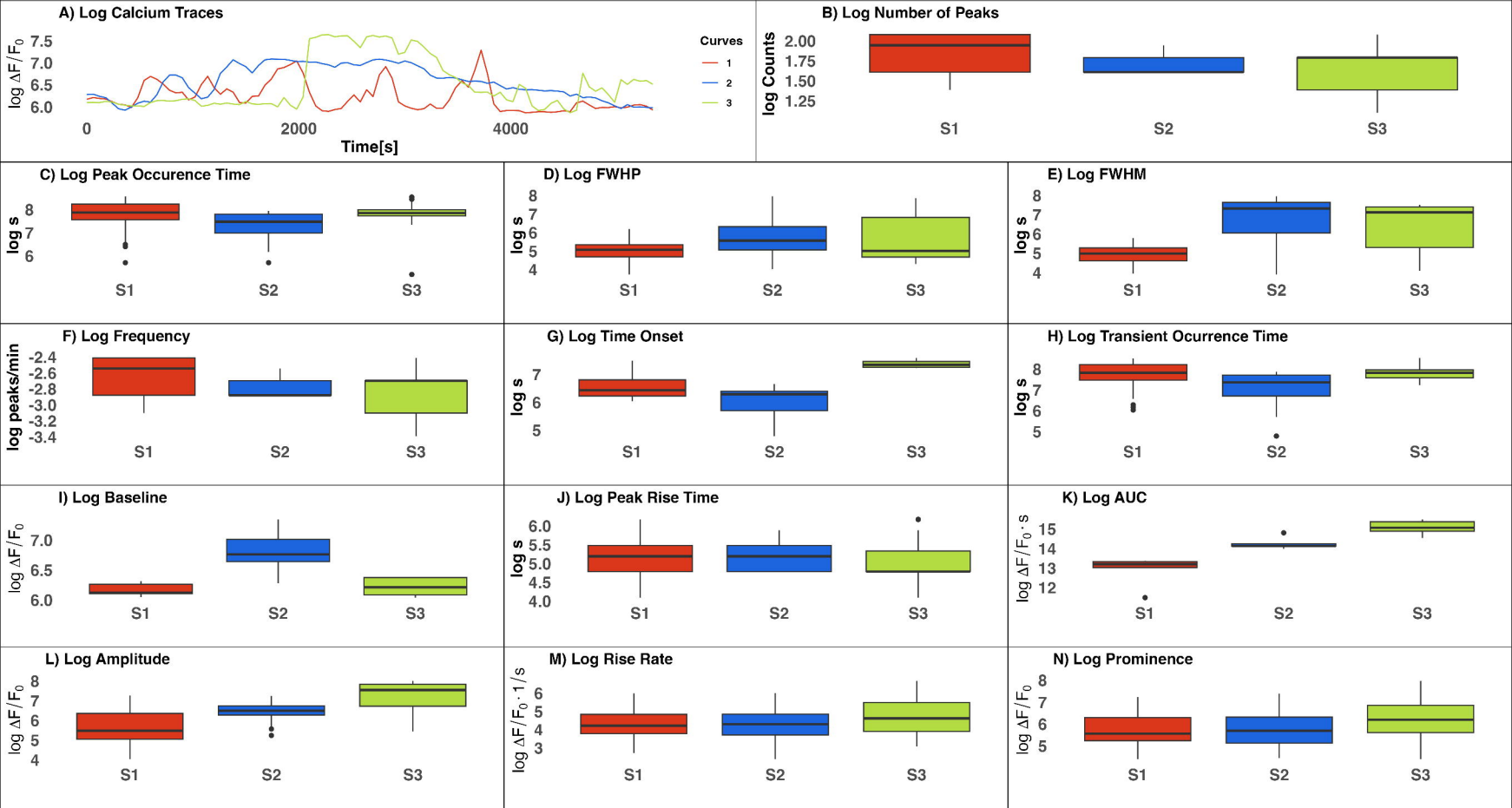
Trace Metrics

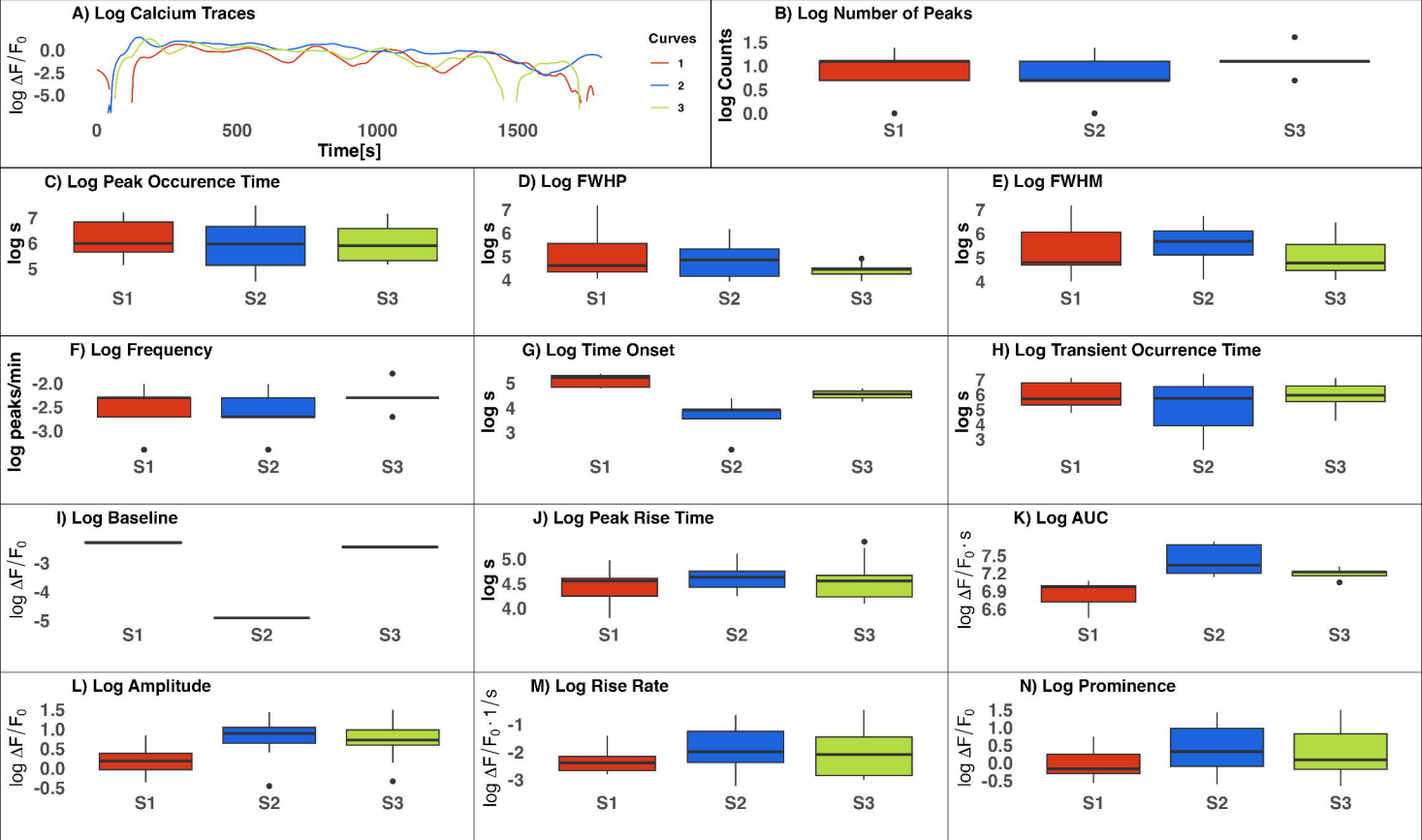
Trace	Metric 1	Metric 2
1		

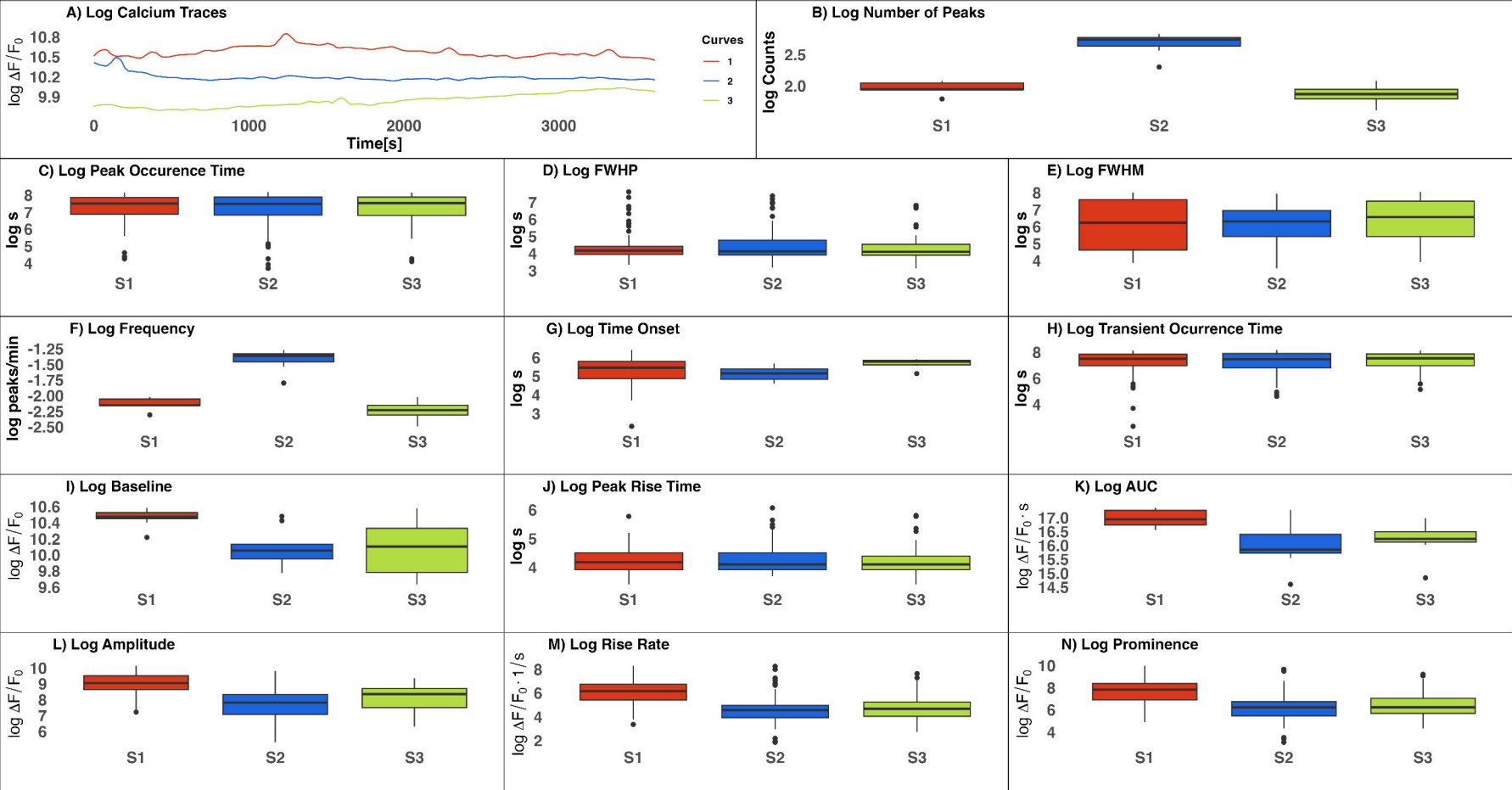
- CSV's ROI comparisons

Data Visualization

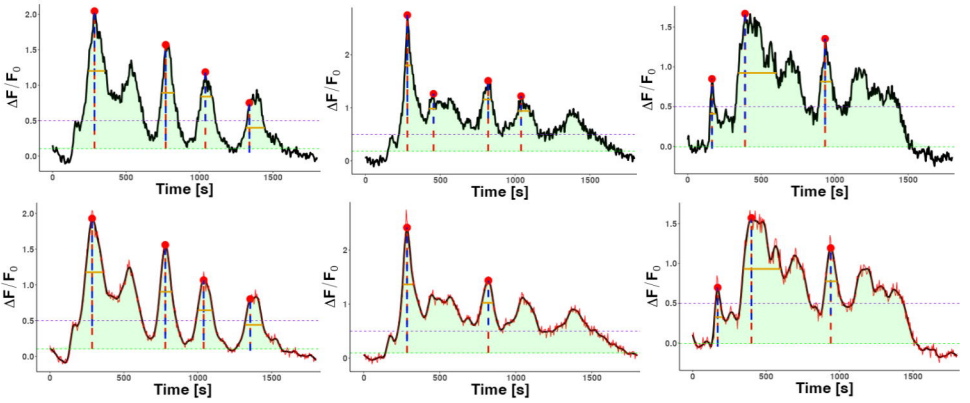



A) Log Calcium Traces
log ΔF/F0
7.5
7.0
6.5
6.0
0
2000
4000
Time[s]
Curves
1
2
3
B) Log Number of Peaks
log Counts
2.00
1.75
1.50
1.25
S1
S2
S3
C) Log Peak Occurence Time
log s
8
7
6
S1
S2
S3
D) Log FWHP
log s
8
7
6
5
4
S1
S2
S3
E) Log FWHM
log s
8
7
6
5
4
S1
S2
S3
F) Log Frequency
log peaks/min
-2.4
-2.6
-2.8
-3.0
-3.2
-3.4
S1
S2
S3
G) Log Time Onset
log s
7
6
5
S1
S2
S3
H) Log Transient Ocurrence Time
log s
8
7
6
5
S1
S2
S3
I) Log Baseline
log ΔF/F0
7.0
6.5
6.0
S1
S2
S3
J) Log Peak Rise Time
log s
6.0
5.5
5.0
4.5
4.0
S1
S2
S3
K) Log AUC
log ΔF/F0 · s
15
14
13
12
S1
S2
S3
L) Log Amplitude
log ΔF/F0
8
7
6
5
4
S1
S2
S3
M) Log Rise Rate
log ΔF/F0 · 1/s
6
5
4
3
S1
S2
S3
N) Log Prominence
log ΔF/F0
8
7
6
5
S1
S2
S3

