Phase Shifting Structured Light Method

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Stereo-vision with Structured Light



Figure: The fringe pattern for the digital projector or the LCD display. The left and right camera optical centers and image planes are (O_1, I_1) and (O_2, I_2) respectively. The projector optical center and image plane are (O_p, I_p) .

Projector Fringe Pattern



(d). $V_1(x, y)$ (e). $V_2(x, y)$ (f). $V_3(x, y)$

Figure: The fringe pattern for the digital projector or the LCD display.

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The horizontal fringe patterns for the projector are given by

$$H_k(u,v) = a + b\cos\left(\Phi(u,v) + \Delta\varphi_k\right), \quad k = 1, 2, 3 \tag{1}$$

where a = 128 and b = 127,

$$(\Delta \varphi_1, \Delta \varphi_2, \Delta \varphi_3) = \left(-\frac{2\pi}{3}, 0, \frac{2\pi}{3}\right),$$
 (2)

and the absolute phase

$$\Phi(u,v) = \frac{2\pi u}{\lambda},\tag{3}$$

where λ is the wavelength.

The vertical fringe patterns for the projector are given by

$$V_{k}(u,v) = a + b\cos(\Phi(u,v) + \Delta\varphi_{k}), \quad k = 1,2,3$$
(4)

where a = 128 and b = 128,

$$(\Delta\varphi_1, \Delta\varphi_2, \Delta\varphi_3) = \left(-\frac{2\pi}{3}, 0, \frac{2\pi}{3}\right), \tag{5}$$

and the absolute phase

$$\Phi(u,v) = \frac{2\pi v}{\lambda},\tag{6}$$

where λ is the wavelength.

Fringe Images



(a). $I_1(x, y)$

(b). $I_2(x, y)$

(c). $I_3(x, y)$

Figure: Fringe images: the top row shows the images captured by the left camera, the bottom row shows those by the right camera.

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The image intensity is formulated as

$$I_{1}(x, y) = I'(x, y) + I''(x, y) \cos(\Phi(x, y) - 2\pi/3)$$

$$I_{2}(x, y) = I'(x, y) + I''(x, y) \cos(\Phi(x, y))$$

$$I_{3}(x, y) = I'(x, y) + I''(x, y) \cos(\Phi(x, y) + 2\pi/3)$$
(7)

where I'(x, y) is the ambient, I''(x, y) the intensity modulation, $\Phi(x, y)$ the absolute phase.

The ambient, the modulation and the relative phase can be obtained by

$$l'(x,y) = \frac{1}{3}[l_1(x,y) + l_2(x,y) + l_3(x,y)]$$

$$l''(x,y) = \frac{1}{3}\sqrt{3(l_1 - l_3)^2 + (2l_2 - l_1 - l_3)^2}$$

$$\varphi(x,y) = \tan^{-1}\frac{\sqrt{3}(l_1 - l_3)}{2l_2 - l_1 - l_3}$$
(8)

where the relative phase $\varphi(x,y)$ is from $-\pi$ to π ,

$$\varphi(x,y) = \Phi(x,y) \mod 2\pi.$$

Image Decomposition



ambient I'(x,y) modulation I''(x,y) wrapped phase $\varphi(x,y)$

Figure: The ambient, modulation and wrapped phase computed from the fringe images in Fig. 3.

Problem (Phase Unwrapping)

Given an image \mathcal{I} of m rows and n columns, each pixel position is represented as a pair of indices (i, j), where $1 \leq i \leq m, 1 \leq j \leq n$. We use $p \in [1, m] \times [1, n]$ to represent a point in the image plane. The wrapped phase at the pixel p is denoted as φ_p , the wrap uncount at p as k_p , the unwrapped phase Φ_p , then

$$\Phi_{p} = \varphi_{p} + 2\pi k_{p}, \quad \forall p \in \mathcal{I}.$$
(9)

The wrap count function $k : \mathcal{I} \to \mathbb{Z}$ is the unkown function.

