

# The endoplasmic reticulum in perisynaptic astrocytic processes: shape, distribution and effect on calcium activity

Audrey Denizot <sup>1\*</sup>, María Fernanda Veloz Castillo <sup>2</sup>, Pavel Puchenkov <sup>3</sup>,  
Corrado Cali <sup>4,5</sup>, Erik De Schutter <sup>1</sup>

<sup>1</sup> Okinawa Institute of Science and Technology, Computational Neuroscience Unit, Onna-Son, Japan

<sup>2</sup> Biological and Environmental Science and Engineering Division, King Abdullah University of Science and Technology, Thuwal, Saudi Arabia

<sup>3</sup> Okinawa Institute of Science and Technology, Scientific Computing and Data Analysis section, Research Support Division, Onna-Son, Japan

<sup>4</sup> Department of Neuroscience, University of Torino, Italy

<sup>5</sup> Neuroscience Institute Cavalieri Ottolenghi, Orbassano, Italy

\*To whom correspondence should be addressed; E-mail: [audrey.denizot3@oist.jp](mailto:audrey.denizot3@oist.jp).

1 **Astrocytes recently emerged as key regulators of information processing in the**  
2 **brain.  $\text{Ca}^{2+}$  signals in perisynaptic astrocytic processes (PAPs) notably allow**  
3 **astrocytes to fine-tune neurotransmission at so-called tripartite synapses. As**  
4 **most PAPs are below the diffraction limit, their content in  $\text{Ca}^{2+}$  stores and**  
5 **the contribution of the latter to astrocytic  $\text{Ca}^{2+}$  activity is unclear. Here, we**  
6 **reconstruct tripartite synapses in 3D from electron microscopy and find that**  
7 **75% of PAPs contain some endoplasmic reticulum (ER), a major astrocytic**  
8  **$\text{Ca}^{2+}$  store, displaying strikingly diverse geometrical properties. To investigate**  
9 **the role of such spatial properties, we implemented an algorithm that creates**  
10 **3D PAP meshes of various ER distributions and constant shape. Reaction-**  
11 **diffusion simulations in those meshes reveal that astrocyte activity is shaped by**  
12 **a complex interplay between the location of  $\text{Ca}^{2+}$  channels,  $\text{Ca}^{2+}$  buffering, ER**  
13 **shape and distribution. Overall, this study sheds new light into mechanisms**  
14 **regulating signal transmission in the brain.**

## 15 Introduction

16 Astrocytes, the most abundant glial cells of the central nervous system, are essential to nu-  
17 merous brain functions [74]. Notably, astrocytes are key modulators of neurotransmission at  
18 so-called tripartite synapses [4, 60]. A single astrocyte in the CA1 region of the mouse hip-  
19 pocampus is in contact with hundreds of thousands of synapses simultaneously, at perisynaptic  
20 astrocytic processes (PAPs) [14]. Around 75 % of cortical and 65 % of hippocampal synapses  
21 are contacted by an astrocytic process [78, 46]. This close contact between astrocytes and neu-  
22 rons allows astrocytes to control various synaptic functions, from glutamate uptake [37], and  
23 spillover [34, 9], to synapse homeostasis [57], stability [11], synaptogenesis [73], and neu-  
24 rotransmission [3, 60]. Those synaptic functions are associated with specific local molecular  
25 expression in PAPs [49, 28], which changes upon fear conditioning [49]. Importantly, the al-  
26 teration of the proximity of PAPs to hippocampal synapses of the CA1 region *in vivo* affects  
27 neuronal activity and cognitive performance [9]. Conversely, neuronal activity has been shown  
28 to induce the remodeling of synaptic coverage by PAPs in various brain regions, both *in vivo*  
29 and in acute slices [34, 51, 54, 11, 46, 56, 29, 76]. Together, those results illustrate that PAPs  
30 are preferential sites of neuron-astrocyte communication. Although the recent emergence of  
31 super-resolution techniques has provided key insights into the properties and functions of PAPs  
32 [33, 5], our understanding of PAP physiology and function in live tissue is hindered by their  
33 nanoscopic size [59, 1].

34  
35  $\text{Ca}^{2+}$  signals are commonly interpreted as a measure of astrocyte activity, notably in re-  
36 sponse to neurotransmitter release at synapses [75, 59, 61]. The recent advances in  $\text{Ca}^{2+}$  imag-  
37 ing approaches have improved the spatio-temporal resolution of  $\text{Ca}^{2+}$  signals in astrocytes [64,  
38 61]. Strikingly, it revealed that astrocytes in acute slices and *in vivo* exhibit spatially-restricted  
39  $\text{Ca}^{2+}$  signals, also referred to as hotspots or microdomains, stable over time and which activity  
40 varies under physiological conditions such as locomotion or sensory stimulation [44, 30, 2, 12,  
41 8, 66, 70, 69, 68, 63, 53, 25, 45]. Growing evidence supports that PAPs are preferential sites dis-  
42 playing spatially-restricted  $\text{Ca}^{2+}$  microdomains in response to neurotransmission [53, 25, 52, 7,  
43 44]. As a single astrocyte can contact hundreds of thousands of synapses simultaneously [14],  
44 such spatially-restricted  $\text{Ca}^{2+}$  microdomains might enable the astrocyte to finely tune synaptic  
45 transmission at the single synapse level.

46  
47 mGluR activation on the astrocytic membrane following neurotransmission at glutamater-  
48 gic synapses results in  $\text{Ca}^{2+}$  transients mediated by  $G_q$  proteins and  $\text{Ca}^{2+}$  stores such as the  
49 endoplasmic reticulum (ER) [64], which can trigger the release of molecules that modulate neu-  
50 rotransmission, referred to as gliotransmitters [15, 48, 3, 60]. Most astrocytic  $\text{Ca}^{2+}$  signals are  
51 mediated by the Inositol 3-Phosphate ( $\text{IP}_3$ ) receptors on the membrane of the endoplasmic retic-  
52 ulum (ER) [62]. Because of their nanoscopic size, the  $\text{Ca}^{2+}$  pathways involved in microdomain  
53  $\text{Ca}^{2+}$  signals in PAPs are still unclear. Notably, the presence of ER in PAPs and its involvement  
54 in microdomain  $\text{Ca}^{2+}$  signals at synapses is highly debated. During the last decade, PAPs have

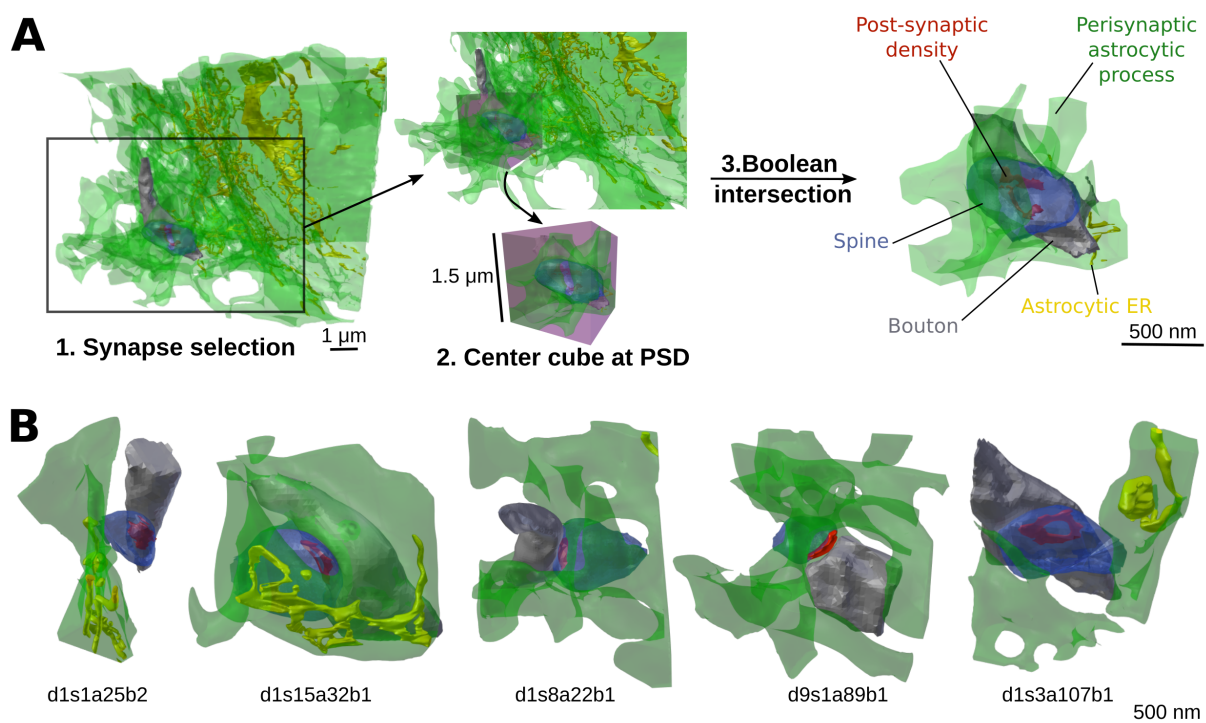
55 notably been regarded as devoid of ER, with a minimum distance between the synapse and the  
56 closest astrocytic ER  $> 0.5 \mu\text{m}$  [55, 59]. In contrast, inhibiting ER-mediated  $\text{Ca}^{2+}$  signaling in  
57 fine processes results in a decreased number of  $\text{Ca}^{2+}$  domains [2] and a decreased  $\text{Ca}^{2+}$  peak  
58 frequency [2, 7, 63]. Furthermore, some astrocytic ER has been detected near synapses in other  
59 EM studies [1, 10]. Yet, the geometrical properties of the ER in PAPs and its distribution re-  
60 main poorly characterized, but could have a strong impact on neuron-astrocyte communication  
61 at tripartite synapses.

62  
63 Here, we use a  $220 \mu\text{m}^3$  hippocampal astrocytic volume from the CA1 stratum radiatum  
64 region (6 nm voxel resolution) [17], reconstructed from electron microscopy (EM), to create 46  
65 three dimensional meshes of tripartite synapses. Strikingly, we find that 75 % of PAPs contain  
66 some ER, which can be as close as 72 nm to the post-synaptic density (PSD). Analysis of the  
67 geometrical features of those meshes reveal the vast diversity of ER shapes and distributions  
68 within PAPs from a single cell. We then used a detailed stochastic reaction-diffusion model of  
69  $\text{Ca}^{2+}$  signals in PAPs to investigate the mechanistic link between the spatial features of the ER  
70 measured in the 3D meshes and the spatio-temporal properties of  $\text{Ca}^{2+}$  microdomain activity  
71 in PAPs. To be able to decipher the effect of ER distribution within the PAP independently  
72 from the effect of its shape, we developed an algorithm that automatically creates realistic 3D  
73 tetrahedral PAP meshes with various ER distributions from the realistic meshes reconstructed  
74 from EM. *In silico* experiments in those meshes reveal that the spatio-temporal properties of  
75  $\text{Ca}^{2+}$  signals in PAPs are tightly regulated by a complex interplay between the clustering of  
76  $\text{Ca}^{2+}$  channels, the ratio between ER surface area and PAP volume,  $\text{Ca}^{2+}$  buffering and ER  
77 spatial distribution. Together, this study provides new insights into the geometrical properties  
78 of hippocampal tripartite synapses and predicts mechanistic links between those features and  
79  $\text{Ca}^{2+}$  microdomain activity at tripartite synapses.

## 80 **Results**

### 81 **Quantification of the main geometrical properties of hippocampal tripar-** 82 **tite synapses**

83 To characterize the presence, shape and distribution of the endoplasmic reticulum (ER) in  
84 perisynaptic astrocytic processes (PAPs), we used a  $220 \mu\text{m}^3$  ( $7.07 \mu\text{m} \times 6.75 \mu\text{m} \times 4.75 \mu\text{m}$ )  
85 hippocampal astrocytic volume from the CA1 stratum radiatum region reconstructed from a  
86 perfectly isotropic EM stack (6 nm voxel resolution) [17]. Elements from the neuropil, i.e. bou-  
87 tons, dendritic spines and post-synaptic densities (PSDs), were also reconstructed. Following  
88 the workflow presented in Fig. 1A, forty four excitatory and two inhibitory tripartite synapse  
89 meshes were created, containing all elements belonging to the astrocyte and to the neuropil  
90 within a cube of  $1.5 \mu\text{m}$  edge length ( $3.375 \mu\text{m}^3$ ) centered at the center of mass of the PSD  
91 (Supplementary Movie 1). Five of those tripartite synapse meshes are displayed in Fig. 1B.



**Fig 1: Reconstruction of 46 tripartite synapse meshes from electron microscopy.** (A) Schematic presenting our tripartite synapse mesh creation workflow, here performed on synapse d10s1a2b1. 1. Synapses in contact with the  $220 \mu\text{m}^3$  astrocytic volume were selected one by one. 2. A cube of  $1.5 \mu\text{m}$  edge length ( $3.375 \mu\text{m}^3$ ) was created and centered at the center of mass of the post-synaptic density (PSD, red). 3. Boolean intersection between the neuronal and astrocytic objects and the cube resulted in the isolation of the elements of the tripartite synapse mesh: the perisynaptic astrocytic process (PAP, green), the astrocytic endoplasmic reticulum (ER, yellow), the bouton (grey) and the spine (blue). This workflow resulted in the creation of 44 excitatory and 2 inhibitory tripartite synapse meshes. (B) Images of five of the 3D tripartite synapse meshes created, d1s1a25b2, d1s15a32b1, d1s8a22b1, d9s1a89b1, d1s3a107b1, revealing their diverse geometrical properties.

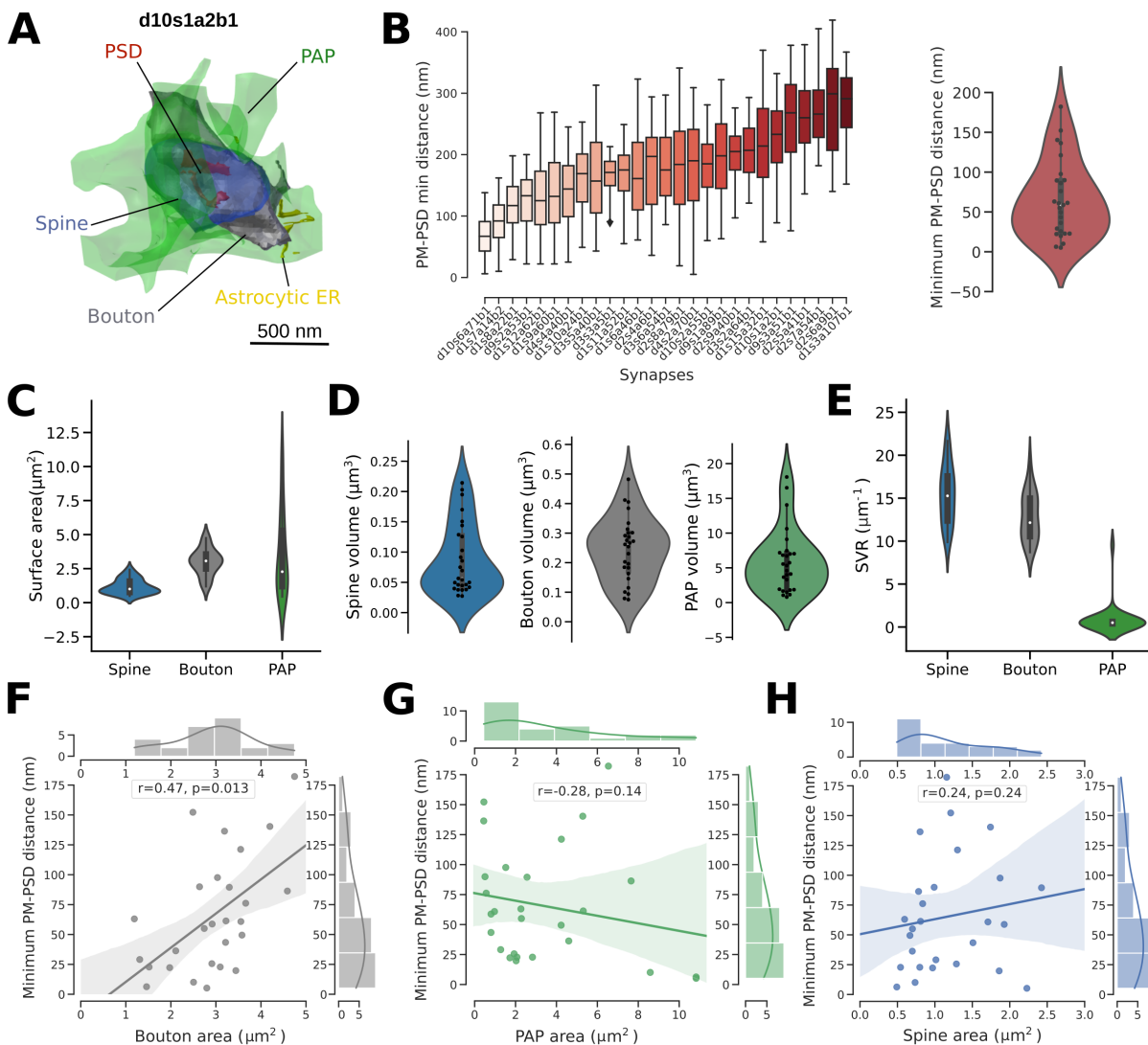
92 Among those meshes, seventeen were located at the borders of the  $220 \mu\text{m}^3$  astrocytic volume.  
93 They were thus omitted from data analysis as synaptic elements in those meshes could not be  
94 fully reconstructed. The volume, surface area and surface-volume ratio (SVR) of each synaptic  
95 element, i.e the PAP, astrocytic ER, spine and bouton, of the remaining twenty seven fully re-  
96 constructed excitatory tripartite synapses are presented in Fig. 2C-E and Supplementary Table  
97 S1. The minimum distance between each vertex on the membrane of the PAP and the center  
98 of mass of the PSD was measured in each of the twenty seven meshes (Fig. 2B), providing  
99 a quantification of the distribution of the astrocyte around the synapse. Our results highlight  
100 the diverse distances between PSDs and PAPs belonging to a single cell. In accordance with  
101 previous studies [46, 50, 55], PAP membrane vertices could be as close as 5 nm to the PSD,  
102 with an average distance between the PSD and the closest PAP vertex of 65 nm. Importantly,  
103 we found that PM-PSD distance is the shortest, i.e PAPs are the closest to the synapse, when  
104 bouton surface area is low (Fig. 2F,  $p=0.013$ ). PAP-PSD distance was not correlated to the  
105 surface area of the PAP (Fig. 2G,  $p=0.14$ ) or spine (Fig. 2H,  $p=0.24$ ).

