Two-Photon Optogenetic Mapping of Excitatory Synaptic Connectivity and Strength

HIGHLIGHTS
- Two-photon optogenetic mapping of excitatory connectivity and strength in neocortex
- Identification of connected neurons in acute slices through numerical optimization
- Synaptic delays align with location of connected presynaptic cell
- Confirmation of predicted connections by dual patch-clamp recordings

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Two-Photon Optogenetic Mapping of Excitatory Synaptic Connectivity and Strength

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SUMMARY

The development of optical methods to activate neurons with single-cell resolution has enabled systematic mapping of inhibitory connections. In contrast, optical mapping of excitatory connections between pyramidal neurons (PCs) has been a major challenge due to their high densities in cortical tissue and their weak and stochastic connectivity. Here we present an optogenetic two-photon mapping method in mouse neocortical slices by activating PCs with the red-shifted opsin C1V1 while recording postsynaptic responses in whole-cell configuration. Comparison of delays from triggered action potentials (APs) with those from synaptic inputs allowed us to predict connected PCs in three dimensions. We confirmed these predictions with paired recordings, and used this method to map strong connections among large populations of layer 2/3 PCs. Our method can be used for fast, systematic mapping of synaptic connectivity and weights.

INTRODUCTION

Decoding the structural and functional principles of microcircuitry appears necessary to understand neural computations. Most experimental studies of connectivity and strength of connections have employed multiple simultaneous patch-clamp recordings, either in slices or in vivo (e.g., Feldmeyer et al., 2006; Lefort et al., 2009; Hofer et al., 2011; Ko et al., 2011; Jiang et al., 2013; Rieubland et al., 2014; Jiang et al., 2015; Jouhanneau et al., 2015). Although this method provided great insights, it is limited by the amount of neurons that can be recorded from simultaneously. Studying connectivity at the anatomical level, employing, e.g., tracing using a modified rabies virus (Wickersham et al., 2007; Ranz et al., 2011; Sun et al., 2014; Cocas et al., 2016), synapse labeling by mGRASP (Kim et al., 2012; Druckmann et al., 2014), RNA barcoding (Kebschull et al., 2016), or electron microscopic reconstructions (Kasthuri et al., 2015), enables probing of circuitry at a larger scale, yet provides limited information about the function, or strength, of a synaptic connection, and often requires significant analytical processing. Mapping the functionally and strength of synaptic connections is necessary to decipher the functional patterns of a microcircuit, since this cannot be derived from structural connectivity alone. In addition, ideally, functional mapping methods should be usable in living tissue, facilitating experimental approaches such as pharmacological testing, learning or plasticity paradigms, or closed-loop optogenetics or optochemistry.

Using optical tools to map neural connections with single-cell precision is a promising road to perform systematic physiological experiments to probe connectivity between hundreds or thousands of neurons. Indeed, two-photon uncaging of caged glutamate has served as a powerful tool to map the connectivity from hundreds of interneurons to pyramidal neurons (PCs) in the neocortex, revealing a dense, unspecific connectivity (Fino and Yuste, 2011; Packer and Yuste, 2011). Similar studies have employed this method to map excitatory connections between PCs (Nikolenko et al., 2007) or stellate cells of layer 4 (Ashby and Isaac, 2011), but these efforts either remained unproven about their precision, or mapped circuitry at relatively small scales. At the same time, the evidence of specific microcircuitry between excitatory neurons (reviewed in Harris and Mrsic-Flogel, 2013) calls for precise, large-scale connectivity studies between PCs. In addition, establishing and refining the tools needed for such projects would also contribute to the development of methods to precisely map or influence brain activity in neuroscience in general (Alivisatos et al., 2013).

Although glutamate uncaging has been successfully used for functional mapping of connections (Callaway and Katz, 1993; Nikolenko et al., 2007), optogenetic stimulation is desirable since it would enable to map connections among genetically specified cell types. Also, although optogenetic tools have been used to map circuits in two dimensions using conventional, one-photon, laser excitation (Petreanu et al., 2009), circuit mapping with single-cell precision and in three dimensions appears necessary. This is possible...
employing the non-linear excitation of ultrafast lasers, but such methods are still difficult. This is partly because evoking APs through conventional channelrhodopsin 2 needs highly specialized optics (Rickgauer and Tank, 2009; Andrasfalvy et al., 2010; Papagiakoumou et al., 2010; Baker et al., 2016). To partly supersede these problems, the slower opsin C1V1 was developed as a tool to optogenetically activate neurons with two-photon excitation (Packer et al., 2012; Prakash et al., 2012). Using this opsin to evoke APs with conventional two-photon scanning, we introduce here a mapping method of PC connectivity in slices of mouse visual cortex. By using the latency of AP generation in response to laser stimulation as an indication of potentially connected presynaptic neurons, we overcome the problem that neurons with different opsin expression levels have different excitation properties, and we predict putative connections and confirm them with paired recordings. We demonstrate the capability of our method by mapping stronger connections in mouse V1. Future refinements of this method could allow mapping weaker connections and using the technique in vivo.

