1. Consider a hemi-sphere specified by the *spherical coordinate system* using parameters  $\theta$  and  $\phi$ , where  $\theta \in [0, \pi/2]$  and  $\phi \in [0, 2\pi)$ . Using two uniform random variables  $\xi_1, \xi_2 \in U(0, 1)$ , derive the mapping f, g

$$\begin{cases} \theta = f(\xi_1) \\ \phi = g(\xi_2), \end{cases}$$

based on which the random variables  $\theta$  and  $\phi$  can generate *cosine-weighted samples* on the surface of the hemisphere, i.e.,  $p(\omega) = C \cdot \cos(\theta)$  where  $C \in \mathbb{R}$  is a constant which should be determined by calculation.

**Answer:** Since  $\sin \theta \cdot p(\omega) = p(\theta, \phi)$ , then  $p(\theta, \phi) = C \sin \theta \cos \theta$ . To sample  $p(\theta, \phi)$ , we first generate a uniform sample on  $\phi$ , obtaining  $\phi = 2\pi \xi_2$ . Then

$$p(\theta \mid \phi) = C' \cdot \sin \theta \cos \theta$$

while its CDF is

$$P(\theta) = \int_0^{\theta} C' \cdot \sin t \cos t \, dt = \sin^2 \theta.$$

From the inverse of the CDF, we can construct

$$\begin{cases} \theta = \arcsin\left(\sqrt{\xi_1}\right) \\ \phi = 2\pi\xi_2 \end{cases}$$

2. Consider The Rendering Equation written in directional form (the position  $\mathbf{p}$  is neglected)

$$L_o(\theta_o, \phi_o) = L_e(\theta_o, \phi_o) + \int_0^{2\pi} \left( d\phi_i \int_0^{\pi/2} \sin \theta_i d\theta_i \cdot f_r(\theta_o, \theta_i, \phi_o, \phi_i) L_i(\theta_i, \phi_i) \cos \theta_i \right).$$

Suppose that  $f_r$  and  $L_i$  are both analytical,  $f_r = \pi^{-1}$  and  $L_i = |\cos \theta_i|$ , derive the *Monte Carlo Estimator*  $I(\theta, \phi)$  that estimates the  $L_o$  above using the random variables  $\theta$  and  $\phi$  obtained from the previous question. Can this estimator be more efficient by altering the mapping of f and g? You can give your reasoning by derivations.

**Answer:** This question only require you to copy down the integral content and the PDF, as the *Monte Carlo Integrator* of the integral

$$I = \int_A f(x) \,\mathrm{d}\mu(x)$$

is

$$\langle I \rangle = \frac{f(X)}{p_{\mu}(X)}$$

where X is arbitrary random variable that spans the domain.

This way, to estimate  $L_o$ , the estimator can be written as

$$\langle L_o \rangle = L_e(\theta_o, \phi_o) + \frac{\sin \theta \cdot \pi^{-1} \cos^2 \theta}{p(\theta, \phi) = \pi^{-1} \sin \theta \cos \theta}$$
$$= L_e(\theta_o, \phi_o) + \cos \theta.$$

where all terms are either constant or a basic random variable.

When  $\langle I \rangle$  is a constant, the estimator is most efficient.