## 106 **Presence and geometrical properties of the endoplasmic reticulum in perisy-** 107 **naptic astrocytic processes**

108 Because of the small size of most PAPs, the  $\text{Ca}^{2+}$  pathways that regulate astrocytic  $\text{Ca}^{2+}$  mi-  
109 crodomain activity at tripartite synapses remain to be uncovered. Notably, the presence of ER  
110 in PAPs is controversial [55, 1, 10, 41]. We have thus analyzed the presence and shape of the  
111 ER in the PAPs from the twenty seven fully reconstructed excitatory tripartite synapse meshes  
112 presented in Fig. 2.

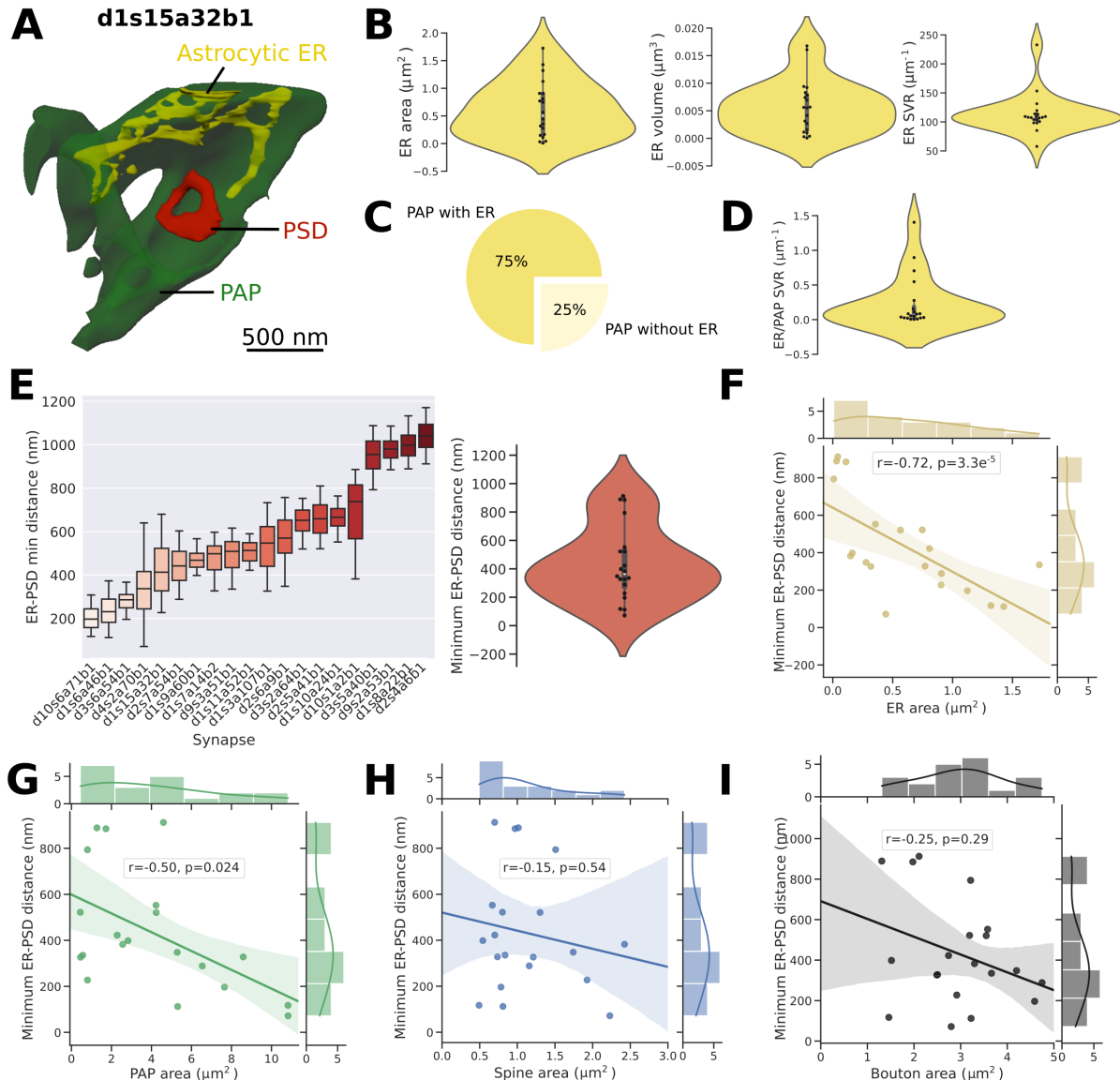
113 75% of PAPs contained some ER (Fig. 3C), which challenges the widespread belief that tri-  
114 partite synapses are devoid of astrocytic ER. ER surface area, volume and SVR were measured  
115 in ER-containing PAPs and highlight that ER shape is highly variable between PAPs from the  
116 same cell (Fig. 3B). Note that there was no significant difference between bouton, spine and  
117 PAP surface area, volume and SVR between synapses with vs without astrocytic ER (Supple-  
118 mentary Fig. S1). We further characterized the vicinity of the astrocytic ER to the synapse. To  
119 do so, we measured the distance between each vertex on the ER membrane to the center of mass  
120 of the PSD ( $n=20$ ). We found that ER-PSD distance varies drastically from synapse to synapse  
121 (Fig. 3E) and can be as little as 72 nm, far below the  $> 0.5 \mu\text{m}$  ER-PSD distance reported pre-  
122 viously [55, 59]. The closest ER vertex was on average 432 nm away from the center of mass  
123 of the PSD. Interestingly, the larger the surface area of the ER, the closer it was to the PSD (Fig.  
124 3F,  $p=0.013$ ). Astrocytic ER was closer to the PSD in PAPs with higher surface area (Fig. 3G,  
125  $p=0.024$ ). The minimum ER-PSD distance was not correlated to the surface area of the spine  
126 (Fig. 3H,  $p=0.54$ ) or bouton (Fig. 3I,  $p=0.29$ ). Overall, our results highlight that most PAPs  
127 contain some ER and that its shape is highly variable, which could have strong implications on  
128 ER-dependent  $\text{Ca}^{2+}$  signaling in PAPs resulting from synaptic transmission.

129



**Fig 2: Characterization of the geometrical properties of hippocampal tripartite synapses.**

(A) Image of a tripartite synapse mesh, d10s1a2b1, containing a bouton (grey), spine (blue), post-synaptic density (PSD, red), perisynaptic astrocytic process (PAP, green) and the astrocytic endoplasmic reticulum (ER, yellow). (B) Left: Boxplots presenting the distribution of the minimum distance between each vertex on the PAP membrane and the center of mass of the PSD, measured in the twenty seven excitatory tripartite synapse meshes fully reconstructed in this study. Right: Distribution of the minimum distance between the PAP and the PSD (n=27). (C-E) Violin plots presenting the distribution of spine, bouton and PAP surface area (C), volume (D) and surface-volume ratio (E). (F-H) Scatterplots presenting the variation of the minimum PAP-PSD distance as a function of bouton surface area (left), PAP surface area (middle) and spine surface area (right). Plots are presented with univariate kernel density estimation curves and a linear regression fit. Spearman correlation coefficients,  $r$ , and p-values,  $p$ , are displayed onto each regression plot, n=27.



**Fig 3: Presence and geometrical properties of the endoplasmic reticulum in perisynaptic astrocytic processes.** (A) Image of the d1s15a32b1 PSD (red) and the neighboring PAP (green), that contains ER (yellow). (B) Violin plots representing the distribution of ER surface area (left), volume (middle) and surface volume ratio (right) within PAPs,  $n=20$ . (C) Among the twenty seven fully reconstructed PAP meshes extracted, 75 % contained some ER. (D) Distribution of the ratio between the ER surface area and PAP volume ( $n=20$ ). (E) Quantitative analysis of the distance between the astrocytic ER and the neighboring PSD,  $n=20$ . (Left) Boxplots presenting the distribution of the distance of ER membrane vertices to the center of mass of the PSD in each PAP. (Right) Distribution of the minimum ER-PSD distance in PAPs,  $n=20$ . The lowest ER-PSD distance measured was 70nm (synapse d4s2a70b1).

Fig 3: (F-I) Scatterplots presenting the variation of the minimum ER-PSD distance as a function of ER surface area (F), PAP surface area (G), spine surface area (H) and bouton surface area (I),  $n=20$ . Plots are presented with univariate kernel density estimation curves and a linear regression fit. Spearman correlation coefficient,  $r$ , and  $p$ -value,  $p$ , are displayed onto each regression plot.

## 130 **Reaction-diffusion simulations reveal different spatio-temporal properties** 131 **of $\text{Ca}^{2+}$ signals in PAPs of the same cell**

132 PAPs are characterized by highly diverse sizes and shapes of the ER (Fig. 3), which could have  
133 strong implications on ER-mediated  $\text{Ca}^{2+}$  signals in PAPs. Because of their nanometric size,  
134 measuring  $\text{Ca}^{2+}$  activity and deciphering the involvement of ER-mediated signals in individual  
135 PAPs in live tissue is extremely challenging [61]. A better understanding of the mechanistic  
136 link between the geometrical properties of the ER and the spatio-temporal properties of  $\text{Ca}^{2+}$   
137 microdomain signals in PAPs is crucial, yet hard to test experimentally. Here, we use the PAP  
138 meshes presented in Fig. 3 together with a spatial stochastic model of  $\text{Ca}^{2+}$  signaling adapted  
139 from the model of Denizot and collaborators [20] to investigate the mechanistic link between  
140 ER shape and  $\text{Ca}^{2+}$  microdomain activity in PAPs.  $\text{Ca}^{2+}$  influx in the PAP cytosol in the model  
141 is mediated by Inositol 3-Phosphate ( $\text{IP}_3$ ) receptors on the membrane of the ER and by  $\text{Ca}^{2+}$   
142 channels at the plasma membrane,  $\text{Ch}_{\text{PM}}$ . The reactions modeled are presented in Fig. 4A and  
143 in the Methods section. Neuronal activity was simulated at  $t=1\text{s}$  by infusing 50  $\text{IP}_3$  molecules  
144 at the PM of the PAP. The implementation of this model with STEPS software [35] allows to  
145 perform simulations in tetrahedral meshes in 3 spatial dimensions, such as the ones created in  
146 this study. Representative Ca-GCaMP traces in a cylindrical mesh, corresponding to the con-  
147 centration of  $\text{Ca}^{2+}$  bound to  $\text{Ca}^{2+}$  indicators added to the cytosol of the model, display similar  
148 spatio-temporal characteristics to  $\text{Ca}^{2+}$  signals measured in organotypic hippocampal astrocytic  
149 cultures [21] (Fig. 4A, right).

150  
151 We performed simulations in six PAP meshes reconstructed from electron microscopy, char-  
152 acterized by various geometrical properties of the ER: d1s3a107b1, d1s8a22b1, d1s10a24b1,  
153 d2s6a9b1, d9s4a34b1 and d10s1a2b1 (Fig. 4B, Table 1). To do so, meshes were pre-processed  
154 to allow their use in reaction-diffusion simulations. The pre-processing workflow is described in  
155 Fig. 4C and in the Methods section, and produced 3D tetrahedral meshes from the 2D triangular  
156 meshes reconstructed from EM. Screenshots of simulations performed in two realistic tetrahe-  
157 dral PAP meshes are presented in Fig. 4B. Ca-GCaMP and free  $\text{Ca}^{2+}$  signals, in simulations  
158 with and without  $\text{Ca}^{2+}$  indicators in the cytosol, respectively, were measured in d1s3a107b1,  
159 d1s8a22b1, d1s10a24b1, d2s6a9b1, d9s4a34b1 and d10s1a2b1 PAP meshes. A simulation in  
160 PAP d9s4a34b1 is presented in Supplementary movie 2. Representative traces are displayed  
161 in Fig. 4E. Signals varied greatly depending on the mesh (Fig. 4F). Note that, in accordance  
162 with previous studies [20, 81], Ca-GCaMP and free  $\text{Ca}^{2+}$  signals displayed different spatio-



**Table 1: Characteristics of the 3D PAP meshes used in the reaction-diffusion simulations.**

$V_{\text{cyt}}$  is the cytosolic volume,  $S_{\text{PM}}$  is the plasma membrane surface area,  $S_{\text{ER}}$  is the ER surface area,  $\text{SVR}_{\text{ER/PAP}}$  is the ratio between the ER surface area and the cytosolic volume.  $ER_c$  is the number of ER vertices at ER-PM contact sites, i.e.  $\leq 20$  nm from the closest PM vertex.  $d1s15a32b1_{f0}$ ,  $d1s15a32b1_{f21}$ ,  $d1s15a32b1_{f64}$  and  $d1s15a32b1_{f250}$  refer to meshes from frames 0, 21, 64 and 250 of the  $d1s15a32b1$  PAP mesh presented in Fig. 8-9.

