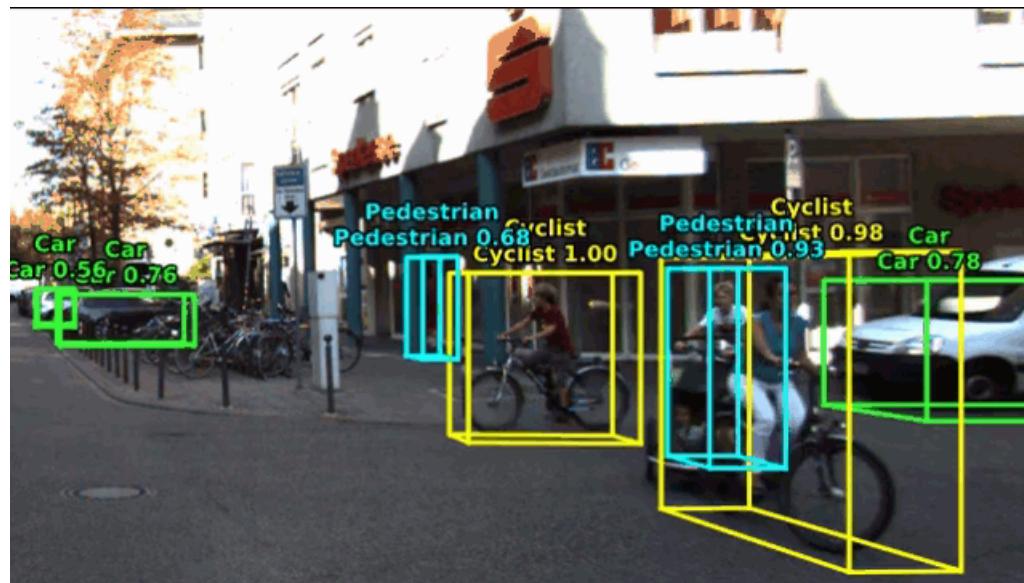


# The Camera Sensor

Course 3, Module 1, Lesson 1 – Part 1

# The Camera Sensor



# The Camera Sensor

- Captures detailed appearance information
- High rate of information capture
- Relatively inexpensive

*More information*

Sam: Can we add this video of a camera sensor here?

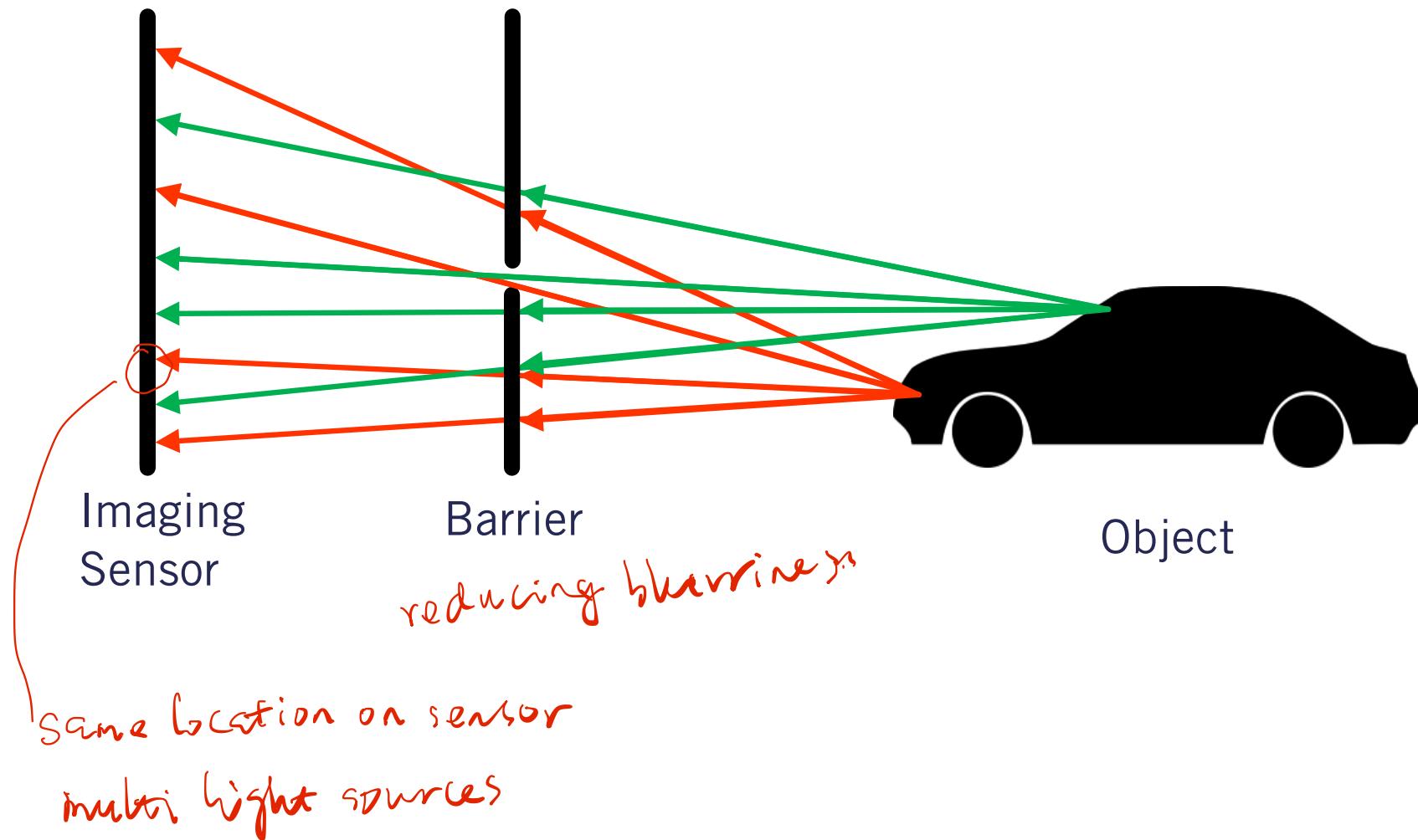
<https://www.shutterstock.com/video/clip-1022389735-camera-repair-sensor-digital-close-up-matrix>