RESULTS
The aim of our study was to develop an optical method to map connectivity between PCs in acute cortical brain slices with single-cell precision. To do so, we first expressed the opsin construct C1V1 in combination with enhanced yellow fluorescent protein (C1V1-p2A-EYFP) in excitatory neurons (under control of the CamKII promoter) in mouse visual cortex via transcranial adeno-associated virus (AAV) injection into neonatal mice. Densely packed expressing neurons could be identified by two-photon-imaging of EYFP at 940 nm. To optically activate neurons, we employed a custom-built two-photon microscope capable of consecutively visiting a pre-programmed set of point stimulations (for stimulation pattern see Transparent Methods, Supplemental Information). For C1V1 stimulation, the Ti:Sapphire laser was set to 1,040 nm.

Calibrating Optical Stimulation Parameters
We characterized stimulation conditions by performing extracellular cell-attached patch-clamp recordings of C1V1-p2A-EYFP-expressing layer 2/3 PCs (Figures 1A and 1B). We first explored different power levels (5, 7, and 17 mW) resulting in 61%, 72%, and 82% of neurons firing, respectively. Overall, we found a clear dependency of the number of evoked APs, as well as the latency of the first AP, on laser power on sample (latency: 5 mW: 38 ± 4 ms, n = 12; 7 mW: 31 ± 3 ms, n = 28; 17 mW: 27 ± 6 ms, n = 9; APs per stimulation: 5 mW: 1.1 ± 0.1, n = 12; 7 mW: 1.3 ± 0.1, n = 28; 17 mW: 1.7 ± 0.3, n = 9; Figures 1B and 1C). As expected, AP delay and number of evoked APs (7 mW on sample) were strongly correlated with EYFP fluorescence of the recorded neuron (r = −0.67, p = 0.0063 and r = 0.51, p = 0.018, respectively), indicating that higher expression of the opsin construct favors faster and stronger activation of a given neuron (Figures 1D–1F). This may result in less accurate photostimulation in a heterogeneously expressing population, as high expression level could potentially lead to AP generation, not only when targeting the soma, but also when targeting the dendritic tree. To avoid this, we chose 7 mW at the sample for all further experiments, as this value resulted in AP evoking in most tested neurons, yet was still relatively low to minimize the probability of evoking APs at sites away from the soma.

To test the spatial resolution of our system with this power, the excitation pattern was moved laterally from the center of the cell soma in steps of 10 μm. AP probability decreased strongly, and latency increased, with distance (latency: r0: 31 ± 6 ms, n = 15; r10: 49 ± 5 ms, n = 10; r20: 61 ± 6 ms, n = 8; r30: 80 ± 10 ms, n = 7; AP per stimulation: r0: 1.1 ± 0.1, r10: 0.5 ± 0.1, r20: 0.3 ± 0.1, r30: 0.3 ± 0.1; n = 14 in all, Figures 2A and 2B). We also observed that stimulations that evoked APs at longer distances from soma were temporally less reliable, as can be seen by an increase of delay standard deviation in Figure 2C. In fact, nearly all stimulations performed at the center of the cell soma resulted in delays with standard deviation below 18 ms. Because of this, we used 18 ms as a threshold. Excluding AP responses with a delay standard deviation higher than 18 ms across repeats resulted in improved lateral resolution (AP per stimulation: r0: 1.1 ± 0.1, r10: 0.13 ± 0.06, r20: 0.1 ± 0.07, r30: 0.1 ± 0.1; n = 10 in all, Figure 2B, gray traces, and Figure 2D). Thus, we concluded that resolution in the lateral plane, improved by variability restriction across repeats, would be sufficient to reliably map synaptic connections. Testing axial resolution by changing the focus of the excitation pattern revealed a similar behavior for AP latency and number, yet displayed a less steep drop in AP generation probability with distance from the soma than observed for lateral resolution (Figures 2E–2G). The delay standard deviation remained relatively low, even at high axial distances from the soma (Figure 2H). We did not observe a dependency between delay standard deviation and expression level or cell depth (r = 0.24,
p = 0.33, n = 18 and r = 0.21, p = 0.41, n = 18, respectively). We felt that lowering the laser power to improve axial resolution would result in a reduction of the number of cells we can activate. We further characterized the change in AP delay in the axial dimension with more detail. This effect was similar across cells, and it could be fitted with a parabolic function, with the minimum centered on the soma position (Figure 2G). These results demonstrate that AP delays indicate the location of the soma of the optically targeted neuron as the point of shortest delay. We also performed stimulations at different axial positions when moving the excitation away from the cell soma in the lateral dimension. The probability of fitting the parabolic function described above (χ² test with an acceptance threshold of 0.1, see below) was much lower compared to the soma position (probability of fitting: r0: 1 ± 0, n = 6; r10: 0.08 ± 0.04, n = 48; r20: 0.06 ± 0.04, n = 48; r30: 0 ± 0, n = 48 stimulations, Figures 2I and 2J). This suggested that by fitting the delay of APs to an axial stimulation position, one could distinguish APs evoked by dendritic stimulation or background activity (which do not follow the parabolic function) from those evoked by somatic stimulation. Thus, using the two established criteria based on AP delay, (1) limiting to 18 ms or less standard deviation across repeats and (2) fitting its dependence on axial position to calibration data.