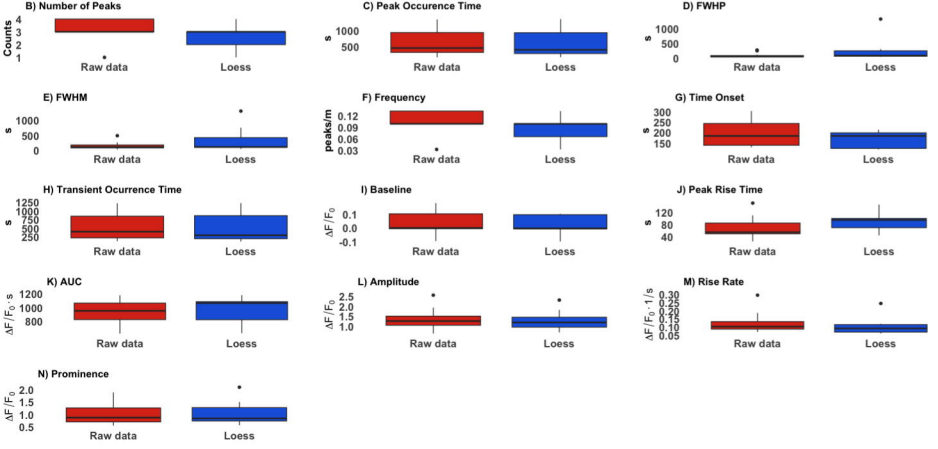

A) Log Calcium Traces
log ΔF/F0
0.0
-2.5
-5.0
0
500
1000
1500
Time[s]
Curves
1
2
3
B) Log Number of Peaks
log Counts
1.5
1.0
0.5
0.0
S1
S2
S3
C) Log Peak Occurence Time
log s
7
6
5
S1
S2
S3
D) Log FWHP
log s
7
6
5
4
S1
S2
S3
E) Log FWHM
log s
7
6
5
4
S1
S2
S3
F) Log Frequency
log peaks/min
-2.0
-2.5
-3.0
S1
S2
S3
G) Log Time Onset
log s
5
4
3
S1
S2
S3
H) Log Transient Ocurrence Time
log s
7
6
5
4
3
S1
S2
S3
I) Log Baseline
log ΔF/F0
-3
-4
-5
S1
S2
S3
J) Log Peak Rise Time
log s
5.0
4.5
4.0
S1
S2
S3
K) Log AUC
log ΔF/F0 · s
7.5
7.2
6.9
6.6
S1
S2
S3
L) Log Amplitude
log ΔF/F0
1.5
1.0
0.5
0.0
-0.5
S1
S2
S3
M) Log Rise Rate
log ΔF/F0 · 1/s
-1
-2
-3
S1
S2
S3
N) Log Prominence
log ΔF/F0
1.5
1.0
0.5
0.0
-0.5
S1
S2
S3

