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Title

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Permalink

<https://escholarship.org/uc/item/6nc5d23b>

Journal

Cell Chemical Biology, 29(11)

ISSN

2451-9456

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Publication Date

2022-11-01

DOI

10.1016/j.chembiol.2022.10.004

Peer reviewed



Published in final edited form as:

Cell Chem Biol. 2022 November 17; 29(11): 1649–1660.e4. doi:10.1016/j.chembiol.2022.10.004.

Multiplexed bioluminescence imaging with a substrate unmixing platform

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SUMMARY:

Bioluminescent tools can illuminate cellular features in whole organisms. Multi-component tracking remains challenging, though, owing to a lack of well-resolved probes and long imaging times. To address the need for more rapid, quantitative, and multiplexed bioluminescent readouts, we developed an analysis pipeline featuring sequential substrate administration and serial image acquisition. Light output from each luciferin is “layered” on top of the previous image, with minimal delay between substrate delivery. A MATLAB algorithm was written to analyze bioluminescent images generated from the rapid imaging protocol and deconvolute (i.e., unmix) signals from luciferase-luciferin pairs. Mixtures comprising 3–5 luciferase reporters were readily distinguished in under 50 minutes; this same experiment would require days using conventional workflows. We further showed that the algorithm can be used to accurately quantify luciferase levels in heterogeneous mixtures. Based on its speed and versatility, the multiplexed imaging platform will expand the scope of bioluminescence technology.

In Brief

Bioluminescent probes are routinely used to track biological processes *in vitro* and *in vivo*. Multiplexed imaging remains challenging, though, owing to a lack of well-resolved probes and

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AUTHORSHIP CONTRIBUTIONS

C.K.B., Z.Y., A.A.I., C.M.R., and J.A.P. were responsible for the conceptualization and methodology of the project. C.K.B., Z.Y., A.A.I., and B.S. generated reagents and performed the experiments. C.K.B. wrote the *SubstrateUnmixing* algorithm. C.K.B., Z.Y., A.A.I., and J.A.P. analyzed data. The initial manuscript draft was written by C.K.B., Z.Y., and J.A.P., and input was provided by all authors. C.K.B., Z.Y., and J.A.P. reviewed and edited the final version of the manuscript. J. A. P. was responsible for project administration and funding acquisition.

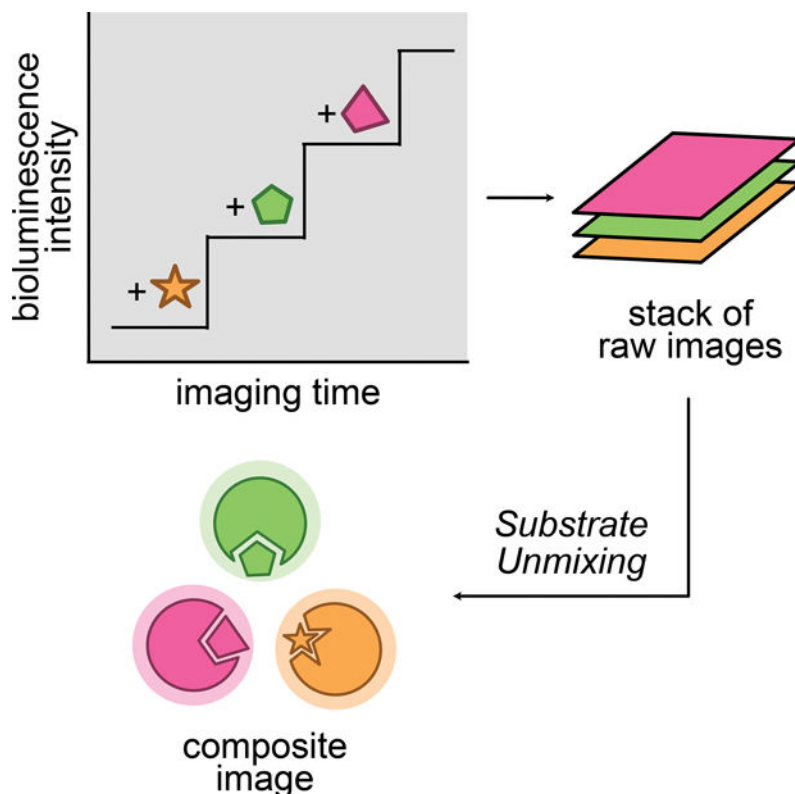
DECLARATION OF INTERESTS

J.A.P. is a member of the Editorial Advisory Board for *Cell Chemical Biology*.

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lengthy times. Brennan, *et al.* report a straightforward platform for rapid, quantitative analyses of bioluminescent mixtures.

Graphical Abstract



INTRODUCTION

Bioluminescence imaging (BLI) with luciferase-luciferin pairs is a popular method for visualizing cells and other biological features in heterogeneous environments (Syed and Anderson, 2021; Yeh and Ai, 2019). BLI relies on photons generated from the oxidation of small molecule luciferins by luciferase enzymes (Figure 1a) (Love and Prescher, 2020; Sanford and Palmer, 2017). Luciferases can be incorporated into many cell types and report on biological phenomena, including cell movements, proliferation, and gene expression (Contag and Bachmann, 2002; Paley and Prescher, 2014). Since no external light is needed, sensitive imaging in thick tissues and whole organisms is possible. As a few as 1–10 cells can be reliably tracked using optimized probes (Liu et al., 2010; Rabinovich et al., 2008). Recent years have also seen a surge in bioluminescent tool development for various applications (Love and Prescher, 2020; Yao et al., 2018). Engineered enzymes and analogs with enhanced light output (Hall et al., 2012; Iwano et al., 2018), red-shifted emission (Chu et al., 2016; Kuchimaru et al., 2016; Yeh et al., 2017), and improved biocompatibility (Su et al., 2020; Yeh et al., 2019), are now widely available (Rathbun and Prescher, 2017; Yeh and Ai, 2019). Additionally, an increasing number of *in vitro* assays have capitalized on the user-friendly features and broad compatibility of BLI (Elledge et al., 2021; Griss et al.,

2014; Quijano-Rubio et al., 2021; Thorne et al., 2010). Recent examples include methods for monitoring protein-protein interactions (Dixon et al., 2016; Machleidt et al., 2015) and high-throughput drug discovery (Fan and Wood, 2007; Thorne et al., 2012), among others (Dale et al., 2019; Kobayashi et al., 2019).

While popular, traditional bioluminescent probes have been slow to transition to imaging more than one target at a time in live cells and organisms. This is due, in part, to a lack of methods and probes for routine multiplexing (Love and Prescher, 2020; Rathbun and Prescher, 2017). Bioluminescent reporters cannot be easily separated based on emission wavelengths alone, and the inherent broadness of their spectra makes spectral resolution difficult to achieve at varying tissue depths. Improved separation can be achieved with more refined optical filters and spectral unmixing algorithms, but most examples have been limited to two probes (Aswendt et al., 2019; Kleinovink et al., 2019; Stowe et al., 2019; Zambito et al., 2021). Luciferases can also be differentiated based on luciferin recognition patterns (substrate resolution) (Jones et al., 2017; Williams and Prescher, 2019). In these cases, light is produced when complementary enzymes and substrates interact. Most examples to date have featured naturally orthogonal probes with perfect selectivity (Maguire et al., 2013; Moroz et al., 2021; Williams and Prescher, 2019). Larger collections of compatible luciferases have been difficult to identify, though, due to the promiscuous nature of the engineered enzymes and cross-reactivities with different substrates.

Routine implementation of orthogonal probes has been further limited by lengthy imaging times (Moroz et al., 2021; Rathbun et al., 2021; Rathbun et al., 2017; Su et al., 2020). Substrates are typically administered at saturating doses to maximize photon output, and the signal from the preceding injection must clear before the next substrate can be delivered (Figure 1b). This process can take ~8–24 h depending on the substrate (Gross et al., 2007; Sim et al., 2011). The imaging window can be shortened by combining spectral and substrate resolution. For example, Mezzanotte and coworkers showed that dual tracking is possible within 3–4 h using NIR probes that were both spectrally distinct and processed by two unique enzymes (Zambito et al., 2021). Imaging a larger number of probes within a practical time frame, though, still remains challenging.

