

Phase Space Ray Tracing for Illumination Optics

Carmela Filosa

Cover art:
Photography:

A catalogue record is available from the Eindhoven University of Technology Library

ISBN: 978-90-386-3972-7

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title

PROEFSCHRIFT

ter verkrijging van de graad van doctor aan de
Technische Universiteit Eindhoven, op gezag van de
rector magnificus prof.dr.ir. F.P.T. Baaijens, voor een
commissie aangewezen door het College voor
Promoties, in het openbaar te verdedigen

door

Carmela Filosa

geboren te Torre del greco, Italië

Dit proefschrift is goedgekeurd door de promotoren en de samenstelling van de promotiecommissie is als volgt:

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leden:

Het onderzoek of ontwerp dat in dit proefschrift wordt beschreven is uitgevoerd in overeenstemming met de TU/e Gedragscode Wetenschapsbeoefening.

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List of symbols

τ	time
Q	Total energy emitted from a light source or received by a target
Φ_r	Radiant flux
Φ	Luminous flux
λ	Wavelength
Ψ_r	Power per wavelength
$\bar{y}(\lambda)$	Luminosity function
E	Illuminance
$d\Omega$	Solid angle
I	Intensity
L	Luminance
U	étendue
ν	Surface normal
n	Index of refraction of the medium in which a surface is immersed
θ	Angle between the direction of the solid angle and the normal ν
n_i	Index of refraction of the medium in which the incident ray travels
$n_r = n_i$	Index of refraction of the medium in which the reflected ray is located
n_t	Index of refraction of the medium in which the transmitted ray travels
$n_{i,t}$	$\frac{n_i}{n_t}$
θ_i	Angle between the incident ray and the normal ν
θ_r	Angle between the reflected ray and the normal ν
θ_t	Angle between the transmitted ray and the normal ν
θ_c	Critical angle
t_i	Direction of the incident ray
t_r	Direction of the reflected ray
t_t	Direction of the transmitted ray
ν_j	Normal to the line j
t_j	Angle that the ray located on line j forms with respect to the optical axis
θ_j	Angle between the ray and the normal ν_j to line j
n_j	Index of refraction of the medium in which line j is located

Chapter 1

Introduction

1.1 Motivation

1.2 Methods and results

1.3 Content of this thesis

Chapter 2

Illumination optics

This chapter provides some concepts of illumination optics used in this thesis. We start explaining the difference between radiometry and photometry. In particular, we focus on the photometric variables, defining them both in three and two dimensions. The reflection and refraction laws and the phenomenon of total internal reflection are explained. The last paragraph of the chapter gives a brief introduction to Fresnel reflection.

2.1 Radiometric and photometric variables

Radiometry is concerned with the measurement of electromagnetic radiation across the entire electromagnetic spectrum. Photometry is the subfield of radiometry that takes into account only the portion of the electromagnetic spectrum corresponding to the visible light, [1]. Radiometry deals with radiometric quantities. An important radiometric quantity is the radiant flux Φ_r (unit watt, [W]) which is the total energy emitted from a source or received by a target per unit time:

$$\Phi_r = \frac{dQ}{dT}, \quad (2.1.1)$$

where Q is the energy and T the time.

In illumination optics the measurement of light is given in terms of the impression that it gives on the human eye. Therefore, illumination optics deals with photometric variables. The most important photometric variables are defined as follows using the same notation adopted by Chaves in [2]. The luminous flux Φ (unit lumen, [lm]) is defined as the perceived power of light by the human eye. The radiant and the luminous flux are related by the luminous efficacy function, unit [lm/W], which tells us how many lumen there are for each Watt of power at a given wavelength. The luminous efficacy reaches its maximum at a wavelength of 555 nm where it is equal to 683 lm/W. We may normalize the luminous efficacy function with its maximum value of 683. This normalized function is the dimensionless luminosity function $\bar{y}(\lambda)$ shown in Figure 2.1 where λ is the wavelength.

The luminous flux corresponding to one Watt of radiation power at any wavelength is given by the product of 683 lm/W and the luminosity function at the same wavelength,

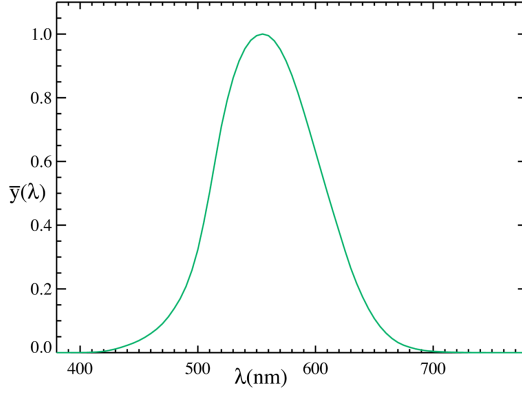


Figure 2.1: Luminosity function $\bar{y}(\lambda)$: relation between the eye's sensitivity and the wavelength of light. The luminosity function is dimensionless, [3].

i.e. $683 \bar{y}(\lambda)$. Hence, the total luminous flux Φ has unit lumen [lm] and it is defined as:

$$\Phi = 683 \int_0^\infty \Psi_r(\lambda) \bar{y}(\lambda) d\lambda \quad (2.1.2)$$

where $\Psi_r(\lambda)$ is the power in watt per unit wavelength (unit [W/m]).

A beam of light can be described as a collection of parallel light rays, where a light ray can be interpreted as a path along which the energy travels. The luminous flux $d\Phi$ incident on a surface is called illuminance E (unit [lm/m²]) and is defined as:

$$E = \frac{d\Phi}{dA}, \quad (2.1.3)$$

where dA is an infinitesimal area receiving radiation. The density of light emitted by a point source in a given direction is determined by the solid angle. The solid angle on a given direction is defined by the infinitesimal surface area dS of a sphere subtended by the radius of that sphere and by the rays emitted by the center on that direction, [4]. Indicating with r the radius of the sphere, the infinitesimal solid angle $d\Omega$ defined by dS is given by:

$$d\Omega = \frac{dS}{r^2}. \quad (2.1.4)$$

The solid angle on the entire sphere is $\Omega = 4\pi$, its unit is the steradian [sr] and it is usually defined on a unit sphere. The luminous intensity I (unit candela (cd), [cd = lm/sr]) is defined as the luminous flux $d\Phi$ per solid angle $d\Omega$ and is given by:

$$I = \frac{d\Phi}{d\Omega}. \quad (2.1.5)$$

The luminance L (unit [cd/m²]) is the luminous flux per unit solid angle $d\Omega$ and per unit projected area $\cos\theta dA$ where θ is the angle that the normal ν to the area dA makes with the direction of the solid angle $d\Omega$, as shown in Figure 2.2. L is given by:

$$L = \frac{d\Phi}{\cos\theta dA d\Omega}. \quad (2.1.6)$$

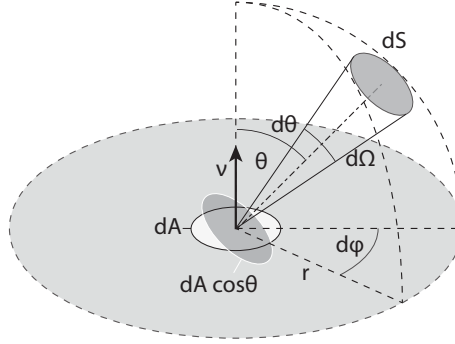


Figure 2.2: Solid angle $d\Omega$ in a direction making an angle θ with the normal to the area dA .

Note that from (2.1.5) and (2.1.6) we can derive a relation between the intensity and the luminance. The intensity I emitted by the infinitesimal area dA is given by:

$$I = \frac{d\Phi}{d\Omega} = L(\mathbf{x}, \theta) \cos \theta dA. \quad (2.1.7)$$

When the luminance is uniform over a finite area A , the luminous intensity emitted in the direction θ is equal to:

$$I(\theta) = L(\theta) A \cos \theta. \quad (2.1.8)$$

Thus, when $L(\mathbf{x}, \theta)$ does not depend on the position and the direction (i.e. $L(\mathbf{x}, \theta) = L$), we obtain Lambert's cosine law:

$$I(\theta) = I_0 \cos \theta. \quad (2.1.9)$$

where $I_0 = I(0) = LA$.

Finally, the étendue U (unit [m sr]) describes the ability of a source to emit light or the capability of an optical system to receive light, [5]. The quantity dU is defined as:

$$dU = n^2 \cos \theta dA d\Omega. \quad (2.1.10)$$

where n is the index of refraction of the medium in which the surface A is immersed. In optics the étendue is considered to be a volume in phase space (or an area for two-dimensional systems). This concept will be clarified in Chapter 4 in which we treat the phase space in more detail. An important property of the étendue is that it is conserved within an optical system in absence of absorption. We now show, using the approach of Chaves in [2], how conservation of this quantity can be derived. Consider a light ray emitted from an infinitesimal area dA_1 to the area dA_2 . Suppose that the centers of dA_1 and dA_2 are located at a distance d to each other, see Figure 2.3. Indicating with $\boldsymbol{\nu}_1$ and $\boldsymbol{\nu}_2$ the normals to the surfaces dA_1 and dA_2 , respectively and with θ_1 and θ_2 the angles that the central ray forms with $\boldsymbol{\nu}_1$ and $\boldsymbol{\nu}_2$, respectively, the flux $d\Phi_1$ passing through dA_2 coming from dA_1 and the corresponding solid angle

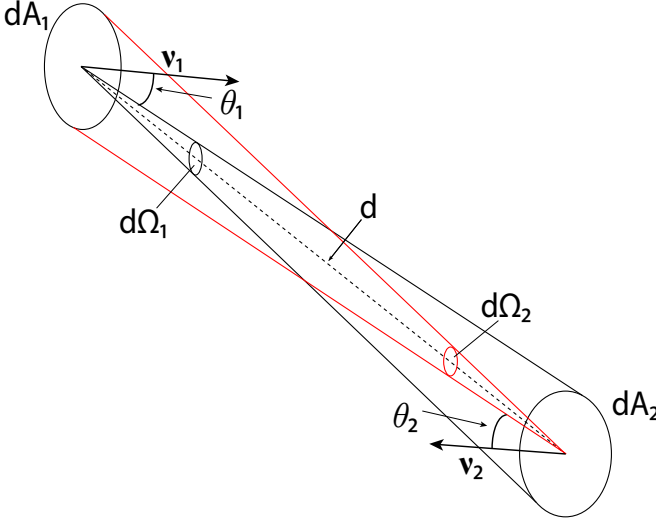


Figure 2.3: dA_1 and dA_2 are two surfaces with normals ν_1 and ν_2 , respectively. Their centers are located at a distance d . θ_1 and θ_2 are the angles made by the central ray with the normals ν_1 and ν_2 , respectively.

