# Bidirectional Rendering of Vector Light Transport 

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#### Abstract

On the foundations of many rendering algorithms it is the symmetry between the path traversed by light and its adjoint path starting from the camera. However, several effects, including polarization or fluorescence, break that symmetry, and are defined only on the direction of light propagation. This reduces the applicability of bidirectional methods that exploit this symmetry for simulating effectively light transport. In this work, we focus on how to include these non-symmetric effects within a bidirectional rendering algorithm. We generalize the path integral to support the constraints imposed by non-symmetric light transport. Based on this theoretical framework, we propose modifications on two bidirectional methods, namely bidirectional path tracing and photon mapping, extending them to support polarization and fluorescence, in both steady and transient state.


Keywords: global illumination, rendering, ray tracing, rendering
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## 1. Introduction

For many years, simulating light transport has exploited the symmetry of most common scattering operators for building efficient methods for rendering. This symmetry involves that, independently on whether the scattering is computed from the propagation direction of the light, or from its adjoint (i.e. the direction from the camera), the throughput of the light transport would be exactly the same [Vea97].

Starting from ray tracing (as opposed to light tracing), these methods have taken advantage of this symmetry to compute only paths that contribute to the image. Bidirectional methods have gone a step further, seamlessly combining paths from the camera and the light, either by connecting vertices of both sub-paths [LW93, VG94], merging them via density estimation [Jen01] or a combination of both approaches [GKDS12, HPJ12], for robustly and efficiently handling most common light transport configurations.

Unfortunately, not all scattering operators can benefit from this symmetry. Effects such as polarization and fluorescence are defined with respect to the incoming radiance. Therefore, their adjoint (importance) cannot be modelled symmetrically, or even cannot be modelled at all, given its dependence on the incoming illumination. While in most common scenes, these effects are negligible, there are many examples where they might play a crucial role on the final
appearance: Rendering birefringent crystals [WW08, LSG12], interreflections between conductors and dielectrics, phosphorescent materials [Gla95] or scattering on turbid organic media [GSMA08] require modelling these non-symmetrical operators on light transport for accurate, predictive results.

Moreover, most of these effects exhibit a strong effect on the temporal domain. Therefore, while for traditional light transport, they might be important, they might become crucial when computing light transport in transient state. Including these effects is, however, non-trivial in bidirectional methods, since these techniques build upon the symmetry between radiance and importance.

In this work, we focus on developing a non-symmetric, but bidirectional rendering system, supporting effects such as polarization or fluorescence, which are dependent on the incoming illumination and therefore break the implicit symmetry of bidirectional methods. We first formalize this type of light transport by generalizing the well-known path integral formulation [Vea97] into what we call the vector light transport. This allows us to discuss the required modifications on bidirectional algorithms formulated within this framework, in particular bidirectional path tracing (BPT). We then show how these changes are applicable to photon mapping (PM). Finally, we extend this formulation to transient state, and show how under reasonable assumptions, transient rendering can be easily modelled within this theoretical framework.

This paper is an extension of our previous work on bidirectional rendering of polarization [JG16], in which we introduced a vector formulation of the path integral and use it to generalize BPT and PM to support polarization. This work was concurrent with the work of Mojzik et al. [MSWK16], which had similar aim and proposed a similar theory. Here, we reformulate the vector path integral to support a wider number of effects, which we demonstrate by rendering fluorescence, and define it in transient state following the work of Jarabo et al. [JMM*14].

## 2. Background and Related Work

Light transport simulation One of the core goals of computer graphics is to synthesize photorealistic images, by accurately simulating how light interacts with matter. While several methods have been introduced to that aim, bidirectional techniques have stood out as the most robust techniques. These methods compute light transport by tracing random walks from the light and camera, which are later connected by means of deterministic connection [LW93, VG94], density estimation [Jen01, JNSJ11] or combinations of both [HPJ12, GKDS12, KGH*14]. These works are formulated to render scalar radiance, and do not support effects such as polarization or elastic scattering.

Polarization Polarized light can be represented using the socalled Stokes vectors. These were introduced in graphics by Sankararayanan [San97], and were adopted by other authors for efficient and practical polarization rendering [FGH99, WTP01, WW12]. Stokes vectors model polarized light (for each wavelength $\lambda)$ as a 4 -vector $S_{\lambda}=\left\langle S_{0}, S_{1}, S_{2}, S_{3}\right\rangle$ defined as [WTP01]:

$$
\begin{align*}
& S_{0}=I_{s}^{2}+I_{p}^{2} \\
& S_{1}=I_{s}^{2}-I_{p}^{2} \\
& S_{2}=2 I_{s}^{2} I_{p}^{2} \cos \left(\phi_{s}-\phi_{p}\right) \\
& S_{3}=2 I_{s}^{2} I_{p}^{2} \sin \left(\phi_{s}-\phi_{p}\right) \tag{1}
\end{align*}
$$

where $I_{s}$ and $I_{p}$ are the intensity at the parallel and perpendicular planes of the electromagnetic wave, and $\phi_{s}$ and $\phi_{p}$ are their respective phases. Each component of $S_{\lambda}$ represents a type of polarization: $S_{0}$ describes the total intensity of the light wave. The second and third component $S_{1}$ and $S_{2}$ model the linear polarization at zero and 45 degrees, respectively. Finally, $S_{3}$ represents the circular polarization. The components in the Stokes vector must hold that $S_{0} \geq \sqrt{S_{1}^{2}+S_{2}^{2}+S_{3}^{2}}$, imposing that $S_{j} \in\left[-S_{0}, S_{0}\right]$ for $j=1$..3. Note that these components are defined on a particular reference frame aligned with the propagation direction. This formulation is compact and allows to represent all polarization states, and explicitly encodes the intensity of the light wave typically used in rendering, which allows an easier integration into current rendering systems. It is important to note that since Stokes vectors represent polarized light, they are only additive under the condition of lying on the same reference frame. As we will see later (Section 3), this has an important implication when integrating polarized light in the pixel.

