

Lecture 17: Wave optics and Fourier optics

Machine Learning and Imaging

BME 548L Roarke Horstmeyer

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What is light and how can we model it?

- Interpretation #1: Radiation (Incoherent)
- Model: Rays







- Interpretation #2: Electromagnetic wave (Coherent)
- Model: Waves





• Real, non-negative

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- Models absorption and brightness
 - $\mathbf{I}_{\text{tot}} = \mathbf{I}_1 + \mathbf{I}_2$

- Complex field
- Models Interference

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\mathsf{E}_{\mathrm{tot}} = \mathsf{E}_1 + \mathsf{E}_2
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- Interpretation #3: Particle
- Model: Photons



Simple mathematical model of incoherent image formation



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Simple mathematical model of incoherent image formation



Summary: simple physical layers for incoherent imaging

- Deal with sample/image intensities I, real and non-negative ۲
- Effect of illumination is element-wise multiplication ٠ λ
- Imaging systems blur the object via point-spread function matrix **H**

- Discrete pixels down-sample the object via
- Add noise into measurement $\mathbf{I}_{\mathbf{N}}(\mathbf{x},\mathbf{y}) = \mathbf{D} \mathbf{I}_{\mathbf{0}}(\mathbf{x},\mathbf{y}) + \mathbf{N}$ ٠
- Different colors add linearly ٠

$$I_{s}(x, y) = \sum I_{0}(x, y, \lambda)$$



$$\mathbf{I_{d}}(\mathbf{x},\mathbf{y}) = \mathbf{D} \mathbf{I_{0}}(\mathbf{x},\mathbf{y})$$

 $\mathbf{H}_{\mathbf{b}}(\mathbf{x},\mathbf{y}) = \mathbf{H}_{\mathbf{0}}(\mathbf{x},\mathbf{y})$



Example #1: Optimized illumination pattern (one color)





Example #2: Optimized color filter for a grayscale camera



Training data: $[I_0(x, y, \lambda), y]$ $I_0:100 \times 100 \text{ pix. x } 30$ Label y: 1x3 - pepper, broccoli, green beans

$$I_{s}(x, y) = \sum_{\lambda} W_{0}(\lambda) I_{0}(x, y, \lambda)$$

Physical Layer



Example 3: learned illumination pattern for improved segmentation



U-Net Architecture

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Example 3: learned illumination pattern for improved segmentation





Example 3: learned illumination pattern for improved segmentation



*If we allow w's here to be trainable weights, then we can find ideal brightnesses for different LEDs to illuminate a sample of interest!

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Example 3: learned illumination pattern for improved segmentation



Optimized illumination for nuclei segmentation



Standard illumination

Learned illumination



See C. Cooke et al., "Physics-enhanced machine learning for virtual fluorescence microscopy," ICCV (2021)

Multiple Patterns for Fluorescence image inference



4 Patterns

 2^{6}

25

 2^{4}



See C. Cooke et al., "Physics-enhanced machine learning for virtual fluorescence microscopy," ICCV (2021)

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Maxwell's equations without any charge

$$\nabla \times \vec{\mathcal{E}} = -\mu \frac{\partial \vec{\mathcal{H}}}{\partial t}$$
$$\nabla \times \vec{\mathcal{H}} = \epsilon \frac{\partial \vec{\mathcal{E}}}{\partial t}$$
$$\nabla \cdot \epsilon \vec{\mathcal{E}} = 0$$
$$\nabla \cdot \mu \vec{\mathcal{H}} = 0.$$



Maxwell's equations without any charge

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$$\nabla \cdot \epsilon \vec{\mathcal{E}} = 0$$
$$\nabla \cdot \mu \vec{\mathcal{H}} = 0.$$

- 1. Take the curl of both sides of first equation
- 2. Substitute 2nd and 3rd equation
- 3. Arrive at the wave equation:

$$\nabla^2 \vec{\mathcal{E}} - \frac{n^2}{c^2} \frac{\partial^2 \vec{\mathcal{E}}}{\partial t^2} = 0 \qquad n = \left(\frac{\epsilon}{\epsilon_0}\right)^{1/2} \qquad c = \frac{1}{\sqrt{\mu_0 \epsilon_0}}.$$



Considering light that isn't pulsed over time, we can use the following solution:

 $u(P,t) = A(P) \cos[2\pi\nu t + \phi(P)]$ $u(P,t) = Re\{U(P) \exp(-j2\pi\nu t)\},\$



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With this particular solution, we get the following important time-independent equation:

Helmholtz Equation

$$(\nabla^2 + k^2)U = 0. \qquad k = 2\pi n \frac{\nu}{c} = \frac{2\pi}{\lambda},$$

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Let's take a step back: how does light propagate?

