## Tutorial 8

TA: Mengyun Liu, Hongtu Xu

1°. F(x) 6 TO, 17 Universality of the Uniform

Iniversality of the Unitorm

(1) Ununifical), 
$$X = F^{-1}(U)$$
, compute CDF of  $X$ . (P(U \le a) = a)

$$F_{X}(x) = P(X \leq x) = P(F^{-1}(U) \leq x) = P(U \leq F(x))$$

=F(x)

#### Theorem

Let F be a CDF which is a continuous function and strictly increasing on the support of the distribution. This ensures that the inverse function  $F^{-1}$  exists, as a function from (0,1) to  $\mathbb{R}$ . We then have the following results.

- Let  $U \sim \text{Unif}(0,1)$  and  $X = F^{-1}(U)$ . Then X is an r.v. with CDF F. P(XEa)= F(a)
- 2 Let X be an r.v. with CDF F. Then  $F(X) \sim \text{Unif } (0,1)$ .

$$(2) Y = F(X) \in Co_{1}) \quad \underline{y \in R} \quad P(Y \leq y) = \underline{0} \quad y \leq 0$$

$$\underline{y \in (o_{1})} \quad P(Y \leq y) = P(F(X) \leq y) = P(X \leq F^{-1}(y)) = F[F^{-1}(y)] = \underline{y}$$

$$Y = C_{1}(X) \cdot (x + y) = \underline{0} \quad y \leq 0$$

$$Y = C_{1}(X) \cdot (x + y) = \underline{0} \quad y \leq 0$$

$$Y = C_{1}(X) \cdot (x + y) = \underline{0} \quad y \leq 0$$

$$Y = C_{1}(X) \cdot (x + y) = \underline{0} \quad y \leq 0$$

$$Y = C_{1}(X) \cdot (x + y) = \underline{0} \quad y \leq 0$$

$$Y = C_{1}(X) \cdot (x + y) = \underline{0} \quad y \leq 0$$

$$Y = C_{1}(X) \cdot (x + y) = \underline{0} \quad y \leq 0$$

$$Y = C_{1}(X) \cdot (x + y) = \underline{0} \quad y \leq 0$$

$$Y = C_{1}(X) \cdot (x + y) = \underline{0} \quad y \leq 0$$

$$Y = C_{1}(X) \cdot (x + y) = \underline{0} \quad y \leq 0$$

$$Y = C_{1}(X) \cdot (x + y) = \underline{0} \quad y \leq 0$$

$$Y = C_{1}(X) \cdot (x + y) = \underline{0} \quad y \leq 0$$

$$Y = C_{1}(X) \cdot (x + y) = \underline{0} \quad y \leq 0$$

$$Y = C_{1}(X) \cdot (x + y) = \underline{0} \quad y \leq 0$$

$$Y = C_{1}(X) \cdot (x + y) = \underline{0} \quad y \leq 0$$

$$Y = C_{1}(X) \cdot (x + y) = \underline{0} \quad y \leq 0$$

$$Y = C_{1}(X) \cdot (x + y) = \underline{0} \quad y \leq 0$$

$$Y = C_{1}(X) \cdot (x + y) = \underline{0} \quad y \leq 0$$

$$Y = C_{1}(X) \cdot (x + y) = \underline{0} \quad y \leq 0$$

$$Y = C_{1}(X) \cdot (x + y) = \underline{0} \quad y \leq 0$$

$$Y = C_{1}(X) \cdot (x + y) = \underline{0} \quad y \leq 0$$

$$Y = C_{1}(X) \cdot (x + y) = \underline{0} \quad y \leq 0$$

$$Y = C_{1}(X) \cdot (x + y) = \underline{0} \quad y \leq 0$$

$$Y = C_{1}(X) \cdot (x + y) = \underline{0} \quad y \leq 0$$

$$Y = C_{1}(X) \cdot (x + y) = \underline{0} \quad y \leq 0$$

$$Y = C_{1}(X) \cdot (x + y) = \underline{0} \quad y \leq 0$$

$$Y = C_{1}(X) \cdot (x + y) = \underline{0} \quad y \leq 0$$

$$Y = C_{1}(X) \cdot (x + y) = \underline{0} \quad y \leq 0$$

$$Y = C_{1}(X) \cdot (x + y) = \underline{0} \quad y \leq 0$$

$$Y = C_{1}(X) \cdot (x + y) = \underline{0} \quad y \leq 0$$

$$Y = C_{1}(X) \cdot (x + y) = \underline{0} \quad y \leq 0$$

$$Y = C_{1}(X) \cdot (x + y) = \underline{0} \quad y \leq 0$$

$$Y = C_{1}(X) \cdot (x + y) = \underline{0} \quad y \leq 0$$

$$Y = C_{1}(X) \cdot (x + y) = \underline{0} \quad y \leq 0$$

$$Y = C_{1}(X) \cdot (x + y) = \underline{0} \quad y \leq 0$$

$$Y = C_{1}(X) \cdot (x + y) = \underline{0} \quad y \leq 0$$

$$Y = C_{1}(X) \cdot (x + y) = \underline{0} \quad y \leq 0$$

$$Y = C_{1}(X) \cdot (x + y) = \underline{0} \quad y \leq 0$$

$$Y = C_{1}(X) \cdot (x + y) = \underline{0} \quad y \leq 0$$

$$Y = C_{1}(X) \cdot (x + y) = \underline{0} \quad y \leq 0$$

$$Y = C_{1}(X) \cdot (x + y) = \underline{0} \quad y \leq 0$$

$$Y = C_{1}(X) \cdot (x + y) = \underline{0} \quad y \leq 0$$

$$Y = C_{1}(X) \cdot (x + y) = \underline{0} \quad y \leq 0$$

- Sampling a unit disk uniformly
  - Wrong approach:  $r = \xi_1, \, \theta = 2\pi \, \xi_2$
  - PDF p(x,y) by normalization is:  $p(x, y) = 1/\pi$
  - Transform into polar coordinate:  $p(r, \theta) = r/\pi$   $p(r, \theta) = r \ p(x, y)$
  - Compute the marginal and conditional densities

$$p(r) = \int_0^{2\pi} p(r, \theta) d\theta = 2r$$
$$p(\theta|r) = \frac{p(r, \theta)}{p(r)} = \frac{1}{2\pi}$$

• Integrating and inverting to find P(r), P<sup>-1</sup>(r), P( $\theta | r$ ) P<sup>-1</sup>( $\theta | r$ )

$$r = \sqrt{\xi_1}$$
$$\theta = 2\pi \, \xi_2$$

$$x = r \cos \theta$$
$$y = r \sin \theta$$

- Suppose we draw samples from some density  $p(r,\theta)$
- Computing the Jacobian

$$J_T = \begin{pmatrix} \frac{\partial x}{\partial r} & \frac{\partial x}{\partial \theta} \\ \frac{\partial y}{\partial r} & \frac{\partial y}{\partial \theta} \end{pmatrix} = \begin{pmatrix} \cos \theta & -r \sin \theta \\ \sin \theta & r \cos \theta \end{pmatrix}$$

- The determinant:  $r(\cos^2\theta + \sin^2\theta) = r$
- So

$$p(x, y) = p(r, \theta)/r$$
  $\longrightarrow$   $p(r, \theta) = r p(x, y)$ 

#### Cosine-weighted hemisphere sampling

- It is useful to have a cosine distribution over the hemisphere (the incident cosine term)
- We require:  $p(\omega) \propto \cos \theta$
- Derive as before:

$$\int_{\mathbb{H}^2} c \ p(\omega) \ d\omega = 1$$

$$\int_0^{2\pi} \int_0^{\frac{\pi}{2}} c \cos \theta \sin \theta \, d\theta \, d\phi = 1$$

$$c \, 2\pi \, \int_0^{\pi/2} \cos \theta \, \sin \theta \, \, \mathrm{d}\theta = 1$$

$$c = \frac{1}{\pi}$$

$$x = r \sin \theta \cos \phi$$
$$y = r \sin \theta \sin \phi$$
$$z = r \cos \theta$$

