一、图形学入门

常用公式:

已知三角形三点坐标, 三角形面积为 $S=\frac{1}{2}[x_1(y_2-y_3)+x_2(y_3-y_1)+x_3(y_1-y_2)]$ 直线Ax+By+C=0, 点 $P(x_0,y_0)$ 到直线距离为 $\frac{|Ax_0+By_0+C|}{\sqrt{A^2+B^2}}$

・
$$ec{a}=(x_1,y_1,z_1), ec{b}=(x_2,y_2,z_2)$$
,则 $ec{a} imesec{b}=i(y_1z_2-z_1y_2)+j(z_1x_2-x_1z_2)+k(x_1y_2-y_1x_2)$

焦距: 眼到图像投影平面的距离

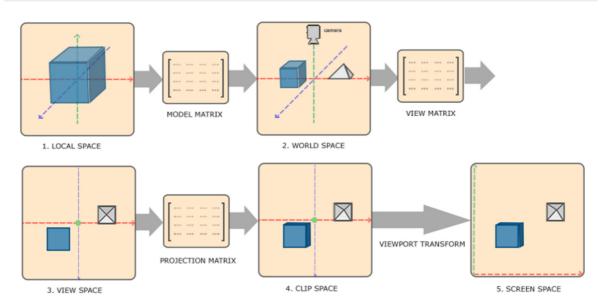
透视投影和平行投影的区别:透视投影焦距有限;平行投影焦距无限

光栅化 (Rasterizaton): 将连续信号转变为离散像素的过程

双缓存图像:设置主要缓存和次要缓存,不断切换屏幕指向的缓存地址,以避免单屏幕显示的"显示,擦 除"产生的闪烁

深度缓存(Z-buffer): 在光栅化每个像素点时,遍历几何体,同时记录该几何体在该像素点上的深 度。如果下一个几何体深度更小,则该像素点替换成下个几何体的颜色。该方法用于解决几何体的先后 问题。

二、坐标空间



在三维空间的计算中,点的坐标表示为:

$$(x,y,z,w)=(rac{x}{w},rac{y}{w},rac{z}{w})$$

注: 当w=0时,该坐标代表向量

变换矩阵

平移

$$T(\Delta x, \Delta y, \Delta z) = egin{pmatrix} 1 & 0 & 0 & \Delta x \ 0 & 1 & 0 & \Delta y \ 0 & 0 & 1 & \Delta z \ 0 & 0 & 0 & 1 \end{pmatrix}$$

拉伸

$$T(x,y,z) = egin{pmatrix} x & 0 & 0 & 0 \ 0 & y & 0 & 0 \ 0 & 0 & z & 0 \ 0 & 0 & 0 & 1 \end{pmatrix}$$

旋转

$$R_x(heta) = egin{pmatrix} 1 & 0 & 0 & 0 & 0 \ 0 & \cos heta & -\sin heta & 0 \ 0 & \sin heta & \cos heta & 0 \ 0 & 0 & 0 & 1 \end{pmatrix}$$
 $R_y(heta) = egin{pmatrix} \cos heta & 0 & \sin heta & 0 \ 0 & 1 & 0 & 0 \ -\sin heta & 0 & \cos heta & 0 \ 0 & 0 & 0 & 1 \end{pmatrix}$ $R_z(heta) = egin{pmatrix} \cos heta & -\sin heta & 0 & 0 \ \sin heta & \cos heta & 0 & 0 \ 0 & 0 & 1 & 0 \ 0 & 0 & 0 & 1 \end{pmatrix}$ $R(heta) = R_x(heta) R_y(heta) R_z(heta)$

注:法向量在不均匀缩放情况下直接乘以model矩阵会不垂直于交点, $N'=mat3(transpose(model^{-1}))*N$

View矩阵

步骤为先将相机平移到原点,再旋转,所以将世界坐标系做逆变换就能得到View矩阵(相机位置为 (0,0,0) ,看向 -z

$$M_v = R^{-1}T^{-1}$$

$$T^{-1} = egin{pmatrix} 1 & 0 & 0 & -x_e \ 0 & 1 & 0 & -y_e \ 0 & 0 & 1 & -z_e \ 0 & 0 & 0 & 1 \end{pmatrix}$$
 , $R^{-1} = egin{pmatrix} x_{\hat{g} imes \hat{t}} & x_{\hat{t}} & x_{-\hat{g}} & 0 \ y_{\hat{g} imes \hat{t}} & y_{\hat{t}} & y_{-\hat{g}} & 0 \ z_{\hat{g} imes \hat{t}} & z_{-\hat{g}} & 0 \ 0 & 0 & 0 & 1 \end{pmatrix}$