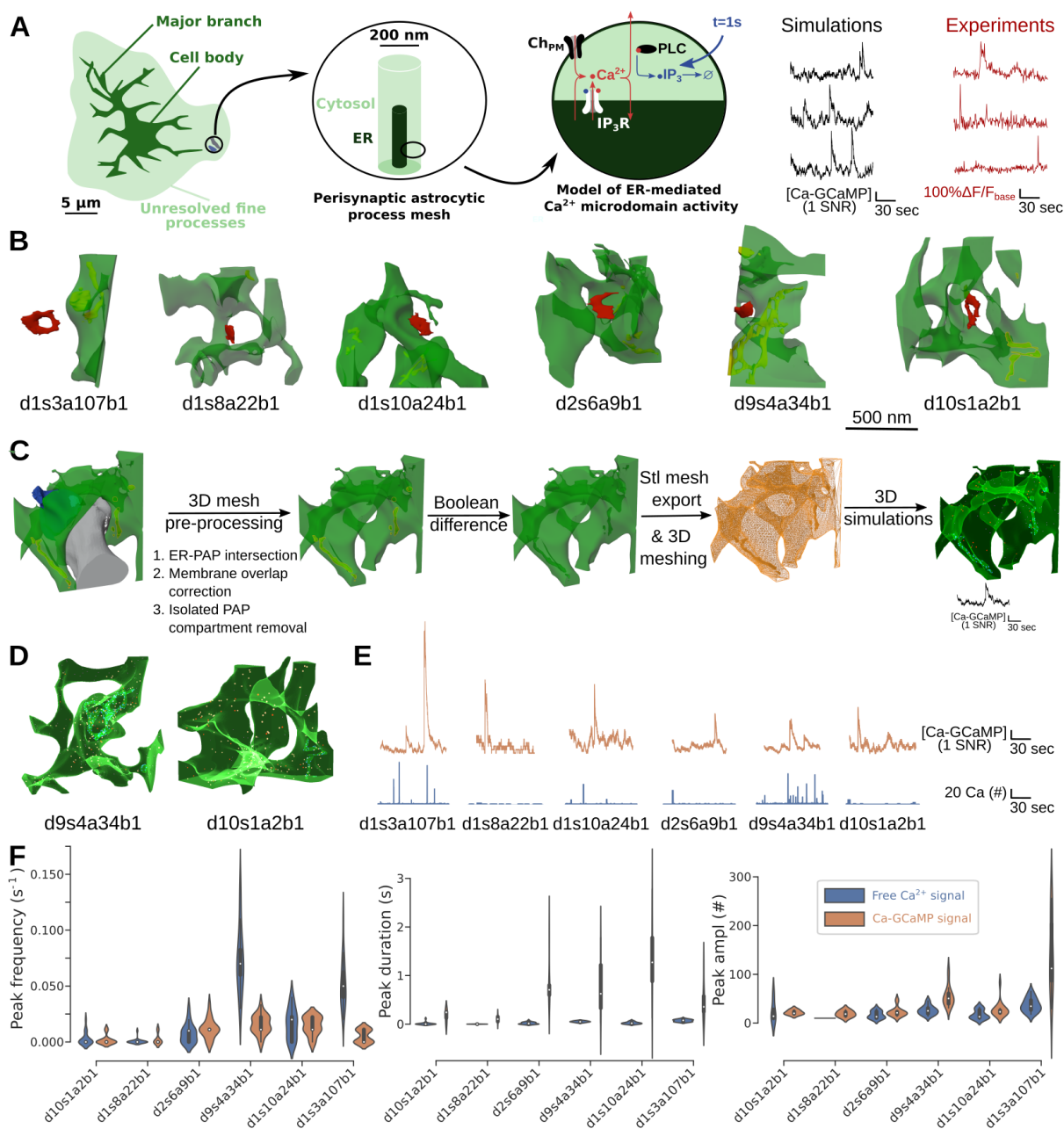
Geom	$V_{\text{cyt}} (\mu\text{m}^3)$	$S_{\text{PM}} (\mu\text{m}^2)$	$S_{\text{ER}} (\mu\text{m}^2)$	$\text{SVR}_{\text{ER/PAP}} (\mu\text{m}^{-1})$	$ER_c$
$d1s3a107b1$	0.112	2.00	0.315	2.81	183
$d1s8a22b1$	0.397	8.60	0.031	0.078	31
$d1s10a24b1$	0.331	5.80	0.344	1.04	0
$d2s6a9b1$	0.505	10.0	0.273	0.54	158
$d9s4a34b1$	0.410	7.05	0.807	1.97	20958
$d10s1a2b1$	0.531	10.0	0.136	0.26	3771
$d1s15a32b1_{f0}$	0.426	6.91	0.85	2.00	295
$d1s15a32b1_{f21}$	0.426	6.91	0.85	2.00	2337
$d1s15a32b1_{f64}$	0.426	6.91	0.85	2.00	1683
$d1s15a32b1_{f250}$	0.426	6.91	0.85	2.00	2408
$\text{PAP1}_v$	0.434	3.55	0.088	0.21	0
$\text{PAP1}_w$	0.432	3.55	0.428	0.99	0
$\text{PAP1}_x$	0.428	3.55	0.834	1.95	125
$\text{PAP1}_y$	0.423	3.55	1.27	3.00	0
$\text{PAP1}_z$	0.418	3.55	1.62	3.88	555

temporal properties (Fig. 4F). Those results suggest that the diverse geometrical features of PAPs and ER reported in this study (Fig. 2-3) strongly influence  $\text{Ca}^{2+}$  microdomain activity at tripartite synapses.

### The effect of IP<sub>3</sub>R clustering differs from PAP to PAP

IP<sub>3</sub>R channels are not randomly distributed on the surface of the ER and form stable clusters of various sizes [77, 72, 71, 67]. Numerous computational studies performed in 2 spatial dimensions have predicted that IP<sub>3</sub>R cluster size shapes  $\text{Ca}^{2+}$  activity (see [58] for a review). Whether this effect still holds in 3D, notably in complex shapes such as that of the PAPs reconstructed in this study, remains to be uncovered. We thus next simulated  $\text{Ca}^{2+}$  signaling in the 6 realistic PAP meshes presented in Fig. 4, reconstructed from EM,  $d1s3a197b1$ ,  $d1s8a22b1$ ,  $d1s10a24b1$ ,  $d2s6a9b1$ ,  $d9s4a34b1$  and  $d10s1a2b1$ , with various distributions of IP<sub>3</sub>Rs on the membrane of the ER.

Simulations of the model were performed with different IP<sub>3</sub>R cluster sizes  $\eta$ .  $\text{Ca}^{2+}$  channels were either randomly placed on the membrane of the PAP ( $\text{cocl}=0$ ) or clustered onto the



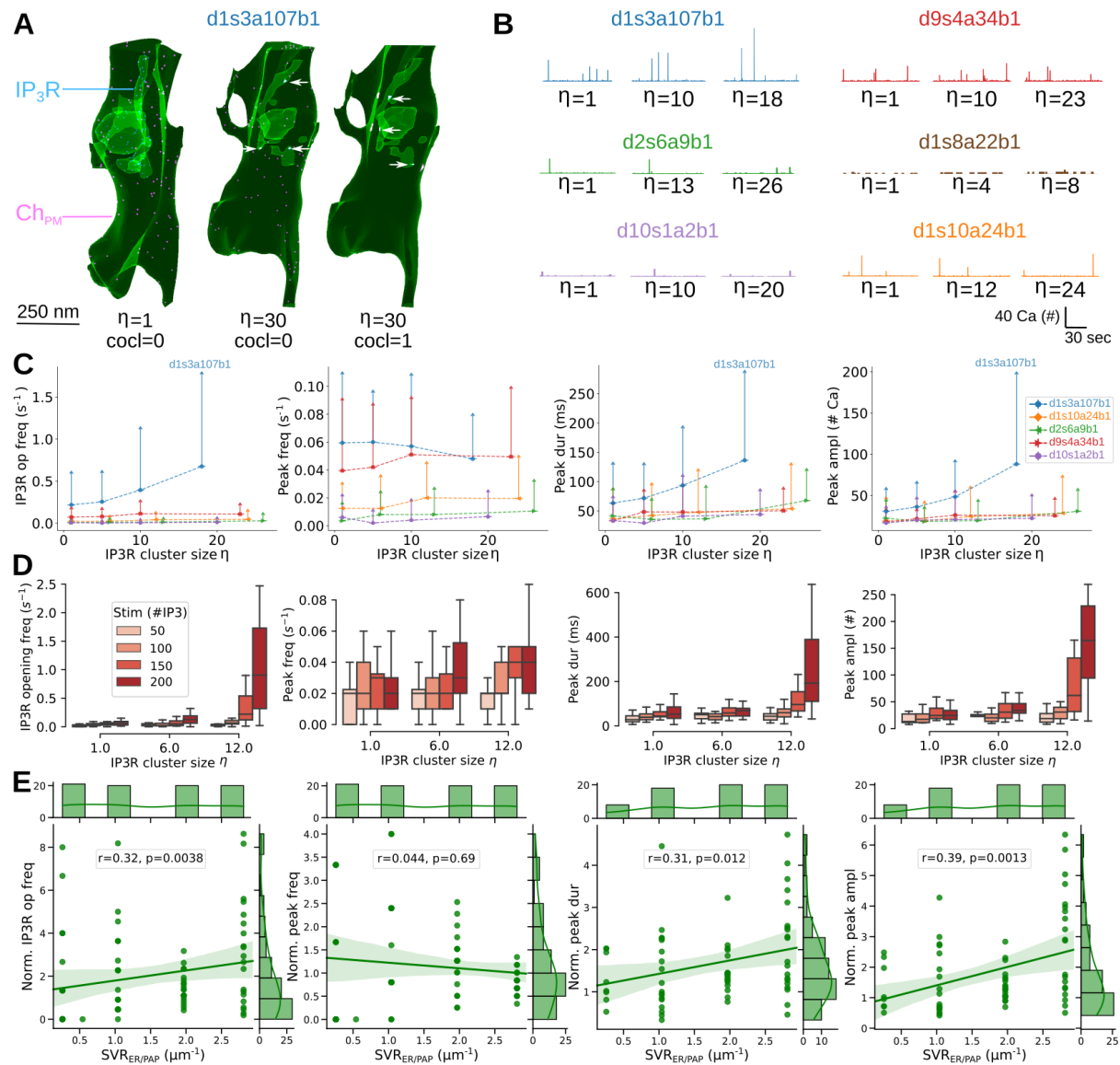
**Fig 4: Reaction-diffusion simulations reveal different spatio-temporal properties of Ca<sup>2+</sup> signals between PAPs of the same cell.** (Left) Schematic representation of the model of Ca<sup>2+</sup> signaling in PAPs used in this study. The model is stochastic, spatially-extended and simulations can be performed in 3D meshes. Ca<sup>2+</sup> influx into the cytosol results from Ca<sup>2+</sup> channels on the plasma membrane and from IP<sub>3</sub>R channels on the ER. At t=1s, 50 IP<sub>3</sub> molecules were injected at the plasma membrane of the PAP, simulating neuronal activity. (Right) Representative Ca-GCaMP traces from simulations in a cylindrical mesh, 200 nm in diameter, 1  $\mu\text{m}$  long (left, black) and experiments (right, red) [21].

Fig 4: (B) Images representing the 6 PAP meshes in which simulations were performed: d1s3a197b1, d1s8a22b1, d1s10a24b1, d2s6a9b1, d9s4a34b1 and d10s1a2b1. (C) Workflow to prepare the PAP meshes for 3D simulations illustrated on d2s6a9b1 mesh (see Methods section). The geometrical features of the resulting PAP meshes are presented in Table 1. (D) Screenshots of simulations in PAP meshes d9s4a34b1 and d10s1a2b1. Note that the darker and lighter greens result from 3D shading and rendering of the meshes. (E) Representative Ca-GCaMP (top, orange) and free  $\text{Ca}^{2+}$  (bottom, blue).  $\text{Ca}^{2+}$  traces were measured in separate simulations, where no GCaMP was added into the cytosol of the PAP.  $\text{IP}_3\text{R}$  channels and  $\text{Ca}^{2+}$  channels at the plasma membrane,  $\text{Ch}_{\text{PM}}$ , were randomly distributed onto the ER membrane and plasma membrane, respectively. (F) Quantification of peak frequency (left), duration (middle) and amplitude (right) of free  $\text{Ca}^{2+}$  (left, blue, n=20) and Ca-GCaMP (right, orange, n=20) signals measured *in silico* in 3D meshes of the PAPs presented in panel B.

177 PM triangles that were the closest to the ER triangles containing an  $\text{IP}_3\text{R}$  cluster (cocl=1), em-  
178 ulating co-localization of  $\text{Ca}^{2+}$  channels, reported in neurons and astrocytes [43]. As  $\text{IP}_3\text{R}$   
179 density was kept constant across simulations,  $3.5e^{-3}/\mu\text{m}^2$  [20], the total number of  $\text{IP}_3\text{Rs}$ ,  
180  $N_{\text{IP}_3\text{R}}$ , varied depending on the mesh: 90, 230, 78, 8, 40 and 96 in PAP meshes from synapses  
181 d1s3a107b1, d9s4a34b1, d2s6a9b1, d1s8a22b1, d10s1a2b1 and d1s10a24b1, respectively. As  
182  $\text{IP}_3\text{R}$  cluster size was a divider of  $N_{\text{IP}_3\text{R}}$ , cluster sizes tested varied slightly depending on the  
183 mesh. The range of  $\text{IP}_3\text{R}$  cluster size tested varied from  $\eta=1-26$ . Representative free  $\text{Ca}^{2+}$   
184 traces measured in d1s3a197b1, d1s8a22b1, d1s10a24b1, d2s6a9b1, d9s4a34b1 and d10s1a2b1  
185 PAP meshes with various  $\text{IP}_3\text{R}$  cluster sizes  $\eta$  are displayed in Fig. 5B. Strikingly,  $\text{IP}_3\text{R}$  cluster-  
186 ing only affected  $\text{Ca}^{2+}$  activity in a subset of the PAP meshes studied (Fig. 5C). Indeed,  $\text{Ca}^{2+}$   
187 peak duration and amplitude increased with  $\text{IP}_3\text{R}$  cluster size in PAP meshes from synapses  
188 d1s3a107b1 (ANOVA,  $p=5.1e^{-5}$  and  $1.9e^{-7}$ ), d2s6a9b1 (ANOVA,  $p=3.16e^{-3}$  and 0.026) and  
189 d9s4a34b1 (ANOVA,  $p=0.018$  and 0.028) but not in d1s10a24b1 (ANOVA,  $p=0.44$  and 0.32)  
190 and d10s1a2b1 (ANOVA,  $p=0.69$  and 0.83). This effect was associated with an increased fre-  
191 quency of  $\text{IP}_3\text{R}$  opening with cluster size in d1s3a107b1, d2s6a9b1 and d9s4a34b1 meshes  
192 (ANOVA,  $p=3.4e^{-4}$ , 0.007 and 0.037, respectively). Conversely, cluster size had no effect on  
193  $\text{IP}_3\text{R}$  opening frequency in d1s10a24b1 (ANOVA,  $p=0.050$ ) and d10s1a2b1 (ANOVA,  $p=0.15$ ).  
194 Interestingly, such differences in  $\text{IP}_3\text{R}$  clustering effects on  $\text{Ca}^{2+}$  activity were still observed  
195 in the absence of co-localization of  $\text{Ca}^{2+}$  channels at the plasma membrane with  $\text{IP}_3\text{R}$  clusters  
196 (Supplementary Fig. S2). Those results highlight that different PAP and ER shapes are associ-  
197 ated with different  $\text{IP}_3\text{R}$  clustering effects.

198

199 Interestingly, increased neuronal stimulation, simulated as an increased amount of  $\text{IP}_3$  in-  
200 fused,  $i$ , in the PAP at  $t=1\text{s}$ , triggered clustering effects in a PAP in which no clustering effect  
201 was observed after a milder neuronal stimulation (Fig. 5D). This effect was characterized by an  
202 increase of  $\text{Ca}^{2+}$  peak amplitude (ANOVA,  $p=8.2e^{-5}$  for  $i=150$  and  $p=4.66e^{-7}$  for  $i=200$ ), fre-  
203 quency (ANOVA,  $p=0.005$  for  $i=150$  and  $p=0.006$  for  $i=200$ ) and duration (ANOVA,  $p=3.4e^{-5}$



**Fig 5: The effect of  $IP_3R$  clustering differs from PAP to PAP.** (A) Screenshots of simulations in d1s3a107b1, with  $IP_3R$  cluster size  $\eta=1$  and  $\eta=30$ . Simulations were performed with ( $cocl=1$ ) and without ( $cocl=0$ ) co-clustering of  $Ca^{2+}$  channels at the plasma membrane (purple) with  $IP_3R$ s on the ER (blue).  $IP_3R$  clusters are indicated with a white arrow. (B) Representative free  $Ca^{2+}$  traces measured *in silico* in d1s3a107b1 (blue), d9s4a34b1 (red), d2s6a9b1 (green), d1s8a22b1 (brown), d10s1a2b1 (purple) and d1s10a24b1 (orange) meshes. (C) Quantification of  $IP_3R$  opening frequency (left),  $Ca^{2+}$  peak frequency (middle left), duration (middle right) and amplitude (right) as a function of  $\eta$ , in d1s3a107b1 (blue circle), d1s10a24b1 (orange diamond), d2s6a9b1 (green triangle), d9s4a34b1 (red triangle) and d10s1a2b1 (purple hexagon),  $cocl=1$ . Data are represented as mean  $\pm$  STD,  $n=20$  for each mesh and cluster size tested. (D) Quantification of  $IP_3R$  opening frequency (left),  $Ca^{2+}$  peak frequency (middle left), duration (middle right) and amplitude (right) as a function of  $IP_3R$  cluster size  $\eta$ , for different stimulation levels (Stim (#IP3): 50, 100, 150, 200). (E) Normalized  $IP_3R$  opening frequency (left), peak frequency (middle left), duration (middle right) and amplitude (right) as a function of  $SVR_{ER/PAP}$  ( $\mu m^{-1}$ ). Correlation coefficients ( $r$ ) and  $p$ -values are shown for each plot.

Fig 5: Lines are guides for the eyes. Note that no peaks were detected in simulations in d1s8a22b1. (D) Quantification of IP<sub>3</sub>R opening frequency (left), Ca<sup>2+</sup> peak frequency (middle left), duration (middle right) and amplitude (right), in d1s10a24b1, for  $\eta=1, 6$  and 12 and various levels of neuronal stimulation: IP<sub>3</sub> infused  $i=50, 100, 150$  and 200 (from left to right, light red to dark red),  $n=20$  for each parameter set tested. (E) Scatterplots presenting the variation of IP<sub>3</sub>R opening frequency (left), Ca<sup>2+</sup> peak frequency (middle left), duration (middle right) and amplitude (right) for  $\eta=18-24$ , normalized with Ca<sup>2+</sup> peak characteristics measured for  $\eta=1$ , as a function of  $SVR_{ER/PAP}$ . Plots are accompanied by univariate kernel density estimation curves and a linear regression fit. Spearman correlation coefficient,  $r$ , and  $p$ -value,  $p$ , are displayed onto each regression plot.

