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# Genetically Encoded Sensors for the In Vivo Detection of Neurochemical Dynamics

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## Abstract

The ability to measure dynamic changes in neurochemicals with high spatiotemporal resolution is essential for understanding the diverse range of functions mediated by the brain. We review recent advances in genetically encoded sensors for detecting neurochemicals and discuss their in vivo applications. For example, notable progress has been made with respect to sensors for second messengers such as cyclic adenosine monophosphate, enabling in vivo real-time monitoring of these messengers at single-cell and even subcellular resolution. Moreover, the emergence of highly sensitive sensors for neurotransmitters and neuromodulators has greatly accelerated the study of these signaling molecules in a wide variety of behavioral models using an array of powerful imaging techniques. Finally, we discuss the future direction of neurochemical sensors, including their ability to measure neurochemical concentrations and the potential for multiplex imaging.



## 1. INTRODUCTION

The nervous system relies on neurotransmitters (NTs) and neuromodulators (NMs) to coordinate signaling between cells. In addition, intracellular second messengers such as calcium ( $\text{Ca}^{2+}$ ) and cyclic adenosine monophosphate (cAMP) play an essential role in regulating synaptic transmission and neuronal plasticity. These chemicals work together in an intricately regulated network to drive various physiological processes, including learning and memory, perception, and the sleep–wake cycle. For the purposes of this review, we collectively refer to NTs, NMs, and intracellular messengers as neurochemicals. In addition to hundreds of neurochemicals that function within the nervous system, a wide range of cell types—ranging from neurons to glia—have specific properties and mediate a variety of functions. Therefore, understanding neural circuitry requires a robust set of tools designed to monitor spatial and temporal changes in neurochemicals within a genetically defined population of cells.

Genetically encoded optical sensors have revolutionized the study of neurochemicals by giving researchers the ability to monitor specific neurochemicals, including extracellular NTs and NMs, as well as intracellular second messengers. In this review, we summarize the array of genetically encoded sensors for neurochemicals and discuss their use *in vivo* in behaving animals for addressing key biological questions. In addition, we discuss future opportunities for developing and applying genetically encoded optical sensors, including their use in (a) multiplex imaging and simultaneous recordings and (b) quantitative measurements.

## 2. THE BIOCHEMICAL NATURE OF SIGNALING PATHWAYS AND NEURONAL ACTIVITY AND TRADITIONAL METHODS FOR DETECTION

### 2.1. Intracellular Second Messengers: $\text{Ca}^{2+}$ and cAMP

Throughout several billion years of evolution, cells have adapted to use  $\text{Ca}^{2+}$  as a crucial signaling species by expressing numerous  $\text{Ca}^{2+}$ -binding proteins to perform various physiological functions (1). Specifically, neurons employ specialized membrane proteins known as synaptotagmins to sense  $\text{Ca}^{2+}$  and mediate rapid vesicle fusion, leading to the release of certain types of NTs and NMs (2). Indeed,  $\text{Ca}^{2+}$  imaging in neurons provides a reliable approximation of neuronal activity, as reviewed extensively by Grienberger & Konnerth (3). Consequently,  $\text{Ca}^{2+}$  sensors are by far the most widely used in neuroscience. Given the close relationship between increased intracellular  $\text{Ca}^{2+}$  concentration ( $[\text{Ca}^{2+}]_i$ ) and vesicle fusion,  $[\text{Ca}^{2+}]_i$  has also been used as an indirect measure of vesicle-dependent NT and NM release. This approach is based on the assumption that each type of neuron releases a specific NT or NM; however, single-cell RNA sequencing data suggest that each neuron expresses at least one small-molecule transmitter and one neuropeptide (4), thereby complicating the interpretation of  $\text{Ca}^{2+}$  imaging in neurons. Furthermore, specific NTs and NMs such as purines (5) and neurolipids (6) can be released via vesicle-independent mechanisms, and the role of  $\text{Ca}^{2+}$  in these processes remains poorly understood. Thus, the use of  $\text{Ca}^{2+}$  indicators may not serve as a definitive method for tracking dynamic changes in NT and NM release.

Another key intracellular signaling molecule that regulates the effector function of many NTs and NMs is cAMP, which orchestrates an array of cellular functions in neurons, including morphological properties, neuronal excitability, and synaptic plasticity. By acting on multiple targets—including ion channels, protein kinases, and transcription factors—cAMP can drive a wide range of biochemical processes over timescales ranging from seconds to several days (7). In cells, cAMP is produced by the enzyme adenylyl cyclase, which converts ATP to cAMP; in contrast, the enzyme phosphodiesterase converts cAMP to AMP, thereby reducing cAMP levels. Although



intracellular concentrations of cAMP were previously believed to increase uniformly throughout the cell due to its high diffusion coefficient, emerging evidence suggests that cAMP may accumulate and function locally within cellular “microdomains” (8). In this respect, the development of cAMP sensors may provide new opportunities to study these cAMP microdomains.

## 2.2. Extracellular Signaling: Chemical Synapses and Neurotransmission

Neurons form complicated networks via chemical synapses, which comprise three structural components: the presynaptic terminal, the synaptic cleft, and the postsynaptic membrane. The presynaptic terminal contains vesicles that are filled with one or more specific NTs and/or NMs, including small molecules and neuropeptides. Note that cell–cell communication via distinct molecules is not unique to neurons but is also used by other cell types, such as glial cells.

Small molecular NTs such as glutamate,  $\gamma$ -aminobutyric acid (GABA), and glycine (9, 10) are classified on the basis of their action on ionotropic (i.e., ion-conducting) receptors, which elicit an electrophysiological response and thereby initiate extremely rapid, spatially restricted synaptic communication [note also that some NTs have corresponding metabotropic G protein–coupled receptors (GPCRs) such as metabolic glutamate receptors and GABA<sub>B</sub> receptors, which bind glutamate and GABA, respectively, to drive intracellular signaling via second messengers]. These properties make it possible to measure the release of NTs using highly sensitive electrophysiological recording techniques, which can capture events on a millisecond timescale. Importantly, however, because changes in membrane potential can result from the activity of various NTs, specific receptor antagonists are needed to confirm their molecular specificity (11). Furthermore, electrophysiological recording is technically challenging and is a relatively low-throughput technique, limiting its use primarily to *in vitro* applications.

In contrast, NMs such as monoamines, neurolipids, and neuropeptides act primarily via metabotropic GPCRs (although some molecules such as serotonin and ATP also have corresponding ionotropic receptors) (12). NMs induce a change in the level of secondary messengers within the recipient cell, thereby changing its electrophysiological properties, metabolism, morphology, and/or transcriptional activity. In contrast to ionotropic signaling, metabotropic signaling has a relatively slow rate and long range due to diffusion, affecting a wider area in the neural circuitry (13, 14). Importantly, not all GPCRs are coupled to ion channels, preventing the use of electrophysiological recordings for detecting the release of some NMs. To overcome this limitation, a range of detection techniques have been developed. Some of them (e.g., Tango assay and cell-based neurotransmitter fluorescent engineered reporters) are reviewed extensively elsewhere (15).

With respect to a suitable detection method, several criteria have been described, including high sensitivity, high molecular specificity, and high spatial and temporal resolution (16). Among the various methods developed to date, optical imaging has shown great potential for meeting most of these criteria, leading to rapid advances in the development of genetically encoded sensors for detecting NTs and NMs (17, 18).

In addition to their distinct modes of action, some NTs and NMs can bind to multiple receptors, with an array of affinities. Thus, even when the same neurochemical is present, its function is determined in part by its concentration. For example, dopamine (DA) receptors are divided into G $\alpha_{s/olf}$ -coupled D1-like receptors and G $\alpha_{i/o}$ -coupled D2-like receptors. Pharmacologically, D1-like receptors have an affinity 10- to 100-fold lower than that of D2-like receptors (19). In addition, two main types of medium spiny neurons in the striatum express either D1 or D2 receptors, with minimal overlap (20). These findings suggest that the effects of D1 versus D2 receptor signaling will vary depending on extracellular DA concentration (21, 22). Moreover, abnormal levels of NTs and/or NMs are associated with numerous neurological disorders, as reviewed by Carhart-Harris



& Nutt (23). Thus, our ability to correlate NT/NM levels with physiological and/or pathological conditions requires tools that can provide information regarding absolute concentration. Although existing optical techniques may not completely meet the technical requirements needed for accurate quantification, several strategies may help overcome this limitation, a topic we address in the following sections.

### 3. PHOTOCHEMISTRY AND THE STRUCTURAL BASIS OF GENETICALLY ENCODED SENSORS

The green fluorescent protein (GFP) was discovered in the early 1960s (24), and Chalfie et al. (25) showed nearly three decades ago that GFP can be heterologously expressed in both prokaryotic and eukaryotic cells. These breakthroughs paved the way for the subsequent development and application of various genetically encoded optical tags and sensors for studying biological processes. Compared with other methods, genetically encoded sensors have several unique advantages. For example, they can be expressed in specific cell types, thus providing high cellular specificity. In addition, these sensors are compatible with long-term imaging for days—or even months—when expressed in living cells and animal models (26). These properties make genetically encoded sensors well suited for use in *in vivo* preparations.