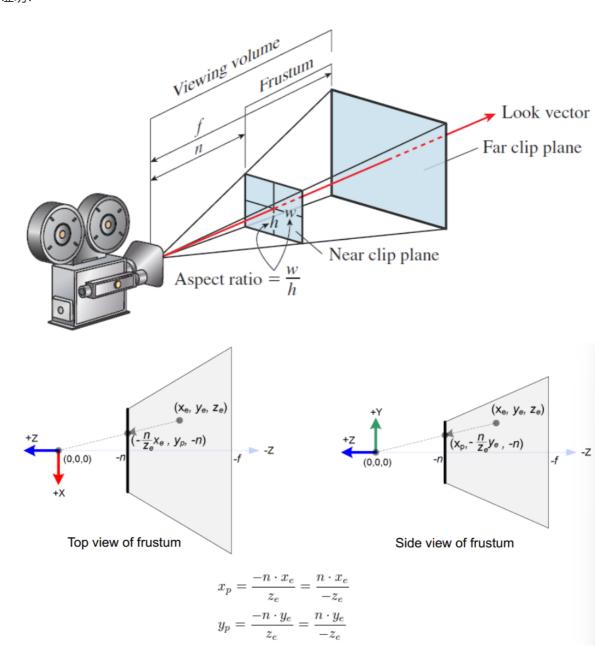
其中, \hat{t} 向y转, \hat{g} 向-z转, $\hat{g} \times \hat{t}$ 向x转

投影矩阵

透视投影

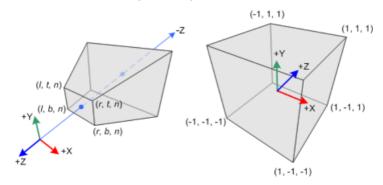
$$M_p = egin{pmatrix} rac{2n}{r-l} & 0 & rac{r+l}{r-l} & 0 \ 0 & rac{2n}{t-b} & rac{t+b}{t-b} & 0 \ 0 & 0 & rac{-(f+n)}{f-n} & rac{-2fn}{f-n} \ 0 & 0 & -1 & 0 \end{pmatrix}, where egin{pmatrix} t = n anrac{fovY}{2} \ r = aspect_ratio imes n anrac{fovY}{2} \ l = -r \ t = -b \end{pmatrix}$$

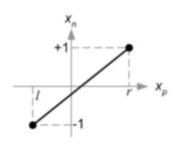
证明:



Normalized device coordinate (NDC)

- Range normalization
 - x-coordinate: [l, r] to [-1, 1]
 - y-coordinate: [b, t] to [-1, 1]
 - z-coordinate: [n, f] to [-1, 1]





$$x_n = \frac{1 - (-1)}{r - l} \cdot x_p + \beta$$

$$1 = \frac{2r}{r - l} + \beta \quad \text{(substitute } (r, 1) \text{ for } (x_p, x_n))$$

$$\beta = 1 - \frac{2r}{r - l} = \frac{r - l}{r - l} - \frac{2r}{r - l}$$

$$= \frac{r - l - 2r}{r - l} = \frac{-r - l}{r - l} = -\frac{r + l}{r - l}$$

$$\therefore x_n = \frac{2x_p}{r - l} - \frac{r + l}{r - l}$$

$$y_n = \frac{1 - (-1)}{t - b} \cdot y_p + \beta$$

$$1 = \frac{2t}{t - b} + \beta \qquad \text{(substitute } (t, 1) \text{ for } (y_p, y_n))$$

$$\beta = 1 - \frac{2t}{t - b} = \frac{t - b}{t - b} - \frac{2t}{t - b}$$

$$= \frac{t - b - 2t}{t - b} = \frac{-t - b}{t - b} = -\frac{t + b}{t - b}$$

$$\therefore y_n = \frac{2y_p}{t - b} - \frac{t + b}{t - b}$$

$$x_{n} = \frac{2x_{p}}{r - l} - \frac{r + l}{r - l} \qquad (x_{p} = \frac{nx_{e}}{-z_{e}})$$