204 for  $i=150$ , and  $p=0.026$  for  $i=200$ ) with IP<sub>3</sub>R cluster size for  $i=150$  and 200, while cluster size  
205 did not affect peak amplitude (ANOVA,  $p=0.27$  for  $i=50$  and  $p=0.08$  for  $i=100$ ), frequency  
206 (ANOVA,  $p=0.86$  for  $i=50$  and  $p=0.72$  for  $i=100$ ) and duration ( $p=0.13$  for  $i=50$  and  $p=0.15$   
207 for  $i=100$ ) for  $i=50$  and 100. This effect was mediated by an increased IP<sub>3</sub>R opening frequency  
208 with cluster size when neuronal stimulation was larger (ANOVA,  $p=0.024, 5.23e^{-7}$  and  $7.8e^{-5}$ ,  
209 for  $i=100, 150$  and 200, respectively), while clustering had no effect on IP<sub>3</sub>R opening frequency  
210 for milder neuronal stimulation (ANOVA,  $p=0.21, i=50$ ). This suggests that IP<sub>3</sub>R clustering in  
211 PAPs could act as an amplifier of neuronal stimulation.

212

213 Unexpectedly, the PAPs in which an IP<sub>3</sub>R clustering effect was observed were not the PAPs  
214 with the highest ER surface area  $S_{ER}$ , i.e with the highest number of IP<sub>3</sub>R channels. Indeed,  
215 although the normalized IP<sub>3</sub>R opening frequency and Ca<sup>2+</sup> peak frequency at high cluster size  
216 were positively correlated with  $S_{ER}$ , normalized Ca<sup>2+</sup> peak amplitude and duration were not  
217 correlated with  $S_{ER}$  (Supplementary Fig. S3). Rather, normalized IP<sub>3</sub>R opening frequency,  
218 Ca<sup>2+</sup> peak amplitude and duration were positively correlated to the ratio between ER surface  
219 area and PAP volume  $SVR_{ER/PAP}$  (Fig. 5E,  $p=0.0038, 0.012$  and  $0.0013$ , respectively). Ca<sup>2+</sup>  
220 peak frequency however did not vary with  $SVR_{ER/PAP}$  (Fig. 5E,  $p=0.69$ ). This probably re-  
221 sults from our peak definition. Indeed, as a peak is considered terminated when the Ca<sup>2+</sup> trace  
222 decreases below peak threshold, a higher frequency of IP<sub>3</sub>R opening events can result in suc-  
223 cessive opening events occurring before peak termination, resulting in a similar peak frequency  
224 but larger peak duration.

225

226 Overall, our simulation results nuance the effect of the clustering of Ca<sup>2+</sup> channels on Ca<sup>2+</sup>  
227 signals in small sub-cellular compartments like PAPs. Strikingly, in contrast with reports from  
228 models in 2 spatial dimensions [23, 58], Ca<sup>2+</sup> activity increased with cluster size in only a  
229 subset of the realistic 3D PAP meshes tested, highlighting the complex interplay between the ER  
230 surface to PAP volume ratio, the intensity of neuronal stimulation and IP<sub>3</sub>R clustering on Ca<sup>2+</sup>  
231 microdomain activity. This highlights the importance of cautious interpretation of simulation  
232 results on geometrical effects depending on the geometry used.

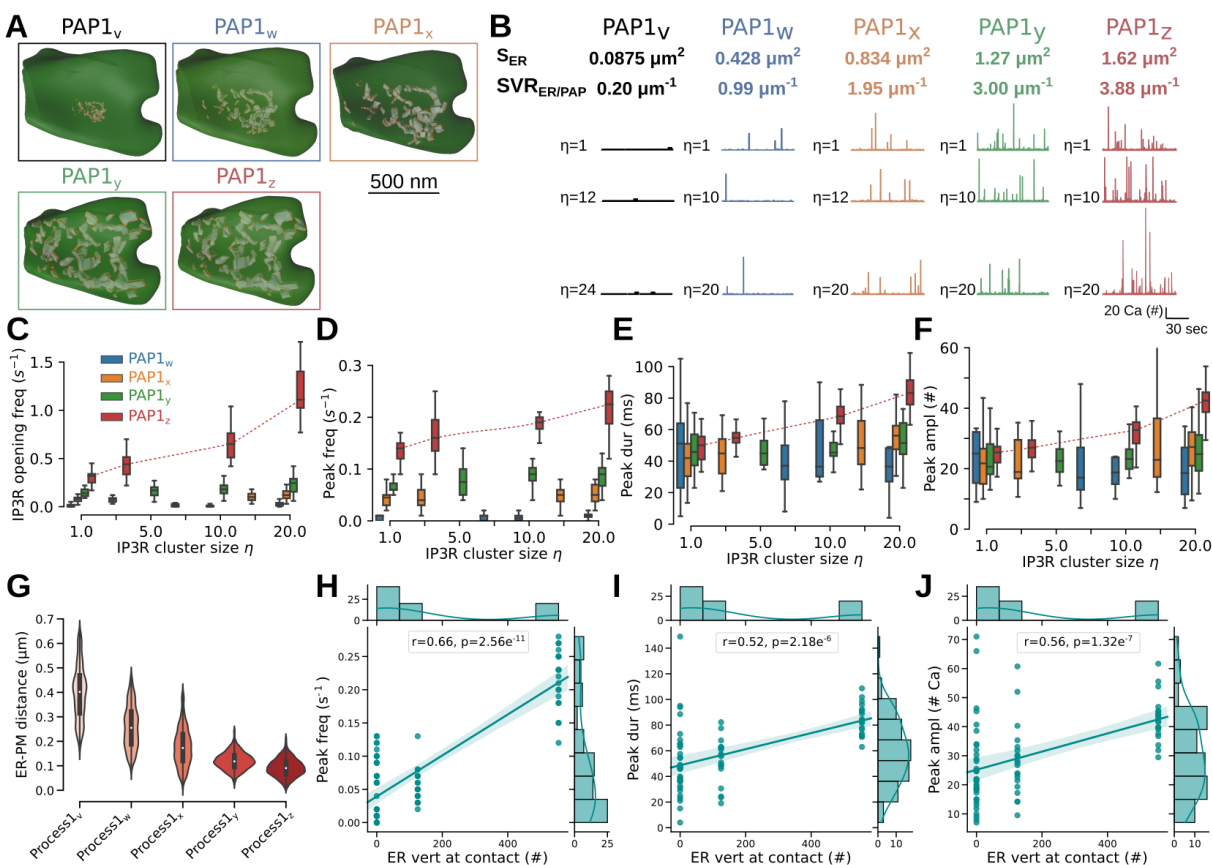
## 233 **The surface-volume ratio of the ER conditions the amplification of $\text{Ca}^{2+}$** 234 **activity by $\text{IP}_3\text{R}$ clustering in 3D**

235 Our simulation results revealed that the increase of  $\text{Ca}^{2+}$  activity resulting from  $\text{IP}_3\text{R}$  cluster-  
236 ing increased with the ratio between the ER surface area and the PAP volume ( $\text{SVR}_{\text{ER}/\text{PAP}}$ , Fig  
237 5E). In the PAP meshes studied in Fig. 5,  $\text{SVR}_{\text{ER}/\text{PAP}}$  varied together with PAP shape and ER  
238 shape. To discern the effect of ER and PAP shape from  $\text{SVR}_{\text{ER}/\text{PAP}}$  on  $\text{Ca}^{2+}$  activity in PAPs,  
239 we created meshes with various ER size and constant ER and PAP shapes. The original mesh  
240 was extracted from the  $220 \mu\text{m}^3$  astrocytic volume, located at the vicinity of the d9s3a51b1  
241 PSD and referred to as PAP1. The location of PAP1 in the  $220 \mu\text{m}^3$  reconstructed hippocampal  
242 astrocytic volume is presented in Supplementary Fig. S4. Meshes with various  $\text{SVR}_{\text{ER}/\text{PAP}}$   
243 were created from PAP1 by rescaling the ER using Blender software. Meshes were then created  
244 following the mesh pre-processing workflow described in Fig. 4C, resulting in the creation of  
245  $\text{PAP1}_v$ ,  $\text{PAP1}_w$ ,  $\text{PAP1}_x$ ,  $\text{PAP1}_y$  and  $\text{PAP1}_z$  meshes (Fig. 6A). The geometrical properties of  
246 those meshes are presented in Table 1.

247

248  $\text{IP}_3\text{R}$  opening frequency,  $\text{Ca}^{2+}$  peak frequency, duration and amplitude increased with  
249  $\text{SVR}_{\text{ER}/\text{PAP}}$  (Fig. 6B-F). This is not surprising as ER surface area increases with  $\text{SVR}_{\text{ER}/\text{PAP}}$   
250 in those meshes, thus resulting in an increase of the amount of  $\text{IP}_3\text{R}$  channels with  $\text{SVR}_{\text{ER}/\text{PAP}}$ .  
251 The total number of  $\text{IP}_3\text{R}$  channels,  $N_{\text{IP}_3\text{R}}$ , thus was 24, 120, 240, 360 and 460, in  $\text{PAP1}_v$ ,  
252  $\text{PAP1}_w$ ,  $\text{PAP1}_x$ ,  $\text{PAP1}_y$  and  $\text{PAP1}_z$  meshes, respectively. Importantly,  $\text{Ca}^{2+}$  peak frequency  
253 (Fig. 6D, ANOVA,  $p=2.39e^{-8}$ ), duration (Fig. 6E, ANOVA,  $p=7.52e^{-17}$ ) and amplitude (Fig.  
254 6F, ANOVA,  $p=1.29e^{-14}$ ) increased with  $\text{IP}_3\text{R}$  cluster size in  $\text{PAP1}_z$  mesh. This resulted  
255 from an increase of  $\text{IP}_3\text{R}$  opening frequency with cluster size in  $\text{PAP1}_z$  (Fig. 6C, ANOVA,  
256  $p=5.93e^{-24}$ ). No clustering effect was observed in  $\text{PAP1}_{w-y}$  meshes, characterized by a lower  
257  $\text{SVR}_{\text{ER}/\text{PAP}}$ , confirming the mechanistic link between  $\text{SVR}_{\text{ER}/\text{PAP}}$  and the amplification of  
258  $\text{Ca}^{2+}$  activity mediated by  $\text{IP}_3\text{R}$  clustering suggested in Fig. 5. Note that no  $\text{Ca}^{2+}$  signals were  
259 detected in  $\text{PAP1}_v$  mesh. Simulations in  $\text{PAP1}_z$  meshes with constant  $\text{IP}_3\text{R}$  channel number  
260 further highlight that this effect results both from the increased number of  $\text{IP}_3\text{R}$  channels in  
261  $\text{PAP1}_z$  and from ER shape (Supplementary Fig. S5). Supplementary Fig. S6 reveals that  $\text{IP}_3\text{R}$   
262 opening frequency and  $\text{Ca}^{2+}$  peak frequency increased with  $\text{SVR}_{\text{ER}/\text{PAP}}$  even when  $\text{IP}_3\text{R}$  chan-  
263 nels were not clustered ( $\eta=1$ ).

264 Increasing ER surface area in PAP1 however also resulted in a decreased distance between the  
265 ER and the plasma membrane (PM) in the PAP (Fig. 6G). Importantly,  $\text{Ca}^{2+}$  peak frequency  
266 (Fig. 6H,  $p=2.56e^{-11}$ ), duration (Fig. 6I,  $p=2.18e^{-6}$ ) and amplitude (Fig. 6J,  $p=1.32e^{-6}$ ) in-  
267 creased with the amount of ER vertices at ER-PM contact sites ( $\leq 20 \text{ nm}$  to the closest PM  
268 vertex [80, 79]). This suggests that the increased  $\text{IP}_3\text{R}$  clustering effect on  $\text{Ca}^{2+}$  microdomain  
269 activity in  $\text{PAP1}_z$  could result either from its increased  $\text{SVR}_{\text{ER}/\text{PAP}}$  or to the decreased ER-PM  
270 distances in this mesh compared to  $\text{PAP1}_{w-y}$  meshes. Overall, our simulation results suggest  
271 that ER shape, notably the ratio between its surface area and PAP volume, shape astrocytic  
272  $\text{Ca}^{2+}$  microdomain activity at synapses.



**Fig 6: The surface-volume ratio of the ER in PAPs conditions the amplification of  $\text{Ca}^{2+}$  activity by  $\text{IP}_3\text{R}$  clustering in 3D.** (A) Images of the different PAP meshes created to investigate the effect of the ratio between ER surface area and PAP volume,  $SVR_{ER/PAP}$ , on  $\text{Ca}^{2+}$  microdomain activity: PAP<sub>1v-z</sub>. Meshes were obtained by rescaling the ER object in PAP1, located at the vicinity of the d9s3a51b1 PSD (Supplementary Fig. S4). Geometrical features of the meshes are presented in Table 1. (B) Representative free  $\text{Ca}^{2+}$  traces measured in PAP<sub>1v</sub> (black), PAP<sub>1w</sub> (blue), PAP<sub>1x</sub> (orange), PAP<sub>1y</sub> (green) and PAP<sub>1z</sub> (red), for  $\text{IP}_3\text{R}$  cluster size  $\eta=1$  (top),  $\eta=12$  (middle) and  $\eta=24$  (bottom). (C-F) Quantification of  $\text{IP}_3\text{R}$  opening frequency (C),  $\text{Ca}^{2+}$  peak frequency (D), duration (E) and amplitude (F), in PAP<sub>1w</sub> (blue), PAP<sub>1x</sub> (orange), PAP<sub>1y</sub> (green) and PAP<sub>1z</sub> (red), for  $\eta=1-23$ . Note that  $\eta$  varies depending on the mesh (see Methods). Lines were added to visualize the effect of  $\eta$  on  $\text{Ca}^{2+}$  peak characteristics in PAP<sub>1z</sub> mesh. (G) Quantification of the variation of the distance between each ER vertex and the closest plasma membrane (PM) vertex in PAP<sub>1v-z</sub> meshes. (H-J) Scatterplots presenting  $\text{Ca}^{2+}$  peak frequency (H), duration (I) and amplitude (J) in PAP<sub>1v-z</sub> meshes for  $\eta=20$ , as a function of the number of ER vertices  $\leq 20$  nm to the closest PM vertex.

Fig 6: Plots are presented with univariate kernel density estimation curves and a linear regression fit. Spearman correlation coefficient,  $r$ , and  $p$ -value,  $p$ , are displayed onto each regression plot.

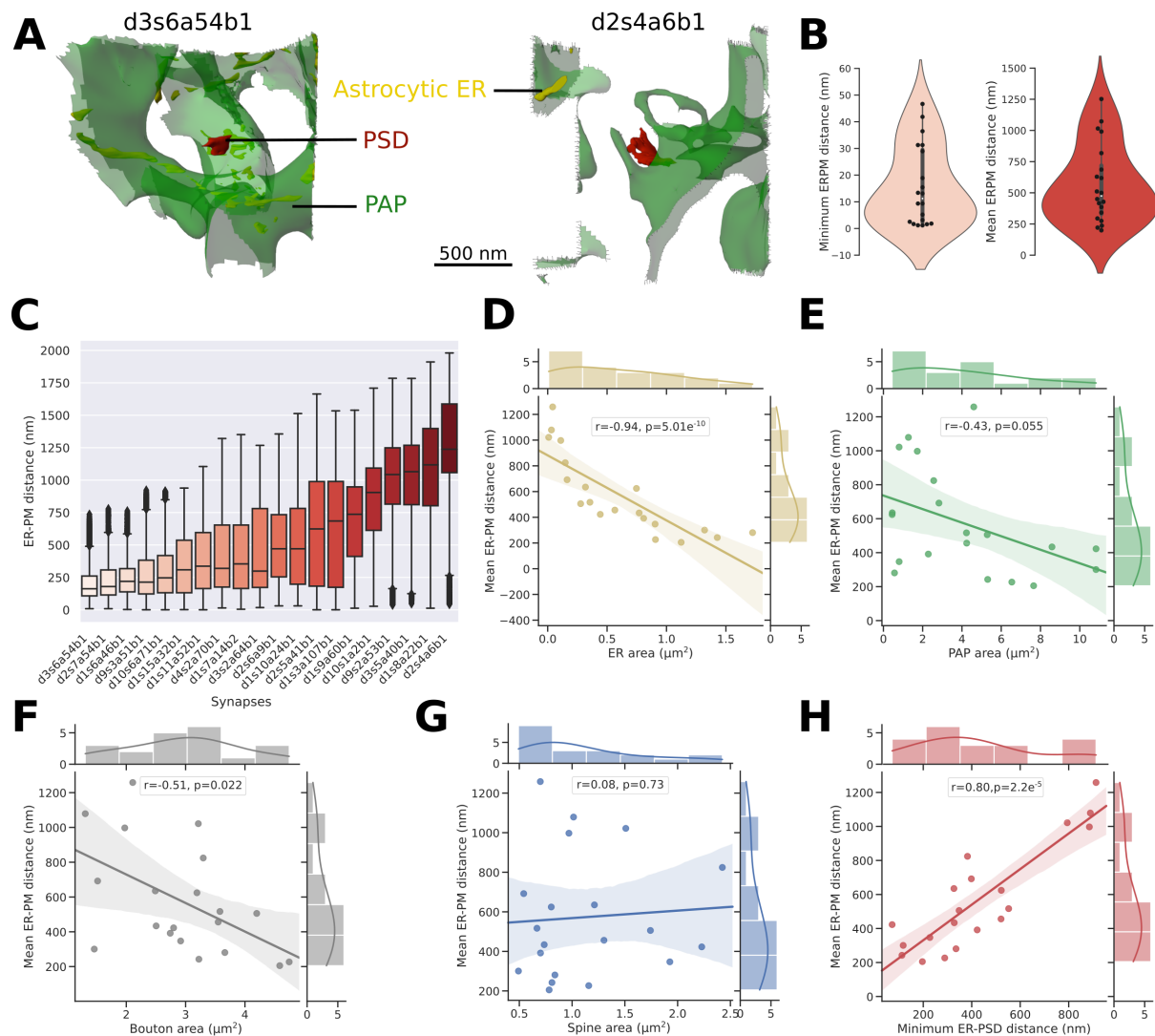
## 273 **Quantification of ER-PM distance distribution within PAPs**

274 As simulation results suggested that  $\text{Ca}^{2+}$  activity in  $\text{PAP1}_{v-z}$  meshes varies depending on  
275 the distribution of the ER within the PAP, we next aimed at quantifying ER distribution in the  
276 twenty ER-containing PAP meshes reconstructed from EM and presented in Fig. 3. To do so,  
277 we measured the distance between each vertex on the plasma membrane (PM) and the closest  
278 vertex on the ER. We found that ER-PM distance is highly variable in PAPs from a single cell,  
279 with an average ER-PM distance within a single PAP from around 200 nm to 1200 nm (Fig.  
280 7B-C). Not surprisingly, mean ER-PM distance decreases as ER (Fig. 7D,  $p=5.01e^{-10}$ ) and  
281 PAP (Fig. 7E,  $p=0.055$ ) surface area increase. Interestingly, ER-PM distance was lower in  
282 PAPs contacting boutons with higher surface area (Fig. 7F,  $p=0.022$ ). Note that there was no  
283 correlation between ER-PM distance and spine surface area (Fig. 7G,  $p=0.73$ ). Importantly,  
284 we found that PAPs closer to the synapse are characterized by lower mean ER-PM distance  
285 (Fig. 7H,  $p=2.2e^{-5}$ ), which, according to simulation results presented in Fig. 6, could result in  
286 enhanced  $\text{Ca}^{2+}$  activity in those PAPs.