$d\Omega_1$ are defined as:

$$\begin{aligned} d\Phi_1 &= L \cos \theta_1 dA_1 d\Omega_1, \\ d\Omega_1 &= \frac{dA_2 \cos(\theta_2)}{d^2}. \end{aligned} \quad (2.1.11)$$

Similarly, the flux $d\Phi_2$ passing through dA_1 coming from dA_2 is equal to:

$$\begin{aligned} d\Phi_2 &= L \cos \theta_2 dA_2 d\Omega_2 \\ d\Omega_2 &= \frac{dA_1 \cos \theta_1}{d^2}. \end{aligned} \quad (2.1.12)$$

Then from Eq. (2.1.10) we obtain the following relations:

$$\begin{aligned} dU_1 &= n^2 dA_1 \cos \theta_1 d\Omega_1 = \frac{n^2 dA_1 \cos \theta_1 dA_2 \cos \theta_2}{d^2}, \\ dU_2 &= n^2 dA_2 \cos \theta_2 d\Omega_2 = \frac{n^2 dA_2 \cos \theta_2 dA_1 \cos \theta_1}{d^2} \end{aligned} \quad (2.1.13)$$

for dA_1 and dA_2 , respectively. From the previous equations we can conclude that $dU_1 = dU_2$ and therefore the étendue dU is conserved along a beam of light. Since also the flux through the areas dA_1 and dA_2 is conserved, the following relation holds:

$$L := n^2 \frac{d\Phi}{dU} = \text{constant}. \quad (2.1.14)$$

In the optical systems we will consider in this work, the source and the target are located in the same medium (air) with $n = 1$, so the luminance L equals the basic

luminance $L^* = L/n^2$ at the source and the target of the system.

In this thesis we consider two-dimensional optical systems. Hence, the definitions of the photometric parameters have to be given in two dimensions. An infinitesimal line segment of length da that emits a light beam and the ray that makes an angle θ with the normal ν are considered, see Fig. 2.4.

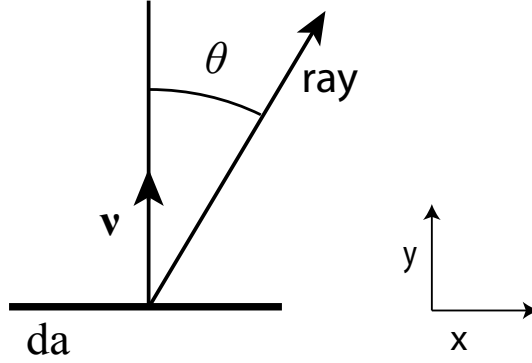


Figure 2.4: Ray emitted by an infinitesimal line segment da that makes an angle θ with respect to the line normal ν .

The two-dimensional illuminance (unit $[\text{lm}/\text{m}]$) denotes the luminous flux falling on an infinitesimal line segment of length da and it is given by:

$$E = \frac{d\Phi}{da} . \quad (2.1.15)$$

The luminous intensity (unit $[\text{lm}/\text{rad}]$) is the luminous flux per angle $d\theta$:

$$I = \frac{d\Phi}{d\theta} . \quad (2.1.16)$$

The two-dimensional luminance (unit $[\text{lm}/(\text{rad} \cdot \text{m})]$) is given by:

$$L = \frac{d\Phi}{\cos \theta da d\theta} . \quad (2.1.17)$$

Thus the following relation holds:

$$I = L(x, \theta) \cos \theta da \quad (2.1.18)$$

where x is a certain position at the light source da . Finally, the étendue dU (unit $[\text{m} \cdot \text{rad}]$) in two dimensions is given by:

$$dU = n \cos \theta da d\theta . \quad (2.1.19)$$

In order to determine the light distribution on a surface and to compute the photometric variables on that surface, we need to understand how the light emitted from the source propagates. In the field of geometric optics the light propagation is described by light rays. The propagation of a light ray traveling through different media is determined by the reflection and refraction law. In the following we introduce these two laws and we explain the total internal reflection phenomenon.

2.2 Reflection and refraction law

A light ray is described by a position vector \mathbf{x} on a surface and a direction vector \mathbf{t} and can be parameterized by the arc length s . Light rays travel in a homogeneous medium along straight lines, once they hit a reflective surface their direction changes. Denoting with \mathbf{t}_i the direction of the incident ray and with $\boldsymbol{\nu}$ the unit normal to the surface at the location of incidence, the direction \mathbf{t}_r of the reflected ray is given by:

$$\mathbf{t}_r = \mathbf{t}_i - 2(\mathbf{t}_i \cdot \boldsymbol{\nu})\boldsymbol{\nu}, \quad (2.2.1)$$

where the vectors \mathbf{t}_i and $\boldsymbol{\nu}$ are unit vectors and $\mathbf{t}_i \cdot \boldsymbol{\nu}$ indicates the scalar product between \mathbf{t}_i and $\boldsymbol{\nu}$. From Eq. (2.2.1) it follows that the vector \mathbf{t}_r is a unit vector too, indeed considering the scalar product $(\mathbf{t}_r, \mathbf{t}_r)$ we conclude:

$$\mathbf{t}_r \cdot \mathbf{t}_r = \mathbf{t}_i \cdot \mathbf{t}_i - 4(\mathbf{t}_i \cdot \boldsymbol{\nu})(\mathbf{t}_i \cdot \boldsymbol{\nu}) + 4(\mathbf{t}_i \cdot \boldsymbol{\nu})^2(\boldsymbol{\nu} \cdot \boldsymbol{\nu}) = 1. \quad (2.2.2)$$

The vectors \mathbf{t}_i , \mathbf{t}_r and $\boldsymbol{\nu}$ live all in the same plane. Defining the incident angle θ_i and the reflective angle θ_r such that $\theta_i, \theta_r \in [0, \pi/2)$. the reflection law states that $\theta_i = \theta_r$, see Fig. 2.5.

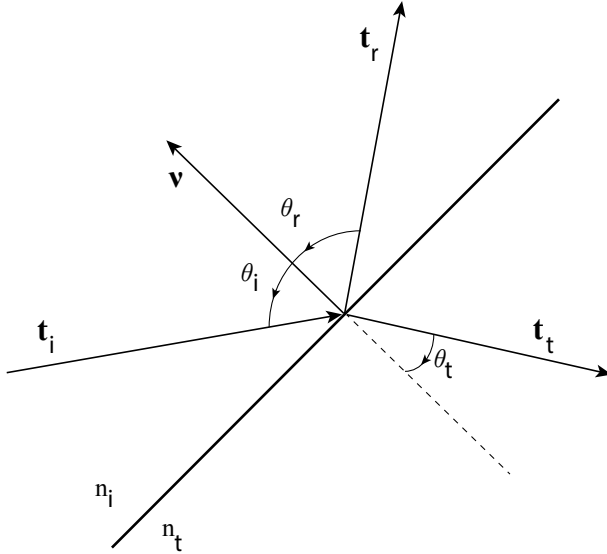


Figure 2.5: Propagation of a ray through two different media with index of refraction n_i and n_t .

When a ray propagates through two different media, its direction changes according to the law of refraction. Indicating with n_i the index of refraction of the medium in which the incident ray travels and with n_t the index of refraction of the medium of the transmitted ray, the direction \mathbf{t}_t of the transmitted ray is given by:

$$\mathbf{t}_t = n_{i,t} \mathbf{t}_i + \left[\sqrt{1 - n_{i,t}^2 + n_{i,t}^2 (\boldsymbol{\nu} \cdot \mathbf{t}_i)^2} - n_{i,t} (\boldsymbol{\nu} \cdot \mathbf{t}_i) \right] \boldsymbol{\nu}, \quad (2.2.3)$$

where $n_{i,t} = n_i/n_t$, [2]. Note that in Eq. (2.2.1) the direction of the normal $\boldsymbol{\nu}$ to the surface is not relevant for the computation of the direction of the reflective ray, since:

$$\mathbf{t}_r = \mathbf{t}_i - 2(\mathbf{t}_i \cdot \boldsymbol{\nu})\boldsymbol{\nu} = \mathbf{t}_i - 2(\mathbf{t}_i \cdot -\boldsymbol{\nu})(-\boldsymbol{\nu}), \quad (2.2.4)$$

however, this is not the case for Eq. (2.2.3), therefore in the latter case we need to specify the direction of $\boldsymbol{\nu}$ which is usually chosen in such a way that the angle that it forms with the incident ray \mathbf{t}_i is smaller than or equal to $\pi/2$. Hence, if $(\mathbf{t}_i, \boldsymbol{\nu}) \leq 0$ the normal $\boldsymbol{\nu}$ directed inside the same medium in which travels the incident ray is taken as in Fig. 2.5, otherwise the normal $-\boldsymbol{\nu}$ directed inside the same medium in which the transmitted ray will travel has to be considered.

Eq. (2.2.3) is only valid for

$$1 - n_{i,t}^2 + n_{i,t}^2(\boldsymbol{\nu} \cdot \mathbf{t}_i)^2 \geq 0 \quad (2.2.5)$$

which implies that

$$\frac{n_t}{n_i} \geq \sqrt{1 - (\boldsymbol{\nu} \cdot \mathbf{t}_i)^2} \quad (2.2.6)$$

from which we obtain:

$$n_t \geq n_i \sin \theta_i. \quad (2.2.7)$$

The angle θ_c for which the equality holds is

$$\theta_c = \arcsin\left(\frac{n_t}{n_i}\right) \quad (2.2.8)$$

and it is called the critical angle, [2]. When the incident angle θ_i is exactly equal to the critical angle θ_c , the square root in Eq. (2.2.3) is zero and the inner product $(\mathbf{t}_t, \boldsymbol{\nu}) = 0$, hence the transmitted ray propagates parallel to the refractive surface. When $\theta_i > \theta_c$ the light ray is no longer refracted but is only reflected by the surface. This phenomenon is called total internal reflection (TIR). When TIR occurs, 100% of light is reflected and there is no loss of energy. Therefore, optical systems designed such that rays are reflected by TIR are very efficient. Light that hits an ordinary refractive surface can be reflected and refracted. The energy that is reflected and refracted is determined by the Fresnel's coefficients. In the next paragraph an overview of the Fresnel coefficients is given.

