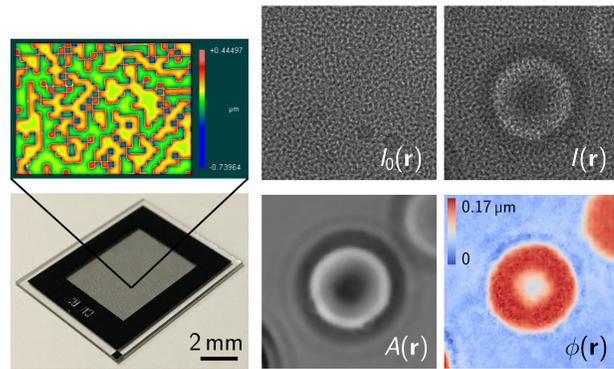
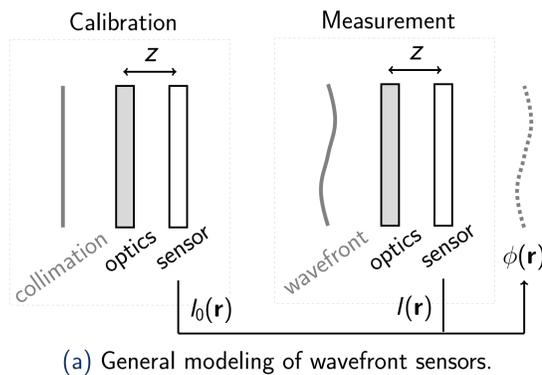


# Snapshot Incoherent Wavefront Sensing

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## Theory: A Model for Classical Wavefront Sensors



Wavefront sensors are instruments that retrieve phase  $\phi(\mathbf{r})$  from wavefront-encoded intensity measurements.

For incoherent light, classical wavefront sensors are grouped in two categories (probing slopes or curvature):

- **Slopes sensors:** Shack-Hartmann [1], lateral shearing interferometers [2], coded wavefront sensors [3, 4, 5].
- **Curvature sensors:** Curvature sensors [6], based on the Transport-of-Intensity Equation (TIE) [7].

Figure 1(a) models wavefront sensors consists of an encoding optics placed distance  $z$  away from an intensity sensor, and a numerical algorithm that decodes intensity images ( $I_0(\mathbf{r})$  and  $I(\mathbf{r})$ ) to retrieve wavefront  $\phi(\mathbf{r})$ . At wavelength  $\lambda$  (though it also works for incoherent light), when  $z$  is small, we show in either ray or wave optics, that the relationship between  $I_0(\mathbf{r})$  and  $I(\mathbf{r})$  is:

$$I\left(\mathbf{r} + \frac{\lambda z}{2\pi} \nabla \phi\right) = |A(\mathbf{r})|^2 \left(1 - \frac{\lambda z}{2\pi} \nabla^2 \phi\right) I_0(\mathbf{r}).$$

It is versatile, e.g., TIE can be derived via linearizing  $I\left(\mathbf{r} + \frac{\lambda z}{2\pi} \nabla \phi\right)$  around  $\mathbf{r}$  at the limit that  $z \rightarrow 0$ . Our model generalizes TIE in: (i) A concise theoretical model for two distanced planes; (ii) Nonlinearity; (iii) More degrees of freedom to allow for a customizable modulation mask (reflected as a user-defined reference image  $I_0(\mathbf{r})$ ).

Table 1: Our equation under different (approximated) forms as commonly seen formulas for each wavefront sensor.  $\delta_p(x)$  is the Dirac comb function with period  $p$  (the lenslet pitch).  $k = 2\pi/\lambda$  is the wave number.

Name	Optics	Commonly seen formula
Shack-Hartmann	micro-lens arrays $I_0(\mathbf{r}) = \delta_p(x)\delta_p(y)$	$I\left(\mathbf{r} + \frac{\lambda z}{2\pi} \nabla \phi\right) = I_0(\mathbf{r})$
Lateral shearing	sinusoid gratings (freq. $\omega$ ) $I_0(\mathbf{r}) = \cos^2(\omega x) \cos^2(\omega y)$	$I(\mathbf{r}) =  A(\mathbf{r}) ^2 I_0\left(\mathbf{r} - \frac{\lambda z}{2\pi} \nabla \phi\right)$
Curvature sensor	none $I_0(\mathbf{r}) = 1$	$\nabla I_2 \cdot \nabla \phi + I_1 \nabla^2 \phi = \frac{k}{2}(I_1 - I_2) \approx -k \frac{\partial I}{\partial z}$
Coded wavefront sensor	random gratings $I_0(\mathbf{r})$ are speckles	$I\left(\mathbf{r} + \frac{\lambda z}{2\pi} \nabla \phi\right) = I_0(\mathbf{r})$ in [5]; or our equation (this work)

Figure 1: (a) The general model. (b) The “optics” for our wavefront sensor (variant of [5]), which recovers amplitude  $A(\mathbf{r})$  and phase  $\phi(\mathbf{r})$  from reference  $I_0(\mathbf{r})$  and measured  $I(\mathbf{r})$ .