## 287 **Effect of ER-PM distance in PAPs on $\text{Ca}^{2+}$ microdomain activity**

288 To discern the effect of  $\text{SVR}_{\text{ER/PAP}}$  from the effect of ER-PM distance on  $\text{Ca}^{2+}$  microdomain  
289 activity in PAPs reported in Fig. 6, we implemented an algorithm that creates realistic tetra-  
290 hedral 3D meshes of PAPs characterized by various distributions of the ER within the same  
291 PAP. The workflow is presented in Fig. 8. Briefly, the ER is split into small portions of simi-  
292 lar size, then resized to match the total ER surface area of the original mesh. Simulations in  
293 meshes with the original ER and with split ER confirmed that this ER splitting algorithm does  
294 not alter  $\text{Ca}^{2+}$  activity in the PAP (Supplementary Fig. S7). A simulation of  $n$  frames is then  
295 generated in Blender, which alters the location of the ER objects within the PAP. Each frame is  
296 thus characterized by a unique distribution of the ER objects within the PAP, while ER and PAP  
297 shape, surface area, volume and SVR are constant across frames (Supplementary movie 3). The  
298 mesh processing workflow presented in Fig. 4C is then automatically applied to each frame  
299 of interest. This workflow allows the creation of numerous realistic 3D PAP meshes, that can  
300 be used for reaction-diffusion simulations in 3D. Fig. 7B-D displays the quantification of ER  
301 distribution in the PAP meshes created with this workflow on PAP d1s15a32b1. The workflow  
302 successfully produced realistic tetrahedral PAP meshes characterized by various ER-PM dis-  
303 tances (Fig. 7B). Note that the distribution of ER-PM distance at ER-PM contact sites (ER-PM  
304 distance  $\leq 20$  nm) did not vary, while the number of ER vertices belonging to contact sites, and  
305 thus the size of the contact sites, increased with frame number (Fig. 7C).





**Fig 7: ER-PM distance in PAPs is highly variable and is decreased in PAPs close to the synapse.** (A) Images of 2 PAP meshes (green), d3s6a54b1 (left) and d2s4a6b1 (right) with the neighboring PSD (red), displaying the diverse shapes and distributions of the ER (yellow) in PAPs from the same cell. (B) Distribution of the minimum (left) and mean (right) distance between each vertex on the plasma membrane (PM) and the closest ER vertex, measured in PAP meshes reconstructed from EM,  $n=20$  (Fig. 3). (C) Quantification of the distance between each PM vertex and the closest ER vertex in each PAP mesh. (D-F) Scatterplots presenting the variation of the mean distance between each PM vertex and the closest ER vertex as a function of the surface area of the ER (D), PAP (E), bouton (F), and spine (G), and as a function of the minimum ER-PSD distance (H),  $n=20$ . Plots are presented with univariate kernel density estimation curves and a linear regression fit. Spearman correlation coefficient,  $r$ , and  $p$ -value,  $p$ , are displayed onto each regression plot.

306

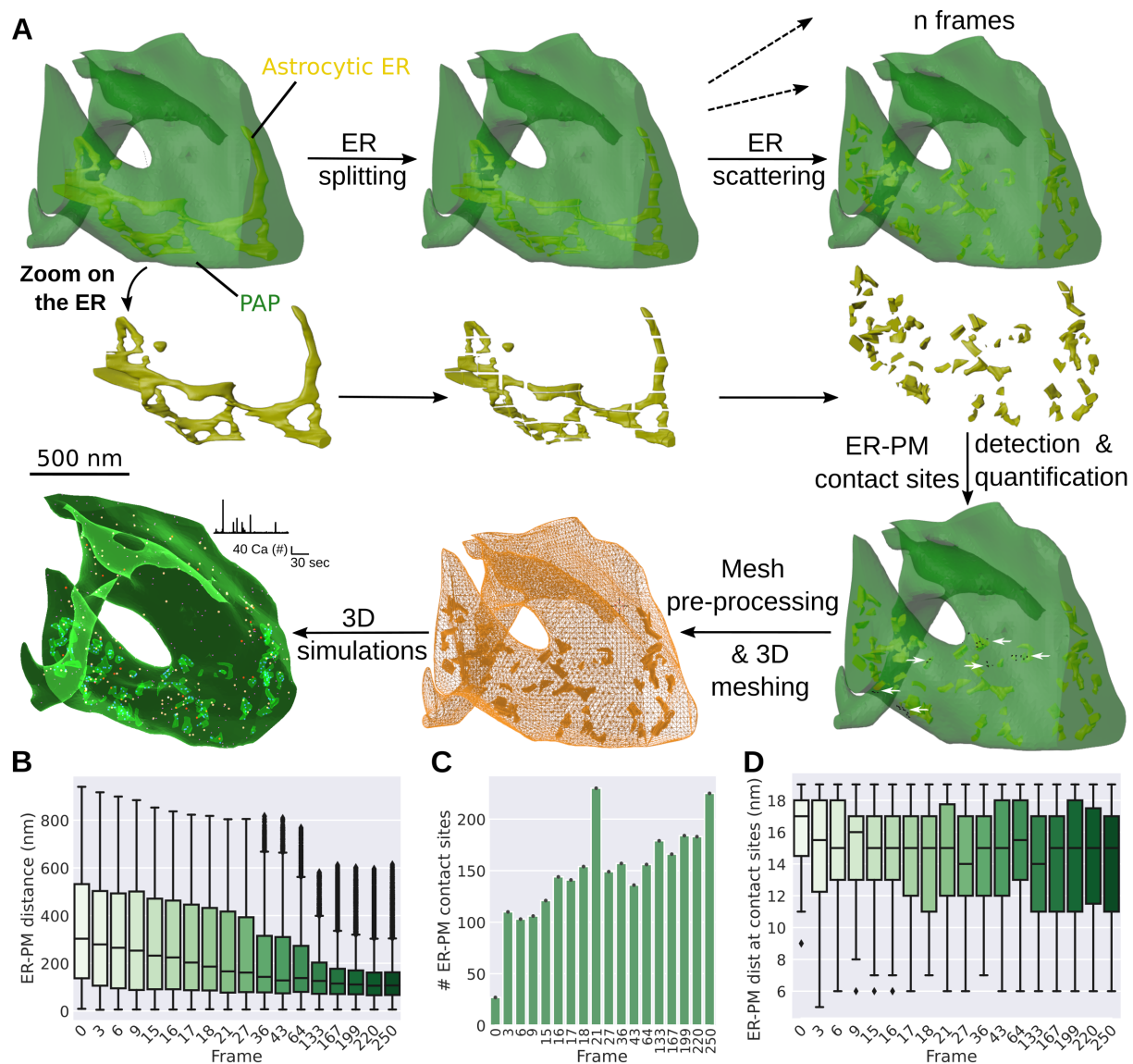
307 To test the effect of ER distribution in PAPs, simulations were performed in meshes from  
308 frames 0, 21, 64 and 250 of PAP d1s15a32b1 (Fig. 9A) with different IP<sub>3</sub>R cluster sizes,  $\eta$ .  
309 Free Ca<sup>2+</sup> signals did not vary depending on ER distribution within the PAP (Fig. 9C). Inter-  
310 estingly, ER distribution did not affect the amplification of IP<sub>3</sub>R opening frequency (ANOVA,  
311  $p=0.059$ ), Ca<sup>2+</sup> peak duration (ANOVA,  $p=0.55$ ) and amplitude (ANOVA,  $p=0.15$ ) with IP<sub>3</sub>R  
312 cluster size when IP<sub>3</sub>R clusters were randomly distributed on the membrane of the ER (Fig.  
313 9D). However, the increase of peak frequency with cluster size was larger in meshes in which  
314 the ER was closer to the plasma membrane (Fig 9D, ANOVA,  $p=0.0048$ ). This suggests that  
315 a distribution of the ER closer to the plasma membrane might allow increased Ca<sup>2+</sup> peak fre-  
316 quency with IP<sub>3</sub>R cluster size.

317

318 As IP<sub>3</sub>R clusters are believed to be preferentially located at ER-PM contact sites [72], we  
319 performed simulations in the same meshes while positioning IP<sub>3</sub>R clusters at ER-PM contact  
320 sites. Screenshots of simulations with different locations of IP<sub>3</sub>R clusters are presented in Fig.  
321 9B. Interestingly, locating IP<sub>3</sub>R clusters at ER-PM contact sites resulted in larger increases of  
322 IP<sub>3</sub>R opening frequency (ANOVA,  $p=0.0019$ ) and Ca<sup>2+</sup> peak frequency (ANOVA,  $p=0.0043$ )  
323 with cluster size in meshes in which the ER was closer to the PM (Fig. 9D). However, the loca-  
324 tion of the ER did not impact the effect of IP<sub>3</sub>R cluster size on Ca<sup>2+</sup> peak duration (ANOVA,  
325  $p=0.45$ ) and amplitude (ANOVA,  $p=0.069$ ). Together, those results suggest that a distribution  
326 of the ER closer to the plasma membrane, coupled with a location of IP<sub>3</sub>R clusters at ER-PM  
327 contact sites, favors an increase of Ca<sup>2+</sup> peak frequency with cluster size. ER-PM contact sites  
328 could act as diffusional barriers. Locating IP<sub>3</sub>R channels at ER-PM contact sites would thus  
329 increase the residency time of Ca<sup>2+</sup> ions and IP<sub>3</sub> molecules at the vicinity of the channels, thus  
330 increasing the probability of Ca<sup>2+</sup> and IP<sub>3</sub> binding to IP<sub>3</sub>Rs and resulting in an increased IP<sub>3</sub>R  
331 opening frequency in meshes with larger ER-PM contact sites. To further test the interplay  
332 between ER distribution, IP<sub>3</sub>R clustering and local diffusional properties, Ca<sup>2+</sup> buffers, here  
333 Ca<sup>2+</sup> indicators GCaMP6s, were added to the model. Strikingly, ER distribution had a greater  
334 effect on buffered Ca<sup>2+</sup> signals compared to free Ca<sup>2+</sup> signals (Fig. 9E). More precisely, the in-  
335 crease of IP<sub>3</sub>R opening frequency (ANOVA,  $p=2.37e^{-7}$ ), Ca-GCaMP peak frequency (ANOVA,  
336  $p=1.14e^{-4}$ ), duration (ANOVA,  $p=7.78e^{-3}$ ) and amplitude (ANOVA,  $p=0.023$ ) with IP<sub>3</sub>R clus-  
337 ter size was significantly larger in meshes in which the ER was located closer to the plasma  
338 membrane. Those results highlight that the effect of ER distribution within the PAP on Ca<sup>2+</sup>  
339 microdomain activity is conditioned by the location of IP<sub>3</sub>R channels at ER-PM contact sites  
340 and by local Ca<sup>2+</sup> buffering.

## 341 Discussion

342 Here, we extracted 3D meshes of tripartite synapses from a 220  $\mu m^3$  hippocampal astrocytic  
343 volume from the CA1 stratum radiatum region, reconstructed from EM [17]. Quantification



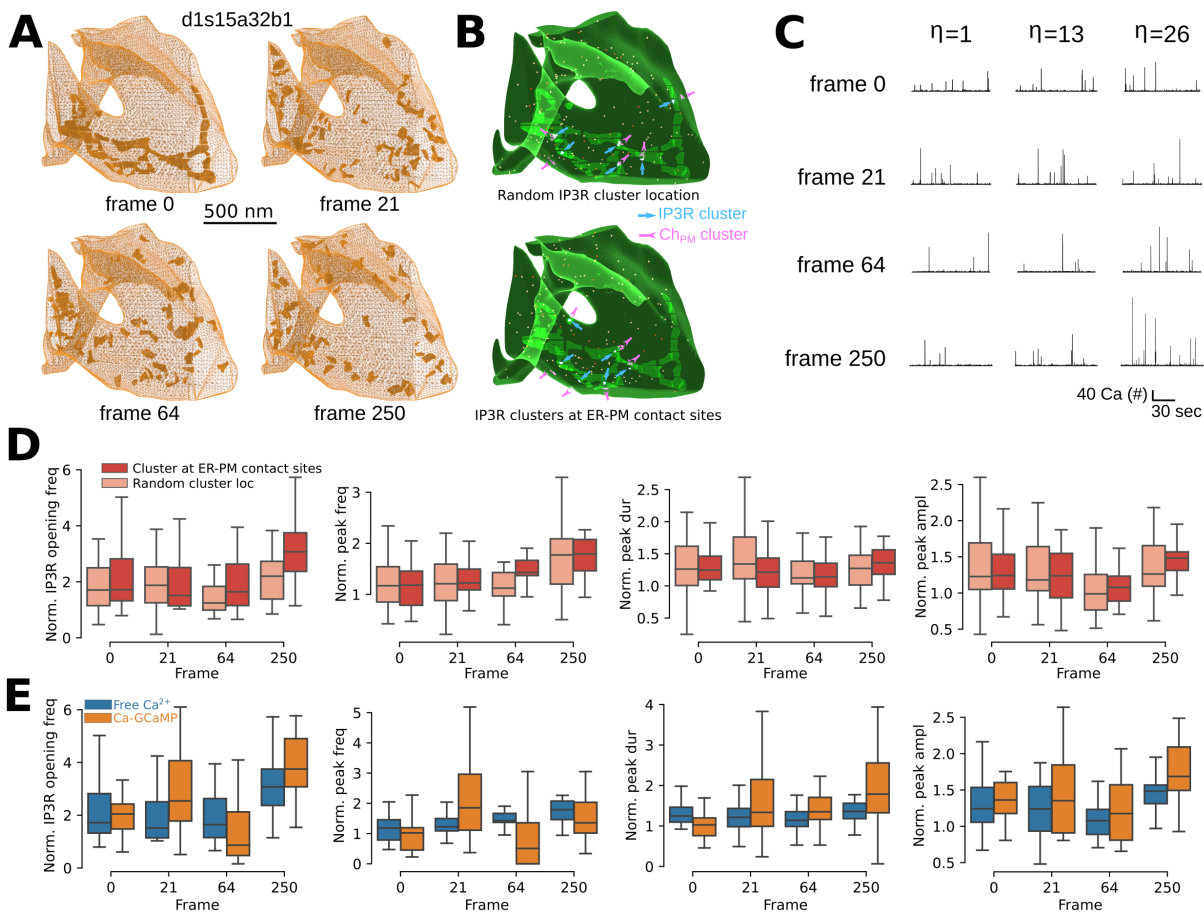
**Fig 8: Automated realistic 3D PAP mesh generation with diverse ER distributions.** (A) Schematic representing the workflow developed in this study to create realistic PAP meshes in 3 spatial dimensions with various ER distributions and constant shape, volume and surface area of PAP and ER, used on the PAP mesh d1s15a32b1. The ER is split and a simulation with  $n$  frames is generated in Blender, in which ER objects are subject to physical forces that alter their spatial distribution. The  $n$  frames are thus characterized by different locations of the ER elements within the PAP, with constant ER and PAP shapes. The pipeline detects, quantifies and exports in a text file the distance between each vertex at the plasma membrane (PM) and the closest vertex at the membrane of the ER. A point cloud can be created to visualize the ER vertices at ER-PM contact sites (ER-PM distance  $\leq 20$  nm, white arrows). The mesh pre-processing workflow presented in Fig. 4C is then applied to the mesh of each desired frame. The resulting 3D tetrahedral meshes can then be used for 3D reaction-diffusion simulations.

