Pose Tracking I



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EE 267 Virtual Reality

Lecture 11

stanford.edu/class/ee267/

Overview

- overview of positional tracking
- camera-based tracking
- HTC's Lighthouse
- VRduino an Arduino for VR, specifically designed for EE 267 by Keenan Molner
- pose tracking with VRduino using homographies

What are we tracking?

- Goal: track pose of headset, controller, ...
- What is a pose?
 - 3D position of the tracked object
 - 3D orientation of the tracked object, e.g. using quaternions or Euler angles
- Why? So we can map the movement of our head to the motion of the camera in a virtual environment – motion parallax!

Overview of Positional Tracking

"inside-out tracking": camera or sensor is located on HMD, no need for other external devices to do tracking

> simultaneous localization and mapping (SLAM) – classic computer & robotic vision problem (beyond this class)

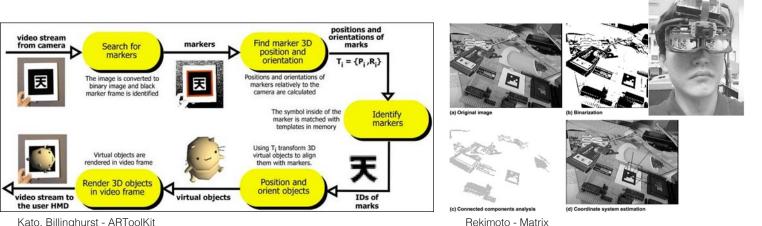
"outside-in tracking": external sensors, cameras, or markers are required (i.e. tracking constrained to specific area)

 used by most VR headsets right now, but everyone is feverishly working on insight-out tracking!

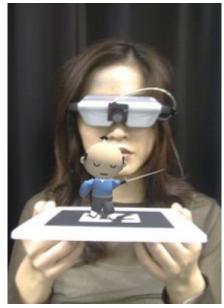
Marker-based Tracking

seminal papers by Rekimoto 1998 and Kato & Billinghurst 1999

widely adopted after introduced by ARToolKit



Marker-based Tracking



ARToolKit

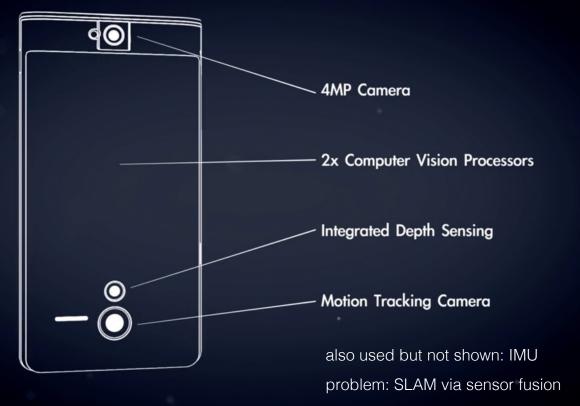


OpenCV marker tracking

Inside-out Tracking



Google's Project Tango



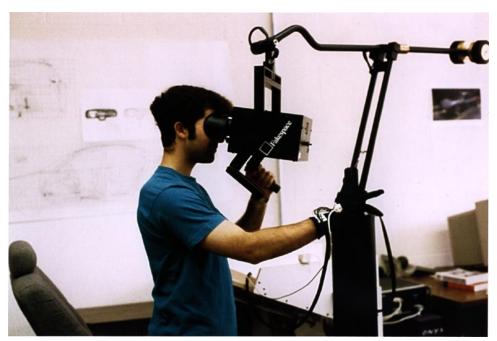
Inside-out Tracking

- marker-less inside-out tracking used by Microsoft HoloLens, Oculus Quest, Magic Leap, ...
- eventually required by all untethered VR/AR systems
- if you need it for your own HMD, consider using Intel's RealSense (small & has SDK)
- if you want to learn more about SLAM, take a 3D computer vision or robotic vision class, e.g. Stanford CS231A

"Outside-in Tracking"

- mechanical tracking
- ultra-sonic tracking
- magnetic tracking
- optical tracking
- GPS
- WIFI positioning
- marker tracking
- ...