As a general principle of photochemistry, fluorescence occurs when photon-excited molecules decay back to their ground state and emit detectable photons (**Figure 1a**). The brightness of a fluorophore is determined primarily by two parameters, namely the extinction coefficient and quantum yield. The extinction coefficient describes the fluorophore's efficiency at absorbing excitation light, while the quantum yield is the fluorophore's ability to produce photons and is defined as the ratio of emitted photons to absorbed photons. Both parameters are intrinsic to the fluorophore's chemical composition and can be affected by the microenvironment. In addition to fluorescence intensity, a fluorophore is characterized by its fluorescence lifetime, typically defined as the time in which an excited molecule stays in the excited state. For a single fluorescent protein (FP), the fluorescence lifetime is proportional to the quantum yield. The fluorescence lifetime can be measured using fluorescence lifetime imaging microscopy (FLIM) (27, 28) and is generally not affected by intensity-related factors, thus circumventing the difficulties associated with intensimetric imaging and drawing increasing interest (29–33).

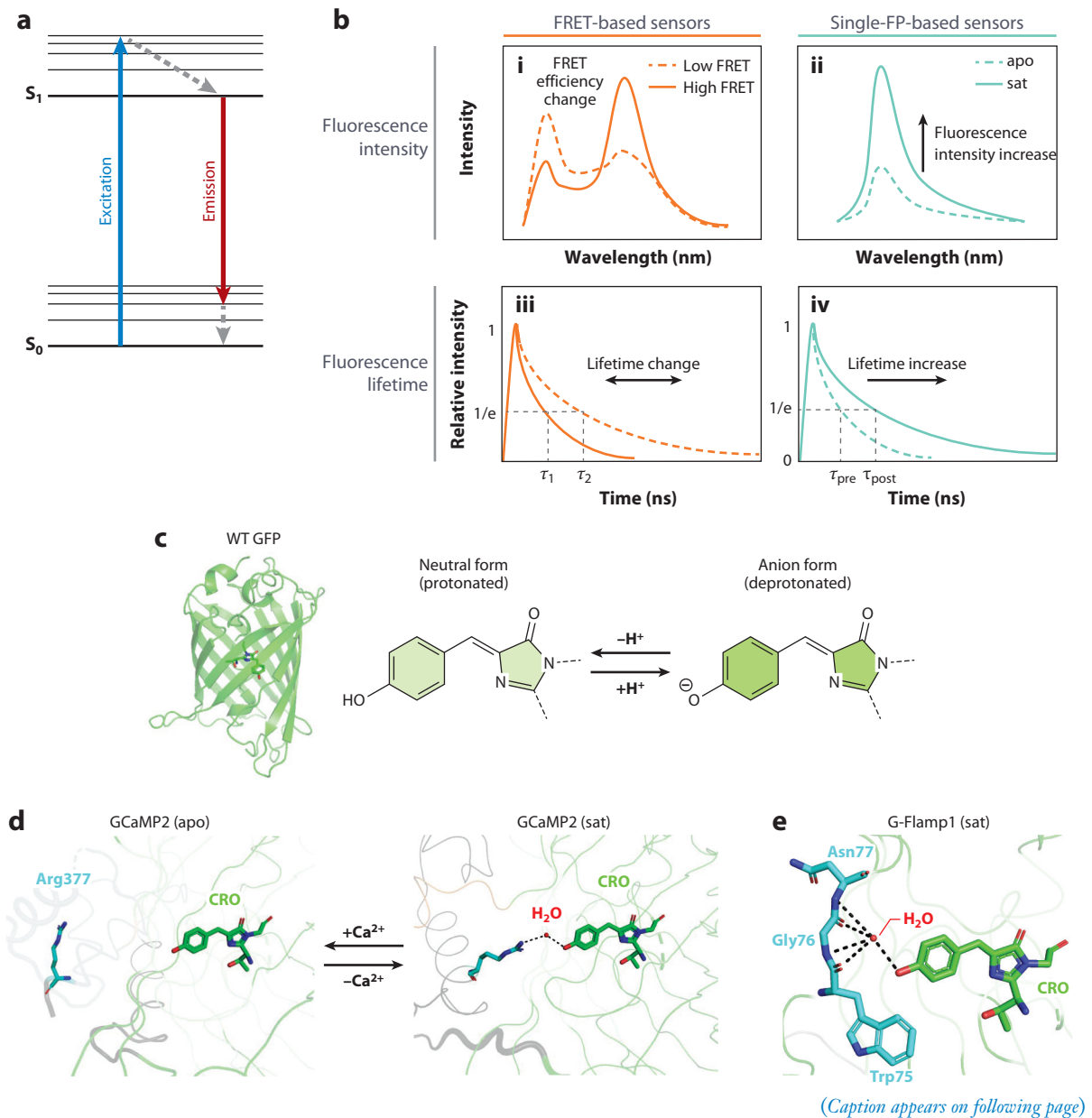
A genetically encoded sensor typically has two components. The first is a sensing module that binds the ligand of interest, and the second is the reporter module, which is usually based on either Förster resonance energy transfer (FRET) or a conformation-sensitive change in a single FP such as GFP. Furthermore, on the basis of readout, these sensors can be further classified as either intensity-based sensors, which provide an intensimetric or ratiometric readout, or FLIM-based sensors, which provide fluorescence lifetime as the readout (**Figure 1b**). More specifically, in most circularly permuted (cp)GFP-based sensors, the predominant factor contributing to the change in fluorescence intensity is a change in the extinction coefficient, while the change in quantum yield is typically not significant. Consequently, such sensors are suitable only for intensimetric measurements, not lifetime measurements. Therefore, the development of novel FLIM-based sensors requires ways to regulate the sensors' quantum yield (31).

FRET-based sensors are generally easier to develop and can theoretically be used to quantify ligand concentration using the emission ratio or FLIM-FRET. In addition, expressing both the donor and acceptor proteins in a single expression construct helps reduce sensor concentration-based effects. In contrast, several factors can limit the ability to accurately quantify the signal, including wavelength-dependent scattering, differing protein maturation rates between the donor and acceptor, and differences in the emission ratio when using different detection tools (34).



Moreover, the dynamic range of FRET-based sensors is typically rather limited and is determined primarily by the physical distance between the donor and acceptor. Finally, most FRET-based sensors have two emission wavelengths, limiting their application in multiplex imaging.

Compared with FRET-based sensors, single-FP-based sensors are more challenging to engineer, but they have recently become more popular because of several key advantages. First, single-FP-based sensors can achieve a higher dynamic range and provide an intensimetric signal, making them more suitable for use in *in vivo* applications. Second, their relatively small size makes single-FP-based sensors easier to deliver in *in vivo* preparations. Moreover, they typically occupy



**Figure 1** (Figure appears on preceding page)

Types and structural basis of genetically encoded sensors. (a) Jablonski diagram illustrating the principle of fluorescent indicators.  $S_0$  and  $S_1$  represent the ground and excited states, respectively. Excitation light shifts the fluorophore to the excited state, and emission light of a different wavelength is released when the fluorophore decays back to the ground state. (b) Overview of the four general types of genetically encoded fluorescent sensors, categorized according to their fluorescence module and signal readout. (i) With a FRET-based ratiometric sensor, FRET efficiency is altered upon ligand binding, inducing a change in the sensor's spectrum and emission ratio. (ii) With a single-FP-based intensimetric sensor, fluorescence intensity generally increases upon ligand binding; in some cases, ligand binding induces a decrease in intensity (i.e., inverse sensors). Apo refers to the ligand-free state, and sat refers to the ligand-saturated state. (iii) Normalized fluorescence lifetime decay curve with the FRET donor in the low-FRET or high-FRET state. The lifetime of the donor FP decreases when FRET occurs, thereby reporting the sensor state. (iv) Normalized fluorescence lifetime decay curve with a single-FP-based sensor. The difference in fluorescence lifetime corresponds to different sensor states. (c) Crystal structure (PDB ID: 1EMA) of WT GFP, highlighting the chemical composition of the chromophore. The GFP chromophore is constantly shifting between the neutral (protonated) phenol state and ionized (deprotonated) phenolate state. The deprotonated state is typically brighter in fluorescence and is therefore the main contributor to the sensor's signal. (d) Structures of the GCaMP2 sensor in (left) the  $Ca^{2+}$ -free state (PDB ID: 3EKJ) and (right) the  $Ca^{2+}$ -bound state (PDB ID: 3EVR). In the  $Ca^{2+}$ -bound state, the Arg-377 residue is close to the chromophore, allowing the guanidinium side chain to stabilize the protonated form of the chromophore via a water bridge interaction. (e) Structure of the G-Flamp1 sensor, highlighting the water bridge between the linker and the chromophore (PDB ID: 6M63). Abbreviations: Arg, arginine; Asn, asparagine; CRO, chromophore; FP, fluorescent protein; FRET, Förster resonance energy transfer; GFP, green fluorescent protein; Gly, glycine; PDB ID, Protein Data Bank identifier; Trp, tryptophan; WT, wild-type.

a single emission wavelength band and are therefore more amenable for use in multiplex imaging. Together, these advantages outweigh their lack of an ability to accurately quantify the signal, and many single-FP-based sensors have been developed.

To create a sensor in which the FP's fluorescence changes based on the local environment, the FP's chromophore must be exposed. This is commonly achieved by circularly permuting the FP—or, in some cases, by splitting the FP—and then linking the FP to the sensing module. The conformational change in the sensing module induced by ligand binding is then transduced to the cpFP, changing the microenvironment around the exposed fluorophore, and causing a measurable, reversible change in fluorescence.