$$= \frac{2 \cdot \frac{n \cdot x_{e}}{-z_{e}}}{r - l} - \frac{r + l}{r - l}$$

$$= \frac{2n \cdot x_{e}}{(r - l)(-z_{e})} - \frac{r + l}{r - l}$$

$$= \frac{2n}{r - l} \cdot \frac{x_{e}}{-z_{e}} - \frac{r + l}{r - l}$$

$$= \frac{2n}{r - l} \cdot \frac{x_{e}}{-z_{e}} - \frac{r + l}{r - l}$$

$$= \frac{2n}{r - l} \cdot \frac{x_{e}}{-z_{e}} + \frac{r + l}{r - l} \cdot \frac{x_{e}}{-z_{e}}$$

$$= \left(\frac{2n}{r - l} \cdot x_{e} + \frac{r + l}{r - l} \cdot z_{e}\right) / - z_{e}$$

$$= \left(\frac{2n}{r - l} \cdot x_{e} + \frac{r + l}{r - l} \cdot z_{e}\right) / - z_{e}$$

$$= \left(\frac{2n}{r - l} \cdot x_{e} + \frac{r + l}{r - l} \cdot z_{e}\right) / - z_{e}$$

$$= \left(\frac{2n}{r - l} \cdot x_{e} + \frac{r + l}{r - l} \cdot z_{e}\right) / - z_{e}$$

$$\begin{split} &= \frac{2x_p}{r-l} - \frac{r+l}{r-l} \qquad (x_p = \frac{nx_e}{-z_e}) \\ &= \frac{2 \cdot \frac{n \cdot x_e}{-z_e}}{r-l} - \frac{r+l}{r-l} \\ &= \frac{2n \cdot x_e}{(r-l)(-z_e)} - \frac{r+l}{r-l} \\ &= \frac{\frac{2n \cdot x_e}{r-l} - \frac{r+l}{r-l}}{-z_e} - \frac{r+l}{r-l} \\ &= \frac{\frac{2n}{r-l} \cdot x_e}{-z_e} - \frac{r+l}{r-l} \\ &= \frac{\frac{2n}{r-l} \cdot x_e}{-z_e} + \frac{\frac{r+l}{r-l} \cdot z_e}{-z_e} \\ &= \left(\frac{\frac{2n}{r-l} \cdot x_e + \frac{r+l}{r-l} \cdot z_e}{x_c} \right) \middle/ - z_e \end{split}$$

$$= \frac{\left(\frac{2n}{r-l} \cdot x_e + \frac{r+l}{r-l} \cdot z_e \right)}{x_c} \middle/ - z_e$$

$$= \left(\frac{2n}{r-l} \cdot x_e + \frac{r+l}{r-l} \cdot z_e \right) \middle/ - z_e$$

$$= \left(\frac{2n}{r-l} \cdot x_e + \frac{r+l}{r-l} \cdot z_e \right) \middle/ - z_e$$

$$\begin{pmatrix} x_c \\ y_c \\ z_c \\ w_c \end{pmatrix} = \begin{pmatrix} \frac{2n}{r-l} & 0 & \frac{r+l}{r-l} & 0 \\ 0 & \frac{2n}{t-b} & \frac{t+b}{t-b} & 0 \\ 0 & 0 & A & B \\ 0 & 0 & -1 & 0 \end{pmatrix} \begin{pmatrix} x_e \\ y_e \\ z_e \\ w_e \end{pmatrix}, \qquad z_n = z_c/w_c = \frac{Az_e + Bw_e}{-z_e}$$

$$\begin{cases} \frac{-An+B}{n} = -1\\ \frac{-Af+B}{f} = 1 \end{cases} \rightarrow \begin{cases} -An+B = -n\\ -Af+B = f \end{cases}$$

平行矩阵

$$M_o = egin{pmatrix} rac{2}{r-l} & 0 & 0 & -rac{r+l}{r-l} \ 0 & rac{2}{t-b} & 0 & rac{t+b}{t-b} \ 0 & 0 & rac{-2}{f-n} & -rac{f+n}{f-n} \ 0 & 0 & 0 & 1 \end{pmatrix}$$

观察矩阵

 $[-1,1]^2 \rightarrow [0,width] \times [0,height]$

$$M_v iewport = egin{pmatrix} rac{width}{2} & 0 & 0 & rac{width}{2} \ 0 & rac{height}{2} & 0 & rac{height}{2} \ 0 & 0 & 1 & 0 \ 0 & 0 & 0 & 1 \end{pmatrix}$$

光栅化

Bresenham算法

在平面直角坐标系上, 假设有一条直线, 方程为

 $f(x,y)=Ax+By+C, A=\Delta y, B=-\Delta x, C=(\Delta x)b$,注意这里斜率不能过大。从 (x_0,y_0) 开始,决定需要填色的像素是 (x_0+1,y_0) 还是 (x_0+1,y_0+1) ,目标是寻找其中一个距离直线更近的点