The interaction between polarized light and matter is modelled with a matrix structure called Müller matrix, which encodes the
effect of the Bidirectional Scattering Distribution Function (BSDF) or the phase function as a $4 \times 4$ matrix, representing the linear transformation occurring to the polarized light. The Müller matrix is defined in a particular incoming and outgoing reference frame, which needs to be aligned with the respective light frames. This matrix form is defined for light paths, and therefore not for its adjoint (importance, more common in rendering).

Previous work has solved this by implementing Müller calculus on single-directional methods, such as recursive path tracing, or light tracing, where it is relatively easy to incorporate each of them [WW12], since there are not multiple cases to consider. Here, we focus on extending the applicability of fully bidirectional methods for polarization-based rendering. In the following section, we go deeper on that.

Polarization is important in a number of effects, including Fresnel-based specular reflection, glowing specular surfaces [WW11] such as reflecting (i.e. no black bodies) incandescent objects, micro-facet-based metal [BWWM12] and layered dielectrics surfaces [WWHN17], light transport in the atmosphere [WZP04] and rainbows [SML*12], or uniaxial [GS04, Hac07, WW08] and biaxial [LSG12, DK13] refringent crystals and gems. Additional details on polarized light transport can be found in optics literature [BW02] or in the excellent tutorial by Wilkie and Weidlich [WW12].

Fluorescence and phosphorescence Polarization is not the only effect that breaks symmetry on light transport: quantum effects such as fluorescence and phosphorescence are also defined as a function of the incoming light, and therefore they cannot be modelled based on their adjoint. These phenomena are the result of a change of its quantum state due to absorption of light. This change is not permanent, and after some time, it returns to its initial state, resulting into re-emission of light, in general with lower energy (i.e. light is re-emitted red-shifted). In particular, Glassner [Gla95] proposed a model for these two effects, based on re-radiation functions. Gutierrez et al. [GSMA08] focused on fluorescence, including its effect as part of the Russian roulette-based termination on PM and a re-radiation term on the density estimation pass. Hullin and colleages [HHA*10] presented a method to capture bi-spectral re-radiation matrices, which can be used for rendering fluorescent materials. Recently, Nalbach et al. [NSR17] introduced a pipeline for acquiring and reproducing phosphorescent materials, including their temporal response. Finally, closer to us, Wilkie et al. [WTP01] used bi-spectral re-radiation BRDFs within a forward path tracing, but do not describe how to extend this work to bidirectional methods.

Transient rendering The emergence of transient imaging [JMMG17] have brought an increased interest on simulating transient light transport in graphics and vision [Jar12, PBSC14, ADY*17]. Jarabo et al. [JMM*14] presented a path-integralbased framework for rendering transient light transport. Ament et al. [ABW14] demonstrated time-resolved light transport based on the refractive radiative transfer equation. Recently, Marco et al. [MJGJ17] proposed an extension of photon beams for transient rendering in participating media. We show time-resolved light transport in the context of the vector path integral, and show how transient rendering can be modelled as a vector-matrix operation.
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## 3. Vector Path Integral

Here, we describe a generalization of the path integral for including polarization. We use the term vector as an analogy of the vector radiative transfer equation used to model polarized radiative transfer [Cha60]. The path integral [Vea97] is a theoretic framework where the pixel intensity $I$ is computed as an integral over the space of light transport paths $\Omega$ contributing in the pixel:

$$
\begin{equation*}
I=\int_{\Omega} f(\overline{\mathbf{x}}) \mathrm{d} \mu(\overline{\mathbf{x}}) \tag{2}
\end{equation*}
$$

where $\overline{\mathbf{x}}=\mathbf{x}_{0} \ldots \mathbf{x}_{k}$ are the $k+1$ vertices of a length- $k$ path with $k \geq 1$ segments. Vertices $\mathbf{x}_{0}$ and $\mathbf{x}_{k}$ lie on a light source and camera sensor, respectively, while $\mathbf{x}_{1} \ldots \mathbf{x}_{k-1}$ are intermediate scattering vertices. The differential measure $\mathrm{d} \mu(\overline{\mathbf{x}})$ denotes area integration. The path contribution function $f(\overline{\mathbf{x}})$ is the product of the emitted radiance $L_{\mathrm{e}}$, path throughput $\mathfrak{T}$ and sensor importance $W_{\mathrm{e}}$ :

$$
\begin{equation*}
f(\overline{\mathbf{x}})=L_{\mathrm{e}}\left(\mathbf{x}_{0} \rightarrow \mathbf{x}_{1}\right) \mathfrak{T}(\overline{\mathbf{x}}) W_{\mathrm{e}}\left(\mathbf{x}_{k-1} \rightarrow \mathbf{x}_{k}\right) \tag{3}
\end{equation*}
$$