2.3 Fresnel's equations

In order to derive Fresnel's equations we need to describe light as an electromagnetic wave. It is therefore useful to study the light propagation from the perspective of electromagnetic theory which gives information about the incident, reflected and transmitted radiant flux density that are denoted with E_i , E_r and E_t , respectively. Any component of the electric field \mathcal{E} can be written as

$$\mathcal{E}(\mathbf{x}, \tau) = \mathcal{E}_0(\mathbf{x})e^{i(\mathbf{k} \cdot \mathbf{x} - \omega \tau)} \quad (2.3.1)$$

where \mathbf{x} is the position vector and T is the time. The amplitude $\mathcal{E}_0(\mathbf{x})$ is constant in time and $\omega = \frac{ck}{n}$ is the value of the angular frequency with c the velocity of light

and n the index of refraction in which the wave is traveling, which is the ratio of the speed of light c in vacuum and the speed of light v in the material. Note that the angular frequency can be also written as $\omega = vk$, in particular when a wave travels in vacuum $n = 1$ and $\omega = ck$. The vector \mathbf{k} has the same direction of the wave and its absolute value $|\mathbf{k}| = k = \frac{2\pi}{\lambda}$ is the wave number in vacuum, with λ the wavelength. Similarly, the magnetic field has the form:

$$\mathcal{B}(\mathbf{x}, \tau) = \mathcal{B}_0(\mathbf{x})e^{i(\mathbf{k} \cdot \mathbf{x} - \omega\tau)}. \quad (2.3.2)$$

Light can be seen as an electromagnetic wave, that is an oscillating electric field \mathcal{E} and an oscillating magnetic field \mathcal{B} which propagates always perpendicular to \mathcal{E} . The electric field oscillates perpendicular to the wave propagation. Light is said to be polarized if the direction of the electric field is well defined. When the electric field propagates in different directions we talk about unpolarized light. By convention, we refer to the light's polarization as the direction of the electric field \mathcal{E} , [6] with respect to the incident plane that is defined by the incident and reflected rays as is shown in Fig. 2.6.

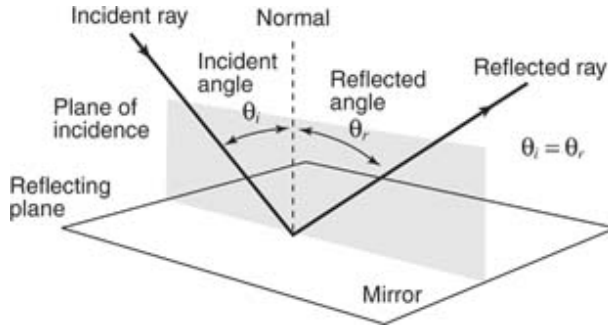


Figure 2.6: Light ray that hits a mirror located on the reflecting plane. The incident and the reflected ray leave in the same plane of the normal to the mirror that is called plane of incidence.

In order to derive the Fresnel's coefficients the polarization of light must to be taken into account. Those coefficients are obtained considering Maxwell's equations and the boundary conditions due to the conservation of energy. The details of Fresnel's equations are widely explained in the literature. In the following we provide Fresnel coefficients and we briefly explain their physical interpretation. We refer the reader to [7, 8] for more details. Fresnel's coefficients can also be derived using a different approach that does not involve Maxwell's equations, this method is explained in [9]. The following particular cases of light's polarization need are considered.

1. \mathcal{E} is perpendicular to the plane of incidence (see Fig. 2.7). In this case light is said to be *s*-polarized.
2. \mathcal{E} is parallel to the plane of incidence (see Fig. 2.8). In this case light is said to be *p*-polarized.

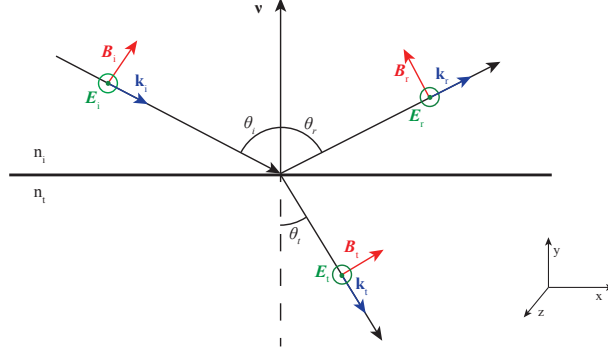


Figure 2.7: Propagation of an electromagnetic wave where \mathcal{E} is perpendicular to the incident plane. The components of \mathcal{E} are indicated with the green circles. The components of \mathcal{B} are indicated with red arrows.

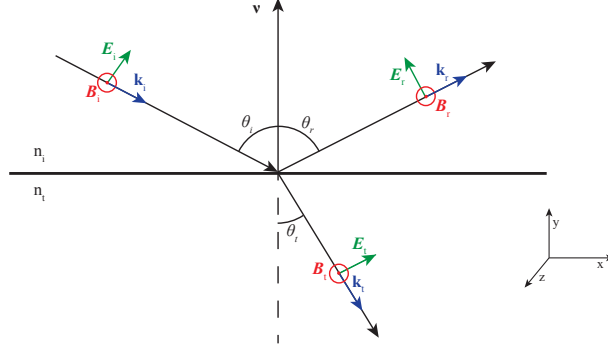


Figure 2.8: Propagation of an electromagnetic wave where \mathcal{E} is parallel to the incident plane. The components of \mathcal{B} are indicated with the red circle. The components of \mathcal{E} are indicated with green arrows.

Energy conservation gives the boundary conditions of the electromagnetic field at the plane of the interface (which is perpendicular to the incident plane). In the following we derive Fresnel's coefficients for case 1. Similarly, the Fresnel's coefficients can be derived for the second case.

For s -polarized light the tangential components of \mathcal{E} and \mathcal{B}/μ across the boundary between the two different media must be continuous. The continuity of the tangential component of \mathcal{E} leads to:

$$|\mathcal{E}_{0i}| + |\mathcal{E}_{0r}| = |\mathcal{E}_{0t}|, \quad (2.3.3)$$

while the continuity of the tangential component of \mathcal{B}/μ gives:

$$-\frac{|\mathcal{B}_{0i}|}{\mu_i} \cos \theta_i + \frac{|\mathcal{B}_{0r}|}{\mu_r} \cos \theta_r = -\frac{|\mathcal{B}_{0t}|}{\mu_t} \cos \theta_t, \quad (2.3.4)$$

where the negative sign in front of $|\mathcal{B}_{0,i}|$ and $|\mathcal{B}_{0,t}|$ is due to the convention that a positive direction is considered with increasing x . Since $\mathcal{B} = \mathcal{E}/v$, Eq. (2.3.4) can be written as

$$\frac{1}{\mu_i v_i} (|\mathcal{E}_{0,i}| - |\mathcal{E}_{0,r}|) \cos \theta_i = \frac{1}{\mu_t v_t} |\mathcal{E}_{0,t}| \cos \theta_t, \quad (2.3.5)$$

where we employed the fact that $v_i = v_r$, and $\theta_i = \theta_r$. Using Eq. (2.3.1) and $n = c/v$, the previous equation becomes:

$$\frac{n_i}{\mu_i} (|\mathcal{E}_{0,i}| - |\mathcal{E}_{0,r}|) \cos \theta_i = \frac{n_t}{\mu_i} |\mathcal{E}_{0,t}| \cos \theta_t \quad (2.3.6)$$

Finally, assuming that $\mu_i = \mu_t = \mu_0$ and employing Eq. (2.3.3) we obtain:

$$\begin{aligned} r_s &= \frac{|\mathcal{E}_{0,r}|_s}{|\mathcal{E}_{0,i}|_s} = \frac{n_i \cos \theta_i - n_t \cos \theta_t}{n_i \cos \theta_i + n_t \cos \theta_t}, \\ t_s &= \frac{|\mathcal{E}_{0,t}|_s}{|\mathcal{E}_{0,i}|_s} = \frac{2n_i \cos \theta_i}{n_i \cos \theta_i + n_t \cos \theta_t}. \end{aligned} \quad (2.3.7)$$

The coefficients r_s and t_s are amplitude coefficients for the reflected and transmitted light. They are the perpendicular components of r and t for s -polarized light. Using Snell's law, that is $n_i \sin \theta_i = n_t \sin \theta_t$, the relations for r_s and t_s are simplified as follows:

$$\begin{aligned} r_s &= -\frac{\sin(\theta_i - \theta_t)}{\sin(\theta_i + \theta_t)}, \\ t_s &= -\frac{2 \sin \theta_t \cos \theta_i}{\sin(\theta_i + \theta_t)}. \end{aligned} \quad (2.3.8)$$

A similar argument for the p -polarized light leads to the calculation of the parallel components r_p and t_p of r and t . In case \mathcal{E} is parallel to the plane of incidence the amplitude coefficients are:

$$\begin{aligned} r_p &= \frac{n_t \cos \theta_i - n_i \cos \theta_t}{n_i \cos \theta_t + n_t \cos \theta_i}, \\ t_p &= \frac{2n_i \cos \theta_i}{n_i \cos \theta_t + n_t \cos \theta_i}, \end{aligned} \quad (2.3.9)$$

and their simplified relations are:

$$\begin{aligned} r_p &= \frac{\tan(\theta_i - \theta_t)}{\theta_i + \theta_t}, \\ t_p &= \frac{2 \sin \theta_t \cos \theta_i}{\sin(\theta_i + \theta_t) \cos(\theta_i - \theta_t)}. \end{aligned} \quad (2.3.10)$$

Furthermore, it can be checked that

$$\begin{aligned} t_s - r_s &= 1, \\ t_p + r_p &= 1. \end{aligned} \quad (2.3.11)$$

The amplitude coefficients are shown in Fig. 2.9 for the case in which light travels from a less dense to a more dense medium ($n_i < n_t$), that is external reflection. In

Fig. 2.10 the reflection coefficients are shown for the case in which $n_i > n_t$, that is internal reflection. Note from Fig. 2.9 that r_p approaches 0 when θ_i approaches to θ_p and it gradually decreases reaching -1 for an incident angle $\theta_i = 90^\circ$. The angle θ_p is called Brewster's angle or polarization angle as only the component perpendicular to the incident plane is reflected at that angle and therefore light is perfectly polarized. Similarly, Fig. 2.10 shows that $r_p = 0$ for $\theta_i = \theta_{p'}$. It can be show that $\theta_p + \theta_{p'} = 90^\circ$. Both r_p and r_s reach 1 when $\theta_i = \theta_c$. θ_c is called the critical angle. Light that hits the incident plane with an incident angle equal to or greater than the critical angle is totally reflected back and no transmitted light is observed. This phenomenon is called total internal reflection.

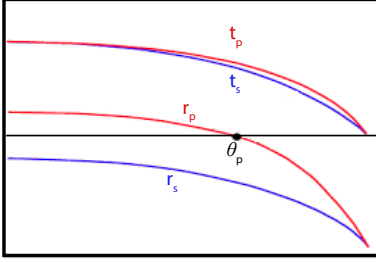


Figure 2.9: Amplitude coefficients of reflection and transmission as a function of the incident angle θ_i in the case of external reflection, i.e. $n_t < n_i$ ($n_t = 1$ and $n_i = 1.5$). θ_p is the polarization angle, [8].

