Volumetric Image Visualization

Alexandre Xavier Falcão

LIDS - Institute of Computing - UNICAMP

afalcao@ic.unicamp.br

Alexandre Xavier Falcão MO815 - Volumetric Image Visualization

Normal estimation for the Phong's illumination model

 In the Phong's illumination model, the angle θ must be computed between -n' = -φ_r⁻¹(n) and the normal vector o.n(p') at the surface point p'.

$$r(p') = k_a r_a + r_d(p') \left(k_d \cos\left(\theta\right) + k_s \cos^{n_s}\left(2\theta\right)\right).$$

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- In this lecture, we will learn when and how to estimate o.n(p') using the two main approaches:
 - scene-based normal estimation and
 - object-based normal estimation.

Scene-based normal estimation

Whenever there exists high contrast in $\hat{I} = (D_I, I)$ between an object and its surroundings, the normal vector at a surface voxel $p \in D_I$ can be estimated from the gradient vector G of the scene.

$$\begin{aligned} \mathsf{G}(p) &= \sum_{\forall q \in \mathcal{A}_r(p)} I(q) - I(p) \frac{q-p}{\|q-p\|}, \\ \mathcal{A}_r(p) &= \{q \in D_I \mid \|q-p\| \leq r, q \neq p\} \end{aligned}$$

for small values $1 \le r \le 5$.



Scene-based normal estimation

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- However, the orientation of the normal vector o.n(p) should always be towards the exterior of the object.
- Given that we have segmented the objects in the scene and output a label image L̂ = (D_I, L), such that L(p) ∈ {0, 1, ..., c} indicates when p ∈ D_I belongs to the background, L(p) = 0, or to one of c objects, L(p) = j, j ∈ [1, c], this information can be used as follows.

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The normal vector o.n(p) can be defined by

$$o.n(p) = \begin{cases} \frac{-G(p)}{\|G(p)\|} & \text{if } L(p + \alpha \frac{G(p)}{\|G(p)\|}) = L(p), \\ \frac{+G(p)}{\|G(p)\|} & \text{if } L(p + \alpha \frac{G(p)}{\|G(p)\|}) \neq L(p), \end{cases}$$

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For a point $p' = (x_{p'}, y_{p'}, z_{p'})$ with real coordinates, such that $(\lceil x_{p'} \rceil, \lceil y_{p'} \rceil, \lceil z_{p'} \rceil) \in D_I$, o.n(p') can be obtained by interpolation from the normal vectors of the nearby spels.

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Interpolation for normal estimation

The gradient vectors of the nearby spels $q_k \in D_I$, k = 1, 2, ..., 8, are used to estimate G(p') and so o.n(p').



$$G(p') = (x_{p'} - x_{q_{1357}})G(q_{2468}) + (x_{q_{2468}} - x_{p'})G(q_{1357})$$

$$G(q_{2468}) = (z_{p'} - z_{q_{24}})G(q_{68}) + (z_{q_{68}} - z_{p'})G(q_{24})$$

$$G(q_{1357}) = (z_{p'} - z_{q_{13}})G(q_{57}) + (z_{q_{57}} - z_{p'})G(q_{13})$$

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Interpolation for normal estimation



$$\begin{aligned} \mathsf{G}(q_{24}) &= (y_{p'} - y_{q_4})\mathsf{G}(q_2) + (y_{q_2} - y_{p'})\mathsf{G}(q_4) \\ \mathsf{G}(q_{68}) &= (y_{p'} - y_{q_8})\mathsf{G}(q_6) + (y_{q_6} - y_{p'})\mathsf{G}(q_8) \\ \mathsf{G}(q_{13}) &= (y_{p'} - y_{q_3})\mathsf{G}(q_1) + (y_{q_1} - y_{p'})\mathsf{G}(q_3) \\ \mathsf{G}(q_{57}) &= (y_{p'} - y_{q_7})\mathsf{G}(q_5) + (y_{q_5} - y_{p'})\mathsf{G}(q_7) \end{aligned}$$

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- After segmentation, these methods depend on the type of object representation.
 - Mesh-based object representation.
 - Voxel-based object representation.

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Normal estimation from a surface mesh

When the object's surface is represented by a mesh of polygons,

- the vertices are stored in a given order for the normal estimation of the faces from the outer product between two edges.
- Normal estimation for the vertices, along the edges and scan-lines on the faces are obtained by interpolation.



Normal estimation from boundary voxels

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- Boundary voxels S may be extracted from each object o, with $\lambda(o) \in [1, c]$, in $\hat{L} = (D_I, L)$ as

$$\mathcal{S} = \{ p \in D_I \mid \exists q \in \mathcal{A}_1(p), L(q) \neq L(p), L(p) = \lambda(o) \}$$

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• One sEDT algorithm must be executed for each object o upto a small distance $\rho \ge 1$ from S, such that even the parts of its surface hidden by other objects can be visualized when the visibility of those objects is turned off.

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The sEDT algorithm

Input: $\hat{L} = (D_I, L), \lambda(o), \rho$, and S of object o. Output: Signed distance image $\hat{C} = (D_I, C)$, initially with zeros. 01. For each $p \in D_I$, $C(p) \leftarrow +\infty$. 02. While $\mathcal{S} \neq \emptyset$ do Remove p from S, $C(p) \leftarrow 0$, $R(p) \leftarrow p$, and insert p in Q. 03. 04. While $Q \neq \emptyset$ do Remove $p = \arg \min_{q \in Q} \{C(q)\}$ from Q. 05. 06. If $\sqrt{C(p)} < \rho$ then For each $q \in \mathcal{A}_{\sqrt{3}}(p) \mid C(q) > C(p)$ do 07. $tmp \leftarrow ||q - R(p)||^2$. 08. If tmp < C(q) then 09. 10. If $q \in Q$ then remove q from Q. $C(q) \leftarrow tmp, R(q) \leftarrow R(p)$, and insert q in Q. 11. 12. For each $p \in D_I$ do, If $C(p) = +\infty$ then $C(p) \leftarrow 0$. Else, If $L(p) \neq \lambda(o)$, then $C(p) \leftarrow -C(p)$. 13. By construction, the gradient of the signed distance image $\hat{C} = (D_I, C)$ at a boundary voxel $p \in S$ should always point towards the interior of the object. The normal vector o.n(p) can then be estimated as

$$o.n(p) = \frac{-G(p)}{\|G(p)\|},$$

$$G(p) = \sum_{\forall q \in \mathcal{A}_{r}(p)} C(q) - C(p) \frac{q-p}{\|q-p\|},$$

$$\mathcal{A}_{r}(p) = \{q \in D_{I} \mid \|q-p\| \le r, q \ne p\}$$

for small values $1 \le r \le 5$.

Scene-based vs. Object-based approaches

Scene-based normal estimation provides renditions of better quality, but object-based normal estimation is a good approximation needed in some situations.



Scene-based (left) and object-based (right) normal estimation. The holes may come from segmentation by thresholding or thin parts of the boundary.

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