The path throughput is itself the product of the scattering function $\rho$ for the inner path vertices and the geometry $G$ and visibility $V$ terms for path segments:

$$
\begin{equation*}
\mathfrak{T}(\overline{\mathbf{x}})=\left[\prod_{i=1}^{k-1} \rho\left(\mathbf{x}_{i}\right)\right]\left[\prod_{i=0}^{k-1} G\left(\mathbf{x}_{i}, \mathbf{x}_{i+1}\right) V\left(\mathbf{x}_{i}, \mathbf{x}_{i+1}\right)\right] . \tag{4}
\end{equation*}
$$

We assume a fractional visibility to account for transmittance within media, as well as opaque objects. The scattering kernel at each vertex is defined as

$$
\rho\left(\mathbf{x}_{i}\right)= \begin{cases}\rho_{\mathrm{s}}\left(\mathbf{x}_{i-1} \rightarrow \mathbf{x}_{i} \rightarrow \mathbf{x}_{i+1}\right) & \mathbf{x}_{i} \text { on surface }  \tag{5}\\ \rho_{\mathrm{p}}\left(\mathbf{x}_{i-1} \rightarrow \mathbf{x}_{i} \rightarrow \mathbf{x}_{i+1}\right) \sigma_{s}\left(\mathbf{x}_{i}\right) & \mathbf{x}_{i} \text { in medium }\end{cases}
$$

where $\sigma_{s}$ is the scattering coefficient in the medium, and $\rho_{\mathrm{s}}$ and $\rho_{\mathrm{p}}$ are the surface BSDF and phase function, respectively.

Given that, in general, there is no analytic solution for Equation (2), Monte Carlo solutions are used to approximate the path integral as:

$$
\begin{equation*}
\langle I\rangle=\frac{1}{n} \sum_{j=1}^{n} \frac{f\left(\overline{\mathbf{x}}_{j}\right)}{p\left(\overline{\mathbf{x}}_{j}\right)} \tag{6}
\end{equation*}
$$

which averages the contribution of $n$ random paths $\overline{\mathbf{x}}_{j}$, sampled with a probability distribution function (pdf) $p\left(\overline{\mathbf{x}}_{j}\right)=p\left(\mathbf{x}_{0} \ldots \mathbf{x}_{k}\right)$ the combined probability density of each path's vertex. The probability density of the path is determined by the sampling technique used to obtain the path: for example, $B P T$ ([LW93, VG94] independently generates a sub-path $\overline{\mathbf{x}}_{w}$ from the eye with pdf $p\left(\overline{\mathbf{x}}_{w}\right)$ and a subpath $\overline{\mathbf{x}}_{l}$ from the light with pdf $p\left(\overline{\mathbf{x}}_{l}\right)$. These are then (optionally) connected using a shadow ray to build the full path $\overline{\mathbf{x}}$ with pdf $p(\overline{\mathbf{x}})=p\left(\overline{\mathbf{x}}_{l}\right) p\left(\overline{\mathbf{x}}_{w}\right)$ (see Figure 1, top).

Vector path integral The vector path integral takes a similar form as Equation (2), with a core difference on the definition of the signal


Figure 1: Schematic description of bidirectional path tracing (BPT, top) and photon mapping (PM, bottom). In both algorithms, a sensor and a light sub-path are traced from the eye and the light sources, respectively; these two sub-paths are then connected to form a full path, via deterministic shadow connection in the case of BPT, and via an additional random segment and density estimation in PM (figure after Georgiev et al. [GKDS12]).
being integrated. While Equation (2) integrates a scalar value (i.e. $I \in \mathbb{R}$ ), the vector form of the path integral integrates a vector $\mathbf{i}$ as:

$$
\begin{equation*}
\mathbf{i}=\int_{\Omega} \mathbf{f}(\overline{\mathbf{x}}) \mathrm{d} \mu(\overline{\mathbf{x}}), \tag{7}
\end{equation*}
$$

with $\mathbf{i}$ and the vector form of the path contribution function $\mathbf{f}(\overline{\mathbf{x}})$ defined by a multi-dimensional vector in $\mathbb{R}^{N}$. As an example $\mathbf{f}(\overline{\mathbf{x}})$, let us consider a polarized, monochrome light. In this case, $\mathbf{f}(\overline{\mathbf{x}})$ would be a Stokes vector defined in $\mathbb{R}^{4}$.

In its vector form, the scattering kernel at $\mathbf{x}_{i}$ is no longer a scalar term $\rho\left(\mathbf{x}_{i}\right) \in \mathbb{R}$, but a high-dimensional matrix $\mathbf{K} \in \mathbb{R}^{N \times N}$ modelling the relationship between incoming and outgoing light vectors defined in $\mathbb{R}^{N}$. Using the same example of monochromatic polarized light, in this case, $\mathbf{K}$ would be the Müller matrix, defined in $\mathbb{R}^{4 \times 4}$. Working with scattering matrices $\mathbf{K}$ breaks transitivity, and therefore we need to define how the operations are concatenated: As opposed to the traditional path integral (Equation (2)), here the order on which the operations are concatenated is important. For that, let us define the concatenation operation as:

$$
\begin{equation*}
\prod_{i=1}^{k-1} \mathbf{K}\left(\mathbf{x}_{i}\right)=\mathbf{K}\left(\mathbf{x}_{k-1}\right) \mathbf{K}\left(\mathbf{x}_{k-2}\right) \ldots \mathbf{K}\left(\mathbf{x}_{2}\right) \mathbf{K}\left(\mathbf{x}_{1}\right) \tag{8}
\end{equation*}
$$

Note that for simplicity, we are using the terms 'vector' and 'matrix' very loosely here: in both cases, they might be high-dimensional tensors, and therefore the product operation in Equation (8) would be a tensor product.

