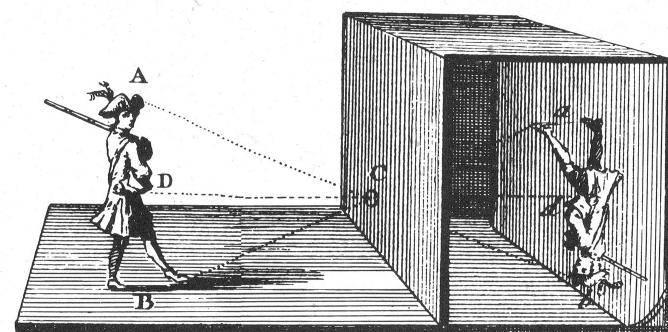
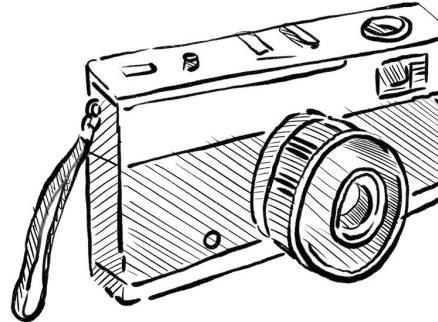


# Computer Vision

Class 02

Raquel Frizera Vassallo



# Class 02

01

Pinhole Model

02

Calibration Matrix

03

Radial Distortion

04

Experiment

01

# Image Formation

# Image Formation – Geometric Model

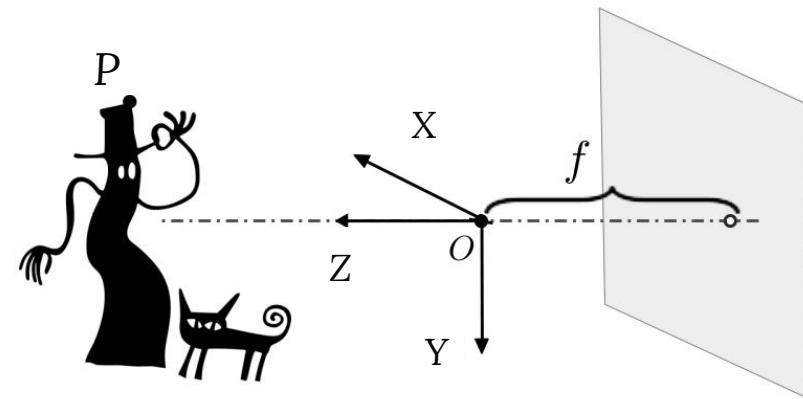
Assumptions:

- Pinhole Camera
- Lambertian Surface (uniform reflection)

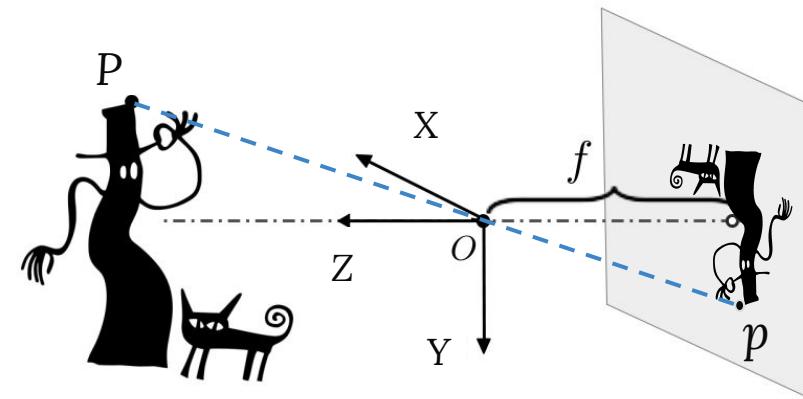
Mathematical Model accounts for 3 types of transformations:

- Coordinate transformations between the camera frame and the world frame;
- Projection of 3D coordinates onto a 2D image;
- Coordinate transformation between possible choices of image frame.

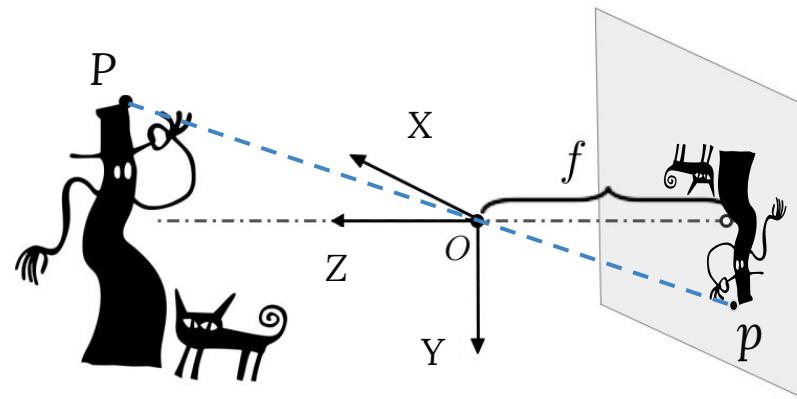
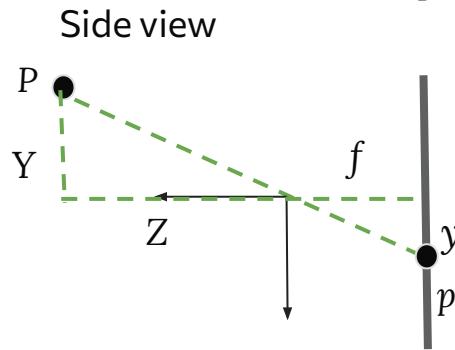
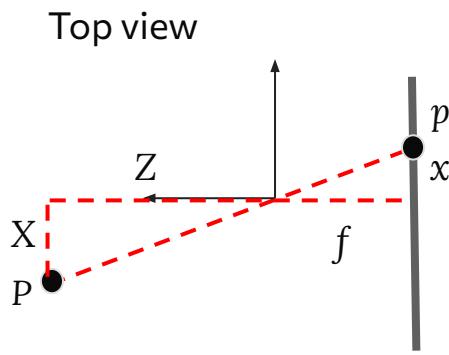
# Pinhole Model