Many single-FP-based sensors have been developed to date and their structural study has helped elucidate how sensors respond to ligand binding. The first mechanistic insights into how single-FP-based sensors respond to ligand binding came from the structure of GCaMP2, a  $Ca^{2+}$  sensor based on cpGFP and the calmodulin (CaM)–M13 peptide complex (35, 36). In wild-type GFP, three consecutive residues (threonine–tyrosine–glycine) form the chromophore and are protected from solvent by surrounding  $\beta$ -sheets. The fluorophore in wild-type GFP exists in two forms depending on the protonation state of the phenol group in the tyrosine residue (37). In the neutral (i.e., protonated) state, the chromophore has an absorption peak at 395 nm, while the ionized (i.e., deprotonated) form has an absorption peak at 475 nm (**Figure 1c**). Of these two forms, the ionized form contributes to the majority of the chromophore's fluorescence. A structural analysis of both the  $Ca^{2+}$ -free and  $Ca^{2+}$ -bound forms of GCaMP2 revealed that Arg-377 in the CaM domain plays a key role in determining the sensor's fluorescence response (35). In the  $Ca^{2+}$ -free state, Arg-377 is located relatively far from—and therefore does not interact directly with—the chromophore, causing the chromophore to reside predominantly in the neutral state.  $Ca^{2+}$  binding induces a conformational change that moves the Arg-377 residue closer to the chromophore, allowing a water bridge to form. As a result, the fluorophore is stabilized in the ionized state and the fluorescence increases. This change in the chromophore's protonation state, combined with changes in the fluorophore's extinction coefficient and quantum yield, provides the mechanism by which GCaMP2 responds to an increase in  $Ca^{2+}$  concentration (**Figure 1d**).

The mechanism by which fluorescence changes in response to ligand binding may vary widely from sensor to sensor, as suggested by structural studies of a large variety of sensors (38–42). For example, in the cAMP-bound state, the cAMP sensor G-Flamp1 (42), which is based on

cpGFP and a bacterial cyclic nucleotide-binding domain, has a unique N-terminal linker structure that contains a highly twisted  $\beta$ -strand and is in close proximity to the chromophore in GFP. A closer examination of the structure revealed that the Trp-75 residue in the N-terminal linker helps stabilize the chromophore, both by forming a water bridge and by protecting the chromophore from solvent quenching (**Figure 1e**). The apparent significance of the Trp-75 residue was further supported by molecular dynamics simulations and saturation mutagenesis (42).

In summary, the change in a sensor's fluorescence can occur via the interaction between the chromophore and adjacent residues in a broad range of sensor domains. Mechanistic insights based on structural studies will likely facilitate the development of high-performance sensors. Such high-performance sensors are expected to have improved optical properties and are better suited for use in biological applications.

## 4. IMAGING NEUROCHEMICALS USING GENETICALLY ENCODED SENSORS

### 4.1. Genetically Encoded $\text{Ca}^{2+}$ Sensors

Given the important role of  $[\text{Ca}^{2+}]_i$  in a wide variety of physiological processes, it is not surprising that a variety of both chemical  $\text{Ca}^{2+}$  sensors and genetically encoded  $\text{Ca}^{2+}$  indicators (GECIs) have been developed, the earliest of which were introduced in the 1960s (43–45). However, the first truly groundbreaking advance in  $\text{Ca}^{2+}$  dyes can be attributed to Tsien and colleagues (46, 47). Prior to the wide application of GECIs,  $\text{Ca}^{2+}$  dyes had already been used to detect  $\text{Ca}^{2+}$  changes both in vitro and ex vivo. Although commercially available  $\text{Ca}^{2+}$  dyes are highly sensitive and have been used quite extensively, as small molecules they cannot selectively label specific cell types and sometimes require highly skilled microinjection techniques for in vivo applications. In contrast, genetically encoded indicators offer a more straightforward means to be readily expressed in living animals. Using cell type-specific promoters and/or recombinase-based approaches, these sensors can achieve specificity in genetically defined cell populations. Moreover, the discovery and subsequent optimization of GFP and similar FPs have opened a new world of possibilities in this field.

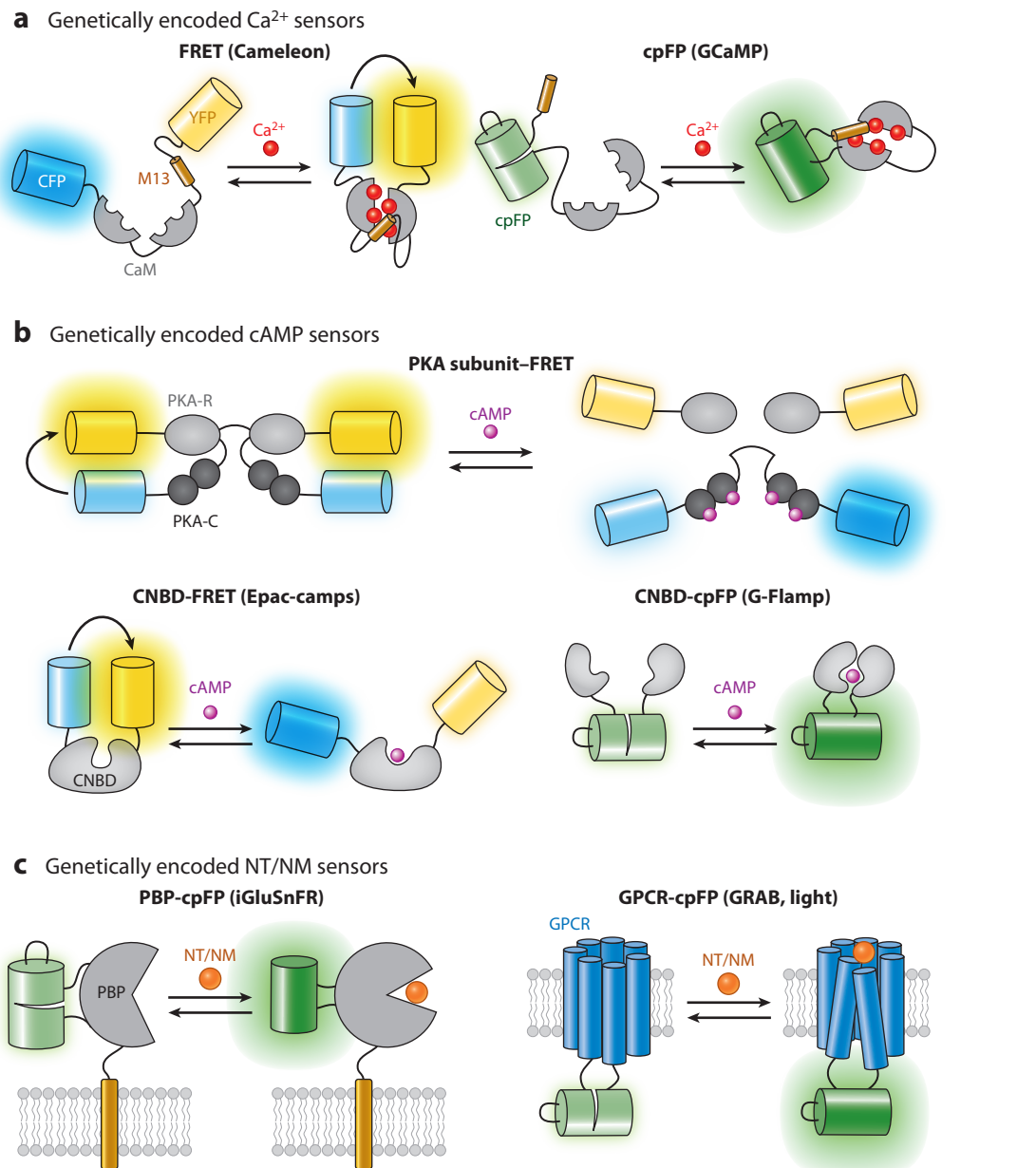
In 1997, the Tsien group (47) developed the first GECI, a fluorescent  $\text{Ca}^{2+}$  indicator in which the cyan fluorescent protein–yellow fluorescent protein (CFP–YFP) FRET pair was flanked by CaM and the CaM-binding peptide (CBP) M13. The rationale for this sensor was based on the reversible “wrapping” of CaM and M13 upon binding  $\text{Ca}^{2+}$  ions, which increased FRET efficiency between CFP and YFP (**Figure 2a**). Further advances led to the development of FRET-based  $\text{Ca}^{2+}$  sensors with a higher dynamic range and improved kinetics; these new sensors were then successfully expressed in several animal models, including *Drosophila* and mice (48, 49).

The design of single-FP-based  $\text{Ca}^{2+}$  indicators is similar in some respects to that of FRET-based sensors. With a single-FP-based  $\text{Ca}^{2+}$  indicator, a single fluorophore is inserted into the CaM–CBP domain (50, 51). This design was based on the discovery of cpFPs by Baird et al. (52) in the late 1990s (**Figure 2a**). Over the course of more than two decades, single-FP-based sensors have been optimized, resulting in sensors that often have superior dynamics compared with those of FRET-based sensors.