Once we have defined the scattering concatenation operator, we can define the throughput of path $\overline{\mathbf{x}}$ in matrix form $\mathfrak{T}_{\mathbf{v}}(\overline{\mathbf{x}}) \in \mathbb{R}^{N \times N}$ as:

$$
\begin{equation*}
\mathfrak{T}_{\mathbf{v}}(\overline{\mathbf{x}})=\left[\prod_{i=1}^{k-1} \mathbf{K}\left(\mathbf{x}_{i}\right)\right]\left[\prod_{i=0}^{k-1} G\left(\mathbf{x}_{i}, \mathbf{x}_{i+1}\right) V\left(\mathbf{x}_{i}, \mathbf{x}_{i+1}\right)\right] \tag{9}
\end{equation*}
$$

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where the result of the second product is a scalar, and therefore does not require to be included in the vector form of the scattering kernels.

Finally, by applying Equation (9) to Equation (2) and defining the vector form of the emission $\mathbf{I}_{\mathrm{e}}\left(\mathbf{x}_{0} \rightarrow \mathbf{x}_{1}\right) \in \mathbb{R}^{N}$, we get the vector form of $\mathbf{f}(\overline{\mathbf{x}}) \in \mathbb{R}^{N}$ as:

$$
\begin{equation*}
\mathbf{f}(\overline{\mathbf{x}})=\mathbf{W}_{\mathrm{e}}\left(\mathbf{x}_{k-1} \rightarrow \mathbf{x}_{k}\right) \mathfrak{T}_{\mathbf{v}}(\overline{\mathbf{x}}) \mathbf{l}_{\mathrm{e}}\left(\mathbf{x}_{0} \rightarrow \mathbf{x}_{1}\right) \tag{10}
\end{equation*}
$$

where $\mathbf{W}_{\mathrm{e}}\left(\mathbf{x}_{k-1} \rightarrow \mathbf{x}_{k}\right) \in \mathbb{R}^{N \times N}$ is the matrix defining the sensor's importance. Similar to Equation (2), the vector version of the path integral defined in Equation (7) can be computed using a Monte Carlo estimator, with the difference of estimating a vector value $\langle\mathbf{i}\rangle \in \mathbb{R}^{N}$.

Transient vector path integral Similarly to the traditional path integral (Equation (2)), we can extend Equation (7) to support time-resolved light transport. Following the work of Jarabo et al. [JMM*14], we generalize Equations (7), (10) and (9) to transient state as:

$$
\begin{gather*}
\mathbf{i}=\int_{\Omega} \int_{\Delta T} \mathbf{f}(\overline{\mathbf{x}}, \overline{\Delta \mathbf{t}}) \mathrm{d} \mu(\overline{\Delta \mathbf{t}}) \mathrm{d} \mu(\overline{\mathbf{x}}),  \tag{11}\\
\mathbf{f}(\overline{\mathbf{x}}, \overline{\Delta \mathbf{t}})=\mathbf{W}_{\mathrm{e}}\left(\mathbf{x}_{k-1} \rightarrow \mathbf{x}_{k}, \Delta t_{k}\right) \mathfrak{T}_{\mathrm{v}}(\overline{\mathbf{x}}, \overline{\Delta \mathbf{t}}) \mathbf{l}_{\mathrm{e}}\left(\mathbf{x}_{0} \rightarrow \mathbf{x}_{1}, \Delta t_{0}\right),  \tag{12}\\
\mathfrak{T}_{\mathrm{v}}(\overline{\mathbf{x}}, \overline{\Delta \mathbf{t}})=\left[\prod_{i=1}^{k-1} \mathbf{K}\left(\mathbf{x}_{i}, \Delta t_{i}\right)\right]\left[\prod_{i=0}^{k-1} G\left(\mathbf{x}_{i}, \mathbf{x}_{i+1}\right) V\left(\mathbf{x}_{i}, \mathbf{x}_{i+1}\right)\right], \tag{13}
\end{gather*}
$$

where $\overline{\Delta \mathbf{t}}=\Delta t_{0} \ldots \Delta t_{k}$ is the sequence of time delays of path $\overline{\mathbf{x}}$, and $\mathrm{d} \mu(\overline{\Delta \mathbf{t}})$ is the integration on the temporal domain at each path vertex. Finally, the total temporal delay is

$$
\begin{equation*}
t_{i}^{-}=\sum_{j=0}^{i-1}\left(t\left(\mathbf{x}_{\mathbf{x}_{j}} \leftrightarrow \mathbf{x}_{\mathbf{x}_{j+1}}\right)+\Delta t_{j}\right) \tag{14}
\end{equation*}
$$

### 3.1. Defining vector and scattering matrices

Polarization Our original formulation of the vector path integral [JG16] was defined to support polarization. In this case, assuming a single wavelength, light is represented using a Stokes vector defined in $\mathbb{R}^{4}$. The scattering interactions $\mathbf{K}$, on the other hand, are modelled by a Müller matrix $\mathbf{M}\left(\mathbf{x}_{i}\right) \in \mathbb{R}^{4 \times 4}$. Note that Müller matrices need to be defined in valid reference systems. Therefore, we need to rotate the frames of the incoming and outgoing electromagnetic waves to match the frame on which $\mathbf{M}\left(\mathbf{x}_{i}\right)$ is defined such that the perpendicular plane of both frames lays in the plane defined by the incoming and outgoing directions. Therefore, the scattering kernel becomes $\mathbf{K}=\mathbf{R}\left(-\alpha_{o}\right) \mathbf{M}\left(\mathbf{x}_{i}\right) \mathbf{R}\left(\alpha_{i}\right)$, where $\mathbf{R}(\alpha)$ is the rotation matrix defined by an angle $\alpha$ defining the rotation along the ray direction, and $\alpha_{i}$ and $\alpha_{o}$ are the rotation angles for the incoming and outgoing frames, respectively.