# Pinhole Model



# Pinhole Model



$$\frac{y}{Y} = -\frac{f}{Z} \quad y = -\frac{fY}{Z}$$

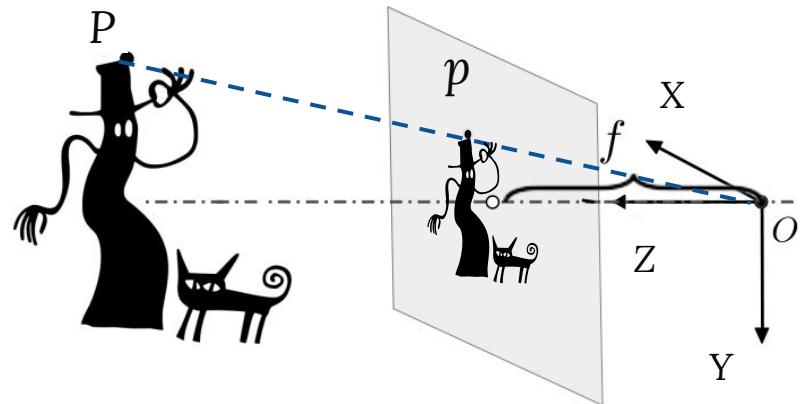
$$\frac{x}{X} = -\frac{f}{Z} \quad x = -\frac{fX}{Z}$$

# Frontal Pinhole Model

$$\frac{y}{Y} = +\frac{f}{Z} \quad y = +\frac{fY}{Z}$$

$$\frac{x}{X} = +\frac{f}{Z} \quad x = +\frac{fX}{Z}$$

$$P = \begin{bmatrix} X \\ Y \\ Z \end{bmatrix} \longrightarrow p = \begin{bmatrix} x \\ y \end{bmatrix} = \frac{f}{Z} \begin{bmatrix} X \\ Y \end{bmatrix}$$



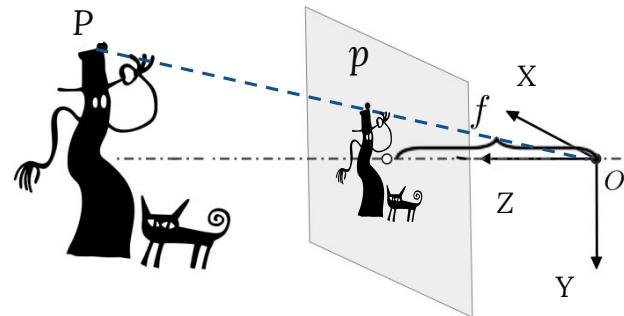
# ~~Frontal~~ Pinhole Model

## Cartesian coordinates:

$$p = \begin{bmatrix} x \\ y \end{bmatrix} = \frac{f}{Z} \begin{bmatrix} X \\ Y \end{bmatrix}$$

## Homogeneous coordinates:

$$\tilde{P} = \begin{bmatrix} X \\ Y \\ Z \\ 1 \end{bmatrix} \quad \tilde{p} = \begin{bmatrix} x \\ y \\ 1 \end{bmatrix} = \frac{1}{Z} \begin{bmatrix} fX \\ fY \\ Z \end{bmatrix}$$



## Image point

$$Z \begin{bmatrix} x \\ y \\ 1 \end{bmatrix} = \underbrace{\begin{bmatrix} f & 0 & 0 \\ 0 & f & 0 \\ 0 & 0 & 1 \end{bmatrix}}_{K_f} \underbrace{\begin{bmatrix} 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & 1 & 0 \end{bmatrix}}_{\Pi_o} \begin{bmatrix} X \\ Y \\ Z \\ 1 \end{bmatrix}$$

3D point

## Standard or canonical projection matrix

# Pinhole Model

Since Z is usually unknown, it may be represented by an arbitrary positive scalar.

$$\lambda \begin{bmatrix} x \\ y \\ 1 \end{bmatrix} = \begin{bmatrix} f & 0 & 0 \\ 0 & f & 0 \\ 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & 1 & 0 \end{bmatrix} \begin{bmatrix} X \\ Y \\ Z \\ 1 \end{bmatrix}$$

**Attention:** P (X,Y,Z) are coordinates represented in the Camera Frame.

We usually know the coordinates of 3D points in the World Frame.

The Camera Frame and the World Frame are related by a rotation and translation.

# Pinhole Model

Since Z is usually unknown, it may be represented by an arbitrary positive scalar.

$$\lambda \begin{bmatrix} x \\ y \\ 1 \end{bmatrix} = \begin{bmatrix} f & 0 & 0 \\ 0 & f & 0 \\ 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & 1 & 0 \end{bmatrix} \begin{bmatrix} X \\ Y \\ Z \\ 1 \end{bmatrix}$$

$$\begin{bmatrix} X \\ Y \\ Z \\ 1 \end{bmatrix} = \begin{bmatrix} R & T \\ \mathbf{0} & 1 \end{bmatrix} \begin{bmatrix} X_w \\ Y_w \\ Z_w \\ 1 \end{bmatrix}$$

Attention: P (X,Y,Z) are coordinates represented in the Camera Frame.

We usually know the coordinates of 3D points in the World Frame.

The Camera Frame and the World Frame are related by a rotation and translation.

Extrinsic  
Parameters

# Pinhole Model

We already have performed two of the three transformations involved in image formation when using a pinhole camera:

- Coordinate transformations between the camera frame and the world frame;
- Projection of 3D coordinates onto a 2D image.

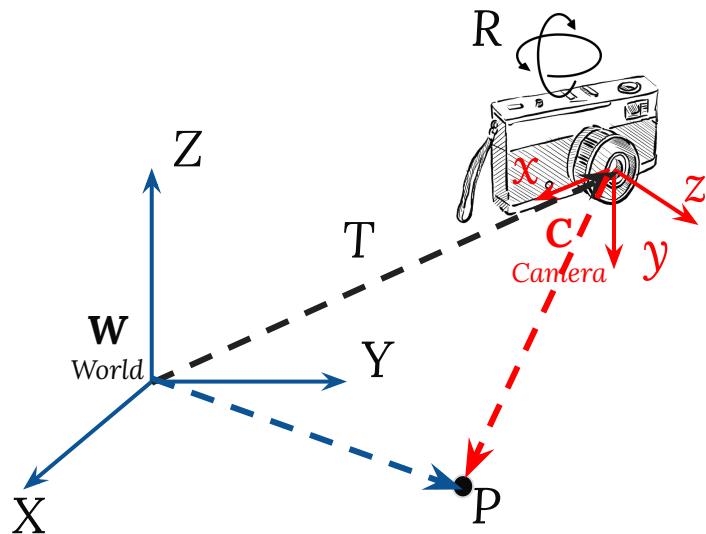
$$\lambda \begin{bmatrix} x \\ y \\ 1 \end{bmatrix} = \begin{bmatrix} f & 0 & 0 \\ 0 & f & 0 \\ 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & 1 & 0 \end{bmatrix} \begin{bmatrix} R & T \\ \mathbf{0} & 1 \end{bmatrix} \begin{bmatrix} X_w \\ Y_w \\ Z_w \\ 1 \end{bmatrix}$$

Projects 3D coordinates onto 2D coordinates losing depth information

Converts from world to camera frame

# Important to Notice

From the previous class...