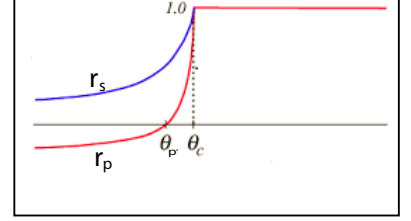


Figure 2.10: Reflection coefficients as a function of the incident angle θ_i in the case of internal reflection, i.e. $n_t > n_i$ ($n_t = 1.5$ and $n_i = 1$). $\theta_{p'}$ is the polarization angle and θ_c is the critical angle, [8].

The we introduce the Poynting vector \mathbf{P} that defines the energy flux of an electromagnetic field. It is measured in $[\text{W}/\text{m}^2]$, and it is given by:

$$\mathbf{P} = \frac{1}{\mu} (\mathbf{\mathcal{E}} \times \mathbf{\mathcal{B}}), \quad (2.3.12)$$

where $\mu = \frac{1}{\epsilon v^2}$ is the permeability and ϵ the permittivity of the medium. In the following, the parameters for vacuum are indicated with the subscript 0. All quantities defined in the media of the incident, reflective and transmitted light are indicated with the subscripts i, r and t, respectively. Optical rays are perpendicular to the wave front of an electromagnetic wave and parallel to the Poynting vector, [10]. The irradiance E is defined as the average energy that crosses in unit time a unit area A perpendicular to the direction of the energy flow. Therefore, defining the average of the vector \mathbf{P} over the time as:

$$\langle \mathbf{P} \rangle_T = \frac{1}{T} \int_0^T \mathbf{P} dt \quad (2.3.13)$$

we can write the irradiance E as:

$$E = \langle \mathbf{P} \rangle_\tau = v \epsilon |\mathbf{\mathcal{E}}|^2. \quad (2.3.14)$$

Considering a beam of light that hits a surface such that an area A is illuminated, the incident, reflected and transmitted beams are $\mathbf{E}_i A \cos \theta_i$, $\mathbf{E}_r A \cos \theta_r$ and $\mathbf{E}_t A \cos \theta_t$,

respectively. The reflectance \mathcal{R} is the ratio of the reflected power to the incident power:

$$\mathcal{R} = \frac{|\mathbf{E}_r| \cos \theta_r}{|\mathbf{E}_i| \cos \theta_i} = \frac{|\mathcal{E}_{0r}|^2}{|\mathcal{E}_{0i}|^2} = r^2 \quad (2.3.15)$$

where the second equality holds because $v_i = v_t$, $\varepsilon_i = \varepsilon_t$ and $\theta_i = \theta_t$. Similarly, the transmittance \mathcal{T} is the ratio between the transmitted to the incident power:

$$\mathcal{T} = \frac{|\mathbf{E}_t| \cos \theta_t}{|\mathbf{E}_r| \cos \theta_r} = \frac{n_t \cos \theta_t}{n_t \cos \theta_i} \frac{|\mathcal{E}_{0t}|^2}{|\mathcal{E}_{0i}|^2} = \frac{n_t \cos \theta_t}{n_t \cos \theta_i} t^2. \quad (2.3.16)$$

Employing total energy conservation, that is:

$$\mathbf{E}_i A \cos \theta_i = \mathbf{E}_r A \cos \theta_r + \mathbf{E}_t A \cos \theta_t, \quad (2.3.17)$$

we can easily prove that:

$$\mathcal{R} + \mathcal{T} = 1. \quad (2.3.18)$$

The parallel and perpendicular components of \mathcal{R} and \mathcal{T} are:

$$\begin{aligned} \mathcal{R}_p &= r_p^2, \\ \mathcal{T}_p &= \frac{n_t \cos \theta_t}{n_t \cos \theta_i} t_p^2, \\ \mathcal{R}_s &= r_s^2, \\ \mathcal{T}_s &= \frac{n_t \cos \theta_t}{n_t \cos \theta_i} t_s^2. \end{aligned} \quad (2.3.19)$$

it can be show that

$$\begin{aligned} \mathcal{R}_s + \mathcal{R}_p &= 1, \\ \mathcal{T}_s + \mathcal{T}_p &= 1. \end{aligned} \quad (2.3.20)$$

For normal incidence, i.e. $\theta_i = 0$, there is no polarization and Eqs. (2.3.19) lead to:

$$\begin{aligned} \mathcal{R} = \mathcal{R}_p = \mathcal{R}_s &= \left(\frac{n_i - n_t}{n_t + n_i} \right)^2, \\ \mathcal{T} = \mathcal{T}_p = \mathcal{T}_s &= \frac{4n_i n_t}{(n_t + n_i)^2}. \end{aligned} \quad (2.3.21)$$

Many common light sources such as sunlight, halogen lighting, LED spotlights, and incandescent bulbs produce unpolarized light. In case of unpolarized light the amount of reflected and transmitted light is given by the average of reflectance \mathcal{R} and transmittance \mathcal{T} calculated considering first p -polarized light and then s -polarization, that is:

$$\begin{aligned} \mathcal{R} &= \frac{\mathcal{R}_p + \mathcal{R}_s}{2}, \\ \mathcal{T} &= \frac{\mathcal{T}_p + \mathcal{T}_s}{2}, \end{aligned} \quad (2.3.22)$$

where \mathcal{R}_p , \mathcal{R}_s , \mathcal{T}_p and \mathcal{T}_s are given in Eqs. (2.3.19).

With this overview we conclude this chapter. The notions given in Section 2.1 will be used in the entire thesis as our goal is to study the distribution of light at the target of some optical systems. In particular we will focus on the computation of the output intensity distribution. The reflection and refraction laws explained in Section 2.2 are needed to determine how the optical system changes the ray's direction every time that it hits a surfaces (or a line in the two-dimensional case). In Chapters 3, 4, 5, 6 and ?? only systems where the reflection and refraction laws play a role are considered. Systems with Fresnel reflection are treated in the last chapter. The amount of reflected and transmitted light is calculated using the Fresnel's equation (introduced in the last paragraph of this chapter). Since, we restrict ourselves to two-dimensional systems, the value of reflectance and transmittance will be computed using Eqs. (2.3.22).

Chapter 3

Ray tracing

Ray tracing is a geometric problem that calculates the transport of light within optical systems. It is a forward method which uses single rays to describe light. Given an optical system and a set of rays at the source, ray tracing relates the input light distribution with the output distribution. The influence of diffraction on the transport of a ray is neglected and geometrical modeling of an optical system needs to be considered. Generally, the method can be implemented for two or more dimensions and for any optical system. In this thesis, we restrict ourselves to the two-dimensional case. Hence, from now on we will refer to optical lines instead of optical surface. The two-dimensional case is particularly relevant because it is a good test case to demonstrate the performance of the new methods. Optical designers often start with 2D systems where only the meridional plane is taken into account because it gives a good prediction of the target distribution of the rays (see [11], chapter 4, p.50 – 65). However, the two-dimensional case also has some limitations, as it may not identify skew rays that are turned back by the system, and so a 2D analysis cannot guarantee a proper treatment of non meridional rays in 3D. In this chapter we consider systems where only reflection and refraction laws are taken into account.

3.1 Ray tracing for two-dimensional optical systems

Ray tracing consists of tracing each ray, which is considered to be a broken line, through the optical system. Every ray emitted from the source is followed until it reaches the target. During its propagation it can encounter optical components which change its direction. The ray tracing procedure is constructed such that the position and the direction of the rays are calculated on every optical lines that they hit until they reach the target. Given a Cartesian coordinate system (x, z) , a two-dimensional optical system symmetric with respect to the z -axis is defined. Hence, usually the optical axis coincides with the z -axis. The optical system is formed by a source S , a target T and some optical components labeled with indexes j where $j \in \{2, \dots, Nl-1\}$ and Nl indicates the number of lines that form the system. S and T are indicated with the indexes 1 and Nl , respectively. The index of refraction of the medium in which line j is located is indicated with n_j . Every ray emitted by S (line 1) can hit some optical components $j \in \{2, \dots, Nl-1\}$ before reaching T (line Nl). The intersection point of

the rays with line j are $(x_j, z_j)_{j=1, \dots, N_l}$ and, $\mathbf{s}_j = (-\sin t_j, \cos t_j)$ indicates the direction vector of the rays that leave j , with t_j the angle that the ray forms with respect to the optical axis measured counterclockwise. As we consider only forward rays, the angles $t_j \in (-\pi/2, \pi/2)$. Therefore, a ray segment between (x_j, z_j) and (x_k, z_k) with $k \neq j$ is parameterized in real space by:

$$\mathbf{r}(s) = \begin{pmatrix} x_j - s \sin t_j \\ z_j + s \cos t_j \end{pmatrix} \quad 0 < s \leq s_{\max}, \quad (3.1.1)$$

where s denotes the arc-length and s_{\max} is the maximum value that it can assume. Fig. 3.1 shows an example where a single ray is traced inside a very simple optical system, the so-called two-faceted cup. The light source $S = [-a, a]$ (line 1) and the

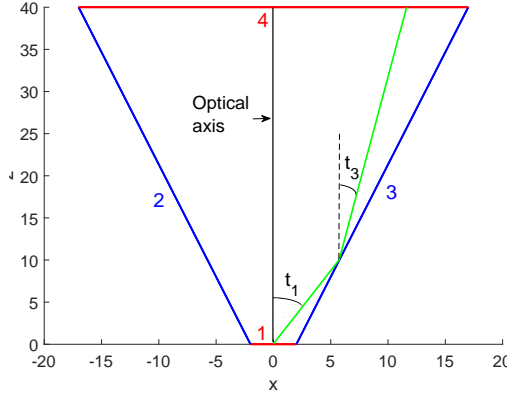


Figure 3.1: Shape of the two-faceted cup. Each line of the system is labeled with a number. The source $S = [-2, 2]$ (line number 1) is located on the x -axis. The target $T = [-17, 17]$ (line 4) is parallel to the source and is located at a height $z = 40$. The left and right reflectors (line 2 and 3) connect the source with the target.

target $T = [-b, b]$ (line 4) are two segments normal to the z -axis, where $a = 2$ and $b = 17$. The left and right reflectors (line 2 and 3) are oblique segments that connect the source and the target. All the optical lines $j \in \{1, \dots, 4\}$ are located in air, therefore the refractive index $n_j = 1$ for every j .

In order to compute the target photometric variables, we need to know how the optical system influences the direction of the rays when they hit an optical line. Ray tracing relates the position coordinates (x_1, z_1) and the direction vector \mathbf{s}_1 of every ray at the source S with the corresponding position (x_{N_l}, z_{N_l}) and direction \mathbf{s}_{N_l} at the target T . As in the following we will use often the target coordinates of the rays, from now on, to simplify the notation, we do not write the subscript N_l for the target coordinates. Hence, from now on, we write (x, z) instead of (x_{N_l}, z_{N_l}) , t instead of t_{N_l} and \mathbf{s} instead of \mathbf{s}_{N_l} for the target coordinates. The ray tracing algorithm can be schematized as follows.