*object detection ,  
segmentation ?  
identification*

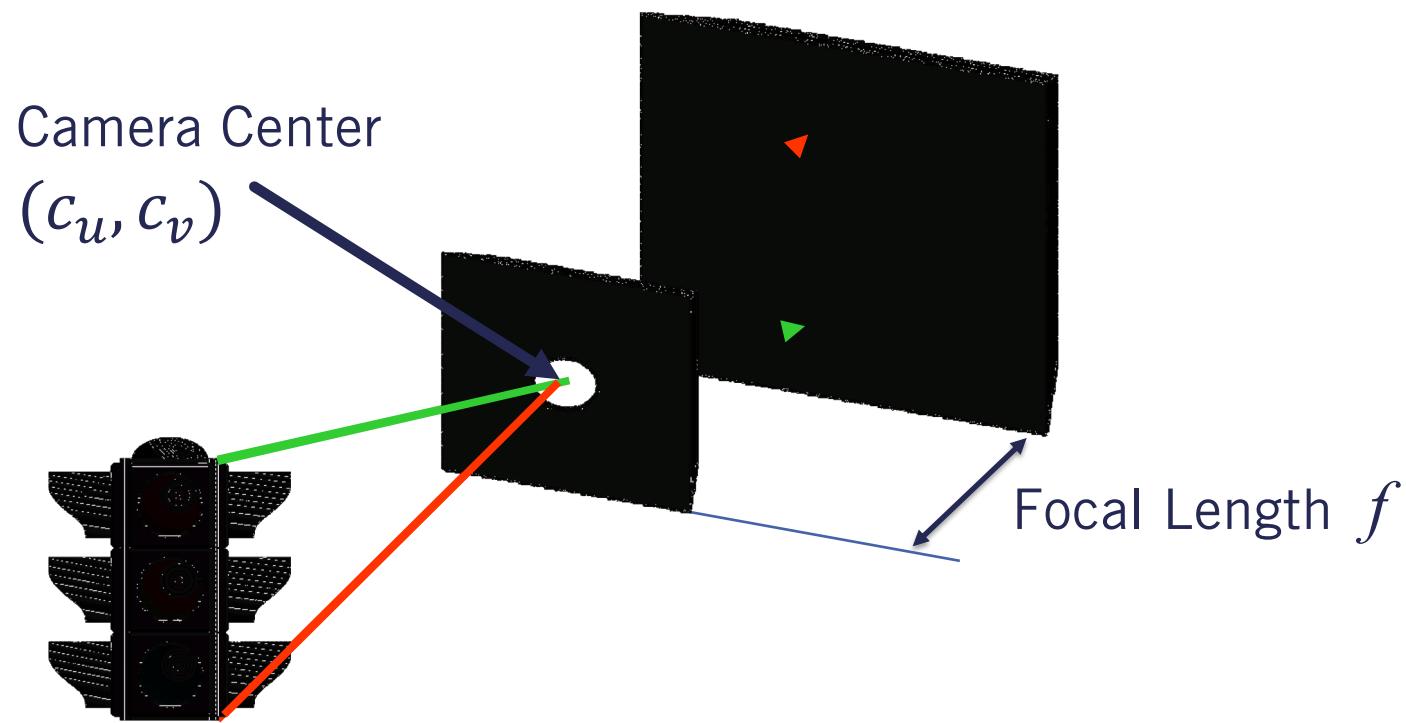
Sam: This image of driving with lots of signs and lights for click 1:

<https://www.shutterstock.com/image-photo/busy-traffic-during-rush-hour-amsterdam-1069610642?src=bLnVCTgF8bcUNe81sHDIVw-1-39>

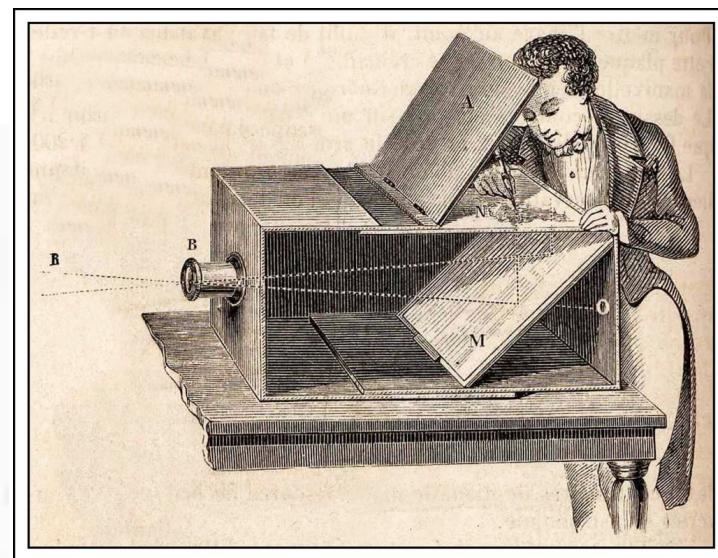
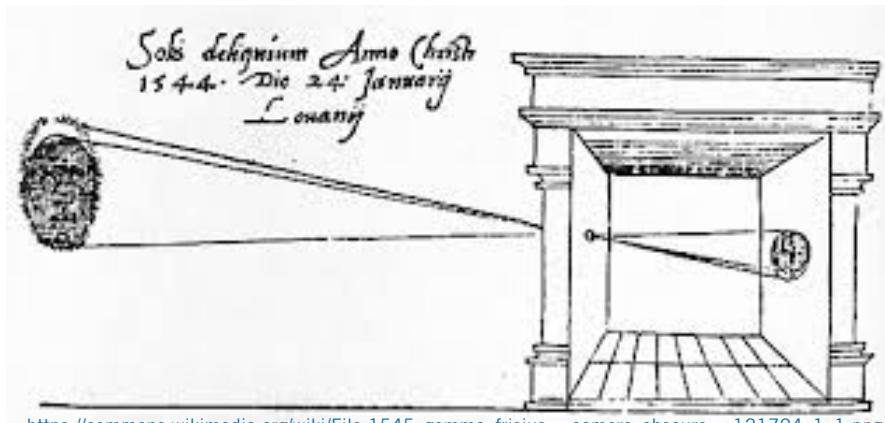
# Image Formation