Among these single-FP-based  $\text{Ca}^{2+}$  sensors, the GCaMP series stands out. As a result of extensive research and development, GCaMP sensors are now considered state-of-the-art tools for imaging  $\text{Ca}^{2+}$  dynamics in vivo (53). For example, GCaMP6f can respond to a single action potential-induced  $\text{Ca}^{2+}$  increase within 50 ms, and the decay half-life is 142 ms. The affinity of GCaMP6f is 375 nM, approximately three- to sevenfold lower than the resting level of  $[\text{Ca}^{2+}]_i$



(50–100 nM). The suitable affinity and fast kinetics allow the sensor to readily catch up with the increase of  $[Ca^{2+}]_i$  during action potential and other intracellular  $Ca^{2+}$  mobilization processes. The dynamic ranges of purified protein of GCaMP6 sensors are around 38- to 63-fold, superior to those of the previous GCaMP3 (12.3-fold) and GCaMP5 (17.4–32.7-fold) versions. Indeed, the outstanding performance of these sensors has led to their widespread use in virtually all model



*(Caption appears on following page)*



**Figure 2** (Figure appears on preceding page)

Genetically encoded indicators for detecting neurochemicals. (a) Schematic depiction of GECIs. Two types of GECIs have been developed. (Left) FRET-based GECIs are created by inserting the CaM domain and a CaM-binding peptide (e.g., M13) into FRET pair FPs such as the CFP-YFP pair used in Cameleon. (Right) In contrast, cpFP-based GECIs are created by inserting a cpFP into M13 and CaM, as shown for the GCaMP series of sensors. Closure of the narrow split in the cpGFP represents the environment of the cpGFP chromophore upon calcium binding. (b) Schematic depiction of genetically encoded cAMP indicators. cAMP indicators have been developed according to several principles, including PKA holoenzyme subunits and CNBDs. (Top row) A FRET-based sensor using PKA holoenzyme subunits was developed by fusing the FRET donor to the PKA-R (regulatory) subunit and fusing the FRET acceptor to the PKA-C (catalytic) subunit. (Bottom left) With FRET-based cAMP indicators that use CNBDs as the single cAMP-binding domain, CFP and YFP are fused to the N-terminal and C-terminal domains of CNBD, and cAMP levels are reported as a change in the FRET ratio. Epac-camps is an example of this sensor type. (Bottom right) Finally, similar to cpFP-based  $\text{Ca}^{2+}$  indicators, cpFP can be inserted into a CNBD to generate an intensimetric cAMP indicators such as G-Flamp1. (c) Schematic depiction of genetically encoded NT/NM indicators. Two general categories of NT/NM indicators have been developed on the basis of the ligand-binding protein used (i.e., PBPs or GPCRs). (Left) With single-fluorophore indicators that use a PBP, the cpFP is inserted into the PBP's hinge region (e.g., iGluSnFR). (Right) With GPCR-based fluorescent indicators, the cpFP is inserted into the receptor's third intracellular loop of the GPCR (e.g., GRAB, Light). Abbreviations: CaM, calmodulin; CNBD, cyclic nucleotide-binding domain; cpFP, circularly permuted FP; FP, fluorescent protein; FRET, Förster resonance energy transfer; GECI, genetically encoded  $\text{Ca}^{2+}$  indicator; GPCR, G protein-coupled receptor; NM, neuromodulator; NT, neurotransmitter; PBP, periplasmic binding protein; PKA, protein kinase A. In panel a, the diagram of FRET-based GECIs is adapted from Reference 47, and the cpFP-based  $\text{Ca}^{2+}$  sensor image is adapted from Reference 50. In panel b, the diagram of the PKA subunit-based cAMP indicators is adapted from Reference 63, the CNBD-based FRET cAMP indicator image is adapted from Reference 150, and the CNBD-based cpFP cAMP indicator image is adapted from Reference 42. In panel c, the diagram of PBP-based NT/NM indicators is adapted from Reference 82, and the GPCR-based indicator image is adapted from References 89 and 90.

organisms, and they are highly compatible with a wide range of imaging techniques, including fiber photometry recording, multiphoton imaging, and the recently developed mesoscopic wide-field imaging.

In many respects, GCaMP sensors revolutionized the field of  $\text{Ca}^{2+}$  imaging, allowing researchers to visualize  $\text{Ca}^{2+}$  dynamics in living organisms with high precision and high spatiotemporal resolution. Recent advances and the expansion of GFP-based  $\text{Ca}^{2+}$  indicators have further improved these sensors' response, brightness, and kinetics. For example, replacing enhanced GFP with mNeonGreen (an exceptionally bright FP derived from the cephalochordate *Branchiostoma lanceolatum*) led to the NEMO series of sensors, with wide dynamic ranges and brightness (54). In addition, by screening various CBP domains, investigators have developed jGCaMP8 sensors with faster on and off kinetics (55). The off kinetics of the jGCaMP8f is  $37.03 \text{ s}^{-1}$ , which is around ten times faster than that of GCaMP6f ( $3.93 \text{ s}^{-1}$ ) and more than five times faster than that of jGCaMP7f ( $7.34 \text{ s}^{-1}$ ).

Several other series of single-FP-based  $\text{Ca}^{2+}$  indicators have been developed. Multiplex imaging requires sensors with distinct, nonoverlapping spectra colors, which prompted the development of the GECO and XCaMP series of GECIs, in which the original cpGFP module was replaced with other FPs (56, 57). Highly sensitive red fluorescent  $\text{Ca}^{2+}$  indicators, such as the jRGECO and jRCaMP series, have also been created (58). These indicators offer clear advantages over green fluorescent indicators, including lower phototoxicity and deeper penetration of light into living tissue. Recently, far-red  $\text{Ca}^{2+}$  imaging became possible with the introduction of far-red FPs (59, 60) and cp-protein tags with far-red chemical dyes (61), greatly expanding the toolkit of  $\text{Ca}^{2+}$  indicators. These advances now provide researchers with various options for  $\text{Ca}^{2+}$  imaging, with improved sensitivity, the ability to perform multiplex imaging, and increased imaging capabilities. For instance, a study by Inoue et al. (57) simultaneously utilized GFP-, blue fluorescent protein-, and red fluorescent protein-based calcium indicators to monitor different cell types in behaving mice. However, the use of blue sensors in *in vivo* applications is less preferred, because their excitation light ( $\sim 385 \text{ nm}$ ) has compromised tissue penetrance and their



emission light (~450 nm) falls within the range of enhanced autofluorescence from biological samples. An alternative approach involves employing far-red sensors, which not only overcome these limitations but also allow for four-color multiplexed imaging.

Another advantage of GECIs is their ability to target specific tissues using cell type-specific promoters and/or transgenic lines. Moreover, through the addition of targeting motifs, these sensors can even be directed to specific subcellular compartments, facilitating the imaging of structures such as the axon, soma, or endoplasmic reticulum so as to provide high versatility and specificity.

## 4.2. Genetically Encoded cAMP Sensors

Various genetically encoded sensors developed over the past 30 years have contributed considerably to our understanding of cAMP signaling. Genetically encoded cAMP sensors have been developed using various downstream effector protein domains as the sensing module. The first such sensor, FICRhR [fluorescein-labeled protein kinase A (PKA) catalytic (PKA-C) subunit and rhodamine-labeled regulatory (PKA-R) subunit (62)], is based on dissociation of the PKA holoenzyme and consists of two fluorescent modules fused to the PKA-R and PKA-C subunits. When cAMP binds to the PKA-R subunit, the PKA-C subunit dissociates, inducing a FRET signal. However, FICRhR is not fully genetically encoded and requires microinjection into the organism together with tagged rhodamine and fluorescein. Subsequent versions of this sensor based on PKA holoenzyme dissociation use FPs instead of fluorescent dyes and are fully genetically encoded. Subsequent optimization included the use of higher-efficiency FRET pairs and fine-tuned the linker domains (63–66) (**Figure 2b**).

Single-polypeptide cAMP sensors exploit the conformational change induced by cAMP binding to its effector proteins such as PKA, Epac (exchange protein directly activated by cAMP) (67–69), and cAMP-dependent ion channels (70). Moreover, FRET efficiency can be optimized by fusing FPs to the N-terminal and C-terminal domains of these effector proteins, or truncated versions containing the cAMP-binding domain. For example, considerable effort has led to optimized Epac-based FRET cAMP sensors with a high FRET ratio (**Figure 2b**), including a series of FLIM-FRET-based cAMP sensors (71). Moreover, the use of cpFPs to develop a sensor led to green fluorescent cAMP sensors based on cpGFP (42, 72) and its red-shifted variants (73, 74), offering a significant change in fluorescence and wide versatility for *in vivo* applications. Despite their relatively brief history, single-FP-based cAMP sensors have made significant strides in the past decade, and their compact size and potential for producing a larger fluorescence response will likely result in even further advances in the field (**Figure 2b**).

Note that a wide array of genetically encoded sensors have been developed to monitor the activity of PKA, a major downstream effector of cAMP. We do not cover PKA sensors in this review because PKA is generally not considered a neurochemical; however, PKA sensors have been reviewed thoroughly by several other groups (32, 75, 76).

## 4.3. Genetically Encoded Sensors for Extracellular Neurochemicals

The mammalian brain uses more than 100 known NTs and NMs, each of which plays an important role in complex processes such as signaling transmission (13). In addition, drugs that target NT and/or NM receptors—or NT and/or NM metabolic enzymes—have shown promise in treating various neurological conditions. For example, decreased serotonin levels are widely believed to play a central role in depression, and the selective serotonin reuptake inhibitor fluoxetine (Prozac) acts as an effective antidepressant by increasing serotonin levels in the synaptic cleft (77). Over the



past two decades, numerous sensors for NTs and NMs have been developed, allowing researchers to perform *in vivo* imaging with high spatiotemporal resolution.