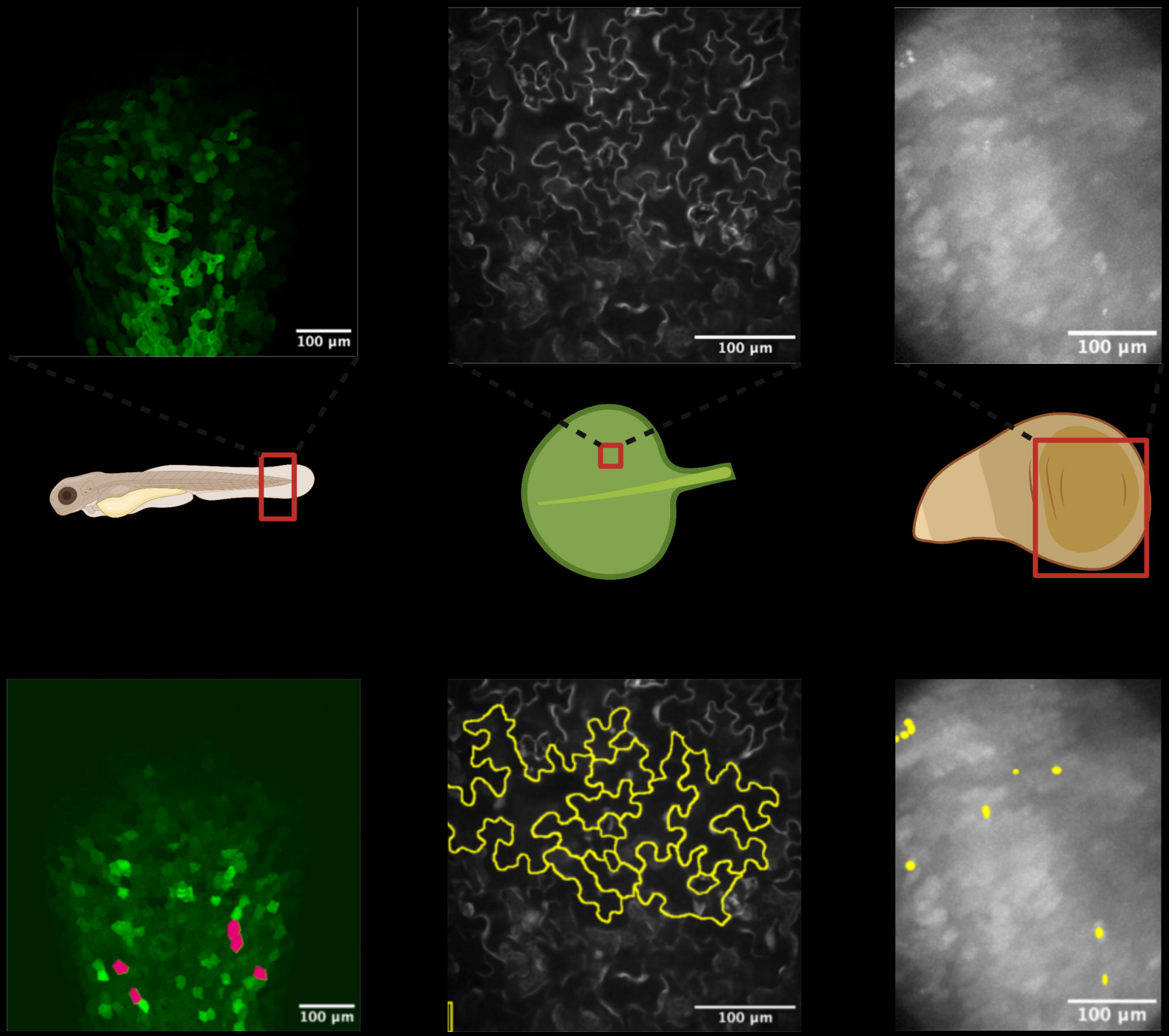

A) Log Calcium Traces
log ΔF/F0
10.8
10.5
10.2
9.9
0
1000
2000
3000
Time[s]
Curves
1
2
3
B) Log Number of Peaks
log Counts
2.5
2.0
S1
S2
S3
C) Log Peak Occurence Time
log s
8
7
6
5
4
S1
S2
S3
D) Log FWHP
log s
7
6
5
4
3
S1
S2
S3
E) Log FWHM
log s
8
7
6
5
4
S1
S2
S3
F) Log Frequency
log peaks/min
-1.25
-1.50
-1.75
-2.00
-2.25
-2.50
S1
S2
S3
G) Log Time Onset
log s
6
5
4
3
S1
S2
S3
H) Log Transient Ocurrence Time
log s
8
6
4
S1
S2
S3
I) Log Baseline
log ΔF/F0
10.6
10.4
10.2
10.0
9.8
9.6
S1
S2
S3
J) Log Peak Rise Time
log s
6
5
4
S1
S2
S3
K) Log AUC
log ΔF/F0 · s
17.0
16.5
16.0
15.5
15.0
14.5
S1
S2
S3
L) Log Amplitude
log ΔF/F0
10
9
8
7
6
S1
S2
S3
M) Log Rise Rate
log ΔF/F0 · 1/s
8
6
4
2
S1
S2
S3
N) Log Prominence
log ΔF/F0
10
8
6
4
S1
S2
S3

