

## Sphere Intersection

We know that a sphere of radius  $r$  is the set of all points a distance  $r$  from sphere's center. In vector notation, this is

$$(\mathbf{P} - \mathbf{O}_S) \cdot (\mathbf{P} - \mathbf{O}_S) = r^2$$

where  $\mathbf{P}$  is a vector touching a point on the sphere and  $\mathbf{O}_S$  is the vector describing the sphere's center.

We describe a ray via

$$\mathbf{P}(t) = \mathbf{O}_R + \mathbf{D}t$$

where  $t \in \mathbb{R}$ . Substituting this into the implicit equation for our sphere yields

$$\begin{aligned} r^2 &= (\mathbf{P}(t) - \mathbf{O}_S) \cdot (\mathbf{P}(t) - \mathbf{O}_S), \\ &= (\mathbf{O}_R + \mathbf{D}t - \mathbf{O}_S) \cdot (\mathbf{O}_R + \mathbf{D}t - \mathbf{O}_S), \\ &= (\mathbf{D}t + (\mathbf{O}_R - \mathbf{O}_S)) \cdot (\mathbf{D}t + (\mathbf{O}_R - \mathbf{O}_S)), \\ &= (\mathbf{D} \cdot \mathbf{D})t^2 + 2\mathbf{D} \cdot (\mathbf{O}_R - \mathbf{O}_S)t + (\mathbf{O}_R - \mathbf{O}_S) \cdot (\mathbf{O}_R - \mathbf{O}_S). \end{aligned}$$

We can define  $\mathbf{U} = \mathbf{O}_R - \mathbf{O}_S$ , collect everything on one side, and group terms by orders of  $t$  to arrive at

$$(\mathbf{D} \cdot \mathbf{D})t^2 + (2\mathbf{D} \cdot \mathbf{U})t + (\mathbf{U} \cdot \mathbf{U} - r^2) = 0$$

which is a simple quadratic equation in  $t$ . Its solution is

$$t = \frac{-2\mathbf{D} \cdot \mathbf{U} \pm \sqrt{(2\mathbf{D} \cdot \mathbf{U})^2 - 4(\mathbf{D} \cdot \mathbf{D})(\mathbf{U} \cdot \mathbf{U} - r^2)}}{2\mathbf{D} \cdot \mathbf{D}}.$$

or, removing a common factor of 2,

$$t = \frac{-\mathbf{D} \cdot \mathbf{U} \pm \sqrt{(\mathbf{D} \cdot \mathbf{U})^2 - (\mathbf{D} \cdot \mathbf{D})(\mathbf{U} \cdot \mathbf{U} - r^2)}}{\mathbf{D} \cdot \mathbf{D}}.$$

Evidently, the ray intersects our sphere when the discriminant of this equation is greater than or equal to 0, i.e.

$$(\mathbf{D} \cdot \mathbf{U})^2 - (\mathbf{D} \cdot \mathbf{D})(\mathbf{U} \cdot \mathbf{U} - r^2) \geq 0$$

If this condition is met, the value of  $t$  corresponding to the first intersection is the smallest value that is still greater than 0.

## Importance Sampling

For a function  $f(x)$  of a random variable  $X$ , its expected value is given by

$$\mathbb{E}_p[f(X)] = \int_{\Omega} f(x)p_X(x) \, dx$$

where  $p_X(x)$  is the probability density function (PDF) of  $X$ . A statistical way to evaluate the expected value is given by the sample mean, defined by

$$\mathbb{E}_p[f(X)] = \lim_{N \rightarrow \infty} \frac{1}{N} \sum_{i=1}^N f(x_i)$$

where each  $x_i$  is a realization of  $X$  drawn from the distribution  $p_X(x)$ . We often want to evaluate integrals of the form  $\int_{\Omega} f(x) \, dx$ , which we can do by taking the expected value of  $f(x)/p_X(x)$  instead of just  $f(x)$ . Truncating the sample mean at a finite  $N$ , we see the correspondence

$$\int_{\Omega} f(x) \, dx \approx \frac{1}{N} \sum_{i=1}^N \frac{f(x_i)}{p_X(x_i)}$$