Depending on the sensing module, NT and NM sensors can be categorized into two major groups, namely sensors based on a periplasmic binding protein (PBP) and sensors based on a GPCR (see **Supplemental Table 1**). PBPs are a group of proteins typically consisting of two large lobes that surround the bound ligand, resembling a Venus flytrap (78). These proteins undergo a large conformational change upon ligand binding, making them a promising scaffold for engineering biosensors.

The strategy for developing the first PBP-based NT sensors FLIPE (79) and GluSnFR (80, 81) involved inserting the glutamate-binding protein into CFP and YFP. Because the PBP is displayed on the outer surface of the cell membrane, any changes in FRET efficiency reflect a change in extracellular glutamate concentration. Inspired by cpGFP-based GECIs, investigators then developed a series of single-FP-based sensors using PBPs; the first, and perhaps the most widely used, are the iGluSnFR series (82–84) (**Figure 2c**). Further engineering of iGluSnFR led to next-generation iGluSnFR3 versions with faster kinetics (approximately twice as fast compared with the previous-generation SF-iGluSnFR) and higher sensitivity under two-photon excitation (~20-fold response improvement compared with SF-iGluSnFR) (see **Supplemental Table 1**). The success of iGluSnFR paved the way for the development of a broad range of sensors for detecting NTs and NMs, including—but by no means limited to—acetylcholine (ACh) (85), serotonin (86), and GABA (87).

Although the large ligand-induced conformational change in PBPs leads to a large dynamic change, not every NT and NM has a corresponding PBP, particularly in the case of neuropeptides. Indeed, most PBP-based sensors developed to date have been designed to detect small molecules and usually require considerable effort to optimize their affinity and molecular specificity.

The introduction of GPCR-based biosensors started a new era in the detection of NTs and NMs. Unlike PBPs, GPCRs serve as the body's own sensors for NTs and NMs. Moreover, as transmembrane proteins, most GPCRs function at the cell surface, reducing the need to engineer the sensor to target the plasma membrane. These advantages make GPCRs attractive scaffolds for building genetically encoded biosensors. Several decades of extensive study have revealed that most GPCRs contain a highly conserved conformational change in the TM5 and TM6 domains upon activation (88). By replacing the receptor's third intracellular loop between TM5 and TM6 with cpFP modules, researchers developed two major families of GPCR-based sensors, namely the GRAB (89) and Light (90) sensors (**Figure 2c**). GPCR-based sensors have recently been developed for an array of NTs and NMs, including monoamines (91–94), purines (95, 96), neurolipids (97), and neuropeptides (98, 99). Because these sensors are derived from their respective GPCRs, they have a good affinity for detecting changes in endogenous NTs and NMs. Moreover, by systematically modifying the linker region, the cpFP module, and/or the GPCR itself, these optimized sensors can have remarkably high sensitivity, making them ideal for monitoring NT/NM release *in vivo* (see **Supplemental Table 1**).

Note, however, that certain GPCRs have a promiscuous response to structurally similar chemicals. For example, several opioid receptors can respond to a wide range of endogenous peptide ligands. The DA receptor D1R also shows modest affinity toward norepinephrine (NE). The native oxytocin (OT) receptor can be activated by arginine vasopressin, which is evolutionarily homologous to OT. Sensors developed with such GPCR backbones also display similar ligand specificity. In order to distinguish between different endogenous ligands, extensive molecular engineering in the GPCR domain of the sensor is often required. By mutating amino acids that interact with the ligands in the ligand binding pocket, one can tune the sensors' selectivity for higher specificity to distinguish between different endogenous ligands.



## 5. DETECTING NEUROCHEMICALS IN VIVO

### 5.1. Approaches for Detecting Neurochemicals In Vivo with Fluorescent Sensors

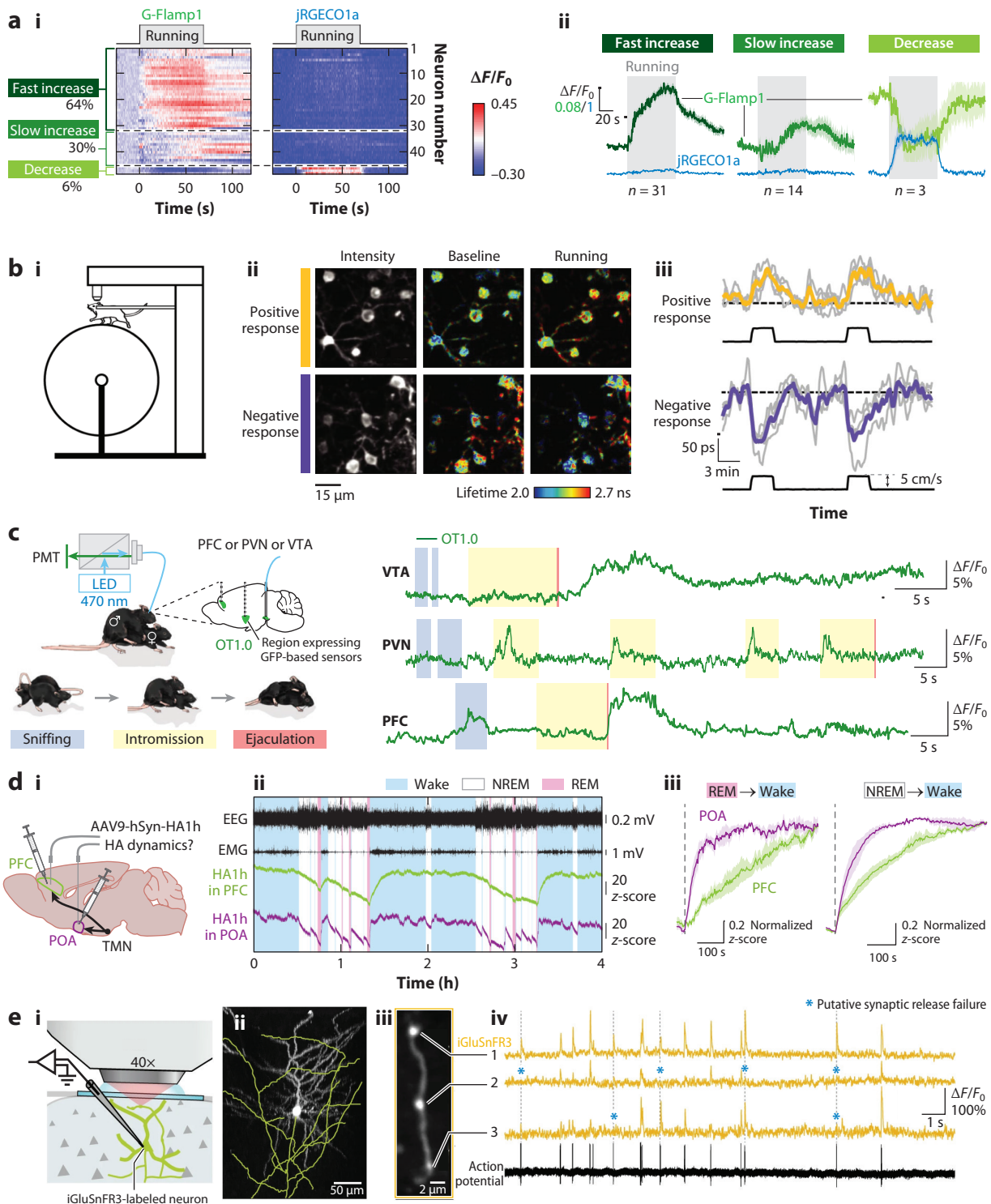
Two primary methods have been employed to capture fluorescence signals from biosensors in living animals: fiber photometry and microscopy. Fiber photometry enables the collection of an integrated fluorescence signal through an optical fiber from a targeted brain region; thus, it lacks spatial resolution. The surgical and imaging procedures associated with fiber photometry are user-friendly, and this technique can easily be applied to freely moving animals. This technique not only allows for deep tissue imaging but also enables simultaneous multichannel recording from different brain regions (100). In contrast, one-photon and multiphoton microscopy have been utilized to visualize neurochemical dynamics with spatial resolution. Prior to imaging, the cortical surface needs to be exposed through relatively complex surgical procedures, such as thinning the skull or replacing it with a cover glass (101, 102). Additionally, the implantation of a gradient index lens is necessary to access deep brain regions (103). While traditional microscopies are typically used on head-fixed animals, the recent development of miniature microscopes has enabled imaging in freely moving animals (104, 105). In summary, these two methods complement each other, and researchers should select the appropriate technique according to their specific scientific inquiry.

### 5.2. Recent Advances in Imaging cAMP Levels In Vivo

Studying the important role of cAMP in key processes requires the use of cAMP sensors in both physiological and pathological conditions. In the past dozen years, cAMP sensors have been used in cultured cells and other *in vitro* applications (32). However, compared with *in vitro* imaging, *in vivo* applications require sensors with higher brightness, a larger dynamic range, a sufficiently high signal-to-noise ratio (SNR), and an affinity for cAMP that is compatible with endogenous cAMP levels under physiological conditions. Considerable efforts by sensor developers and input from their end users have given rise to cAMP sensors suitable for *in vivo* imaging. In this section, we summarize recent advances in the use of genetically encoded cAMP sensors to monitor cAMP in living animals during biologically relevant behaviors.