Positional Tracking - Mechanical



some mechanical linkage, e.g.

- fakespace BOOM
- microscribe



Positional Tracking - Mechanical

pros:

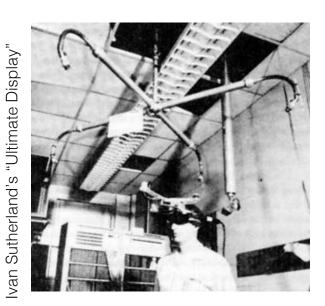
- super low latency
- very accurate

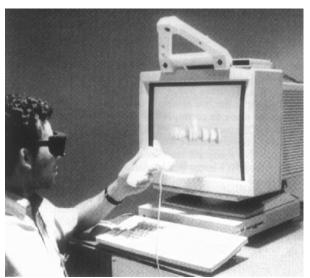
cons:

- cumbersome
- "wired" by design

Positional Tracking – Ultra-sonic

1 transmitter, 3 receivers → triangulation





Logitech 6DOF

Positional Tracking – Ultra-sonic

pros:

· can be light, small, inexpensive

cons:

- line-of-sight constraints
- susceptible to acoustic interference
- low update rates

Positional Tracking - Magnetic

- reasonably good accuracy
- position and orientation
- 3 axis magnetometer in sensors
- need magnetic field generator (AC, DC, ...),
 e.g. Helmholtz coil
- magnetic field has to oscillate and be sync'ed with magnetometers



3 axis Helmholtz coil www.directvacuum.com

Positional Tracking - Magnetic

pros:

- small, low cost, low latency sensors
- no line-of-sight constraints

cons:

- somewhat small working volume
- susceptible to distortions of magnetic field
- not sure how easy it is to do this untethered (need to sync)



3 axis Helmholtz coil

Positional Tracking - Magnetic



Magic Leap One controller tracking:

- magnetic field generator in controller
- magnetometer in headset





https://www.ifixit.com/Teardown/Magic+Leap+One+Teardown/112245

Positional Tracking - Optical

OR

 track passive retro-reflectors with IR illumination around camera

 both Oculus Rift and HTC Vive come with optical tracking



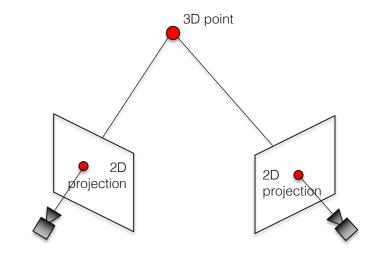
Oculus Rift
https://www.ifixit.com/Teardown/Oculus+Rift+C



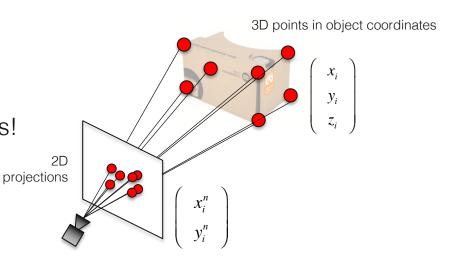
Understanding Pose Estimation - Triangulation

 for tracking individual 3D points, multi-camera setups usually use triangulation

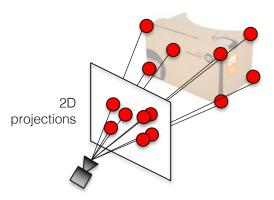
 this does not give us the pose (rotation & translation) of camera or object yet



 for pose estimation, need to track multiple points with known relative 3D coordinates!