解决:将 $x=x_0+1,y=y_0+\frac{1}{2}$ 代入直线方程,若 $f(x_0+1,y_0+\frac{1}{2})<=0$,则选择 (x_0+1,y_0) 涂色,若 $f(x_0+1,y_0+\frac{1}{2})>0$,则选择 (x_0+1,y_0+1) 涂色

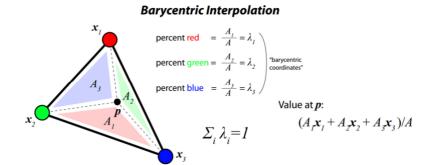
三角形取点

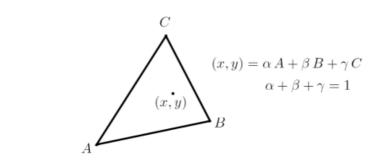
对于三角形每一条直线,做Bresenham算法: $P_i=(X_i,Y_i)$, $dX_i=X_{i+1}-X_i$, $dY_i=Y_{i+1}-Y_i$

$$E_i(x,y) = (x-X_i)dY_i - (y-Y_i)dX_i \left\{ egin{aligned} &= 0: point \ on \ edge \ &> 0: outside \ edge \ &< 0: inside \ edge \end{aligned}
ight.$$

取所有点,当三条直线 $E_i(x,y)$ 同时满足< 0时,该点为三角形内的点

上色





$$\alpha = \frac{-(x - x_B)(y_C - y_B) + (y - y_B)(x_C - x_B)}{-(x_A - x_B)(y_C - y_B) + (y_A - y_B)(x_C - x_B)}$$
$$\beta = \frac{-(x - x_C)(y_A - y_C) + (y - y_C)(x_A - x_C)}{-(x_B - x_C)(y_A - y_C) + (y_B - y_C)(x_A - x_C)}$$
$$\gamma = 1 - \alpha - \beta$$

三、曲线和三角化

德劳内三角剖分法 (构造Mesh) : 任意三角形的外接圆内没有第四个点

翻转 Flipping: 如果两个三角形未满足德劳内,则换中间连线

泰森多边形 Voronoi Diagram:将平面划分为不同区域(每个区域只有一个点),使该区域到区域中的点距离最近,这些区域叫做Voronoi Cells,可由中垂线法生成。

泰森多边形和德劳内三角是对偶关系,连接泰森多边形相邻的点即可获得德劳内三角剖分

四、几何模型

椭球模型: $\frac{x^2}{a^2} + \frac{y^2}{b^2} + \frac{z^2}{c^2} = 1$

椭球坐标变换: $x = a\cos(u)\cos(v), y = b\cos(u)\sin(v), z = c\sin(u)$, $-\frac{\pi}{2} \leq u \leq \frac{\pi}{2}, -\pi \leq v \leq \pi$

对于曲线某点的切线,直接求偏导;对于曲面某点的法向量,先按照两个不同方向求偏导,得两个切 线,再对两个切线的方向向量做叉乘

插值问题:给定值的映射序列,找到满足该映射的函数

多项式插值:对于n+1个数据,构造一个最大项为n次的多项式,将数据带入求解

多项式插值的震荡现象: 当多项式阶数很高时, 会在边缘点发生强烈的震荡

样条插值: 样条插值是一种**分段定义**的低阶多项式函数,但阶数也不能过低,需要保证一定的曲线连续 性,通常考虑三阶的样条插值

三阶样条插值: 定义 $p(x) = a + bx + cx^2 + dx^3$, 其具有二阶导数连续性, 在结点处一致

给定定义在[a,b]上的函数f(x)和结点集合 $a=x_0 < x_1 < \ldots < x_n = b$,样条曲线为:

$$S(x) = egin{cases} a_0 + b_0(x - x_0) + c_0(x - x_0)^2 + d_0(x - x_0)^3, & if \ x_0 \leq x \leq x_1 \ a_1 + b_1(x - x_1) + c_1(x - x_1)^2 + d_1(x - x_1)^3, & if \ x_1 \leq x \leq x_2 \ \dots \ a_{n-1} + b_{n-1}(x - x_{n-1}) + c_{n-1}(x - x_{n-1})^2 + d_{n-1}(x - x_{n-1})^3, & if \ x_{n-1} \leq x \leq x_n \end{cases}$$

