# Three-Dimensional Sensors Lecture 2: Projected-Light Depth Cameras

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### Outline

- The geometry of active stereo.
- Surface reflectivity.
- Camera-projector calibration
- Active stereo matching and reconstruction

## Generalities

- The basic principle is to associate a light projector with a standard camera.
- The projector and the camera share the same geometric model.
- Optical (non-linear distorsions) are present with both devices. They will be treated separately.
- 3D reconstruction is based on the same principle as stereo: sensor calibration and matching.
- Instead of matching two images, we match a pattern with "itself" after being backscattered by scene objects.

# The Geometry of Active Stereo

- Perspective effects (foreshortening).
- Surface discontinuities.
- Cluttered scenes.

# Foreshortening



### Discontinuities



# Clutter



# Surface Reflectivity and External Illumination

- This is a scientific topic in its own right.
- Reflectance (or absorption) depends on the material properties and of the wavelength of the incident light.
- A rough way to think of surface reflectivity is to divide surfaces into **diffuse** (matte) and **specular** (mirror).
- In general a surface is a combination of these effects plus other ones: metal, glossy, rough, transparent, translucent, etc. cause problems in practice.
- External illumination (daylight, artificial light, etc.) acts as *noise* added to the infra-red light emitted by the projector.

# Reflectivity



# Artifacts with the Kinect Camera



# Back to Geometry

- Consider an IR camera and a projector in general positions (before rectification).
- Parameters to be estimated:
  - The internal parameters of the IR camera
  - The external parameters of the IR camera
  - The projection matrix of the light projector

# IR-Camera and Projector Geometry and Calibration



### External IR-Camera Calibration

• The basic equation is  $m \approx \begin{bmatrix} R & t \end{bmatrix} M'$  where M' lies on the plane Z' = 0, hence we obtain:

$$\left(\begin{array}{c} sx\\ sy\\ s\end{array}\right) = \left[\begin{array}{cc} \boldsymbol{r}_1 & \boldsymbol{r}_2 & \boldsymbol{t}\end{array}\right] \left(\begin{array}{c} X'\\ Y'\\ 1\end{array}\right)$$

• Each correspondence between an image point  $m_i$  and a calibration point  $M'_i$  yields two homogeneous linear equations in the unknown extrinsics:

$$\begin{cases} x_i(X'_ir_{13} + Y'_ir_{23} + t_3) &= X'_ir_{11} + Y'_ir_{12} + t_1 \\ y_i(X'_ir_{13} + Y'_ir_{23} + t_3) &= X'_ir_{21} + Y'_ir_{22} + t_2 \end{cases}$$

In practice it is useful to constrain the parameters of the rotation matrix, i.e., ||r<sub>1</sub>|| = ||r<sub>2</sub>|| = 1, r<sub>1</sub><sup>⊤</sup>r<sub>2</sub> = 0. Various optimization methods can be used in practice.

#### **External IR-Camera Parameters**

• The external camera matrix is:

• The coordinates of the planar set of points in the IR-camera frame:

$$\left(\begin{array}{c}X\\Y\\Z\end{array}\right) = \left[\begin{array}{cc}r_1 & r_2 & t\end{array}\right] \left(\begin{array}{c}X'\\Y'\\1\end{array}\right)$$

• Recall the simple relationship between m and M when the latter is in camera frame:

$$\left(\begin{array}{c} X\\Y\\Z\end{array}\right) = Z \left(\begin{array}{c} x\\y\\1\end{array}\right)$$

# Point/Plane Duality in Projective Space

• The point M belongs to the plane  $\mu$ :

$$\alpha X + \beta Y + \gamma Z + \delta H = 0$$
 or  $\boldsymbol{\mu}^{\top} \tilde{\boldsymbol{M}} = 0$ 

with the notation  $\tilde{M}=\left( egin{array}{ccc} X & Y & Z & H \end{array} 
ight)$  and H=1.

The 4 × 4 point-transformation D (a rotation followed by a translation) maps the coordinates of a point from the calibration frame to the camera frame: *M* = D*M*<sup>'</sup>
Let µ<sup>'</sup><sup>⊤</sup>*M*<sup>'</sup> = 0 with µ<sup>'</sup> = ( 0 0 1 0 ) be the plane of the calibration pattern. Substituting *M*<sup>'</sup> we obtain (D<sup>-⊤</sup>µ<sup>'</sup>)<sup>⊤</sup>*M* = 0 with:

$$\mathbf{D}^{- op} = \left[egin{array}{cc} \mathbf{R} & \mathbf{0} \ -m{t}^ op \mathbf{R} & 1 \end{array}
ight]$$
 and  $m{\mu} = \mathbf{D}^{- op} m{\mu}'$ 

• The latter is the plane-transformation or the dual of the point-transformation.

## **Projector Calibration**

- We assume that the planar calibration pattern is in the same position as before. A ray of light through the projector image-point  $\boldsymbol{p} = (u \; v \; 1)^{\top}$  is  $\boldsymbol{p} \approx \mathbf{P} \tilde{\boldsymbol{M}}$ .
- This ray of light is observed in the IR-camera image at m. Recall that M = Zm and that  $\mu^{\top} \tilde{M} = 0$ . Hence the depth Z is:

$$Z = -\frac{\delta}{\alpha x + \beta y + \gamma}$$

- For each camera-projector point correspondence  $oldsymbol{m}_i \leftrightarrow oldsymbol{p}_i$  we obtain two equations:
- $\begin{cases} u_i[Z_i(x_ip_{31} + y_ip_{32} + p_{33}) + p_{34}] &= Z_i(x_ip_{11} + y_ip_{12} + p_{13}) + p_{14} \\ v_i[Z_i(x_ip_{31} + y_ip_{32} + p_{33}) + p_{34}] &= Z_i(x_ip_{21} + y_ip_{22} + p_{23}) + p_{24} \end{cases}$ 
  - This requires several such correspondences and several parallel planes.