Phase Unwrapping - Multiple Wavelength Method

One can use multiple projection fringe pattern with different wavelengths to recover the absolute phase. The wavelength λ_i 's satisfy the relation

$$\lambda_i=2\lambda_{i+1},\quad i=0,1,2,\ldots,n.$$

 λ_0 is big enough to cover the whole scanning range, so the relative phase φ_0 equals the absolute phase Φ_0 , $\Phi_0 \leftarrow \varphi_0$. By the relation

$$\Phi_i(p)\lambda_i = \Phi_{i+1}(p)\lambda_{i+1} \implies 2\Phi_i = \Phi_{i+1}$$

We obtain

$$\Phi_{i+1} - \varphi_{i+1} = 2\Phi_i - \varphi_{i+1} \implies k_{i+1} = \operatorname{round}\left(\frac{\Phi_i}{\pi} - \frac{\varphi_{i+1}}{2\pi}\right)$$

and

$$\Phi_{i+1} = 2\pi k_{i+1} + \varphi_{i+1}$$

Multiple Wavelength



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Phase Unwrapping - Double Wavelength Method

One can use double projection fringe pattern with different wavelengths to recover the absolute phase. The wavelengths are λ_1 and λ_2 , $\lambda_1 < \lambda_2$, then

$$\Phi_1(p) - \Phi_2(p) = \frac{2\pi u}{\lambda_1} - \frac{2\pi u}{\lambda_2} = \frac{2\pi u}{\lambda_{eq}} = \Phi_{eq}$$
(10)

where λ_{eq} is the equivalent wavelength,

$$\lambda_{eq} = \frac{\lambda_1 \lambda_2}{\lambda_2 - \lambda_1} \tag{11}$$

Once can choose λ_1 and λ_2 to be close enough, such that λ_{eq} covers the whole scanning range. So

$$\Phi_{eq} = \varphi_{eq} = \varphi_1 - \varphi_2 \quad (\mod 2\pi),$$

and

$$\Phi_{eq}\lambda_{eq} = \Phi_i\lambda_i \implies k_i = \frac{1}{2\pi} \left(\Phi_{eq} \frac{\lambda_{eq}}{\lambda_i} - \varphi_i \right)$$

Double Wavelength Method



 φ and Φ $\lambda_1 = 18$

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Treat the modulation as the quality of the pixel.

- Pick the pixel p with the highest quality as the base pixel,
 Φ(p) ← φ(p); insert p to the priority queue Q;
- ${f 2}$ while the Q is not empty
 - **1** $p \leftarrow Q.pop();$
 - **2** Find each unprocessed neighbor pixel q of p, unwrap the phase of q

$$k(q) \leftarrow \operatorname{argmin}_{k \in \mathbb{Z}} (\Phi(p) - (\varphi(q) + 2\pi k))^2;$$

 $\Phi(q) \leftarrow \varphi(q) + 2\pi k(q);$

Solution Insert q to the priority queue Q

The quality guided path follower is vulnerable to bad pixels.

Phase Unwrap - Path Follower



Figure: phase unwrapping mistake.

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Phase Unwrap - Path Follower



Figure: phase unwrapping mistake.

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The energy for wrap count function $k:\mathcal{I} \to \mathbb{Z}$ is defined as

$$E(k) := \sum_{e_{pq}} |\Phi_p - \Phi_q|^2, \qquad (12)$$

where e_{pq} is the edge connecting the pixel p and the pixel q.

$$\begin{split} E(k) &:= \sum_{e_{pq}} E_{pq} \\ E_{pq} &:= |(\varphi_p + 2\pi k_p) - (\varphi_q + 2\pi k_q)|^2 \\ &= \sum_p k_p \sum_{e_{pq}} 4\pi (\varphi_p - \varphi_q) + \sum_{e_{pq}} 4\pi^2 (k_p - k_q)^2. \end{split}$$

The energy based phase unwrapping problem is formulated as a combinatorial optimization method using Ising model, which is a mathematical model of ferromagnetism in statistical mechanics to study phase transition.



Figure: Ising model for phase transition.

