

Advances in Ultrafast Optics and Imaging Applications

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ABSTRACT

Ultrafast imaging has been a key enabler to many novel imaging modalities, including looking behind corners and imaging behind scattering layers. With picosecond time resolution and unconventional sensing geometries, ultrafast imaging can fundamentally impact sensing capabilities in industrial and biomedical applications. This paper reviews the fundamentals, recent advances, and the future prospects of ultrafast imaging-based modalities.

Keywords: Ultrafast Imaging, Computational Imaging

1. INTRODUCTION

The use of ultrafast imaging has been a fundamental piece to many advances in various imaging applications, including looking behind corners,¹ imaging behind scattering layers²⁻⁴ and pose estimation.⁵ Critical components to these advances are emerging image sensors with picosecond (ps) time resolution, which enable accurate temporal information acquisition without the need for conventional interferometric geometries.⁶ Typical computational imaging techniques in traditional scene analysis exploit sensor parameters such as spatial resolution,⁷ angular sensitivity,⁸ wavelength, and polarization.⁹ However, these parameters alone are limited in their ability to capture the complex dynamics of light propagation. Ultrafast time-resolved sensors overcome this limitation and enable complicated analysis of light-in-flight in various imaging geometries.

One of the most notable applications of ultrafast optics is imaging beyond the conventional field of view of the camera. The sensor measures the time of flight (ToF) of indirect reflections from the hidden object, and a reconstruction algorithm is used to invert the measurements to recover the object. Based on the Plenoptic function,¹⁰ light transport theory¹¹ and the rendering equation,¹² it has been shown that time-resolved imaging is especially suited for these applications.¹³

To use ultrafast measurements for sensing beyond line-of-sight or conventional field of view, we usually need to solve a complex inverse problem. A key insight in solving the inverse problem is to treat light propagation between the scene and sensors in a five dimensional space¹⁴ comprising of space (2D), angle (2D), and time (1D). Thus, by combining forward models of light propagation and advanced signal processing and optimization techniques, we are able to invert the measurement and recover the hidden scene.

In this paper we demonstrate how ultrafast imaging has enabled simultaneous localization and identification of objects with temporal signatures hidden behind scattering layers. The paper is structured as follows: Section 2 serves as an introduction to ultrafast optical measurement techniques. Section 3 discusses non-line of sight imaging in cases of discrete scattering events. Section 4 discusses non-line of sight imaging in cases where the time dependency is a continuous function. Section 5 extends the discussion to the THz regime. Section 6 discusses novel imaging architectures, and section 7 provides an insight to future ultrafast sensors and their imaging applications.

2. IMAGING WITH ULTRAFAST OPTICS

A key component to ultrafast imaging is the ultrafast sensor. There is a broad range of sensors and sensor arrays that can be used for time-resolved imaging with temporal resolution as low as the ultrafast pulse cycle itself.¹⁵⁻¹⁷ Despite the large diversity of sensors for this application, electronically-triggered sensors, such as streak camera and single photon avalanche diode arrays (SPADs), are more common due to convenience in alignment and

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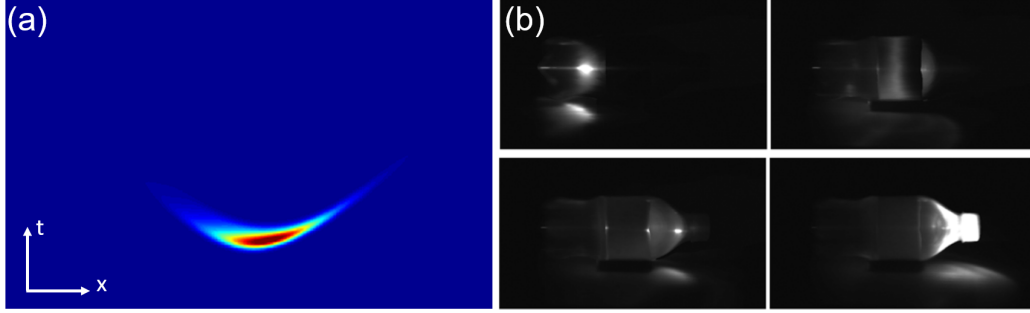


Figure 1. Measurement and result of ultrafast imaging. (a) Streak image, y-axis represents time, x-axis represents spatial coordinates. The target is a patch behind a thin diffuser. Each row represent a time window of $2ps$. (b) Four frames from an ultrafast measurement. The target is a pulse of light propagating through a bottle. Measurements were taken using a streak camera and mechanical moving mirrors (scanning through the spatial y-axis).