To speed up multi-substrate BLI, we recently adopted a strategy termed substrate unmixing that features sequential acquisition and signal deconvolution (Gammon et al., 2006; Rathbun et al., 2021). In this approach, each substrate is added with minimal delay, and a final processing step is used to delineate the signals (Figure 1c). Each luciferin thus provides a unique emission fingerprint (or “barcode”) with the collection of enzymes. Conceptually, substrate unmixing is similar to time-domain fluorescence imaging in which the lifetimes of excited state fluorophores are used to deconvolute multiple spectral elements (Berezin and Achilefu, 2010; Datta et al., 2020). Analogous lifetime measurements are difficult to obtain for bioluminescence probes, but barcoding changes in reactivity (via orthogonal substrate application) is a sufficient substitute. Related approaches have expanded the number of detectable targets in fluorescence-based transcriptomics and synaptic profiling (Eng et al., 2019; Guo et al., 2019; Seo et al., 2022). The substrate unmixing approach was compatible with existing spectral unmixing algorithms for *in vitro* and *in vivo* BLI (Rathbun et al., 2021; Sarrion-Perdigones et al., 2019). However, the method required

multiple post-processing steps, including the generation of simulated input images. The algorithm also relied on pseudo-linear equations, where the number of input images could exceed the number of output images. We aimed to use perfectly linear equations (one input to one output) for improved and streamlined multi-substrate BLI. The ideal method would exclude unnecessary inputs and provide easy-to-analyze outputs compatible with ImageJ and other popular processing platforms.

Here we report a platform for rapid, quantitative orthogonal imaging featuring a MATLAB-based linear unmixing algorithm. This algorithm, dubbed *SubstrateUnmixing*, was first validated using three engineered firefly luciferase mutants: Cashew, Pecan and Akaluc (Kitada et al., 2018; Rathbun et al., 2021). We demonstrated that the analysis is suitable for deconvoluting cell mixtures of multiple luciferases in heterogeneous environments. In all cases, complementary substrates were added in quick succession and acquisitions were complete on the minutes time scale (~10 min per target). *SubstrateUnmixing* also enabled five-component BLI tracking in one hour, based *solely* on substrate preference – a record in the field. We further established a quantification platform by coupling the algorithm with calibration curves in tandem. The versatility and speed of the imaging method expands the capabilities of bioluminescence for visualizing dynamic, multicomponent processes.

RESULTS

MATLAB linear unmixing algorithm for multi-substrate BLI.

We set out to develop a custom MATLAB algorithm, dubbed *SubstrateUnmixing*, for rapid, multi-component BLI. In previous work, we co-opted existing algorithms for deconvoluting spectrally resolved probes and applied them to collections of orthogonal luciferases (Gammon et al., 2006; Rathbun et al., 2021). These approaches required an input image where no emission filter was present. The “no filter” image was generated *de novo*, adding unnecessary computing time. Because this input is not required for substrate unmixing, we decided to remove it in our next-generation algorithm. The number of inputs would therefore equal the number of probes being resolved, better capitalizing on the linear acquisition workflow and unique reactivity patterns of engineered bioluminescent pairs. In addition, we included image processing steps (e.g., pixel value normalization) in the algorithm itself, making it turn-key for the end user.

SubstrateUnmixing was modeled after algorithms used to differentiate among ultrasound and related reporters (Farhadi et al., 2019; Gammon et al., 2006) (Figure 2). For an experiment featuring i luciferase-luciferin pairs, a total of i consecutive images would be acquired (one per pair). The signal in a given image would encompass photons generated from the corresponding luciferin (with all luciferases) *and* from the previous substrates. Consequently, the photon intensity value (p) in each pixel could be specified using a set of linear equations:

$$p(s_1) = c_1 I_1(s_1) + c_2 I_2(s_1) + \dots + c_i I_i(s_1) + R \quad (1)$$

where s is the designated substrate for an image (s_1 being the first substrate), I is the luciferase in the image, c is the relative contribution of each enzyme present, and R is

the residual signal and image noise from the previous acquisition. This equation could be written more simply for each pixel as:

$$p(s_1) = \sum_i c_i l_i(s_1) + R \quad (2)$$

The collective intensities for all pixels across the entire image could thus be represented in matrix format:

$$P = C \times L \quad (3)$$

where matrix P comprises columns for the pixel intensity recorded after each substrate addition, and rows for each pixel in the image. All signals collected from the experiment would be encoded in P . L is a reference matrix containing the light emission values from all possible luciferase-luciferin combinations in the experiment, including cross-reactivities between unmatched pairs. Each column in L corresponds to the luciferin used to acquire the image, and each row denotes the possible luciferase involved. Biochemically, the values in each column of matrix L represent the reactivity fingerprint, or “emission barcode,” of each substrate with the collection of enzymes. Each barcode would be unique for a given probe set. The values in L could be influenced by substrate pharmacokinetics and signal dynamics (e.g., duration of residual signal from previous images), and would be determined empirically for each experiment. Based on these considerations, we anticipated that *SubstrateUnmixing* could differentiate probes based on a spectrum of reactivity, and would be well suited for multi-component imaging with larger sets of engineered (and imperfectly selective) luciferases. Promiscuous luciferase activity could actually be desirable, as the entire pattern would be considered when resolving the bioluminescent reporters.

Lastly, C is the matrix representing the unknown contribution of each luciferase reporter, the key variable in every experiment. Solving equation (3) for C would provide the luciferase identity and abundance (relative to the reference L) present in each pixel:

$$P \times L^{-1} = C \times L \times L^{-1} \quad (4)$$

and rearranging:

$$C = P \times L^{-1} \quad (5)$$

This solution could be used to generate a set of false color images for each luciferase-luciferin pair.

In previous work, we established that the unmixing approach required probes to be illuminated from dimmest to brightest. Each luciferin that is layered-in must be brighter than the preceding one (Kleinovink et al., 2019; Rathbun et al., 2021). We termed this criterion “intensity resolution” as the probes need to exhibit a range of photon outputs. Intensity resolution pushes residual signal from earlier probes into the background when successive, brighter probes are applied. This requirement is captured by equation (2), where in order for

an emission barcode to be registered and used for unmixing, the following expression must hold:

$$\sum_i c_i l_i(s_1) \gg R \quad (6)$$

SubstrateUnmixing similarly requires probes that are intensity resolved. The exact differential necessary varies based on the luciferase-luciferin pairs employed. Less intensity resolution is required for luciferases with more distinct luciferin usage patterns. By contrast, less selective luciferases require greater intensity differentials. The degree of resolution required for each probe set must be empirically determined, similar to matrix L . We initially focused on resolving probes with 10-fold or greater differences in emission intensity. This cutoff was selected based on the recognition that imaging reporters are typically used within their respective linear ranges of detection. Ten-fold differences in intensity would ensure that signals from recently applied probes would be clearly distinct from earlier ones. Moreover, previous work from us and Gammon, *et al.*, showed that such differentials can be readily unmixed with standard bioluminescent detectors (Gammon et al., 2006; Rathbun et al., 2021). Many sets of popular bioluminescent reporters also meet the criterion for 10-fold intensity resolution, including Fluc/D-luc, Akaluc/AkaLumine, and NanoLuc/furimazine (Hall et al., 2012). In assays where reporters are used outside of the linear range (e.g. $c_1 \gg c_2$), it is important to re-evaluate the order of substrate addition and choose one that satisfy the requirement set by equation (6).

Validating *SubstrateUnmixing* for rapid BLI.

With the custom algorithm in hand, we tested whether we could resolve a mixture of three established orthogonal reporters: Pecan, Cashew, and Akaluc (Figure 3a). Pecan and Cashew were previously engineered to preferentially process 4'-BrLuc and D-luc, respectively (Jones et al., 2017; Rathbun et al., 2017). Akaluc was engineered to use AkaLumine, a red-shifted luciferin analog (Iwano et al., 2013; Kitada et al., 2018). Based on the substrate preference profiles, we reasoned that this combination of luciferases would be well suited for validating *SubstrateUnmixing*. We further ranked the probes from dimmest to brightest based on their reported catalytic efficiency: Pecan/4'-BrLuc < Cashew/D-luc < Akaluc/AkaLumine (Iwano et al., 2018; Rathbun et al., 2021). Consequently, the maximum resolution would be achieved when the luciferins were added in the following order: 4'-BrLuc, then D-luc, followed by AkaLumine. Cross-reactivities between unmatched luciferase-luciferin pairs (i.e., Cashew/4'-BrLuc) would likely be observed, but we anticipated a substantial increase in light emission when the brighter luciferin was added.