If you are using **R** and **T** to represent the initial rotation and translation applied to the camera in the world frame, note that they **are not directly the Extrinsic Parameters**, since they convert from the Camera Frame to the World Frame.

$$\tilde{P}_w = \begin{bmatrix} R & T \\ \mathbf{0} & 1 \end{bmatrix} \cdot \tilde{P}_c$$

In this case, the Extrinsic Parameters will actually be:

$$\tilde{P}_c = \begin{bmatrix} R & T \\ \mathbf{0} & 1 \end{bmatrix}^{-1} \cdot \tilde{P}_w$$

02

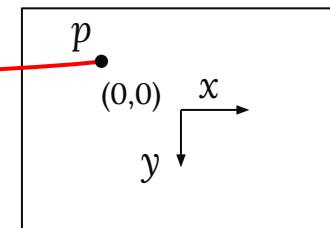
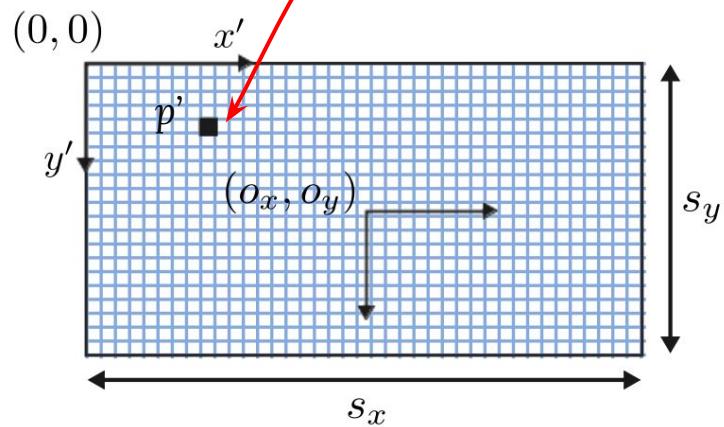
# Calibration Matrix

# Pinhole Model

Points in images are given in pixels, not in centimeters or millimeters.

Pixel  
Coordinates

$$p' = \lambda \begin{bmatrix} x' \\ y' \\ 1 \end{bmatrix}$$



$$p = \lambda \begin{bmatrix} x \\ y \\ 1 \end{bmatrix}$$

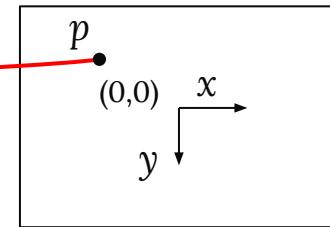
Metric  
Coordinates

# Pinhole Model

Points in images are given in pixels, not in centimeters or millimeters.

Pixel  
Coordinates

$$p' = \lambda \begin{bmatrix} x' \\ y' \\ 1 \end{bmatrix}$$



$$p = \lambda \begin{bmatrix} x \\ y \\ 1 \end{bmatrix}$$

Metric  
Coordinates

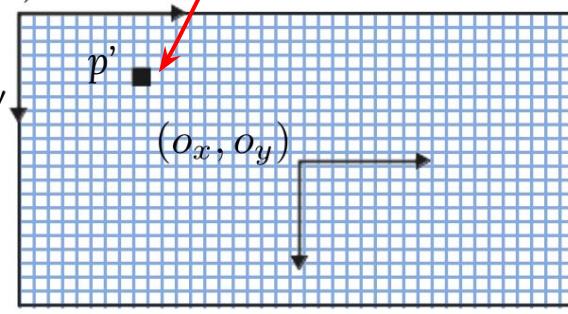
(0, 0)

$p'$

$x'$

$y'$

$(o_x, o_y)$



$s_y$

$s_x$

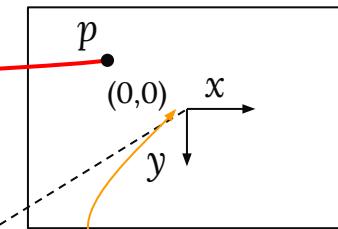
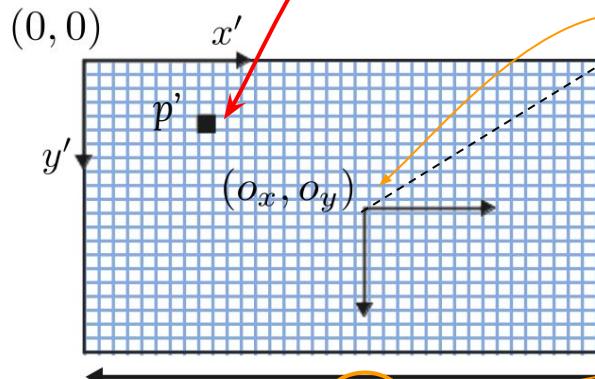
**Scale factors:** number of pixels per millimeter

# Pinhole Model

Points in images are given in pixels, not in centimeters or millimeters.

Pixel  
Coordinates

$$p' = \lambda \begin{bmatrix} x' \\ y' \\ 1 \end{bmatrix}$$



$$p = \lambda \begin{bmatrix} x \\ y \\ 1 \end{bmatrix}$$

Metric  
Coordinates

**Principal point:** where the optical axis meets the image plane

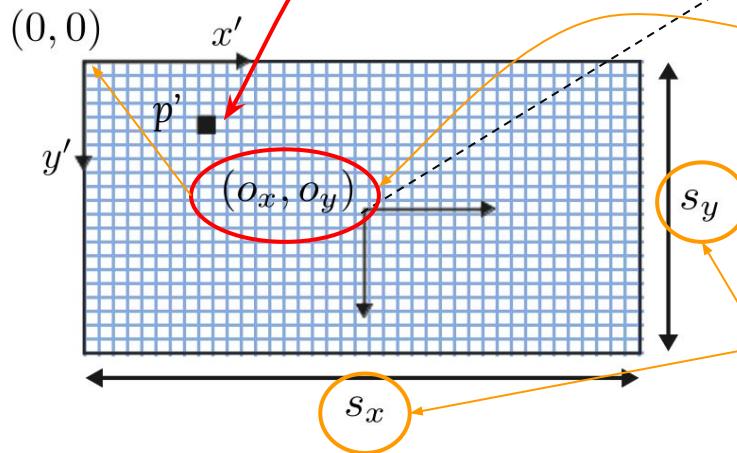
**Scale factors:** number of pixels per millimeter

# Pinhole Model

Points in images are given in pixels, not in centimeters or millimeters.