First, the cpGFP-based sensor G-Flamp1 has been expressed in live zebrafish, *Drosophila*, and mice (42). When G-Flamp1 was coexpressed with the red  $\text{Ca}^{2+}$  sensor jRGECO1a to simultaneously monitor cAMP and  $\text{Ca}^{2+}$ , respectively, using two-photon *in vivo* microscopy in the mouse M1 cortex during forced running, differences between cAMP and  $\text{Ca}^{2+}$  dynamics were observed, suggesting that these two signals may be regulated via relatively independent processes (**Figure 3a**). G-Flamp1 is also compatible with *in vivo* fiber photometry and has been used in mice to report changes in cAMP levels during learning and memory in the nucleus accumbens in the ventral striatum. In addition, the FRET-based sensor cAMPFIRE (71) was combined with FRET-FLIM and two-photon microscopy for long-term imaging of cAMP levels in the mouse cortex for more than a month. The authors of this study (71) also used cAMPFIRE to monitor cell-specific changes in cAMP during forced running and identified three distinct groups of L2/3 pyramidal neurons that respond differently during running; specifically, one group of neurons had a positive response, another group had a negative response, and the third group had a small response (**Figure 3b**). Although further advances in sensor development are needed, these studies clearly illustrate the potential of using existing cAMP sensors in living animals during behavioral activities and in other biological contexts to better understand the functional role of cAMP and its interactions with other neurochemicals in the central nervous system.





**Figure 3** (Figure appears on preceding page)

In vivo applications of genetically encoded neurochemical sensors. (a) Two-photon dual-color in vivo imaging of cAMP and  $\text{Ca}^{2+}$  levels in M1 cortical neurons in mice during forced running. The neurons can be classified into three groups on the basis of their intrinsically heterogeneous cAMP and  $\text{Ca}^{2+}$  signals. (b) Two-photon FRET-FLIM imaging of cAMP in layer 2/3 cells during forced running. (i) Schematic of the enforced running behavioral paradigm. A head-fixed mouse is put on a motorized treadmill. (ii) Representative image of fluorescence intensity and lifetime of a cAMP sensor of individual neurons during forced running. (iii) Traces of a cAMP sensor from positive- and negative-response neurons during forced running. (c) Simultaneous recording of OT in the PFC, PVN, and VTA during various mating behaviors. (i) Schematic of the fiber photometry setup for recording OT signals. The optical fibers record signals in three different regions expressing the OT sensor. The mating process consists of three stages: sniffing, intromission, and ejaculation. (ii) The fiber photometry signals of the OT sensor in three brain regions during mating. (d) Long-term simultaneous recording of histamine in mouse PFC and POA during the sleep–wake cycle. (i) Schematic of the fiber photometry setup for recording histamine dynamics in the PFC and POA, which receive histaminergic projection from the TMN. (ii) Representative traces of EEG, EMG, and histamine sensor signal in the PFC and POA during the sleep–wake cycle. The wake state is shown in blue, REM sleep is in pink, and NREM sleep is unshaded. Overall, the changes in histamine are similar between these two brain regions, but the histamine signal in the POA has relatively fast kinetics during the sleep-to-wake transition. (iii) Averaged traces of histamine dynamics in the PFC and POA during the sleep state transition. (e) Simultaneous electrophysiological recording and iGluSnFR3 imaging of a single cortical neuron, with iGluSnFR3 reporting glutamate release from individual synaptic boutons. (i) Experimental schematic for simultaneous two-photon axonal imaging and cell-attached recording in vivo. (ii) A z-projection of iGluSnFR3-labeled neuron in mouse V1, with an axon outlined in yellow. (iii) Magnified image of a neuron axon expressing iGluSnFR3. Three neighboring boutons are labeled. The yellow traces represent the iGluSnFR3 signal from three neighboring boutons, asterisks indicate putative synaptic release failures, and the black trace represents the membrane potential of the same neuron. Abbreviations: EEG, electroencephalogram; EMG, electromyography; FRET-FLIM: Förster resonance energy transfer–fluorescence lifetime imaging microscopy; HA, histamine; LED, light-emitting diode; LT, lifetime; NREM, non-REM; OT, oxytocin; PFC, prefrontal cortex; PMT, photomultiplier tube; POA, preoptic area; PVN, paraventricular nucleus of the hypothalamus; REM, rapid eye movement; TMN, tuberomammillary nucleus; VTA, ventral tegmental area. Panel a adapted with permission from Reference 42; copyright 2022 Springer Nature. Panel b adapted with permission from Reference 71; copyright 2022 Springer Nature. Panel c adapted with permission from Reference 99; copyright 2023 Springer Nature. Panel d adapted with permission from Reference 93; copyright 2021 Elsevier. Panel e adapted from Reference 84 (CC BY 4.0).

### 5.3. Using Neurotransmitter and Neuromodulator Sensors to Gain Insights into In Vivo Biological Processes

Fully understanding how the brain works requires the ability to observe neural activity in vivo with high spatiotemporal resolution and minimal experimental interference. Thus, our growing toolbox of neurochemical sensors now allows researchers to tackle many key questions and design experiments with increased flexibility, thereby advancing our understanding of brain function.

The ability to capture dynamic changes in neurochemicals in mice engaged in natural and learned behaviors can increase our understanding of the neural mechanisms that underlie these behaviors. As mentioned above, fiber photometry is an exceptionally good tool for studying neurochemical changes of deep brain regions in freely moving mice. Several studies have used DA sensors to study changes in DA release during both innate and learned behaviors. For example, Dai et al. (106) found sexual dimorphism in dopaminergic signaling in the nucleus accumbens during mating and aggression. In addition, by measuring DA and cAMP signaling in the hypothalamus, Zhang et al. (107) identified a group of specialized dopaminergic neurons that motivate mating behaviors in male mice by activating the cAMP-PKA pathway. The peptide hormone and neuromodulator OT have also been studied using two independently developed sensors called GRAB<sub>OT</sub> (99) and MATRIA<sub>OT</sub> (98). MATRIA<sub>OT</sub> has been used to detect changes in OT under several conditions, including social interactions and acute stress, while GRAB<sub>OT</sub> has been used to simultaneously record the soma and projection targets during mating behaviors, revealing compartmentalized OT release (Figure 3c).

How neurochemical signaling regulates the sleep–wake cycle is another significant yet open question in neuroscience. Studies have shown that specific brain regions play an essential role in maintaining the sleep–wake cycle (108, 109); however, the specific changes that occur in

various NTs and NMs during transitions between various sleep states—critical for understanding the function of individual neurochemicals—remain largely unexplored. The emergence of neurochemical sensors is now providing researchers the opportunity to tackle these questions. For example, using a newly developed genetically encoded adenosine sensor, Peng et al. (96) simultaneously recorded  $\text{Ca}^{2+}$  signals and adenosine levels in the mouse basal forebrain; surprisingly, they discovered a neuronal activity–dependent rapid increase in extracellular adenosine levels, even during REM (rapid eye movement) sleep. Dong et al. (93) recently used GRAB<sub>HA</sub> sensors to identify differences in histamine release between the prefrontal cortex and preoptic area during the sleep–wake cycle in freely moving mice, suggesting possible heterogeneity among histaminergic neurons and/or local regulation of histaminergic terminals in specific brain regions (**Figure 3d**). Another monoamine that plays an important role in arousal, NE, has also been studied during the sleep–wake cycle using GRAB<sub>NE</sub> sensors; specifically, fiber photometry recordings of NE dynamics revealed unique oscillations in NE release during non–rapid eye movement (NREM) sleep, which may play a critical role in shaping the microarchitecture of sleep (110). With respect to DA sensors, Hasegawa et al. (111) measured DA release in the basolateral amygdala during the sleep–wake cycle and found that a transient increase in dopaminergic activity during NREM sleep is essential for transitioning from NREM sleep to REM sleep. Finally, several groups have studied other extracellular neurochemicals, including ACh (112), serotonin (91), DA, ATP (113), and orexin (114).

Fluorescent sensors can also provide remarkably rich information regarding the spatial dynamics of neurochemicals. To study neuromodulation at cellular resolution, investigators have applied one-photon and multiphoton microscopy to neurochemical imaging. For example, the iGluSnFR series has successfully been used to image glutamate release *in vivo*. The recently developed iGluSnFR3, which has a larger dynamic range when excited using a two-photon laser, has been used to monitor glutamate release in individual synaptic boutons (84) (**Figure 3e**). Using a highly sensitive ATP sensor, Wu et al. (95) found a localized increase in ATP release during lipopolysaccharide-induced systemic inflammation; a further analysis indicated that the early release of ATP in the cortex occurs near blood vessels. Patriarchi et al. (90, 115) used dLight sensors to measure cortical DA release during a reward-associated visuomotor task and found that DA release was linked to specific motor or reward behaviors, with a spatially heterogeneous distribution.