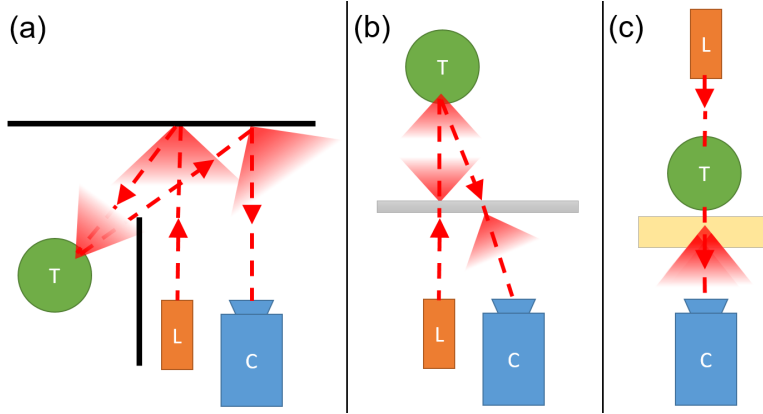


Figure 2. Acquisition geometries for non-line of sight imaging using ultrafast optics. ‘T’ represents the target. ‘C’ represents the camera (sensor), and ‘L’ represents the pulsed laser source. (a) Looking behind corner setup. Black lines are opaque diffusive walls. (b) Imaging through diffuser, reflection mode. Gray line is a thin diffusive sheet. (c) Imaging through diffuser, transmission mode. Yellow box is a volumetric diffuser.

acquisition. The use of electronic triggering through photodiode signal eliminates the need for precise optical delay lines and interferometric geometries.

Streak cameras are photocathode tubes that map the time axis onto the y-axis of a sensor (with $2ps$ time resolution). This is achieved by deflecting photoexcited electrons inside the photocathode tube. The streak image is thus an $x - t$ image (Fig. 1a). In order to acquire the full $x - y - t$ data cube, vertical scanning of the scene is needed. This can be done in a single shot by optical multiplexing^{18,19} or it can be done in periodic mode through mechanical scanning means. An example of an ultrafast scene captured with a streak camera and mechanical scanning of the y-axis is shown in Fig. 1b.²⁰ Full $x - y - t$ scanning is not always necessary. Depending on the application, illumination scanning may replace the vertical scanning. This is known as dual photography²¹ and will be demonstrated in multiple applications below.

Streak camera is especially proper for non-line of sight acquisition geometries as it provides nanosecond (ns) time window along with ps time resolution and $\sim 1K$ pixels for spatial resolution. This is not the case for some other electronically triggered ultrafast sensors such as ICCDs²²⁻²⁴ and SPADs. Another aspect to consider is spectral sensitivity. The majority of the above sensors is broadband; however, since they are mostly based on direct band gap photoexcitation in semiconductors, they lose their sensitivity in far IR and THz.²⁵ For these frequencies, nonlinear optoelectronic approaches are paving the way for ultrafast imaging.^{26,27}

Fig. 2 reviews three main geometries that use ultrafast measurement for non-line of sight imaging. As explained in the following sections, each geometry is better suited for imaging through a certain type of scattering

barriers. For example, reflective geometries allow imaging through discrete scattering barriers (Fig. 2a,b). Transmission geometries (Fig. 2c) are desired in case of volumetric scattering for improved signal-to-noise ratio (SNR).

3. IMAGING AFTER DISCRETE SCATTERING EVENTS

3.1 Looking Around Corners

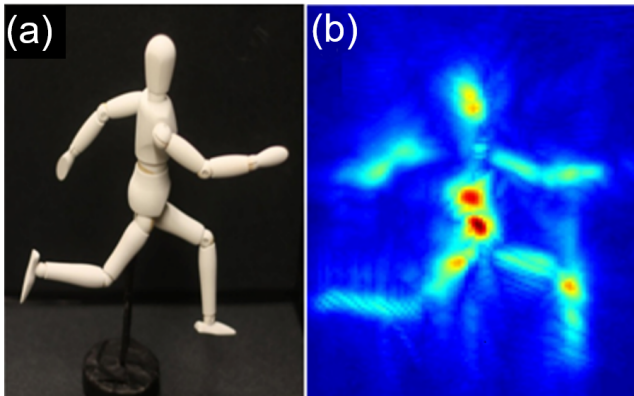


Figure 3. Recovered geometry of hidden object. a) A photograph of a hidden mannequin. b) The recovered geometry using ultrafast time-resolved measurement.

Consider an optical geometry of looking around a corner (Fig. 2a). Using ultrafast time-resolved measurement allows us to recover a mannequin (Fig. 3) hidden behind the corner.^{1,28} This is achieved by illuminating the first surface in front of the camera (a door) by a short laser pulse ($\sim 50fs$). The pulse bounces off the door and scatters in all directions. A small fraction of the light will travel into the room, scatter from the hidden object there and travel back, first to the door, and then to the camera. To increase the measurement diversity, it is repeated 60 times, each measurement taken with a different illumination position on the door (using the concept of dual photography²¹).