We first evaluated the triplet set using bacterial and mammalian cells expressing Pecan, Cashew, or Akaluc. The cells were distributed across a 96-well plate as single populations (Figure 3b, Figure S1). 4'-BrLuc was added first, and an image was acquired. When D-luc was added, the resulting image contained the expected signal from Cashew. Residual signal from 4'-BrLuc was also present, but the output was much dimmer than the signal generated from D-luc. This same trend was observed when AkaLumine was added to the wells, with Akaluc emission dominating the final signal output. The raw images collected were then

subjected to *SubstrateUnmixing*. All three enzyme-substrate pairs were readily discerned in the resulting unmixed images (Figure 3c). In this particular case, the probe set was sufficiently resolved such that the identities of the reporters were apparent from the raw images alone, prior to *SubstrateUnmixing*. It is important to note, though, that the unmixing platform did not interfere with the interpretation of the images. Signal deconvolution often becomes more necessary as the number of probes or overall complexity of the sample increases, resulting in images with undesired signals (e.g., Cashew lanes in Figure 3b). For these reasons, we recommend that *SubstrateUnmixing* be employed across all probe sets.

We anticipated that the imaging platform would be able to resolve complex mixtures of reporters. To confirm, we varied the number of bacterial cells across a 96-well plate (Figure 3d). Some wells contained two reporters, where the concentration of one was higher than the other. Substrates were administered sequentially, and an image was acquired between each addition. The reporters were successfully resolved using *SubstrateUnmixing*. For samples containing mixtures of reporters, the extracted pixel values correlated linearly with reporter load, suggesting efficient unmixing. Cross-talk was eliminated, as the algorithm uses the individual luciferin reactivity patterns (“barcodes”) to differentiate the luciferase reporters (Figure 3e). Notably, the imaging sessions for all triplet sets took ~30 min to complete, a drastic improvement over standard protocols requiring >24 h (Maguire et al., 2013; Rathbun et al., 2021). A second triplet set of reporters (Pecan/Akaluc/Antares) was validated using *SubstrateUnmixing* (Figure S2) (Yao et al., 2021b). In this case, luciferase-expressing mammalian cells were plated in gradients across a 96-well plate. Images were acquired as described above, and raw images were unmixed using a previously reported algorithm (Gammon et al., 2006), in addition to *SubstrateUnmixing*. Signals from “unmatched” luciferase-luciferin pairs were removed more effectively using the modified platform. Collectively, the *in vitro* tests demonstrated that *SubstrateUnmixing* can effectively deconvolute signals from luciferase reporters for multiplexed BLI.

We also examined whether *SubstrateUnmixing* could be used for processing *in vivo* images. Bioluminescence is routinely used for examining cell behavior and other biological processes in rodents and other preclinical models. In these cases, photons are typically acquired using a cooled CCD camera. The raw images generated from such acquisitions can be immediately uploaded to *SubstrateUnmixing*. Single populations of each luciferase reporter would then be used to generate matrix *L*. To evaluate the algorithm in this context, we analyzed a recently published dataset (Rathbun et al., 2021). In this experiment, different ratios of Pecan- and Cashew-expressing cells were implanted in mice. Images were acquired within ~1 h via sequential administration of 4'-BrLuc and D-luc. Subjecting the data to *SubstrateUnmixing* successfully deconvoluted and identified the composition of engrafted mixtures (Figure S3). Collectively, these results suggest that *SubstrateUnmixing* can be implemented to analyze data from both *in vitro* and *in vivo* multi-component studies.

Rapid imaging with an expanded probe set.

We envisioned that four- and five-component imaging could be readily achieved by combining the Pecan-Cashew-Akaluc triplet with other well-established bioluminescent tools. One potential candidate was PhOH-Luc, a pi-extended luciferin, and its

complementary luciferase, Almond (Yao et al., 2020). Given its unique structure, PhOH-Luc would likely exhibit a different emission barcode than 4'-BrLuc, D-luc, or AkaLumine with a collection of luciferases. Indeed, we had previously shown that Almond/PhOH-Luc could be resolved from other bioluminescent reporters using a conventional workflow, where each substrate was supplied and imaged separately (Yao et al., 2020). The photon output from Almond/PhOH-Luc is also lower than that of the other probes within the triplet set, providing the necessary intensity resolution for rapid imaging. When the four reporters were treated with luciferins in succession, multiplexed imaging was complete in ~40 minutes (Figure S4a–c).

In addition to Almond/PhOH-Luc, we selected a marine luciferase, NanoLuc, to form a quartet of bioluminescent reporters (Hall et al., 2012). NanoLuc and FRZ are commonly used in combination with insect luciferases for dual-component imaging (Rathbun et al., 2021; Stacer et al., 2013; Su et al., 2020). The unique structural features of FRZ and its blue emission make NanoLuc a versatile reporter for both spectral and substrate multiplexing. NanoLuc is also substantially brighter than the Pecan, Cashew, and Akaluc probes, and can thus be intensity resolved. When integrated into the substrate unmixing workflow, combinations of NanoLuc and the other luciferases were resolved in less than 1 h (Figure S4d–f). This result highlights the “plug-and-play” nature of the *SubstrateUnmixing* algorithm, in that different combinations of reporters could be imaged without the need for additional enzyme or substrate engineering.

We further generated a quintet of reporters by combining Almond, Pecan, Cashew, Akaluc, and NanoLuc (Figure 4a). The order of luciferin addition necessary to achieve intensity resolution was PhOH-Luc, 4'-BrLuc, D-luc, AkaLumine, then FRZ. When subjected to the luciferins, bacterial cells produced five sets of distinguishable barcodes (Figure 4b). Mixtures of luciferase-expressing bacteria were also examined. In this case, varying numbers of Almond-, Pecan-, Cashew, Akaluc- or NanoLuc-expressing cells were combined (Figure 4c–d). Serial imaging with the corresponding luciferins and unmixing enabled five-component imaging. Linear regression with the unmixed, pseudo-colored pixels further revealed the expected compositions (Figure S5). Similar to the preceding examples, the multicomponent experiments were completed within 1 hour (~10 min per reporter). Using conventional imaging conditions, such analyses would take several days.

Deconvoluting changes in reporter ratio.

Bioluminescent probes are often used to monitor relative changes in cell count or reporter expression within biological samples (Chaincy et al., 2007; Doyle et al., 2004; Gammon et al., 2006). We aimed to examine whether *SubstrateUnmixing* could report on such changes. Toward this end, we expressed the triplet set of reporters (Pecan, Cashew, and Akaluc) and diluted each culture to achieve altered reporter concentrations. The samples were then mixed such that the amount of one reporter was systematically varied and the other two were held constant (Figure 5a–c). Images were acquired and analyzed via *SubstrateUnmixing* (Figure S6a–b). As shown in Figure 5, the expected changes in luciferase signal were reflected in the intensity of the pseudo-colored pixels. Signals from channels where the reporters were held constant also remained stable across the range of ratios tested. The unmixed

signals were highly correlated with the amount of culture present, as revealed by linear regression analyses (Figure 5d and Figure S6c). Similar results were obtained using mixtures of mammalian cells (Figure S7).

SubstrateUnmixing was further applied to monitor changes in substrate concentration. Bacteria cells expressing Pecan, Cashew, or Akaluc were plated in equal ratios (Figure S8). During sequential substrate administration, cells were treated with either a constant amount of luciferin or varying levels. Changes in substrate concentration were successfully deconvoluted with *SubstrateUnmixing*. In all cases, unmixed signals correlated linearly with the substrate being varied.

Quantitative multicomponent analysis via *SubstrateUnmixing*.

In addition to multiplexed imaging in heterogeneous samples, *SubstrateUnmixing* can be used for quantitative analyses of bioluminescent mixtures. Determining reporter concentrations from bioluminescent images requires standard curves and external validation. Such curves are rarely generated for *in vivo* experiments due to the complexities of photon absorption and scatter in tissue environments (Doyle et al., 2004; Gammon et al., 2006). Thus, BLI typically involves tracking *relative* changes in expression levels or cell growth (similar to Figure 5). Standardizing bioluminescent signals is more common for *in vitro* analyses, but traditionally limited to one reporter at a time. Each reporter must be separately calibrated with each substrate, and this approach becomes impractical when larger numbers of reporters are involved.