Fluorescence and phosphorescence Both fluorescence and phosphorescence involve elastic scattering, which means that light incoming with a given wavelength $\lambda$ can be reflected with different
colour. More precisely, for an incoming light defined on the continuous spectrum $L_{\mathrm{i}}(\lambda)$, the outgoing light $L_{\mathrm{o}}(\lambda)$ is

$$
\begin{equation*}
L_{\mathrm{o}}(\lambda)=\int_{0}^{\infty} L_{\mathrm{o}}\left(\lambda^{\prime}\right) \rho\left(\lambda^{\prime} \rightarrow \lambda\right) \mathrm{d} \lambda^{\prime} \tag{15}
\end{equation*}
$$

where $\rho$ is the bi-spectral and bidirectional scattering function (note that we omit the directional dependence for simplicity). To model this effect, we therefore need to leave the monochromatic assumption. By discretizing Equation (15), we can pose $L_{0}(\lambda)$ and $L_{\mathrm{i}}(\lambda)$ as vectors $\mathbf{I}_{0} \in \mathbb{R}^{L}$ and $\mathbf{I}_{i} \in \mathbb{R}^{L}$, respectively, with $L$ the number of discretizations of the continuum $\lambda \in \mathbb{R}$. Following the same discretization, the bi-spectral scattering operator $\rho$ is also modelled as the matrix $\mathbf{K} \in \mathbb{R}^{L \times L}$, which relates incoming and outgoing wavelengths $\lambda$. The diagonal of $\mathbf{K}$ models the common inelastic scattering (i.e. no energy transfer between wavelengths), while off-diagonal terms model elastic scattering. Note that Equation (15) models local transport in a single vertex of the path. However, it is easy to see that we can concatenate scattering operators $\mathbf{K}$ following Equation (8), resulting on a path throughput $\mathfrak{T}_{\mathbf{v}}(\overline{\mathbf{x}}) \in \mathbb{R}^{L \times L}$, and a vector integral $\mathbf{i}(\lambda) \in \mathbb{R}^{L}$ defined as a function of $\lambda$.

Time-resolved light transport Equation (11) models theoretically the integration in transient state for vector light transport. However, the vector form defined by Equation (7) can also be used for transient rendering, eliminating the need of sampling $\mathrm{d} \mu(\overline{\Delta \mathbf{t}})$. The most important operations in the temporal domain in this context are the delays due to light propagation and scattering. Assuming that we can define the incoming and outgoing light as continuous function in the temporal domains $L_{\mathrm{i}}(t)$ and $L_{\mathrm{o}}(t)$, respectively, both operations can be modelled as convolutions on the temporal domain as:

$$
\begin{equation*}
L_{\mathrm{o}}(t)=\int_{0}^{t} T\left(t^{\prime}\right) L_{\mathrm{i}}\left(t-t^{\prime}\right) \mathrm{d} t^{\prime} \tag{16}
\end{equation*}
$$

where $T$ is the convolution kernel, which for propagation is a delta function centred at $t_{i}^{-}$(Equation (14)), and for scattering is the temporal smoothing function due to, e.g. microscopic multiple scattering [HHdD16]. Similar to fluorescence or phosphorescence, we can discretize $L_{\mathrm{o}}(t)$ and $L_{\mathrm{i}}(t)$ in the temporal domain, as $\mathbf{I}_{\mathrm{o}} \in \mathbb{R}^{T}$ and $\mathbf{l}_{\mathrm{i}} \in \mathbb{R}^{T}$, with $T$ the number of frames. With this formulation, we can then just easily compute both propagation and scattering delays as scattering matrices $\mathbf{K} \in \mathbb{R}^{T \times T}$, characterizing the outgoing temporal profile due to scattering and propagation of an incoming temporal profile. For example, a propagation delay of $t$ frames in an input temporal profile would be modelled by a transport matrix $\mathbf{K}$ defined as the identity matrix shifted to the right $t$ frames. Similar to Equation (15), Equation (16) models the effect locally, but it can be similarly integrated in Equation (8), resulting on a vector inte$\operatorname{gral} \mathbf{i}(\lambda) \in \mathbb{R}^{T}$. This means that we keep track of the full temporal profile for each path contribution. Interestingly, this novel formulation allows for an easy integration with other light transport effects, without the need of temporal sampling the scattering delays. Note that, as opposed to the previous phenomena, the propagation delays are bidirectional. This symmetry can be broken in the scattering delays.

Combining different domains In order to combine different domains, such as wavelength and Stokes parameters, we only need
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to elevate the dimensionality of both the vector light and the matricial operators, while taking into account the operation order of each dimensions on the high-dimensional tensors. For example, for coloured polarized light, our vector integral $\mathbf{i}$ would have dimensionality $\mathbf{i} \in \mathbb{R}^{4 \times L}$, resulting in a Stokes vector for each colour sample. In order to handle both polarization and reradiation, we would need four-dimensional scattering tensors $\mathbf{K} \in \mathbb{R}^{4^{2} \times L^{2}}$ (i.e. each component of the Müller matrix would be a reradiation matrix itself). This results into high-dimensional throughput tensors $\mathfrak{T}_{\mathbf{v}}(\overline{\mathbf{x}}) \in \mathbb{R}^{4^{2} \times L^{2}}$, and might result into costly computations. However, as we discussed later, the cost can be severely alleviated by doing the expensive tensor computations only when strictly required.