If the samples we use are not a good fit to  $f$ , we run the risk of adding up a lot of terms that contribute very little to the final result. We can work around this by including a simple multiplicative factor of 1 in the expected value in the form of  $q_X(x)/q_X(x)$ , where  $q_X(x)$  is a PDF that better matches  $f$ . Specifically, we see

$$\mathbb{E}_p \left[ \frac{f(X)}{p_X(X)} \frac{q_X(X)}{q_X(X)} \right] = \int_{\Omega} \frac{f(x)}{p_X(x)} \frac{q_X(x)}{q_X(x)} p_X(x) dx = \int_{\Omega} \frac{f(x)}{q_X(x)} q_X(x) dx = \mathbb{E}_q \left[ \frac{f(X)}{q_X(X)} \right]$$

which gives us the correspondence

$$\int_{\Omega} f(x) dx \approx \frac{1}{N} \sum_{i=1}^N \frac{f(x_i)}{q_X(x_i)}$$

where, in this case, each  $x_i$  is sampled from the distribution defined by  $q_X(x)$ . There are, of course, restrictions on the possible choices of  $q_X(x)$ , but they are surprisingly soft: we simply require that  $q_X(x)$  be nonzero wherever  $f(x)p_X(x)$  is nonzero.

In path tracing, we use this to estimate the indirect lighting present in the scene. Specifically, we take  $f(x)$  to be the integrand of the rendering equation and  $N$  to be 1,

$$\int_{\Omega} f_r(\mathbf{x}, \hat{\omega}_i, \hat{\omega}_o, \lambda, t) L_i(\mathbf{x}, \hat{\omega}_i, \lambda, t) |\hat{\omega}_i \cdot \mathbf{n}| d\hat{\omega}_i \approx \frac{f_r(\mathbf{x}, \hat{\omega}_i, \hat{\omega}_o, \lambda, t) L_i(\mathbf{x}, \hat{\omega}_i, \lambda, t) |\hat{\omega}_i \cdot \mathbf{n}|}{p_{\omega}(\hat{\omega}_i)}$$

## Lambertian Reflectance

Recall the rendering equation,

$$L_o(\mathbf{x}, \hat{\omega}_o, \lambda, t) = L_e(\mathbf{x}, \hat{\omega}_o, \lambda, t) + \int_{\Omega} f_r(\mathbf{x}, \hat{\omega}_i, \hat{\omega}_o, \lambda, t) L_i(\mathbf{x}, \hat{\omega}_i, \lambda, t) |\hat{\omega}_i \cdot \mathbf{n}| d\hat{\omega}_i$$

where  $f_r(\mathbf{x}, \hat{\omega}_i, \hat{\omega}_o, \lambda, t)$  describes the bidirectional reflectance distribution function (BRDF) of the material under consideration and  $\Omega$  is the hemisphere centered at the geometry's normal vector.

A diffuse, Lambertian material has an equal chance to scatter light in all directions. This implies that  $f_r(\mathbf{x}, \hat{\omega}_i, \hat{\omega}_o, \lambda, t) = \rho_0$  where  $\rho_0$  is a constant. Because of this, the direction of the incoming radiance does not matter and we may take  $L_i(\mathbf{x}, \hat{\omega}_i, \lambda, t)$  outside of the integral. Assuming all light is reflected imposes the additional requirement that

$$\int_{\Omega} \rho_0 |\hat{\omega}_i \cdot \mathbf{n}| d\hat{\omega}_i = \rho_0 \int_{\Omega} \cos \theta d\hat{\omega}_i = 1$$

where  $\theta$  is the angle between  $\hat{\omega}_i$  and the normal. On the unit hemisphere, with  $\phi$  describing the azimuthal angle and  $\theta$  the altitude, this requirement becomes

$$\begin{aligned} \rho_0 \int_0^{2\pi} \int_0^{\pi/2} \cos \theta \sin \theta d\theta d\phi &= \frac{\rho_0}{2} \int_0^{2\pi} \int_0^{\pi/2} \sin(2\theta) d\theta d\phi \\ &= \frac{\rho_0}{4} \int_0^{2\pi} -\cos(2\theta) \Big|_0^{\pi/2} d\phi \\ &= \frac{\rho_0}{2} \int_0^{2\pi} d\phi \\ &= \rho_0 \pi \\ &= 1 \end{aligned}$$