The phase unwrapping energy statisfies the following property:

$$E_{pq}(1,0) + E_{pq}(0,1) - E_{pq}(0,0) - E_{pq}(1,1) \ge 0,$$
(13)

the energy is sub-modular. So the energy is graph representable.

- Each pixel p is represented by a node n_p ;
- Each pair of adjacent pixels *p*, *q* are represented as two oriented edges e_{pq} and e_{qp} ;
- One source noce is added as S; One sink is added as T;
- For each pixel p, an oriented edge e_{Sp} is added, and an oriented edge e_{pT} is added.



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Definition (Network)

A network consists of

- a finite directed graph N = (V, E), where V denotes the finite set of vertices and E ⊂ V × V is the set of directed edges;
- a source $s \in V$ and a sink $t \in V$;
- a capacity function, which is a mapping $c : E \to \mathbb{R}^+$ denoted by c(u, v) for $(u, v) \in E$, representing the maximum amount of flow that can pass through an edge.

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Definition (Flows)

A flow through a network is a mapping $f : E \to \mathbb{R}^+$ denoted as f(u, v), subject to the following constraints:

- Capacity constraint: for every edge $(u, v) \in E$, $f(u, v) \leq c(u, v)$;
- Conservation of flows: for each vertex v apart from s and t, the following equality holds:

$$\sum_{(u,v)\in E} f(u,v) = \sum_{(v,w)\in E} f(v,w).$$

The value of flow is defined by

$$|f| = \sum_{(s,v)\in E} f(S,v) = \sum_{(v,t)\in E} f(v,T).$$

Definition (Cut)

An s - t cut C = (S, T) is a partition of V, such that $s \in S$ and $t \in T$. The cut-set X_C of a cut C is the set of oriented edges that connect the source part to the sink part:

$$X_C := \{(u,v) \in E : u \in S, v \in T\} = (S \times T) \cap E.$$

The *capacity* of an s - t cut is the sum of the capacities of the edges in its cut-set,

$$c(S,T) = \sum_{(u,v)\in X_C} c(u,v).$$

Problem (Max Flow)

Given a network, maximize |f|, route as much flow as possible from s to t.

Problem (Min Cut)

Given a network, minimize c(S, T), determine S and T such that the capacity of the s-t cut is minimal.

Theorem (Max-flow min-cut)

The maximum value of an s-t flow is equal to the minimum capacity over all s-t cuts.

Define the residue network $G_f(V, E_f)$ to be the network with capacity $c_f(u, v) = c(u, v) - f(u, v)$ and no flow.

Ford-Fulkerson Algorithm

Input: Given a network G = (V, E) with flow capacity c, a source node s and a sink node (t);

Output: Compute a flow f from s to t of maximum value;

•
$$f(u, v) \leftarrow 0$$
 for all edges (u, v)

- While there is a path p from s to t in G_f, such that c_f(u, v) > 0 for all edges (u, v) ∈ p; (p can be found using BFS or DFS)
 - Find $c_f(p) = \min\{c_f(u, v) : (u, v) \in p\}$
 - **2** For each edge $(u, v) \in p$

$$f(u, v) \leftarrow f(u, v) + c_f(p)$$

$$f(v, u) \leftarrow f(v, u) - c_f(p)$$

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Combinatorial Optimization

Edge energy

$$E_{pq} = k_p 4\pi (\varphi_p - \varphi_q) + k_q 4\pi (\varphi_q - \varphi_p) + 4\pi^2 (k_p - k_q)^2 + (\varphi_p - \varphi_q)^2$$

= $\psi_p(k_p) + \psi_q(k_q) + \psi_{pq}(k_p, k_q).$



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Combinatorial Optimization

Graph Construction

- Solution For each edge (p, q) ∈ I, construct a graph G_{p,q} with a source, since, the edge capacities are given by ψ_p(k_p), ψ_q(k_q) and ψ_{pq}(k_p, k_q);
- **②** The graph for the whole image $G_{\mathcal{I}}$ is obtained by merging the edge graphs

$$G_{\mathcal{I}} = \bigcup_{(p,q)\in E} G_{pq},$$

the capacities on the same edge are accumulated together;

Global Optimization

- At each step, the edge capacity is updated by the one-step optimal wrap count;
- The optimal wrap count (for one step) is obtained by the max-flow/min-cut algorithm;
- S The global optimal wrap count can be found by iterations.