# Pinhole Camera Model



# Camera Obscura: 1544 A.D.



# Modern Day Cameras



<https://www.pexels.com/photo/4k-alone-andaman-and-nicobar-black-774343/>

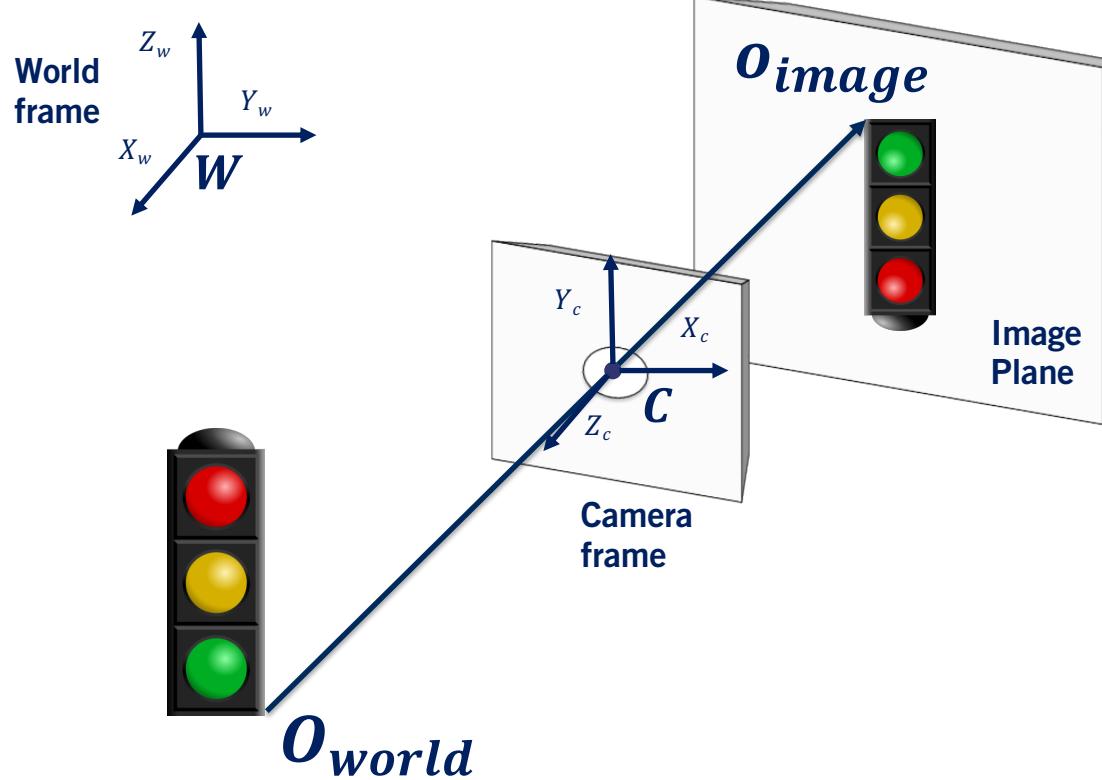
# Summary

- The camera is important as a sensor due to the amount of information it can capture, and its relatively small price tag
- The basic camera model has existed since the 1500s!
- **Next: Projective Geometry**