With substrate unmixing, we reasoned that simplified standard curves could be readily integrated into the rapid imaging pipeline. Only one set of standardized reporters would be required per experiment. This hypothesis was tested using a blinded setup with the Almond-Pecan-Cashew-Akaluc reporter set. Bacterial cultures comprising each luciferase were randomly mixed to create 12 samples of unknown composition (Figure 6a). Each mixture was plated in duplicate over a 96-well plate, and a standard curve was included. Substrates were administered to both the unknowns and standards simultaneously with minimal delay (Figure S9a–b, total experiment time = 40 min). *SubstrateUnmixing* was then used to analyze the images. Pixel values were quantified from each well, and the amount of luciferase present was calculated using the standard curve. The approach successfully predicted the composition of the 12 mixtures ($m = 0.78–0.99$, Figure 6b), with the calculated values in agreement with the actual quantity of each reporter present (Figure S9c). *SubstrateUnmixing* was also successfully able to deconvolute mixtures of reporters in heterogeneous matrices. Bacterial cultures comprising Pecan, Cashew, or Akaluc were suspended in an agarose matrix to mimic a tissue environment (Figure S10a). The cultures were plated over a 96-well plate either as random mixtures or as a standard curve, similar to the experiment in Figure 6. Images were acquired at peak emission, approximately 20 minutes after each substrate administration. The amount of culture present in the mixtures was accurately estimated based on the pixel intensities after unmixing (Figure S10b). These data set the stage for further applications for streamlined, multiplexed imaging in whole animal models.

Monitoring dynamic changes over time.

BLI is uniquely suited for longitudinal imaging, as fluctuations in photon intensity correlate with biological changes (e.g., cell growth or gene expression) (Aalipour et al., 2019; Matta et al., 2018). However, monitoring heterogeneous samples over time remains difficult with traditional protocols. Most multiplexed BLI applications require sequential administration of the complementary luciferins, with lengthy times required for substrate clearance between each administration. Imaging times can be dramatically improved via rapid substrate application in conjunction with *SubstrateUnmixing*. To showcase the utility of the multi-component workflow, we imaged a heterogeneous mixture of breast cancer cells. The cells derive from organ-specific metastases in the MMTV-PyMT mouse model, and serve as useful reagents for analyzing cancer growth in complex environments (Ionkina et al., 2021). Tumor cells from the mammary fat pad (MFP), lymph node (LN), and lung were engineered to express different luciferase-fluorescent protein fusions (MFP Pecan-eGFP, LN Cashew-mNeptune, and lung Akaluc-TagBFP). The cells were then plated as single populations or three-component mixtures. The samples were then monitored over time via sequential substrate administration and *SubstrateUnmixing* (Figure 7a). For all time points, the populations were successfully deconvoluted (Figure 7b). Changes in luminescence correlated with changes in fluorescence (as measured by flow cytometry) and cell counts (Figures 7c–d and Figures S11–S14). Interestingly, we observed that the lymph node derived cells grew more rapidly than cells derived from the primary tumor or lung. Differences in proliferation were also observed depending on whether the cells were grown in isolation or as mixtures.

DISCUSSION

We developed a general platform for rapid, multiplexed imaging with bioluminescent probes. A MATLAB-based linear unmixing algorithm (*SubstrateUnmixing*) was written and used to deconvolute mixtures of luciferases in bacterial cells, mammalian cells, and mice. Probe differentiation relies on the unique emission barcodes generated from luciferin administration. The luciferins are simply added from dimmest to brightest, and the resulting patterns of emission provide a readout on the luciferases present. Using this approach, combinations of 3–5 luciferase reporters were readily distinguished in under one hour. *SubstrateUnmixing* is compatible with probes commonly used in the field (e.g., Akaluc/AkaLumine and NanoLuc/FRZ) and quantitative imaging is also possible. We further applied the imaging platform to monitoring changes in reporter ratios over time, including in a model of heterogeneous breast cancer.

The improved speed and multiplexing capabilities of *SubstrateUnmixing* will enable new BLI applications, including heterogeneous biological processes such as immune system activation, host-pathogen interactions, and cancer metastasis. In these experiments, multiple cell types or reporters can be monitored simultaneously in a single sample rather than over multiple days. Dynamic changes substrate ratios could also be easily deconvoluted, suggesting that the imaging platform can be integrated with studies where measuring luciferin levels is critical (e.g., caged probe release). Calibration curves can be readily integrated with *SubstrateUnmixing*, increasing the range of quantitative experiments that can

be applied to the platform. In addition, we anticipate that additional engineered luciferase-luciferin pairs can be integrated with this approach, thus maximizing the number of reporters that can be imaged in tandem.

Future studies will address whether more generalized reference signals (and associated matrices for unmixing) can be developed and used broadly between experiments. Such studies will be especially important in the context of tissues and *in vivo* models, where reporter attenuation is likely to be more pronounced. Bioluminescent probes with longer emission wavelengths (e.g., Akaluc/AkaLumine) are less susceptible to these effects and are thus desirable for *in vivo* work. Several efforts are underway in our lab and others to generate additional red-emitting probes. We will also explore whether additional modifications to the algorithm (e.g., using pseudo-imaging algorithms) can expand the number of imaging targets or improve the accuracy of multi-substrate calibration (Seo et al., 2022). Such advances will further expand the scope of the *SubstrateUnmixing* platform.

LIMITATIONS OF THE STUDY

While multi-substrate BLI with *SubstrateUnmixing* provides many advantages over conventional imaging protocols, limitations remain. For example, monitoring changes in mixed populations without a calibration curve or fluorescent protein standard remains challenging. Pixel values in unmixed images are quantified relative to a reference population, so they cannot be directly compared to other channels. Additionally, the requirement for a reference can be limiting in certain applications, especially *in vivo*. The bioavailability of luciferins in desired imaging locales is also an important consideration for future multiplexed imaging applications. Differences in pharmacokinetics and biodistribution could influence the observed reactivity patterns in tissues and other environments. While *SubstrateUnmixing* should be able to deconvolute signals as long as the fingerprint remains distinct, the reference matrix will likely need to be empirically determined to account for differences in substrate accessibility.

SIGNIFICANCE

Bioluminescence imaging (BLI) is a powerful tool for sensitive detection of cellular changes over time *in vitro* and *in vivo*. However, traditional BLI workflows do not allow for simultaneous imaging of live cell populations containing three or more reporters. We developed a platform, *SubstrateUnmixing*, for multi-substrate imaging of heterogeneous cell mixtures. The approach is compatible with readily available BLI tools and can provide quantitative, rapid deconvolution of 3–5 reporters. The imaging strategy was used to monitor the dynamics of cell growth in a heterogenous model of breast cancer.

STAR METHODS

RESOURCE AVAILABILITY

Lead contact—Further information and requests for resources and reagents should be directed to and will be fulfilled by the lead contact, Jennifer A. Prescher (jpresche@uci.edu)

Materials availability—Materials generated in this study are available upon request. Depending on the reagent and institution of origin, an MTA might be required.

Data and code availability

- All data reported in this paper will be shared by the lead contact upon request.
- All original code has been deposited at <https://github.com/ckbrenna/Substrate-Unmixing> and is publicly available as of the date of publication. DOIs are listed in the key resources table.
- Any additional information required to reanalyze the data reported in this work is available from the lead contact upon request.

EXPERIMENTAL MODEL AND SUBJECT DETAILS

All data are generated from the datasets provided in the KRT.

Cell culture conditions—MMTV-PyMT primary cell lines (MFP, LN, and lung) were originally derived from 10–12-week female FVB/NJ MMTV-PyMT mice (courtesy of the Kessenbrock lab, UCI) as reported in previous work (Ionkina et al., 2021). Both CD44^{low}/EpCAM^{high} and CD44^{high}/EpCAM^{high} cells were used for imaging. These cells or DB7 cells were engineered to express Pecan-eGFP, Cashew-mNeptune, or Akaluc-TagBFP via CRISPR-mediated gene insertion as previously described (Rathbun et al., 2021). All cells were cultured in DMEM (Corning) supplemented with 10% (v/v) fetal bovine serum (FBS, Life Technologies), penicillin (100 U/mL), and streptomycin (100 µg/mL). Cells expressing luciferases were further cultured with puromycin (2 µg/mL) to ensure gene integration. All cells were maintained in a 5% CO₂ water-saturated incubator at 37 °C. Cells were serially passaged using trypsin-EDTA (0.25% in HBSS, Gibco). The cell lines were not authenticated prior to use.