Pixel  
Coordinates

$$p' = \lambda \begin{bmatrix} x' \\ y' \\ 1 \end{bmatrix}$$



$$p = \lambda \begin{bmatrix} x \\ y \\ 1 \end{bmatrix}$$

Metric  
Coordinates

**Principal point:** where the optical axis meets the image plane

**Scale factors:** number of pixels per millimeter

# Pinhole Model

Points in images are given in pixels, not in centimeters or millimeters.

Scale factors      Skew factor      Principal point

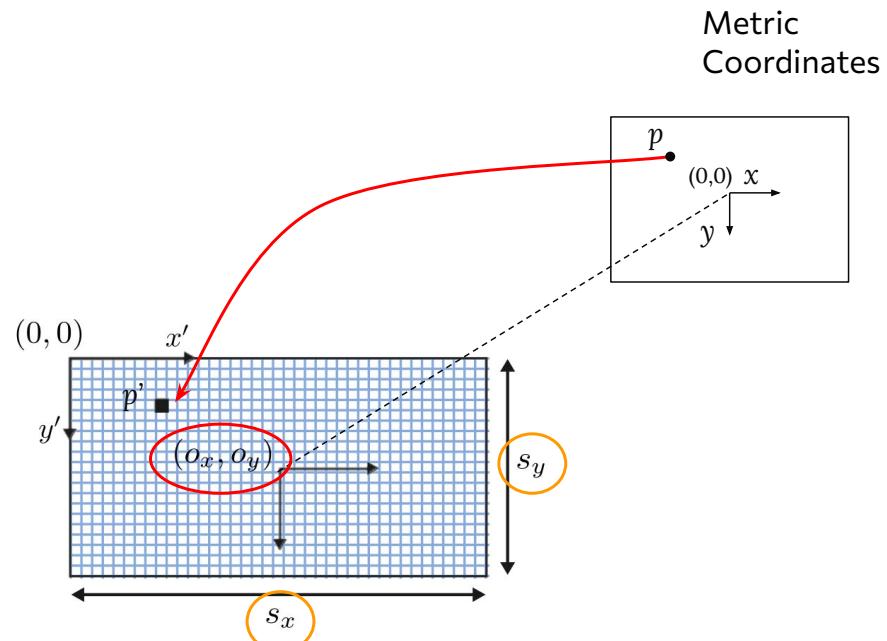
$$\begin{bmatrix} x' \\ y' \\ 1 \end{bmatrix} = \underbrace{\begin{bmatrix} s_x & s_\theta & o_x \\ 0 & s_y & o_y \\ 0 & 0 & 1 \end{bmatrix}}_{K_s} \begin{bmatrix} x \\ y \\ 1 \end{bmatrix}$$

Skew pixel

$s_\theta \rightarrow \square$

Usually  $s_\theta \sim 0$   
Square pixels

Pixel Coordinates



# Pinhole Model

$$\lambda \begin{bmatrix} x' \\ y' \\ 1 \end{bmatrix} = \begin{bmatrix} s_x & s_\theta & o_x \\ 0 & s_y & o_y \\ 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} f & 0 & 0 \\ 0 & f & 0 \\ 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & 1 & 0 \end{bmatrix} \begin{bmatrix} R & T \\ \mathbf{0} & 1 \end{bmatrix} \begin{bmatrix} X_w \\ Y_w \\ Z_w \\ 1 \end{bmatrix}$$


$$\lambda \begin{bmatrix} x' \\ y' \\ 1 \end{bmatrix} = \begin{bmatrix} fs_x & fs_\theta & o_x \\ 0 & fs_y & o_y \\ 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & 1 & 0 \end{bmatrix} \begin{bmatrix} R & T \\ \mathbf{0} & 1 \end{bmatrix} \begin{bmatrix} X_w \\ Y_w \\ Z_w \\ 1 \end{bmatrix}$$


Intrinsic Parameters Matrix or Calibration Matrix

# Pinhole Model

$$\lambda \begin{bmatrix} x' \\ y' \\ 1 \end{bmatrix} = \begin{bmatrix} fs_x & fs_\theta & o_x \\ 0 & fs_y & o_y \\ 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & 1 & 0 \end{bmatrix} \begin{bmatrix} R & T \\ \mathbf{0} & 1 \end{bmatrix} \begin{bmatrix} X_w \\ Y_w \\ Z_w \\ 1 \end{bmatrix}$$

Diagram illustrating the Pinhole Model equation:

- Image point:**  $\begin{bmatrix} x' \\ y' \\ 1 \end{bmatrix}$  (indicated by a blue arrow pointing to the first column)
- Intrinsic Parameters:**  $\begin{bmatrix} fs_x & fs_\theta & o_x \\ 0 & fs_y & o_y \\ 0 & 0 & 1 \end{bmatrix}$  (indicated by a blue arrow pointing to the second column)
- Canonical Projection Matrix:**  $\begin{bmatrix} 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & 1 & 0 \end{bmatrix}$  (indicated by a blue arrow pointing to the third column)
- Extrinsic Parameters:**  $\begin{bmatrix} R & T \\ \mathbf{0} & 1 \end{bmatrix}$  (indicated by a blue arrow pointing to the fourth column)
- 3D point:**  $\begin{bmatrix} X_w \\ Y_w \\ Z_w \\ 1 \end{bmatrix}$  (indicated by a blue arrow pointing to the fifth column)

$$\lambda p' = K \Pi_0 g(R, T) P_w$$

PS:  $fs_x$  and  $fs_y$  are called the focal distance in pixels at the horizontal and vertical directions

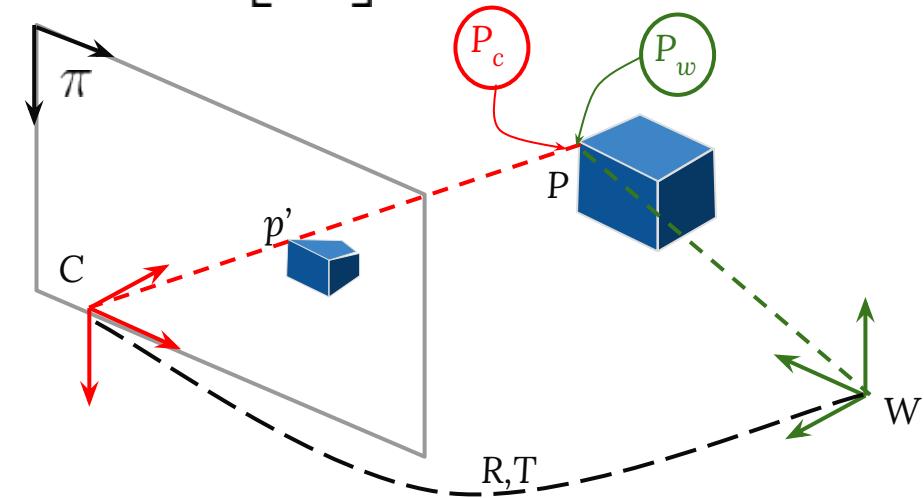
# Pinhole Model

$$\lambda \begin{bmatrix} x' \\ y' \\ 1 \end{bmatrix} = \begin{bmatrix} fs_x & fs_\theta & o_x \\ 0 & fs_y & o_y \\ 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & 1 & 0 \end{bmatrix} \begin{bmatrix} R & T \\ \mathbf{0} & 1 \end{bmatrix} \begin{bmatrix} X_w \\ Y_w \\ Z_w \\ 1 \end{bmatrix}$$