To reconstruct the hidden object using the ultrafast measurements, we first develop a mathematical model which describes the image formation. Consider a hidden patch in position x' , and illumination point on the wall x_l . The captured time-resolved measurement at sensor position x and time t will be:

$$I_l(x, t) = I_0 \int g(x_l, x, x') R(x') * \delta(t - c^{-1}(r_l(x') + r_c(x'))) dx' \quad (1)$$

where, I_0 is the source intensity, and $g(x_l, x, x')$ is a geometric factor which accounts for scene geometry. $R(x')$ is the reflectance of the patch. The $\delta(\cdot)$ function enforces the information cone defined by the speed of light c . Finally, r_l and r_c are the distances from the source to the patch and patch to camera, respectively. The goal is to recover the reflectance distribution $R(x')$ from the set of ultrafast measurements $\{I_l(x, t)\}_{l=1..60}$.

Using this forward model allows to scan through the target volume (x') and compare the expected measurement to the actual measurement. The amount of overlap provides a confidence measure to the existence of object in that location. Repeating this process on the target volume and for all illumination points results in the reconstruction in Fig. 3b.

3.2 Recovering Material Reflectances Behind Scattering Layer

Measuring the reflectance properties of materials—in the form of simple albedo or diffuse reflectivity (as discussed earlier), or complex Bidirectional Reflectance Distribution Functions—is useful for a variety of applications in Optics, Medical Imaging, and Computer Graphics. To demonstrate accurate recovery of these material properties behind scattering layer,^{2,29,30} we employ the reflection optical geometry (Fig. 2b).

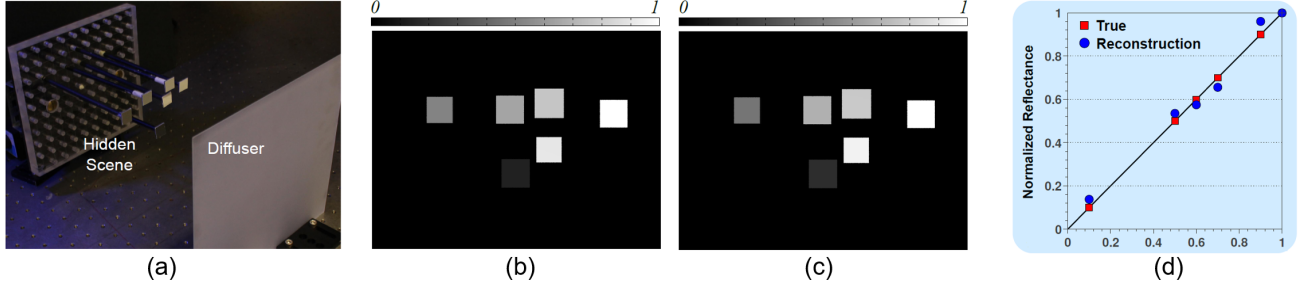


Figure 4. Recovering albedo behind diffuser. (a) Diffuser and hidden scene composed of multiple patches with different albedos. (b) Ground truth albedos. (c) Recovered albedos. (d) Quantitative comparison of recovery to ground truth.

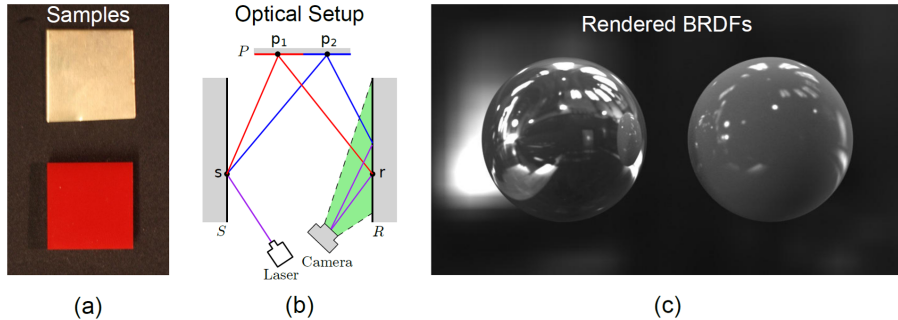


Figure 5. Acquiring parametric BRDFs. (a) Multiple material samples. (b) Indirect-imaging setup. (c) The recovered BRDFs match well with the data acquired with traditional methods.

Similarly to the “looking around corners” case, we acquire time-space streak images by focusing on a single line on the diffuser and illuminating it with pulsed laser on several locations. Using Eq. 1 we render synthetic streak images based on the scene geometry, where the only unknown is the albedo of scene points ($R(x')$). We solve a nonlinear optimization problem for scene point albedos, that minimizes the error norm between the real streak images and the rendered streak images. We are able to accurately measure the albedo of several scene points in complex scenes (Fig. 4).

The Bidirectional Reflectance Distribution Function (BRDF) is a four-dimensional function that characterizes the relationship between the reflected and incident radiance for opaque surfaces. BRDFs are primarily used in the field of computer graphics for photorealistic rendering, image relighting, and material editing, as well as for matching and material identification. Traditional techniques in graphics directly illuminate and image a small sample of the material from various angles, to acquire material BRDFs (see³¹ for a survey on acquisition techniques). Traditional BRDF acquisition methods are time consuming, need complex equipment encircling small material samples, and typically image only a single material at a time.