METHOD DETAILS

Compound handling and preparation—All reagents purchased from commercial supplies were of analytical grade and used without further purification. 4'-BrLuc and PhOH-Luc were prepared and used as described previously (Steinhardt et al., 2017; Yao et al., 2020). All compounds were stored dry at –80 °C until needed and then dissolved in DMSO or phosphate buffer (100 mM buffer, pH 8). Stock solutions were prepared at concentrations that varied with compound apparent solubility (10 mM in phosphate buffer for D-luc, 4'-BrLuc; 50 mM in DMSO for PhOH-Luc and AkaLumine).

Synthesis of 4'-Br-Luc—4'-Br-Luc was synthesized as described in the literature from commercially available 4-isopropoxy aniline (Steinhardt et al., 2017). The desired luciferin was isolated via acidification with 1 M NaHSO₄ and extraction with ethyl acetate, yielding a yellow solid. ¹H NMR (500 MHz, D₂O) δ 7.01 (m, 2H), 5.21 (m, 1H), 3.84 (m, 1H), 3.64 (m, 1H); ¹³C NMR (500 MHz, D₂O) δ 180.4, 168.3, 169.8, 159.3, 146.7, 139.9, 123.3, 119.2, 109.0, 82.8, 39.3. In accordance with literature (Steinhardt et al., 2017).

Synthesis of PhOH-Luc—PhOH-Luc was synthesized as described in the literature from commercially available 6-methoxybenzo[*d*]thiazol-2-amine (Yao et al., 2020). The final compound was isolated from the reaction mixture following acidification with 1 M NaHSO₄ and extraction with ethyl acetate. PhOH-Luc was isolated as an orange solid. ¹H NMR (600 MHz, (CD₃)₂SO) δ 11.7 (s, 1H), 10.0 (s, 1H), 8.25 (d, *J* = 8.3 Hz, 1H), 7.85 (d, *J* = 8.2 Hz, 1H), 7.53 (d, *J* = 8.8 Hz, 1H), 7.44 (d, *J* = 1.7 Hz, 1H), 7.40 (dd, *J* = 8.3, 1.7 Hz, 1H), 7.03 (dd, *J* = 8.8, 2.4 Hz, 1H), 5.33 (dd, *J* = 9.5, 8.3 Hz, 1H), 3.74 (dd, *J* = 11, 9.6 Hz, 1H), 3.64 (dd, *J* = 11, 8.2 Hz, 1H); ¹³C (150 MHz, (CD₃)₂SO) δ 172.2, 168.1, 160.3, 156.2, 155.9, 145.6, 137.0, 135.0, 129.0, 123.6, 122.3, 119.9, 116.8, 116.5, 106.7, 78.9, 35.3. In accordance with literature (Yao et al., 2020).

Bioluminescence imaging—All assays were performed in black 96-well plates (Grenier Bio One). Plates containing luminescent reagents were imaged in a light-proof chamber with an IVIS Lumina (Xenogen) CCD camera chilled to −90 °C. The stage was kept at 37 °C during the imaging session, and the camera was controlled using Living Image software. For all assays, exposure times were set to 1–180 s, and data binning levels were set to medium. Total flux values for regions of interest were analyzed using Living Image software. Integrated pixel values were analyzed using ImageJ (Installed under the FIJI package, NIH). The data were analyzed using GraphPad Prism (version 9.0 for Macintosh, GraphPad Software).

SubstrateUnmixing analysis—Substrate unmixing experiments were designed as previously described (Rathbun et al., 2021). Substrate unmixing was conducted with MATLAB R2020a (See Supplementary Discussion). Luminescence images containing the raw CCD counts (as TIFF files) were loaded into MATLAB. Images were subjected to a 2-pixel median filter (using the *medfilt2* function with a 5×5 neighborhood around the corresponding pixel). Next, the signal at each pixel was normalized to lie between 0 and 65536 (the maximum value that can be stored in a 16-bit image). As a result, the brightest pixel in each image had a value of 65536, and the dimmest had a value of 0. Regions of interest (ROIs) were generated by identifying the image coordinate of the reference well and input dimensions. Once assigned, the MATLAB algorithm was run to perform the unmixing. After unmixing, text images were imported into ImageJ (installed under the FIJI package). Integrated pixel values for regions of interest were analyzed using the “Measure” tool. Pseudocolors were assigned with the “Merge Channels” tool. In some cases, Bland-Altman plots were used to evaluate deviations in measurements obtained from unmixed images versus control samples. The averages of the deviations were shown as biases, and the standard deviations were indicated as 95% limits of agreements. All analyzes were performed using GraphPad Prism 9.

Bacterial cell analysis of luciferase mutants—*E. coli* BL21 cells expressing mutant luciferases (glycerol stocks, 50% v/v) were streaked on agar plates containing kanamycin sulfate (Kan, 40 µg/mL final concentration, Fisher Scientific). After overnight growth, colonies were picked and incubated in LB media (Genesee Scientific) supplemented with Kan (LB-Kan) at 37 °C with shaking (250 rpm) for 16–18 h. Aliquots of the starter cultures (100–200 µL) were used to inoculate 5 mL of LB-Kan media and grown to OD₆₀₀ = 0.8–0.9.

Protein expression was induced with 1 M isopropyl β -D-1-thiogalactopyranoside (IPTG, 2.5 μ L, 500 μ M final concentration, Gold Biotechnology), and the cultures were grown for 18 h with shaking (250 rpm). The cells were pelleted at 4000 rpm for 5 min, and then resuspended in 600 μ L buffer (50 mM Tris-HCl, 500 mM NaCl, 0.5% (v/v) Tween® 20, 5 mM MgCl₂, pH 7.4). Unless otherwise noted, samples were diluted prior to plating to remain in the linear range of detection (1:10 for Cashew, Pecan, Akaluc and Almond, 1:100 for NanoLuc). Cell lysate (90 μ L) was added to black 96-well plates, followed by luciferin solution (10 μ L, 100 μ M for D-luc, 4'-BrLuc, and AkaLumine, 250 μ M for PhOH-Luc with 1 mM ATP final concentration, 1:100 dilution for furimazine (Promega Corporation)). Plates were imaged and analyzed as described above.

Analysis of luciferase mutants in a tissue mimic—*E. coli* BL21 cells expressing Cashew, Pecan, or Akaluc were induced for protein expression as described above. After pelleting, the cells were resuspended in 2.5 mL buffer (50 mM Tris-HCl, 500 mM NaCl, 0.5% (v/v) Tween® 20, 5 mM MgCl₂, pH 7.4). Calibration curves and mixtures were prepared in resuspension buffer and 50 μ L of each suspension was mixed with 50 μ L of agarose (1% w/v in H₂O). The mixtures were plated across black 96-well plates and allowed to solidify. Buffer (90 μ L) was added, followed by luciferin solution (10 μ L, 100 μ M with 1 mM ATP final concentration). Plates were imaged and analyzed as described above.

Mammalian cell analysis of luciferase mutants—DB7 or MMTV-PyMT cells stably expressing luciferases were added to black 96-well plates. Cells were treated with a luciferin solution (10 μ L, 100 μ M for D-luc, 4'-BrLuc, and AkaLumine). Plates were imaged and analyzed as described above.

Flow cytometry methods—Samples were transferred to Eppendorf tubes and pelleted (500 \times g, 5 min) using a tabletop centrifuge (Thermo Fisher Sorvall Legend Micro 17). The resulting supernatants were discarded, and cells were washed with PBS (2 \times 100 μ L). Cells were then analyzed on a Novocyte Quanteon flow cytometer (ACEA Biosciences Inc). Live cells were gated and TagBFP+, eGFP+, and mNeptune+ cells were further gated. For each sample, 10,000 events were collected on the “Live cell” gate (see Figure S10 for raw plots). TagBFP, eGFP, and mNeptune fluorescence were analyzed and quantified using NovoExpress software (ACEA Biosciences Inc.).

Longitudinal imaging of luciferase-expressing cells—MMTV-PyMT cells derived from either the mammary fat pad (MFP), lymph node (LN), or lung expressing Pecan-eGFP, Cashew-mNeptune, or Akaluc-TagBFP, respectively, or a mixture of the three cell lines were seeded in triplicate in tissue culture treated 24-well plates (Corning, 5×10^4 cells per well). On day of seeding, cells were added to 96-well plates (100 μ L, 2.5×10^4 cells per well). On each subsequent day, cells were lifted with trypsin (100 μ L) with DMEM (Corning, 100 μ L) supplemented with 10% (v/v) fetal bovine serum (FBS, Life Technologies), penicillin (100 U/mL), and streptomycin (100 μ g/mL). A portion of the cells (100 μ L) were analyzed via flow cytometry as described above. The remaining cells (100 μ L) were added to 96-well plates. On each day including the day of seeding, a portion of the cells (10 μ L) were counted using Trypan Blue (10 μ L, 0.4%, Gibco) and Countess II FL Automated Cell Counter

(Fisher Scientific). The remaining cells (90 μ L) were imaged and analyzed as described above.

