# Introduction to 3D Imaging: Perceiving 3D from 2D Images 

How can we derive 3D information from one or more 2D images?

There have been 2 approaches:

1. intrinsic images: a 2D representation that stores some 3 D properties of the scene
2. 3D shape from X : methods of inferring 3 D depth information from various sources

What can you determine about

1. the sizes of objects
2. the distances of objects from the camera?


# What objects are shown in this image? <br> How can you estimate distance from the camera? <br> What feature changes with distance? 



## Intrinsic Images: 2.5 D

The idea of intrinsic images is to label features of a 2D image with information that tells us something about the 3D structure of the scene.


## Contour Labels for Intrinsic Images

- convex crease (+)
- concave crease (-)
- blade (>)

- limb (>>)
- shadow (S)

- illumination boundary (I)
- reflectance boundary (M)


## Labeling Simple Line Drawings

- Huffman and Clowes showed that blocks world drawings could be labeled (with +, -, >) based on real world constraints.
- Labeling a simple blocks world image is a consistent labeling problem!
- Waltz extended the work to cracks and shadows and developed one of the first discrete relaxation algorithms, known as Waltz filtering.


## Problems with this Approach

- Research on how to do these labelings was confined to perfect blocks world images
- There was no way to extend it to real images with missing segments, broken segments, nonconnected junctions, etc.
- It led some groups down the wrong path for a while.


## 3D Shape from $X$

- shading
- silhouette
- texture
- stereo
- light striping
- motion
mainly research
used in practice


## Perspective Imaging Model: 1D



This is the axis of the real image plane.

O is the center of projection.
This is the axis of the front image plane, which we use.

$$
\frac{x i}{f}=\frac{x c}{z c}
$$

## Perspective in 2D (Simplified)

3D object point $\mathrm{P}=(\mathrm{xc}, \mathrm{yc}, \mathrm{zc})$ =(xw,yw,zw)


Here camera coordinates equal world coordinates.

$$
\frac{y i}{f}=\frac{y c}{z c}
$$

$$
\begin{aligned}
& \mathrm{xi}=(\mathrm{f} / \mathrm{zc}) \mathrm{xc} \\
& \mathrm{yi}=(\mathrm{f} / \mathrm{zc}) \mathrm{yc}
\end{aligned}
$$

## 3D from Stereo

O 3D point

left image
right image
disparity: the difference in image location of the same 3D point when projected under perspective to two different cameras.
d = xleft - xright

## Depth Perception from Stereo Simple Model: Parallel Optic Axes



$$
\frac{\mathrm{z}}{\mathrm{f}}=\frac{\mathrm{x}}{\mathrm{xl}} \quad \frac{\mathrm{z}}{\mathrm{f}}=\frac{\mathrm{x}-\mathrm{b}}{\mathrm{xr}} \quad \frac{\mathrm{z}}{\mathrm{f}}=\frac{\mathrm{y}}{\mathrm{yl}}=\frac{\mathrm{y}}{\mathrm{yr}} \quad \begin{gathered}
\mathrm{y} \text {-axis is } \\
\text { perpendicular } \\
\text { to the page. }
\end{gathered}
$$

## Resultant Depth Calculation

For stereo cameras with parallel optical axes, focal length f , baseline b , corresponding image points ( $\mathrm{xl}, \mathrm{yl}$ ) and (xr,yr) with disparity d :

$$
\begin{aligned}
& \mathrm{z}=\mathrm{f} * \mathrm{~b} /(\mathrm{xl}-\mathrm{xr})=\mathrm{f} * \mathrm{~b} / \mathrm{d} \\
& \mathrm{x}=\mathrm{xl} \mathrm{l}^{*} \mathrm{z} / \mathrm{f} \text { or } \mathrm{b}+\mathrm{xr}^{*} \mathrm{z} / \mathrm{f} \\
& \mathrm{y}=\mathrm{yl} * \mathrm{z} / \mathrm{f} \quad \text { or } \quad \mathrm{yr} \mathrm{r}^{*} \mathrm{z} / \mathrm{f}
\end{aligned}
$$

## Finding Correspondences

- If the correspondence is correct, triangulation works VERY well.
- But correspondence finding is not perfectly solved. for the general stereo problem.
- For some very specific applications, it can be solved for those specific kind of images, e.g. windshield of a car.



## 2 Main Matching Methods

1. Cross correlation using small windows.


## dense

2. Symbolic feature matching, usually using segments/corners.

sparse

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## Epipolar Geometry Constraint: 1. Normal Pair of Images

The epipolar plane cuts through the image plane(s) forming 2 epipolar lines.


The match for P1 (or P2) in the other image, must lie on the same epipolar line.

## Epipolar Geometry: General Case



## Constraints

1. Epipolar Constraint: Matching points lie on corresponding epipolar lines.
2. Ordering Constraint: Usually in the same order across the lines.

## Structured Light

3D data can also be derived using

- a single camera
- a light source that can produce stripe(s) on the 3D object

light

source
camera

## Structured Light 3D Computation

3D data can also be derived using

- a single camera
- a light source that can produce stripe(s) on the 3D object
 source


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## Depth from Multiple Light Stripes



What are these objects?

## Our (former) System 4-camera light-striping stereo