同时要满足:对于每个结点,S值与f值一致;连续两条样条曲线共有点的值一致;连续两条样条曲线在 共有点的一阶导数一致;连续两条样条曲线在共有点的二阶导数一致

还要满足下述其一: $S''(x_0) = S''(x_n) = 0$ (自然样条) 或 $S'(x_0) = f'(x_0), S'(x_n) = f'(x_n)$ (紧凑样条)

Bernstein Basis:
$$b_{v,n}(x) = \binom{n}{v} x^v (1-x)^{n-v}$$

贝塞尔曲线:
$$B(t)=\sum_{i=0}^n b_{i,n}(t)P_i=\sum_{i=0}^n \binom{n}{v}t^i(1-t)^{n-i}P_i$$
,其中 P_i 是控制点

de Casteljau:
$$eta_i^{(0)}=eta_i, i=0,\dots,n$$
 and $eta_i^{(j)}=eta_i^{(j-1)}(1-t_0)+eta_{i+1}^{(j-1)}t_0, i=0,\dots,n-j, j=1,\dots,n$

贝塞尔曲面: $P(u,v) = \sum_{i=1}^n \sum_{j=0}^m B_i^n(u) B_j^m(v) k_{i,j}$,其有(n+1)(m+1)个控制点定义

B样条曲线: $S_{n,t}(x) = \sum_{i=0}^{degree} \alpha_i B_{i,n}(x)$

Coxde Boor法解B样条曲线:

$$B_{i,1}(x) = egin{cases} 1, & t_i \leq x \leq t_{i+1} \ 0, & otherwise \end{cases} \ B_{i,k}(x) = rac{x-t_i}{t_{i+k-1}-t_i} B_{i,k-1}(x) + rac{t_{i+k}-x}{t_{i+k}-t_{i+1}} B_{i+1,k-1}(x) \end{cases}$$

Catmull细分曲面:

- 增加面中心点: $v_F = \sum_{i=1}^n \frac{1}{n} v_i$ 增加边点: $v_E = \frac{v+w+v_{F_1}+v_{F_2}}{4}$, 其中v, w是边端点, v_{F_1} , v_{F_2} 是生成的相邻面中心点
 改变原来点的值: $v' = \frac{1}{n}Q + \frac{2}{n}R + \frac{n-3}{n}v$, 其中n为面的顶点数, 如正方形为n = 3, Q为所 有相邻生成面中心点的均值,R为所有和该点相邻的生成的边点的均值

五、几何渲染

冯模型 Phong reflection model

光的组成:

• 环境光Ambient: 常量光源 (与观看角度无关)

• 漫射光Diffuse: 均匀发射向任意方向的光源 (与观看角度无关)

• 镜面光Specular: 向特定方向反射的光源 (和观看角度有关)

注: 对于物体也定义了如上三种属性

$$I_p = k_a i_a + \sum_{m \in lights} (k_d (\hat{L_m} \cdot \hat{N}) i_{m,d} + k_s (\hat{R_m} \cdot \hat{V})^lpha i_{m,s})$$

其中 k_a,k_d,k_s 分别是环境光、漫反射、镜面反射常量,和光属性相关; $i_a,i_{m,d},i_{m,s}$ 和物体属性相关; \hat{L}_m 代表归一后光射入的反向向量, \hat{N} 代表归一后光和物体交点物体的法向量,二者点乘得夹角 θ ; \hat{R}_m 是出射光线,且 $\hat{R}_m=2(\hat{L}_m\cdot\hat{N})\hat{N}-\hat{L}_m$, \hat{V} 代表从交点指向相机的向量, α 控制shininess,值越低,图形越善亮

布林冯模型 Blinn-Phone

定义 $H=rac{\hat{L}_m+\hat{V}}{||\hat{L}_m+\hat{V}||}$,用 $(\hat{N}\cdot H)^{lpha'}$ 代替冯模型中 $(\hat{R_m}\cdot\hat{V})^{lpha}$

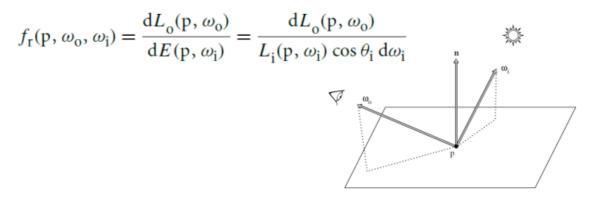
XYZ颜色和RGB颜色

根据能量,定义 $X=k\int_{\lambda}\Phi(\lambda)ar{x}(\lambda)d\lambda$, $Y=k\int_{\lambda}\Phi(\lambda)ar{y}(\lambda)d\lambda$, $Z=k\int_{\lambda}\Phi(\lambda)ar{z}(\lambda)d\lambda$