Fig 8: (B) Quantification of the distance between each PM vertex and the closest ER vertex in the meshes generated by the workflow presented in panel A, applied to the d1s15a32b1 PAP mesh. (C) Quantification of the number of ER vertices located at ER-PM contact sites, i.e.  $\leq 20$  nm to the closest PM vertex, in each frame from mesh d1s15a32b1. (D) Quantification of the distance between each PM vertex and the closest ER vertex at ER-PM contact sites, in each frame from mesh d1s15a32b1.

344 of the geometrical features of those meshes highlighted the diverse geometrical properties of  
345 PAPs from a single astrocyte and revealed, contrary to a widespread belief that PAPs are devoid  
346 of ER [55, 59], that 75 % of PAPs contained some ER. Interestingly, we found that PAPs are  
347 the closest to the synapse when bouton surface area is low, which could result from the spa-  
348 tial constraints imposed by larger boutons, preventing the PAP from getting in close contact to  
349 the PSD. Reaction-diffusion simulations in the realistic PAP 3D meshes reconstructed in this  
350 study provided key insights into the effect of the diverse shapes and distributions of the ER in  
351 PAPs on microdomain  $\text{Ca}^{2+}$  activity. As reactive astrocytes, hallmark of brain diseases [26],  
352 are characterized by a remodelling of ER volume and shape [39], our results suggest that such  
353 geometrical alterations of the ER could be one of the factors responsible for the altered astro-  
354 cytic  $\text{Ca}^{2+}$  activity reported in pathological conditions [65].

355  
356 Fine-tuning the spatial distribution of  $\text{Ca}^{2+}$  channels, monitoring channel opening events at  
357 each channel, while independently manipulating ER shape and distribution, such as performed  
358 in this study, is not feasible experimentally. It is yet essential to understand the mechanistic link  
359 between the spatial features of the astrocyte and its  $\text{Ca}^{2+}$  microdomain activity. Combining our  
360 detailed biophysical model of  $\text{Ca}^{2+}$  signals in PAPs, the PAP meshes that we extracted from  
361 EM and the realistic PAP meshes with various ER distributions generated by our automated  
362 mesh generator allowed us to provide key insights into  $\text{Ca}^{2+}$  signaling in PAPs. Notably, we  
363 predict how the complex interplay between the clustering of  $\text{Ca}^{2+}$  channels, the ratio between  
364 ER surface area and PAP volume,  $\text{Ca}^{2+}$  buffering and the size and location of ER-PM  
365 contact sites shapes  $\text{Ca}^{2+}$  microdomain signals at tripartite synapses. This study is the first to  
366 our knowledge to model  $\text{Ca}^{2+}$  activity in astrocytes with realistic shapes in 3D at the nanoscale  
367 that accounts for the complex and diverse spatial characteristics of  $\text{Ca}^{2+}$  stores in PAPs. Fur-  
368 thermore, our results highlight the impact of the modeling choices on simulation results, notably  
369 when investigating spatial effects. Importantly, our results nuance the effect of the clustering of  
370  $\text{Ca}^{2+}$  channels, which is stronger in 2D or simple 3D shapes than in more realistic 3D meshes.  
371 This is crucial as, until now, modeling studies on PAPs were conducted in 1D, 2D or in simple  
372 3D shapes, notably cylinders [47, 24, 13, 20, 22, 32]. The 3D meshes provided by this study,  
373 together with our realistic 3D PAP mesh generator, pave the way for future modeling studies in  
374 realistic 3D meshes to investigate the mechanisms governing neuron-astrocyte communication  
375 at tripartite synapses.

376



**Fig 9: Ca<sup>2+</sup> buffering and IP<sub>3</sub>R channel location mediate the effect of ER-PM distance on Ca<sup>2+</sup> microdomain activity in PAPs.** (A) Images presenting different meshes created from PAP d1s15a32b1 using the automated workflow presented in Fig. 8: frames 0, 21, 64 and 250, characterized by diverse ER distributions within the PAP with constant PAP and ER shape, volume and surface area. Characteristics of ER-PM distance in those meshes are displayed in Fig. 8B-D. (B) Screenshots of simulations performed in d1s15a32b1<sub>f0</sub> mesh (frame 0), with IP<sub>3</sub>R clusters (blue, arrows) distributed randomly on the ER membrane (top) or at ER-PM contact sites (bottom). IP<sub>3</sub>R clusters were co-localized with Ch<sub>PM</sub> clusters at the plasma membrane (purple, inverse arrows):  $\text{cocl}=1$ . (C) Representative free Ca<sup>2+</sup> traces in frames 0, 21, 64 and 250, with IP<sub>3</sub>R cluster size  $\eta=1$ , 13 and 26 and random distribution of IP<sub>3</sub>R clusters on the ER. (D) Quantification of IP<sub>3</sub>R opening frequency (left), Ca<sup>2+</sup> peak frequency (middle left), duration (middle right) and amplitude (right), in frames 0, 21, 64 and 250, for  $\eta=26$ , normalized by Ca<sup>2+</sup> peak characteristics for  $\eta=1$  in each mesh, with IP<sub>3</sub>R clusters at random locations on the ER (left, light red) or at ER-PM contact sites (right, dark red).

Fig 9: (E) Quantification of IP<sub>3</sub>R opening frequency (left), peak frequency (middle left), duration (middle right) and amplitude (right) of free Ca<sup>2+</sup> signals (blue) and Ca-GCaMP signals (orange), in frames 0, 21, 64 and 250, for  $\eta=26$ , normalized by Ca<sup>2+</sup> peak characteristics for  $\eta=1$ . IP<sub>3</sub>R clusters were located at ER-PM contact sites. n=20 for each parameter set tested.

377 The geometrical data used here were extracted from electron microscopy, which is the only  
378 tool that can resolve PAP and ER shape at a high spatial resolution (6 nm here), yet results  
379 in potential alterations of the ultrastructure of the extracellular space [42] and cannot be used  
380 to study live cells. Furthermore, the model used in this study, focusing on the effect of the ER  
381 shape and distribution on Ca<sup>2+</sup> activity, describes with great details the kinetics of ER-mediated  
382 Ca<sup>2+</sup> signals while simplifying other Ca<sup>2+</sup> sources. Other Ca<sup>2+</sup> sources and channels however  
383 contribute to Ca<sup>2+</sup> microdomain activity in PAPs, including mitochondria, the Na<sup>+</sup>/Ca<sup>2+</sup> ex-  
384 changer, transient receptor potential ankyrin 1 channels, L-type voltage gated channels and  
385 other pathways [64, 62]. According to our model's predictions, the spatial distribution of  
386 Ca<sup>2+</sup> channels can alter the spatio-temporal properties of Ca<sup>2+</sup> microdomain signals in PAPs  
387 as well as their amplification upon neuronal stimulation. Further quantification of the Ca<sup>2+</sup>  
388 channels expressed in PAPs, their density, location and remodeling in live tissue under (patho-  
389 )physiological conditions is thus essential to better understand astrocyte activity at synapses.  
390 The recent advances in super-resolution techniques, notably single particle tracking methods,  
391 provide a promising avenue to overcome current limitations in obtaining such data [33, 6].

392  
393 Recent super-resolution studies in live neurons revealed dynamical remodeling of ER-PM  
394 contact sites [27] and diffusional trapping of molecules resulting from the ER remodeling [18]  
395 in neurons. Those observations, together with our model predictions, highlight the need for fur-  
396 ther quantification of the dynamical shape and distribution of the ER in astrocytes in live tissue  
397 to fully grasp its influence on Ca<sup>2+</sup> microdomain activity in astrocytes. According to our model  
398 predictions, preferential location of IP<sub>3</sub>Rs at ER-PM contact sites might be essential to allow  
399 signal amplification with IP<sub>3</sub>R cluster size and could thus strongly alter the spatio-temporal  
400 properties of astrocytic Ca<sup>2+</sup> signals evoked by neurotransmitters, potentially affecting the sub-  
401 sequent modulation of neuronal activity by astrocytes. Our results, in accordance with previous  
402 computational studies in other cell types [81], highlight that Ca<sup>2+</sup> buffering plays a crucial role  
403 in shaping Ca<sup>2+</sup> activity at ER-PM contact sites. The Ca<sup>2+</sup> buffering effect described here was  
404 mediated by Ca<sup>2+</sup> indicators. Future experimental and computational studies will be essential  
405 to assess Ca<sup>2+</sup> buffering mechanisms in astrocytes and PAPs, which are still poorly understood,  
406 yet, according to our simulation results, play crucial roles in shaping Ca<sup>2+</sup> microdomain activ-  
407 ity in astrocytes.

408  
409 Overall, this study provides new insights into astrocytic activity at tripartite synapses by  
410 characterizing the presence, shape and distribution of the ER in PAPs and by shedding light  
411 to the mechanistic link between those features and microdomain Ca<sup>2+</sup> activity at tripartite

412 synapses. The realistic 3D meshes of tripartite synapses created in this study pave the way for  
413 new modeling studies of neuron-astrocyte communication in the synaptic micro-environment,  
414 allowing the study of various processes, such as glutamate spillover or gliotransmission. Such  
415 studies will be crucial to decipher whether the various nano-architectures displayed by tripartite  
416 synapses reflect distinct functional identities.

## 417 **Methods**

### 418 **3D reconstruction from electron microscopy**

#### 419 **Sample preparation and imaging**

420 The original dataset used in this work (EM stack and 3D reconstructions) was previously pub-  
421 lished in [17]. The block was a gift from Graham Knott (BioEM imaging facility at EPFL,  
422 Lausanne, Switzerland). All procedures were performed according to the Swiss Federal Laws.  
423 One P90 Sprague-Dawley rat was deeply anesthetized with isoflurane and transcardially per-  
424 fused using 2% paraformaldehyde and 2.5% glutaraldehyde in PBS 0.1M. Coronal sections  
425 ( $100\ \mu\text{m}$ ) were obtained and washed in cacodylate buffer, followed by a post-fixation using  
426 osmium tetroxide and uranyl acetate. Finally, the sections were embedded in Durcupan. Re-  
427 gions of the hippocampus were dissected under a stereoscopic microscope, mounted onto a  
428 blank resin slab, and trimmed using an ultramicrotome (Laica Ultracut UC-7). Imaging was  
429 performed using an NVision 40 FIB-SEM (Carl Zeiss) with an acceleration voltage of 1.5 kV, a  
430 current of 350 pA, and a dwell time of  $10\ \mu\text{s}/\text{pixel}$ . Serial images were obtained using backscat-  
431 tered electrons and collected at a  $6\ \text{nm}/\text{pixel}$  magnification and 5 nm of milling depth between  
432 images.

#### 433 **3D reconstruction and rendering**

434 The serial micrographs were first registered using Multistackreg, a freely available plug-in  
435 for Fiji [17]. Then, using those micrographs, we proceeded to the image segmentation and  
436 3D model reconstructions by using TrackEM2 (a plug-in for Fiji) for manual segmentation,  
437 and iLastik, for a semi-automated segmentation. The extracted models were then imported to  
438 Blender software for visualization and rendering purposes [16].

#### 439 **Extraction of tripartite synapse meshes**

440 For each synapse in contact with the  $220\ \mu\text{m}^3$  astrocytic volume, a cube of edge length  $1.5\ \mu\text{m}$   
441 ( $3.375\ \mu\text{m}^3$ ) was created and centered at the center of mass of the PSD. All the elements of  
442 the mesh (astrocyte, astrocytic ER, spine and bouton) that were within the cubic volume were  
443 isolated using a boolean intersection operator available in Blender, forming what we refer to as  
444 a tripartite synapse mesh. The size of the cube was chosen to be large enough to contain the

445 whole spine and bouton elements while containing a single synapse, taking into consideration  
446 that the neuropil is believed to contain around one synapse per micrometer cube. This workflow  
447 resulted in the creation of 44 excitatory and 2 inhibitory synapse meshes.

### 448 **3D mesh manipulation**

449 All 3D mesh manipulations were performed with open-access, open-source software. All 3D  
450 PAP meshes used in this study will be available online upon paper acceptance.

### 451 **3D PAP mesh processing for reaction-diffusion simulations**

452 PAP meshes from tripartite synapse meshes were pre-processed using Blender software so that  
453 they could be used for reaction-diffusion simulations. The workflow is illustrated in Fig. 4C.  
454 Intersection between ER and PAP membranes was prevented by using a boolean intersection  
455 operator. ER was relocated a few nanometers away from the plasma membrane. PAP compart-  
456 ments that did not belong to the main PAP volume were deleted. Boolean difference operation  
457 between PAP and ER elements was performed. Non-manifold vertices were repaired. The  
458 resulting PAP mesh was exported in .stl format, which was then converted into a .msh 3D tetra-  
459 hedral mesh using TetWild software [38]. Lastly, the mesh was imported into Gmsh software  
460 to be converted into 2.2 ASCII format, format supported by the STEPS mesh importer.

### 461 **Automated 3D PAP mesh generation**

462 We have implemented a workflow to generate realistic 3D tetrahedral PAP meshes characterized  
463 by various ER locations and constant ER shape. The algorithm is written in python, can be  
464 imported in Blender and is available at <https://bit.ly/3Nc2Qin>. The workflow is presented in Fig.  
465 8. First, all elements of the mesh, i.e the PAP and the ER, are relocated so that their center of  
466 mass is centered at the origin. Then, the ER is split into smaller ER objects using a custom-made  
467 function. Briefly,  $n$  cubes of a given size are placed along the ER object. Intersection boolean  
468 operation is then performed between the ER and each cube, resulting in the creation of  $n$  ER  
469 objects. ER objects smaller than  $30 \text{ nm}^3$  are deleted. The remaining ER objects are rescaled  
470 so that the sum of their surface areas matches the area of the original ER element, measured  
471 with the Blender 3D Print add-on. The number and size of cubes can be altered depending on  
472 the size of the original ER and on the mesh characteristics desired. Using Blender's physics  
473 engine, a simulation with  $n$  frames is generated, in which ER objects are subject to physical  
474 forces that alter their location between each frame. Inputs of the 'RunPhysics' function include  
475 parameters that affect how close objects can get, which can be altered to prevent membrane  
476 intersection. Note that successful scattering of the ER depends on the geometrical properties of  
477 each mesh so that adjusting the parameters of physics simulation might be necessary depending  
478 on the mesh used. Details are provided in comments of the code to allow the user to adjust  
479 the code to the mesh under study. Examples of frames generated by this workflow applied to



480 d1s15a32b1 PAP mesh are presented in Supplementary movie 3. For each selected frame, the  
481 mesh pre-processing steps presented in Fig. 4C are performed automatically, resulting in the  
482 export of a .stl triangular mesh. 3D meshing and format conversion can then be performed using  
483 TetWild and Gmsh software, as described above. The resulting meshes can be used to perform  
484 reaction-diffusion simulations.

### 485 **Analysis of the geometrical properties of 3D meshes**

486 The volume and surface area of each synaptic element, i.e the PAP, astrocytic ER, spine and  
487 bouton, were measured using the Blender add-on Neuromorph [40]. We implemented a python  
488 script that can be imported in Blender software that measures distances between mesh elements  
489 of interest. The distance between each vertex of the plasma membrane of the PAP and the  
490 center of mass of the neighboring PSD was computed in Blender and stored in a list. Similarly,  
491 ER-PSD distance was quantified by measuring the distance between each vertex of the ER  
492 membrane and the center of mass of the PSD. To characterize ER-PM distance, for each vertex  
493 on the PM, the closest ER vertex was detected and its distance to the PM vertex was stored in  
494 a list. PM-PSD, ER-PSD and ER-PM distance lists were exported to a text file for analysis and  
495 visualisation. The analysis code, implemented in python and imported in Blender, is available  
496 at <https://bit.ly/3Nc2Qin>.