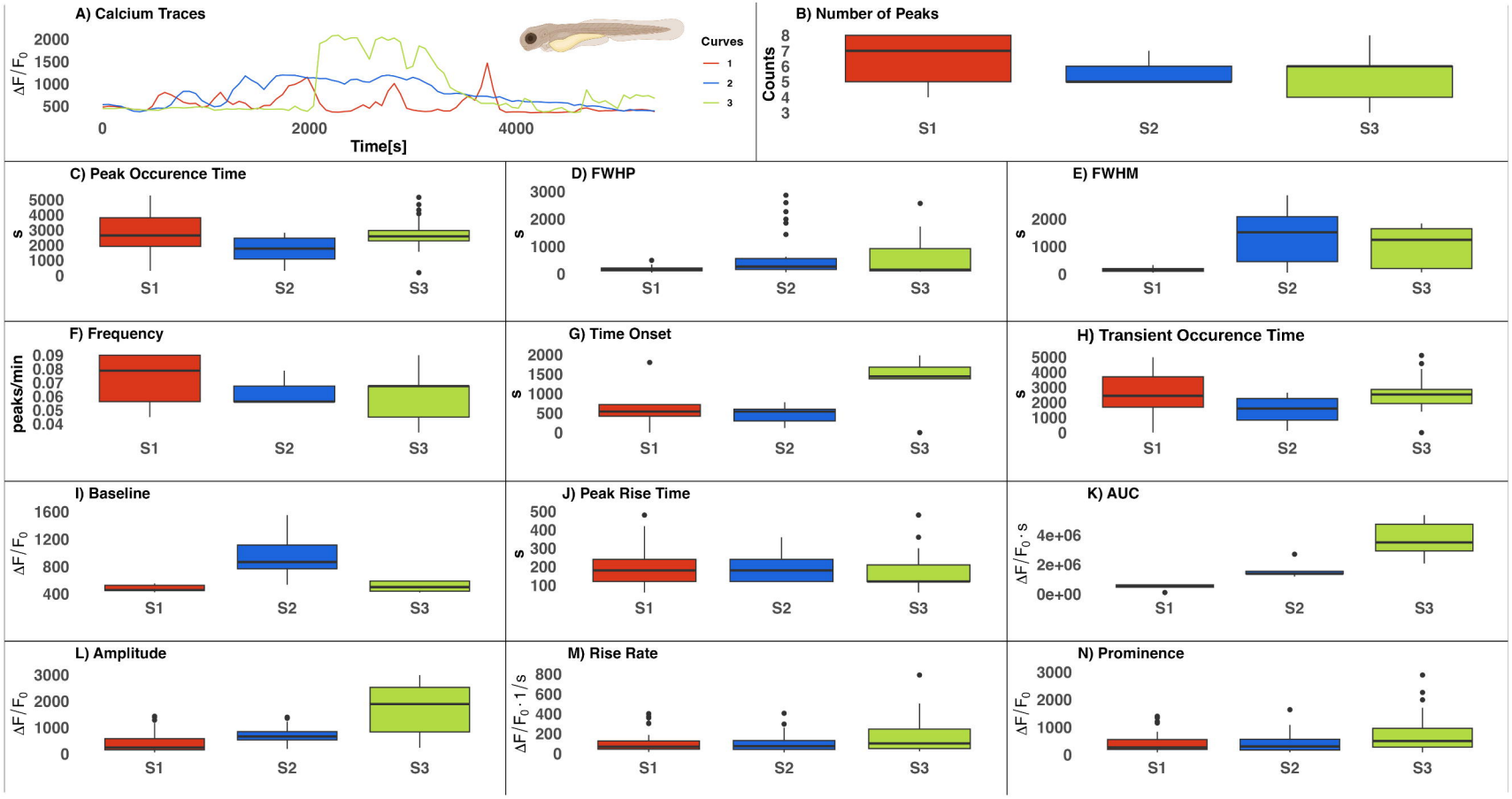
A) Calcium Traces

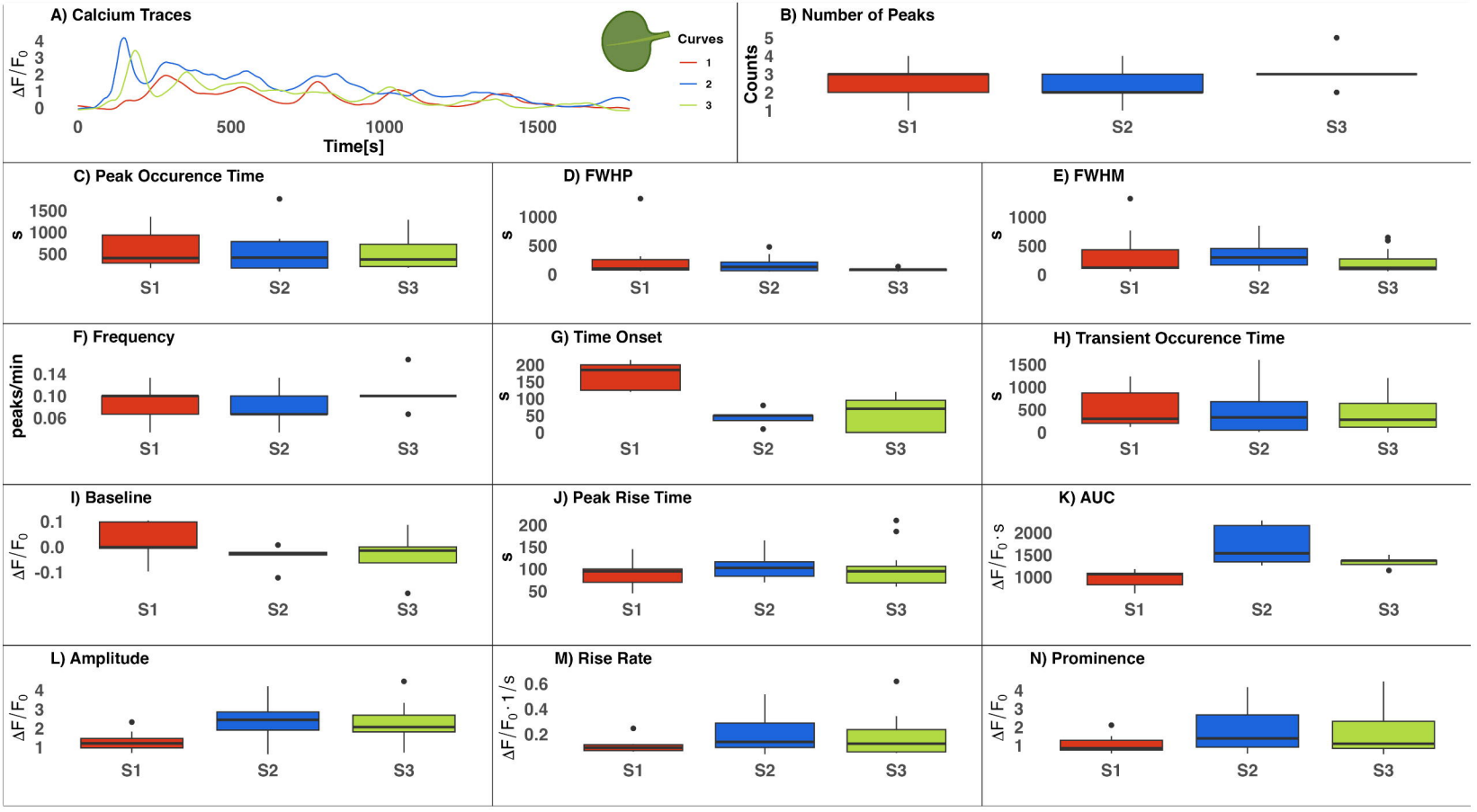