Figure 1. Two-Photon Stimulation of C1V1-Expressing Neurons

(A) Left: differential interference contrast image of a layer 2/3 PC recorded in cell-attached configuration. Right: fluorescent image, showing in green expression of C1V1-p2A-EYFP and in magenta Alexa 594 from the pipette solution. White dots represent the pattern of point stimulations performed to excite a neuron. Scale bar, 10 μm.

(B) Example of a response of an EYFP-expressing neuron to a stimulation during cell-attached current recording. Stimulation, red bar, lasts for 54 ms, 1,040 nm, and 7 mW on sample. To characterize the efficiency of stimulation we analyzed the delay of the first AP and the number of APs.

(C) Dependence of mean delay (filled circles) and number of APs (empty circles) on laser power at the sample. In all cases, only neurons that fired after stimulation were considered; n = 12 for 5 mW, n = 28 for 7 mW, and n = 9 for 17 mW. Dots depict mean ± S.E.M.

(D) Distribution of normalized fluorescence intensity of EYFP from cell-attached patched neurons, n = 21 cells.

(E) Plot of AP delay against normalized EYFP fluorescence intensity. Negative correlation is observed, n = 15 cells. Dotted line depicts the linear fitting of the data.

(F) Plot of the number of APs triggered during one stimulation against normalized EYFP fluorescence intensity. Positive correlation is observed, n = 21 cells. Dotted line depicts linear fitting of the data.
A. Diagram showing a network of neurons with labeled distances and time intervals. 

B. Graph showing delay (ms) vs. radius (μm) with error bars. 

C. Graph showing delay standard deviation (ms) vs. radius (μm). 

D. Scatter plot showing x radius (μm) vs. y radius (μm) with color indicating #AP/stim @ z=0. 

E. Graph showing z position (μm) vs. z position (μm) with #AP/stim. 

F. Graph showing AP Delay (ms) vs. z position (μm). 

G. Graph showing probability of fit to Z calibration vs. z position (μm). 

H. Graph showing delay standard deviation (ms) vs. z position (μm). 

I. Graph showing y radius (μm) vs. z position (μm). 

J. Graph showing probability of fit to Z calibration vs. radius (μm).
could in combination be used to identify the soma of a stimulated cell in three dimensions. We used these observations to establish a mapping strategy that reliably identified the exact position of presynaptically connected neurons to the recorded cell.

**Mapping Excitatory Synaptic Microcircuits**

We then explored the application of this photostimulation method to map synaptic microcircuits. To record excitatory synaptic inputs in postsynaptic cells we performed whole-cell patch-clamp recordings of layer 2/3 PCs in slices expressing C1V1-p2A-EYFP, patching non- or low-expressing neurons (to minimize optically stimulating them), localized at 24 to 66 μm depth from the slice surface. Since the internal solution contained Alexa 594, we could visualize the recorded cell via two-photon imaging at 800 nm in the red channel and the expressing neurons surrounding it in the green channel while imaging at 940 nm. We recorded images of 400 × 300 μm in stacks with a step size of 6 μm, spanning 24 to 60 μm. In such a volume, we then manually located 70–200 C1V1-p2A-EYFP-expressing neurons using custom-written MATLAB code. A non-neighbor path in which the neurons would be stimulated was generated (see Transparent Methods, Supplemental Information, for details). In this process, we treated the volume as a 2D image, projecting the z axis down into one plane (Figure 3A). We then performed sequential stimulation of all cell locations at the more superficial end (in z) of the stack and repeated it 5–10 times. Subsequently, we moved 6 μm in z toward the other end of the stack, repeated the stimulations again, and moved again 6 μm in z until we reached the end of the imaged stack. This resulted in each targeted cell being stimulated not only in the focal plane of its soma but also above and below it throughout the stack, performing essentially a resolution test in z on each of the targeted neurons.