## 497 **Computational modeling**

### 498 **Modeled reactions and computational approach**

499 Astrocytic  $\text{Ca}^{2+}$  signals in PAPs were simulated using the reaction-diffusion voxel-based model  
500 of ER-dependent  $\text{Ca}^{2+}$  signaling from Denizot and colleagues ([20] Table 2, Fig. 6-7). Briefly,  
501 the model describes  $\text{Ca}^{2+}$  fluxes in and out of the astrocytic cytosol. The opening of  $\text{IP}_3\text{R}$   
502 channels on the ER membrane triggers  $\text{Ca}^{2+}$  influx in the cytosol.  $\text{IP}_3$  can be synthesized  
503 by the  $\text{Ca}^{2+}$ -dependent activity of Phospholipase C (PLC)  $\delta$ .  $\text{IP}_3$  removal from the cytosol is  
504 described by a decay rate.  $\text{IP}_3\text{R}$  dynamics is derived from the De Young & Keizer's model  
505 [19]. Each  $\text{IP}_3\text{R}$  has 3 binding sites: one to  $\text{IP}_3$  and two to  $\text{Ca}^{2+}$  (activating and inhibiting).  
506 The channel can thus be in 8 different states. The open state is  $\{110\}$ :  $\text{IP}_3$  and  $\text{Ca}^{2+}$  bound  
507 to the activating sites and the  $\text{Ca}^{2+}$  inactivating site is unbound. In a subset of simulations,  
508 GCaMP6s, genetically-encoded  $\text{Ca}^{2+}$  indicators [64], were added to the cytosol and variations  
509 of  $[\text{Ca-GCaMP}]$  concentration, mimicking experimental  $\text{Ca}^{2+}$  imaging, were measured. For  
510 further details on the kinetic scheme, parameter values and model assumptions, please refer to  
511 the original paper presenting the model [20]. We slightly altered this model to better describe  
512 and control  $\text{IP}_3\text{R}$ -independent  $\text{Ca}^{2+}$  fluxes. To do so,  $\text{IP}_3\text{R}$ -independent  $\text{Ca}^{2+}$  influx was mod-  
513 eled as an influx through  $\text{Ca}^{2+}$  channels at the plasma membrane,  $\text{Ch}_{\text{PM}}$ . For simplicity, the  
514 amount of  $\text{Ch}_{\text{PM}}$  channels equals the total number of  $\text{IP}_3\text{R}$  channels,  $N_{\text{IP}_3\text{R}}$ .  $\text{Ca}^{2+}$  influx rate  
515 at  $\text{Ch}_{\text{PM}}$  channels,  $\gamma_{\text{Ch}_{\text{PM}}}$ , is  $15 \times 10^{-8} \text{ s}^{-1}$ . The reactions modeled here are illustrated in Fig. 4A.

516

517 The model was implemented using the STochastic Engine for Pathway Simulation (STEPS)  
518 python package (<http://steps.sourceforge.net/>) [36]. This software uses a spatialized version of  
519 Gillespie's SSA algorithm [31] to perform exact stochastic simulations of reaction-diffusion  
520 systems. Simulations in STEPS allow the diffusion of molecules in 3D tetrahedral meshes and  
521 onto the surfaces of the mesh, such as the ER and plasma membrane. STEPS allows volume  
522 and surface reactions. Reactions can only occur between molecules within the same tetrahe-  
523 dron (volume reactions) or in adjacent triangle and tetrahedron (surface reactions). Boundary  
524 conditions were reflective. Simulation time was 100s. The states and amounts of all molecular  
525 species were measured at each time step (1 ms).

### 526 **Neuronal stimulation simulation**

527 Unless specified otherwise, glutamatergic transmission at the synapse was modeled and oc-  
528 curred at simulation time  $t=1s$ . To do so,  $IP_3$  molecules were injected in tetrahedra below  
529 the plasma membrane of the PAP, emulating  $IP_3$  synthesis resulting from the activation of  
530 metabotropic glutamatergic receptors at the membrane of the PAP. Supplementary movie 4  
531 presents a visualization of a simulation at neuronal stimulation time, in the d2s6a9b1 PAP mesh.

### 532 $Ca^{2+}$ channel clustering algorithm

533 Surfaces correspond to triangular meshes. To simulate  $IP_3R$  clustering,  $N_{IP_3R}/\eta$   $IP_3R$  clusters  
534 were randomly positioned onto the membrane of the ER, where  $N_{IP_3R}$  is the total number of  
535  $IP_3Rs$  and  $\eta$  is the number of channels per cluster. As  $\eta$  is an integer, it must be a divider  
536 of  $N_{IP_3R}$ . As  $IP_3R$  density was kept constant across simulations,  $3.5e^{-3}/\mu m^2$  [20], the total  
537 number of  $IP_3Rs$ ,  $N_{IP_3R}$ , and  $IP_3R$  cluster size  $\eta$  varied depending on the mesh. Each  $IP_3R$   
538 cluster was located within a region of interest, as defined in STEPS, consisting in 4 triangles.  
539 Clusters could not overlap. In a subset of simulations,  $IP_3R$  clusters were located at ER-PM  
540 contact sites. To do so, ER triangles were sorted depending on the distance between their center  
541 of mass and the closest PM triangle. Cluster center was then located at the ER triangle in which  
542 no cluster was already located characterized by the lowest ER-PM distance in the mesh. The  
543 cluster ROI consisted in this cluster center triangle and the neighboring triangles. Similarly,  
544 clusters could not overlap. The number of  $IP_3R$  opening events at each cluster ROI was mea-  
545 sured at each time step.

546  
547  $IP_3R$  channels were co-clustered with  $Ch_{PM}$   $Ca^{2+}$  channels at the plasma membrane ( $cocl=1$ ),  
548 unless specified otherwise. If  $cocl=0$ ,  $Ch_{PM}$  channels were randomly distributed onto the  
549 plasma membrane. If  $cocl=1$ ,  $Ch_{PM}$  channels were co-clustered with  $IP_3Rs$ . To do so,  $Ch_{PM}$   
550 cluster center was defined as the triangle on the plasma membrane that was the closest to the  
551  $IP_3R$  cluster center triangle on the ER. The cluster ROI then consisted in this  $Ch_{PM}$  cluster  
552 center and the neighboring triangles. Similarly to  $IP_3R$  cluster ROIs,  $Ch_{PM}$  clusters could not  
553 overlap. For simplicity,  $Ch_{PM}$  cluster size was identical to  $IP_3R$  cluster size:  $\eta$ .

## 554 **Simulation code**

555 Simulations were performed using the model of  $\text{Ca}^{2+}$  signals in fine processes from Denizot  
556 and collaborators [20], available at <http://modeldb.yale.edu/247694>. The simulation code used  
557 in this study is available at <https://bit.ly/3Nc2Qin>.

## 558 $\text{Ca}^{2+}$ peak detection and characterization

559  $\text{Ca}^{2+}$  peaks were considered initiated and terminated when  $\text{Ca}^{2+}$  concentration increased above  
560 and decreased below peak threshold, respectively. Peak threshold was  $[Ca]_b + n\sigma_{Ca}$ , where  
561  $[Ca]_b$  is the basal  $\text{Ca}^{2+}$  concentration and  $\sigma_{Ca}$  is the standard deviation of the  $[Ca^{2+}]$  histogram  
562 in the absence of neuronal stimulation.  $n$  varied depending on signal/noise ratio of the simula-  
563 tion of interest, notably when measuring Ca-GCaMP signals, noisier than free  $\text{Ca}^{2+}$  signals (see  
564 (e.g.) Fig 4E).  $\text{Ca}^{2+}$  peak frequency, duration and amplitude were measured in each simulation.  
565  $\text{Ca}^{2+}$  peak duration corresponds to the time between peak initiation and termination,  $\text{Ca}^{2+}$  peak  
566 amplitude corresponds to the maximum number of  $\text{Ca}^{2+}$  ions in the cytosol measured within  
567 peak duration time and  $\text{Ca}^{2+}$  peak frequency corresponds to the amount of peaks detected dur-  
568 ing simulation time. The number of  $\text{IP}_3\text{R}$  peak opening events was recorded at each time step,  
569 in the whole cell as well as at each  $\text{IP}_3\text{R}$  cluster ROI.

## 570 **Statistical analysis**

571 Data analysis and statistics were performed using open-access and open-source software: the  
572 SciPy and pandas python libraries. Data visualization was performed using Seaborn and Mat-  
573 plotlib python libraries. Sample size for each analysis,  $n$ , is described in the figure legend.  
574 Prior to statistical analysis, normality of data distribution was inferred using the Shapiro-Wilk  
575 test. Relationship between  $\text{Ca}^{2+}$  peak characteristics and parameter values was inferred using  
576 one-way ANOVA if values followed a Gaussian distribution, Kruskal-Wallis one-way ANOVA  
577 otherwise. Note that the effect of  $\text{IP}_3\text{R}$  clustering was quantified by measuring the ratio be-  
578 tween the  $\text{Ca}^{2+}$  peak characteristic of interest measured at a given  $\text{IP}_3\text{R}$  cluster size,  $\eta > 1$   
579 and its mean value for  $\eta=1$ . The linear relationship between two datasets was evaluated using  
580 Spearman's correlation coefficient. The test and p-value,  $p$ , associated with each analysis is  
581 described in the figure legend or in the main text.

## 582 **References**

- 583 [1] Amina Aboufares El Alaoui, Molly Jackson, Mara Fabri, Luisa de Vivo, and Michele  
584 Bellesi. "Characterization of Subcellular Organelles in Cortical Perisynaptic Astrocytes".  
585 English. In: *Frontiers in Cellular Neuroscience* 14 (2021). Publisher: Frontiers.

- 586 [2] Amit Agarwal et al. “Transient Opening of the Mitochondrial Permeability Transition  
587 Pore Induces Microdomain Calcium Transients in Astrocyte Processes”. English. In:  
588 *Neuron* 93.3 (Feb. 2017), 587–605.e7.
- 589 [3] Alfonso Araque, Giorgio Carmignoto, Philip G. Haydon, Stéphane H. R. Oliet, Richard  
590 Robitaille, and Andrea Volterra. “Gliotransmitters travel in time and space”. eng. In:  
591 *Neuron* 81.4 (Feb. 2014), pp. 728–739.
- 592 [4] Alfonso Araque, Vladimir Parpura, Rita P. Sanzgiri, and Philip G. Haydon. “Tripartite  
593 synapses: glia, the unacknowledged partner”. In: *Trends in Neurosciences* 22.5 (May  
594 1999), pp. 208–215.
- 595 [5] Misa Arizono, V. V. G. Krishna Inavalli, Stéphane Bancelin, Mónica Fernández-Monreal,  
596 and U. Valentin Nägerl. “Super-resolution shadow imaging reveals local remodeling of  
597 astrocytic microstructures and brain extracellular space after osmotic challenge”. en.  
598 In: *Glia* 69.6 (2021). \_eprint: <https://onlinelibrary.wiley.com/doi/pdf/10.1002/glia.23995>,  
599 pp. 1605–1613.
- 600 [6] Misa Arizono et al. “Nanoscale imaging of the functional anatomy of the brain”. en. In:  
601 *Neuroforum* (Mar. 2021). Publisher: De Gruyter Section: Neuroforum.
- 602 [7] Misa Arizono et al. “Structural basis of astrocytic Ca<sup>2+</sup> signals at tripartite synapses”. en.  
603 In: *Nature Communications* 11.1 (Apr. 2020). Number: 1 Publisher: Nature Publishing  
604 Group, pp. 1–15.
- 605 [8] Akiko Asada et al. “Subtle modulation of ongoing calcium dynamics in astrocytic mi-  
606 crodomains by sensory inputs”. eng. In: *Physiological Reports* 3.10 (Oct. 2015).
- 607 [9] Aina Badia-Soteras et al. *Proximity of astrocyte leaflets to the synapse determines mem-  
608 ory strength*. en. Tech. rep. Section: New Results Type: article. bioRxiv, Jan. 2022,  
609 p. 2022.01.30.478393.
- 610 [10] L. H. Bergersen et al. “Immunogold Detection of L-glutamate and D-serine in Small  
611 Synaptic-Like Microvesicles in Adult Hippocampal Astrocytes”. en. In: *Cerebral Cortex*  
612 22.7 (July 2012), pp. 1690–1697.
- 613 [11] Yann Bernardinelli et al. “Activity-dependent structural plasticity of perisynaptic astro-  
614 cytic domains promotes excitatory synapse stability”. eng. In: *Current biology: CB* 24.15  
615 (Aug. 2014), pp. 1679–1688.
- 616 [12] Erika Bindocci, Iaroslav Savtchouk, Nicolas Liaudet, Denise Becker, Giovanni Carriero,  
617 and Andrea Volterra. “Three-dimensional Ca<sup>2+</sup> imaging advances understanding of as-  
618 trocyte biology”. en. In: *Science* 356.6339 (May 2017), eaai8185.
- 619 [13] Kevin Breslin et al. “Potassium and sodium microdomains in thin astroglial processes:  
620 A computational model study”. en. In: *PLOS Computational Biology* 14.5 (May 2018),  
621 e1006151.

- 622 [14] Eric A. Bushong, Maryann E. Martone, Ying Z. Jones, and Mark H. Ellisman. “Proto-  
623 plasmic astrocytes in CA1 stratum radiatum occupy separate anatomical domains”. eng.  
624 In: *The Journal of Neuroscience: The Official Journal of the Society for Neuroscience*  
625 22.1 (Jan. 2002), pp. 183–192.
- 626 [15] Corrado Cali, Jan Lopatar, Francesco Petrelli, Luca Pucci, and Paola Bezzi. “G-Protein  
627 Coupled Receptor-Evoked Glutamate Exocytosis from Astrocytes: Role of Prostaglandins”.  
628 en. In: *Neural Plasticity 2014* (Jan. 2014). Publisher: Hindawi, e254574.
- 629 [16] Corrado Cali et al. “A Method for 3D Reconstruction and Virtual Reality Analysis of  
630 Glial and Neuronal Cells”. In: *JoVE (Journal of Visualized Experiments)* 151 (Sept.  
631 2019), e59444.
- 632 [17] Corrado Cali et al. “Three-dimensional immersive virtual reality for studying cellular  
633 compartments in 3D models from EM preparations of neural tissues”. en. In: *Journal of*  
634 *Comparative Neurology* 524.1 (Jan. 2016), pp. 23–38.
- 635 [18] Tingting Cui-Wang et al. “Local Zones of Endoplasmic Reticulum Complexity Confine  
636 Cargo in Neuronal Dendrites”. en. In: *Cell* 148.1 (Jan. 2012), pp. 309–321.
- 637 [19] G. W. De Young and J. Keizer. “A single-pool inositol 1,4,5-trisphosphate-receptor-based  
638 model for agonist-stimulated oscillations in Ca<sup>2+</sup> concentration.” en. In: *Proceedings of*  
639 *the National Academy of Sciences* 89.20 (Oct. 1992), pp. 9895–9899.
- 640 [20] Audrey Denizot, Misa Arizono, U. Valentin Nägerl, Hédi Soula, and Hugues Berry.  
641 “Simulation of calcium signaling in fine astrocytic processes: Effect of spatial proper-  
642 ties on spontaneous activity”. en. In: *PLOS Computational Biology* 15.8 (Aug. 2019),  
643 e1006795.
- 644 [21] Audrey Denizot, Misa Arizono, U. Valentin Nägerl, Hédi Soula, and Hugues Berry.  
645 “Spontaneous astrocytic calcium traces from organotypic hippocampal slices”. en. In:  
646 (July 2019). Publisher: figshare Type: dataset.
- 647 [22] Audrey Denizot, Misa Arizono, Valentin Nägerl, Hugues Berry, and Erik De Schutter.  
648 “Astrocyte nanoscale morphology controls Ca<sup>2+</sup> signals at tripartite synapses”. en. In:  
649 *bioRxiv* (Feb. 2021). Publisher: Cold Spring Harbor Laboratory Section: New Results,  
650 p. 2021.02.24.432635.
- 651 [23] Audrey Denizot, Hugues Berry, and Sharmila Venugopal. “Computational Modeling of  
652 Intracellular Ca<sup>2+</sup> Signals in Astrocytes.” In: *Encyclopedia of Computational Neuro-*  
653 *science* (2019). Submitted, Submitted.
- 654 [24] Audrey Denizot, Hugues Berry, and Sharmila Venugopal. “Intracellular Calcium Signals  
655 in Astrocytes, Computational Modeling of”. en. In: *Encyclopedia of Computational Neu-*  
656 *roscience*. Ed. by Dieter Jaeger and Ranu Jung. New York, NY: Springer, 2020, pp. 1–  
657 12.
- 658 [25] Maria Amalia Di Castro et al. “Local Ca<sup>2+</sup> detection and modulation of synaptic release  
659 by astrocytes”. eng. In: *Nature Neuroscience* 14.10 (Oct. 2011), pp. 1276–1284.