Ultrafast measurements enable us to tackle these challenges.³⁰ Unlike traditional techniques which rely on direct measurement of reflected light off the material surface, ToF measures all the bounces of light arriving at the diffuse surface, after interacting with material samples (Fig. 5). We acquire multiple streak images of indirect reflections from samples. The measurements and scene geometry are used in a linear system to solve for a low-dimensional parametric BRDF. We solve the linear system using unconstrained linear optimization, to recover BRDFs, which match well with BRDFs obtained with traditional methods, both in simulations and real experiments.

4. IMAGING AFTER CONTINUOUS SCATTERING

This section extends the previous imaging applications to cases in which there is significant time blur (due to fluorescence or volumetric scattering). Thus, the time dependency is not parameterized by a discrete function.

4.1 Recovering Fluorescence Lifetime Behind Scattering Layer

The ability to control and manipulate luminescent probes enables new capabilities in many imaging systems, such as high-resolution results in biological microscopy³² and anti-fraud measures or covert tracking.³³ These applications can benefit from fluorescence lifetime imaging (FLI) measurement. While FLI requires more complex hardware, it provides information on the environment of the probes.^{34,35} It also overcomes cases in which pure spectral signature is insufficient.³⁶ In particular, the extra information provided by FLI makes it attractive for imaging through complex media.

To demonstrate simultaneous recovery of location and fluorescence lifetime we consider again a reflection optical geometry (Fig. 2b).³⁷ The targets in this case are a set of three $1.5 \times 1.5 \text{cm}^2$ square patches hidden behind the diffuser. The first patch (NF) is non-fluorescent. The second patch (QD) is painted with a quantum dot solution ($\tau = 32 \text{ns}$, $\lambda_{\text{emission}} \sim 652 \text{nm}$). The third patch (PI) is painted with Pyranine ink ($\tau = 5 \text{ns}$, $\lambda_{\text{emission}} \sim 510 \text{nm}$).

In order to incorporate the fluorescence profile into our mathematical model, we assume a time-dependant reflectance profile with an exponential decay in Eq. 1:

$$R(x', t) = \rho(x')\tau^{-1}(x')e^{-t/\tau(x')}u(t) \tag{2}$$

where $\tau(x')$ is the local fluorescence lifetime, and $u(t)$ is a unit step function, imposed to satisfy causality.

The main challenge in the reconstruction process is the coupling between the geometrical information (high frequency data encoded in space and time) and fluorescence profile (low frequency data encoded in time). While previously the time profile encoded just geometry (as seen in Fig. 6a), now the geometrical features are not readily observable since they are masked by the fluorescence profile (Fig. 6b). Ideally we want to recover locations and lifetimes simultaneously. However the problem is ill-posed and the search space is too large. In order to overcome this challenge we first aim to recover a coarse location map of possible locations, followed by a step to recover both locations and lifetimes simultaneously. The first step narrows the search space of the second step, thus making the entire process robust.³⁸ To computationally solve these two steps we assume the patches are sparse in space and use orthogonal matching pursuit.³⁹ To demonstrate our method we show results for three different configurations in Table 1. We are able to correctly classify all patches, and recover their locations.

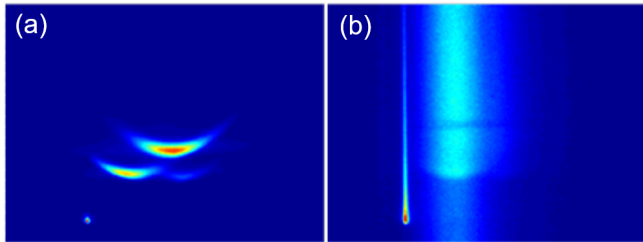


Figure 6. Fluorescing tags behind scattering layer. a) Streak measurement of the targets, due to strong direct reflection from the patches we observe only geometrical features and no fluorescence profile. b) Streak measurement taken with a UV filter to block the direct reflection reveals the fluorescence profile which obscures the pure geometrical data.

	Configuration 1				Configuration 2				Configuration 3			
Patch	ΔX	ΔY	ΔZ	$\Delta \tau$	ΔX	ΔY	ΔZ	$\Delta \tau$	ΔX	ΔY	ΔZ	$\Delta \tau$
QD	8.1	9.9	1.6	0	5.8	14.6	1.67	0	4.8	7.0	2.6	0
PI	5.7	10.9	1.6	0	2.4	14.3	2.13	0	6.9	16.7	1.7	0

Table 1. Reconstruction error; the numbers represent distances from the center of each ground truth patch in space to the center of the corresponding reconstructed patch. Length units are millimeters.