$$\lambda p' = K \Pi_0 g(R, T) P_w$$

$\downarrow$

$$\lambda p' = \Pi P_w$$



# Pinhole Model

Considering just the extrinsic parameters, if we are not working in pixels:

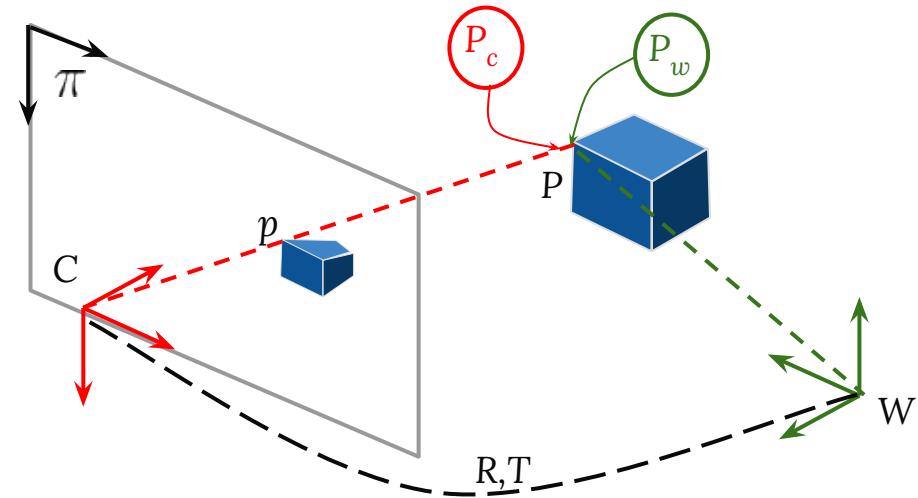
$$\lambda p = \Pi_0 g(R, T) P_w$$

$$\lambda p = \Pi P_w \rightarrow \boxed{\lambda p = [R, T]_{3 \times 4} P_w}$$

Considering the complete model, if we are working in pixels:

$$\lambda p' = K \Pi_0 g(R, T) P_w$$

$$\lambda p' = \Pi P_w \rightarrow \boxed{\lambda p' = [KR, KT]_{3 \times 4} P_w}$$