# Projector's Intrinsics and Extrinsics

- If enough correspondences are available, one can estimate the projection matrix **P** either linearly or with the help of a non-linear optimization method (bundle adjustment).
- Intrinsic and extrinsic parameter can be made explicit by identification:

$$\mathsf{P} pprox \mathsf{K} \begin{bmatrix} \mathsf{R} & t \end{bmatrix}$$

or

$$\left[ egin{array}{ccc} {\bf P}_{3 imes 3} & {p_4} \end{array} 
ight] pprox \left[ egin{array}{ccc} {\sf KR} & {\sf Kt} \end{array} 
ight]$$

 Making use of **R**R<sup>⊤</sup> = **I** one can easily extract the intrinsic and extrinsic parameters from **P**.

# Camera-Projector Calibration Step-by-Step

- Calibrate the internal parameters of the IR camera (OpenCV).
- Calibrate the external parameters with respect to a single planar calibration pattern (OpenCV).
- Transform the calibration plane in the camera coordinate frame.
- Sing the same position of the calibration pattern to obtain  $m_i \leftrightarrow p_i$  correspondences.
- Move the calibration pattern in several parallel positions.
- Estimate the calibration parameters of the projector in the camera frame.
- Rectify the camera-projector setup to align the camera image with the projector image.

#### Active Stereo Reconstruction

• With a calibrated & rectified camera-projector setup, the relationship between the depth Z and the disparity d is:

$$Z = rac{b}{d}$$
 with  $d = x_p - x_c$ 

where b is the baseline,  $x_p$  and  $x_c$  are the horizontal coordinates of a projector image-point and a camera image-point.

• The coordinates of a point are:

$$\left(\begin{array}{c} X\\Y\\Z\end{array}\right) = Z \left(\begin{array}{c} x_c\\y_c\\1\end{array}\right)$$

## Active Stereo Matching

- The problem is to find  $x_c \leftrightarrow x_p$  correspondences along epipolar lines (along the rows).
- Each pixel of the projector image must be **signed** such that it can be easily **recognized** in the IR-camera image:
  - Temporal signature;
  - Spatial signature, or
  - a combination of both
- Define a  $K \times K \times T$  window around each pixel.
- Arrange the pixels in a vector of size  $N=K\times K\times T$

### **Cross-Correlation**

- Let the entries of two N-dimensional vectors v and w correspond to N realizations of two random variables v and w.
- Expectation:  $E[v] = \frac{1}{N} \sum_{n=1}^{N} p_n v_n$  with  $\sum_{n=1}^{N} p_n = 1$ .

• Variance: 
$$var[v] = E[v^2] - E[v]^2$$

- Standard deviation:  $\sigma_v = \sqrt{var[v]}$
- Standardization:  $\hat{v} = \frac{v E[v]}{\sigma_v}$

• Cross-correlation: 
$$\rho_{vw} = corr[\hat{v}, \hat{w}] = E\left[\frac{v - E[v]}{\sigma_v} \frac{w - E[w]}{\sigma_w}\right]$$

$$\rho_{vw} = \sum_{n=1}^{N} p_n \frac{(v_n - E[v])(w_n - E[w])}{\sigma_v \sigma_w}$$

# Matching Two Pixels

- Vector v contains pixel information associated with the projected light:  $v = (L(n))_{n=1}^{N}$ .
- Vector w contains pixel information associated with the IR image:  $w = (I(n))_{n=1}^N$ .
- The entries of these two vectors are non-null, hence:

• 
$$0 \le \rho_{vw} \le 1.$$

• 
$$\rho_{vw} = 1 \Leftrightarrow \boldsymbol{v} = \boldsymbol{w}$$
.

- $\rho_{vw} = 0$  if v and w are uncorrelated.
- v is associated with IR image pixel  $(x_c, y)$
- ${m w}$  is associated with projector image pixel  $(x_p,y)$
- A match is accepted if  $\rho_{vw} > t$ .

# Matching Two Images

- Based on the cross-correlation coefficient it is possible to implement various global image-matching techniques.
- Because of the epipolar geometry holds and if the images are rectified, the matching can be done row-wise.
- The most straightforward way to match two corresponding row is to use a sequence-alignment algorithm.
- Sequence-alignment algorithms are generally based on dynamic programming (DP).

### Conclusions

- To calibrate a Kinect-like camera, it is necessary to have access to the projector image, but this is not the case with Kinect because the manufacturer (Primesense) does not provide this access.
- The principles behind calibration allow to understand the weaknesses and the current limitations of this type of cameras: limited range, limited to indoor lighting, **extremely sensitive to saturated light**, glossy surfaces, occlusions, transparent and translucent objects, scattering effects (wool, hair, etc.).
- Projected-light depth cameras have good resolution and good depth accuracy, both are better than time-of-flight cameras.