i.e. max reflectance occurs when  $\rho_0 = 1/\pi$ . If we allow this to vary by wavelength via  $f_r(\mathbf{x}, \hat{\omega}_i, \hat{\omega}_o, \lambda, t) = \rho(\lambda)/\pi$  with  $\rho(\lambda) \in [0, 1]$ , the indirect part of the path tracing equation becomes

$$\frac{\rho(\lambda) L_i(\mathbf{x}, \hat{\omega}_i, \lambda, t) |\hat{\omega}_i \cdot \mathbf{n}|}{\pi p_{\omega}(\hat{\omega}_i)}$$

If we sample directions uniformly in the unit hemisphere, for example,  $p_{\omega}(\hat{\omega}_i) = 1/2\pi$  and the above simplifies to

$$2\rho(\lambda) L_i(\mathbf{x}, \hat{\omega}_i, \lambda, t) |\hat{\omega}_i \cdot \mathbf{n}|$$

## Uniformly Sampling the Unit Sphere (and Unit Hemisphere)

The easiest way to generate uniformly spaced points on the unit sphere is to build a cumulative distribution function (CDF), then use inverse transform sampling. To put it another way, suppose we want to generate samples from a PDF  $p_X(x) = dF_X(x)/dx$  but only have access to  $U$ , a random variable uniformly distributed between 0 and 1. An obvious way to map from this domain to the range of  $X$  is to define  $Y = F_X^{-1}(U)$ . The CDF of this new random variable is

$$F_Y(y) = \Pr(Y \leq y) = \Pr(F_X^{-1}(U) \leq y) = \Pr(U \leq F_X(y)) = F_X(y),$$

where we have used the fact that all CDFs are monotonic and that  $\Pr(U \leq u) = u$  for a random variable uniformly distributed between 0 and 1. Given the uniqueness of the derivative, we can be confident that  $Y$  has the same distribution as  $X$ .

Moving on to the problem at hand, the PDF of the unit sphere with regard to solid angle  $\Omega$  must be

$$p_\Omega(\Omega) = \frac{1}{4\pi}$$

We can, of course, write this in  $\phi$ - $\theta$  space as

$$p_\Omega(\Omega) d\Omega = p_{\theta,\phi}(\theta, \phi) d\theta d\phi$$

where, upon identifying  $d\Omega = \sin \theta d\theta d\phi$ , we find

$$p_{\theta,\phi}(\theta, \phi) = \frac{\sin \theta}{4\pi}$$

We can integrate over each variable to obtain the PDF of the other,

$$p_\theta(\theta) = \int_0^{2\pi} \frac{\sin \theta}{4\pi} d\phi = \frac{\sin \theta}{2}$$

$$p_\phi(\phi) = \int_0^\pi \frac{\sin \theta}{4\pi} d\theta = \frac{1}{2\pi}$$

which can then be used to find their respective CDFs

$$F_\theta(\theta) = \int_0^\theta \frac{\sin \theta'}{2} d\theta' = \frac{1 - \cos \theta}{2}$$

$$F_\phi(\phi) = \int_0^\phi \frac{d\phi'}{2\pi} = \frac{\phi}{2\pi}$$

whose inverse functions are

$$F_\theta^{-1}(u) = \cos^{-1}(1 - 2u) \equiv \theta$$

$$F_\phi^{-1}(v) = 2\pi v \equiv \phi$$

We can substitute these into the conversions from Cartesian to spherical coordinates to get

$$x = \sin \theta \cos \phi = \sqrt{1 - z^2} \cos(2\pi v)$$

$$y = \sin \theta \sin \phi = \sqrt{1 - z^2} \sin(2\pi v)$$

$$z = \cos \theta = 1 - 2u$$

Another method for uniformly sampling the unit sphere is to draw random values from the unit normal distribution for each component of a vector. One then simply needs to normalize the resulting vector. This works because the combined Gaussian PDF is rotationally invariant, and thus must correspond to a uniform distribution on the sphere.