1. Given a ray that leaves S with initial position (x_1, z_1) and initial direction $\mathbf{s}_1 = (-\sin t_1, \cos t_1)$, use Eq. (3.1.1) to implement the ray parametrization $\mathbf{r}(s_1)$;

2. Compute the coordinates $(x_k, z_k)_{k=1 \dots, N_l}$ of the intersection point of the parameterized ray $\mathbf{r}(s_1)$ with all the lines that it hits;
 - a) If the shape of the lines is described by an analytical equation, the intersection points are determined analytically;
 - b) If there is no analytic description for the optical lines, the intersections need to be determined using iterative methods;
3. Determine the closest line j that the forward ray encounters;
4. If $j = N_l$ stop the procedure, the target ray's coordinates (x, z) and \mathbf{s} are found.
5. Calculate the normal $\boldsymbol{\nu}_i$ to line i at the point (x_i, z_i) ;
6. Compute the new ray direction \mathbf{s}_j of the ray that leaves line j at the point (x_i, z_i) ;
 - a) If the incident line is a reflective line, \mathbf{s}_j is given by Eq. (2.2.1);
 - b) If the incident line is a refractive line, \mathbf{s}_j is given by Eq. (2.2.3);
7. Restart the procedure from 1. for the ray that leaves line j instead of S . Consider as initial ray position coordinates (x_i, z_i) instead of (x_1, z_1) and as initial ray direction $\mathbf{s}_j = (-\sin t_j, \cos t_j)$ instead of \mathbf{s}_1 .

The procedure explained above is repeated for every ray traced through the system, [12]. Once the target position and the direction of every ray traced are computed, the target photometric variables can be calculated.

There are different ways to implement the ray tracing procedure. The efficiency of the ray tracing can be related with the distribution of the rays at the source. If the initial position and direction of the rays are chosen randomly we have Monte Carlo (MC) ray tracing. This is a very common method in non-imaging optics as it is very powerful and easy to implement. MC ray tracing will be explained in details in the next paragraph of this chapter. If the rays are chosen from a so-called low discrepancy sequence we have the Quasi-Monte Carlo (QMC) ray tracing. This will be clarified in Section 3.3.

3.2 Monte Carlo ray tracing

Before explain MC ray tracing we give a general introduction about MC methods. MC simulations are often used to approximate the integral of a given function f considering random variables in the domain of f . Given an interval $D = [\mathbf{a}, \mathbf{b}]$ with $\mathbf{a} = (a_1, \dots, a_d)$ and $\mathbf{b} = (b_1, \dots, b_d)$ elements of \mathbb{R}^d such that $[\mathbf{a}, \mathbf{b}] = [a_1, b_1] \times \dots \times [a_d, b_d]$, a function $f : [\mathbf{a}, \mathbf{b}] \subset \mathbb{R}^d \mapsto \mathbb{R}$ and a random variable $\mathbf{y} \in D$ with probability density function $\rho(\mathbf{y})$, the expected value of f is defines as:

$$\mathbb{E}[f] = \int_D f(\mathbf{y})\rho(\mathbf{y})d\mathbf{y}. \quad (3.2.1)$$

Note that if ρ is a uniform probability density function, we have:

$$\mathbb{E}[f] = \int_D f(\mathbf{y})\rho(\mathbf{y})d\mathbf{y} = \frac{1}{(\mathbf{b} - \mathbf{a})} \int_D f(\mathbf{y})d\mathbf{y}. \quad (3.2.2)$$

In Monte Carlo methods N independent random variables $\{\mathbf{y}_i\}_{i=1,\dots,N} \in D$ are generated from the density function ρ and, the sum

$$S_N(f) = \frac{1}{N} \sum_{i=1}^N f(\mathbf{y}_i) \quad (3.2.3)$$

is calculated, [13]. The strong law of large numbers tells us that:

$$\Pr\left(\lim_{N \rightarrow \infty} \sum_{i=1}^N f(\mathbf{y}_i) = \mathbb{E}[f(\mathbf{y})]\right) = 1. \quad (3.2.4)$$

Therefore the following relation holds:

$$\mathbb{E}[f] \approx \frac{1}{N} \sum_{i=1}^N f(\mathbf{y}_i). \quad (3.2.5)$$

Suppose that f has variance $\text{Var}[f] = \sigma^2[f]$, then

$$\text{Var}[S_N(f)] = \mathbb{E}[(S_N(f) - \mathbb{E}[S_N(f)])^2] = \mathbb{E}[(S_N(f) - \mathbb{E}[f])^2] = \sigma[f]^2/N, \quad (3.2.6)$$

where the second equality follows from the linearity of the expected value, that is:

$$\mathbb{E}[S_N(f)] = \frac{1}{N} \sum_{i=1}^N \mathbb{E}[f(\mathbf{y}_i)]. \quad (3.2.7)$$

Indicating the integration error with:

$$\text{err}(f, S_N) = \left| \int_D f(\mathbf{y}) \rho(\mathbf{y}) d\mathbf{y} - S_N(f) \right|, \quad (3.2.8)$$

we obtain that

$$\mathbb{E}[\text{err}(f, S_N)] \leq \sqrt{\mathbb{E}[\text{err}(f, S_N)^2]} = \frac{\sigma[f]}{\sqrt{N}}. \quad (3.2.9)$$

Hence, the absolute value of the integration error is, on average, bounded by $\sigma[f]/\sqrt{N}$, where $\sigma[f]$ is the standard deviation of f , [14]. It is worthy to note that $\text{err}(f, S_N)$ does not depend on the dimension d of f .

MC technique can be combined with the ray tracing procedure in order to compute the light distribution at the target of an optical system. In MC ray tracing the position and the direction of every ray at the source are chosen randomly. In the two-dimensional case ($d=2$), for every ray we need to choose one position coordinate x_1 at the source and one angular coordinate t_1 at the target, indeed the z_1 coordinate of every ray at the source is always given (for instance, for the two-faceted cup in Fig. 3.1, $z_1 = 0$ for every ray). Therefore, given a set of random variables $\{\mathbf{y}_1, \dots, \mathbf{y}_N\} \in [\mathbf{a}, \mathbf{b}] \subset \mathbb{R}^2$, the initial position coordinate x_1 of the k -th ray corresponds to the first component of the k -th random variable \mathbf{y}_k and, the starting angular coordinate tt_1 of the k -th ray corresponds to the second component of the k -th random variable \mathbf{y}_k . Next, rays with those random coordinates at S are traced from S to T and,

a probabilistic interpretation of the output photometric variables is provided. In particular, we are interested in the total target intensity I which is computed as a function of the angular coordinate t . The MC intensity is calculated dividing the target into intervals of the same length, the so-called bins. A partitioning $P_1 : -\pi/2 = t_0 < t_1 < \dots < t_{\text{Nb}} = \pi/2$ of the interval $[-\pi/2, \pi/2]$ is defined where Nb is the number of bins in P_1 . We remark that, with a slight abuse of notation, we indicated the angular coordinates of the rays at the target (line Nl) with t_j instead of $t_{\text{Nl},j}$ for every $j \in \{0, \dots, \text{Nb}\}$. The normalized approximated intensity $I_{\text{MC}}(t)$ is a piecewise constant function and its value over the j -th bin is the ratio between the number of rays that fall into that bin $\text{Nr}[t_{j-1}, t_j]$ and the total number of rays traced $\text{Nr}[-\pi/2, \pi/2]$. Hence, I_{MC} is defined by:

$$I_{\text{MC}}(t) = \frac{\text{Nr}[t_{j-1}, t_j]}{\text{Nr}[-\pi/2, \pi/2]} \quad \text{for } t \in [t_{j-1}, t_j]. \quad (3.2.10)$$

Furthermore, the output intensity is computed from the value of the intensity $I_{\text{MC}}(t_{j-1/2})$ along the direction $t_{j-1/2} = (t_{j-1} + t_j)/2$ for every bin $[t_{j-1}, t_j]_{j=1, \dots, \text{Nb}}$. The intensity $I_{\text{MC}}(t_{j-1/2})$ gives an estimate of the probability that a ray reaches the target with an angle in the j -th interval $[t_{j-1}, t_j]$ of the partitioning P_1 . This probability $P_{j,\Delta t}$ is given by:

$$P_{j,\Delta t} = \Pr(t_{j-1} \leq t < t_j) = \frac{\int_{t_{j-1}}^{t_j} I(t) dt}{\int_{-\pi/2}^{\pi/2} I(t) dt}, \quad (3.2.11)$$

where $I(t)$ is the output intensity (not normalized). Note that $\sum_{j=1}^{\text{Nb}} P_{j,\Delta t} = 1$. From the mean value theorem for the function $I(t)$ continuous in $[t_{j-1}, t_j]$, we know that there exists a value $t_k \in [t_{j-1}, t_j]$ for which the integral at the numerator of the previous equation can be written as:

$$\int_{t_{j-1}}^{t_j} I(t) dt = \Delta t I(t_k). \quad (3.2.12)$$

Hence, $P_{j,\Delta t}$ is proportional to the size $\Delta t = (t_{\text{Nb}} - t_0)/\text{Nb}$ of the intervals, i.e., inversely proportional to the number of bins Nb of the partitioning P_1 . Indicating with $\Phi = \int_{-\pi/2}^{\pi/2} I(t) dt$ the total flux (measured in lumen [lm]), the error between the intensity $I(t_{j-1/2})$ and the averaged MC intensity $\Phi I_{\text{MC}}(t_{j-1/2})/\Delta t$ is given by:

$$\begin{aligned} \left| I(t_{j-1/2}) - \frac{\Phi}{\Delta t} I_{\text{MC}}(t_{j-1/2}) \right| \leq \\ \left| I(t_{j-1/2}) - \frac{1}{\Delta t} \int_{t_{j-1}}^{t_j} I(t) dt \right| + \\ \frac{1}{\Delta t} \left| \int_{t_{j-1}}^{t_j} I(t) dt - \Phi I_{\text{MC}}(t_{j-1/2}) \right|. \end{aligned} \quad (3.2.13)$$

The first term of the right hand side of inequality (3.2.13) gives an estimate of how much the averaged intensity $\frac{1}{\Delta t} \int_{t_{j-1}}^{t_j} I(t) dt$ differs from the exact intensity $I(t_{j-1/2})$. This term is due to the discretization of the target and therefore it depends on the number of bins Nb considered. Substituting $I(t)$ with its Taylor expansion around

the point $t_{j-1/2}$ we obtain that this term is proportional to the square of the size of the bins, therefore the following equality holds:

$$\left| I(t_{j-1/2}) - \frac{1}{\Delta t} \int_{t_{j-1}}^{t_j} I(t) dt \right| = C_1 / \text{Nb}^2 \quad (3.2.14)$$

with $C_1 > 0$ a certain constant.