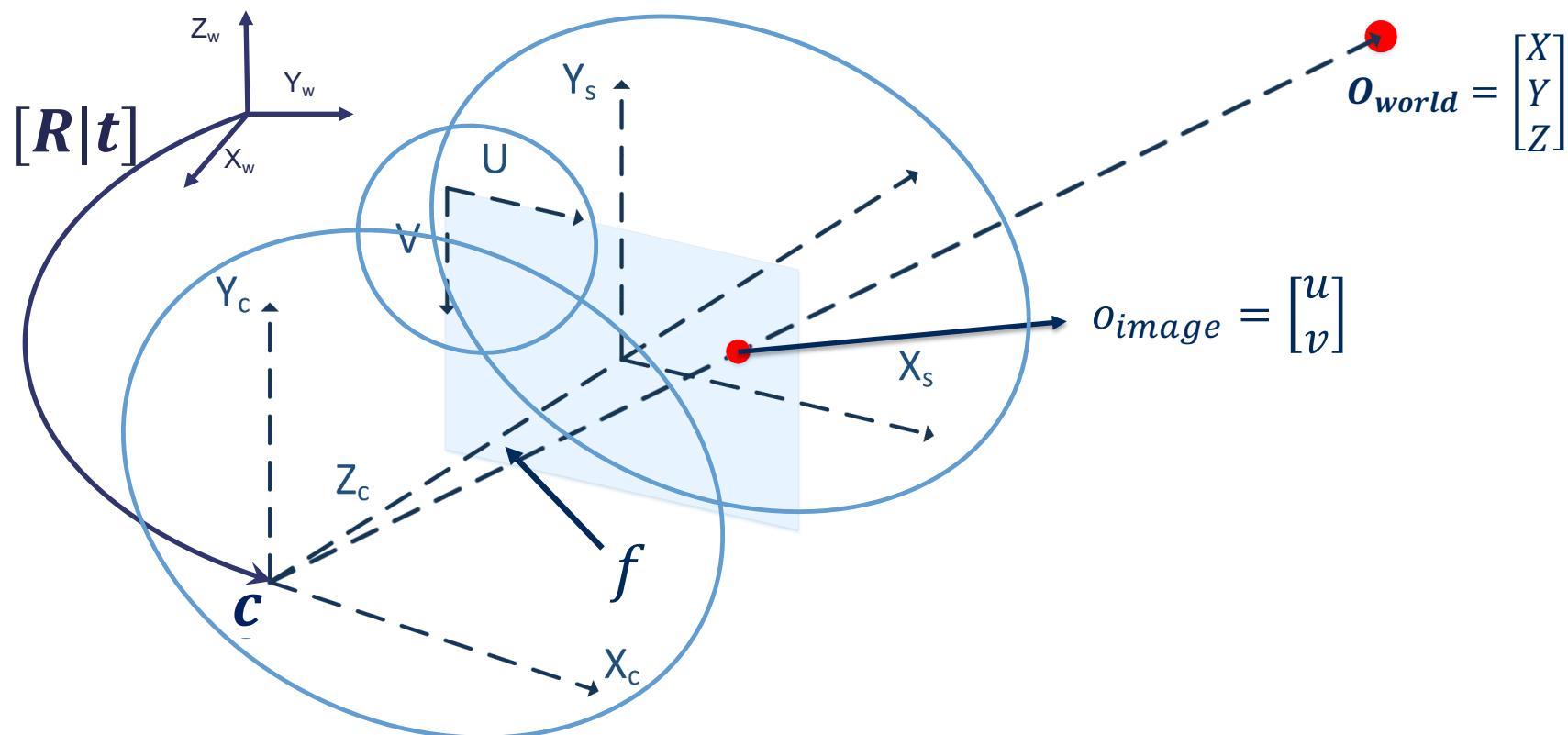
# Camera Projective Geometry

Course 3, Module 1, Lesson 1 – Part 2

## Projection: World → Image (Real Camera)

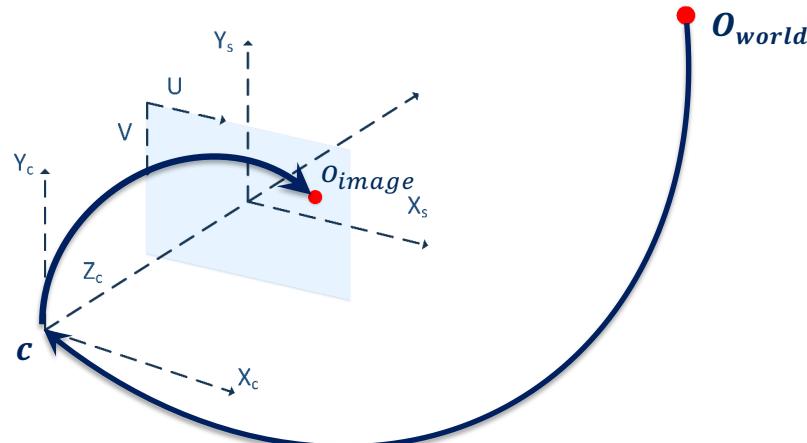


## Projection: World → Image (Simplified Camera)



# Computing the Projection

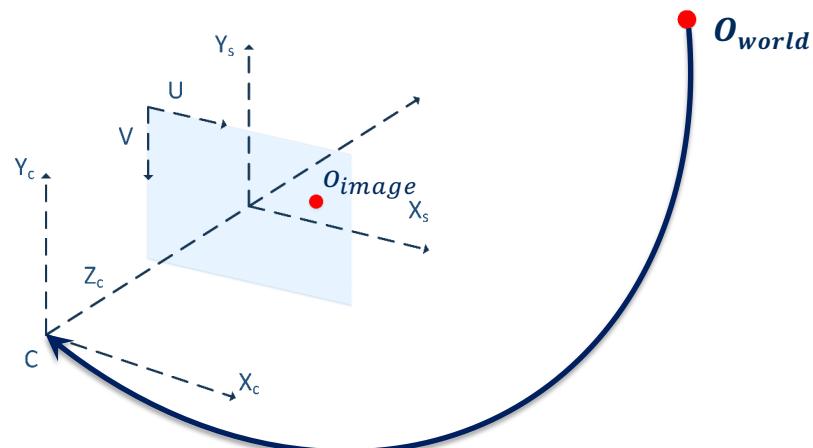
- Projection from World coordinates → Image coordinates:
  1. Project from World coordinates → Camera coordinates