QUANTIFICATION AND STATISTICAL ANALYSIS

Unless indicated otherwise, experiments were performed with at least 3 independent replicates (n) and analyzed using Graphpad Prism (V. 9.2.1 or 7). Values are expressed as mean \pm standard deviation of the mean or SEM of the replicates, as indicated in the figure legends.

Supplementary Material

Refer to Web version on PubMed Central for supplementary material.

ACKNOWLEDGEMENTS

This work was supported by the U.S. National Institutes of Health (R01 GM107630 to J.A.P.). C.K.B was supported by the UCI Physical Sciences Machine Learning NEXUS program. Z.Y. was supported by the National Science Foundation via the BEST IGERT (DGE-1144901) program. Z.Y. and C.M.R. were supported by National Science Foundation Graduate Research Fellowships (DGE-1321846). A.I. was supported by the U.S. National Institutes of Health institutional Cancer Biology Training Grant (T32-CA009054). B.S. was supported by the UCI Undergraduate Research Opportunities Program. We thank M. Shapiro (Caltech) for help with writing the unmixing algorithm. We also thank members of the Prescher Lab for helpful discussions.

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Highlights

- Multiplexed bioluminescence imaging is possible with streamlined algorithm
- *SubstrateUnmixing* provides rapid readouts on mixture composition
- Mixtures of luciferase reporters can be readily quantified
- *SubstrateUnmixing* enables serial tracking of heterogenous cell populations

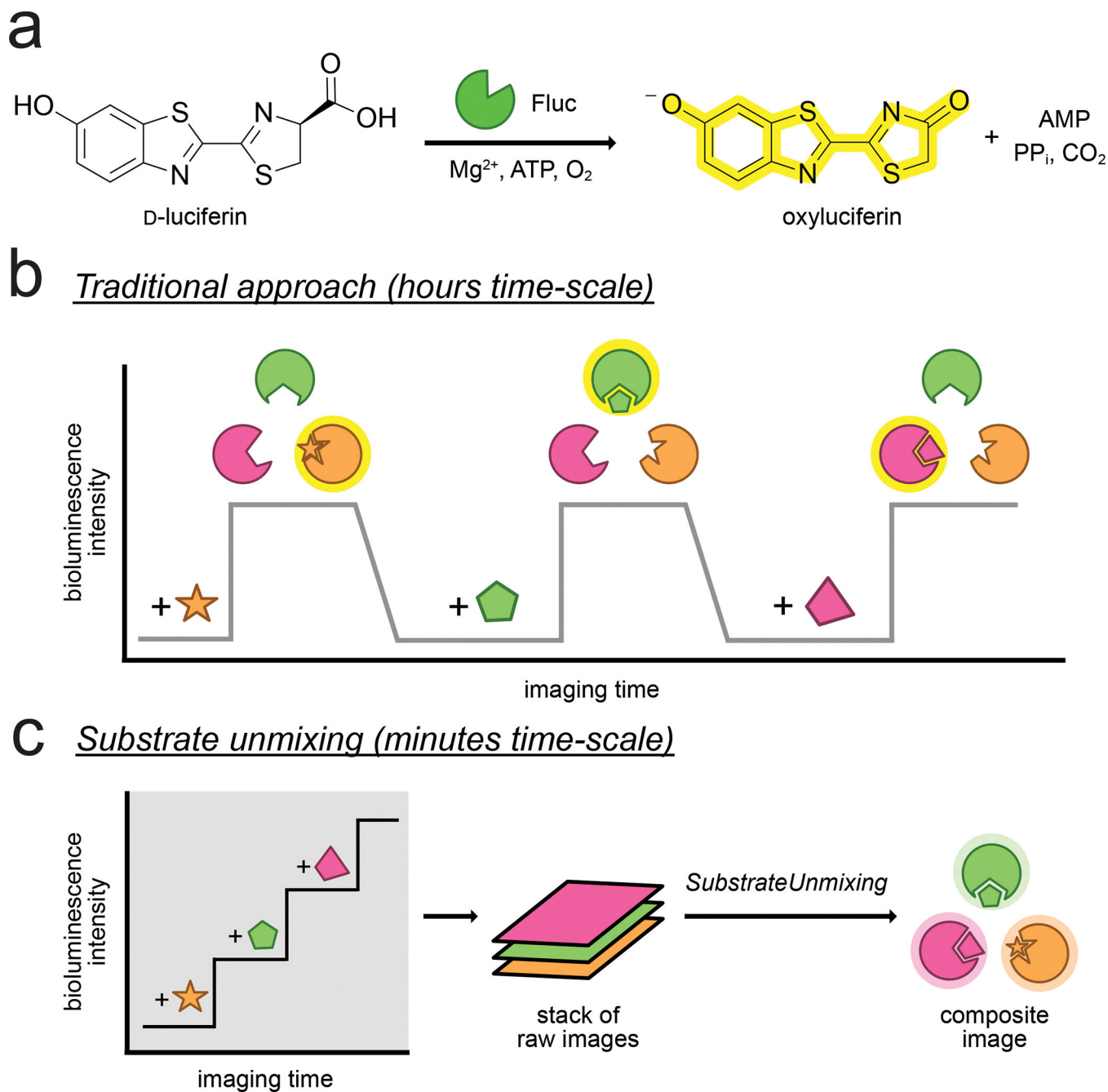


Figure 1. Rapid, multiplexed bioluminescence imaging via sequential substrate administration and serial acquisition.

(a) Optical imaging with bioluminescent probes. A photon of light is produced when D-luciferin (D-luc) is oxidized by firefly luciferase (Fluc). (b) Traditional approach for resolving multiple bioluminescent reporters. Signal from one luciferin must clear before addition of the next luciferin. The required imaging time scales with the number of probes, and can be impractical when more than three targets are involved. (c) Imaging times can be shortened by consecutive substrate application. The resulting images comprise multiple layers of photon output, and require an unmixing step to deconvolute the signal source.

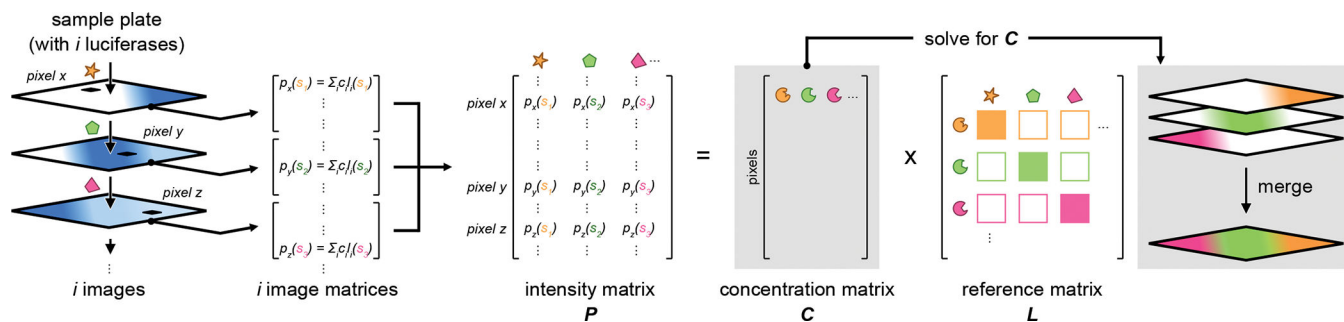


Figure 2. Derivation of the *SubstrateUnmixing* algorithm.

The collection of acquired images is first transformed to an array of matrices. Photon signal at each pixel is defined as the sum of photons produced from the given substrate with the plausible luciferase present. Next, the compiled image matrices are converted into an intensity matrix P . This matrix can be re-defined as a system of linear equations according to equation (4). Lastly, solving for the concentration matrix C affords a stack of unmixed images, representing the abundance of individual luciferases.

Cashew was 0.996. In the AkaLumine channel, the R^2 value for signal from Akaluc was 0.999. Error bars represent the standard error of the mean for $n = 2$ replicate experiments.