## 4. Bidirectional Rendering of Vector Light Transport

In Section 3, we have described the mathematical framework within which we will work, making explicit the differences between traditional scalar rendering and the novel vector formulation. Here, we describe the algorithmic and implementation details for developing a bidirectional rendering within this framework.

Bidirectional methods [VG94, Jen01, GKDS12, HPJ12] compute the path integral by sampling several light paths joining the light and the sensor. This is done by generating two different random walks (sub-paths), each starting from the initial and final vertices of the path. These are then joined by means of a deterministic shadow connection, creating a full contributing path. These two random walks have different probabilities, depending on the sampling strategy used to create the sub-path.

When extending these methods to our framework, we detect that the random walk from the light source (the light sub-path $\overline{\mathbf{x}}_{l}$ ) fits naturally in the vector path integral, since it follows the sequence of events occurring to light since it is emitted. For each new scattering event in point $\mathbf{x}_{i}$, we would compute its scattering kernel $\mathbf{K}\left(\mathbf{x}_{i}\right)$, and apply it to the accumulated throughput of the path following Equation (8). Generating the random walk from the sensor, as well as performing the shadow connection, is a bit trickier. The key difference between them is that, as discussed by Mojzik et al. [MSWK16], while the light sub-path is defined by a light vector, the sensor subpath is defined by an importance matrix. In the following, we explain them on more details.

Sensor sub-path When computing the sensor sub-path $\overline{\mathbf{x}}_{w}$, we need to take into account that we are starting the sequence of events on the reverse order. Thus, we are not tracking vector magnitudes, but matricial operators. This is key, since it affects on how the scattering kernels at each vertex of the sub-path are defined. The main difficulty is to keep track on whether the incoming or outgoing directions follow the light direction ( $\omega_{\mathrm{i}}$ and $\omega_{0}$, respectively), or its adjoint (which we denote by $\hat{\omega}_{\mathrm{i}}$ and $\hat{\omega}_{0}$ ). For each new scattering event in the random walk, we sample the sub-path new direction $\hat{\omega}_{0}$ based on the previous direction $\hat{\omega}_{\mathrm{i}}$. We use the same sampling routine for light and importance tracing, based on intensity in the case of polarization, and in the diagonal of the scattering matrix in the case of fluorescence. The reason is that it allows to sample intensity in the former case, and that re-emission is practically Lambertian [HHA*10], so most of the directionality is kept in the diagonal of the re-emission matrix. Then, we create the scattering
kernel $\mathbf{K}\left(\mathbf{x}_{i}\right)$ in the frame defined by the light incoming and outgoing directions, $\omega_{\mathrm{i}}=-\hat{\omega}_{\mathrm{o}}$ and $\omega_{\mathrm{o}}=-\hat{\omega}_{\mathrm{i}}$, respectively.

With that in place, and taking into account that for the sensor's sub-path, we decrement the indices of the sub-path vertices (i.e. the sub-path vertices are generated in the reverse order, starting by vertex $k$ to vertex 0 ), we then compute the sub-path throughput using Equation (8). Therefore, as opposed to multiplying $\mathbf{K}\left(\mathbf{x}_{i}\right)$ to the left to the accumulated throughput as in the light random walk, we need to apply each new scattering kernel on the right.

Shadow connection In order to join the light and sensor sub-paths, we again need to be careful on the reference frame of the scattering operator, and on the order at which the events are computed. In this case, for a light sub-path $\overline{\mathbf{x}}_{l}$ with length $m$ and partial throughput $\mathfrak{T}_{\mathrm{v}}\left(\overline{\mathbf{x}}_{l}\right)$, and sensor sub-path $\overline{\mathbf{x}}_{w}$ with length $k-m$ and partial throughput $\mathfrak{T}_{\mathbf{v}}\left(\overline{\mathbf{x}}_{w}\right)$, we obtain the final throughput as
$\mathfrak{T}_{\mathbf{v}}(\overline{\mathbf{x}})=\mathfrak{T}_{\mathbf{v}}\left(\overline{\mathbf{x}}_{w}\right) \mathbf{K}\left(\mathbf{x}_{k-m}\right) G\left(\mathbf{x}_{k-m}, \mathbf{x}_{m}\right) V\left(\mathbf{x}_{k-m}, \mathbf{x}_{m}\right) \mathbf{K}\left(\mathbf{x}_{m}\right) \mathfrak{T}_{\mathbf{v}}\left(\overline{\mathbf{x}}_{l}\right)$.

As discussed earlier, note that $\mathfrak{T}_{\mathbf{v}}\left(\overline{\mathbf{x}}_{w}\right) \in \mathbb{R}^{N \times N}$ defines matricial importance, while $\mathfrak{T}_{\mathrm{v}}\left(\overline{\mathbf{x}}_{l}\right) \in \mathbb{R}^{N}$ represents vectorial light.