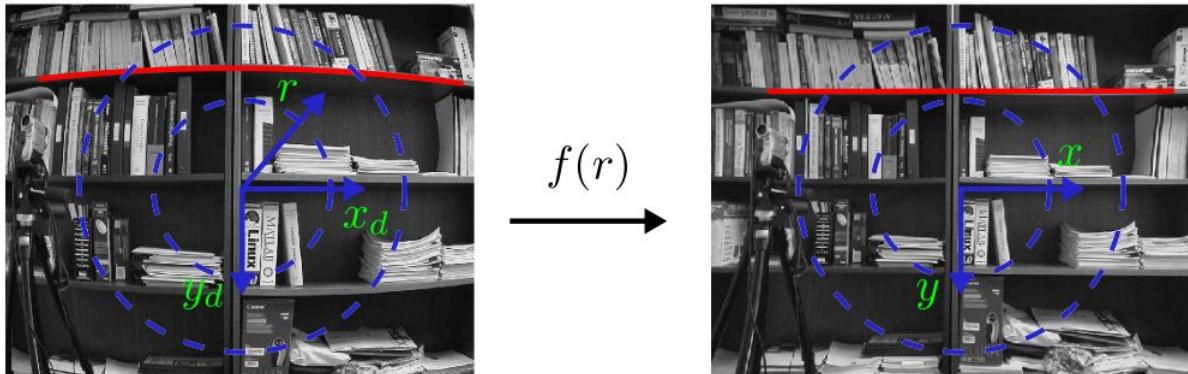


03

# Radial Distortion

# Radial Distortion

Non-linear Transformation along radial direction



Correction:  
straight lines

Corrected point  $\rightarrow p = c + f(r)(p_d - c)$ , with  $r = \|p_d - c\|$

Distorted point

$$f(r) = 1 + a_1 r + a_2 r^2 + a_3 r^3 + a_4 r^4 + \dots$$

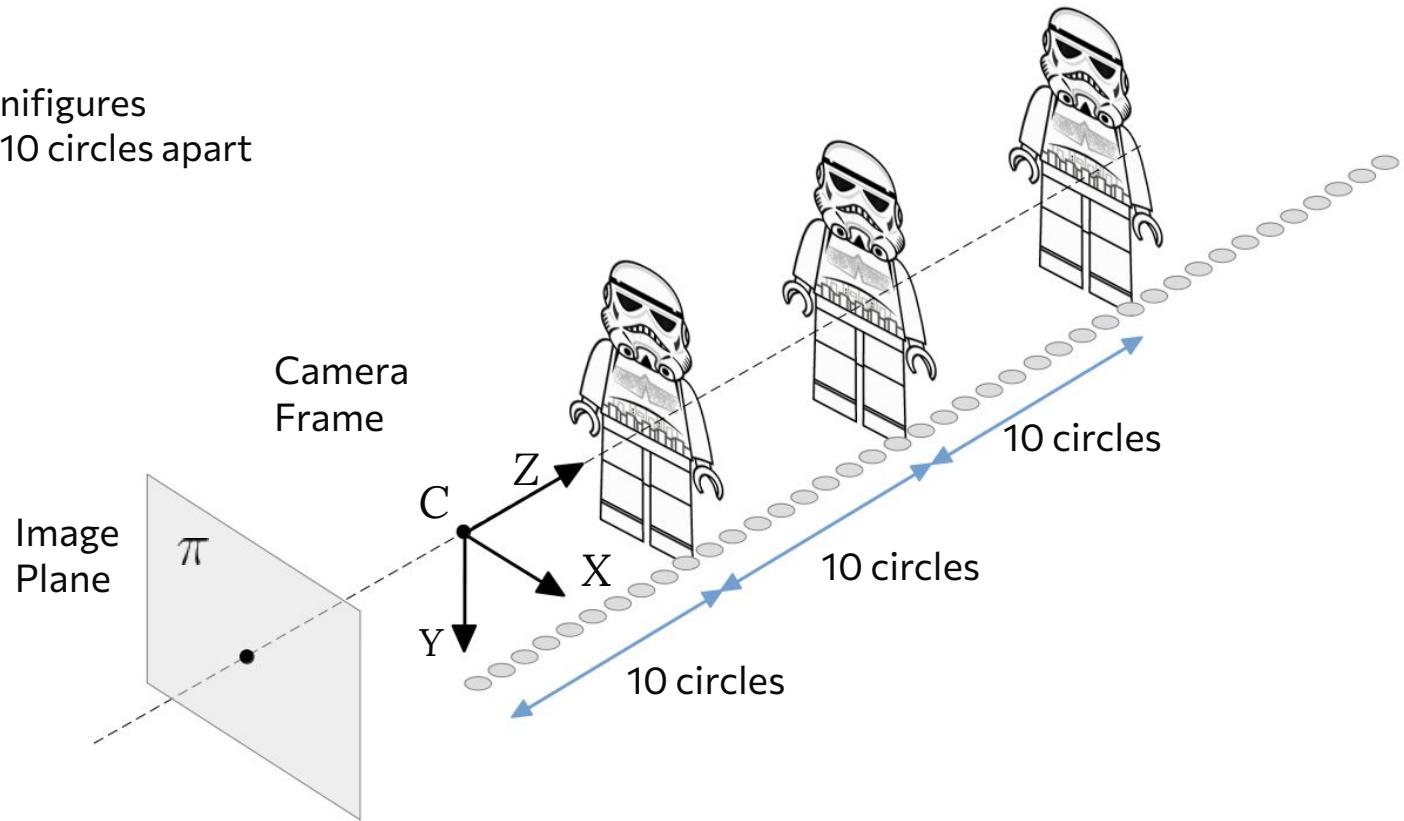
04

# Experiment



# Experiment

- 3 Lego Minifigures
- Distance: 10 circles apart



# Experiment

- 3 Lego Minifigures
- Same dimensions



# Experiment

- 3 Lego Minifigures
- Same dimensions
- 10 circles apart to each other



# Experiment

Basic Perspective Projection

$$x = \frac{fX}{Z} \quad y = \frac{fY}{Z}$$



# Experiment

## Basic Perspective Projection

All 3 minifigures have the same dimensions, but different depth (distance) from the camera.

$$x = \frac{fX}{Z}$$
$$y = \frac{fY}{Z}$$



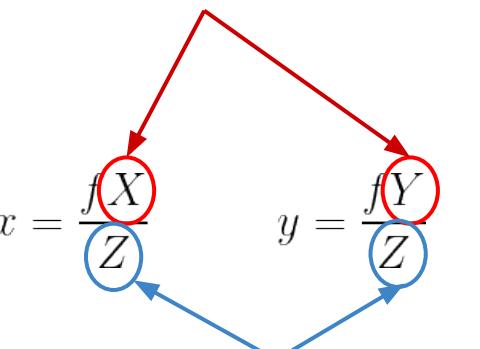
# Experiment

## Basic Perspective Projection

All 3 minifigures have the same dimensions, but different depth (distance) from the camera.

$$x = \frac{fx}{Z}$$
$$y = \frac{fy}{Z}$$

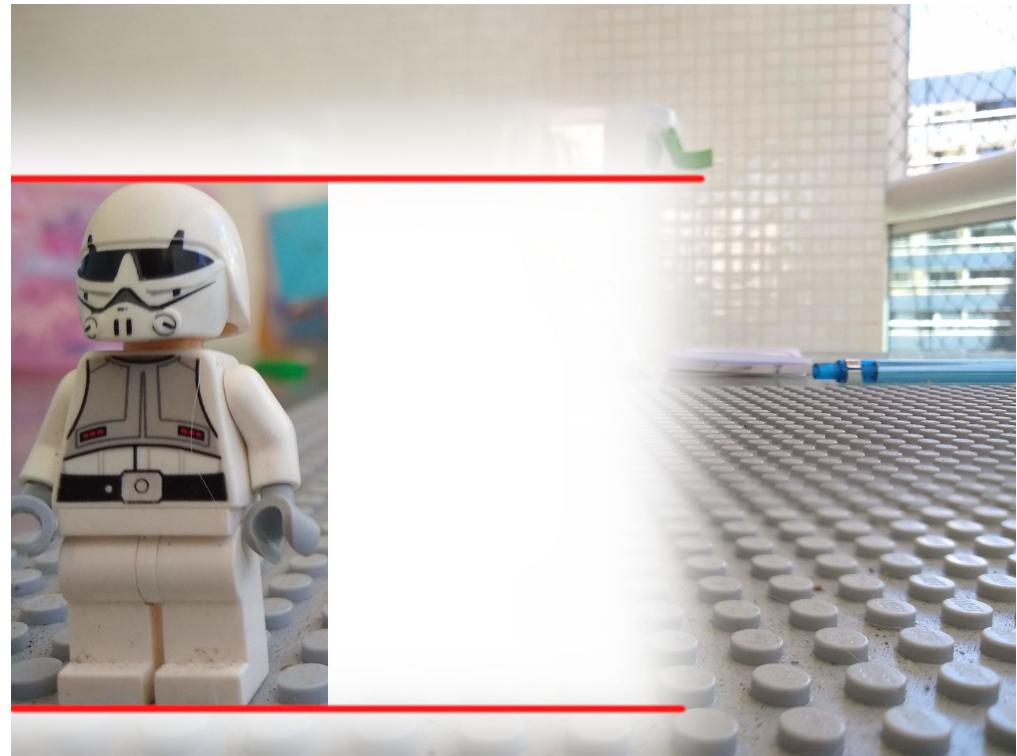
Inversely proportional to the depth.