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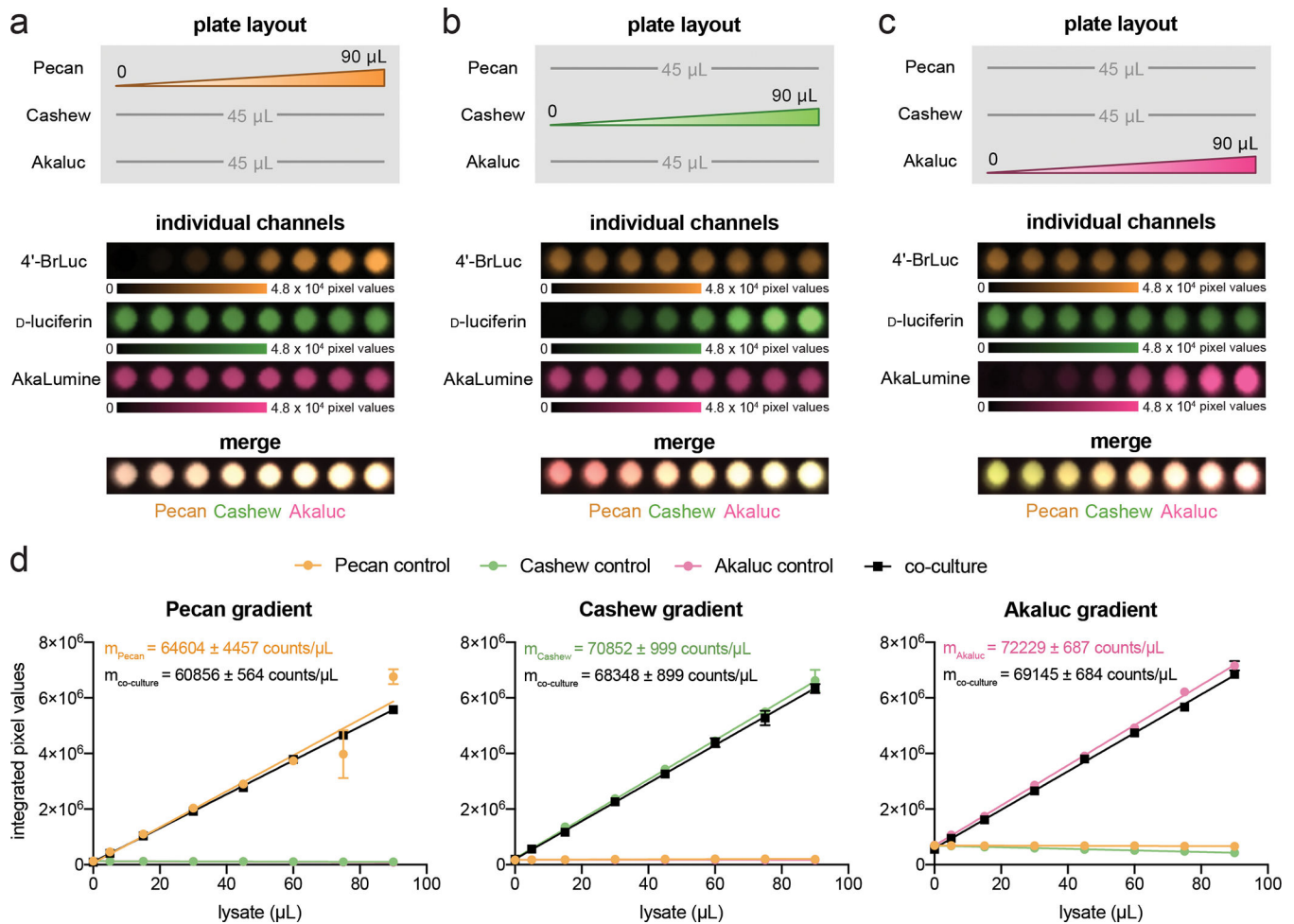


Figure 5. Analyzing changes in reporter ratio with *SubstrateUnmixing*.

(a) Pecan, (b) Cashew, and (c) Akaluc were plated in a gradient (as shown). The amount of one reporter was diluted to mimic a change in reporter expression over time. The other two reporters were kept constant. The samples were treated with 4'-BrLuc (100 μM), D-luc (100 μM), and AkaLumine (100 μM) in succession. Raw images were acquired after each substrate addition and processed by the algorithm. The substrate-specific signals were unmixed, assigned false colors, and overlaid. (d) Quantification of images from (a)–(c), processed via *SubstrateUnmixing* and fit via linear regression. Pecan, Cashew, and Akaluc controls represent samples that only contained a gradient of one reporter (not shown). In each scenario, the unmixed signals correlated linearly with the amount of reporter in the single population ($R^2 = 0.938$ for 4'-BrLuc channel, $R^2 = 0.997$ for D-luc channel, and $R^2 = 0.999$ for AkaLumine channel) and co-culture samples ($R^2 = 0.999$ for Pecan channel, $R^2 = 0.997$ for D-luc channel, $R^2 = 0.999$ for AkaLumine channel).

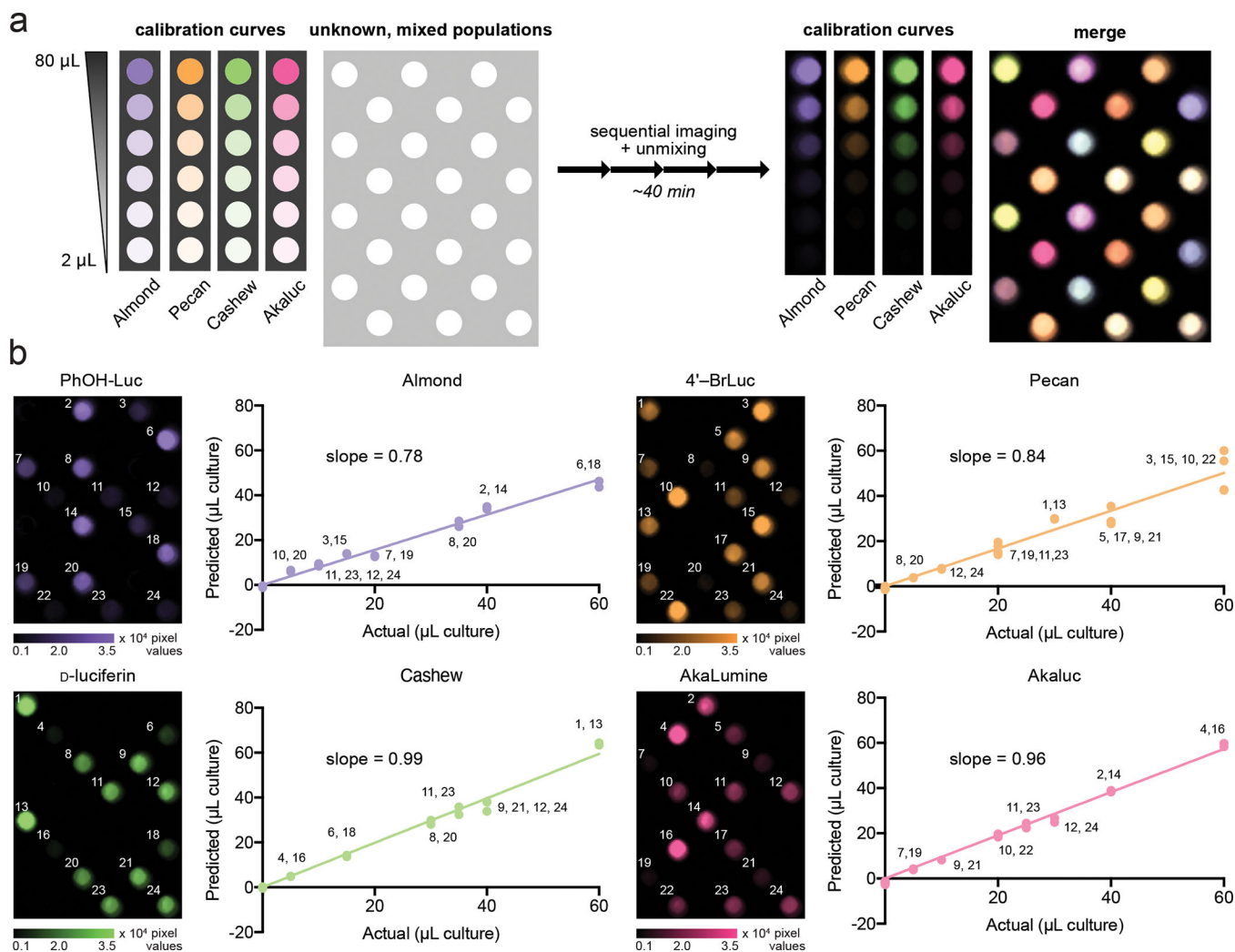


Figure 6. Multicomponent, quantitative BLI via *SubstrateUnmixing*.

