



Visual Data Science @  
**SILICON  
BEACH** 

# Introduction to Polarization for Rendering and Vision

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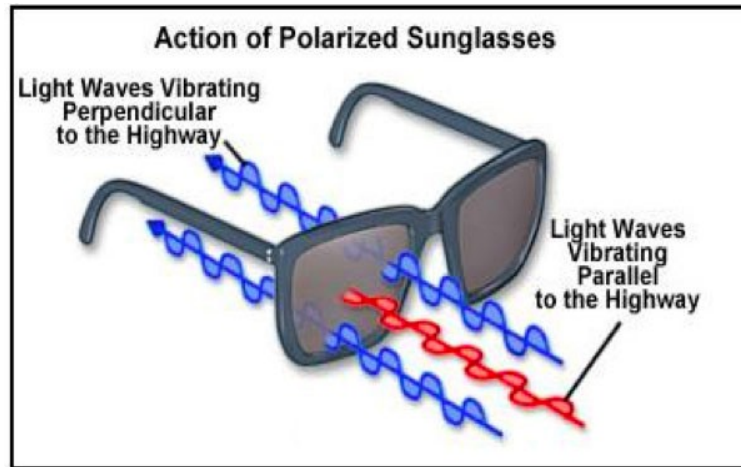
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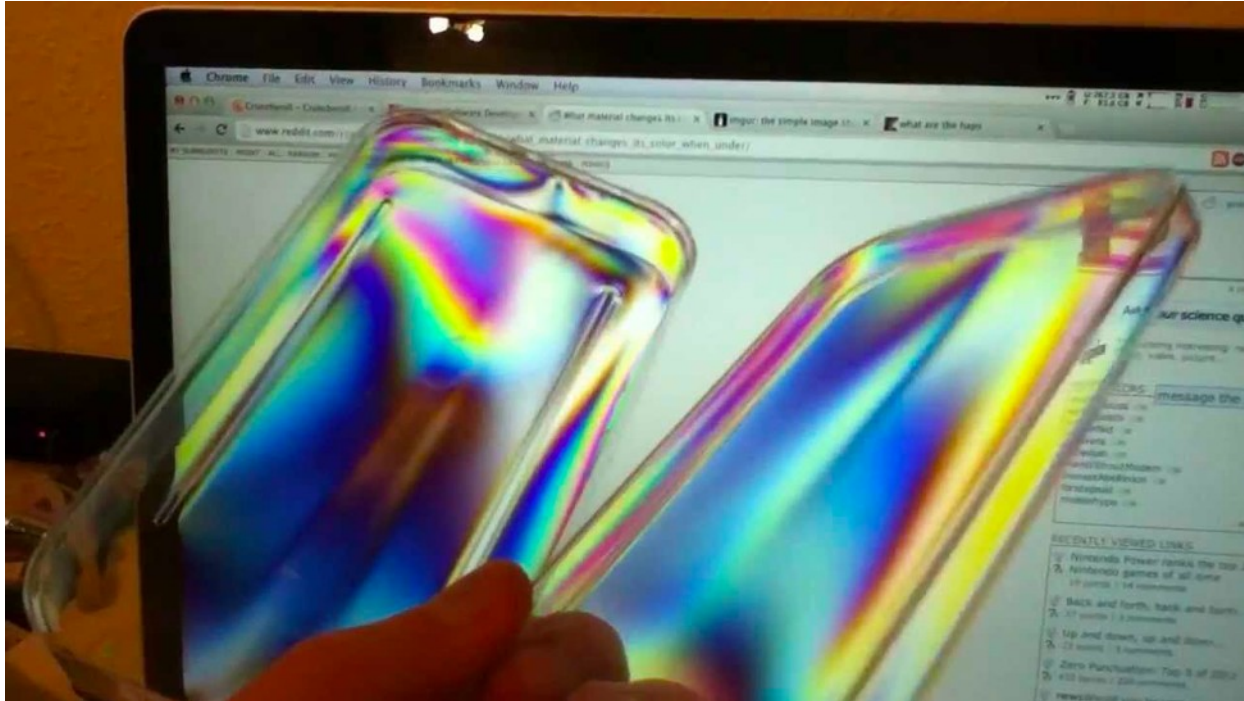
## PART 1: Course Overview Rendering

- **Overview of real-world polarization effects**
- Polarization in nature
- RayTracer Normal vs. Polarized
- Overview on Jones Calculus
- Overview on Mueller Calculus
- Build up of the RayTracer

# Polarized Sunglasses



# Birefringence Effects



©Peter Wasilewski



# Sky is differently polarized depending on time of day



# Polarization in Art



©Peter Waselewski

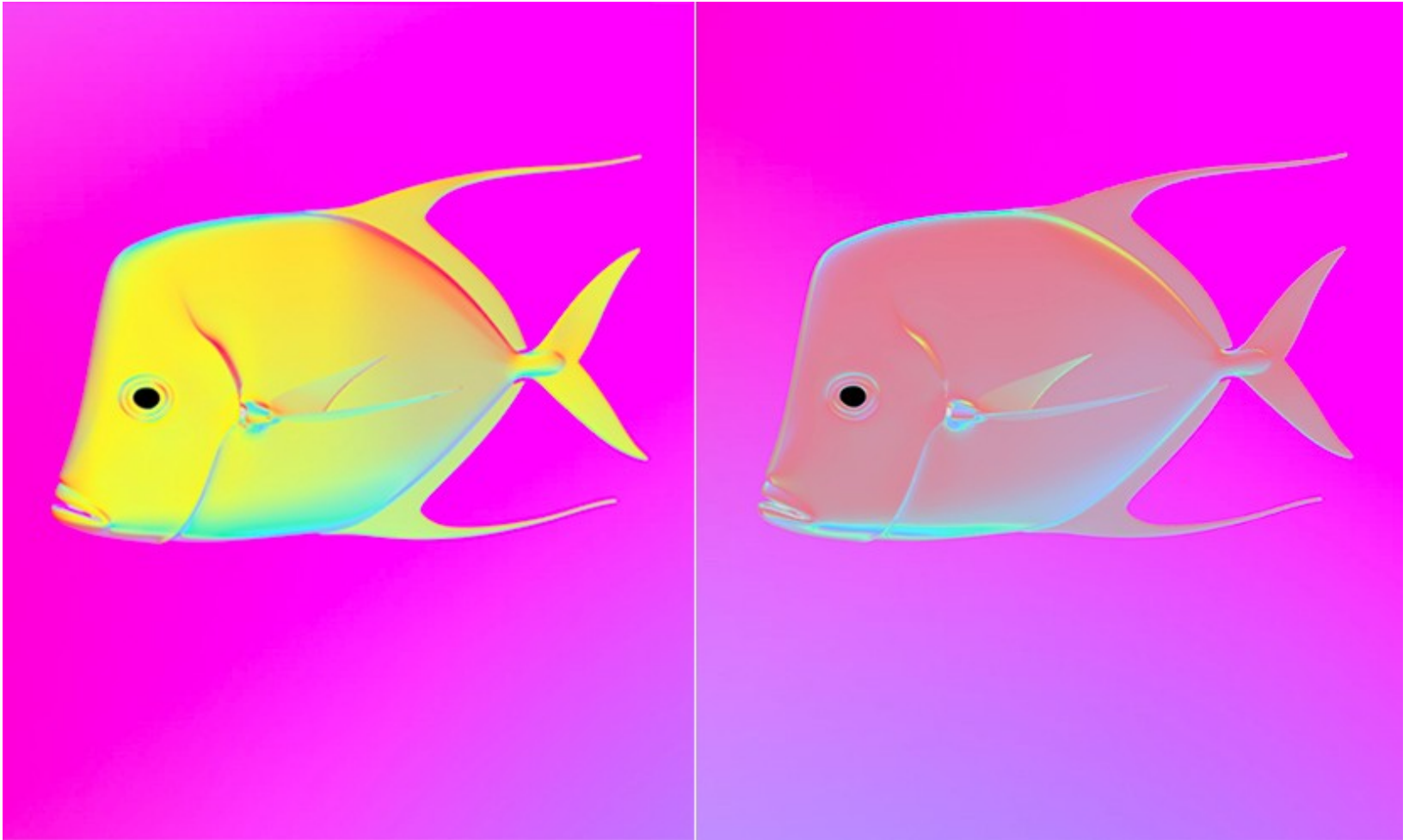
# Course Overview

- Overview of real-world polarization effects
- **Polarization in nature**
- RayTracer Normal vs. Polarized
- Overview on Jones Calculus
- Overview on Mueller Calculus
- Build up of the RayTracer

# Fish skin polarization

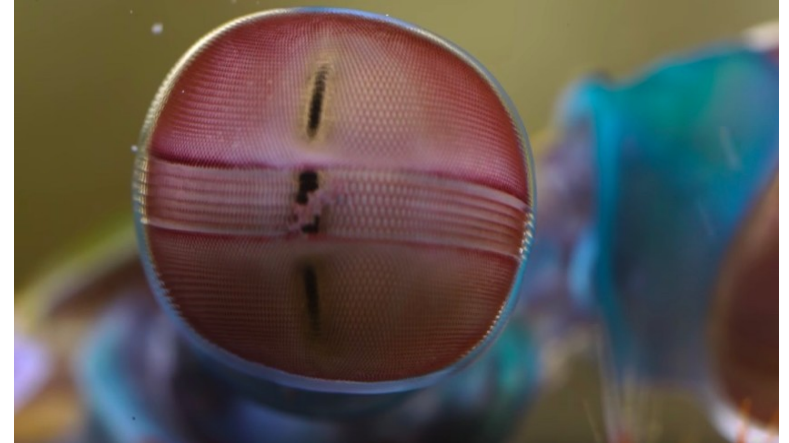


# Simulated Polarization





# Mantis Shrimp



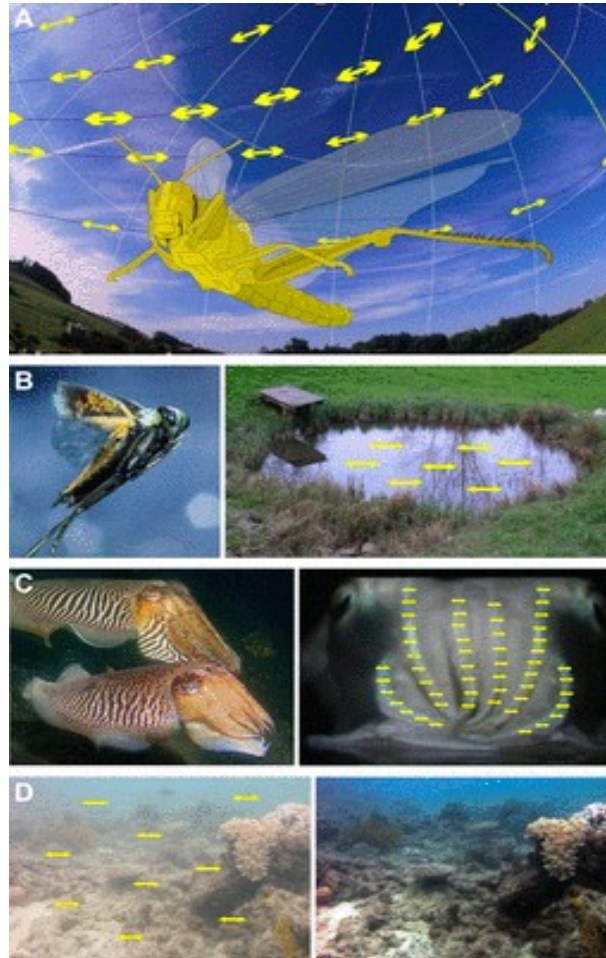
# Seeing and being seen – the butterfly



# Polarized Skin for communication – the Cuttlefish



# Polarization for Navigation



**Can invertebrates see the e-vector of polarization as a separate modality of light?**

Thomas Labhart

Journal of Experimental Biology 2016 219: 3844-3856; doi: 10.1242/jeb.139899

# Course Overview

- Overview of real-world polarization effects
- Polarization in nature
- **Ray Tracer Normal vs. Polarized**
- Overview on Calculus and Formulae
- Build up of the RayTracer



# Ray Tracer Normal vs. Polarized



Your Turn!

- Live demo for standard Raytracer
- Live demo for polarization aware Raytracer

# Course Overview

- Overview of real-world polarization effects
- Polarization in nature
- RayTracer Normal vs. Polarized
- **Overview on Calculus and Formulae**
- Build up of the RayTracer
- Use Case Scenario: Polarized Light for improving Stereo Vision

# Stokes Vector Representation of incident polarized light

$$S_0 = I_0 + I_{90}$$

$$S_1 = I_0 - I_{90}$$

$$S_2 = I_{45} - I_{135}$$

$$S_3 = \text{Chirality, not measured}$$

$I_0, I_{45}, I_{90}, I_{135}$

From a polarization aware camera such as PolarCam/ Bumblebee with Polarizer

# Stokes Vector Representation of incident polarized light

(Spherical coordinate representation)

$$S_0 = I$$

$$S_1 = I p \cos 2\phi \cos 2x$$

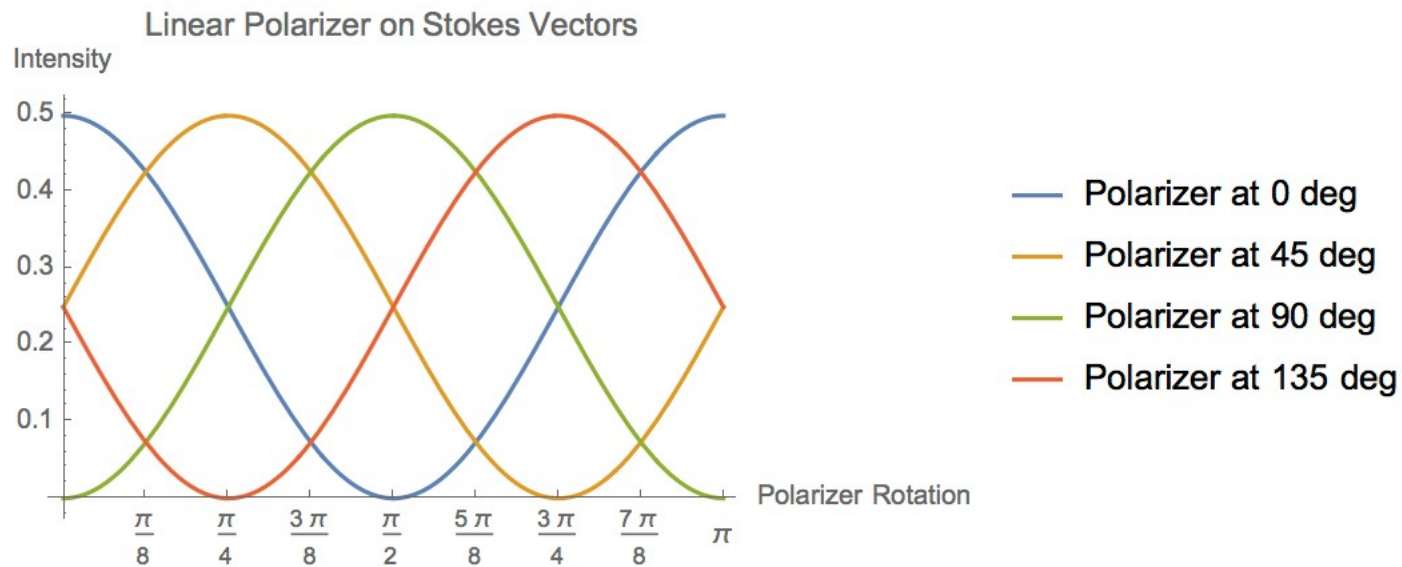
$$S_2 = I p \sin 2\phi \cos 2x$$

$$S_3 = I p \sin 2x$$

I the Intensity, p the DOLP,  $\phi$  the AOP, x the ellipticity - not measured

Source: The Poincare Sphere [https://spie.org/publications/fg05\\_p10-11\\_poincare\\_sphere](https://spie.org/publications/fg05_p10-11_poincare_sphere)

# Linear Polarizer vs. Intensity





# Mueller Matrix representation of light-transport through glass

- Light transport for transmissive and reflective glass

$$M_{\text{reflect}} \cdot S_{\text{reflect}} + M_{\text{transmit out}} \cdot M_{\text{transmit in}} \cdot S_{\text{transmit}}$$

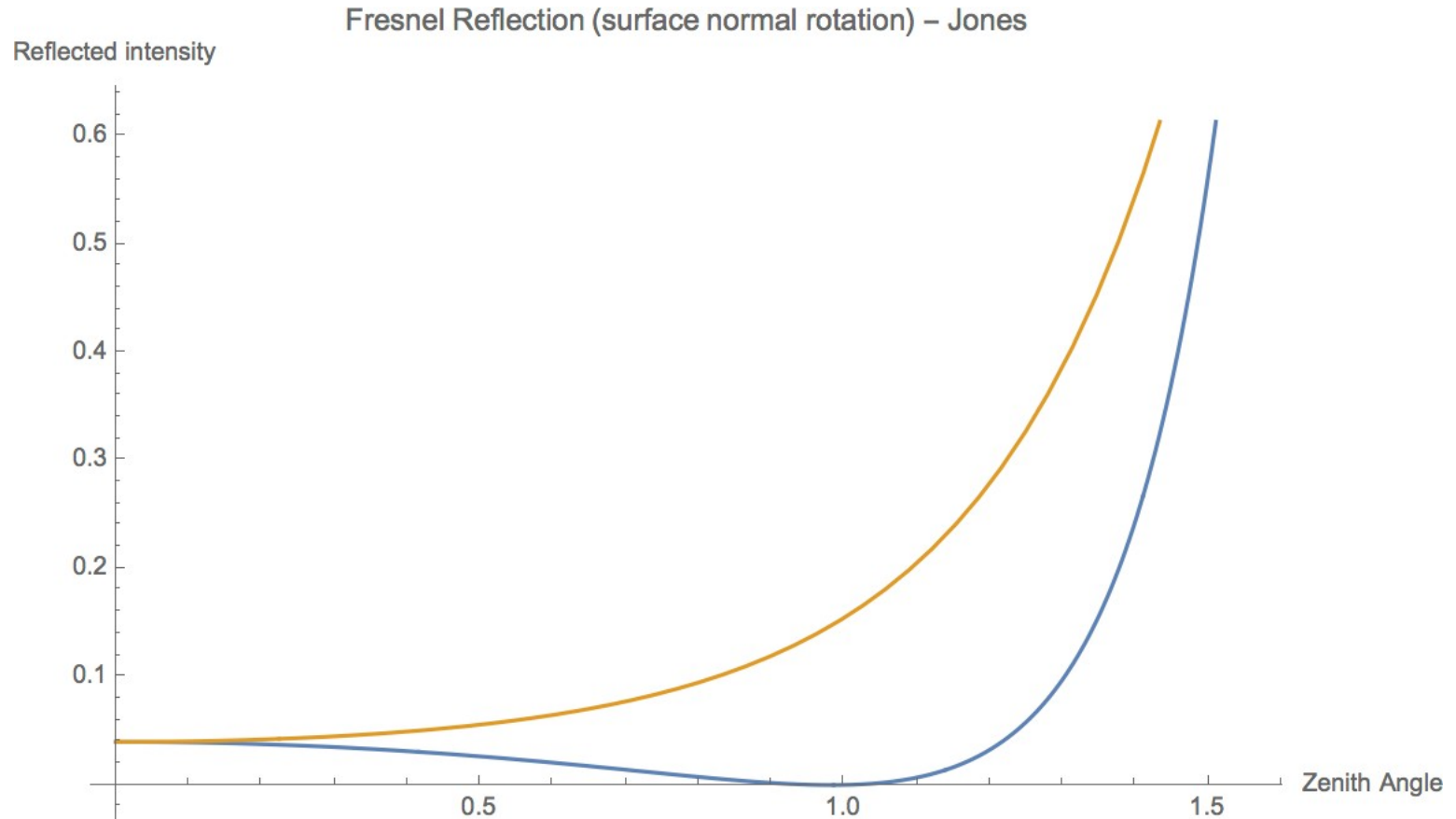
- Unknown incident light  $S_{\text{reflect}}$ ,  $S_{\text{transmit}}$ , potentially already polarized

$$M_{\text{reflect}} = \frac{1}{2} \begin{pmatrix} \cos^2(\theta_-) + \cos^2(\theta_+) & \cos^2(\theta_-) - \cos^2(\theta_+) & 0 & 0 \\ \cos^2(\theta_-) - \cos^2(\theta_+) & \cos^2(\theta_-) + \cos^2(\theta_+) & 0 & 0 \\ 0 & 0 & -2 \cos(\theta_-) \cos(\theta_+) & 0 \\ 0 & 0 & 0 & -2 \cos(\theta_-) \cos(\theta_+) \end{pmatrix} \left( \frac{\tan(\theta_-)}{\sin(\theta_+)} \right)^2$$

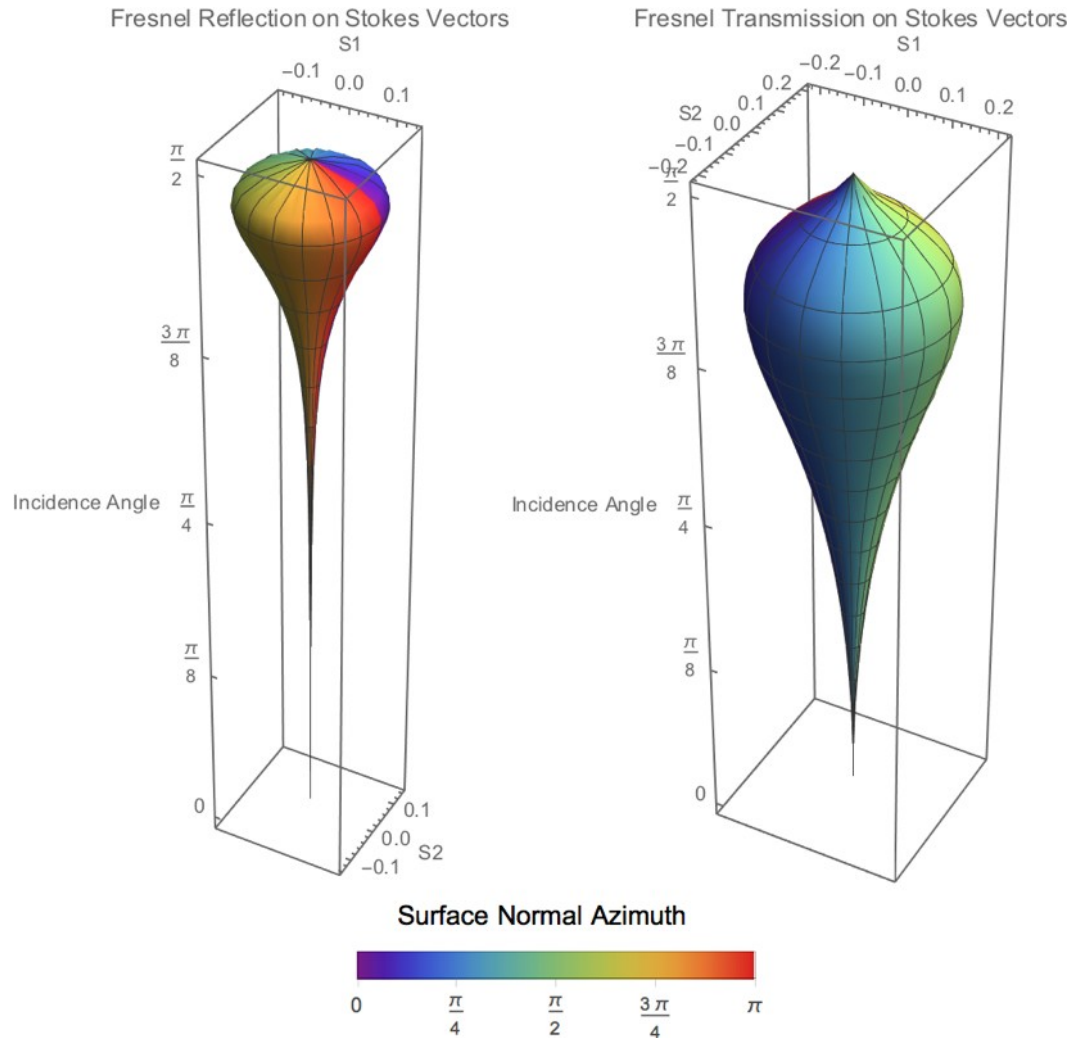
$$\theta_+ = \theta + \sin^{-1} \left( \frac{n_1 \sin(\theta)}{n_2} \right)$$

$$\theta_- = \theta - \sin^{-1} \left( \frac{n_1 \sin(\theta)}{n_2} \right)$$

# Fresnel Reflection Terms



# Reflection and Transmission Intensity



# Jones representation of light-transport for reflective surfaces

$$\mathbf{E}_{0^\circ} = (1, 0), \mathbf{E}_{90^\circ} = (0, 1) \text{ and } \mathbf{E}_{45^\circ} = (1, 1) \cdot \frac{1}{\sqrt{2}}.$$

$$M = \begin{pmatrix} rtm & 0 \\ 0 & rte \end{pmatrix}$$

$$M_\theta = \begin{pmatrix} \cos\theta & -\sin\theta \\ \sin\theta & \cos\theta \end{pmatrix} \cdot \begin{pmatrix} 0 & 0 \\ 0 & 1 \end{pmatrix} \cdot \begin{pmatrix} \cos(-\theta) & -\sin(-\theta) \\ \sin(-\theta) & \cos(-\theta) \end{pmatrix}$$

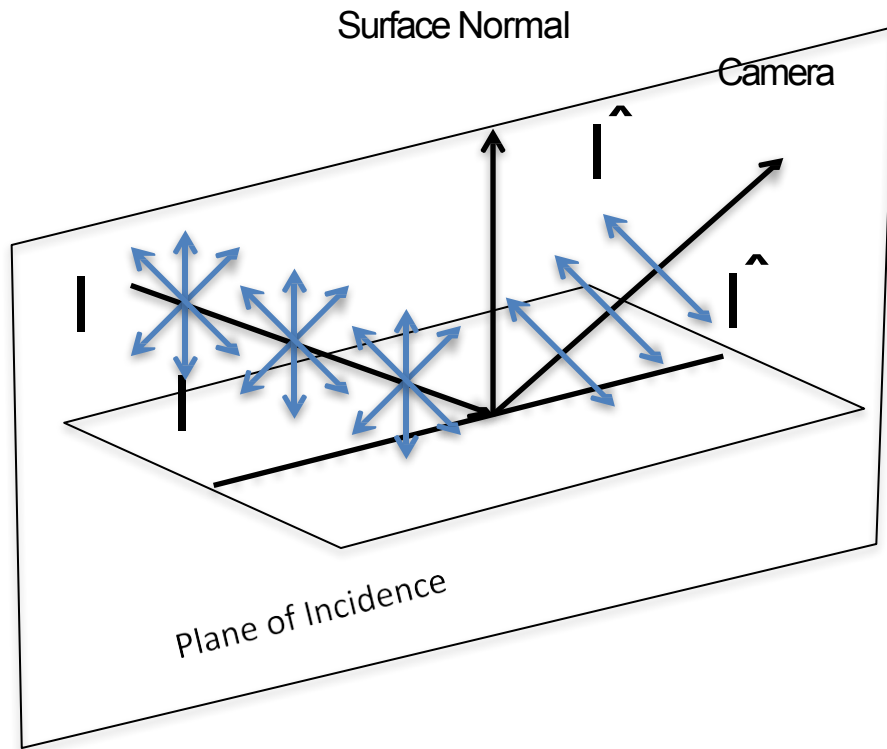
$$Na = 1/\sqrt{\cos(\theta)^2/n^2 + \sin(\theta)^2/n_1^2}$$

$$xe = (n_1 * \sin(\theta))^2, \quad xm = (Na * \sin(\theta))^2$$

$$rte = \frac{n_1 * \cos(\theta) - \sqrt{n_2^2 - xe}}{n_1 * \cos(\theta) + \sqrt{n_2^2 - xe}}, \quad rtm = \frac{n_1 * n_1 * \sqrt{n_2^2 - xm} - n_2 * n_2 * \sqrt{n_1^2 - xm}}{(n_1 * n_1) * \sqrt{n_2^2 - xm} + n_2 * n_2 * \sqrt{n_1^2 - xm}}$$

# Light-Surface Interaction

Polarization behavior changes at specular surfaces

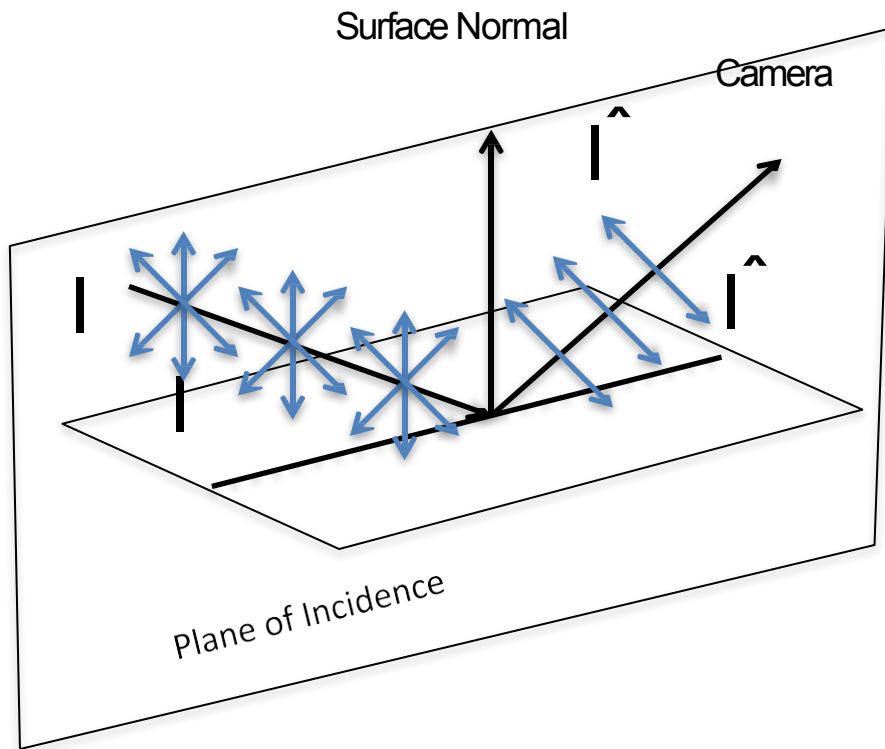


Left view, Polarizer at 0 degrees



# Light-Surface Interaction

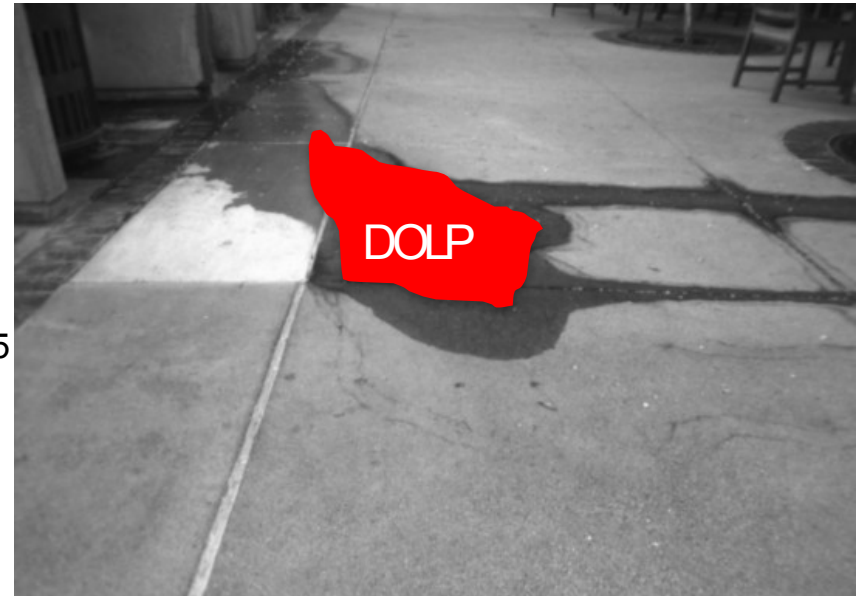
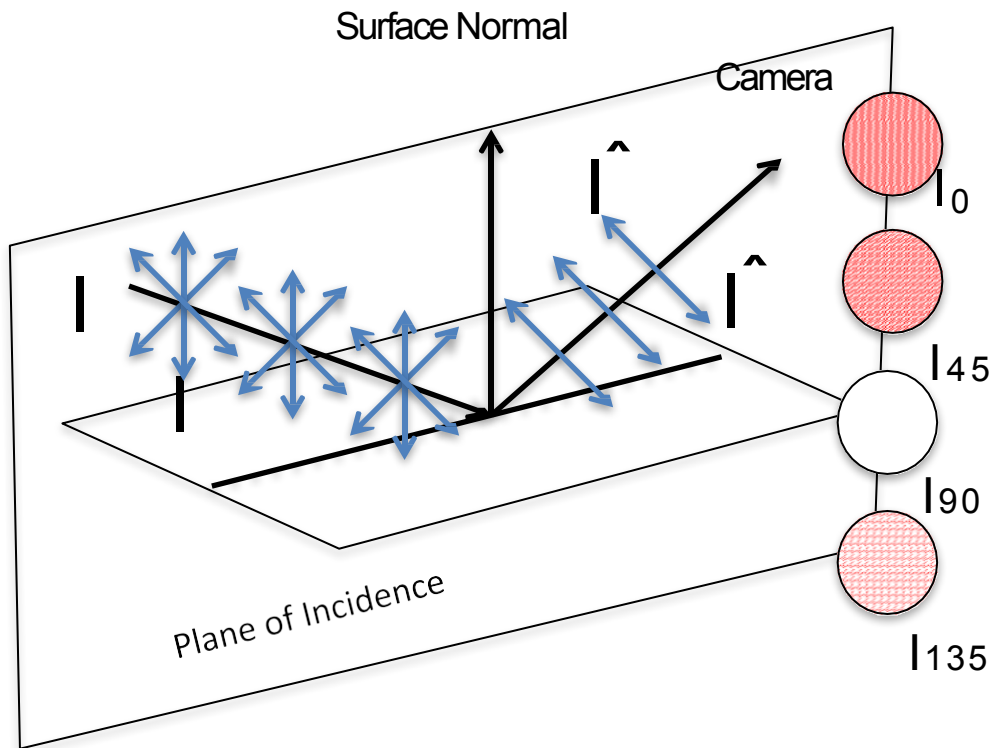
Polarization behavior changes at specular surfaces



Left view, Polarizer at 90 degrees

# Light-Surface Interaction

Polarization behavior changes at specular surfaces

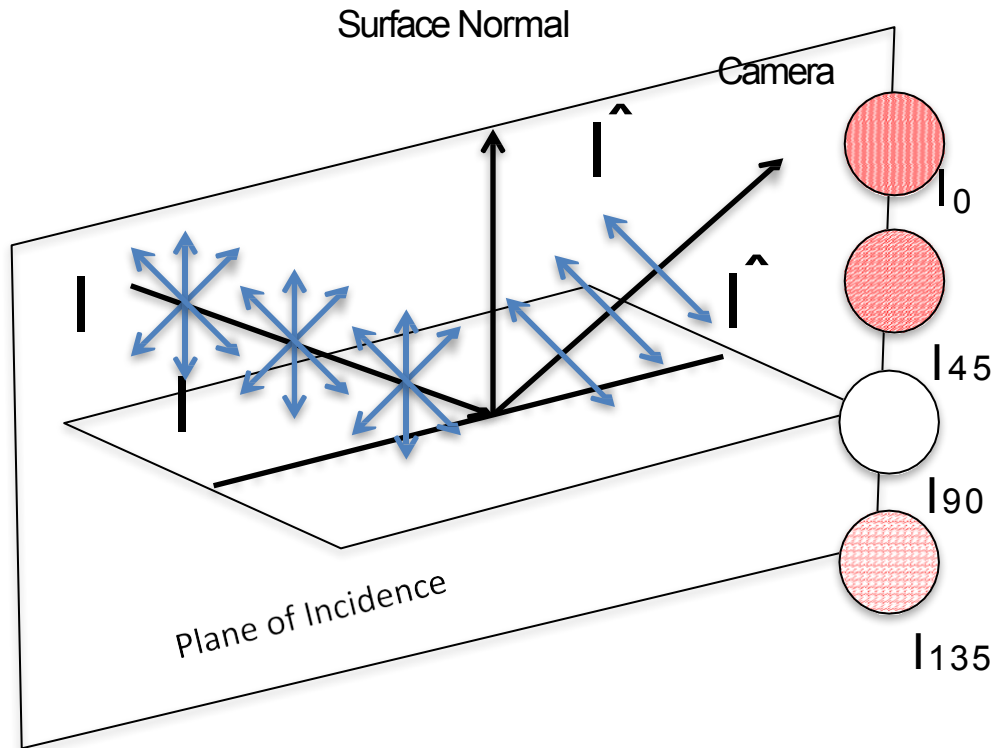


Left view, Polarizer at 0 degrees

$$\text{DOLP} = \frac{\rho \left( (I_{90} - I_0)^2 + (I_{135} - I_{45})^2 \right)}{I_{90} + I_0}$$

# Light-Surface Interaction

Polarization behavior changes at specular surfaces



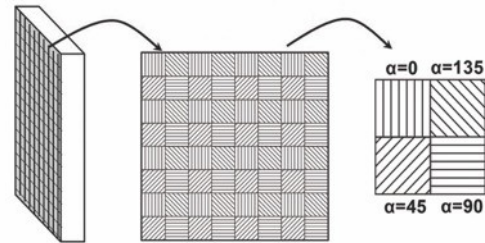
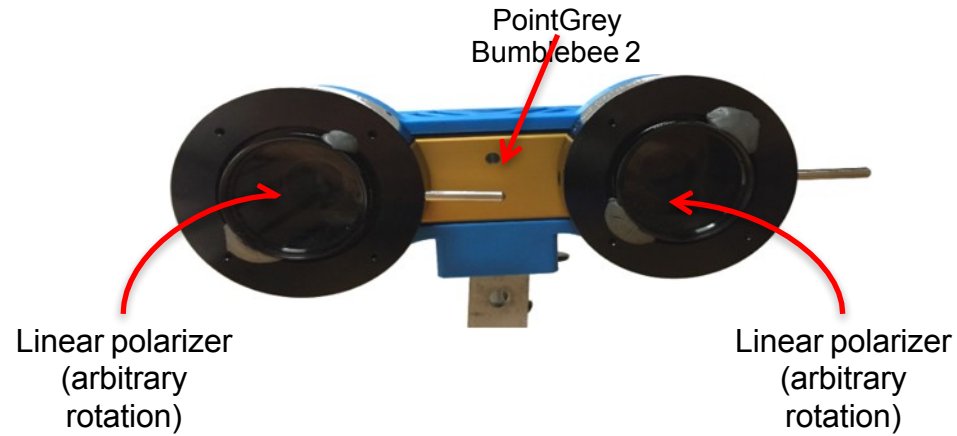
Left view, Polarizer at 0 degrees

$$AOP = \frac{1}{2} \text{ArcTan} \left( \sqrt{\frac{I_{135} - I_{45}}{I_{90} - I_{0}}} \right)$$

Surface normal is constrained to lie in plane of incidence

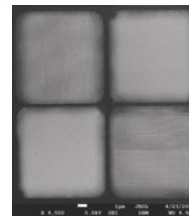
# How to capture polarization

Polarization behavior changes at specular surfaces



Polarizer array matched  
to detector pixels

Unit cell  
(super pixel)

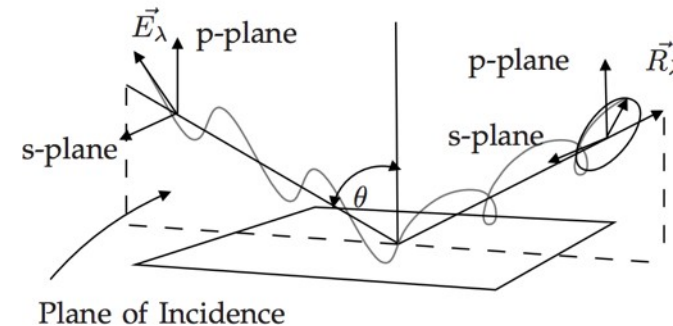
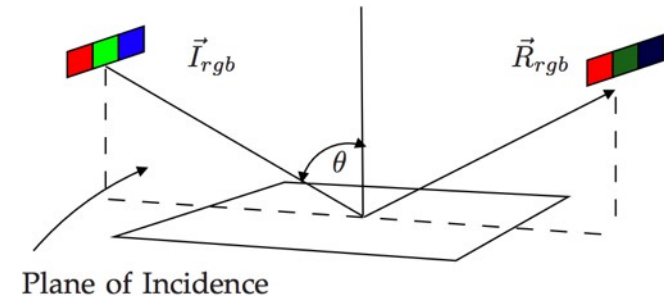


# Course Overview

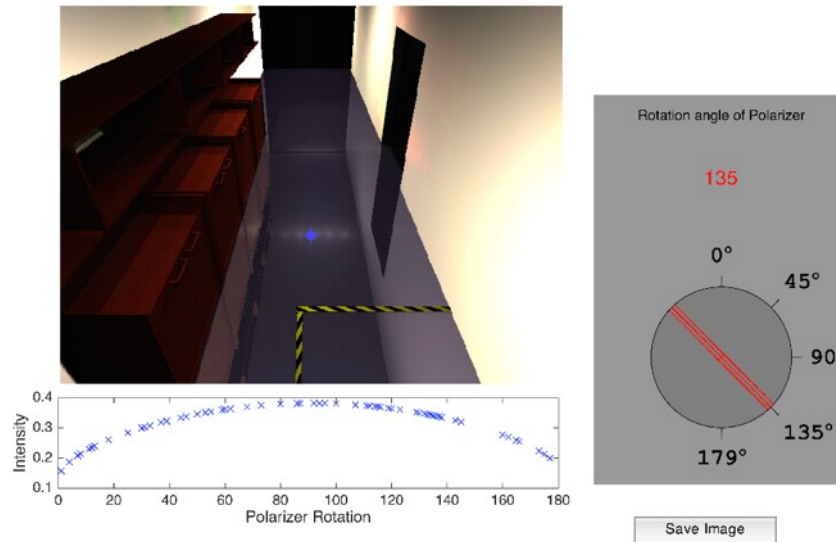
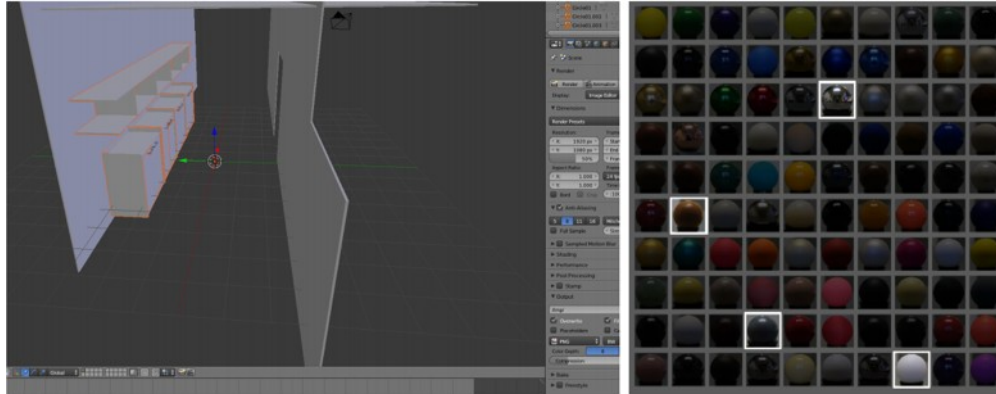
- Overview of real-world polarization effects
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- **Build up of the Ray Tracer**

# Take Whitted RT and alter the following

- RGBvector to Stokes/Jones
- Fresnel for Reflectance
- Fresnel for Transmittance
- Matrix interaction for Reflectance
- Matrix interaction for Transmittance
- Linear Polarizer for Viewing



# Simulate indoor polarized light transport in hallway scene





# Step 1) Change RGB to Stokes



Your Turn!

- Please open the first code sample and follow the instructions
- You will be able to generate an image with enhanced information (Stokes/Jones) after completion of the instructions

## Step 2) Add the Fresnel Term



Your Turn!

- Please open the second code sample and follow the instructions
- You will be able to generate the angular plot that shows slightly different behavior for perpendicular and parallel reflectance.

## Step 3) Add the Matrix interaction



Your Turn!

- Please open the third code sample and follow the instructions
- You will be able to model the reflectance interaction of a polarized light vector with Fresnel behavior now

## Step 4) Implement the linear polarizer



Your Turn!

- Please open the fourth code sample and follow the instructions
- You will be able to simulate a linear polarizer in front of your camera that you can rotate to a given angle.

# Break

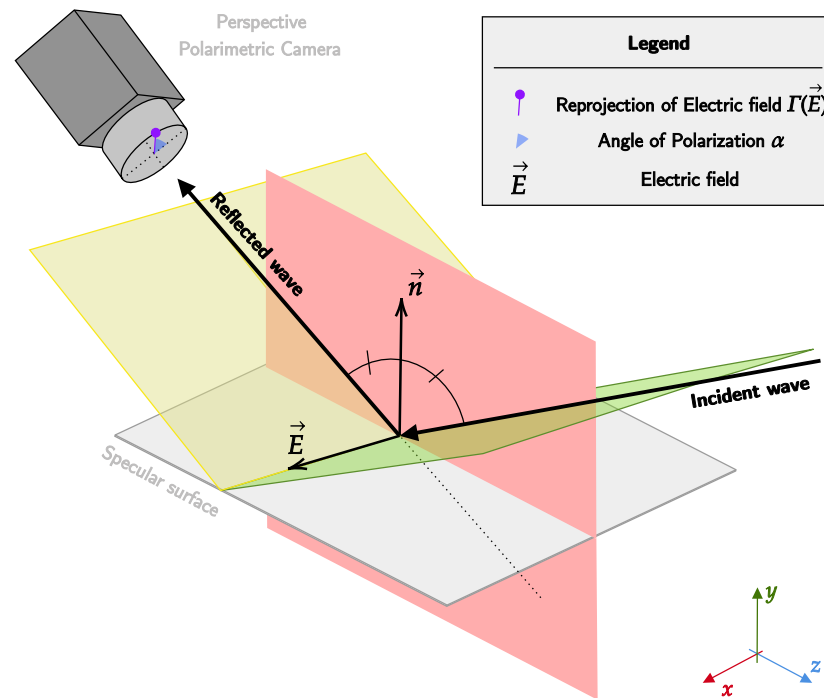
This concludes the first part  
of the course

# Part 2 - Computer Vision

- Polarization basics
  - An unconventional sensor
  - Dealing with super-pixels
  - From raw information to informative images
- Image representation
- Applications in Computer Vision
- Dealing with the data - Augmentation

# Polarization Basics

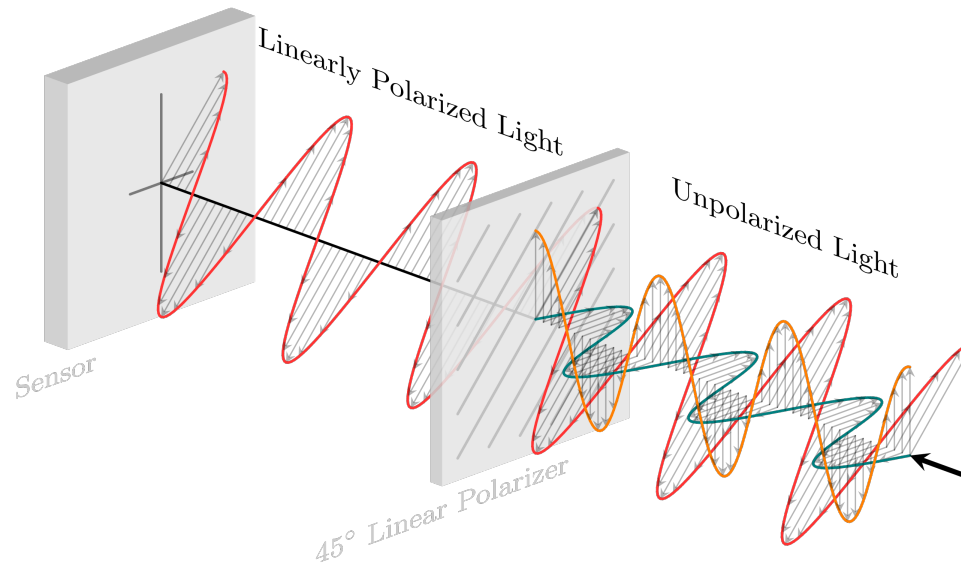
## Polarimetry in Computer Vision





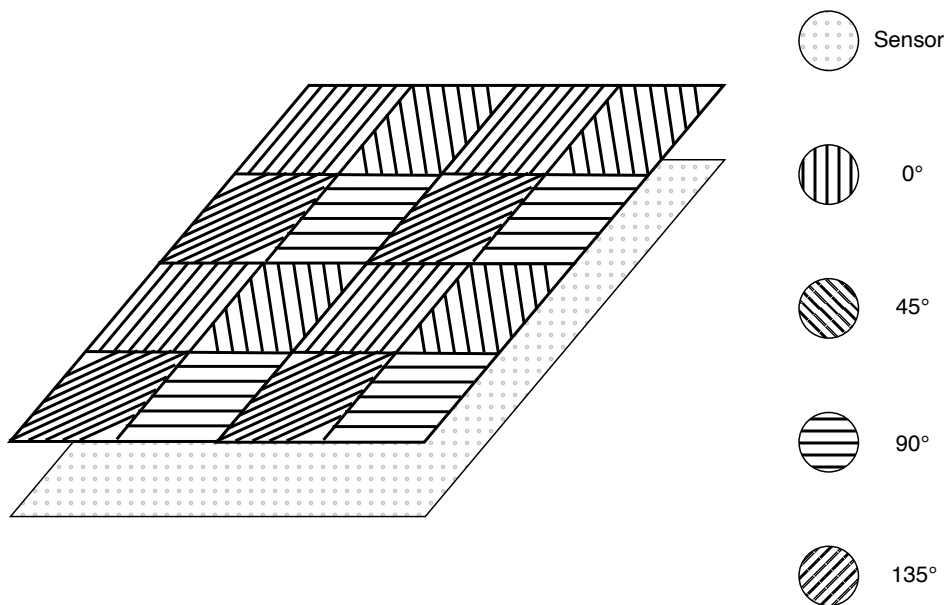
# An unconventional sensor

## Polarimetry in Computer Vision



# An unconventional sensor

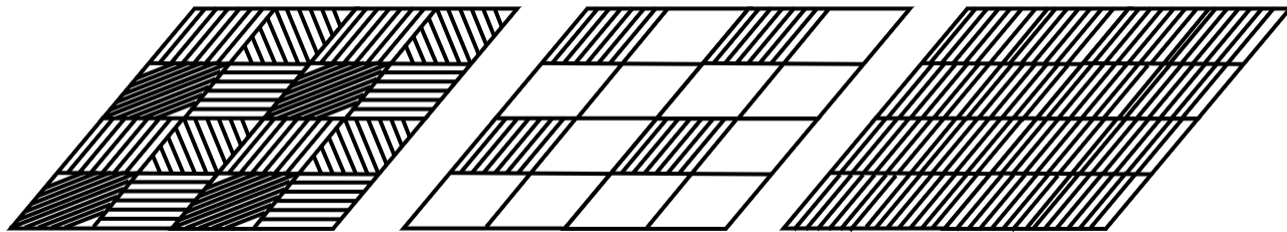
## Polarimetry in Computer Vision



# Dealing with super-pixels

## Polarimetry in Computer Vision

- The extracted raw images are sparse due to pixel organization.
  - We subsequently need to interpolate<sup>1,2</sup> images (bilinear, bicubic,...).



1 - Bradley M. Ratliff, Charles F. LaCasse, and J. Scott Tyo , Interpolation strategies for reducing IFOV artifacts in microgrid polarimeter imagery

2 - <https://github.com/BlanchonMarc/InterPol>

# Dealing with super-pixels



Your Turn!

## Polarimetry in Computer Vision

Please open the first code sample and follow the instructions

Implement the function allowing to open and interpolate polarimetry.

You should obtain 4 distinct images, each representing a unique polarization angle,

and observing the same size as the original raw image.

# From raw information to informative images

## Polarimetry in Computer Vision

Polarimetric information is conventionally indexed to Stokes parameters  $S$  depending on the intensities of polarizers oriented at different angles  $P_{\{0,45,90,135\}}$ .

$$S = \begin{pmatrix} s_0 \\ s_1 \\ s_2 \\ s_3 \end{pmatrix} = \begin{pmatrix} P_0 + P_{90} \\ P_0 - P_{90} \\ P_{45} - P_{135} \\ 0 \end{pmatrix}, \quad (1)$$

with  $s_3$  remaining null since the circular polarization is not acquired.

# From raw information to informative images

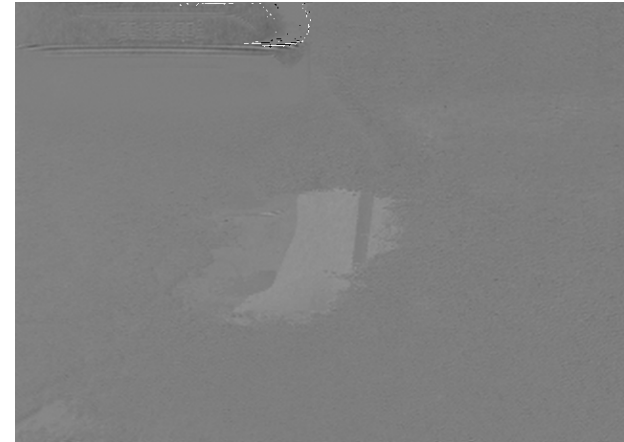
## Polarimetry in Computer Vision



S0



S1



S2

# From raw information to informative images

## Polarimetry in Computer Vision

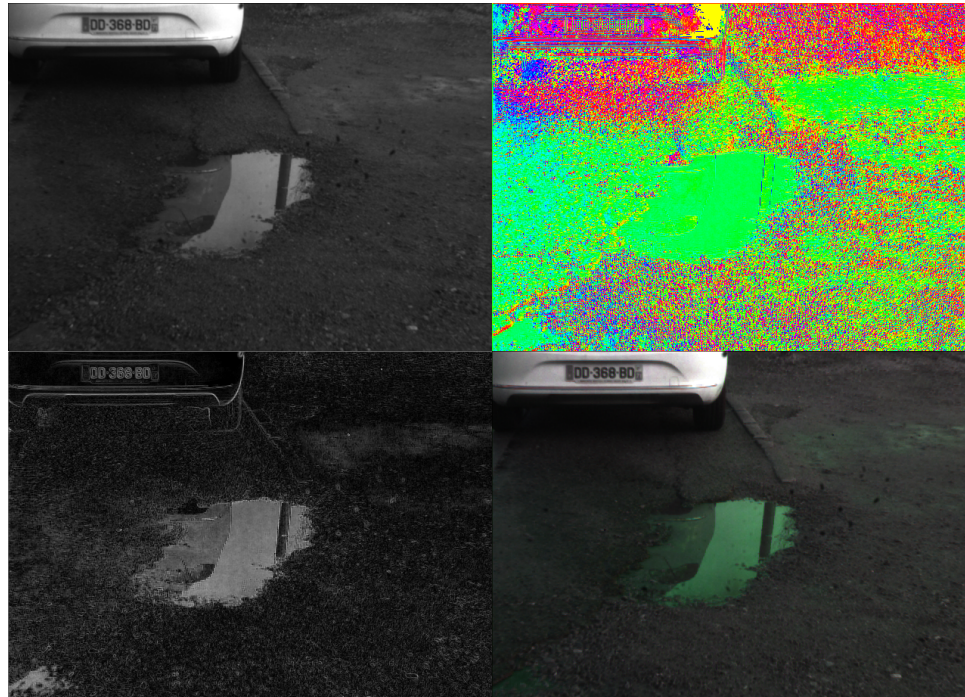
From Stokes parameters, one can derive three informative images characteristic of polarimetry: polarization angle  $\alpha$ , degree of polarization  $\rho$  and intensity  $\iota$ :

$$\begin{cases} \rho = \sqrt{\left(\frac{s_1}{s_0}\right)^2 + \left(\frac{s_2}{s_0}\right)^2} \\ \alpha = \frac{1}{2} \arctan2(s_1, s_2) \\ \iota = \frac{P_0 + P_{45} + P_{90} + P_{135}}{2} \end{cases} .$$



# From raw information to informative images

## Polarimetry in Computer Vision



# From raw information informative images



Your Turn!

## Polarimetry in Computer Vision

Please open the second code sample and follow the instructions

Implement the Stokes parameters function and the Polarization parameters function.

You should obtain two functions, one can return the three normalized Stokes parameters.

The other returns the intensity, the degree and the angle of polarization.

# Image representation

## Polarimetry in Computer Vision

- How can we use those images?
- What do they represent?
- Can we map this data to a specific colour space such that the image is representing the physics?

# Image representation

## Polarimetry in Computer Vision

- Raw concatenation



# Image representation

## Polarimetry in Computer Vision

- Hue - Luminance - Saturation (HSL) representation.



# Image representation

## Polarimetry in Computer Vision

Informative images are combined as proposed by Wolff and Andreou for display:

$$H \longrightarrow 2 * \text{AoP}, \quad S \longrightarrow \text{DoP}, \quad L \longrightarrow I/255.$$



# **Image representation**

## **Polarimetry in Computer Vision**

What are the drawbacks of those image representations?



# Image representation

## Polarimetry in Computer Vision



Your Turn!

Please open the third code sample and follow the instructions

Implement the concatenation and HSL image representation using the informative images.