进行归一化得
$$x=rac{X}{X+Y+Z},y=rac{Y}{X+Y+Z},z=rac{Z}{X+Y+Z}$$

将红绿蓝单独提取,定义 $R=\int_{\lambda}S(\lambda)ar{r}(\lambda)d\lambda$, $G=\int_{\lambda}S(\lambda)ar{g}(\lambda)d\lambda$, $B=k\int_{\lambda}S(\lambda)ar{b}(\lambda)d\lambda$

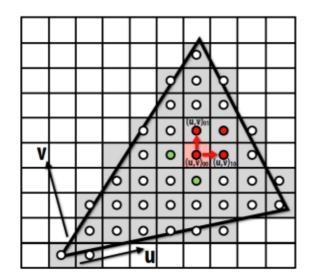
双向反射分布函数 BRDF

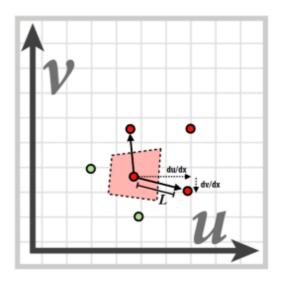


$$L_0(p,\omega_0) = \int_{s^2} f(p,\omega_0,\omega_i) L_i(p,\omega_i) |\cos heta_i| d\omega_i$$

折射: $\eta_i \sin \theta_i = \eta_t \sin \theta_t$

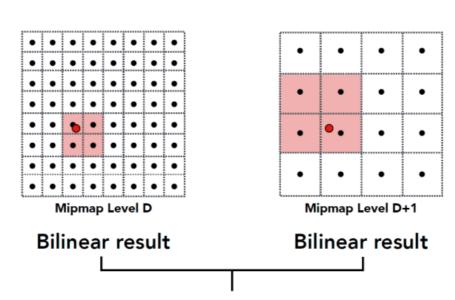
六、纹理





$$D = \log_2 L$$

$$L = \max\left(\sqrt{\left(\frac{du}{dx}\right)^2 + \left(\frac{dv}{dx}\right)^2}, \sqrt{\left(\frac{du}{dy}\right)^2 + \left(\frac{dv}{dy}\right)^2}\right)$$



Linear interpolation based on continuous D

七、光线追踪

1. 计算光线r(t)=o+td和圆 $f(x)=|x|^2-1$ 交点:

将光线方程带入圆方程得
$$f(r(t)) = |o+td|^2 - 1 \rightarrow |d|^2 t^2 + 2(o \cdot d)t + |o|^2 - 1 = 0$$

$$t = -o \cdot d \pm \sqrt{(o \cdot d)^2 - |o|^2 + 1}$$

2. 计算光线r(t) = o + td和平面 $N^T x = c$ 交点:

$$N^T r(t) = c
ightarrow N^T(o+td) = c
ightarrow t = rac{c-N^To}{N^Td}
ightarrow r(t) = o + rac{c-N^To}{N^Td} d$$

3. 判断一个点是否在三角形内部? 需要满足

$$p(b_1,b_2) = (1-b_1-b_2)p_0 + b_1p_1 + b_2p_2, b_1 \ge 0, b_2 \ge 0, b_1+b_2 \le 1$$

将直线方程代入得

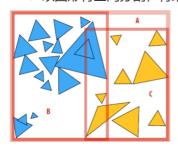
 $o+td=(1-b_1-b_2)p_0+b_1p_1+b_2p_2, b_1\geq 0, b_2\geq 0, b_1+b_2\leq 1$

$$(-\mathbf{d} \quad \mathbf{e}_1 \quad \mathbf{e}_2) \begin{bmatrix} t \\ b_1 \\ b_2 \end{bmatrix} = \mathbf{s}$$

$$\begin{bmatrix} t \\ b_1 \\ b_2 \end{bmatrix} = \frac{1}{(\mathbf{d} \times \mathbf{e}_2) \cdot \mathbf{e}_1} \begin{bmatrix} (\mathbf{s} \times \mathbf{e}_1) \cdot \mathbf{e}_2 \\ (\mathbf{d} \times \mathbf{e}_2) \cdot \mathbf{s} \\ (\mathbf{s} \times \mathbf{e}_1) \cdot \mathbf{d} \end{bmatrix} \longrightarrow \begin{bmatrix} t \\ b_1 \\ b_2 \end{bmatrix} = \frac{1}{\mathbf{s}_1 \cdot \mathbf{e}_1} \begin{bmatrix} \mathbf{s}_2 \cdot \mathbf{e}_2 \\ \mathbf{s}_1 \cdot \mathbf{s} \\ \mathbf{s}_2 \cdot \mathbf{d} \end{bmatrix}$$