# Experiment

$$x = \frac{fX}{Z} \quad y = \frac{fY}{Z}$$

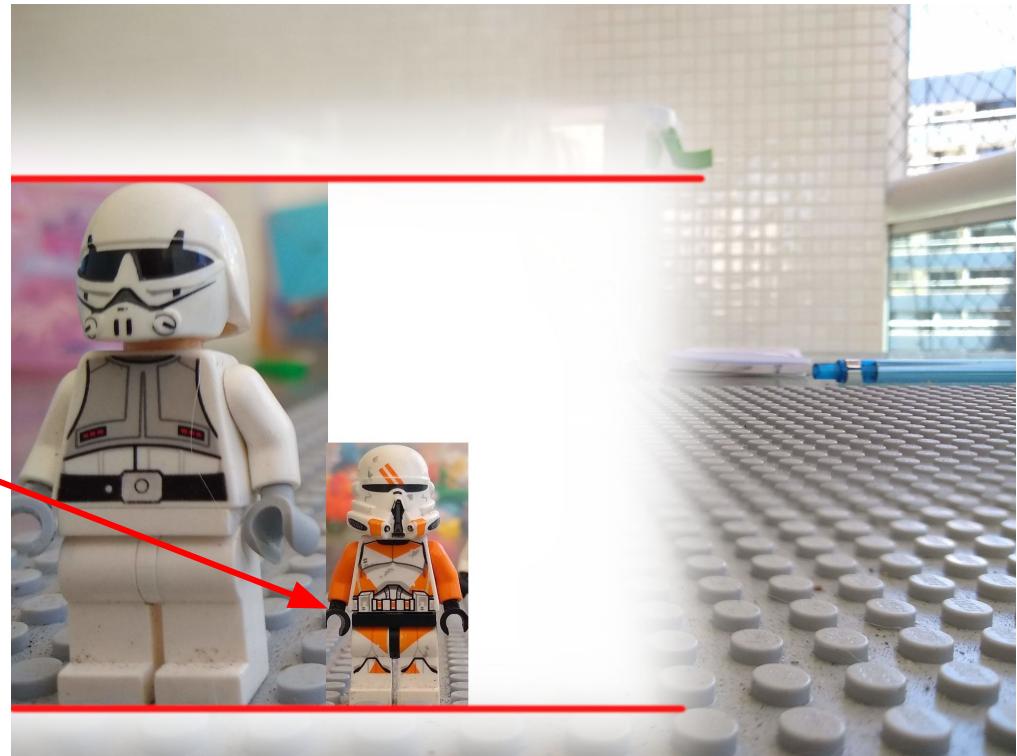


# Experiment

$$x = \frac{fX}{Z} \quad y = \frac{fY}{Z}$$

Twice the distance

$$x' = \frac{fX}{2Z} \quad y = \frac{fY}{2Z}$$



# Experiment

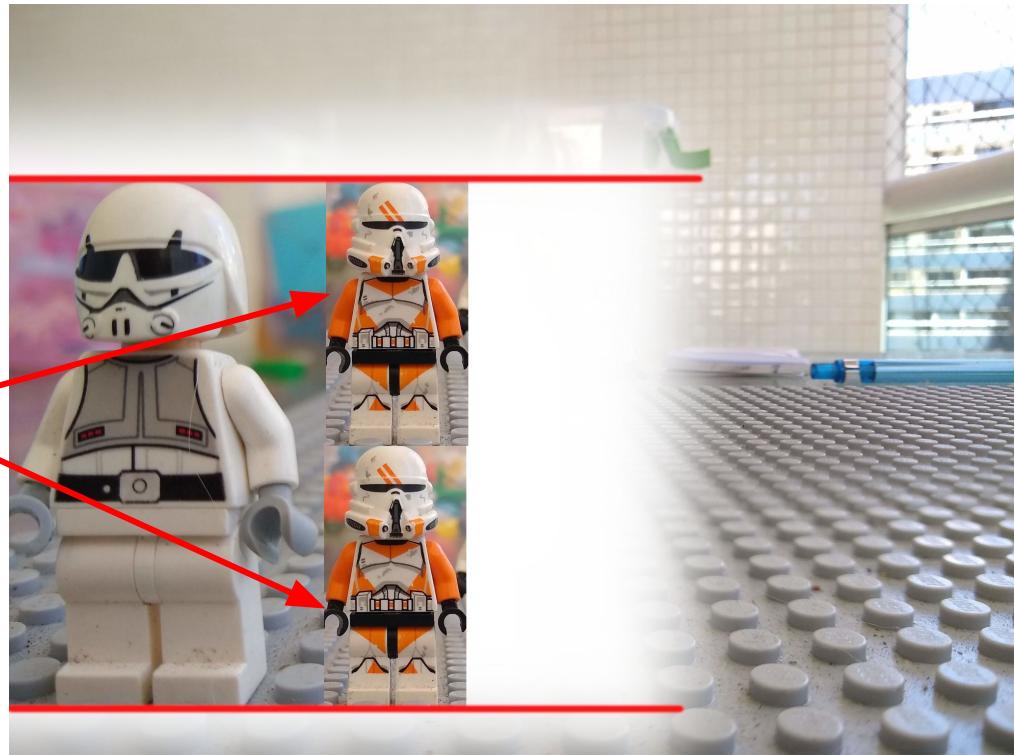
$$x = \frac{fX}{Z} \quad y = \frac{fY}{Z}$$