4.2 Imaging Through Volumetric Scattering

Another case of ultrafast imaging with significant blurring in time occurs when the signal goes through volumetric scattering. While previous examples looked into cases of discrete scattering events (in the fluorescence case it is a discrete event overlaid by a continuous function), here we consider a case of continuous scattering that occurs when light propagates through thick biological tissue. Ultrafast measurement allows to overcome these challenges by measuring the imaging system's point spread function (PSF) in space as well as in time.⁴

Here, the optical setup is transmission mode (Fig. 2c), where we use the rotating mirrors to capture a complete measurement of the $x - y - t$ space. The first step in the reconstruction process is to measure the system's PSF. This is achieved by placing a pin hole mask behind the thick diffuser. The following forward model is empirically fit to the measurement:

$$PSF(x, y, t) = \frac{1}{t} \exp\left(-\frac{(\ln t - \mu)^2}{2\sigma^2}\right) \exp\left(-\frac{x^2 + y^2}{4(D_0 + D_1 t)}\right) \quad (3)$$

where μ, σ, D_0, D_1 are the model parameters. Fig. 7. shows the measured PSF and the fitted model.

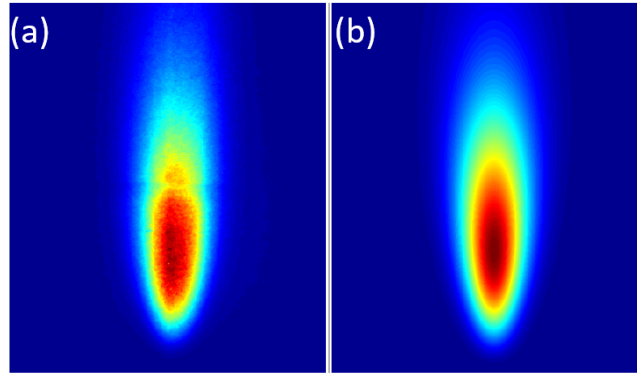


Figure 7. PSF estimation (a) Streak measurement of the PSF (showing a cross section for $y = 0$). (b) The corresponding cross section of our empirical PSF.

This forward model allows to cast the general problem of scene recovery as an optimization problem. The goal is to find a target which minimizes the difference between the $x - y - t$ measurement to a predicted measurement. We demonstrate our method and compare it to other techniques in Fig. 8.

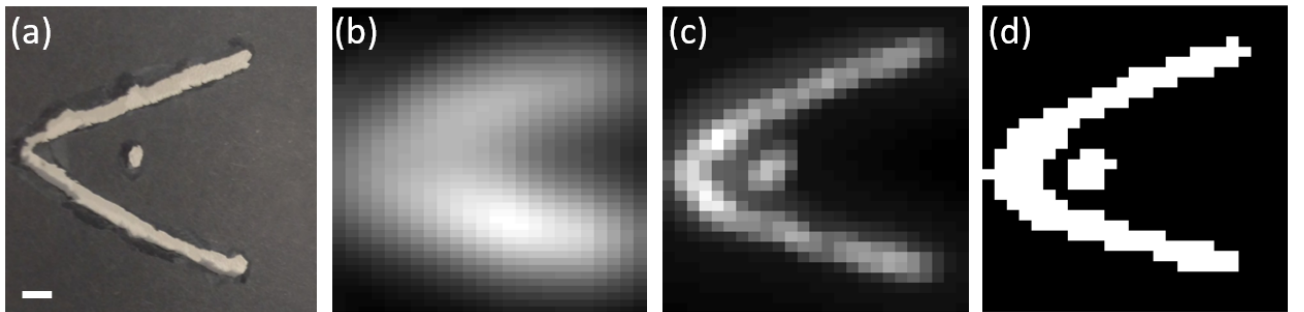


Figure 8. Imaging through volumetric scattering. (a) The mask hidden behind the diffuser (white scale bar: 4mm). (b) Result of imaging without using time-domain information. (c) Reconstruction using our algorithm. (d) Applying a threshold to generate a binary image from the reconstruction in panel (c).

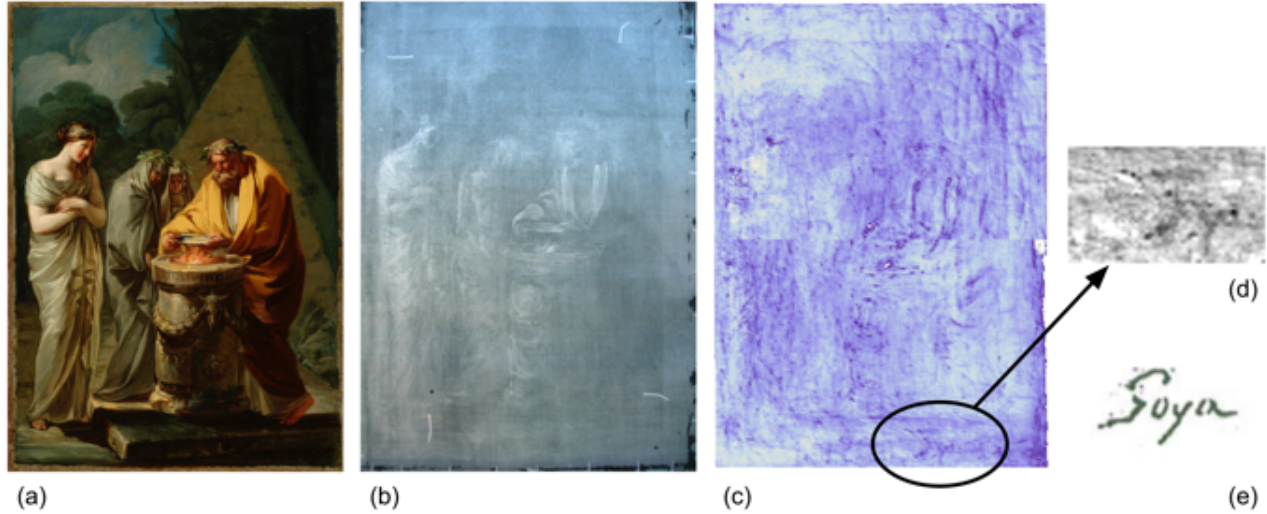


Figure 9. Results of inspecting Goya’s “Sacrifice to Vesta” with Terahertz. (a) Painting in the optical range. (b) X-ray image of the same painting. (c) Terahertz amplitude image of a deep layer. (d) Zoom in area with a feature that resembles the signature of the artist. (e) Registered signature of the artist.

5. IMAGING LAYERED STRUCTURES WITH PULSED TERAHERTZ WAVES

All imaging methods described in the paper thus far make use of visible and near IR wavelengths. We now describe the use of Terahertz (THz) range of the electromagnetic spectrum, which spans the frequencies from 0.1 THz to 10 THz.^{40,41} THz waves offer some unique features, such as the ability to penetrate dielectric materials and wavelengths short enough to resolve sub-mm spatial features (i.e., 1 THz is 300 μm). THz time-domain comprises the methods and techniques to generate and detect sub-picosecond pulses in time. The frequency components of such picosecond pulses extend into the THz range and, therefore, they are also referred to as THz pulses or THz waves. THz time-domain technology is attractive for non-destructive testing (NDT) applications,^{42,43} for example detecting structural defects in foam, wooden objects,⁴⁴ plastic components⁴⁵ and cultural artifacts.^{46–48}

One particular field that benefits from the properties of time-domain THz is cultural heritage. For example, paintings comprise of different layers made of different materials, which may have different content. Current methods (visible, infrared, ultraviolet, and X-ray) are not able to retrieve the content of deep layers; under-paints or other features remain blocked by the top layers. THz waves are more sensitive to chemical composition and offer sub-millimeter resolution to resolve these small details. THz ToF data of the entire depth section of the painting makes them suitable to analyze the different layers of a painting. However, many challenges arise due to the thickness of layers and gaps which are comparable to the wavelength of the THz pulse. For example, the SNR degrades very quickly as the number of layers increases, the contrast of the content in each layer is low and comparable to the contrast from inter-reflection noise, and, content from one layer occludes and causes shadowing effects in the signals coming from deeper layers. We tackle these challenges with computational approaches. The result allow us to retrieve the content of each different layer in the sample.

Fig. 9 shows the results of using THz to unravel the signature of master Goya in his early painting “Sacrifice to Vesta”, 1771 (Fig. 9a). The signature is not visible in the optical nor in the infrared or ultraviolet domain, since it is blocked by a thick layer of lacquer that becomes dark over time. X-ray image (Fig. 9b) shows areas with a high content of lead-based paint, and the nails and frame, but fails to catch subtle features. The THz image (Fig. 9c) is able to capture texture of brush strokes and other structural features that indicate stress in the canvas. However, the most interesting feature is captured in the lower right part of the painting. This feature (Fig. 9d) resembles the signature of the painter (Fig. 9e) and, thus, provides evidence of the authenticity of the piece.⁴⁹

6. NOVEL IMAGING ARCHITECTURES

While ultrafast imaging has been widely used to image through scattering barriers, new studies suggest that it can be used to change the imaging interface itself. Currently most of imaging systems are lens-based systems that are suitable for imaging through air or other transparent media. However, when imaging and sensing through complex harsh environments (e.g. porous media, diffusive liquids, high temperature and pressure, etc.) and complex geometries (geometries proper for endoscopy), fiber-based imaging can be a better imaging interface. Conventionally fiber-based image guides such as coherent fiber bundles have been used for such conditions. In longer length, however, (more than 20cm) coherent fiber bundles are rather rigid and fragile and they cannot provide wide field of view. We use ToF to enable imaging through randomly distributed and permuted sets of fibers.⁵⁰ This imaging interface (named optical brush) provides a brush-like flexible form factor.