  2. Project from Camera coordinates to Image coordinates



# Computing the Projection

- World → Camera:

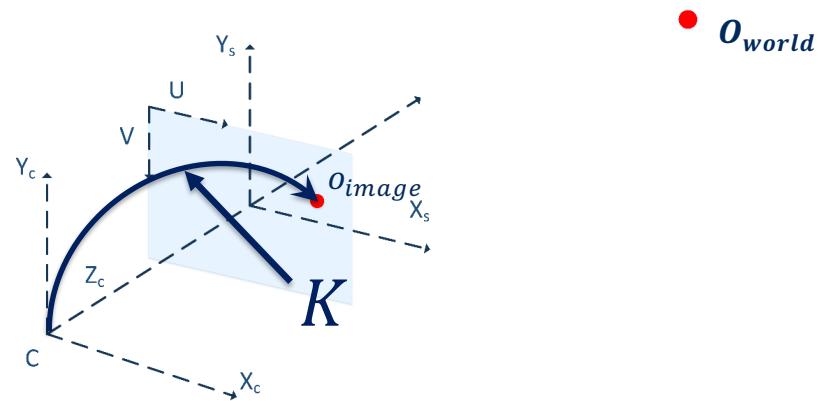
$$o_{camera} = [R|t]O_{world}$$



# Computing the Projection

- Camera → Image:

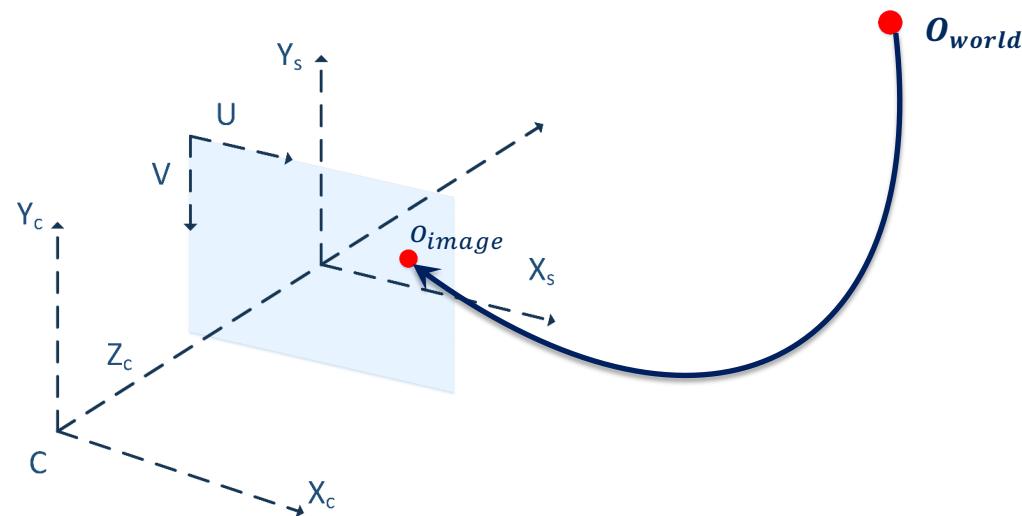
$$o_{image} = \begin{bmatrix} f & 0 & u_0 \\ 0 & f & v_0 \\ 0 & 0 & 1 \end{bmatrix} o_{camera} = K o_{camera}$$