Yet another method is the so-called ‘rejection’ method. In this, we draw values for the vector’s components from a random variable uniformly distributed between  $-1$  and  $1$ . If the vector lies within the unit ball, we normalize it. Otherwise, we repeat the process.

After we have setup a method to sample the unit sphere, sampling the unit hemisphere is just as easy. We simply dot the sampled vector with the desired normal vector: if this result is less than 0, we negate the sampled vector.

## Uniformly Sampling the Unit Disc

We follow a similar strategy to above. Over the unit disc, a uniform sampling strategy must yield

$$p_A(A) = \frac{1}{\pi}.$$

Identifying  $dA = r dr d\phi$  and equating the above PDF with the equivalent PDF in  $r$ - $\phi$  spaces gives

$$p_A(A) dA = \frac{r}{\pi} dr d\phi = p_{r,\phi}(r, \phi) dr d\phi,$$

i.e.  $p_{r,\phi}(r, \phi) = r/\pi$ . The PDF of each variable is

$$p_r(r) = \int_0^{2\pi} \frac{r}{\pi} d\phi = 2r$$

$$p_\phi(\phi) = \int_0^1 \frac{r}{\pi} dr = \frac{1}{2\pi}$$

and so the corresponding CDFs are

$$F_r(r) = \int_0^r 2r' dr' = r^2$$

$$F_\phi(\phi) = \int_0^\phi \frac{d\phi'}{2\pi} = \frac{\phi}{2\pi}$$

and so the inverse CDFs are

$$F_r^{-1}(u) = \sqrt{u} \equiv r$$

$$F_\phi^{-1}(v) = 2\pi v \equiv \phi$$

In Cartesian coordinates, then, a randomly sampled point on the unit disc is given by

$$x = r \cos \phi = \sqrt{u} \cos(2\pi v)$$

$$y = r \sin \phi = \sqrt{u} \sin(2\pi v)$$

## Cosine Weighted Sampling of the Unit Hemisphere

Looking back to the section on importance sampling, we see that sampling directions from a cosine weighted distribution has the advantage of removing the  $|\hat{\omega}_i \cdot \mathbf{n}|$  term from the rendering equation. Let's figure out how to do this.

A cosine weighted distribution over the unit hemisphere should obey

$$p_\Omega(\Omega) d\Omega = A \cos \theta d\Omega = A \cos \theta \sin \theta d\theta d\phi$$

where  $A$  is a normalization constant. We can find it by integrating over the unit hemisphere,

$$\begin{aligned} \int_\Omega p_\Omega(\Omega) d\Omega &= A \int_0^{2\pi} \int_0^{\pi/2} \cos \theta \sin \theta d\theta d\phi \\ &= \frac{A}{2} \int_0^{2\pi} \int_0^{\pi/2} \sin(2\theta) d\theta d\phi \\ &= -\frac{A}{2} \int_0^{2\pi} \left. \frac{\cos(2\theta)}{2} \right|_0^{\pi/2} d\phi \\ &= \frac{A}{2} \int_0^{2\pi} d\phi \\ &= A\pi \\ &= 1 \end{aligned}$$

i.e.  $A = 1/\pi$ . Using this, we may rewrite the PDF as

$$p_{\theta,\phi}(\theta, \phi) = \frac{\cos \theta \sin \theta}{\pi} = \frac{\sin(2\theta)}{2\pi}$$

The individual PDFs are

$$\begin{aligned} p_{\theta}(\theta) &= \int_0^{2\pi} \frac{\sin(2\theta)}{2\pi} d\phi = \sin(2\theta) \\ p_{\phi}(\phi) &= \int_0^{\pi/2} \frac{\sin(2\theta)}{2\pi} d\theta = -\frac{\cos(2\theta)}{4\pi} \Big|_0^{\pi/2} = \frac{1}{2\pi} \end{aligned}$$

and their respective CDFs are

$$\begin{aligned} F_{\theta}(\theta) &= \int_0^{\theta} \sin(2\theta') d\theta' = \frac{1 - \cos(2\theta)}{2} \\ F_{\phi}(\phi) &= \int_0^{\phi} \frac{d\phi'}{2\pi} = \frac{\phi}{2\pi} \end{aligned}$$