Interestingly, NM signals can even be measured in deep brain structures by implanting imaging cannulas. For example, Hamid et al. (116) combined  $\text{Ca}^{2+}$  imaging with DA imaging in the mouse dorsal striatum and observed wavelike DA signals during instrumental and Pavlovian tasks, with opposing wave directions. This observation suggests that the spatiotemporal propagation of DA signaling is critical for shaping learning behaviors. Using miniature two-photon microscopy, which capitalizes on the high spatial sensitivity of neurochemical sensors in freely moving mice while behaving naturally, Jing et al. (112) observed increased ACh release during locomotion, but not in response to visual or auditory stimuli.

At the other end of the scale, measuring changes in neurochemicals at the mesoscopic level has been instrumental in addressing questions regarding how various brain regions coordinate to support overall brain function. Although mesoscopic imaging techniques do not offer single-cell resolution, they facilitate the simultaneous monitoring of neurochemicals over relatively large areas (117). In early studies, voltage-sensitive dyes were widely used with wide-field imaging to monitor cortical function *in vivo* (118, 119), revealing patterns of spontaneous and sensory-evoked activity across large regions. The widespread use of GECIs now enables measurements of neuronal  $\text{Ca}^{2+}$  signals throughout the entire dorsal cortex. Neurochemical sensors for NTs and NMs were recently used to monitor changes in neurochemicals on a large scale.



For example, Lohani et al. (120) used dual-color wide-field imaging to study the relationship between ACh and  $\text{Ca}^{2+}$  signaling throughout the dorsal cortex and found spatially heterogeneous cholinergic activity that was differentially associated with distinct behavioral states. In addition, using wide-field imaging, we recently capitalized on the rapid kinetics of neurochemical sensors and found wavelike propagation of  $\text{Ca}^{2+}$ , serotonin, and endocannabinoids during seizures (121).

Several NM sensors have also been used to measure neurochemicals in *Drosophila* and zebrafish (89, 92, 95). For example, we recently imaged serotonin and ACh release in the *Drosophila* mushroom body (MB), which serves as the learning center, and found spatially heterogeneous serotonin signals that bidirectionally regulate the coincidence time window in a Pavlovian learning paradigm (122). In another study, Stahl et al. (123) examined olfactory learning in *Drosophila* and found bidirectional modulation of ACh signals in the MB that was specific to the valence of the learning event (i.e., aversive versus appetitive) and the MB compartment.

## 6. FUTURE PERSPECTIVES

The rapid development of genetically encoded sensors has in many ways revolutionized the field of neuroscience. However, single-FP-based sensors and their intensimetric imaging modalities have intrinsic limitations in terms of quantitative measurement and multiplexed imaging. In this section, we describe three novel imaging modalities and the corresponding sensor designs that aim to overcome these challenges.

### 6.1. Quantitative Imaging

Many of the currently popular neurochemical sensors used for in vivo imaging use a single fluorophore and rely heavily—if not entirely—on a change in fluorescence intensity to report ligand binding. Intensity-based measurements are affected by perturbations and extrinsic factors such as the local fluorophore concentration, wavelength-dependent light scattering, and fluctuations in excitation. Therefore, sensors based on intensity can report only qualitative, relative changes in the ligand of interest, irrespective of their basal concentration.

To overcome these limitations in the context of neurochemical sensors, the sensor's readout should be linked directly to ligand concentration. Although doing so remains a challenge, several strategies have been proposed. Each is discussed below.

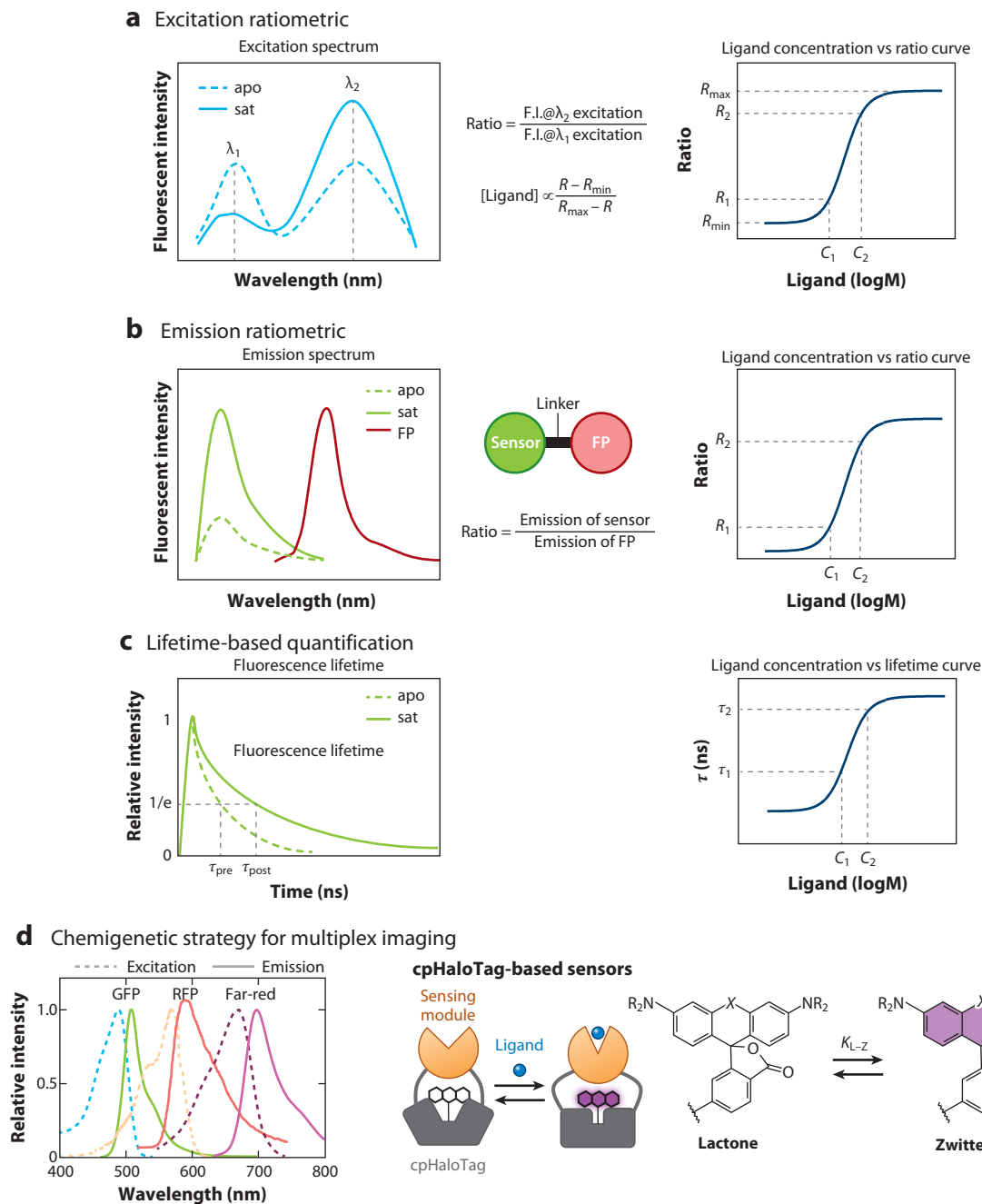
**6.1.1. Ratiometric sensors.** Ratiometric sensors are characterized by a change in either their excitation or emission wavelength in response to a change in the sensor's state. Ideally, this change in excitation or emission wavelength should be determined solely by the number of ligand molecules bound to the sensors. Therefore, ratiometric sensors have an intrinsic reference system that can mitigate the unwanted effects of external factors and can effectively provide a sensitive, quantitative measurement of the corresponding ligand. Typically, intensities at two or more wavelengths of either the excitation or emission spectrum are measured. Ratiometric sensors are therefore categorized into excitation and emission ratiometric sensors.

With excitation ratiometric sensors, the sensor's excitation curve shifts when bound to the respective ligand. Thus, by exciting the sensor using two distinct wavelengths and recording a fixed emission wavelength, one can calculate the emission ratio between the two excitation wavelengths and determine the sensor's state, which—when calibrated properly—should correspond to the ligand concentration (**Figure 4a**). The idea of excitation ratiometric sensors emerged from the  $\text{Ca}^{2+}$  dye fura-2, developed by Tsien and collaborators in 1985 (46). This dye's excitation peak undergoes a blue shift from  $\sim 362$  nm to  $\sim 340$  nm when saturated with  $\text{Ca}^{2+}$ , and it has been widely





used to image changes in  $\text{Ca}^{2+}$  concentration. Although fura-2 is not a genetically encoded sensor, the underlying principle has been used to develop a number of excitation ratiometric sensors (50, 124–127). In theory, this approach can be used to determine ligand concentration, with the calculated ratio corresponding to the percentage of chromophores residing in each of the neutral and



(Caption appears on following page)

**Figure 4** (Figure appears on preceding page)

Developments in neurochemical sensors for use in quantitative measurements and multiplex imaging. (a) Schematic representation of excitation ratiometric sensors in ligand-free (apo) and ligand-saturated (sat) states. (Left) Representative excitation spectrum of an excitation ratiometric sensor. The sensor is excited using two wavelengths ( $\lambda_1$  and  $\lambda_2$ ) with different absorptions, and the emission light is collected from each excitation wavelength. (Right) The ratio between the emission intensity excited by wavelength  $\lambda_1$  and the emission intensity excited by  $\lambda_2$  can then be used to quantify ligand concentration. (b) Schematic representation of emission ratiometric sensors. By fusing a constitutively fluorescing FP to an intensimetric sensor serving as a reference, one can use the emission from two chromophores to calculate an emission ratio and quantify ligand concentration. (c) Principle of FLIM-based quantification, in which a difference in fluorescence lifetime directly corresponds to ligand concentration. (d) Schematic diagram of using a HaloTag-based strategy in multiplex imaging. Ligand binding to the sensing domain induces a conformation change in the sensor, causing the equilibrium of the chromophore (rhodamine dye) in the sensor to shift to the fluorogenic zwitterion state and thereby increasing F.I. Abbreviations: cp, circularly permuted; F.I., fluorescence intensity; FLIM, fluorescence lifetime imaging microscopy; FP, fluorescent protein; GFP, green fluorescent protein; RFP, red fluorescent protein. Figure adapted from Reference 61.

ionized states. This approach can also help reduce the effects of differences in sensor expression levels and movement-related artifacts.