# Computing the Projection

- World → Image:

$$P = K [R|t]$$



## Computing the Projection

- Projection from World Coordinates → Image Coordinates:

$$o_{image} = PO_{world} = K[R|t]O_{world}$$

$\underbrace{\quad\quad}_{3 \times 4} \quad \underbrace{\quad\quad}_{3 \times 4} \quad \underbrace{\quad\quad}_{4 \times 1} \quad \xrightarrow{\hspace{1cm}}$

$$\begin{bmatrix} X \\ Y \\ Z \\ 1 \end{bmatrix}$$

# Computing the Projection

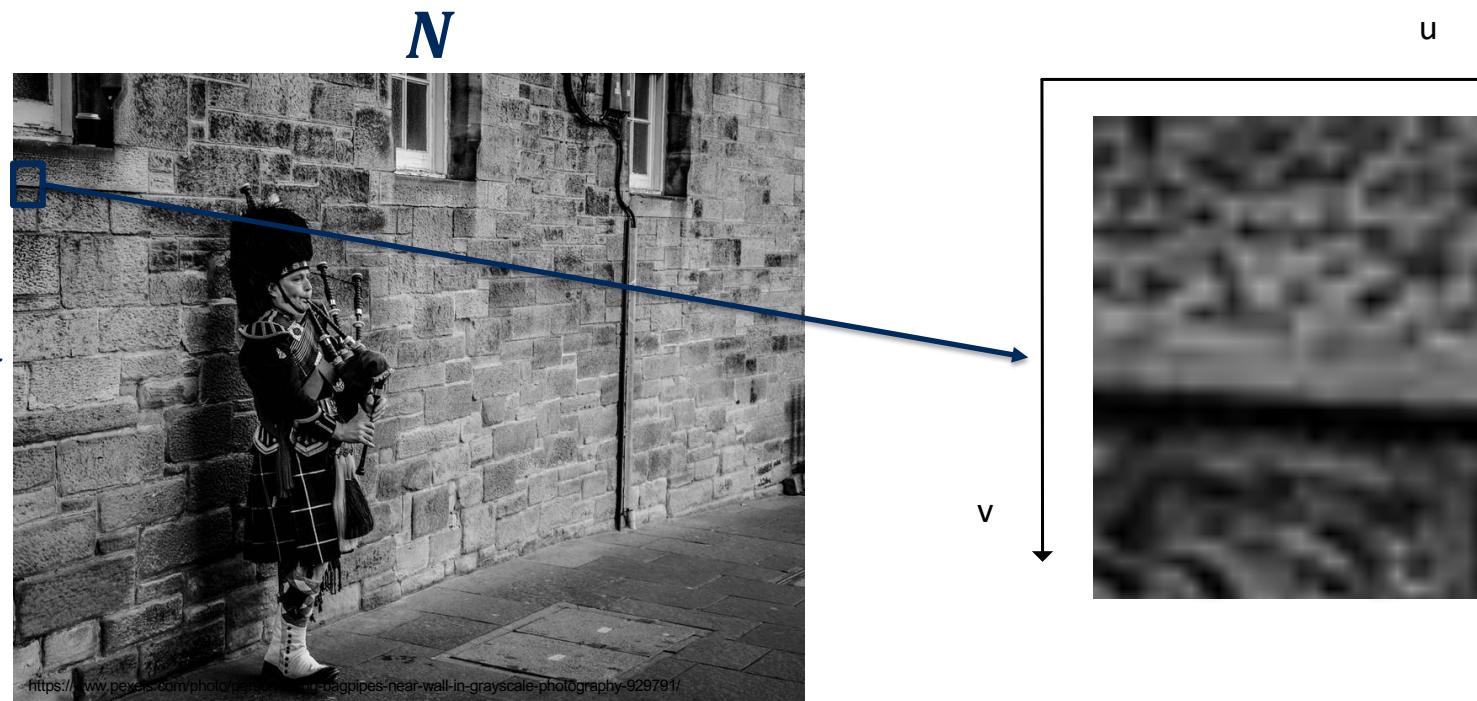
- **World** coordinates to **Image** coordinates:

$$o_{image} = \begin{bmatrix} x \\ y \\ z \end{bmatrix} = K[R|t] \begin{bmatrix} X \\ Y \\ Z \\ 1 \end{bmatrix}$$