### 4.1. Photon mapping

As shown by Georgiev et al. [GKDS12] and Hachisuka et al. [HPJ12], PM [Jen01] can be understood as a variant of BPT, which differs from the standard formulation on how the sensor and light sub-paths are connected. While in BPT, we connect the last two vertices by means of a deterministic shadow connection, and in PM, we merge the last two vertices by using a density estimation kernel. While this introduces bias, it has been shown that in the limit, the algorithm is consistent (i.e. converging to the correct solution, we refer to a recent course [HJG*13] for details).

This means that we can define PM under the path integral framework, with some small modifications. Therefore, we can introduce PM in our vector formulation of the path integral. In fact, the main difference with respect to BDPT is how the eye and light sub-paths are joint. While for BPT, we make use of Equation (4), here we define the throughput of the path resulting from merging the sensor $\overline{\mathbf{x}}_{w}$ and light $\overline{\mathbf{x}}_{l}$ sub-paths as:

$$
\begin{equation*}
\mathfrak{T}_{\mathbf{v}}^{\prime}(\overline{\mathbf{x}})=\mathfrak{T}_{\mathbf{v}}\left(\overline{\mathbf{x}}_{w}\right) \mathcal{K}_{R}\left(\left\|\mathbf{x}_{k-m}-\mathbf{x}_{m}\right\|\right) \mathbf{K}\left(\mathbf{x}_{k-m}\right) \mathfrak{T}_{\mathbf{v}}\left(\overline{\mathbf{x}}_{l}\right) \tag{18}
\end{equation*}
$$

where $\mathcal{K}_{R}$ is the spatial smoothing kernel with bandwidth $R$. Note that given that we are merging vertices $\mathbf{x}_{k-m}$ and $\mathbf{x}_{m}$ (see Figure 1), we only have to apply one scattering kernel $\mathbf{K}\left(\mathbf{x}_{k-m}\right)$. Additionally, note that the scattering kernel $\mathbf{K}\left(\mathbf{x}_{k-m}\right)$ is defined with incoming direction the one from the light sub-path (the incoming direction of $\mathbf{x}_{m}$ ), while the outgoing direction is the inverse of the (virtual) incoming direction for the sensor sub-path's last vertex $\mathbf{x}_{k-m}$.

### 4.2. Implementation

We implemented our vector-based rendering on top of an inhouse physically based renderer written in $\mathrm{C}++$. In many cases,
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working with the full scattering matrices $\mathbf{K}$ was not needed, given that, e.g. light was unpolarized, the scattering kernel itself was a depolarizer, or the scattering event was inelastic. We add a flag to both the Stokes vectors, Müller matrices and reradiation matrices to discard computations depending on the type of light and interaction being computed. This significantly increases performance while not affecting the accuracy of results.

While our renderer supports spectral rendering, our tests have been performed using RGB (i.e. $L=3$ ). In case of using a large number of wavelengths, the costs of the multiplication of scattering kernels might be prohibitive. However, even considering their small size ( $3 \times 3 \times 4 \times 4$ ), scattering matrices supporting polarization and fluorescence are relatively sparse for most common operations, which can be exploited to increase the performance. Note that pre-integrating the reradiation matrix into RGB is accurate for single reflection, but might lead to problems in full global illumination solutions, as discussed by Meng et al. [MSHD15].

Finally, we have included five types of scattering kernels: a Lambertian BRDF acting as a depolarizer [WW12], a smooth conductor BRDF with complex index of refraction modelled with the Fresnel equations for conductors [WW12], a Fresnel smooth dielectric BSDF with support to both transmission and reflection [Azz04], a fluorescent material based on measured data for
chlorophyll [GSMA08] and a weakly polarizing Mie phase function obtained using MiePlot [Lav15].

## 5. Results

We demonstrate our implementation by rendering a set of scenes with complex light interactions, including dielectric, conductors and participating media. The selected results feature polarization, fluorescence and time-resolved light transport.

Figures 2-4 show different scenes showing polarization. These scenes include several diffuse surfaces, as well as both conductor and dielectric mirrors, dielectric transparent objects and participating media with different degrees of scattering. For visualizing the polarization, we use the techniques described by Wilkie and Weidlich [WW10], depicting the degree of polarization, the degree of circular polarization and orientation of linear polarization. In both cases, this visualization is super-imposed to the luminance image: brighter red means higher degree of polarization, while colour codes the orientation of linear polarization. We can observe how conductors are bad polarizers, and polarize light only slightly. However, dielectrics are very good polarizers. Moreover, in Figure 2, we can see how linear polarization switches to circular polarization when the frames of reference of reflection are disaligned (e.g. in poles of


Figure 2: Mirror scene, depicting a scene with a dielectric sphere and a conductor mirror reflected on a spherical conductor mirror. From left to right: scene setup, radiance, degree of polarization, degree of circular polarization and orientation of linear polarization (colour codes orientation). The degree of polarization and circular polarization is encoded in red (brighter is higher). While the conductors only polarize light weakly, the dielectric specular reflection is highly polarized. In addition, the concatenation of multiple reflections on curved surfaces shifts the linear polarization towards circular polarization.


Figure 3: Cups scene, depicting a scene with a dielectric mirror, and a set of glass cups filled with absorbing, scattering and a nonparticipating dielectric media. From left to right: radiance, degree of polarization and orientation of linear polarization. The scene shows the slightly polarizing effect of very scattering and optically thick media, specially visible in the middle cup.
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Figure 4: Cornell box with a conductor mirror in the floor, and two spheres (one conductor and one dielectric). From left to right: scene setup, radiance, degree of polarization, degree of circular polarization and orientation of linear polarization. The accumulative effect of polarization, in particular, how it switches from linear to circular polarization, is clearly visible in the multiple reflections on both spheres.