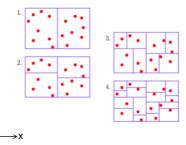
图形空间分割

- Octree: 正方形不断分割成四个正方形,直至分割后正方形中图形数少于一定值,树状存储
- BVH: 以图形将空间分割,将结点的基元分成不相连集合

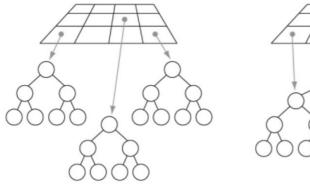


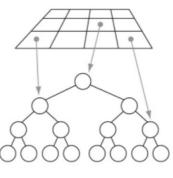


• K-D树:基元表示法,每个基元都由其中心点表示



• Hybrid:每一个划分空间包含独立或是混合的树





差异定义

给定N维序列P和一个采样序列B,P关于B的差异定义为

$$D_N(B, P) = \sup_{b \in B} \left| \frac{\sharp \{x_i \in b\}}{N} - \lambda(b) \right|$$

$$B = \{[0, v_1] \times [0, v_2] \times \dots \times [0, v_s]\}$$

$$P = x_1, \dots, x_N$$

 $\sharp\{x_i \in b\}$ is the number of points in b

 $\lambda(b)$ is the volume of b

八、采样

球分布变换: $x=r\sin\theta\cos\phi, y=r\sin\theta\sin\phi, z=r\cos\theta$,假设我们想根据 $p(r,\theta)$ 采样,先计算 Jacobian矩阵的模, $|J_T|=r^2\sin\theta$,则有 $p(r,\theta,\phi)=r^2\sin\theta p(x,y,z)$ 。根据球坐标定义的固定角 $d\omega=\sin\theta d\theta d\phi$,则 $p(\theta,\phi)d\theta d\phi=p(\omega)d\omega\to p(\theta,\phi)=\sin\theta p(\omega)$

- 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), $P^{-1}(r)$, $P(\theta)$, and $P^{-1}(\theta)$

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

- Uniformly sampling a hemisphere
 - Uniform sampling means $p(\omega) = c$
 - Normalization:

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

$$\int_{\mathbb{H}^2} p(\omega) \, d\omega = 1 \Rightarrow c \int_{\mathbb{H}^2} d\omega = 1 \Rightarrow c = \frac{1}{2\pi} \longrightarrow p(\omega) = 1/(2\pi) \longrightarrow p(\theta, \phi) = \sin \theta/(2\pi)$$

Consider sampling θ:

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

• Compute the conditional density for φ :

$$p(\phi|\theta) = \frac{p(\theta, \phi)}{p(\theta)} = \frac{1}{2\pi}$$

• Use 1D inversion technique to sample:

$$P(\theta) = \int_0^\theta \sin \theta' \, d\theta' = 1 - \cos \theta$$
$$P(\phi|\theta) = \int_0^\phi \frac{1}{2\pi} \, d\phi' = \frac{\phi}{2\pi}$$

· Inversion is straightforward

$$\theta = \cos^{-1} \xi_1$$

$$\phi = 2\pi \xi_2.$$

$$x = \sin \theta \cos \phi = \cos (2\pi \xi_2) \sqrt{1 - \xi_1^2}$$

$$y = \sin \theta \sin \phi = \sin (2\pi \xi_2) \sqrt{1 - \xi_1^2}$$

$$z = \cos \theta = \xi_1$$

• 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:

$$\int_{\Im C} c \ p(\omega) \ d\omega = 1 \qquad d\omega = \sin \theta \ d\theta \ d\phi \qquad p(\theta, \phi) = \sin \theta \ p(\omega)$$

$$\int_{0}^{2\pi} \int_{0}^{\frac{\pi}{2}} c \cos \theta \sin \theta \ d\theta \ d\phi = 1$$

$$c \ 2\pi \int_{0}^{\pi/2} \cos \theta \sin \theta \ d\theta = 1$$

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

重要性采样:

$$F_N = rac{1}{N} \sum_{i=1}^N rac{f(X_i)}{p(X_i)}$$

如果p(x)和f(x)近似,则从p(x)中采样收敛很快

九、重建

在一个像素附近插值采样以进行重建, 计算一个加权平均数:

$$I(x, y) = \frac{\sum_{i} f(x - x_{i}, y - y_{i}) L(x_{i}, y_{i})}{\sum_{i} f(x - x_{i}, y - y_{i})}$$