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Helm Equa

The pholous pholographic states
$$(\nabla^2 + k^2)U = 0.$$
 $k = 2\pi n \frac{\nu}{c} = \frac{2\pi}{\lambda},$

This is an important equation in physics. We won't go into the details, but it leads to the Huygen-Fresnel principle:

$$U(P_2) = \frac{1}{j\lambda} \iint_{\Sigma} U(P_1) \frac{\exp(jkr_{21})}{r_{21}} \cos\theta \, ds$$



The Huygens-Fresnel Equation

ation
$$U(P_2) = \frac{1}{j\lambda} \iint_{\Sigma} U(P_1) \frac{\exp(jkr_{21})}{r_{21}} \cos\theta \, ds$$



Aperture



The Huygens-Fresnel Equation

$$U(P_2) = \frac{1}{j\lambda} \iint_{\Sigma} U(P_1) \frac{\exp(jkr_{21})}{r_{21}} \cos\theta \, ds$$



Generally connects two points in 3D:

 $U(P_1) = U(x_1, y_1, z_1)$

 $U(P_2) = U(x_2, y_2, z_2)$

We are usually concerned about propagation between two planes (almost always in an optical system):

 $U(P_1) = U(x_1, y_1, z_1 = z_{p1})$ $U(P_2) = U(x_2, y_2, z_2 = z_{p2})$ $U(P) = E(x, y, z)e^{ikz}$





We are usually concerned about propagation between two planes (almost always in an optical system):



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Paraxial approximation:

$$abla_{ot}^2 U + 2ikrac{dU}{dz} = 0$$
 Substitute in $U(P) = E(x,y,z)e^{ikz}$ and crank the wheel,

 $\nabla_{\perp}^2 E + 2ik\frac{dE}{dz} + 2k^2 E = 0$

Paraxial Helmholtz Equation. This has an exact integral solution:



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Paraxial Helmholtz Equation. This has an exact integral solution:

$$E(x,y,z) = rac{e^{ikz}}{i\lambda z} \iint_{-\infty}^{+\infty} E(x',y',0) e^{rac{ik}{2z} \left[(x-x')^2 + (y-y')^2
ight]} dx' dy'$$

Fresnel diffraction integral

This is how light propagates from one plane to the next. It's a convolution!

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Fresnel light propagation as a convolution

$$E(x,y,z) = rac{e^{ikz}}{i\lambda z} \iint_{-\infty}^{+\infty} E(x',y',0) e^{rac{ik}{2z} \left[(x-x')^2 + (y-y')^2
ight]} dx' dy'$$

$$h(x,y,z)=rac{e^{ikz}}{i\lambda z}e^{irac{k}{2z}\left(x^2+y^2
ight)}$$

$$E(x,y,z)=E(x,y,0)st h(x,y,z)$$

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 $W_{d20} = -3\lambda/2$

Fresnel light propagation as a convolution

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 $W_{620} = 0$



From the Fresnel approximation to the Fraunhofer approximation

Fresnel Approximation:

$$E(x,y,z) = rac{e^{ikz}}{i\lambda z} \iint_{-\infty}^{+\infty} E(x',y',0) e^{rac{ik}{2z} \left[(x-x')^2 + (y-y')^2
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Lets assume that the second plane is "pretty far away" from the first plane. Then,





From the Fresnel approximation to the Fraunhofer approximation

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Lets assume that the second plane is "pretty far away" from the first plane. Then,



1. Expand the squaring

$$E(x,y,z) = \frac{e^{ikz}}{i\lambda z} \iint E(x',y',0) e^{\frac{ik}{2z}(x^2+y^2)} e^{\frac{ik}{2z}(x'^2+y'^2)} e^{\frac{ik}{2z}(xx'+yy')} dx' dy'$$



From the Fresnel approximation to the Fraunhofer approximation

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2. Front term comes out, assume second term goes away, then,

$$E(x,y,z) = C \iint E(x',y',0)e^{\frac{ik}{2z}(xx'+yy')}dx'dy'$$

$$C = \frac{e^{ikz}}{i\lambda z} e^{\frac{ik}{2z}(x^2 + y^2)}$$

Fraunhofer diffraction is a Fourier transform!!!!!!!

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This is the aperture



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function as an image

of the 2-D rectangle



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Cheetah

Zebra

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magnitude of zebra

phase of zebra

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Last piece of the puzzle: what happens from lens to sensor?





$E_s(x_s, y_s, 0)$ $E'_a(x_d, y_d) = E_a(x_d, y_d)A(x_d, y_d)$ 2D inverse Fourier Transform 2D Fourier Transform

Last piece of the puzzle: what happens from lens to sensor?

inverse Fourier transform!
This process should sound familiar....