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Figure: Computational example.



Figure: Phase unwrapping based on combinatorial optimization.



Figure: Phase unwrapping based on combinatorial optimization.



Figure: Reconstructed point cloud.



Figure: phase shifting structured light.



Figure: Reconstructed point cloud.

One can assume the scanned surface is smooth, the smoothness is measured by the harmonic energy. Since $\Phi(p) = \varphi + 2\pi k_p$, locally, we obtain $\nabla \Phi = \nabla \varphi$. Due to the noise, $\nabla \varphi$ may not be a gradient field. By Hodge decomposition,

$$\nabla \varphi = d\Phi + \nabla \cdot \mathbf{v},\tag{14}$$

for some vector field v, namely $\nabla\cdot v$ is the divergence free component. Therefore, we obtain

$$\Delta \Phi(p) = \nabla \cdot \nabla \varphi(p). \tag{15}$$

The absolute phase Φ can be obtained by solving the Poisson equation.

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We sample Ω by regular grid points with horizontal and vertical step lengths h_x and h_y respectively, the finite difference operators are defined as

$$\begin{aligned} \mathcal{D}_{xx}^{2} u_{ij} &= \frac{1}{h_{x}^{2}} (u_{i+1,j} + u_{i-1,j} - 2u_{i,j}) \\ \mathcal{D}_{yy}^{2} u_{ij} &= \frac{1}{h_{y}^{2}} (u_{i,j+1} + u_{i,j-1} - 2u_{i,j}) \\ \mathcal{D}_{xy}^{2} u_{ij} &= \frac{1}{4h_{x}h_{y}} (u_{i+1,j+1} + u_{i-1,j-1} - u_{i-1,j+1} - u_{i+1,j-1}) \end{aligned}$$

Then we obtain discrete Poisson equation:

$$u_{i+1,j} + u_{i-1,j} + u_{i,j+1} + u_{i,j-1} - 4u_{i,j} = \rho_{i,j},$$

with Neumann boundary condition.

Energy Method - Poisson Equation

The 2D Discrete Cosine Transformation is defined as

$$\tilde{u}(m,n) = c(m,n) \sum_{i,j} \sum_{i,j} \cos \frac{(2i+1)m\pi}{2M} \cos \frac{(2j+1)n\pi}{2N},$$

the Inverse Discrete Cosine Transformation is given by

$$u(i,j) = \sum_{m,n} c(m,n) \tilde{u}(m,n) \cos \frac{(2i+1)m\pi}{2M} \cos \frac{(2j+1)n\pi}{2N},$$

where m, i = 0, 1, ..., M - 1 and n = 0, 1, ..., N - 1,

$$c(m,n) = \left\{ egin{array}{c} rac{\sqrt{2}}{\sqrt{MN}} & m=0, n=0 \ rac{2}{\sqrt{MN}} & ext{otherwise} \end{array}
ight.$$

Given a discrete Poisson equation with Neumann boundary condition:

$$\left\{\begin{array}{rcl}
\Delta u &=& \rho \\
\frac{\partial u}{\partial n} &=& 0
\end{array}\right.$$

Let $\tilde{\rho} = DCT(\rho)$, $\tilde{u} = DCT(u)$, then we have

$$\widetilde{\mu}(m,n) = rac{\widetilde{
ho}(m,n)}{2[\cosrac{m\pi}{M} + \cosrac{n\pi}{N} - 2]}$$

Different solutions differ by a constant. Let $\tilde{u}(0,0) = 0$, we get the unique solution. The DCT, IDCT can be computed using FFT on GPUs.



Figure: phase shifting structured light.

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Figure: Reconstructed point cloud.



Figure: phase shifting structured light.



Figure: Reconstructed point cloud.