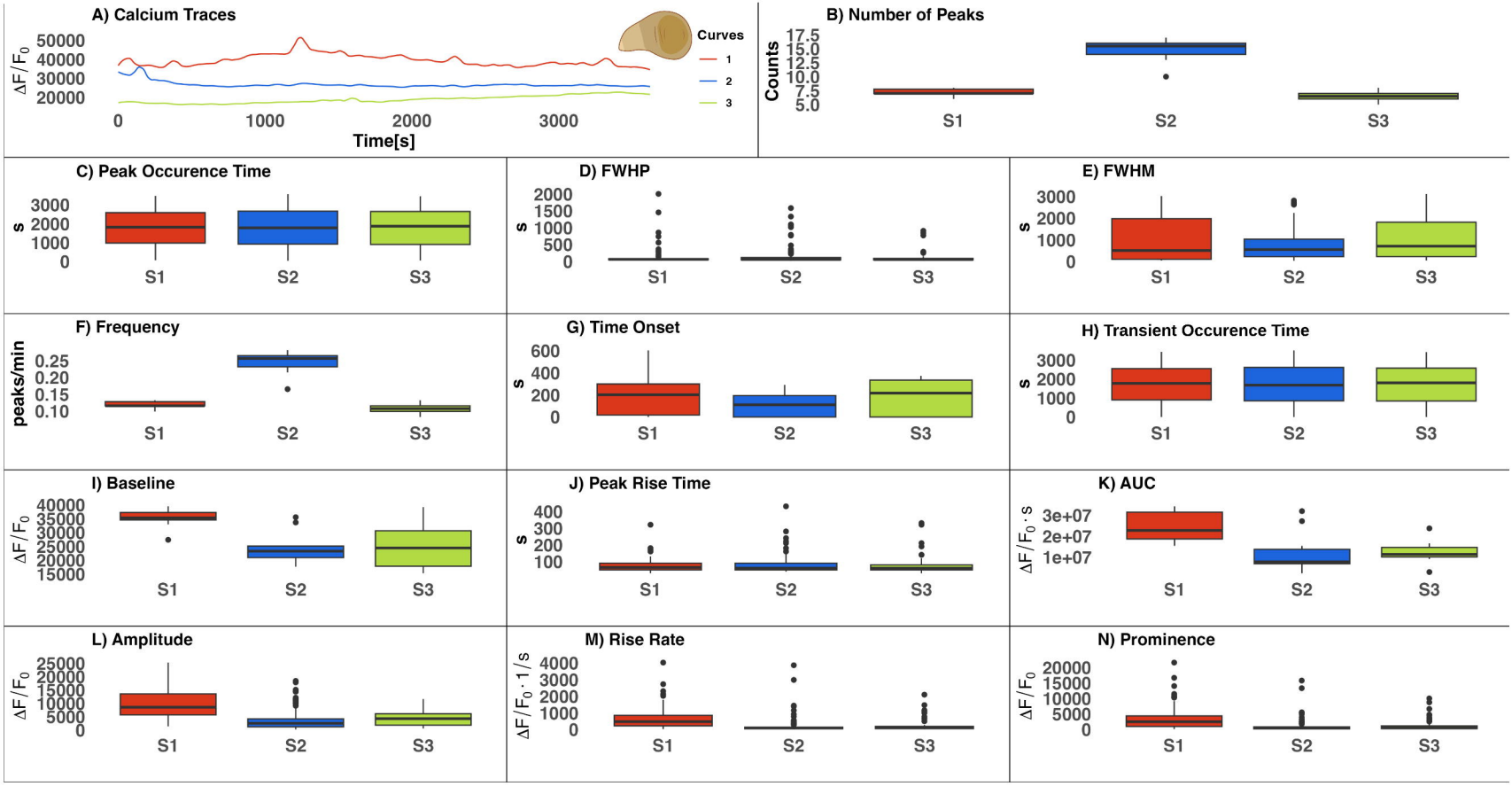
B) Comparison of Calcium Metrics: Raw vs. Smoothed (Loess) Data in *Arabidopsis thaliana*