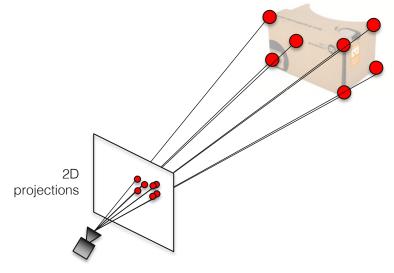


 when object is closer, projection is bigger



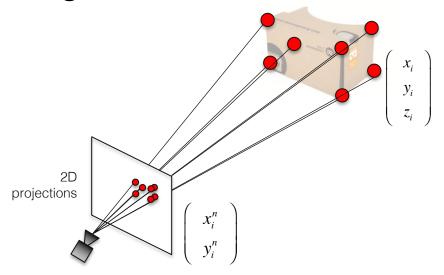
 when object is father, projection is smaller

... and so on ...



Estimating 6-DoF pose from 2D projections is known as the *Perspective-n-point problem*!

- how to get projected
 2D coordinates?
- 2. image formation
- 3. estimate pose with linear homography method
- estimate pose with nonlinear Levenberg-Marquardt method (next class)



- how to get projected
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- HTC Lighthouse
- VRduino

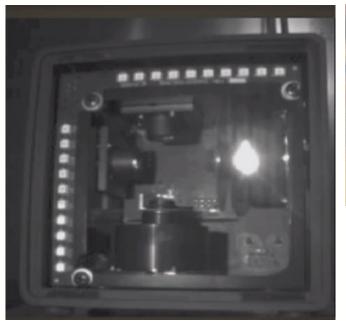
HTC Lighthouse



https://www.youtube.com/watch?v=J54dotTt7k0

HTC Lighthouse







important specs:

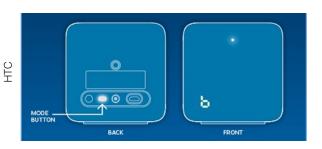
- runs at 60 Hz
 - i.e. horizontal & vertical update combined 60 Hz
 - broadband sync pulses in between each laser sweep (i.e. at 120 Hz)
- each laser rotates at 60 Hz, but offset in time
- useable field of view: 120 degrees

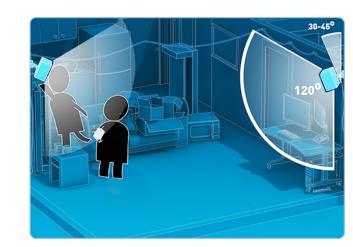
- can use up to 2 base stations simultaneously via time-division multiplexing (TDM)
- base station modes:

A: TDM slave with cable sync

B: TDM master

C: TDM slave with optical sync





- sync pulse periodically emitted (120 times per second)
- each sync pulse indicates beginning of new sweep
- length of pulse also encodes additional 3 bits of information:

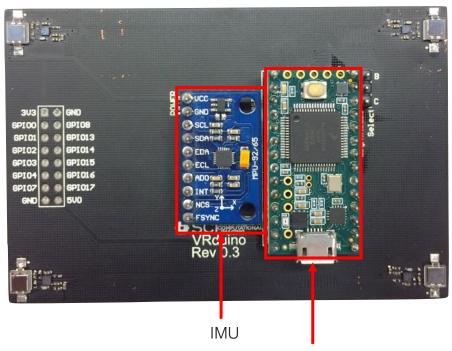
- axis: horizontal or vertical sweep to follow
- <u>skip</u>: if 1, then laser is off for following sweep
- data: data bits of consecutive pulses yield OOTX frame

Name	skip	data	axis	length (ticks)	length (µs)
jO	0	0	0	3000	62.5
k0	0	0	1	3500	72.9
j1	0	1	0	4000	83.3
k1	0	1	1	4500	93.8
j2	1	0	0	5000	104
k2	1	0	1	5500	115
јЗ	1	1	0	6000	125
k3	1	1	1	6500	135

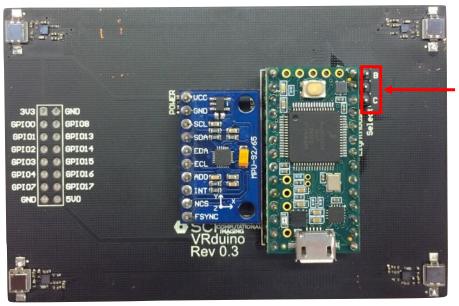
 in this class, we use the HTC Lighthouse base stations but implement positional tracking (i.e., pose estimation) on the VRduino