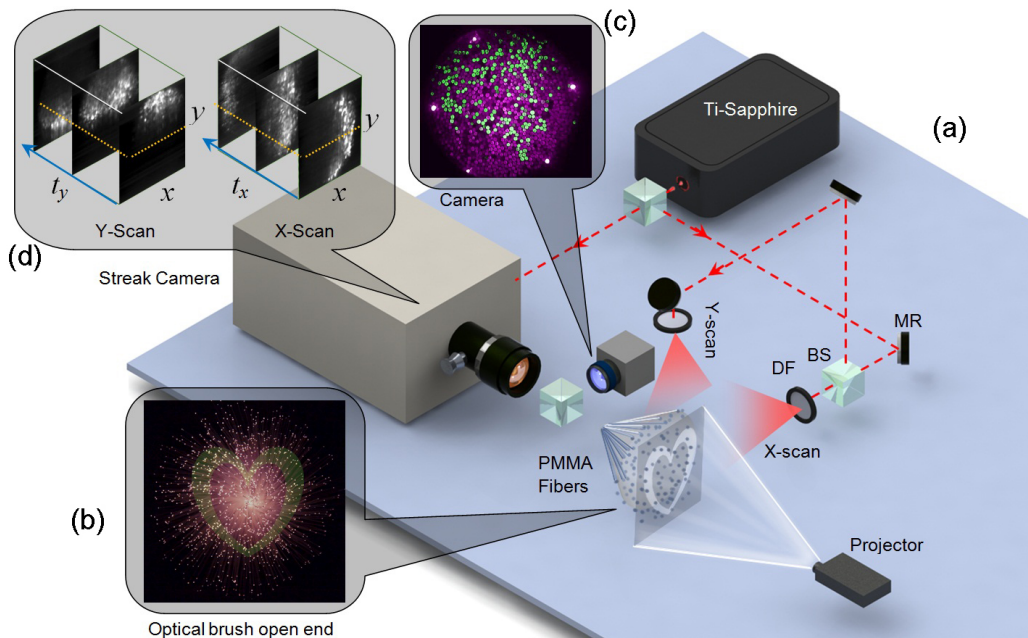


Figure 10. Setup for an optical brush enabled with ultrafast imaging. (a) Setup consists of a streak camera triggered by Ti-Sapphire pulses and an ordinary camera placed on the closed end of the optical brush. The other end is fed with a synthetic scene via a projector. BS stands for beam splitter and DF stands for diffuser. (b) Front view of the open-end of the optical brush with image of the heart projected on to the fibers. The infrared pulses are propagating in off-axis (perpendicular) with the plane that the fibers are distributed in. (c) Shuffled output of the brush seen by an ordinary camera. (d) Streak camera output. Each x-t slice is a streak image.

The optical brush uses a pulsed ToF technique for non-coaxial calibration of randomly distributed fibers to reconstruct the image of a scene for a secondary camera. Fig. 10a shows the setup for an optical brush: the fibers bristles are scanned with pulsed planar wavefronts. A projector is used to make a synthetic 2D scene for the brush (Fig. 10b). Since the fibers are randomly distributed in the 2D scene plane, the camera sees a lossy permuted or “shuffled” image of the scene as in Fig. 10c.

The streak measurements allow us to map between input and output positions of the fibers to reconstruct the target. This map is based on the time in which the pulse is received in the streak image for each fiber. This time correlates directly to the position of that fiber at the open-end. For instance, since all the fibers are equal in length, a fiber that outputs the pulsed signal later in time by the sweeping X-scan pulse (propagating from right to left) should be also positioned further away to the left.

Fig. 11a show an example of an input scene that is fed into the optical brush. As seen in Fig. 11b, the input is completely shuffled and some of the pixels are lost. Fig. 11c shows the deshuffled result by ToF technique. Fig. 11d shows ToF results superimposed on top of a lower resolution reference obtained by raster scanning

the fibers (cyan color). The ToF technique provides a 400×400 lateral resolution ($X - Y$) based on the time resolution of streak data cubes.

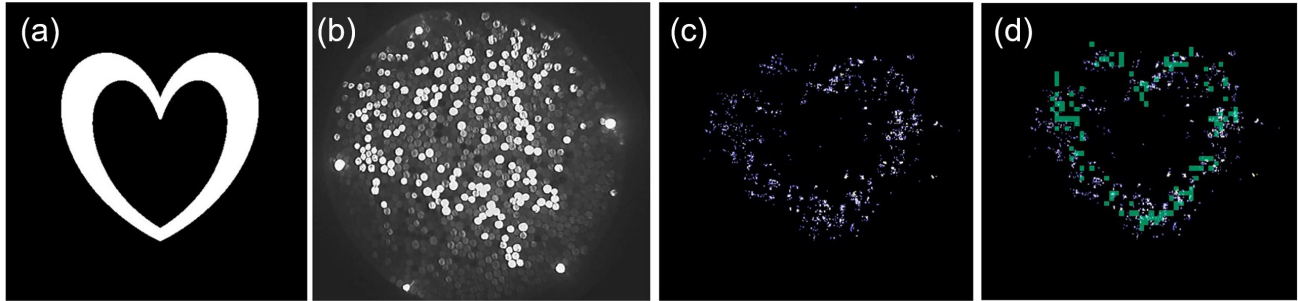


Figure 11. Ultrafast imaging used to reorganize pixels in imaging with and optical brush. (a) Input image. (b) Shuffled image that is output from the brush. (c) Deshuffled image based on off-axis ToF deshuffling technique. The spots are weighed based on the original intensity of the fibers in the streak data cubepoints with higher intensity represent fibers with brighter IR output from the sweeping pulses. (d) Deshuffled (400×400) image is superimposed on coaxial raster scan reference (60×80 resolution) for comparison.

