

Computer Vision Systems Programming VO 3D Vision

Christopher Pramerdorfer
Computer Vision Lab, Vienna University of Technology

Topics

Image formation

3D data acquisition



mages from wikipedia.org, createiveapplications.net

Motivation

Many CV applications rely on knowledge of scene geometry

This lecture covers

- ▶ How scene geometry and images are related
- ▶ How this relation can be used to recover scene geometry

Image Formation

Pinhole Camera Model

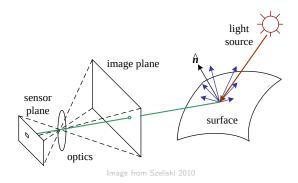
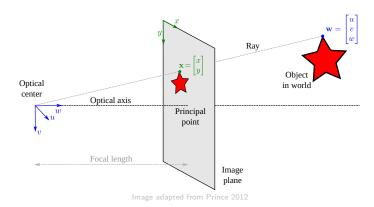


Image Formation

Pinhole Camera Model



We obtain $x = fu/w + p_x$, $y = fv/w + p_y$

- ightharpoonup f: focal length in pixels
- $ightharpoonup p_x, p_y$: image coordinates of the principal point

This mapping is linear in homogeneous coordinates

$$\lambda \tilde{\mathbf{x}} = (\mathbf{\Lambda} \quad \mathbf{0}) \tilde{\mathbf{w}}$$

$$\lambda \begin{pmatrix} x \\ y \\ 1 \end{pmatrix} = \begin{pmatrix} f & 0 & p_x & 0 \\ 0 & f & p_y & 0 \\ 0 & 0 & 1 & 0 \end{pmatrix} \begin{pmatrix} u \\ v \\ w \\ 1 \end{pmatrix}$$

World and camera coordinate systems generally differ

► Transform w to camera coordinates before projection

$$\mathbf{w}' = \mathbf{\Omega}\mathbf{w} + \mathbf{\tau}
\begin{pmatrix} u' \\ v' \\ w' \end{pmatrix} = \begin{pmatrix} \omega_{11} & \omega_{12} & \omega_{13} \\ \omega_{21} & \omega_{22} & \omega_{23} \\ \omega_{31} & \omega_{32} & \omega_{33} \end{pmatrix} \begin{pmatrix} u \\ v \\ w \end{pmatrix} + \begin{pmatrix} \tau_u \\ \tau_v \\ \tau_w \end{pmatrix}$$

We obtain the full pinhole camera model

$$\lambda \tilde{\mathbf{x}} = \begin{pmatrix} \mathbf{\Lambda} & \mathbf{0} \end{pmatrix} \begin{pmatrix} \mathbf{\Omega} & \mathbf{\tau} \\ \mathbf{0}^\top & 1 \end{pmatrix} \tilde{\mathbf{w}}$$

Standard camera model in CV

Usually together with radial distortion correction

Approximation to actual image formation

▶ In practice w is not mapped to a single x



We can obtain w by inverting the pinhole camera model

ightharpoonup But we don't know w

To this end, we must

- Utilize information from multiple images
- lacktriangle Use sensors that capture w directly



Image by John Kratz / flickr

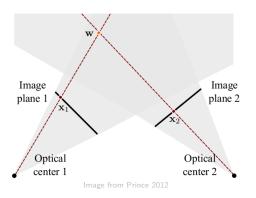
In stereo reconstruction we have

- ▶ Point correspondences $\{(\mathbf{x}_1, \mathbf{x}_2)\}$ in two images
- ▶ Taken with calibrated cameras (known Λ, Ω, au)

Goal is to estimate corresponding world coordinates $\ensuremath{\mathbf{w}}$

Accomplished via triangulation





The challenge is finding correspondences

We typically want

- Many correspondences
- ► High accuracy and low noise

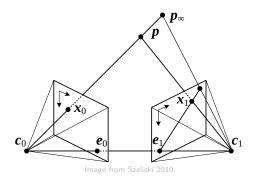
Usually accomplished via

- Dense feature matching along epipolar lines
- ► Followed by local or global optimization



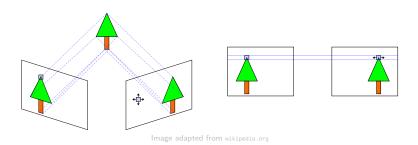
x_1 must lie on the **epipolar line**

ightharpoonup Given by \mathbf{x}_0 and the camera parameters



Images are rectified before correspondence search

- ▶ Relation between x-offset (**disparity** d) and w, d = fb/w
- b is the distance between the cameras



Dense matching on rectified images results in a disparity map





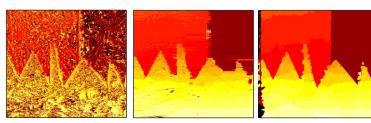




Raw disparity maps are noisy

Quality can be improved by encouraging smoothness

Accomplished via graphical models (e.g. MRFs)



Images from Prince 2012

Limitations of image-based (passive) stereo

- ▶ No correspondences in regions without texture
- Relies on proper illumination (no dark living rooms)
- Computational complexity



Computing Scene Geometry Depth Sensors

Alternatively, we can use sensors that capture w directly

▶ Usually together with brightness or color

These depth sensors

- Do not rely on texture
- Work under dim or dark conditions
- Save computational resources



Released by Microsoft for Xbox 360 in late 2010

Fastest selling consumer electronics device to date



Works by replacing one sensor with an IR source

Projects a speckle pattern onto objects https://www.youtube.com/watch?v=t5joFtzEYpo

Pattern is observed using an IR sensor

Stereo (epipolar) geometry still applies

lacktriangle Shift between patterns corresponds to d respectively w

Does not work in sunlight due to IR radiation







Image from https://www.youtube.com/watch?v=o0CMr0D7Bq

Depth Sensors – Kinect v2

Released by Microsoft in late 2013



Image from wikipedia.org

Works based on the time of flight principle

- lacksquare A light pulse is emitted at time t_0
- lacktriangle Pulse is reflected and observed at time t_1
- w proportional to delay, $t_1 t_0 = 2w/c$

Depth Sensors - Kinect v2



mage from https://www.youtube.com/watch?v=Ja01Ua57BWs

We now have an image encoding w (depth map)

▶ Can be generated from disparity map as shown above

We can use this information to obtain points in the world $\ensuremath{\mathbf{w}}$

- By inverting the pinhole camera model
- Resulting in a point cloud



Image from creativeapplications.net

Summary

We have covered

- ▶ How the scene geometry and images are related
- ightharpoonup Means for estimating point distances w
- ▶ How scene geometry can be recovered on this basis

This information enables interesting CV applications

▶ Next, we will go over some examples

3D Vision Lecture

Interested in 3D vision?

▶ There is an own VO (183.129) and UE (183.130)

Bibliography

Prince, S.J.D. (2012). **Computer Vision: Models Learning and Inference**. Cambridge University Press.

Szeliski, Richard (2010). **Computer vision: algorithms and applications**. Springer.