The inverse CDFs are

$$\begin{aligned} F_{\theta}^{-1}(u) &= \frac{\cos^{-1}(1 - 2u)}{2} \equiv \theta \\ F_{\phi}^{-1}(v) &= 2\pi v \equiv \phi \end{aligned}$$

Before expressing this in Cartesian coordinates, it is useful have the half-angle identities on hand

$$\cos\left(\frac{x}{2}\right) = \pm \sqrt{\frac{1 + \cos x}{2}} \quad \sin\left(\frac{x}{2}\right) = \pm \sqrt{\frac{1 - \cos x}{2}}$$

Using the above, we see

$$\cos \theta = \sqrt{1 - u} \quad \sin \theta = \sqrt{u}$$

where we have chosen the positive square root in each case so as to agree  $\cos \theta$  and  $\sin \theta$  when  $\theta \in [0, \frac{\pi}{2}]$ . Altogether this gives us

$$\begin{aligned} x &= \sin \theta \cos \phi = \sqrt{u} \cos(2\pi v) \\ y &= \sin \theta \sin \phi = \sqrt{u} \sin(2\pi v) \\ z &= \cos \theta = \sqrt{1 - u} \end{aligned}$$

Comparing this to our strategy for randomly sampling the unit disc reveals that this is simply a projection of the former onto the unit hemisphere! Using this sampling method gives a particularly nice form to the integrand in the rendering equation in the case of a Lambertian surface,

$$\rho(\lambda) L_i(\mathbf{x}, \hat{\omega}_i, \lambda, t)$$

## Lack of Cosine Term in Specular Reflection

Specular reflection requires that the integral term in the rendering equation,

$$\int_{\Omega} f_r(\mathbf{x}, \hat{\omega}_i, \hat{\omega}_o, \lambda, t) L_i(\mathbf{x}, \hat{\omega}_i, \lambda, t) |\hat{\omega}_i \cdot \mathbf{n}| d\hat{\omega}_i$$

conserve energy. In this particular case, the we know that  $f_r(\mathbf{x}, \hat{\omega}_i, \hat{\omega}_o, \lambda, t) \propto \delta(\mathbf{n} - (\hat{\omega}_i + \hat{\omega}_o)/2)$ , and so, to conserve energy, we must have

$$\int_{\Omega} A \delta(\mathbf{n} - (\hat{\omega}_i + \hat{\omega}_o)/2) L_i(\mathbf{x}, \hat{\omega}_i, \lambda, t) |\hat{\omega}_i \cdot \mathbf{n}| d\hat{\omega}_i = L_i(\mathbf{x}, 2\mathbf{n} - \hat{\omega}_o, \lambda, t)$$

where  $A$  is a normalization constant. We see that the distribution must be

$$f_r(\mathbf{x}, \hat{\omega}_i, \hat{\omega}_o, \lambda, t) = \frac{\delta(\mathbf{n} - (\hat{\omega}_i + \hat{\omega}_o)/2)}{|\hat{\omega}_i \cdot \mathbf{n}|}$$

and hence  $|\hat{\omega}_i \cdot \mathbf{n}|$  does not appear in implementations of specular reflection (or indeed any delta function distribution).

## Russian Roulette

Suppose we have an estimator  $F$  that we replace with an estimator  $F'$

$$F' = \begin{cases} \frac{F}{p} & u \leq p \\ 0 & u > p \end{cases}$$

where  $u$  is a uniform random value between 0 and 1 and  $p$  is some valid probability. In words,  $F'$  returns  $F/p$  with chance  $p$  and 0 with chance  $(1-p)$ . This new estimator has the same expected value as  $F$ , because

$$\mathbb{E}[F'] = p \cdot \frac{\mathbb{E}[F]}{p} + (1-p) \cdot 0 = \mathbb{E}[F]$$

If we apply this to the indirect lighting integrand in our path tracer, we see that it allows the termination of a loop that would otherwise need to continue until hitting a light source—a process that could easily have no end. This has the advantage of allowing us to run our program in a finite time without incurring bias.

## Reflection and Refraction