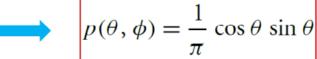
- Computing the Jacobian determinant:  $|J_T| = r^2 \sin \theta$
- The corresponding density function

$$p(r, \theta, \phi) = r^2 \sin \theta \ p(x, y, z)$$

- Solid angle defined with spherical coordinates  $d\omega = \sin\theta \; \mathrm{d}\theta \, \mathrm{d}\phi$
- If we have a density function defined over a solid angle

$$p(\theta, \phi) d\theta d\phi = p(\omega) d\omega \implies p(\theta, \phi) = \sin \theta \ p(\omega)$$

$$d\omega = \sin\theta \, d\theta \, d\phi$$
  $p(\theta, \phi) = \sin\theta \, p(\omega)$ 



#### Cosine-weighted hemisphere sampling

- It is useful to have a cosine distribution over the hemisphere (the incident cosine term)
- We require:  $p(\omega) \propto \cos \theta$
- Derive as before:

$$\int_{\mathbb{H}^2} c \ p(\omega) \ d\omega = 1$$

$$\int_0^{2\pi} \int_0^{\frac{\pi}{2}} c \cos \theta \sin \theta \, d\theta \, d\phi = 1$$

$$c \, 2\pi \, \int_0^{\pi/2} \cos \theta \, \sin \theta \, \, \mathrm{d}\theta = 1$$

$$c = \frac{1}{\pi}$$

$$x = r \sin \theta \cos \phi$$
$$y = r \sin \theta \sin \phi$$
$$z = r \cos \theta$$

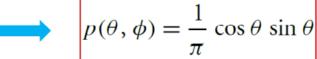
- Computing the Jacobian determinant:  $|J_T| = r^2 \sin \theta$
- The corresponding density function

$$p(r, \theta, \phi) = r^2 \sin \theta \ p(x, y, z)$$

- Solid angle defined with spherical coordinates  $d\omega = \sin\theta \; \mathrm{d}\theta \, \mathrm{d}\phi$
- If we have a density function defined over a solid angle

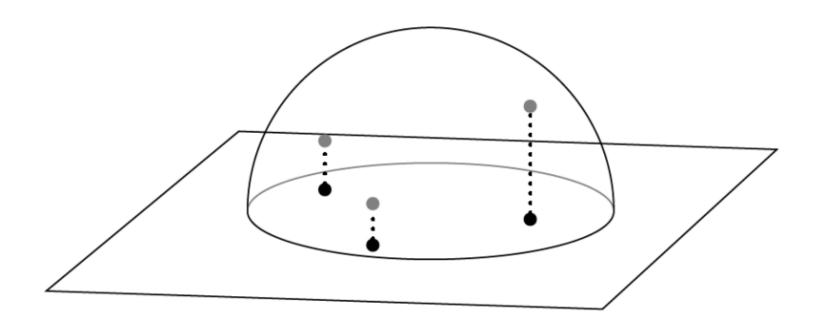
$$p(\theta, \phi) d\theta d\phi = p(\omega) d\omega \implies p(\theta, \phi) = \sin \theta \ p(\omega)$$

$$d\omega = \sin\theta \, d\theta \, d\phi$$
  $p(\theta, \phi) = \sin\theta \, p(\omega)$ 



## Cosine-weighted hemisphere sampling

- Malley's method
  - Sampling a unit disk and project onto the sphere



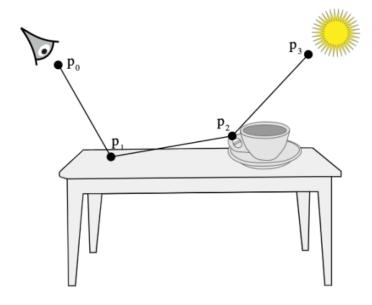
#### Monte-Carlo Integration

$$I = \int f(x) dx = \lim_{N o \infty} rac{1}{N} \sum_{i=1}^N rac{f(x_i)}{p(x_i)}$$

- p can be arbitrarily chosen
- The shape of p similar to  $f \rightarrow$  less variance

#### LTE

- Light transport equation
  - $L_o(p, \omega_o) = L_e(p, \omega_o) + \int_{\Omega} f(p, \omega_o, \omega_i) L_i(p, \omega_i) |\cos \theta_i| d\omega_i$
  - Basic idea: out = reflection + emission
  - Notice that  $L_o(p, \omega_o)$  and  $L_i(p, \omega_i)$  share the same light field but from different solid angles.
  - => Recursive relation



# Thanks