Figure 5: Fluorescent bunny made of chlorophyll. Although the direct reflection is bright green, some blue light is absorbed and re-emitted in the red band, resulting a paler yellowish appearance.
the dielectric sphere or in the top of the planar mirror). In addition, we can see how scattering media tends to slightly polarize light transport (Figure 3), although the main polarizers in that scene are still dielectrics.

Figure 5 shows the effect of introducing fluorescence, where a chlorophyll bunny is illuminated by white light. The direct reflection of the bunny is mostly green, while red and blue are mainly absorbed. However, most of the absorbed blue is later re-emitted in the red band, which, in the end, changes the appearance of the bunny. This effect can be seen more clearly in transient state: Figure 6 shows several frames of a time-resolved render of the same scene. Here, we can see the temporal behaviour of fluorescence, where light that is absorbed by the bunny is later re-emitted red-shifted. The full animation can be seen in the accompanying supplemental video.

Given that the illumination comes from a point light source and the caustic paths due to smooth dielectric and conductors are dominant, we use a stochastic progressive [HJ09, KZ11] version of our vector PM (Section 4.1) in scenes (Figures 2, 3, and 4), although we also account for multiple bounces on the sensor sub-path and perform deterministic shadow connections with the light. We computed 1000 iterations, with 16 samples per pixel and 5 M photon random walks on each iteration. We compute the initial kernel radius using the 20 nearest photons. Note that without bidirectional methods such as PM these scenes, where caustic light paths dominate, could be very hard, or even impossible, to render. For the bunny scene (Figures 5 and 6), we use a vector-based implementation of BPT, given that there is less caustic transport, with 25 K samples per pixel for rendering the full temporal profile.

In terms of performance, adding vector-based effects such as polarization or fluorescence increases the cost with respect to scalar


Figure 6: Time-resolved render for the bunny scene: From left to right, light travels from the light source to the bunny, which reflects most green light and absorbs red and blue light. Note that this reflection before reradiation would be the result if a scalar version of the algorithm was used. Light continues propagating, illuminating the scene via multiple reflections, while the mirror starts reflecting the bunny. After several diffuse reflections, light loses its directionality, becoming mainly diffuse lighting. Finally, after a few nano-seconds, absorbed light is re-emitted by the fluorescent bunny red-shifted. We refer to the supplemental video for the full animation.
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Table 1: Comparison of the average cost (in seconds) per iteration between scalar and vector rendering for the render examples shown in the paper. Note that Figure 2 was rendered in a lower end workstation.

| Figure | Support | Technique | Scalar | Vector |
| :--- | :--- | :--- | :---: | :---: |
| $2^{*}$ | Polarization | Photon mapping | 8 | 14 |
| 3 | Polarization | Photon mapping | 0.72 | 0.83 |
| 4 | Polarization | Photon mapping | 0.79 | 0.79 |
| 5 | Transient fluorescence | BDPT | - | 1.27 |



Figure 7: Convergence plots for our vector light transport (polarization) compared against traditional scalar rendering, for Figure 3 (left) and Figure 4 (right).
rendering (Table 1), due to the vector-to-matrix and the matrix-to-matrix products. However, we avoid these costly computations unless they are strictly needed, although with multiple scattering, these operations are common. The need of tracking the reference frame in the case of polarization also introduces additional costs. Note, however, that the vector and matrix operations are not optimized; using vectorized code could reduce the costs significantly. Additionally, while vector representation of light and scattering is relatively compact (or sparse), the memory cost is much higher than in traditional rendering: For example, for polarized photons, we need to store four floats for a single wavelength, in contrast to the single float per wavelength in scalar rendering, plus the need of storing the full frame, not only the photon's direction. Finally, in Figure 7, we analyse the convergence with respect to the number of samples for Figures 3 and 4 (polarized light transport), compared against scalar rendering: Given the additional dimensions introduced by rendering Stokes vectors, the error is slightly higher than scalar rendering.

## 6. Conclusions and Future Work

In this work, we have generalized the path integral formulation and its transient counterpart, to support vector-based light transport. This is needed to support effects such as polarization or fluorescence, and imposes a set of constraints on the scattering operators and their concatenation. Interestingly, these modifications do not reduce generality, but extend its range of applicability to transport operators breaking symmetry. Based on this theoretical framework, we have described the required changes to well-known (scalar) bidirectional rendering methods defined within Veach's [Vea97], and the unified path integral formulation [GKDS12, HPJ12]. In particular, we have shown how to include vector light transport in both BPT and PM.

We have shown that this form of representing stochastic light transport might be powerful for representing effects beyond traditional vector-based light (i.e. polarization or spectral rendering): In particular, we have shown that transient light transport can also be modelled in a vector-matrix representation. We believe that this is an interesting approach for rendering the full plenoptic function.

There is, of course, several future works ahead. First of all, we have proposed just a few areas of application to the new form of vector light transport. Modelling other non-symmetric effects in a vector-based representation might increase the range of effects that are representable in current render engines. From a practical perspective, developing sampling techniques aware of the particularities of the effect is a promising avenue of future work. Here, we have just used standard scalar-based techniques for both the random walk sampling and deterministic connection of light and sensor paths. Finally, proposing new stochastic shadow connections or optimal combinations of sampling techniques, based on, e.g. polarization state or the bi-spectral scattering kernel, would significantly increase the efficiency when rendering such effects.

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## Video S1