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Can also model this using the Convolution Theorem





Two modeling choices for the camera:





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Linear systems and the black box

The optical black box system and the point-spread function:

Light $g_1(x_i, y_i)$ entering "black box" optical system modified by system point-spread function



$$g_2(x_2, y_2) = \iint_{-\infty}^{\infty} g_1(x_1, y_1) h(x_2 - x_1, y_2 - y_1) dx_1 dy_1$$

Assume shift invariance: This is the system point-spread function



Summary of two models for image formation

- Interpretation #1: Radiation (Incoherent)
- Model: Rays







- Real, non-negative
- Models absorption and brightness

$$\mathbf{I}_{tot} = \mathbf{I}_1 + \mathbf{I}_2$$

Mathematical model of for incoherent image formation

• All quantities are real, and non-negative

Object absorption:

S₀(x,y)

Illumination brightness:



100

photons

60%

transmission

60

photons

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• All quantities are real, and non-negative

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S₀(x,y)

Illumination brightness:

B(x,y)

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 $\mathbf{I}_{s} = \mathbf{B} \mathbf{S}_{0}$

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- Complex field
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$$E_{tot} = E_1 + E_2$$



• Pretty much the same thing, but now we have an amplitude and a complex phase





• Pretty much the same thing, but now we have an amplitude and a complex phase

Sample absorption = S(x,y)



 $C(x,y) = A_i(x,y) \exp[ik\varphi_i(x,y)] \quad U(x,y) = A_i(x,y) S(x,y) \exp[ik\varphi_t(x,y)]$

New: complex phase delay

- Needed to represent wave
- Represents wave delay across space

• Pretty much the same thing, but now we have an amplitude and a complex phase





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• Pretty much the same thing, but now we have an amplitude and a complex phase



• Pretty much the same thing, but now we have an amplitude and a complex phase

Sample absorption = S(x,y)



Output phase is sum of phase delays, product of phasors

 $\phi_t(x,y) = \phi(x,y) + \phi_i(x,y)$ $e^{ik\phi t(x,y)} = e^{ik\phi i(x,y)} \times e^{ik\phi(x,y)}$

Multiply phases!



Pretty much the same thing, but now we have an amplitude and a complex phase •

Sample absorption = S(x,y)



Conclusion:

Transmitted field = incident field x complex sample

 $A_{t}(x,y) \exp[ik\phi_{t}(x,y)] = A_{i}(x,y) \exp[ik\phi_{i}(x,y)] \times S(x,y) \exp[ik\phi(x,y)]$







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- Complex field
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$$\mathsf{E}_{tot} = \mathsf{E}_1 + \mathsf{E}_2$$

$$\mathbf{U} = \mathbf{C} \mathbf{S}_{\mathbf{0}}$$

J, C and S are complex



Additional Information about sample index of refraction, spatial frequency and Fourier optics

Microscope illumination and sample index of refraction

So far: illuminate the sample and create a field that is equivalent to the sample's absorption and phase – why is this true???



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Sample index of refraction $n(x,y,z) = 1 + ia(x) + \phi(x)$



*For more information, see D. Paganin, Coherent X-Ray Optics, Section 2.2

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Thin sample approximation:

Sample's effect on light is multiplication with exp[-ik * n(x,y)]

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In 1D: Emerging field U(x) = incident field $U_i(x)$ * sample function s(x)

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 $U(x) = U_i(x) *exp[-ik n(x)] = U_i(x) A(x) exp[ik\phi(x)]$ A(x) = exp[k a(x)]Horstmeyer (2024)absorptionphase shift: new term for laser



Sample absorption = A(x) Sample phase = exp[ikφ(x)]



Emerging field U = incident field $U_i(x)$ * sample function s(x)

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Q: When is the emerging field equal to the absorption and phase?





Q: When is the emerging field equal to the absorption and phase?

Sample absorption = A(x)Sample phase = $exp[ik\phi(x)]$

A: When the incident wave = 1, means uniform in amplitude and phase:

 $U_i(x) = 1 \longrightarrow U(x) = A(x) \exp[ik\phi(x)]$



Emerging field U = incident field $U_i(x)$ * sample function s(x)

Incident field U_i Emerging field U

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Q: When is the emerging field equal to the absorption and phase?

Sample absorption = A(x) Sample phase = exp[ikφ(x)]



A: When the incident wave = 1, means uniform in amplitude and phase:











From before: Spatial frequencies = "stripes" within each image







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Equivalent coordinates in the Fourier domain and at the Fourier plane



General rules for applying the Fourier transform in optics



Situation 1: From an object to a plane "really far away"



General rules for applying the Fourier transform in optics

Situation 1: From an object to a plane "really far away"



Situation 2: From an object to the back focal plane of the microscope objective lens





General rules for applying the Fourier transform in optics



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Situation 3: From an object to a plane 1 focal length away from a lens (1f-1f system)









f f 2D Fourier Transform

The Fourier plane provides a measure of the **ray angles at the image plane**



Image plane ray angle $\boldsymbol{\theta}$





The Fourier plane provides a measure of the **ray angles at the image plane**

Doesn't contain info about spatial distribution light



Image plane ray angle θ





The Fourier plane provides a measure of the **ray angles at the image plane**

Rays are leaving image plane at +20 degrees



Image plane ray angle θ





The Fourier plane provides a measure of the **ray angles at the image plane**

Rays are coming in at +20 degrees and -15 degrees



Image plane ray angle θ



You typically go between 4 functions to describe one imaging system:

