

# **Underwater image processing, how to improve visibility?**

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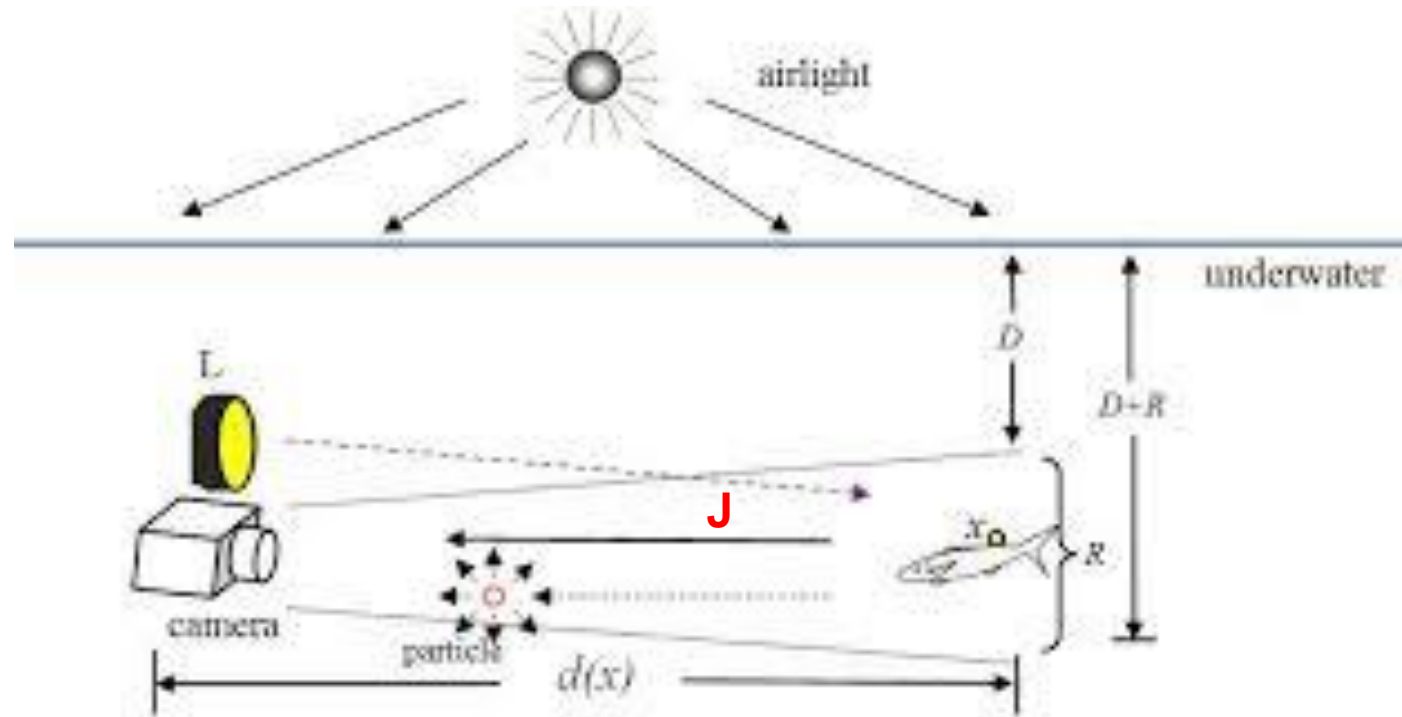
# Exploring the underwater world (issues)

Oceans cover about 70% of the earth's surface. It contains animal, mineral and raw material resources. Exploring its resources is a key topic around the world.

- Latvia has a coastline stretching over 531 km
- 12,500 rivers, which stretch for 38,000 km. Major rivers include the Daugava River, Gauja, and Salaca, the largest spawning ground for salmon in the eastern Baltics.
- Fishing potential estimated by FAO (Food and Agriculture Organization) at around 82,300 tonnes/year.
- The sector employs 13 900 people, which is 1.2% of the total economically active population of Latvia.

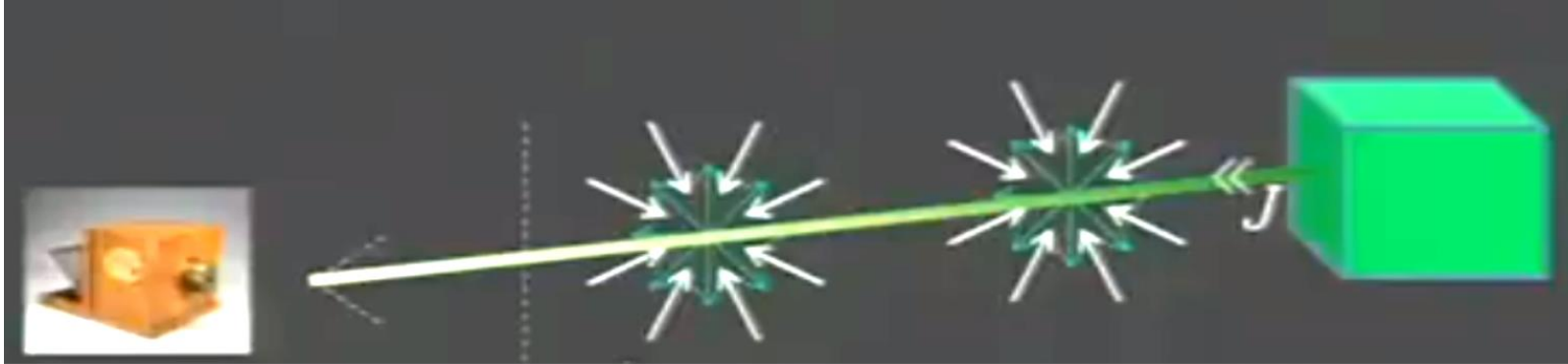


# Problem



- Light reflected by an object (*noted  $J$* ) undergo scattering along its way to the camera.

# Problem

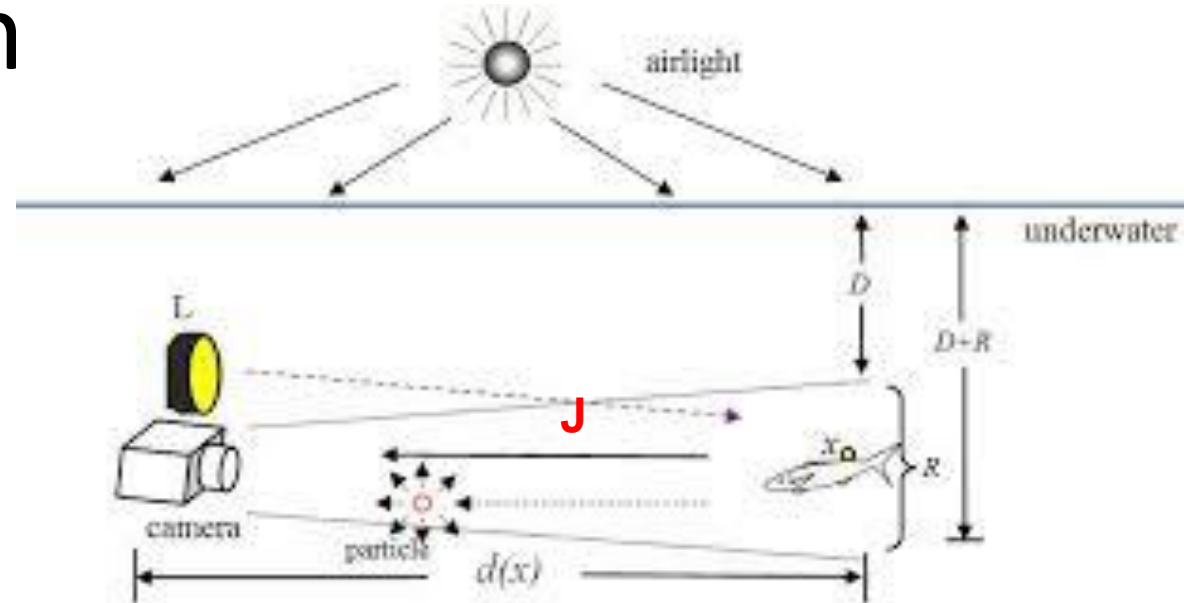


J is the original light  
B scattering effect

**Effects** : light produce a distinctive gray or bluish hue and affects visibility

**How to restore the original penetrated light ?**

# Problem



$$I(x) = J(x)t(x) + (1 - t(x))A \quad (1)$$

- $I(x)$  : raw image
- $t(x)J(x)$ : light reflected by the object
- $A$  : environment light (Air light)
- $t(x)$ : transmission

$$t(x) = e^{-\beta d(x)}$$

# Proposed model

## Mathematical Model:

$$\mathbf{I}(\mathbf{x}) = \mathbf{J}(\mathbf{x})t(\mathbf{x}) + \mathbf{A}(1 - t(\mathbf{x}))$$

$$t(\mathbf{x}) = e^{-\beta d(\mathbf{x})}$$

- $\mathbf{I}(\mathbf{x})$  is the observed radiance at  $\mathbf{x}$
- $\mathbf{J}(\mathbf{x})$  is the original scene radiance at  $\mathbf{x}$
- $\mathbf{A}$  is the *environmental light*
- $t(\mathbf{x})$ , scalar called *transmission*: describes how the radiance of a point in the scene is attenuated according to its distance  $d$  from the observer
- Note that  $\mathbf{I}$ ,  $\mathbf{J}$ ,  $\mathbf{A}$  are (R,G,B) triplets

# ***The Atmospheric Scattering Model***

## Mathematical Model:

$$\mathbf{I}(\mathbf{x}) = \mathbf{J}(\mathbf{x})t(\mathbf{x}) + \mathbf{A}(1 - t(\mathbf{x}))$$
$$t(\mathbf{x}) = e^{-\beta d(\mathbf{x})}$$

- In order to remove these effects, we must recover  $J(x)$
- $\mathbf{I}(\mathbf{x})$  is known
- Quantities  $\mathbf{A}$  and  $t$  are typically unknown

# State of the art

- Can be grouped into several categories
  1. *With multiple images*
  2. *With one image + depth-map*
  3. *Single image*
- We are mostly interested in Single-Image methods



# Summary

## Multiple image methods

- require special equipment (polarizers) or same scene under different weather conditions.
- They don't necessarily produce better results than single-images approaches

# Summary

## Single-Image methods

- do not require special equipment, nor extra information
- They either make assumption on the nature of the scene, or require little interaction by the user

# Multiple Image Approaches

## Narasimhan & Nayar's method

- Assumes 2+ bad weather images are given
- Uses geometric constraints to estimate  $\mathbf{A}$
- The airlight component  $[1-t(\mathbf{x})]$  is estimated from corresponding pixels of the two bad weather images

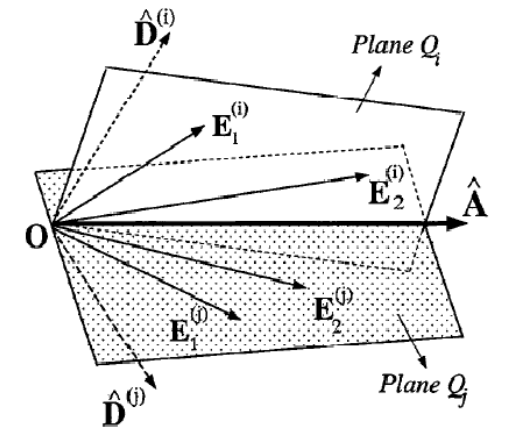


Figure 13. Intersection of two different dichromatic planes yields



# One Image + Depth + Texture

## Kopf et al. Method: *Deep Photo* project from SIGGRAPH 2008

- Assumes a 3D model of the scene is given (e.g.: from Google Maps)
- Assumes textures of the scene are given (from satellite or aerial photos)
- Requires user interaction to align the 3D model with the scene
- Very accurate results



# Proposed approach

- **Limitation:**

Different equations with different unknowns.