Twice the distance

$$x' = \frac{fX}{2Z}$$

$$y = \frac{fY}{2Z}$$

Half the size



# Experiment

$$x = \frac{fX}{Z} \quad y = \frac{fY}{Z}$$

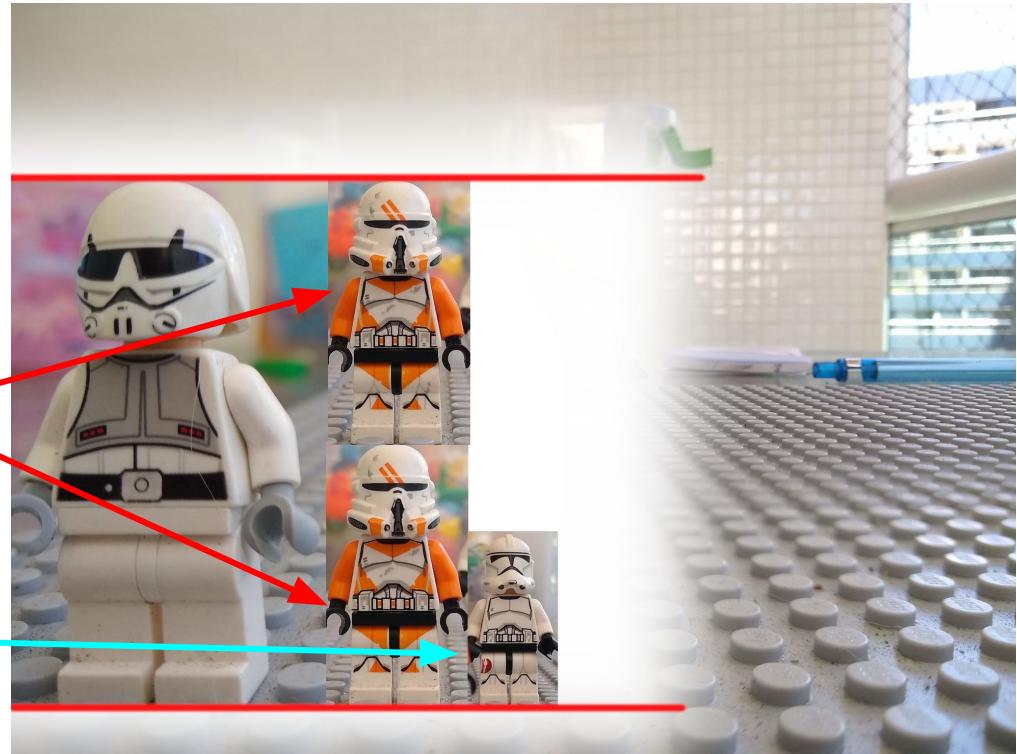
Twice the distance

$$x' = \frac{fX}{2Z} \quad y = \frac{fY}{2Z}$$

Half the size

Triple the distance

$$x'' = \frac{fX}{3Z} \quad y = \frac{fY}{3Z}$$



# Experiment

$$x = \frac{fX}{Z} \quad y = \frac{fY}{Z}$$

Twice the distance

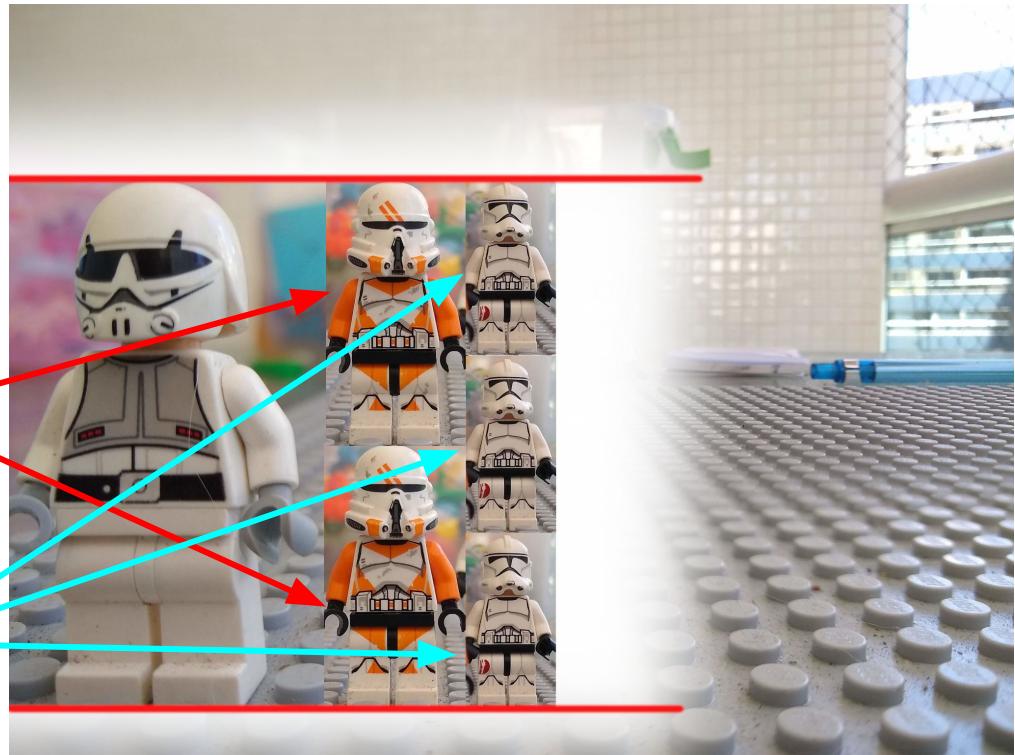
$$x' = \frac{fX}{2Z} \quad y = \frac{fY}{2Z}$$

Half the size

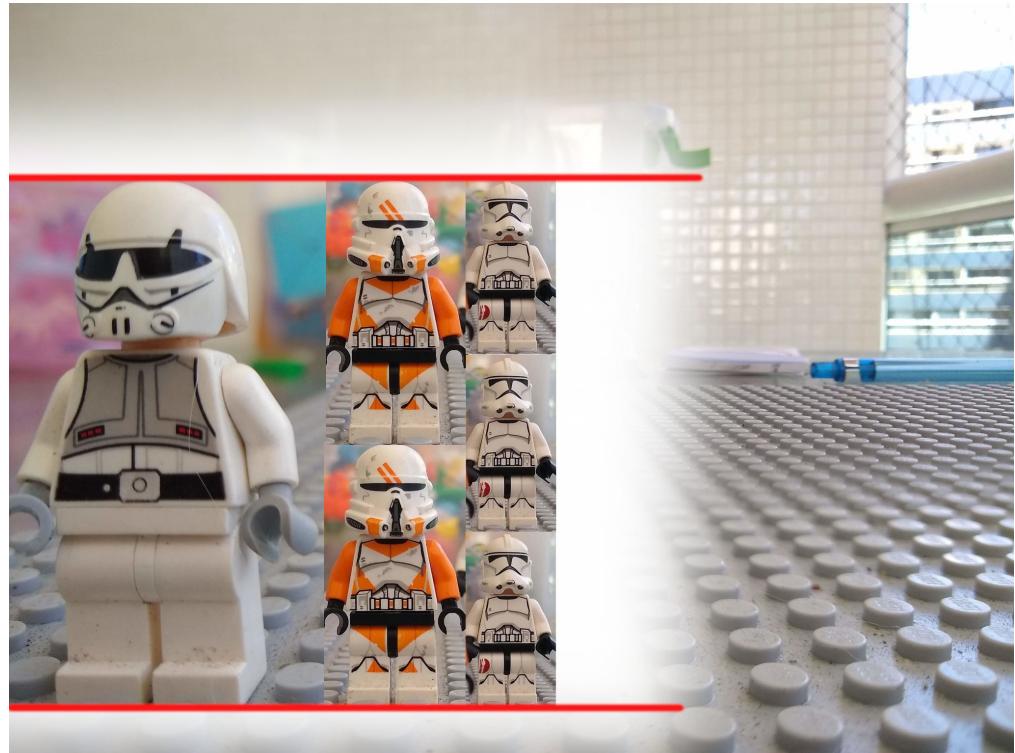
Triple the distance

$$x'' = \frac{fX}{3Z} \quad y = \frac{fY}{3Z}$$

Third the size



# Experiment



# Credits



Yi Ma, Stefano Soatto, Jana Kosecka e S. Shankar Sastry. An Invitation to 3D Vision: From Images to Geometric Models.  
Springer, ISBN 0387008934