(a) Different amounts of Almond, Pecan, Cashew, and Akaluc were mixed, and distributed across a 96-well plate as shown. A total of twelve unique mixtures, comprising two, three, or four reporters, were analyzed. A calibration curve for each luciferase was also plated on the same plate. The samples were treated sequentially with PhOH-Luc (250 μ M), 4'-BrLuc (100 μ M), D-luc (100 μ M), and AkaLumine (100 μ M). Raw images were acquired after each substrate addition, unmixed, and overlaid. (b) From each unmixed channel, a standard curve was computed using unmixed signal from the calibration wells. The amount of luciferase in the unknown wells was computed, and plotted against the actual amount of reporter plated. Values indicate the well number of the mixed population.

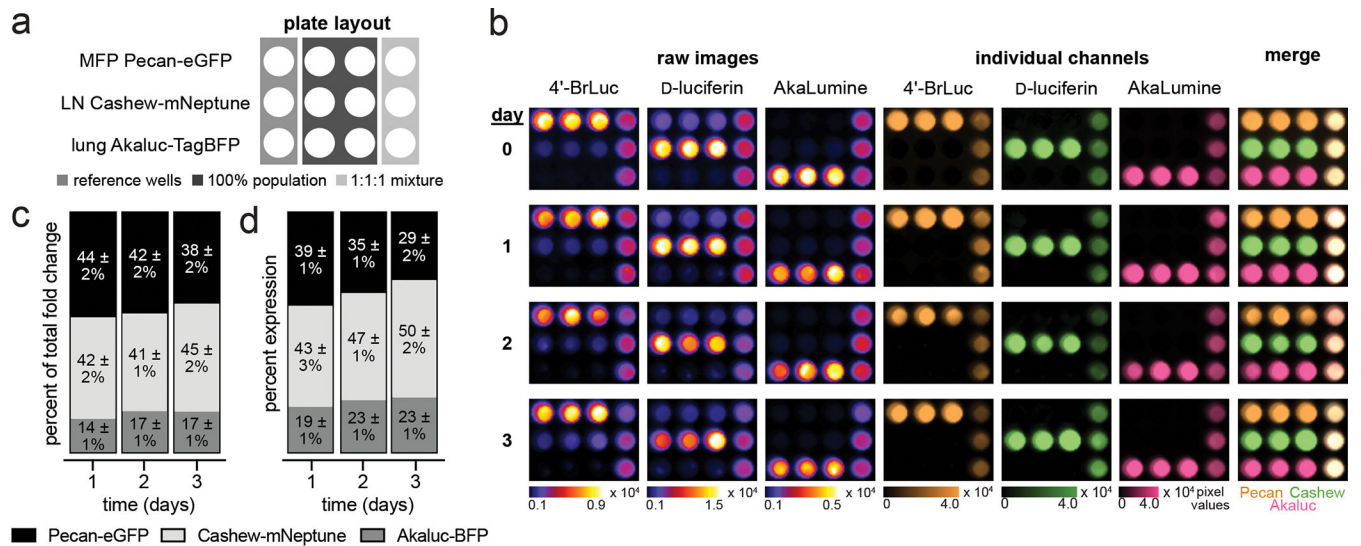


Figure 7. Heterogeneous cell populations can be monitored over time with *SubstrateUnmixing*. (a) MFP cells expressing Pecan-eGFP, LN cells expressing Cashew-mNeptune, and lung cells expressing Akaluc-BFP were plated as shown at the time of imaging. Cells were imaged on the day of seeding (day 0) and three subsequent days. (b) On each day the samples were treated with 4'-BrLuc (100 μ M), D-luc (100 μ M), and AkaLumine (100 μ M) in succession. Raw images were acquired after each substrate addition and processed by *SubstrateUnmixing*. The substrate-specific signals were unmixed, assigned false colors, and overlaid. (c) Quantification of luminescence values (p/s) plotted as the fold change versus day 0 & normalized to expression of fluorescent reporter. (d) Quantification of 1:1:1 mixture by flow cytometry normalized to expression of the single population.

KEY RESOURCE TABLE

REAGENT or RESOURCE	SOURCE	IDENTIFIER
Bacterial and virus strains		
BL21(DE3) Competent E. coli	New England BioLabs	C2527I
Chemicals, peptides, and recombinant proteins		
D-luciferin Firefly, potassium salt	Biosynth Carbosynth	Cat#L-8220
(S)-2-(4-Bromo-6-hydroxybenzo[d]thiazol-2-yl)-4,5-dihydrothiazole-4-carboxylic acid (4'-BrLuc)	(Steinhardt et al., 2017)	N/A
AkaLumine-HCl (TokeOni)	Aobious	AOB9983
(S)-2-(3-hydroxy-4-(6-hydroxybenzo[d]thiazol-2-yl)phenyl)-4,5-dihydrothiazole-4-carboxylic acid (PhOH-Luc)	(Yao et al., 2020)	N/A
Nano-Glo® Luciferase Assay System (furmazine)	Promega Corporation	Cat#N1120
Dulbecco's Modified Eagle's Medium (DMEM)	Corning	Cat#10-017-CV
Fetal Bovine Serum	Gibco	Cat#10082147
Penicillin-Streptomycin (10,000 U/mL)	Gibco	Cat#15140122
Puromycin Dihydrochloride	Gibco	Cat#A1113802
Trypsin-EDTA (0.05%), phenol red	Gibco	Cat#25300054
Glycerol	Fisher Scientific	Cat#S25342A
Kanamycin Monosulfate	GoldBio	Cat#K-120-5
LB Broth (Miller)	Genesee Scientific	Cat#11-121
Isopropyl-beta-D-thiogalactoside	GoldBio	Cat#I2481C
Tris Base	Fisher Scientific	Cat#BP152-500
NaCl	Fisher Scientific	Cat#S271-500
Tween® 20	Fisher Scientific	Cat#BP337-100
MgCl•H ₂ O	Fisher Scientific	Cat#M33-500
Adenosine-5'-Triphosphate (ATP)	GoldBio	Cat#A-081-25
Agarose	Fisher Scientific	Cat#BP160-500
Trypan Blue Stain (0.4%)	Gibco	Cat#15250061
Experimental models: Cell lines		
MMTV-PyMT MFP	(Ionkina et al., 2021)	N/A
MMTV-PyMT LN	(Ionkina et al., 2021)	N/A
MMTV-PyMT Lung	(Ionkina et al., 2021)	N/A
MMTV-PyMT MFP Pecan-eGFP	This paper	N/A
MMTV-PyMT LN Cashew-mNeptune	This paper	N/A
MMTV-PyMT Lung Akaluc-BFP	This paper	N/A
DB7 Pecan-eGFP	(Rathbun et al., 2021)	N/A
DB7 Cashew-mNeptune	(Rathbun et al., 2021)	N/A
DB7 Akaluc-BFP	(Rathbun et al., 2021)	N/A
Recombinant DNA		
pET28a Cashew	(Rathbun et al., 2017)	N/A

REAGENT or RESOURCE	SOURCE	IDENTIFIER
pET28a Pecan	(Rathbun et al., 2017)	N/A
pET28a Akaluc	(Yao et al., 2021a)	N/A
pET28a NanoLuc	(Yao et al., 2021a)	N/A
pET28a Almond	(Yao et al., 2021a)	N/A
AAVS1 Pecan-eFGP-T2A-Puro	(Rathbun et al., 2021)	N/A
AVVS1 Cashew-mNeptune-T2A-Puro	(Rathbun et al., 2021)	N/A
AAVS1 Akaluc-TagBFP-T2A-Puro	(Rathbun et al., 2021)	N/A
hCas9 (Addgene Plasmid #41815)	Gift from George Church	N/A
pSQT1313 (Addgene Plasmid #53370)	Gift from Keith Joung	N/A
Software and algorithms		
Living Image Analysis Software	Perkin Elmer	N/A
ChemDraw v21.0.0	Perkin Elmer	N/A
ImageJ (installed under the FIJI package)	NIH	N/A
NovoExpress Software	ACEA Biosciences	N/A
MATLAB R2019b	MathWorks	N/A
GraphPad Prism 9	GraphPad Prism Software, Inc.	N/A
<i>SubstrateUnmixing</i> pipeline	This paper	DOI: 10.5281/zenodo.7072808

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