

3D Reconstruction from Single 2D Image Based on Silhouette Optimization

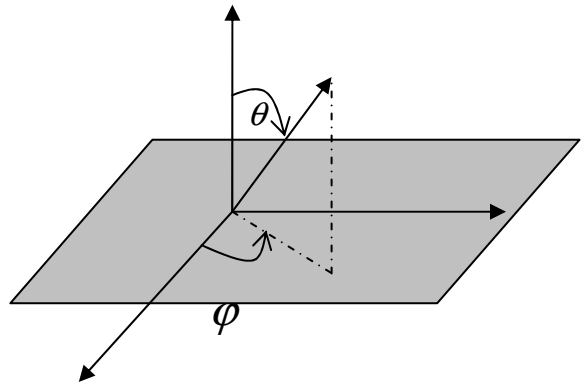
Abstract—In order to overcome the shortcomings of the shape from shading (SFS) algorithm in terms of clearness and smoothness of silhouette in reconstructed objects, a modified SFS algorithm is proposed in this paper. Firstly, we determine the primary factors affecting reconstruction by analyzing the Lambertian light reflection model. Secondly, we perform image restoration and background smoothing by extracting object silhouette using edge detection to separate the object and background. Finally the object is reconstructed using the SFS algorithm. Theory and simulation results show that, compared with the traditional algorithms, the performance of the modified algorithm was efficiently improved, which can improve the accuracy of the reconstructed shape and the continuity of the silhouette of reconstructed objects by reducing the reconstruction error effectively.

$$(x, y) R(p, q)$$

$$(x, y)$$

$$x \quad p \quad q \quad p \quad q$$

$$y$$



B. SFS algorithm

$$n = p \ q -$$

$$(x, y) \ n_s = p_i \ q_i -$$

θ

$$\theta = \frac{n_i \times n}{\|n_i \times n\|}$$

$$= \rho \frac{pp_i + qq_i +}{\sqrt{p_i + q_i +} \sqrt{p + q +}}$$

A. Lambertain illumination model

$$I(x, y) = R(p, x, y, q, x, y)$$

$$I(x, y)$$

$$E(x, y)$$

$$I(x, y) = R(p, q)$$

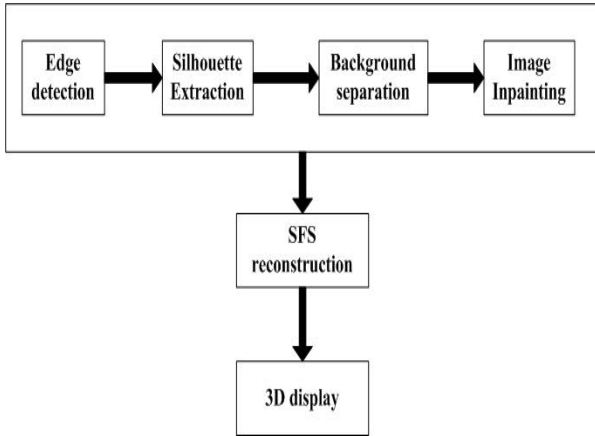
$$= \rho \frac{pp_i + qq_i +}{\sqrt{p_i + q_i +} \sqrt{p + q +}}$$

$$R(p, q) = I_e \rho \quad \theta \quad I_e \rho \quad \angle \quad n \quad s$$

$$I_e \quad \rho$$

$$\theta \quad s$$

$$n$$



A. Silhouette extraction of 2D information

$$E_1 \subset E_2 \subset E$$

B. Background separation and image inpainting

$$E_{avg} = \frac{E_1 + E_2}{2}$$

$$E_{new}(x, y) = \begin{cases} E_{avg}(x, y) & x, y \in E_{background} \\ E_{old}(x, y) & x, y \in E_{substance} \end{cases}$$

$E_{substance}$

$$\frac{\partial u}{\partial t} = \nabla \cdot \left(\frac{\nabla u}{|\nabla u|} \right) + \lambda_e (u - u_0)$$

D

C. 3D image reconstruction

$$p = \frac{\partial z}{\partial x} = z_x - z_{x_0}$$

$$q = \frac{\partial z}{\partial y} = z_y - z_{y_0}$$

$$z = f(x, y), \quad (x, y) \in (Oxyz)$$

$$R(p, q) = E(x, y)$$

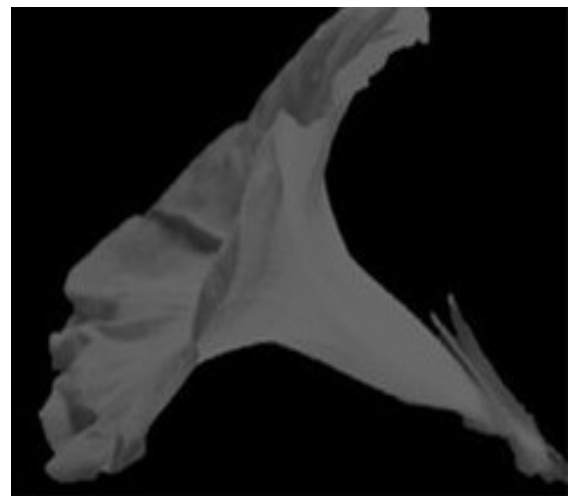
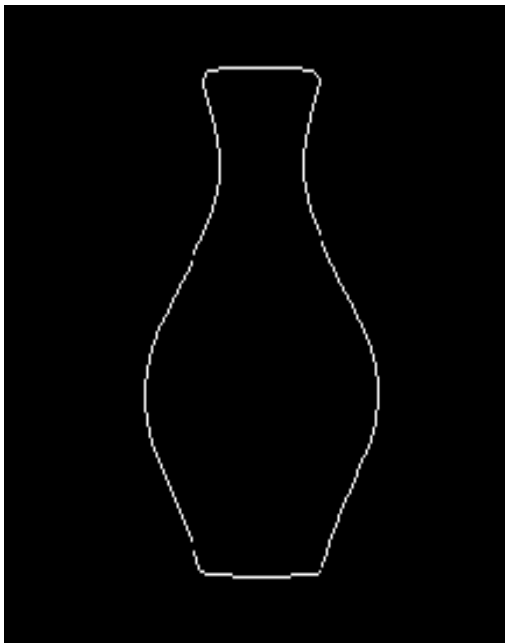
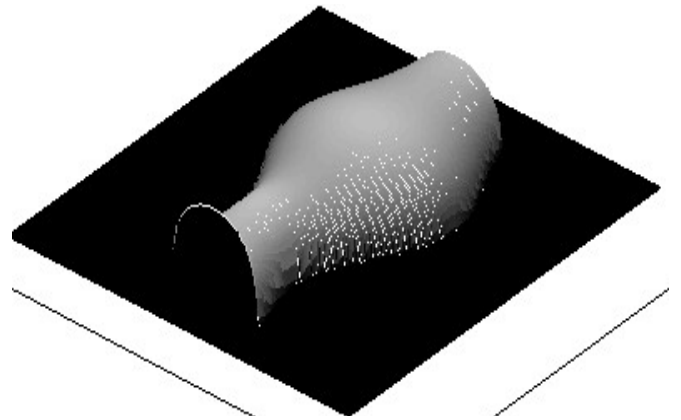
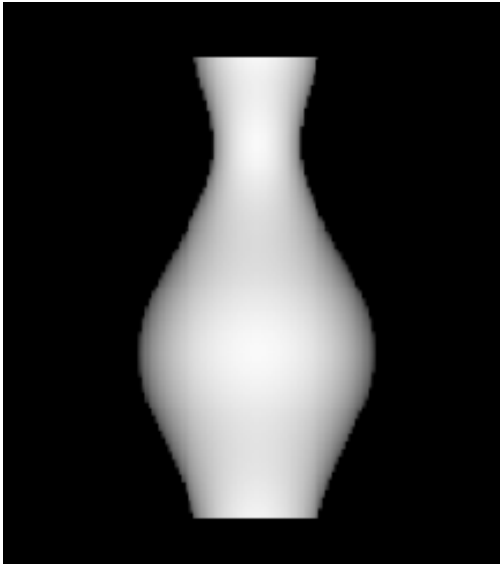
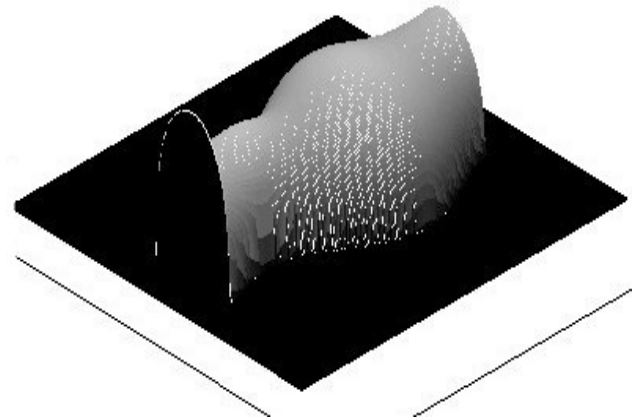
$$f(z, x, y) \approx f(z^{n-1}, x, y) + \frac{dz}{dz} z^{n-1} x, y$$

$$z^n(x, y) = z^{n-1}(x, y) + \frac{-f(z^{n-1}, x, y)}{\frac{d}{dz} z^{n-1}(x, y)}$$

$$\frac{df(z^{n-1}, x, y)}{dz(x, y)} = \frac{p_i + q_i}{\sqrt{p_i^2 + q_i^2} + \sqrt{p^2 + q^2}} - \frac{pp_i + qq_i}{\sqrt{p_i^2 + q_i^2} + \sqrt{p^2 + q^2}}$$

$$z^0(x, y) = 0$$

Algorithm	Error of height		Time/s
	<i>average</i>	<i>maximum</i>	





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*IEEE Proceedings of Computer Vision and Pattern
Recognition*

fl
IAPR International Workshop on Machine Vision Applications

Image and Vision Computing Journal

International Journal of Computer Vision

Pattern

Recognition Letters

International Journal of Computer Vision

*Some Remarks on Perspective
Shape From Shading Models*

Computer Vision—Algorithm and System.

Journal of Image and Graphics

Computer Vision and Image Understanding

*Computer Vision Graphics And Image Processing: Image
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Journal of Mathematical Imaging and Vision

*Engineering in Medicine and Biology Society
(EMBC), 2010 Annual International Conference of the IEEE*

fl *Computer Vision and Image Understanding*

*8th European Conference on Computer Vision, Prague, Czech
Republic,*

*IEEE
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