Instead of using ToF to reveal 3D depth information, an optical brush uses this parameter to change the physical form of the imaging or sensing interface for a second camera. This enabling perspective on ToF parameter can be combined with emerging and preexisting ToF techniques such as continuous wave ToF,^{9,51,52} sequentially time all-optical mapping,⁵³ coherent interferometric depth imaging techniques^{54,55} and pulsed ToF methods.¹ Such change in physical form can affect acquisition capability with applications in biomedical imaging,⁵⁶ photophysics,¹⁹ and industrial sensing.¹ The ToF enabled optical brush, therefore, has significant potential for endoscopy, imaging in turbid media^{23,57,58} and near-field batch probing.⁵⁹

7. FUTURE SENSORS AND SOURCES

Conventionally, single-photon avalanche diodes (SPADs) have been one of the key players for applications that required high sensitivity. While streak cameras are making steady progress in their time resolution, with the introduction of Time-Correlated Single-Photon Counting (TCSPC) and improvement in circuit architecture, time-sensitive SPAD cameras are now entering the realm of ToF applications.⁶⁰ One of the most promising directions is incorporation of these cameras for biomedical imaging. However, while the sensitivity of SPADs promises imaging through thick pieces of living tissue, the time resolution and spatial resolution are still a burden. Fig. 12 demonstrates the advantage of single-photon sensitivity when encountering a thick tissue sample.

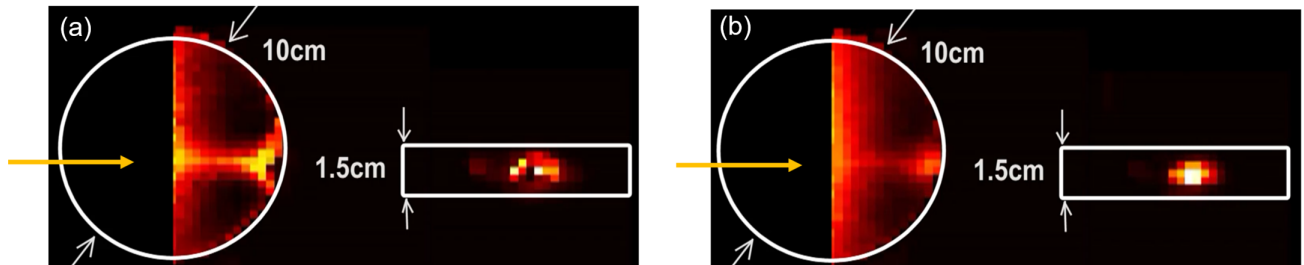


Figure 12. IR pulse ($20mW/cm^2$, $790nm$, $50fs$) passing through $10cm$ of tissue phantom disk. (a),(b) showing different time frames separated by $1.3ns$. Half of the phantom is covered to avoid camera saturation. Pixels at the center of the bright region are black due to saturation. The orange arrow shows the direction of pulse propagation. There is an exponential decay in the pulse intensity as it travels through the sample; this along with strong scattering at the phantom-air interface required us to block the left half of the phantom disk to investigate the other half.

As the sensitivity of pulsed-mode ToF sensors is increasing, the need for higher power laser sources is now decreasing, enabling lower cost longer pulse width lasers to be used in ultrafast imaging systems.⁶¹ On the other side of the spectrum as the electronics speed is increasing, continuous wave ToF cameras are starting to be applied in some of the ultrafast imaging applications such as high resolution depth sensing⁶² and fluorescent lifetime imaging.⁵¹ However, both continuous wave cameras and SPADs are fundamentally incapable of providing single shot ultrafast information as they depend on periodicity of acquisition. Therefore, unlike streak cameras these new technologies are not suitable for study of irreversible phenomena.^{18,19}

8. CONCLUSIONS

We presented how ultrafast detectors and sources enable non-line of sight imaging, which results in novel imaging applications. Different detection solutions provide a wide range of trade-offs, for example SPAD provides great sensitivity and a 2D array but doesn't allow a single shot acquisition and currently has limited time resolution. We also demonstrated non-traditional acquisition modality in the form of an optical brush as well as emerging opportunities in the THz spectrum. Various non-line of sight applications are enabled by ultrafast measurement, such as recovering material properties (albedo, fluorescence lifetime) behind scattering layers.

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