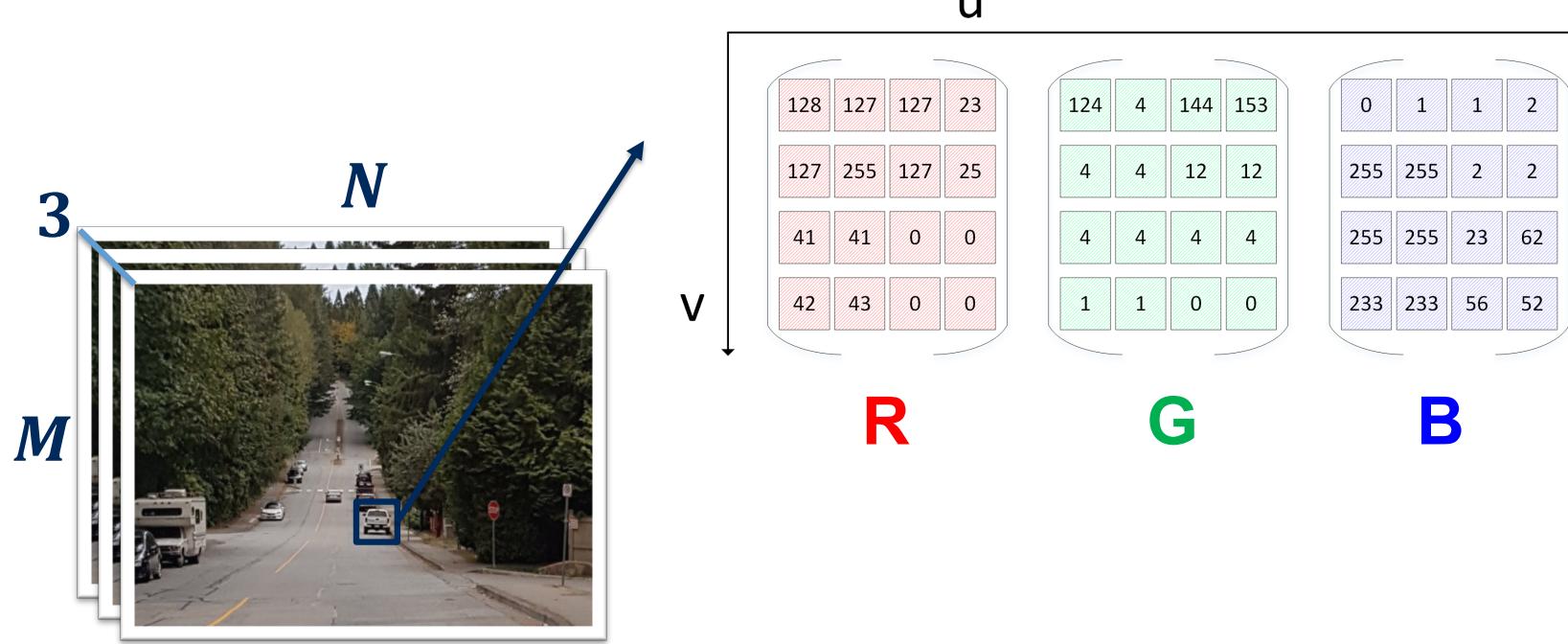
- **Image** coordinates to **Pixel** coordinates:

$$\begin{bmatrix} x \\ y \\ z \end{bmatrix} \rightarrow \begin{bmatrix} u \\ v \\ 1 \end{bmatrix} = \frac{1}{z} \begin{bmatrix} x \\ y \\ z \end{bmatrix}$$

# The Digital Image: Greyscale



# The Digital Image: Color



# Summary

- 3D points in the world coordinate frame can be projected to 2D points in the image coordinate frame using projective geometry equations
- These equations rely on the camera intrinsic parameters, as well as its extrinsic location in the world coordinate frame
- A color camera image is represented digitally as an  $M \times N \times 3$  array of unsigned 8 bit or 16 bit integers
- **Next: Camera Calibration**