The second part of the right hand side of inequality (3.2.13) gives an estimate of the MC error and therefore it depends also on the number of rays traced. In order to show how this term decreases as a function of the number of rays traced, we define the random variable $X_j(t)$ as the variable that is equal to 1 if the ray with angular coordinate t is inside the interval $[t_{j-1}, t_j]$ and equal to 0 otherwise,

$$X_j(t) = \begin{cases} 1 & \text{if } t \in [t_{j-1}, t_j], \\ 0 & \text{otherwise.} \end{cases} \quad (3.2.15)$$

The Bernoulli trial X_j follows a binomial distribution $B(1, P_{j,\Delta t})$. Considering a sample of Nr rays, the variable $Y_j = \sum_{k=1}^{\text{Nr}} X_j(t_k)$ follows a binomial distribution $B(\text{Nr}, P_{j,\Delta t})$, where t_k is the angle that the k -th ray forms with the optical axis. Then, using the de Moivre-Laplace theorem, we conclude that the variable Y_j is approximated by a normal distribution with mean value $E[Y_j] = \text{Nr}P_{j,\Delta t}$ and variance $\sigma^2[Y_j] = \text{Nr}P_{j,\Delta t}(1 - P_{j,\Delta t})$ when a large number of rays is considered, see [15, 16]. Thus, the normalized intensity along the direction $t_{j-1/2}$ is given by:

$$I_{\text{MC}}(t_{j-1/2}) = \sum_{k=1}^{\text{Nr}} X_j(t_k) / \text{Nr}. \quad (3.2.16)$$

The mean value $E[I_{\text{MC}}(t_{j-1/2})] = P_{j,\Delta t}$ and the variance $\sigma^2[I_{\text{MC}}(t_{j-1/2})] = P_{j,\Delta t}(1 - P_{j,\Delta t}) / \text{Nr}$. Note that the standard deviation $\sigma_j := \sigma[I_{\text{MC}}(t_{j-1/2})]$ equals:

$$\sigma_j = \sqrt{P_{j,\Delta t}(1 - P_{j,\Delta t}) / \text{Nr}} = \frac{C_2}{\sqrt{\text{NbNr}}}, \quad (3.2.17)$$

for some $C_2 > 0$. σ_j can be used to give an estimate of the difference between the intensity $I_{\text{MC}}(t_{j-1/2})$ and its mean value $P_{j,\Delta t}$. Therefore, the second term of the right hand side of relation (3.2.13) becomes:

$$\begin{aligned} \frac{1}{\Delta t} \left| \int_{t_{j-1}}^{t_j} I(t) dt - \Phi I_{\text{MC}}(t_{j-1/2}) \right| &= \\ \frac{\Phi}{\Delta t} \left| P_{j,\Delta t} - I_{\text{MC}}(t_{j-1/2}) \right| &\propto \\ \frac{\Phi}{\Delta t} \sigma_j [I_{\text{MC}}(t_{j-1/2})] &= C_3 \frac{\text{Nb}}{\sqrt{\text{NbNr}}} = C_3 \sqrt{\frac{\text{Nb}}{\text{Nr}}}, \end{aligned} \quad (3.2.18)$$

for some $C_3 > 0$, where the approximation holds because σ_j gives a measure for the error between $I_{\text{MC}}(t_{j-1/2})$ and the probability $P_{j,\Delta t}$, [17]. The second equality follows from Eq. (3.2.17). The MC error over the j -th bin is estimated by:

$$\left| I(t_{j-1/2}) - \frac{\Phi}{\Delta t} I_{\text{MC}}(t_{j-1/2}) \right| = \frac{C_1}{\text{Nb}^2} + C_4 \sqrt{\frac{\text{Nb}}{\text{Nr}}}, \quad (3.2.19)$$

for $C_4 > 0$. Considering a fixed number of rays, we obtain that the minimal error is reached when $Nb \approx Nr^{1/5}$. Hence, if 10^{10} rays are considered the target has to be divided into 10^2 bins to minimize the MC error. This leads to computational efforts resulting in a very slow procedure.

We conclude this chapter implementing MC ray tracing for the two-faceted cup the profile of which is depicted in Fig. 3.1. Considering a set of $Nr = 10^3$ random rays at the source, we obtain the distribution on the (x, t) -plane shown in Fig. 3.2.

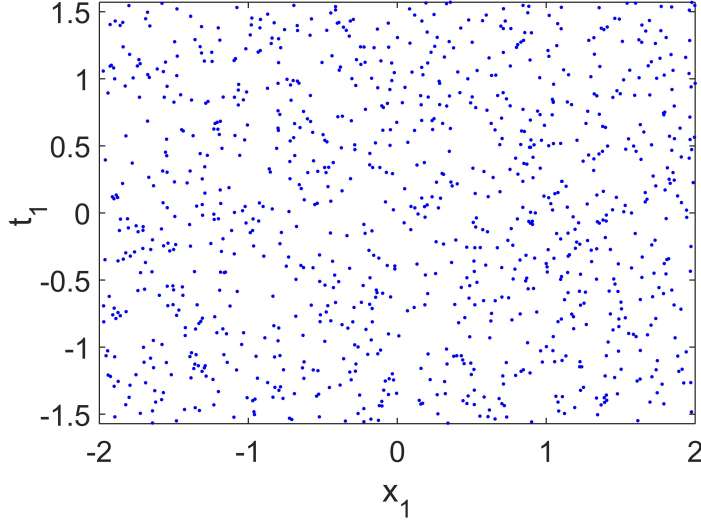


Figure 3.2: Rays at the source of the two-faceted cup with random position coordinate x and random angular coordinates t . 10^4 rays are depicted in this figure.

Then, every sample ray is traced inside the system using the ray tracing procedure. The target $T = [-b, b]$ is divided into $Nb = 100$ bins. Using Eq. (3.2.10), the normalized intensity I_{MC} is computed. I_{MC} is a piecewise constant function, therefore the averaged normalized intensity $\hat{I}_{MC}(t_{j-1/2}) = \frac{1}{\Delta t} I_{MC}(t_{j-1/2})$ is given considering the values that the intensity I_{MC} assumes on the middle point $(t_{j-1/2})_{j=0, \dots, Nb}$ of every bin of the partitioning P . The profile of \hat{I}_{MC} is depicted in Fig. 3.3 with the red line. The exact intensity (analytic intensity) is shown with the green line in the same figure. MC ray tracing has the advantages of being very easy to implement and it does not require too much regularity of the function that has to be approximate. Furthermore, the error convergence does not depend on the dimension of the domain in which the function is defined. On the other hand, MC method is time consuming as the error, for a fixed number of bins, has a speed of convergence of the order $O(1/\sqrt{Nr})$. Furthermore, MC ray tracing is a binning procedure, therefore the error depends also on the number of bins in which the target is divided. It is a statistical procedure and the error bound is only a "probabilistic" error as shown in Eq. (3.2.9). This means that, to calculate the value of the error, several simulations have to be repeated and the average of the errors obtained in every simulation has to be calculated.

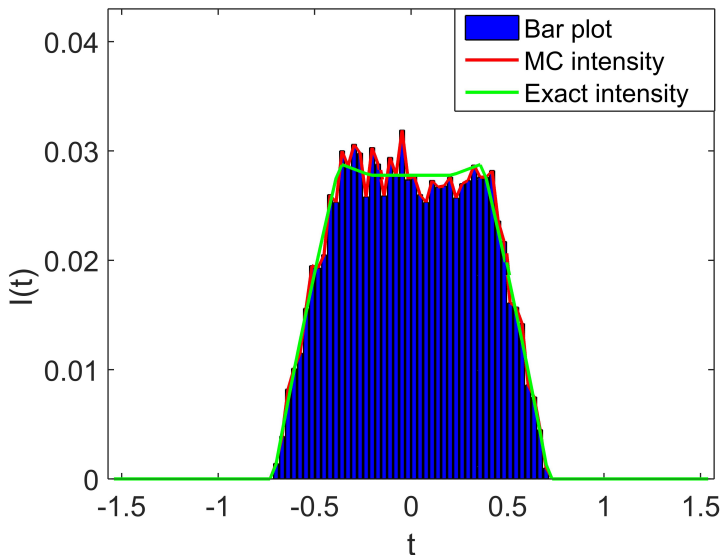


Figure 3.3: Comparison between the averaged normalized MC intensity and the normalized exact intensity.

MC method can be improved using a deterministic approach. Instead of considering random variables, the sample of rays can be defined in such a way that they have a regular distribution on the domain $D \subseteq \mathbb{R}^d$ of the function f of which we want to compute the integral. Methods based on this deterministic approach are called Quasi Monte Carlo (QMC) methods.

3.3 Quasi-Monte Carlo ray tracing

Quasi-Monte Carlo (QMC) methods were proposed for the first time in the 1950s in order to deal with the issues encountered in MC. Likewise MC methods, QMC procedures can be used to approximate the integral of a function. It is useful to restrict ourselves to intervals of the form $[\mathbf{a}, \mathbf{b}] \subseteq [0, 1]^d$ and introduce the concept of sequences uniformly distributed modulo one.

Definition 3.3.1. An infinite sequence $\{y_n\}_{n \in \mathbb{N}_0} \in [0, 1]^d$ is said to be *uniformly distributed modulo one* (or *equidistributed*), if for every interval $[\mathbf{a}, \mathbf{b}] \subseteq [0, 1]^d$ it holds

$$\lim_{N \rightarrow \infty} \frac{\text{card}(A([\mathbf{a}, \mathbf{b}], N))}{N} = \lambda_d([\mathbf{a}, \mathbf{b}]) \quad (3.3.1)$$

where $\text{card}(A([\mathbf{a}, \mathbf{b}], N))$ is the cardinality of the following set

$$A([\mathbf{a}, \mathbf{b}], N) = \{n \in \mathbb{N}_0 : 0 \leq n \leq N - 1 \text{ and } y_n \in [\mathbf{a}, \mathbf{b}]\}, \quad (3.3.2)$$

and $\lambda_d([\mathbf{a}, \mathbf{b}]) = \prod_{j=1}^d (b_j - a_j)$ is the d -dimensional Lebesgue measure of the interval $[\mathbf{a}, \mathbf{b}]$.