Emission ratiometric sensors typically use two distinct emission wavelengths; one changes in response to ligand binding, and the other is used as a fixed reference to minimize nonspecific perturbations (**Figure 4b**). An emission ratiometric sensor can be constructed by, for example, fusing a chemically stable FP with a spectrally distinguishable emission wavelength to an intensity-based sensor (128, 129). Alternatively, some sensors undergo a shift in their emission peak in response to ligand binding, allowing for ratiometric imaging (56). Despite their advantages, excitation and emission ratiometric sensors are still subject to wavelength-dependent scattering, and differences in experiment parameters (e.g., the excitation and/or emission wavelength) can also affect the signal readout.

**6.1.2. Fluorescence lifetime-based sensors.** Fluorescence lifetime is defined as the average time in which an excited fluorophore remains in the excited state. Because the number of photons emitted by a group of excited fluorophores decays exponentially, fluorescence lifetime is defined as the time it takes for the number of emitted photons to decrease to  $1/e$  (or approximately 36.8%) of the original fluorescence. Unlike intensity-based measurements, fluorescence lifetime is independent of factors such as fluorophore concentration and excitation intensity and can be measured quantitatively using FLIM. Therefore, fluorescence lifetime measurements have intrinsic advantages over intensity measurements, as they can provide a robust signal readout that is relatively unaffected by external factors that would normally affect intensity. As a result, fluorescence lifetime may be suitable for quantitative measurements (**Figure 4c**).

The idea of using fluorescence lifetime for quantitative imaging has a fairly long history; indeed, researchers first used this approach in the 1990s with pH-sensitive chemical dyes to quantitatively measure intracellular pH in living cells (130). Much more recently, Zheng et al. (29) used the chemical dye Oregon green BAPTA-1 to map  $[Ca^{2+}]_i$  in cultured neurons, acute brain slices, and living mice. Although genetically encoded fluorescence lifetime sensors have not yet been used for quantitative imaging, recent progress has brought this goal one step closer. For example, as noted above, the FRET-FLIM-based sensor cAMPFIRE developed by Massengill et al. (71) has a relatively robust lifetime readout lasting more than 1 month, an essential feature for future applications using quantitative imaging. In addition, van der Linden et al. (31) used a single-FP-based genetically encoded  $Ca^{2+}$  sensor as a lifetime-based biosensor to quantitatively image intracellular  $Ca^{2+}$ .

The commercial availability of specialized FLIM instruments is increasing, and studies are attempting to optimize these tools to increase the temporal resolution of FLIM microscopes. With these and other advances in both sensor development and imaging instruments, the use of

FLIM-based sensors and imaging techniques will likely lead to major advances in our ability to visualize and quantify neurochemicals.

## 6.2. Expanding the Spectra of Sensors for Use in Multiplex Imaging

The ability to use multiplex imaging to measure multiple neurochemicals simultaneously will represent a major step toward obtaining a comprehensive understanding of signaling in neural networks. To reach this goal, sensors must be developed with spectrally distinct spectra. Currently, most intensity-based sensors use cpGFP as the fluorophore, with an excitation peak at 488 nm and an emission peak at 507 nm (37), and some  $\text{Ca}^{2+}$ , cAMP, and DA sensors use red-shifted fluorophores, as discussed above. Developing sensors with unique spectra will allow researchers to perform multiplex imaging of neurochemicals, providing important new insights into complex neural processes. For example, red-shifted sensors in the far-red or near-infrared (NIR) region of the spectrum are needed to expand multiplex imaging. These sensors will also have practical advantages over existing sensors, including reduced light scattering and deeper tissue penetration, thereby facilitating their use in *in vivo* imaging.

One approach to developing genetically encoded far-red and NIR sensors involves the use of cpFPs with these spectral properties. Over the past two decades, several far-red and NIR FPs have been developed (131–136), and some of them have been used to generate genetically encoded sensors. Although these far-red and NIR sensors have been used for *in vivo* imaging in various organisms (137), they have a relatively low SNR. This is not unexpected, given that most far-red and NIR FPs have relatively low molecular brightness, which hinders their use *in vivo*. Therefore, steps to further optimize current far-red and NIR FPs should focus on increasing their brightness, and thereby increasing their functionality, particularly *in vivo*.

Another viable strategy is based on the combination of chemical dyes and chemically selective genetically encoded self-labeling proteins, which provides cellular specificity. Chemical dyes generally have higher molecular brightness and photostability compared with FPs; moreover, researchers have made strides toward increasing their solubility and their ability to cross cell membranes, as well as their ability to cross the blood–brain barrier, making them suitable for *in vivo* applications in the brain (138). For example, the Janelia fluor (JF) series of fluorescent dyes, developed by the Lavis group (139, 140) at Janelia Research Campus (Ashburn, Virginia), meets many of these requirements.

An array of genetically encoded self-labeling proteins has been developed to overcome the lack of cell specificity with chemical dyes (141–144). For example, HaloTag, which is derived from a bacterial haloalkane dehalogenase enzyme, binds specifically to synthetic HaloTag ligands that contain chloroalkane groups (141). Originally designed to label proteins, this strategy has been adapted to develop chemigenetic hybrid sensors, which combine cellular specificity with a high SNR (**Figure 4d**). The genetically encoded voltage indicator Voltron, developed by the Schreier group (145), uses the voltage-sensitive microbial rhodopsin Ace2N domain and the HaloTag protein. Upon the binding of HaloTag to a JF dye, a change in the cell's membrane potential induces a conformational change in the Ace2N domain, thereby altering FRET efficiency between the HaloTag–JF dye complex and Ace2N. Importantly, Voltron has been successfully used in several *in vivo* models, including mice, zebrafish larvae, and *Drosophila*. Furthermore, protein tags can be circularly permuted in a fashion similar to cpFPs to expose the chemical dyes and modulate their intensity, as demonstrated by the development of rHCaMP (146) and HaloCaMP (61).

These advances in chemigenetic hybrid sensors are paving the way toward the development of next-generation sensors designed to visualize neurochemicals. Choosing the ideal combination



of excitation and emission wavelengths will enable the addition of multiplex imaging to future studies.

## 7. CLOSING REMARKS

### 7.1. Limitations of Genetically Encoded Sensors

In this section, we discuss the limitations of current genetically encoded sensors. First, ethical concerns prohibit their use in humans, and such use is unlikely to be allowed in the foreseeable future. Second, existing intensimetric sensors provide information only about relative changes in chemical dynamics, without offering absolute concentration measurements. Third, the capability for multichannel imaging is hindered by the lack of spectrally orthogonal sensors, as most sensors are based on GFP. Fourth, deep brain imaging remains a challenge because of the complex procedures involved in gradient index lens implantation and the potential tissue damage that accompanies it. Although three-photon imaging allows for deeper tissue penetration (up to 1.6 mm), it also presents a new challenge in the development of customized sensors for multiphoton imaging (147–149).

### 7.2. Future Directions for Developing and Using Genetically Encoded Sensors

Thanks to numerous advances in the development of genetically encoded sensors, researchers can now observe dynamic changes in numerous neurochemicals at an unprecedented level of spatial and temporal resolution. Furthermore, GPCR-based sensors hold significant promise for use in drug screening studies, particularly because they are widely targeted by many therapeutic compounds.

Although the development of neurochemical sensors has made significant progress, several key challenges remain, including expanding the repertoire of sensors, achieving extensive multiplex imaging, and translating sensor readouts into quantitative measurements (e.g., ligand concentration, pH); reaching these important goals will likely require innovation and collaboration. Fortunately, the fields of biology and physiology are witnessing a surge in technologies and tools that can facilitate sensor development and help unravel the mysteries surrounding the role of neurochemicals in the central nervous system. With respect to developing new sensors, structure–function predictions based on computational algorithms (86) and insights provided by structural biologists can be used as a frame of reference for the rational design, evolution, and optimization of these sensors. For end users, genetically encoded sensors can be exploited using a wide range of tools, including—but not limited to—optogenetic and chemigenetic manipulation, pharmacology, and CRISPR/Cas9-mediated gene editing. By integrating these tools and engaging in multidisciplinary collaborations, we can push past the boundaries around our knowledge of neurochemicals and their role in both health and disease.

## DISCLOSURE STATEMENT

Y.L. is listed as an inventor on a pending patent application filed by Peking University (international patent number PCT/CN2018/107533), the value of which might be affected by this publication. The other authors are not aware of any affiliations, memberships, funding, or financial holdings that might be perceived as affecting the objectivity of this review.

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