100 µm
100 µm
100 µm
100 µm
100 µm
100 µm


A) Calcium Traces
ΔF/F0
2000
1500
1000
500
0
2000
4000
Time[s]
Curves
1
2
3
B) Number of Peaks
Counts
8
7
6
5
4
3
S1
S2
S3
C) Peak Occurence Time
s
5000
4000
3000
2000
1000
0
S1
S2
S3
D) FWHP
s
3000
2000
1000
0
S1
S2
S3
E) FWHM
s
2000
1000
0
S1
S2
S3
F) Frequency
peaks/min
0.09
0.08
0.07
0.06
0.05
0.04
S1
S2
S3
G) Time Onset
s
2000
1500
1000
500
0
S1
S2
S3
H) Transient Occurence Time
s
5000
4000
3000
2000
1000
0
S1
S2
S3
I) Baseline
ΔF/F0
1600
1200
800
400
S1
S2
S3
J) Peak Rise Time
s
500
400
300
200
100
S1
S2
S3
K) AUC
ΔF/F0 · s
4e+06
2e+06
0e+00
S1
S2
S3
L) Amplitude
ΔF/F0
3000
2000
1000
0
S1
S2
S3
M) Rise Rate
ΔF/F0 · 1/s
800
600
400
200
0
S1
S2
S3
N) Prominence
ΔF/F0
3000
2000
1000
0
S1
S2
S3


A) Calcium Traces
ΔF/F0
Time[s]
Curves
1
2
3
B) Number of Peaks
Counts
S1
S2
S3
C) Peak Occurence Time
s
D) FWHP
E) FWHM
F) Frequency
peaks/min
G) Time Onset
H) Transient Occurence Time
I) Baseline
J) Peak Rise Time
K) AUC
ΔF/F0·s
L) Amplitude
M) Rise Rate
ΔF/F0·1/s
N) Prominence


A) Calcium Traces
ΔF/F0
50000
40000
30000
20000
0
1000
2000
3000
Time[s]
Curves
1
2
3
B) Number of Peaks
Counts
17.5
15.0
12.5
10.0
7.5
5.0
S1
S2
S3
C) Peak Occurence Time
s
3000
2000
1000
0
D) FWHP
2000
1500
1000
500
E) FWHM
F) Frequency
peaks/min
0.25
0.20
0.15
0.10
G) Time Onset
600
400
200
H) Transient Occurence Time
I) Baseline
40000
35000
30000
25000
20000
15000
J) Peak Rise Time
400
300
200
100
K) AUC
ΔF/F0 · s
3e+07
2e+07
1e+07
L) Amplitude
25000
20000
15000
10000
5000
M) Rise Rate
ΔF/F0 · 1/s
4000
3000
2000
1000
N) Prominence