- $L(x_i, y_i)$: the radiance value of the i-th sample at (x_i, y_i)
- f is a filter function

高斯滤波: 轻微模糊

米切尔滤波:接收负值,提高边缘锐度

十、全局光照

双向路径追踪:

Fixed-length bounce rendering

- Enumerating all possibilities
 - s: number of light source ray path
 - t : number of camera ray path
 - Rendering is based on a fixed length L: s+t=L
- s+t=1: direct emission

光子估计: 光源发射, 环境存储光子信息, 当作二次光源

Radiance estimate

- Incoming flux is approximated using the photon map
- Searching the nearest n photons
- Each photon p has equal power (energy)

$$L_r(x,\vec{\omega}) = \int_{\Omega_r} f_r(x,\vec{\omega}',\vec{\omega}) \frac{d\Phi_i(x,\vec{\omega}')}{dA_i} \approx \sum_{p=1}^n f_r(x,\vec{\omega}_p,\vec{\omega}) \frac{\Delta\Phi_p(x,\vec{\omega}_p)}{\Delta A}$$

- Assuming that the surface is locally flat
 - Projecting the sphere onto the tangent surface

$$\Delta A = \pi r^2$$

十一、体积渲染

体积渲染方程:

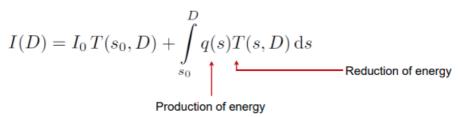
Volume rendering equation

- Integrating along the direction of light

$$I(D) = I_0 e^{-\int_{s_0}^{D} \kappa(t) dt} + \int_{s_0}^{D} q(s) e^{-\int_{s}^{D} \kappa(t) dt} ds$$

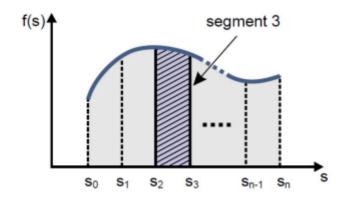
Integral in terms of transparency





离散化

$$I(s_i) = I(s_{i-1})T(s_{i-1}, s_i) + \int_{s_{i-1}}^{s_i} q(s)T(s, s_i) ds$$



Ray integral

Quantity definition

$$I(s_i) = I(s_{i-1})T(s_{i-1}, s_i) + \int_{s_{i-1}}^{s_i} q(s)T(s, s_i) ds$$

$$T_i = T(s_{i-1}, s_i), \quad c_i = \int_{s_{i-1}}^{s_i} q(s)T(s, s_i) ds$$



$$I(D) = I(s_n) = I(s_{n-1})T_n + c_n = (I(s_{n-2})T_{n-1} + c_{n-1})T_n + c_n = \dots$$



$$I(D) = \sum_{i=0}^{n} c_i \prod_{j=i+1}^{n} T_j$$
, with $c_0 = I(s_0)$

Compositing schemes

- Front-to-back composition
 - Viewing rays are traversed from eye into the volume

$$\hat{C}_i = \hat{C}_{i+1} + \hat{T}_{i+1}C_i \\ \hat{T}_i = \hat{T}_{i+1}(1 - \alpha_i)$$
 Alpha blending
$$C_{\mathrm{dst}} \leftarrow C_{\mathrm{dst}} + (1 - \alpha_{\mathrm{dst}})C_{\mathrm{src}} \\ \alpha_{\mathrm{dst}} \leftarrow \alpha_{\mathrm{dst}} + (1 - \alpha_{\mathrm{dst}})\alpha_{\mathrm{src}}$$

- Back-to-front composition

 Viewing rays are traversed from back of the volume into the eye

$$\begin{split} \hat{C}_i &= \hat{C}_{i-1}(1-\alpha_i) + C_i \\ \hat{T}_i &= \hat{T}_{i-1}(1-\alpha_i) \end{split} \quad \text{Alpha blending} \quad C_{\mathrm{dst}} \leftarrow (1-\alpha_{\mathrm{src}})C_{\mathrm{dst}} + C_{\mathrm{src}}$$

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