

# Ultra-Precision 6DoF Pose Estimation Using 2-D Interpolated Discrete Fourier Transform

## Supplementary Material

### A. Detailed Model Formulations

#### A.1. Full Camera Model Derivation

The standard pinhole camera model is mathematically represented as

$$s \begin{bmatrix} x_i \\ y_i \\ 1 \end{bmatrix} = \begin{bmatrix} f & 0 & 0 \\ 0 & f & 0 \\ 0 & 0 & 1 \end{bmatrix} [\mathbf{R} \quad \mathbf{t}] \begin{bmatrix} x_w \\ y_w \\ z_w \\ 1 \end{bmatrix} \quad (\text{S1})$$

Here  $[x_w, y_w, z_w, 1]^T$  and  $[x_i, y_i, 1]^T$  denote the homogeneous coordinates of a point in the world coordinate frame and image coordinate frame, respectively.  $f$  represents the focal length of the camera and  $s$  is a scale factor.  $\mathbf{R}$  and  $\mathbf{t}$  are the rotation and translation matrices respectively that define the transformation from the world coordinate frame to the camera coordinate frame. They can be expressed as

$$\begin{aligned} \mathbf{R} &= \mathbf{R}_z(\alpha) \mathbf{R}_y(\beta) \mathbf{R}_x(\gamma) \\ \mathbf{t} &= [t_x, t_y, t_z]^T \end{aligned} \quad (\text{S2})$$

where  $\alpha$ ,  $\beta$ , and  $\gamma$  are the rotation angles about the  $z$ ,  $y$ , and  $x$  axes and  $t_x$ ,  $t_y$ , and  $t_z$  are the translations along the  $x$ ,  $y$ , and  $z$  axes, respectively.

Assuming the pattern plane coincides with the  $xy$  plane of the world coordinate frame (i.e.,  $z_w = 0$ ), and letting  $\mathbf{r}_j$  denote the  $j$ -th column vector of  $\mathbf{R}$ , Eq. (S1) becomes

$$s \begin{bmatrix} x_i \\ y_i \\ 1 \end{bmatrix} = \begin{bmatrix} f & 0 & 0 \\ 0 & f & 0 \\ 0 & 0 & 1 \end{bmatrix} [\mathbf{r}_1 \quad \mathbf{r}_2 \quad \mathbf{t}] \begin{bmatrix} x_w \\ y_w \\ 1 \end{bmatrix} \quad (\text{S3})$$

For  $z_w = 0$ ,  $s = t_z - x_w \sin \beta + y_w \cos \beta \sin \gamma$ .

#### A.2. Pattern Model Fourier Expansions

Checkerboard and grid patterns are superpositions of sinusoidal components. The spatial representation of a checkerboard pattern in the world coordinate frame is defined as a piecewise function:

$$g(x_w, y_w) = \begin{cases} A, & 0 \leq \frac{2\pi x_w}{T_x} < \pi, 0 \leq \frac{2\pi y_w}{T_y} < \pi \\ A, & -\pi \leq \frac{2\pi y_w}{T_y} < 0, -\pi \leq \frac{2\pi x_w}{T_x} < 0 \\ 0, & \text{otherwise} \end{cases} \quad (\text{S4})$$

The Fourier series expansion of  $g(x_w, y_w)$  is given by

$$g(x_w, y_w) = \frac{A}{2} + \frac{4A}{\pi^2} \sum_{n=1}^{\infty} \sum_{m=1}^{\infty} \frac{1}{mn} \left[ \cos \left( 2\pi n \frac{x_w}{T_x} - 2\pi m \frac{y_w}{T_y} \right) \right. \\ \left. + \cos \left( 2\pi n \frac{x_w}{T_x} + 2\pi m \frac{y_w}{T_y} + \pi \right) \right] \quad (\text{S5})$$

where summations are performed on odd positive integers  $m$  and  $n$ .

Similarly, the spatial representation of a grid pattern in the world coordinate frame is

$$g(x_w, y_w) = \begin{cases} A, & 0 \leq \frac{2\pi x_w}{T_x} < \pi, 0 \leq \frac{2\pi y_w}{T_y} < \pi \\ 0, & \text{otherwise} \end{cases} \quad (\text{S6})$$

The corresponding Fourier series expansion is

$$g(x_w, y_w) = \frac{A}{4} + \frac{A}{\pi} \sum_{m=1}^{\infty} \frac{\sin \left( 2\pi m \frac{y_w}{T_y} \right)}{m} + \frac{A}{\pi} \sum_{n=1}^{\infty} \frac{\sin \left( 2\pi n \frac{x_w}{T_x} \right)}{n} \\ + \frac{2A}{\pi^2} \sum_{n=1}^{\infty} \sum_{m=1}^{\infty} \frac{\cos \left( 2\pi n \frac{x_w}{T_x} - 2\pi m \frac{y_w}{T_y} \right) - \cos \left( 2\pi n \frac{x_w}{T_x} + 2\pi m \frac{y_w}{T_y} \right)}{mn} \quad (\text{S7})$$

where the summations in the final term are again performed on odd positive integers  $m$  and  $n$ .

### B. Pose Estimation Model Derivation

#### B.1. Quasi-Orthographic Projection Model Derivation

Starting from Eq. (S3) and expanding  $s$ , we obtain

$$x_i = f \frac{x_w r_{11} + y_w r_{12} + t_x}{-x_w \sin \beta + y_w \cos \beta \sin \gamma + t_z} \quad (\text{S8})$$

$$y_i = f \frac{x_w r_{21} + y_w r_{22} + t_y}{-x_w \sin \beta + y_w \cos \beta \sin \gamma + t_z} \quad (\text{S9})$$

where

$$r_{11} = \cos \alpha \cos \beta$$

$$r_{12} = \cos \alpha \sin \beta \sin \gamma - \sin \alpha \cos \gamma$$

$$r_{21} = \sin \alpha \cos \beta$$

$$r_{22} = \sin \alpha \sin \beta \sin \gamma + \cos \alpha \cos \gamma$$

Under the quasi-orthographic assumption,  $t_z \gg x_w \sin \beta$  and  $t_z \gg y_w \cos \beta \sin \gamma$ . Thus, the denominator is  $\approx t_z$ . This leads to

$$\begin{aligned} \frac{x_i}{f} &= \frac{x_w \cos \alpha \cos \beta + y_w (\cos \alpha \sin \beta \sin \gamma - \sin \alpha \cos \gamma) + t_x}{t_z} \\ \frac{y_i}{f} &= \frac{x_w \sin \alpha \cos \beta + y_w (\sin \alpha \sin \beta \sin \gamma + \cos \alpha \cos \gamma) + t_y}{t_z} \end{aligned} \quad (\text{S11})$$

which is Eq. (3) in the main paper.

## B.2. Relationships between Pattern Parameters and 6DoF Pose

Substituting Eq. (S11) into Eq. (2) yields the pattern image expression in Eq. (4). The full expressions for  $\omega_{x_1}, \omega_{y_1}, \omega_{x_2}, \omega_{y_2}, \varphi_1$ , and  $\varphi_2$  are

$$\omega_{x_1} = 2\pi \frac{t_z}{f} \left( \frac{\sin \alpha \sin \beta \sin \gamma + \cos \alpha \cos \gamma}{T_x \cos \beta \cos \gamma} - \frac{\sin \alpha}{T_y \cos \gamma} \right) \quad (\text{S12a})$$

$$\omega_{y_1} = 2\pi \frac{t_z}{f} \left( \frac{\sin \alpha \cos \gamma - \cos \alpha \sin \beta \sin \gamma}{T_x \cos \beta \cos \gamma} + \frac{\cos \alpha}{T_y \cos \gamma} \right) \quad (\text{S12b})$$

$$\omega_{x_2} = 2\pi \frac{t_z}{f} \left( \frac{\sin \alpha \sin \beta \sin \gamma + \cos \alpha \cos \gamma}{T_x \cos \beta \cos \gamma} + \frac{\sin \alpha}{T_y \cos \gamma} \right) \quad (\text{S12c})$$

$$\omega_{y_2} = 2\pi \frac{t_z}{f} \left( \frac{\cos \alpha \sin \beta \sin \gamma - \sin \alpha \cos \gamma}{T_x \cos \beta \cos \gamma} + \frac{\cos \alpha}{T_y \cos \gamma} \right) \quad (\text{S12d})$$

$$\begin{aligned} \varphi_1 &= 2\pi \left( \frac{-t_x (\sin \alpha \sin \beta \sin \gamma + \cos \alpha \cos \gamma)}{T_x \cos \beta \cos \gamma} \right. \\ &\quad \left. + \frac{t_y (\cos \alpha \sin \beta \sin \gamma - \sin \alpha \cos \gamma)}{T_x \cos \beta \cos \gamma} + \frac{t_x \sin \alpha - t_y \cos \alpha}{T_y \cos \gamma} \right) \end{aligned} \quad (\text{S12e})$$

$$\begin{aligned} \varphi_2 &= 2\pi \left( \frac{-t_x (\sin \alpha \sin \beta \sin \gamma + \cos \alpha \cos \gamma)}{T_x \cos \beta \cos \gamma} \right. \\ &\quad \left. + \frac{t_y (\cos \alpha \sin \beta \sin \gamma - \sin \alpha \cos \gamma)}{T_x \cos \beta \cos \gamma} - \frac{t_x \sin \alpha - t_y \cos \alpha}{T_y \cos \gamma} \right) \end{aligned} \quad (\text{S12f})$$

## B.3. Solution of Parameter $\eta$

The constraint equation is

$$(\omega_{x_1} + \omega_{x_2})^2 + (\omega_{y_2} - \omega_{y_1})^2 = \frac{\eta^2 (1 - \cos^2 \beta \sin^2 \gamma)}{T_x^2 \cos^2 \beta \cos^2 \gamma} \quad (\text{S13})$$

Substituting expressions for  $\cos \beta$  and  $\cos \gamma$  (from Eq. (6)) into Eq. (S13) leads to the quadratic equation in  $\eta^2$

$$\rho_1 \eta^4 - \rho_2 \eta^2 + 1 = 0 \quad (\text{S14})$$

where

$$\begin{aligned} \rho_1 &= \frac{1}{T_x^2 T_y^2 (\omega_{y_1} + \omega_{y_2})^2 (\omega_{x_1} + \omega_{x_2} - (\omega_{y_2} - \omega_{y_1}) \tan \alpha)^2} \\ \rho_2 &= \frac{(\omega_{x_1} + \omega_{x_2})^2 + (\omega_{y_2} - \omega_{y_1})^2}{T_y^2 (\omega_{y_1} + \omega_{y_2})^2 (\omega_{x_1} + \omega_{x_2} - (\omega_{y_2} - \omega_{y_1}) \tan \alpha)^2} \\ &\quad + \frac{1}{T_x^2 ((\omega_{x_1} + \omega_{x_2}) \cos \alpha - (\omega_{y_2} - \omega_{y_1}) \sin \alpha)^2} \end{aligned} \quad (\text{S15})$$

Solving this quadratic equation gives

$$\eta = \sqrt{\frac{\rho_2 \pm \sqrt{\rho_2^2 - 4\rho_1}}{2\rho_1}} \quad (\text{S16})$$

## C. Pattern Parameter Estimation Algorithm (2D-IpDFT) Derivation

### C.1. Window Function Details

The 2D maximum sidelobe decay window function  $w(u, v)$  is separable:  $w(u, v) = w_1(u)w_1(v)$ , where  $w_1(\cdot)$  is a 1D

window. For a window of length  $L$ ,  $w_1(\cdot)$  has the form

$$w_1(x) = \sum_{h=0}^{H-1} (-1)^h a_h \cos \frac{2\pi x h}{L} \quad (\text{S17})$$

where the coefficients  $a_h$  are

$$a_0 = \frac{C_{2H-2}^{H-1}}{2^{2H-2}}, \quad a_h = \frac{C_{2H-2}^{H-h-1}}{2^{2H-3}}, \quad h = 1, \dots, H-1 \quad (\text{S18})$$

The DFT of  $w(u, v)$  can be expressed as

$$W(k, l) = W(k)W(l) = \frac{D(k)}{P(k)} \frac{D(l)}{P(l)} \quad (\text{S19})$$

where

$$\begin{aligned} D(\cdot) &= \frac{M(2H-2)!}{\pi 2^{2H-2}} \sin(\pi \cdot) e^{-j\pi \cdot} \\ P(\cdot) &= \cdot \prod_{h=1}^{H-1} (h^2 - \cdot^2) \end{aligned} \quad (\text{S20})$$

Note that the function  $D(\cdot)$  exhibits the property

$$D(\cdot - 1) = D(\cdot) = D(\cdot + 1) \quad (\text{S21})$$

### C.2. Full DFT of Windowed Image Signal

The DFT of the windowed image signal  $g(u, v)$  is

$$\begin{aligned} G(k, l) &= \frac{1}{2} [W(k - \lambda_1, l - \mu_1) e^{j\varphi_1} + W(k + \lambda_1, l + \mu_1) e^{-j\varphi_1} \\ &\quad + W(k - \lambda_2, l + \mu_2) e^{j\varphi_2} + W(k + \lambda_2, l - \mu_2) e^{-j\varphi_2}] \end{aligned} \quad (\text{S22})$$

### C.3. Linear System for Frequency Estimation

Rearranging Eq. (S22) and evaluating at DFT frequencies  $k-1, k, k+1$  yields a system of linear equations  $\mathbf{A}\mathbf{d} = \mathbf{b}$ , where

$$\mathbf{A} = \begin{bmatrix} \frac{1}{P(k-1-\lambda_1)} & \frac{1}{P(k-1+\lambda_1)} & G(k-1, l) \\ \frac{1}{P(k-\lambda_1)} & \frac{1}{P(k+\lambda_1)} & G(k, l) \\ \frac{1}{P(k+1-\lambda_1)} & \frac{1}{P(k+1+\lambda_1)} & G(k+1, l) \end{bmatrix} \quad (\text{S23})$$

$$\mathbf{d} = \begin{bmatrix} \frac{D(k-\lambda_1)D(l-\mu_1)e^{j\varphi_1}}{P(l-\mu_1)} \\ \frac{D(k+\lambda_1)D(l+\mu_1)e^{-j\varphi_1}}{P(l+\mu_1)} \\ -2 \end{bmatrix} \quad (\text{S24})$$

$$\mathbf{b} = \begin{bmatrix} -W(k-1-\lambda_2, l+\mu_2)e^{j\varphi_2} - W(k-1+\lambda_2, l-\mu_2)e^{-j\varphi_2} \\ -W(k-\lambda_2, l+\mu_2)e^{j\varphi_2} - W(k+\lambda_2, l-\mu_2)e^{-j\varphi_2} \\ -W(k+1-\lambda_2, l+\mu_2)e^{j\varphi_2} - W(k+1+\lambda_2, l-\mu_2)e^{-j\varphi_2} \end{bmatrix} \quad (\text{S25})$$

The condition  $\det(\mathbf{A}) = 0$  for a non-trivial solution of  $\mathbf{d}$  (when  $\mathbf{b}$  is considered part of the augmented matrix for homogeneous system or by Cramer's rule if solving for components of  $\mathbf{d}$ ) leads to the frequency estimation formula Eq. (9).