Given a sequence $\{\mathbf{y}_i\}_{i=1, \dots, N} \in [0, 1]^d$ uniformly distributed modulo one and a Riemann integrable function $f : [0, 1]^d \mapsto \mathbb{R}$, the integral of f can be approximate as the average of the values that f assumes on $\{\mathbf{y}_i\}$ for every $j = \{1, \dots, N\}$, that is:

$$\lim_{N \rightarrow \infty} \frac{1}{N} \sum_{i=1}^N f(\mathbf{y}_i) = \int_{[0, 1]^d} f(\mathbf{y}) d\mathbf{y}. \quad (3.3.3)$$

The idea of QMC methods is to generate the set of points in $[\mathbf{a}, \mathbf{b}]$ such that they are not random distributed but also not exactly uniformly distributed. To measure how much the distribution of these points differs from a uniform distribution, the concept of discrepancy was introduced. Intuitively, discrepancy measures how much the samples differ from a uniform distribution. Therefore, random sequences have a very high discrepancy, while uniform distributed sequences have zero discrepancy. The definition of discrepancy in more mathematical terms is provided below.

Definition 3.3.2. Given a set $\mathcal{S} = \{\mathbf{y}_1, \dots, \mathbf{y}_N\}$ of N points in $[0, 1]^d$. The discrepancy $D_N(\mathcal{S})$ of \mathcal{S} is defined as

$$D_N(\mathcal{S}) = \sup_{\mathbf{a}, \mathbf{b} \in [0, 1]^d} \left| \frac{\text{card}(A([\mathbf{a}, \mathbf{b}], N))}{N} - \lambda_d([\mathbf{a}, \mathbf{b}]) \right| \quad (3.3.4)$$

where $\lambda_d([\mathbf{a}, \mathbf{b}]) = \prod_{j=1}^d (b_j - a_j)$ is the d -dimensional Lebesgue measure of the interval $[\mathbf{a}, \mathbf{b}]$.

Often, it is enough to consider the discrepancy in the intervals $[0, \mathbf{a}] \subseteq [0, 1]^d$, in that case we talk about star discrepancy.

Definition 3.3.3. Let $\mathcal{S} = \{\mathbf{y}_1, \dots, \mathbf{y}_N\}$ be a set of N points in $[0, 1]^d$. The star discrepancy $D_N^*(\mathcal{S})$ of \mathcal{S} is defined as:

$$D_N^*(\mathcal{S}) = \sup_{\mathbf{a} \in [0, 1]^d} \left| \frac{\text{card}(A([0, \mathbf{a}], N))}{N} - \lambda_d([0, \mathbf{a}]) \right| \quad (3.3.5)$$

Sequences constructed such that the corresponding star discrepancy has an order of $O(\log(N)^d/N)$ are called *low-discrepancy sequences*, [13]. An important results shows that, using a low-discrepancy sequence $\{\mathbf{y}_i\}_{i=1, \dots, N}$, the absolute error of a QMC algorithm:

$$\text{err}(f, S_N) = \left| \int_{[0, 1]^d} f(\mathbf{y}) d\mathbf{y} - \frac{1}{N} \sum_{i=1}^N f(\mathbf{y}_i) \right| \quad (3.3.6)$$

can be bounded by the product of a term that depends on f and another term that depends on the discrepancy of the set $\{\mathbf{y}_i\}_{i=1, \dots, N}$. This is the result provided by the Koksma-Hlawka inequality which gives the following estimation of the error:

$$\left| \int_{[0, 1]^d} f(\mathbf{y}) d\mathbf{y} - \frac{1}{N} \sum_{i=1}^N f(\mathbf{y}_i) \right| \leq V(f) D_N^*(\mathcal{S}) \quad (3.3.7)$$

where $V(f)$ is the so-called variation function of f in the sense of Hardy-Krause (see [18, 19] for details). From the definition of low-discrepancy sequences and from the Koksma-Hlawka inequality we can state that:

$$\text{err}(f, S_N) < C \frac{\log(N)^d}{N}. \quad (3.3.8)$$

For small dimensions d , QMC performs much better than MC methods, while for large dimension d the factor $\log(N)$ could be very big. The convergence of QMC method can depend on which kind of low-discrepancy sequence is used. Indeed, there are many ways to generate such sequences. The most common QMC approach uses the so-called Sobol' sequence. The algorithm for generating Sobol' sequences is widely explained in the literature, (see for instance, [20]). In appendix A we give an overview of how these kind of sequences can be constructed.

Based on QMC methods, QMC ray tracing considers as position and angular coordinates of the rays at the source, the coordinates of the corresponding points of a low-discrepancy sequence. Therefore, to implement QMC ray tracing in two-dimensions we need to construct a low-discrepancy sequence in two-dimensions. Given, for instance, a Sobol' sequence $\{\mathbf{y}_i\}_{i=1,\dots,N}$ with $\mathbf{y}_i \in [0, 1]^2$ for every $i = 1, \dots, N$, the two dimensional QMC ray tracing consider the position coordinate of the i -th ray at the source equal to the first component of the i -th point \mathbf{y}_i of the Sobol' sequence $\{\mathbf{y}_i\}_{i=1,\dots,N}$ and, the direction coordinate of the i -th ray at the source equal to the second component of the i -th point \mathbf{y}_i of the same sequence. A set of $N_r = N$ rays with these initial coordinates is traced within the system and, once the target coordinates of all the rays traced are computed, the output intensity is calculated using the same approach used for MC ray tracing, see Eqs 3.2.10 and 3.2.13. The difference between MC and QMC ray tracing consists only on the choice of the initial ray set.

In Fig. 3.4 we show the distribution of the position and direction coordinates of the rays at the source of the two-faceted cup in Fig. 3.1. A set of 10^3 rays generated from a 2D Sobol sequence is considered, the coordinates (x_1, t_1) of every ray at the source are depicted with blue dots. We note that the rays have a regular distribution on the (x, t) -plane. We need to remark that, for the system in Fig. 3.1, $x_1 \in [-2, 2]$ and the angular coordinates $t_1 \in [-\pi/2, \pi/2]$. Since Sobol' sequences are defined inside intervals of the $[\mathbf{a}, \mathbf{b}] \subseteq [0, 1]^2$, we scaled the points of the sequence \mathbf{y}_i in order to take all the possible positions and directions that the rays can assume at the source.

Dividing the target into $N_b = 100$ bins, we computed the target intensity. In Fig. 3.5 we show the profile of the output intensity at the target of the two-faceted cup computed using QMC ray tracing with 10^4 rays. The QMC intensity is depicted with the red line. It is compared to the analytic intensity shown in the same figure with the green dotted line. A comparison between Fig. 3.3 and 3.5 shows that for the two-faceted cup and for a set of $N_r = 10^4$ rays, QMC ray tracing performs much better than MC ray tracing. The results shown for a simple optical system are indeed consistent with what we expected from the theoretical analysis. Although QMC ray tracing is an improvement of MC ray tracing for small dimensions, it has two main disadvantages. First, its convergence is strongly related with the dimension in which it is implemented. Second, likewise MC ray tracing, QMC ray tracing is a binning procedure, therefore the error still depends on the number of bins in which the target is divided and only the averaged value of the intensity over every bin is provided.

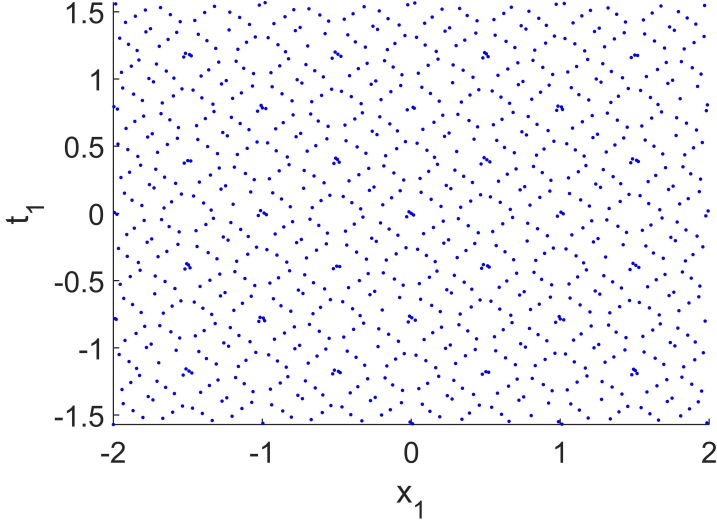


Figure 3.4: 10^3 rays at the source of the two-faceted cup with position x_1 and angular t_1 coordinates with a regular distributions. They are distributed as the points of a Sobol' sequence in two-dimensions.

From the results provided in this chapter we can conclude that the choice of the initial ray set can make a big impact on the performance of the ray tracing procedure. Based on the idea of taking a smart choice of the initial ray set, we develop a new ray tracing method which is based on phase space. The phase space (PS) concept will be introduced in the next chapter. The new ray tracing method employs the PS of the source and the target of the optical systems. We will show in this thesis that phase space ray tracing allows to trace only few rays inside the system to obtain the desired accuracy of the target intensity.

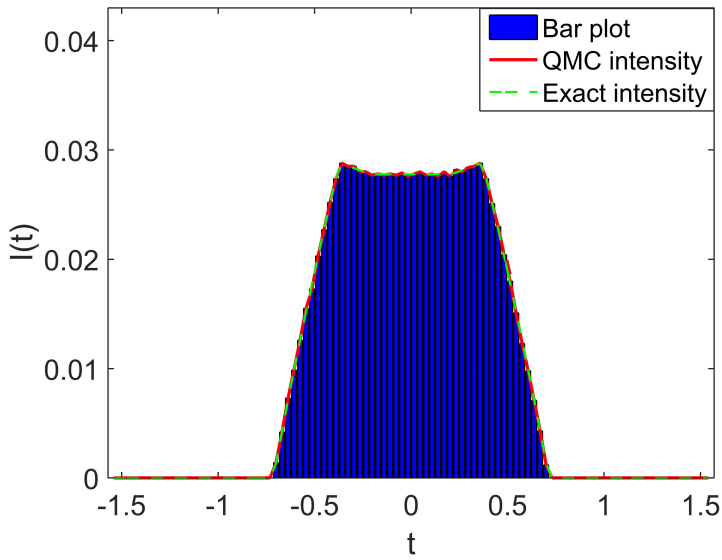


Figure 3.5: QMC intensity for the two-faceted cup obtained tracing $N_r = 10^4$ rays and dividing the target into $N_b = 100$ bins.