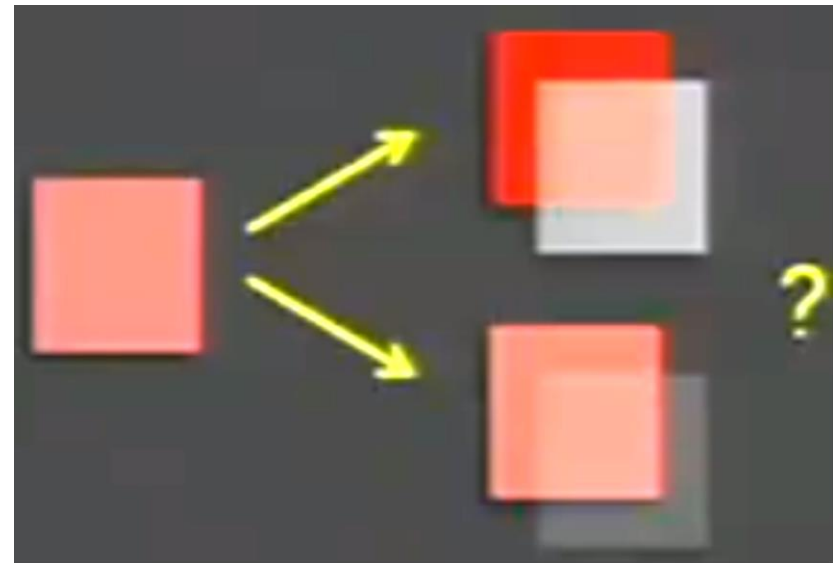
$$I_1 = t_1 J + (1 - t_1) A$$

$$I_2 = t_2 J + (1 - t_2) A$$

In this case we have 5 unknowns ( $t_1, t_2, J_R, J_G, J_B$ ) and 6 equations

# Proposed approach

- **Airlight-albedo ambiguity**
- Deeply structured red color
- Masked with tiny layer of haze



# Proposed approach

- In this model we suppose that the original image is composed by two components: shading  $l$  (*scalar*) and reflectance  $R$  (3D vector).

$$J(x) = l(x)R(x) \quad (3)$$

Redefined model:

$$I(x) = t(x)l(x)R(x) + (1 - t(x))A \quad (4)$$

# Proposed approach

$$I(x) = t(x)l(x)R(x) + (1 - t(x))A \quad (4)$$

- We assume that the reflectance  $R$  is constant in each region.

$$R(x) = R(\text{const})$$

In order to measure the two unknowns (shading  $l$  and medium light  $A$ ) independently, we proceed by multiple derivation using **Piece-wise constant albedo**. The above calculation process is described by the following formula.

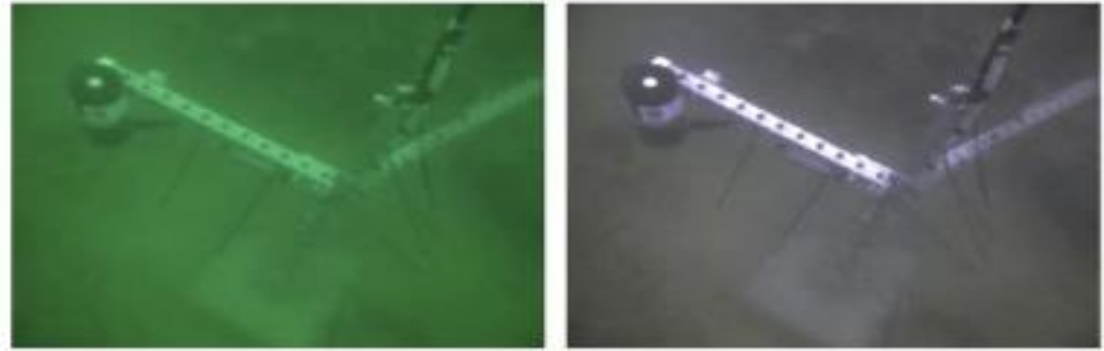
$$I_A(x) = \langle I(x), A \rangle = t(x)l'(x)\eta + (1 - t(x))A \quad (5)$$

$$I_{R'}(x) = \langle I(x), R' \rangle = t(x)l'(x) \quad (6)$$



# Simulation results

The proposed algorithm is tested in ordinary computer (Core i5, 2.3GHz, RAM: 4Go). The input images are in JPG format (600×400 images). This figure presents simulation results applied on underwater images.



(a)

(b)

# Simulation results



Raw image



By **Narasimhan et.al**



By **the proposed approach**

**THANK YOU**