When analyzing recorded synaptic events that occurred during mapping in the patch-clamped neuron, we could clearly observe cases of putative responses from activated presynaptic neurons (see an example in Figures 3B and 3C). To distinguish spontaneous excitatory postsynaptic currents (EPSCs) from ones evoked by our photostimulation, we selected synaptic inputs time-locked to stimulation across repeats according to the calibration data as per the following procedure: (1) we selected those EPSCs with a delay standard deviation lower than 18 ms within the same z position and (2) discarded them if success rate was equal or smaller than 0.2. Latencies of synaptic delays were then compared with the parabolic function of AP latencies obtained in the calibration experiments (Figure 2G) with the lowest point of the fit being within 6 μm of the targeted soma (Figure 3D). We used a χ² test with an acceptance threshold of 0.1 to compare the delay characteristics of evoked EPSCs with the calibration curve. This threshold excluded targeting sites that did not follow the calibration curve, but still included connections with delay characteristics that, e.g., contained an outlier (Figures 4A and 4B). Using these criteria, we identified targeted neurons as being putatively connected to the recorded neuron (Figures 3C and 3D, target #22 and #97) as well as sites at which targeting in planes lower or higher than the soma location resulted in EPSCs that did
Figure 3. Identifying Putatively Connected Presynaptic Neurons
(A) Image of C1V1-p2A-EYFP-expressing cells in layer 2/3. Sum projection of a z stack of images taken at 940 nm, ranging from 42 μm (on the more superficial side of the slice) down to the focus of the patched cell, in Δz of 6 μm. We selected EYFP-expressing cells for mapping stimulations (red circles, 148 targets). Imaged area is 400 × 300 μm; gray line indicates pia at the top right corner.
(B) Magnification of the mapping area with the localization of some of the stimulation targets, numbered according to the stimulation path, at the different z positions focused during the experiment. The patched neuron is located at z0, and the cell and pipette positions are indicated with a dotted line. Positive z plane values indicate focus on positions superficial to the recorded cell. Scale bar, 20 μm.
(C) Whole-cell voltage-clamp recording of the postsynaptic neuron during stimulation (red area) of 4 representative targeted cells (target numbers 22, 87, 97, and 111). At each z position, the stimulation of all targets was repeated 5 times, repeats are grouped and colored in different tones of gray. Target #22 and...
Figure 3. Continued

#97 are identified as connections as shown in panel D. Gray area depicts time after stimulation considered for the detection of synaptic events (100 ms after the start of stimulation, see Transparent Methods, Supplemental Information, for details), as we were probably not able to evoke APs in the second (presynaptic) cell via current injections resulting in synaptic currents in the first (postsynaptic) cell. In all cases, we were able to verify our prediction in layer 2/3 (Figure 5). In addition, we performed one more successful confirmation in a layer 4 mapping experiment (data not shown), for a total of 6/6. In all cases, evoking APs in the second (presynaptic) cell via current injections resulted in synaptic currents in the first (postsynaptic) cell.

Having proven the validity of our technique, we went on to build presynaptic connectivity maps of neurons in layer 2/3 of primary visual cortex as a proof-of-principle demonstration of the capabilities of our technique. To reduce false-negatives, we excluded targeted neurons with low EYFP fluorescence (7% of targeted neurons, see Transparent Methods, Supplemental Information, for details), as we were probably not able to evoke APs in these, due to low C1V1 expression (Figure 1F). We found EPSCs satisfying the parabola fit criteria in 15 mapping experiments (Figure 4). In those, 27 of 1,729 targeting sites were identified as presynaptically connected neurons. Due to the overall low rate of connectivity (1.6%), partly owing to our strict conservative criteria to ensure that every putative connection was correctly predicted, we pooled all maps into one (Figures 6A and 6B), centering them on the postsynaptic cell. For further analysis, we excluded border regions of the map, because these were mapped at low density. Interestingly, even though the space close to the recorded cell (up to about 50 μm) was mapped at high density, we observed large patches without connections. Connectivity was most robust at about 50–130 μm away from the soma (Figures 6A and 6B). Connection probability (analyzed in bins of 10 μm; Figure 6C) showed no correlation with overall distance from the soma of the recorded neuron (r = 0.34, p = 0.13; Figure 6D), but was correlated with distance when only the lateral (horizontal) distance was taken into account (r = −0.61, p = 0.03). The vertical distance from soma was not correlated with connection probability (r = 0.35, p = 0.13). Peak amplitude of synaptic responses (Figure 6B) was neither correlated with absolute distance from soma (r = −0.16, p = 0.42; Figure 6E) nor with lateral distance (r = −0.28, p = 0.16) or vertical distance (r = −0.08, p = 0.71).

DISCUSSION

In this work, we present a strategy to precisely map connectivity and synaptic strength between PCs in cortical brain slices. For this we use two-photon optical activation of neurons expressing the red-shifted opsin C1V1. Employing a spiral-like point scanning, we demonstrate an inverse correlation between the expression level of C1V1 and latency of APs generated. In addition, AP latency also increased when moving the laser away from the cell soma, both in the lateral and axial planes. We took advantage of a parabolic fit of this latency distribution to accurately map excitatory synaptic connections. By moving the laser to and away from optically targeted cell somata in the axial dimension during the mapping process, and
comparing the latency changes of evoked synaptic inputs with those obtained in calibration experiments, we were able to identify connections between PCs. We verified this method by performing targeted patch-clamp recordings of the putative presynaptically connected neuron. In all 6 tested cases, including one performed in layer 4, the predicted connections were verified.

These results demonstrate the ability to precisely map synaptically connected neurons using two-photon optogenetics, even in conditions in which single-cell excitation is technically difficult to obtain. This is often the case in densely packed neuronal populations with heterogeneous opsin expression, because weak expressing neurons will need a certain level of laser intensity to fire, whereas stronger expressing neurons will, at such intensities, also fire APs when their dendrites or axons are targeted. Our strategy, implementing a detailed analysis of the AP delay characteristics, demonstrates that computational approaches can overcome some of these experimental limitations. Other strategies to approach this problem are possible. For example, using temporal-focused excitation or holographic stimulation can limit the problem of dendritic activation (Baker et al., 2016; Shemesh et al., 2017; Forli et al., 2018; Mardinly et al., 2018). However, our method uses relatively simple point stimulations, which, as opposed to complex scanning patterns, can be performed with most conventional laser scanning microscopes. Also, improvements in opsins may enable more selective optogenetic excitation of neurons. In fact, key to two-photon optogenetics was
the red-shifted, slowly closing opsin C1V1, designed precisely for two-photon neuronal activation (Packer et al., 2012; Prakash et al., 2012). Indeed, C1V1 has recently been successfully employed in in vivo studies using simultaneous two-photon imaging and activation (Packer et al., 2015; Carrillo-Reid et al., 2016; Yang et al., 2018). In addition, targeting opsin expression to the neuron somata, in combination with some of these optical strategies, might lead to reliable mapping techniques and perhaps enable implementation of our optical mapping technique in vivo. Finally, for mapping synaptic strengths, data collection with a stable access resistance will be especially important.

We used our mapping protocol to perform the first preliminary study of PC connectivity in the visual cortex of juvenile to young adult mice. We highlight that, to preserve the accuracy of the predicted connections, we imposed severely conservative criteria that only enabled us to optically map the strongest and most reliable synaptic connections. This may explain our overall low connection probability, as we are likely discarding the majority of connections that could be weak or unreliable (see Limitations of the Study). Nevertheless, with this caveat, we found in our dataset connection probability negatively correlated with the lateral (horizontal) distance between tested cells, but found no correlations of the strength of synaptic connections with intersomatic distance. These results, generated from a relatively low number of obtained maps, should be seen as an initial demonstration of the capability of our technique.

Conclusions about the connectivity rules between PCs of the neocortex, mostly studied by using multiple patch-clamp recordings, have been diverse. In layer 2/3 of the barrel cortex, connection probability, tested for up to 150 μm, was reported to not depend on intersomatic distance (Lefort et al., 2009; Avermann et al., 2012). In contrast, connection probability, but not connection strength, appears to be negatively correlated with distance in layer 2/3 of the auditory cortex (Oswald and Reyes, 2008; Levy and Reyes, 2012). Holmgren et al. (2003) also reported distance dependence of connectivity (and possibly also synaptic strength) in visual and somatosensory cortex. In an in vivo study, Kwan and Dan (2012) imaged the activity of a group of neurons while inducing bursts in a single neuron close by. Neurons close to the activated cell were more likely to display neuronal activity than those farther away, arguing for a network favoring close-range connections over medium-range ones, whether on the level of absolute connectivity or strength of connections. These differences in experimental results are probably partly explainable by differences of ages, species, and cortical areas used. These considerations might also partly explain the low connection probability we calculated from our mapping data (1.6%). Also, studies in acute slices usually underestimate connectivity rates, due to the processes of superficial neurons being cut (Levy and Reyes, 2012). The studies mentioned above, and including those by Jouhanneau et al. (2015) and Ko et al. (2011), reported connectivity rates ranging from 6% to 19%. We actually achieve similar connectivity rates when mapping at –50 to –70 μm depths in the slice (Figure 4E), a depth that probably corresponds to conditions used in patch-clamp studies. Our present study, however, also included neurons that were located more superficially, as we wanted to explore the capabilities of our method throughout the complete observed volume. Thus, the low observed connectivity rate could to some extent be a result of slicing artifacts. Nevertheless, our calculated connectivity rate is in remarkable agreement with the value of 1.8% in the visual cortex of mice reported in Jiang et al. (2015), although this study used older mice then we did. It needs to be stressed that we did not limit our sampling to the region immediately adjacent to the patched neurons, as is often done in dual patch recordings, but stimulated neurons across the entire field of view. This further limits the comparability of our connection probability calculation to prior studies. However, the distribution of distances between soma tested for connectivity in Jiang et al. (2015) (see Figure S13B, right panel, of the article) is actually very similar to those tested in our study (see Figure 6C).

Although one has to be aware of the limitations of our technique in its current state (see below), our results can be viewed as an important step in the design of optical methods to decipher neuronal connectivity.
Figure 6. Mapping Layer 2/3 Connectivity among Pyramidal Cells
(A) Two-dimensional projection of all identified connections pooled from all mapping experiments. The recorded postsynaptic cell is depicted by the gray triangle. Red dots mark the presynaptically connected neurons. The background grid has a spacing of 10 µm. The color code depicts the number of targeted cells pooled from all experiments. Medial is to the left. The pia is located at the top. Orange dotted line indicates the area used for further analysis.

(B) Cumulative frequency of targeted and connected cells as a function of distance from soma. The linear trend lines represent the correlation coefficients and their p-values.

(C) Distribution of peak amplitude (pA) as a function of distance from soma, lateral distance from soma, and vertical distance from soma.
Such methods are necessary especially for in vivo experiments, in which space for using multiple patch-clamp electrodes can be limited, and thus reaching all neurons in the field of view is very challenging. In addition, even if testing multiple connections sequentially with the same electrode in cell-attached configuration (Barbour and Isope, 2000), moving the electrode around in the brain tissue, or moving it in and out several times, will strongly impair the health of the tested circuits.

Close to the cell soma, we often find patches without connections, despite a high mapping density. At the same time, connection probability does, within our tested volume, not depend on the total intersomatic distance. As pointed out above, the number of experiments needs to be increased to draw final conclusions. However, this observation is in contrast to the connectivity of parvalbumin-positive interneurons on pyramidal cells, which is the highest at short intersomatic distance and declines with higher distance (Packer and Yuste, 2011). This would favor the view that connectivity between excitatory neurons in the cerebral cortex is, in contrast to most inhibitory projections (reviewed in Karnani et al., 2014), non-random and forms specific networks to perform distinct computational tasks. Evidence for this is the observation that excitatory neurons in visual cortex are more likely to form strong connections when their response to sensory stimuli is correlated (Cossell et al., 2015). Also, bidirectional connections are overrepresented in excitatory cortical networks (Jouhanneau et al., 2015) and the output target of cortical neurons is related to the patterns of their intercortical connectivity (Brown and Hestrin, 2009). Wertz et al. (2015) performed in vivo activity imaging of the network presynaptic to a single layer 2/3 PC and found layer-specific modules of motion direction preferences, either aligned or shifted to the postsynaptic neuron. Additional work needs to be done to understand the rules of connectivity in the cerebral cortex, including structural and functional aspects. The mapping technique presented in this work may help to tackle these questions, which are critical to decipher the functional logic of cortical circuitry.

Limitations of the Study
As mentioned in the Discussion, our method, although has the advantage of enabling the testing of hundreds of connections in one experiment, is still not suited to rigorously identify weak synaptic connections or those with high failure rates. Multiple patch-clamp recordings can overcome this limitation by averaging a high number of repeats to the onset of the electrically induced AP. However, as optical activation often has some jitter, with our method one needs a relatively reliable synaptic response, with a clearly detectable postsynaptic signal, to identify a cell as being putatively connected. Due to this limitation, our maps likely favor stronger synaptic connections. We purposefully set high threshold criteria to define putative connections, to avoid false-positives, as confirmed by our dual recordings. This strategy, however, increases the false-negative rate and focuses on identifying strong synaptic connections, which have to be taken into account when considering the observed connectivity rate of 1.6%. With regard to mapping experiments, performing experiments in deeper tissue will reduce underestimation of connectivity rates. In addition, further data collection will be especially important for the analysis of synaptic strength, which can be greatly influenced by access resistance.

METHODS
All methods can be found in the accompanying Transparent Methods supplemental file.
SUPPLEMENTAL INFORMATION

Supplemental Information includes Transparent Methods and can be found with this article online at https://doi.org/10.1016/j.isci.2018.09.008.

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AUTHOR CONTRIBUTIONS

J.J.H. and R.Y. conceived the original idea. M.I.-S., J.J.H., and B.S. designed the experiments. J.J.H. conducted the initial experiments. M.I.-S. conducted the majority of experiments, M.I.-S. and J.J.H. performed analysis and wrote the manuscript. B.S. contributed custom-written code. R.Y. assembled and directed the team, guided the project, edited the manuscript, and provided laboratory space, equipment, and funding.

DECLARATION OF INTERESTS

The authors declare no competing interests.

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Supplemental Information

Two-Photon Optogenetic Mapping
of Excitatory Synaptic Connectivity and Strength
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Supplemental Information

Transparent Methods

Animal handling and experimentation were done according to the US National Institutes of Health and Columbia Institutional Animal Care and Use Committee guidelines. Animals of both sexes were used and were housed and maintained in a temperature-controlled environment on a 12-h light-dark cycle, with *ad libitum* food and water, in a Columbia University Animal Facility.

C57BL/6 mice aged postnatal day (P)0 to P1 were injected with AAV-DJ-CaMKIIa-C1V1(E162T)-TS-P2A-eYFP-WPRE or AAV8(Y733F)-CaMKIIa-C1V1(E162T)-TS-P2A-eYFP-WPRE at concentrations of 1.8·10^{13}–5.5·10^{13} GC·mL^{-1} (Neuroscience Gene Vector and Viral Core, Stanford, CA). Pups were anesthetized using hypothermia. They were placed on aluminum foil on a cooling block placed on crushed ice. Viral injections were made using a 10 µL PCR glass pipette (Drummond Scientific Company, PA) pulled to a sharp micropipette and attached to a custom made manipulator. After penetration of the pup's skin and skull, 400 nL of virus containing solution were expelled into left visual cortex using repetitive 30 ms pulses of 7-8 psi applied with a Picospritzer II (Parker Hannifin, NJ). After retracting the pipette, animals were rewarmed on a circulating warm water blanket. Before returning to the home cage, pups were placed on the cage/dam bedding to facilitate acceptance by the dam due to familiar smell.

On day 18-34 after virus injection, coronal sections of the neocortex of the injected animals were prepared using a Leica VT1200S vibratome. Following deep anesthesia via inhalation of isoflurane, the animal was decapitated, and the brain quickly removed. Slices of 300 µm thickness were prepared in ice-cold slicing solution containing (in mM): 93 N-Methyl-D-glucamine, 2.5 KCl, 1.2 NaH2PO4, 30 NaHCO3, 20 HEPES, 25 glucose, 5 Na-ascorbate, 3 Na-pyruvate, 10 MgSO4, 0.5 CaCl2, pH adjusted with HCl to 7.3, bubbled with 95% O2 and 5% CO2 (modified after Ting et al., 2014). After a short recovery period (4-8 min) in 35-37 °C warm slicing solution, slices were kept at room temperature in artificial cerebral spinal fluid (ACSF) until transferred into a recording chamber. ACSF contained (in mM): 126 NaCl, 26 NaHCO3, 1.145 NaH2PO4, 10 glucose, 3 KCl, 1 MgSO4 and 2 CaCl2, 0.1 Na-pyruvate, 0.8 ascorbic acid, bubbled with 95% O2 and 5% CO2. Osmolarity ~300 mOsm.

Experiments were performed with a custom-made two-photon laser scanning microscope based on a modified Olympus BX50WI microscope equipped with a water immersion 40x/0.8 NA objective (Olympus). Patch pipettes were pulled from borosilicate glass (1.5 mm O.D, 0.86 mm I.D, Sutter Instruments Co., CA) using a DMZ Universal Electrode Puller (Zeitz-Instruments, Martinsried, Germany) with a resistance of 4-5 MΩ when filled with internal
solution containing (in mM): 130 K-gluconate, 5 NaCl, 2 MgSO\(_4\), 10 HEPES, 5 EGTA, 4 MgATP, 0.4 Na\(_2\)GTP, 7 Na\(_2\)-phospocreatine, 2 pyruvic acid, 0.007 Alexa 594 hydrazide, pH adjusted to 7.3, ~280-290 mOsm). Whole cell patch-clamp recordings were established using a Multiclamp 700B amplifier (Molecular Devices, Union City, CA). For mapping experiments cells were clamped to -70 mV. Experiments were performed at room temperature. Access resistance was left uncompensated. Voltage and current signals were acquired with a sampling rate of 10k Hz. Current was offline filtered with a cutoff frequency of 1 kHz.

All experiments were performed using a Ti:sapphire laser as the light source (Coherent Chameleon Ultra II, 140-fs pulses, 80-MHz repetition rate). Laser power was modulated by a Pockels cell (350-160, Conoptics, Danbury, CT). Images were acquired using Fluoview 2.1.22 with 800 nm excitation to visualize neurons filled with fluorescent dye via whole cell recordings, or 940 nm excitation to visualize C1V1-EYFP-expressing neurons. Appropriate emission filters were used to separate the red (Alexa 594) and green (EYFP) signal. Optical signals were amplified through photomultiplier tubes (H7422-P40 Hamamatsu) connected to a signal preamplifier (Model 5113, Signal Recovery AMETEK Advanced Measurement Technology, PA). To normalize fluorescence intensity across experiments, we added 5 µL of 6.0 µm–diameter polystyrene microspheres of 0.3% relative intensity fluorescence with excitation/emission wavelength of 505/515 nm (InSpeck Green Calibration Kit, Thermo Fisher Scientific) on top of each slice at the beginning of the experiment. Fluorescence of the microspheres was collected simultaneously when imaging EYFP expression at 940 nm. Mean fluorescence intensity of EYFP-expressing neurons was calculated using ImageJ (National Institutes of Health, MD) and normalized by the mean fluorescence intensity of microspheres from the same image stack. To calculate mean fluorescence intensity, we made an individual sum projection of the image stack. We manually draw a region of interest around the neuron or microsphere and calculated mean fluorescence intensity and subtracted background fluorescence before normalizing the data.

To excite C1V1, the laser was tuned to 1040 nm. For mapping experiments, cell bodies identified by EYFP expression were selected manually using custom-written Matlab code. A random path in which the neurons were stimulated one after the other was generated. A “non-neighbor” algorithm was implemented, which prevented a neuron from being chosen as the next one to be stimulated if another neuron had been stimulated within 15 µm in the last 10 seconds. If no neuron fulfilled this criterion, the minimal time of 300 ms between stimulations was increased to satisfy the criterion. The path file was imported into custom-written software (Nikolenko et al., 2003) used to move the laser beam to the chosen points in the field of view. To stimulate a neuron, a pattern of point stimulations (53 points, lasting about 54 ms total) was carried out, starting at the border of the cell, moving in a spiral inwards with an external
diameter of 9 µm (Fig. 1A). The same pattern was used for calibration experiments. For calibration of lateral resolution, the stimulation pattern was first placed on the center of the soma. Subsequently, stimulations were performed in a radius of 10, 20, and 30 µm around the soma, with 8 stimulations per circle as is shown in Figure 2A. Interstimulation time was 500 ms, and 2 s between repeats of the complete calibration run. For axial calibration of resolution the stimulation placed at the center of the soma and was repeated three times with a 2 s interstimulation time, then the laser focus was moved 3 µm up or down. Finally, in Figure 2I and J, we performed a calibration of lateral resolution in 3 dimensions. In these experiments the lateral stimulation pattern shown in Figure 2A was repeated 3 times at the soma position, then the laser focus was moved 6 µm up or down and stimulation was repeated 3 times per position.