Chapter 4

Ray tracing on phase space

4.1 The edge-ray principle

4.2 Phase space ray tracing

4.3 Comparison between MC QMC and PS ray tracing

Chapter 5

Two different approaches to compute the boundaries in target phase space

5.1 The α -shapes approach

Given a finite set \mathcal{S} of points we want to determine the shape formed by these points. α -shapes are geometrical objects which give us a good approximation of the shape of a given point set \mathcal{S} . Before giving a formal definition we explain an intuitive interpretation of α -shapes. As mentioned in [21] we can think of an α -shape as a mass of ice-cream with several chocolate pieces. The mass making up the space \mathbb{R}^3 and the chocolate pieces are the point set \mathcal{S} . Then the aim is to find the shape formed by the chocolate pieces. We can use a spoon with a spherical shape and carve out all parts of the ice-cream without removing the chocolate pieces. We will obtain a shape formed by arcs and points (see figure below for the two-dimensional case). Straightening the arcs to triangles and line segments we have an intuitive description of what is called the α -shape of \mathcal{S} . In our example, the parameter α determines the radius of the carving spoon. If α is equal to 0 the shape degenerates to the point set \mathcal{S} . On the other hand, when $\alpha \rightarrow \infty$ the α -shape is simply the convex hull. More precisely the process is summarized as follows. Given a point cloud \mathcal{S} we start with a triangulation of it (a possible choice could be the Delaunay triangulation described in the next section). For each triangle we calculate the radius of the circumcircle. If the radius is larger than α the triangle is removed from the shape. The rule of the parameter α is highly significant in this procedure. Hence we have to choose it in such a way to get a better approximation. The choice of the parameter α is closely related to the radius of the circumcircles. A possible strategy is to find the radius of the greater empty circumcircle. Thus α is related to the density of the points. In particular we have:

$$\alpha = C \frac{1}{\Delta}, \quad (5.1.1)$$

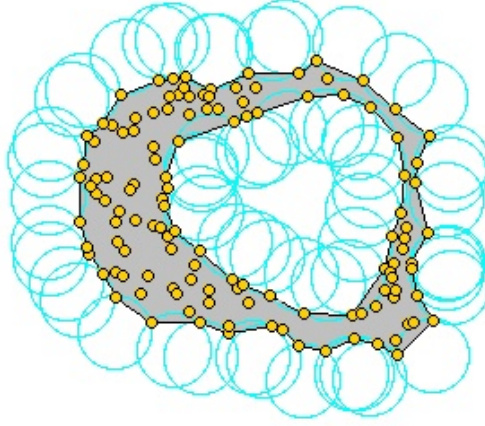


Figure 5.1: Construction of α -shape given a set of points in \mathbb{R}^2 .

with C a constant that can be determined by a simulation and Δ the density of the point set \mathcal{S} defined as:

$$\Delta = \frac{N}{\text{surface area}}, \quad (5.1.2)$$

where N is the number of points in \mathcal{S} and the surface area is the area inside the boundaries of the region formed by the points cloud. Hence Δ is a constant. As mentioned above to find the α -shape of a point cloud we need a triangulation and a possible choice could be the Delaunay triangulation. As explained in [22] we can see a Delaunay triangulation as the dual of a Voronoi diagram. Let us define a Voronoi diagram in a metric space.

Definition 5.1.1. Let X be a space endowed with a distance d and $\mathcal{S} = \{S_1, \dots, S_n\}$ a set formed by subsets of X . The Voronoi cell R_k associated with the set S_k where $k \in \{1, \dots, n\}$ is defined as follows:

$$R_k = \{\mathbf{x} \in X \mid d(\mathbf{x}, S_k) \leq d(\mathbf{x}, S_j) \quad \forall j \neq k\}, \quad (5.1.3)$$

where $d(x, A) = \inf\{d(x, a) \mid a \in A\}$. The Voronoi diagram is defined as the tuple of the cells $(R_k)_{k \in \{1, \dots, n\}}$ that are assumed to be disjoint.

The simplest case that we can have is the two-dimensional case that is the case where $X = \mathbb{R}^2$. The tuple $\mathcal{S} = \{1, \dots, n\} \subset \mathbb{R}^2$ is now a set of points. The Voronoi diagram of \mathcal{S} is a subsection of \mathbb{R}^2 such that every other region around a point $p \in \mathcal{S}$ contains all points that are closer to p than to every point in \mathcal{S} . A triangulation of the point set \mathcal{S} is a set of edges \mathcal{E} whose extremes are points of \mathcal{S} such that the faces of each triangle are bounded by three edges and any edge that is not in \mathcal{E} intersects one of the existing edges. The Delaunay triangulation is the dual graph of the Voronoi diagram: it consists of vertices (the points in \mathcal{S}) and it has an edge between two vertices if the two corresponding faces share an edge.

The Delaunay triangulation triangulates the convex hull of the point set \mathcal{S} . Instead,

the α -shape of a point set is formed only by the triangles (taken from the Delaunay triangulation) that satisfy the " α -test" and therefore is a suitable method to reconstruct the surface formed by a point cloud. Even if α -shapes are a powerful tool to reconstruct surfaces, some simulations show that there exist surfaces that are not described well by α -shapes. Indeed for some particular surface there exist no value of α that includes all desired triangles and deletes all undesired triangles. For instance, since the parameter α depends on the density of the point cloud, is intuitively clear that using α -shapes for a non-uniform points set we won't get a good approximation of the surface. Furthermore, the α -shape method doesn't work well when there is a sharp turn or a joint. In this case α -shapes often give a "webbed-foot" appearance at such joints since they improperly connect the adjacent surfaces. Hence a generalization of "classical" α -shapes is required. In the next section a method to solve the "density problem" for two separated and close objects is described. In [23] Teichmann and Capps present "Density-scaled α -shapes". The first step of this method is to make a triangulation of the point cloud. Then the key idea is to compute somehow the point-density of each point and use this to get an approximation of the point density of a triangle. In this way one can reduce the α -value in areas where the triangle's point density (see equation 5.1.6 for the definition) is higher than average in such a way that is possible to obtain a finer level of detail for areas that have an higher density. More precisely, each point $\mathbf{p} \in \mathcal{S}$ has a local point density defined as

$$\delta(\mathbf{p}) = \sum_{\mathbf{q} \in \mathcal{S}} \left(1 - \frac{d(\mathbf{q}, \mathbf{p})}{\lambda}\right) \quad \forall \mathbf{q} \text{ such that } d(\mathbf{p}, \mathbf{q}) < \lambda, \quad (5.1.4)$$

where λ is the constant radius of the local neighborhood and $d(\mathbf{x}, \mathbf{y})$ is the Euclidean distance. When local density is larger than the average, that is when

$$\delta(\mathbf{p}) > \frac{1}{|\mathcal{S}|} \sum_{\mathbf{q} \in \mathcal{S}} \delta(\mathbf{q}) \quad (5.1.5)$$

we know some properties about the region surrounding \mathbf{p} . For instance, if the point set is uniformly distributed then it is possible to find areas with a high-density in the case where there are two closely separated surfaces. In point sets of non-uniform distribution, high densities are found when the surface presents a joint discontinuity. The algorithm developed by Teichmann and Capps is structured as follow. After computing density information for each point they make a triangulation of the point set. Then they calculate the average density $\delta(t)$ for each triangle Δ_{abc} defined as:

$$\delta(t) = \frac{\delta(a) + \delta(b) + \delta(c)}{3\mu}, \quad (5.1.6)$$

where μ is the global average density of the entire point set \mathcal{S} . If $\delta(t)$ is greater than 1 the density of the point cloud is higher. Hence is necessary to define another value of α :

$$\alpha' = \frac{\alpha}{\delta(t)^\sigma} \quad (5.1.7)$$

where σ is a value that is adjusted by the user. If δ is less than 1 the α -value is not modified. In this way it is possible to have a finer precision on the shape formed by the point set where the density is higher than the average density. Hence it is possible to distinguish two separated objects with different density.

- 5.2 The two-faceted cup
- 5.3 Results for a TIR collimator
- 5.4 The triangulation refinement approach
- 5.5 The two-faceted cup
- 5.6 Results for a TIR collimator
- 5.7 Results for a Parabolic reflector
- 5.8 Results for the Compound Parabolic Concentrator (CPC)

Chapter 6

The inverse ray mapping method: analytic approach

6.1 Explanation of the method

6.2 The two-faceted cup

6.3 Results for the two-faceted cup

6.4 Results for the multi-faceted cup

6.5 Discussions

Chapter 7

Extended ray mapping method to systems with Fresnel reflection

Chapter 8

Discussion and conclusions

Appendix A

Implementation of Sobol' sequences

A.1 Van der Corput sequences

In the following we show a particular construction of a low-discrepancy sequence for $d = 1$ that was introduced the first time by Van der Corput in 1935. This kind of sequences, called *van der Corput* sequences, are particular interesting not only because they give an intuition of how to construct low discrepancy sequences but also because many other kind of sequences in higher dimensions are based on this one-dimensional case. Before introducing these sequences we need to give the concept of radical inverse function. Let $b \geq 2$ be an integer base. Any natural number $n \in \mathbb{N}_0$ can be decomposed in base b as follows:

$$n = \sum_{i=0}^{\infty} d_i b^i \quad (\text{A.1.1})$$

where $d_i \in \{0, 1, \dots, b-1\}$ are the digit numbers. The radical inverse function $\phi_b : \mathbb{N}_0 \mapsto [0, 1)$ in base b is defined as:

$$\phi_b(n) = \sum_{i=1}^{\infty} \frac{d_{i-1}}{b^i}. \quad (\text{A.1.2})$$

As an example we provide in the following the radical inverse function $\phi_b(5)$ in base $b = 2$. The digit expansion in base b of $n = 5$ is:

$$5 = 1 \cdot 2^0 + 1 \cdot 2^2. \quad (\text{A.1.3})$$

Therefore, $d_0 = 1, d_1 = 0$ and $d_2 = 1$. The radical inverse function $\phi_2(5)$ is:

$$\phi_2(5) = \frac{1}{2} + \frac{1}{8} = \frac{5}{8}. \quad (\text{A.1.4})$$

Definition A.1.1. The Van der Corput sequence in base b is defined as $\{\phi_b(n)\}_{n \in \mathbb{N}_0}$.

For example, suppose we have the finite sequence of numbers $n \in \{0, 1, \dots, 8\}$ the corresponding Van der Corput sequence $\{\phi_b(n)\}_{n \in \{0, 1, \dots, 8\}}$ in base $b = 2$ is:

$$\{\phi_2(n)\}_{n \in \{0, 1, \dots, 8\}} = \left\{0, \frac{1}{2}, \frac{1}{4}, \frac{3}{4}, \frac{1}{8}, \frac{5}{8}, \frac{3}{8}, \frac{7}{8}, \frac{1}{16}\right\}. \quad (\text{A.1.5})$$

It can be proved that the Van der Corput sequence in base b is uniformly distributed modulo one, [14]. The van der Corput sequence has been extended to higher dimensions. The most common QMC approach uses Sobol sequence which is one an extended Van der Corput sequence in base $b = 2$ to $d \geq 2$.

A.2 Sobol' sequences

Acknowledgments

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