- 660 [26] Carole Escartin et al. “Reactive astrocyte nomenclature, definitions, and future direc-  
661 tions”. en. In: *Nature Neuroscience* (Feb. 2021). Publisher: Nature Publishing Group,  
662 pp. 1–14.
- 663 [27] Rubén Fernández-Busnadiego, Yasunori Saheki, and Pietro De Camilli. “Three-dimensional  
664 architecture of extended synaptotagmin-mediated endoplasmic reticulum–plasma mem-  
665 brane contact sites”. en. In: *Proceedings of the National Academy of Sciences* 112.16  
666 (Apr. 2015), E2004–E2013.
- 667 [28] Joshua B. Foster et al. “Pyridazine-derivatives enhance structural and functional plasticity  
668 of tripartite synapse via activation of local translation in astrocytic processes”. eng. In:  
669 *Neuroscience* (July 2018).
- 670 [29] Christel Genoud, Charles Quairiaux, Pascal Steiner, Harald Hirling, Egbert Welker, and  
671 Graham W Knott. “Plasticity of Astrocytic Coverage and Glutamate Transporter Expres-  
672 sion in Adult Mouse Cortex”. en. In: *PLoS Biology* 4.11 (Oct. 2006). Ed. by Charles  
673 Stevens, e343.
- 674 [30] Leonidas Georgiou, Anaí Echeverría, Achilleas Georgiou, and Bernd Kuhn. “Ca<sup>+</sup> activity  
675 maps of astrocytes tagged by axoastrocytic AAV transfer”. In: *Science Advances* 8.6  
676 (2022). Publisher: American Association for the Advancement of Science, eabe5371.
- 677 [31] Daniel T. Gillespie. “Exact stochastic simulation of coupled chemical reactions”. In: *The*  
678 *Journal of Physical Chemistry* 81.25 (Dec. 1977), pp. 2340–2361.
- 679 [32] László Héja, Zsolt Szabó, Márton Péter, and Julianna Kardos. “Spontaneous Ca<sup>2+</sup> Fluc-  
680 tuations Arise in Thin Astrocytic Processes With Real 3D Geometry”. English. In:  *Fron-*  
681 *tiers in Cellular Neuroscience* 15 (2021). Publisher: Frontiers.
- 682 [33] Janosch P. Heller and Dmitri A. Rusakov. “The Nanoworld of the Tripartite Synapse:  
683 Insights from Super-Resolution Microscopy”. English. In: *Frontiers in Cellular Neuro-*  
684 *science* 11 (2017).
- 685 [34] Christian Henneberger et al. “LTP Induction Boosts Glutamate Spillover by Driving  
686 Withdrawal of Perisynaptic Astroglia”. en. In: *Neuron* (Sept. 2020).
- 687 [35] Iain Hepburn, Robert Cannon, and Erik De Schutter. “Efficient calculation of the quasi-  
688 static electrical potential on a tetrahedral mesh and its implementation in STEPS”. In:  
689 *Frontiers in Computational Neuroscience* 7 (Oct. 2013).
- 690 [36] Iain Hepburn, Weiliang Chen, Stefan Wils, and Erik De Schutter. “STEPS: efficient sim-  
691 ulation of stochastic reaction–diffusion models in realistic morphologies”. en. In: *BMC*  
692 *Systems Biology* 6.1 (2012), p. 36.
- 693 [37] Michel K. Herde et al. “Local Efficacy of Glutamate Uptake Decreases with Synapse  
694 Size”. en. In: *Cell Reports* 32.12 (Sept. 2020), p. 108182.
- 695 [38] Yixin Hu, Qingnan Zhou, Xifeng Gao, Alec Jacobson, Denis Zorin, and Daniele Panozzo.  
696 “Tetrahedral meshing in the wild”. In: *ACM Transactions on Graphics* 37.4 (July 2018),  
697 60:1–60:14.

- 698 [39] Jeffrey R. Jones et al. “Mutations in GFAP Disrupt the Distribution and Function of  
699 Organelles in Human Astrocytes”. In: *Cell Reports* 25.4 (Oct. 2018), 947–958.e4.
- 700 [40] Anne Jorstad, Biagio Nigro, Corrado Cali, Marta Wawrzyniak, Pascal Fua, and Graham  
701 Knott. “NeuroMorph: A Toolset for the Morphometric Analysis and Visualization of 3D  
702 Models Derived from Electron Microscopy Image Stacks”. en. In: *Neuroinformatics* 13.1  
703 (Jan. 2015), pp. 83–92.
- 704 [41] Conrad M. Kiyoshi et al. “Ultrastructural view of astrocyte-astrocyte and astrocyte-synapse  
705 contacts within the hippocampus”. en. In: *bioRxiv* (Oct. 2020). Publisher: Cold Spring  
706 Harbor Laboratory Section: New Results, p. 2020.10.28.358200.
- 707 [42] Natalya Korogod, Carl CH Petersen, and Graham W Knott. “Ultrastructural analysis of  
708 adult mouse neocortex comparing aldehyde perfusion with cryo fixation”. In: *eLife* 4  
709 (2015), e05793.
- 710 [43] Lubomira Lencesova, Andrea O’Neill, Wendy G. Resneck, Robert J. Bloch, and Mordecai  
711 P. Blaustein. “Plasma membrane-cytoskeleton-endoplasmic reticulum complexes in  
712 neurons and astrocytes”. eng. In: *The Journal of Biological Chemistry* 279.4 (Jan. 2004),  
713 pp. 2885–2893.
- 714 [44] Annamaria Lia et al. “Calcium Signals in Astrocyte Microdomains, a Decade of Great  
715 Advances”. English. In: *Frontiers in Cellular Neuroscience* 15 (2021). Publisher: Fron-  
716 tiers.
- 717 [45] Barbara Lykke Lind, Alexey R. Brazhe, Sanne Barsballe Jessen, Florence C. C. Tan,  
718 and Martin J. Lauritzen. “Rapid stimulus-evoked astrocyte Ca<sup>2+</sup> elevations and hemo-  
719 dynamic responses in mouse somatosensory cortex in vivo”. en. In: *Proceedings of the*  
720 *National Academy of Sciences* (Nov. 2013), p. 201310065.
- 721 [46] Irina Lushnikova, Galina Skibo, Dominique Muller, and Irina Nikonenko. “Synaptic po-  
722 tentiation induces increased glial coverage of excitatory synapses in CA1 hippocampus”.  
723 eng. In: *Hippocampus* 19.8 (Aug. 2009), pp. 753–762.
- 724 [47] Tiina Manninen, Riikka Havela, and Marja-Leena Linne. “Computational Models for  
725 Calcium-Mediated Astrocyte Functions”. In: *Frontiers in Computational Neuroscience*  
726 12 (Apr. 2018).
- 727 [48] Julie Marchaland et al. “Fast Subplasma Membrane Ca<sup>2+</sup> Transients Control Exo-Endocytosis  
728 of Synaptic-Like Microvesicles in Astrocytes”. en. In: *Journal of Neuroscience* 28.37  
729 (Sept. 2008), pp. 9122–9132.
- 730 [49] Noémie Mazaré et al. “Local Translation in Perisynaptic Astrocytic Processes Is Specific  
731 and Changes after Fear Conditioning”. en. In: *Cell Reports* 32.8 (Aug. 2020), p. 108076.
- 732 [50] Nikolai Medvedev, Victor Popov, Christian Henneberger, Igor Kraev, Dmitri A. Rusakov,  
733 and Michael G. Stewart. “Glia selectively approach synapses on thin dendritic spines”.  
734 en. In: *Phil. Trans. R. Soc. B* 369.1654 (Oct. 2014), p. 20140047.

- 735 [51] Linnaea E. Ostroff, Mustfa K. Manzur, Christopher K. Cain, and Joseph E. LeDoux.  
736 “Synapses lacking astrocyte appear in the amygdala during consolidation of Pavlovian  
737 threat conditioning”. In: *The Journal of comparative neurology* 522.9 (June 2014), pp. 2152–  
738 2163.
- 739 [52] Aude Panatier, Misa Arizono, and U. Valentin Nägerl. “Dissecting tripartite synapses  
740 with STED microscopy”. en. In: *Phil. Trans. R. Soc. B* 369.1654 (Oct. 2014), p. 20130597.
- 741 [53] Aude Panatier, Joanne Vallée, Michael Haber, Keith K. Murai, Jean-Claude Lacaille,  
742 and Richard Robitaille. “Astrocytes are endogenous regulators of basal transmission at  
743 central synapses”. eng. In: *Cell* 146.5 (Sept. 2011), pp. 785–798.
- 744 [54] Aude Panatier et al. “Glial-Derived d-Serine Controls NMDA Receptor Activity and  
745 Synaptic Memory”. In: *Cell* 125.4 (May 2006), pp. 775–784.
- 746 [55] Ilya Patrushev, Nikolay Gavrillov, Vadim Turlapov, and Alexey Semyanov. “Subcellular  
747 location of astrocytic calcium stores favors extrasynaptic neuron-astrocyte communica-  
748 tion”. eng. In: *Cell Calcium* 54.5 (Nov. 2013), pp. 343–349.
- 749 [56] Alberto Perez-Alvarez, Marta Navarrete, Ana Covelo, Eduardo D. Martin, and Alfonso  
750 Araque. “Structural and functional plasticity of astrocyte processes and dendritic spine  
751 interactions”. eng. In: *The Journal of Neuroscience: The Official Journal of the Society  
752 for Neuroscience* 34.38 (Sept. 2014), pp. 12738–12744.
- 753 [57] Andreas Reichenbach, Amin Derouiche, and Frank Kirchhoff. “Morphology and dynam-  
754 ics of perisynaptic glia”. eng. In: *Brain Research Reviews* 63.1-2 (May 2010), pp. 11–  
755 25.
- 756 [58] Sten Rüdiger and Jianwei Shuai. “Modeling of Stochastic Ca<sup>2+</sup> Signals”. en. In: *Com-  
757 putational Glioscience*. Ed. by Maurizio De Pittà and Hugues Berry. Springer Series in  
758 Computational Neuroscience. Cham: Springer International Publishing, 2019, pp. 91–  
759 114.
- 760 [59] Dmitri A. Rusakov. “Disentangling calcium-driven astrocyte physiology”. en. In: *Nature  
761 Reviews Neuroscience* 16.4 (Apr. 2015), pp. 226–233.
- 762 [60] Iaroslav Savtchouk and Andrea Volterra. “Gliotransmission: Beyond Black-and-White”.  
763 en. In: *Journal of Neuroscience* 38.1 (Jan. 2018), pp. 14–25.
- 764 [61] Alexey Semyanov, Christian Henneberger, and Amit Agarwal. “Making sense of astro-  
765 cytic calcium signals — from acquisition to interpretation”. en. In: *Nature Reviews Neu-  
766 roscience* (Sept. 2020). Publisher: Nature Publishing Group, pp. 1–14.
- 767 [62] Mark W. Sherwood, Misa Arizono, Aude Panatier, Katsuhiko Mikoshiba, and Stéphane  
768 H. R. Oliet. “Astrocytic IP3Rs: Beyond IP3R2”. English. In: *Frontiers in Cellular Neu-  
769 roscience* 0 (2021). Publisher: Frontiers.
- 770 [63] Mark William Sherwood et al. “Astrocytic IP3Rs: Contribution to Ca<sup>2+</sup> signalling and  
771 hippocampal LTP”. en. In: *Glia* 65.3 (Mar. 2017), pp. 502–513.



- 772 [64] Eiji Shigetomi, Sandip Patel, and Baljit S. Khakh. “Probing the Complexities of Astrocyte Calcium Signaling”. eng. In: *Trends in Cell Biology* 26.4 (Apr. 2016), pp. 300–312.  
773
- 774 [65] Eiji Shigetomi, Kozo Saito, Fumikazu Sano, and Schuichi Koizumi. “Aberrant Calcium Signals in Reactive Astrocytes: A Key Process in Neurological Disorders”. en. In: *International Journal of Molecular Sciences* 20.4 (Jan. 2019), p. 996.  
775  
776
- 777 [66] Eiji Shigetomi et al. “Imaging calcium microdomains within entire astrocyte territories and endfeet with GCaMPs expressed using adeno-associated viruses”. eng. In: *The Journal of General Physiology* 141.5 (May 2013), pp. 633–647.  
778  
779
- 780 [67] Ian F. Smith, Steven M. Wiltgen, Jianwei Shuai, and Ian Parker. “Ca<sup>2+</sup> Puffs Originate from Preestablished Stable Clusters of Inositol Trisphosphate Receptors”. en. In: *Sci. Signal.* 2.98 (Nov. 2009), ra77–ra77.  
781  
782
- 783 [68] Rahul Srinivasan et al. “Ca(2+) signaling in astrocytes from Ip3r2(-/-) mice in brain slices and during startle responses in vivo”. eng. In: *Nature Neuroscience* 18.5 (May 2015), pp. 708–717.  
784  
785
- 786 [69] Jillian L. Stobart et al. “Cortical Circuit Activity Evokes Rapid Astrocyte Calcium Signals on a Similar Timescale to Neurons”. In: *Neuron* 98.4 (May 2018), 726–735.e4.  
787
- 788 [70] Min-Yu Sun et al. “Astrocyte calcium microdomains are inhibited by Bafilomycin A1 and cannot be replicated by low-level Schaffer collateral stimulation in situ”. In: *Cell Calcium* 55.1 (Jan. 2014), pp. 1–16.  
789  
790
- 791 [71] Taufiq-Ur-Rahman, A. Skupin, M. Falcke, and C. W. Taylor. “Clustering of InsP3 receptors by InsP3 retunes their regulation by InsP3 and Ca<sup>2+</sup>.” eng. In: *Nature* 458.7238 (Apr. 2009), pp. 655–659.  
792  
793
- 794 [72] Nagendra Babu Thillaiappan, Alap Chavda, Stephen Tovey, David Prole, and Colin Taylor. “Ca<sup>2+</sup> signals initiate at immobile IP3 receptors adjacent to ER-plasma membrane junctions”. In: *Nature Communications* 8 (Dec. 2017).  
795  
796
- 797 [73] Alexei Verkhratsky and Maiken Nedergaard. “Astroglial cradle in the life of the synapse”. In: *Philosophical Transactions of the Royal Society B: Biological Sciences* 369.1654 (Oct. 2014).  
798  
799
- 800 [74] Alexei Verkhratsky and Maiken Nedergaard. “Physiology of Astroglia”. eng. In: *Physiological Reviews* 98.1 (Jan. 2018), pp. 239–389.  
801
- 802 [75] Alexei Verkhratsky, Alexey Semyanov, and Robert Zorec. “Physiology of Astroglial Excitability”. In: *Function* 1.zqaa016 (Sept. 2020).  
803
- 804 [76] Jürgen Wenzel, Grit Lammert, Ursula Meyer, and Manfred Krug. “The influence of long-term potentiation on the spatial relationship between astrocyte processes and potentiated synapses in the dentate gyrus neuropil of rat brain”. en. In: *Brain Research* 560.1 (Sept. 1991), pp. 122–131.  
805  
806  
807

- 808 [77] Steven M. Wiltgen, Ian F. Smith, and Ian Parker. “Superresolution localization of single  
809 functional IP3R channels utilizing Ca<sup>2+</sup> flux as a readout”. eng. In: *Biophysical Journal*  
810 99.2 (July 2010), pp. 437–446.
- 811 [78] Mark R. Witcher, Sergei A. Kirov, and Kristen M. Harris. “Plasticity of perisynaptic  
812 astroglia during synaptogenesis in the mature rat hippocampus”. eng. In: *Glia* 55.1 (Jan.  
813 2007), pp. 13–23.
- 814 [79] Haoxi Wu, Pedro Carvalho, and Gia K. Voeltz. “Here, there, and everywhere: The impor-  
815 tance of ER membrane contact sites”. en. In: *Science* 361.6401 (Aug. 2018), eaan5835.
- 816 [80] Minnie M. Wu, JoAnn Buchanan, Riina M. Luik, and Richard S. Lewis. “Ca<sup>2+</sup> store  
817 depletion causes STIM1 to accumulate in ER regions closely associated with the plasma  
818 membrane”. en. In: *J Cell Biol* 174.6 (Sept. 2006), pp. 803–813.
- 819 [81] S. Zeller et al. “Modeling of the modulation by buffers of Ca<sup>2+</sup> release through clusters  
820 of IP3 receptors”. eng. In: *Biophysical Journal* 97.4 (Aug. 2009), pp. 992–1002.

## 821 **Acknowledgements**

822 This work used the computing resources of the Scientific Computing and Data Analysis section  
823 from the Okinawa Institute of Science and Technology. We thank Iain Hepburn and Weiliang  
824 Chen of the Computational Neuroscience Unit, OIST, Okinawa, Japan for discussion and advice  
825 on STEPS software. We thank Pierre Magistretti (KAUST, Thuwal) for the financial support to  
826 MFVC and to CC, and Graham Knott for kindly sharing the original EM dataset.

827 The work from AD was supported by a JSPS (Japan Society for the Promotion of Science)  
828 Standard Postdoctoral Fellowship for Research in Japan (21F21733).

829  
830 **Author Contributions** AD, CC and EDS conceived the research, provided resources and su-  
831 pervised the work. AD and EDS designed the analysis. MFVC and CC curated the electron  
832 microscopy data. PP wrote the Blender code for automatic 3D mesh handling. AD did the com-  
833 putational modeling work. AD performed data analysis and visualisation of experimental and  
834 computational data. AD wrote the first draft of the manuscript. All authors read and reviewed  
835 the manuscript.

836  
837 **Competing Interests Statement** The authors declare that they have no competing financial in-  
838 terests.

839