To perform a mapping experiment, we targeted C1V1-EYFP expressing neurons across different focus planes of the mapping volume, repeating stimulations between 5 to 10 times per plane, while recording whole cell currents from the patched neuron. To identify connected neurons, we offline identified EPSCs occurring in response to each stimulation. Considering that maximal latencies of evoked APs in calibration experiments were observed around 70 ms, only EPSCs occurring within 100 ms from the start of stimulation were considered to potentially result from the stimulation of a connected neuron. We selected a single EPSC per stimulation, such that the variation of delays across repeats in the same plane was lowest. We discarded EPSC sets with a standard deviation of delay higher than 18 ms (see results and Figure 2C) and with success rate equal or lower than 0.2 to reduce interferences from background activity or stimulation of neighboring neurons. Positions of targeted cells were confirmed offline. Only targets at which we could clearly detect a cell were considered for further analysis.

To reduce false negatives, we obtained the fluorescence of each targeted neuron automatically within a 5x5 µm square centered at the target positions. For each map, we chose the connected neuron displaying the longest EPSC delays compared to the other observed connections. We then compared this delay to the AP delays obtained in calibration data (Fig. 2G), adding 9 ms to the AP delays, as this was the mean time between AP peak and EPSC peak observed in confirmation experiments (Fig. 5). From these calculations, we assigned a normalized fluorescence intensity value to the connected neuron, based on the linear fit of AP delay and normalized fluorescence in calibration data. Next we compared the absolute fluorescence of all targeted neurons to the absolute fluorescence of the connected neuron. With this, we were able to assign normalized fluorescence values to all targeted neurons. We excluded those with less than 0.15 normalized fluorescence from analysis, as calibration neurons beyond this threshold did not fire APs upon optical stimulation (Fig. 1F).
For the axial calibration data set (n=9 cells), the AP delay increase in dependency on the z position was fitted to a parabolic function with a cubic term:

\[ \Delta \text{Delay} = a + b(z-z_{\text{min}})^2 + c(z-z_{\text{min}})^3 \]

Where \( \Delta \text{Delay} \) is the observed AP delay increase, \( z \) corresponds to the axial z position and \( z_{\text{min}} \) to the z position where the delay was shortest. The constants where fitted to the following values considering their ±95% confidence interval: \( a = (-2 \pm 2) \times 10^{-1} \text{ ms} \), \( b = 0.1 \pm 0.1 \text{ ms} \cdot \mu \text{m}^{-2} \) and \( c = (1.3 \pm 14) \times 10^{-3} \text{ ms} \cdot \mu \text{m}^{-3} \). The cubic term in the parabolic function was used to account for the slight asymmetry in the delay increase dependence on z position. We evaluated the similitude of the EPCSs delays from targeted cells to the calibration function using a \( \chi^2 \)-test. \( \chi^2 \) was calculated using the following formula:

\[ \chi^2 = \sum_{i=1}^{n} \frac{(O_i - C_i)^2}{\sigma_i^2} \]

Where \( O \) is the EPCSs delay observed, \( C \) is the calibration delay and \( \sigma \) the standard deviation of calibration delay. Then we calculated the associated p-value for a \( \chi^2 \) cumulative distribution function with \( n-2 \) degrees of freedom in \( \chi^2 \). We assumed the position of the targeted cell within ±6 \( \mu \text{m} \) of the point of lowest delay in the calibration function. Targeted cells with \( p < 0.1 \) were discarded as possible connections (See Figure 4 with examples of different fittings obtained).

In the cases where we observed neurons located on top of each other, we fitted the data to two calibration functions, each one centered at each position identified (\( n=3 \) cases, see Figure 4C). When two calibration curves overlapped, we always considered the calibration function with lowest delay value at that z position. Then, as we did for single calibration curve, we evaluated the similitude between the function with two calibration curves and EPSC delay data by calculating the associated p-value, now considering \( n-4 \) degrees of freedom. Stimulations with \( p < 0.1 \) were considered as not featuring two connected cells. The discarded ones were tested further for similitude to single calibration function considering each of the soma locations identified. The soma location resulting in the higher p-value was chosen as the connected neuron.

For analysis of synaptic current peak amplitudes, the values obtained when targeting the soma directly as those within 6 \( \mu \text{m} \) axial distance were averaged. Correlation coefficients were obtained using MATLAB.

Data were acquired and analyzed using PackIO (Watson et al., 2016), Ephysviewer (Watson et al., 2016), MiniAnalysis (Synaptosoft, NJ), IgorPro (Wavemetrics, OR) and custom-written MATLAB code (Mathworks, MA).
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