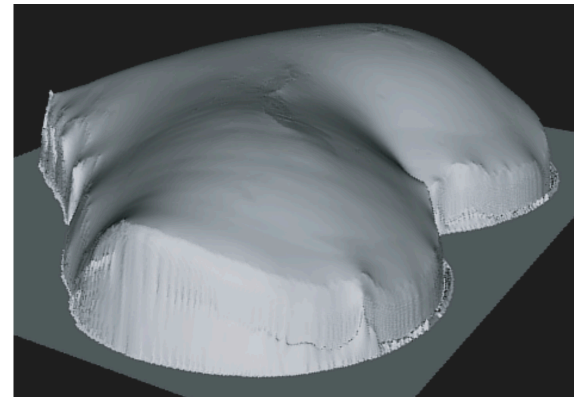
# Applications in Computer Vision

## Polarimetry in Computer Vision

- Shape from Polarization - Transparent objects



(a)



(b)

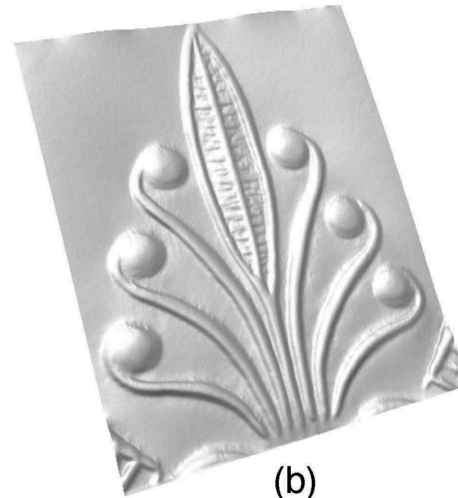
# Applications in Computer Vision

## Polarimetry in Computer Vision

- Shape from Polarization - Specular / Metallic objects



(a)



(b)

# Applications in Computer Vision

## Polarimetry in Computer Vision

- Pixel-wise semantic segmentation



# Applications in Computer Vision

## Polarimetry in Computer Vision

- Pixel-wise semantic segmentation with multimodal fusion

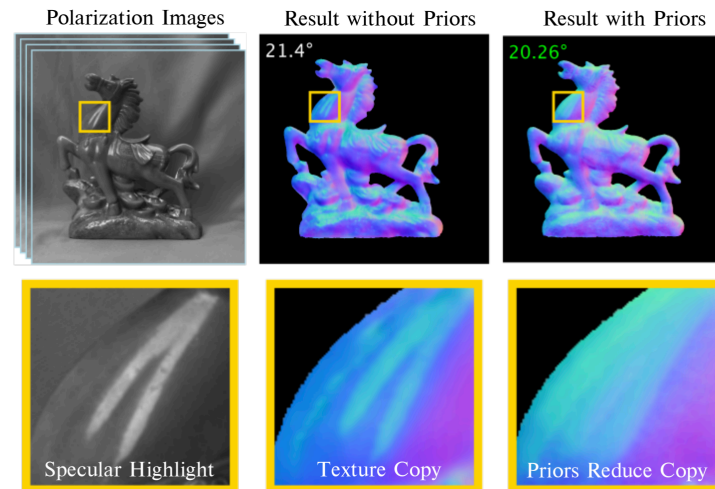




# Applications in Computer Vision

## Polarimetry in Computer Vision

- Deep Learning based Shape from Polarimetry



# Dealing with the data - Augmentation

## Polarimetry in Computer Vision

- Deep Learning and new learning based approach requires a sufficient amount of data.
- There is few dataset in physic based modality.
  - They are very often specialized and/or contain few images<sup>1</sup>.

1 - Polabot - <http://vibot.cnrs.fr/polabot.html>



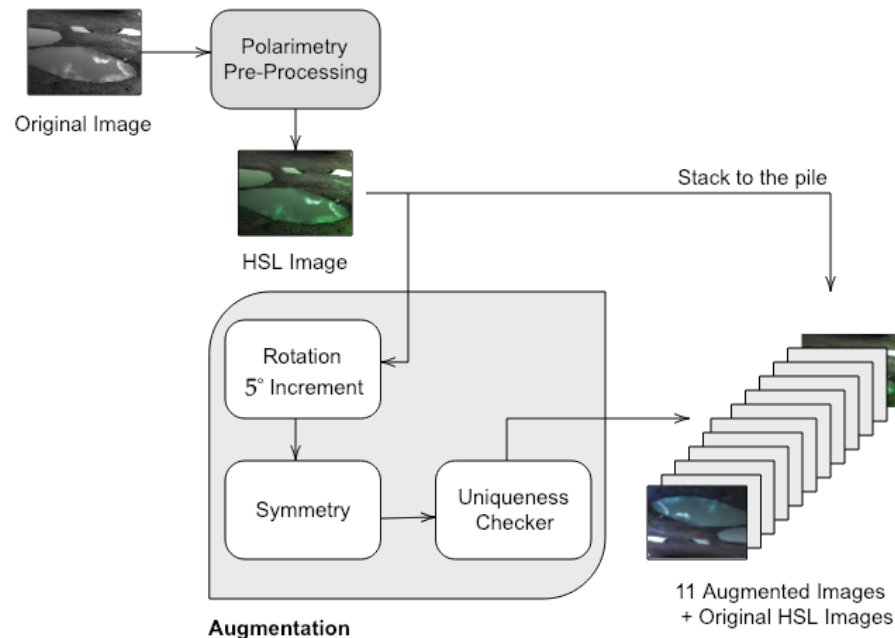
# Dealing with the data - Augmentation

## Polarimetry in Computer Vision

- The solution commonly adapted to deep learning and which is frequently used is **augmentation**.
  - It avoids overfitting
  - The principle is to create new images.
  - The augmentation **must** maintain the physical integrity of the modality.
    - Understanding the physics of the sensor.
    - Adapting the transformations.

# Dealing with the data - Augmentation

## Polarimetry in Computer Vision



Blanchon, Marc, Olivier Morel, Fabrice Meriaudeau, Ralph Seulin, and Désiré Sidibé. "Polarimetric image augmentation." arXiv preprint arXiv:2005.11044 (2020).

# Dealing with the data - Augmentation

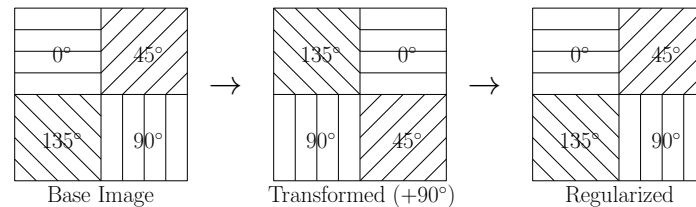
## Polarimetry in Computer Vision

- Two types of augmentation will be discussed.
  - Rotation
  - Flipping

# Dealing with the data - Rotation

## Polarimetry in Computer Vision

- Let's remember the pixel organization in polarimetry. Let's adapt the rotation to keep the polarimetric information.



$$H_{\text{rotated}} = R_{\theta}(H_{\text{prev}} - 2 * \mathbf{1}\theta).$$

# Dealing with the data - Rotation

## Polarimetry in Computer Vision

Base Image



Rotated Image



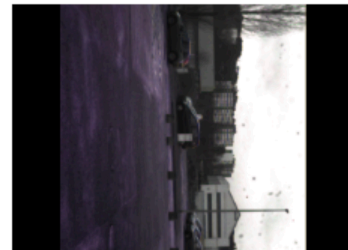
Regularized Image



(a) Acquisition  $0^\circ$

(b) Acquisition  $90^\circ$

(c) Augmented  $90^\circ$



# Dealing with the data - Flipping

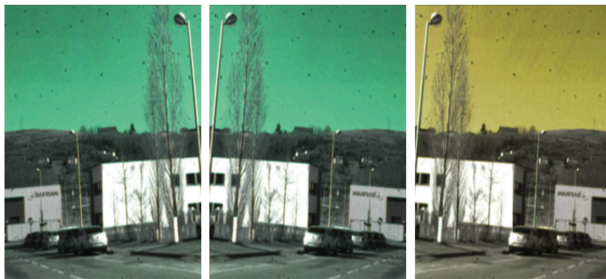
## Polarimetry in Computer Vision

- And let's apply the same approach to symmetry. This time, let's look at the impact of a transformation on the AoP.

Base Image

Flipped Image

Regularized Image

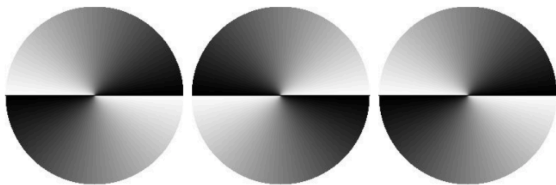


$$H_{\text{flipped}} = -H_{\text{prev}}.$$

Base Image

Flipped Image

Regularized Image



# Dealing with the data - Regularizing to HSL space

## Polarimetry in Computer Vision

- Use the color space to your advantage.

$$H_{final} = H_{transformed} \pmod{360}$$

- Verifying the physical integrity.

$$I_{\theta} = \frac{l}{2}(1 + \rho \cos(2\phi - 2\theta)),$$

# Dealing with the data - Regularizing to HSL space

## Polarimetry in Computer Vision



Your Turn!

Please open the fourth code sample and follow the instructions  
Implement the augmentation procedure in the designed spot.  
Do not deal with uniqueness and/or combination.



**Fin.**

Thank you for your attention.