## Application: Snapshot Incoherent Phase & Intensity Quantitative Phase Microscopy

Our theory reveals a potential to retrieve  $|A(\mathbf{r})|^2$  and  $\phi(\mathbf{r})$  directly from raw data such as from [5]. As a demonstration, by taking off the condenser lens, we turn an ordinary low-budget bright field microscopy into a simultaneous intensity and phase microscopy, under collimated halogen lamp (HPLS245, Thorlabs) illumination. A prototype coded wavefront sensor is employed which consists of a bare sensor (1501M-USB-TE, Thorlabs) and a random binary phase mask (Fig. 1(b): pixel size  $12.9 \mu\text{m}$ , either 0 or  $\pi$  phase modulation at  $\lambda = 550 \text{ nm}$ , placed  $z = 1.43 \text{ mm}$  away from the sensor). Figure 2 shows simultaneous amplitude and phase recovery of transparent thin cells. Note how the speckles have been removed and the plausibility of the reconstructed smooth phase maps.

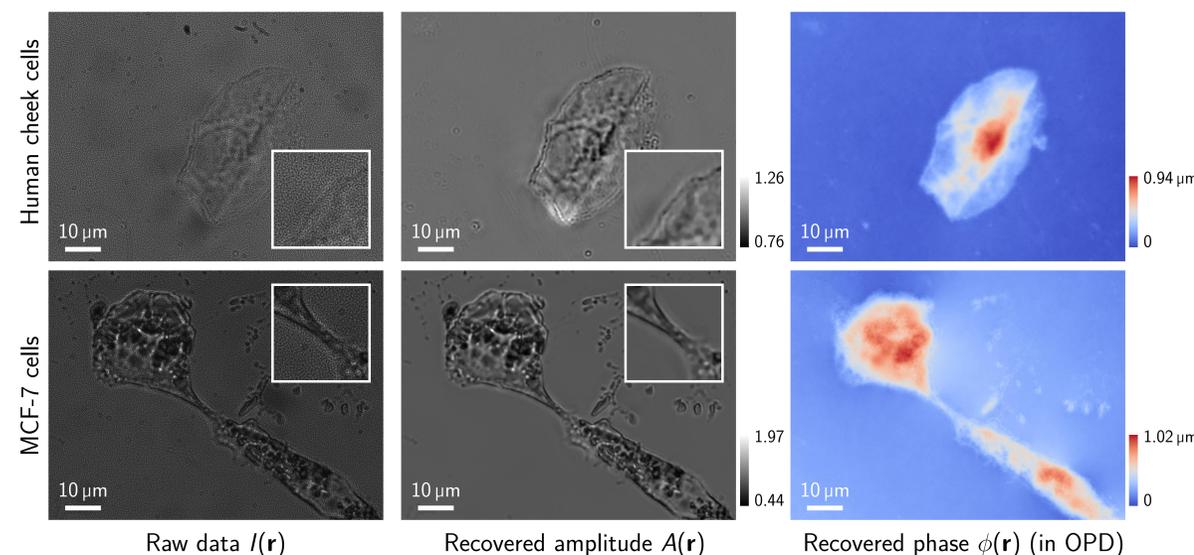


Figure 2: Quantitative phase imaging using our wavefront sensor. Images were taken under an  $\times 100$  Mitutoyo plan apochromat objective, 0.70 NA. Inset close-up images show that the speckle patterns have been fully removed from the original raw data.

## Computation: Numerical Solver

The forward model gives  $N$  equations but number of unknowns is  $2N$  ( $A$  and  $\phi$ ). For input images  $I_0(\mathbf{r})$  and  $I(\mathbf{r})$ , let  $|\tilde{A}|^2 = |A|^2 \left(1 - \frac{\lambda z}{2\pi} \nabla^2 \phi\right)$ , we solve the following optimization problem to obtain  $\tilde{A}$  and  $\phi$ :

$$\min_{\tilde{A}, \phi} \left\| I\left(\mathbf{r} + \frac{\lambda z}{2\pi} \nabla \phi\right) - |\tilde{A}|^2 I_0(\mathbf{r}) \right\|_2^2 + \Gamma_{\text{prior}}(\phi, |\tilde{A}|^2),$$

where the prior terms are (parameters  $\alpha, \beta, \gamma, \tau$ ):

$$\Gamma_{\text{prior}}(\phi, |\tilde{A}|^2) = \Gamma_{\text{phase}}(\phi) + \Gamma_{\text{intensity}}(|\tilde{A}|^2),$$

$$\Gamma_{\text{phase}}(\phi) = \alpha \|\nabla \phi\|_1 + \beta \left\| \frac{\nabla}{\nabla^2} \right\| \phi\|_2^2,$$

$$\Gamma_{\text{intensity}}(|\tilde{A}|^2) = \gamma \left\| \frac{\nabla}{\nabla^2} \right\| |\tilde{A}|^2\|_1 + \tau \left\| \frac{\nabla}{\nabla^2} \right\| |\tilde{A}|^2\|_2^2.$$

Using modern optimization the ADMM scheme [8], our solver typically elapses within 100 ms on a Nvidia Titan X (Pascal) GPU for megapixel input images.

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