VRduino is a shield (hardware add-on) for the Arduino Teensy
 3.2; custom-designed for EE 267 by Keenan Molner



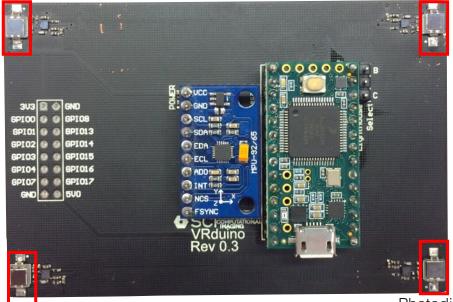


Teensy 3.2



Lighthouse Select

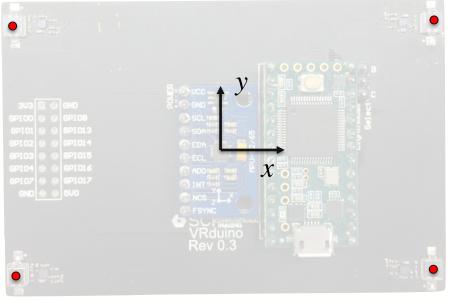
Photodiode 0 Photodiode 1

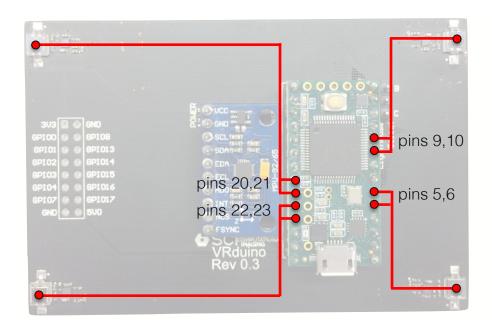


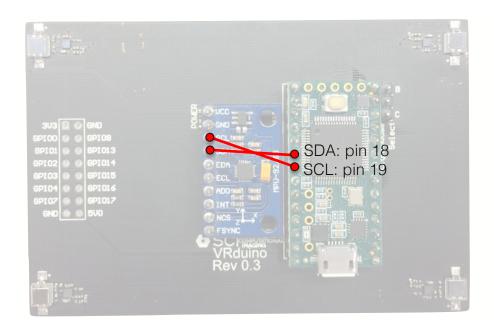
Photodiode 2

x=-42mm, y=25mm

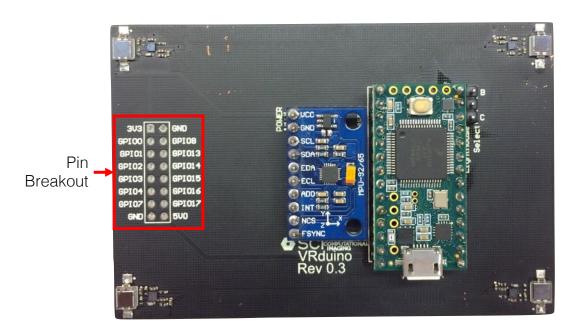
x=42mm, y=25mm











3.3V power, 200mA MAX digital R/W, Serial, cap. sense digital R/W, Serial, cap. sense digital R/W digital R/W, PWM, CAN digital R/W, PWM, CAN



SPI, Serial, digital R/W digital R/W SPI, analog read, digital R/W SPI, analog read, digital R/W I2C, analog read, digital R/W I2C, analog read, digital R/W 5V power, 500mA MAX

For more details, see Lab Writeup

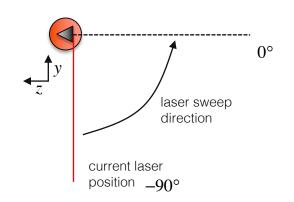
Pose Estimation with the VRduino

 timing of photodiodes reported in Teensy "clock ticks" relative to last sync pulse

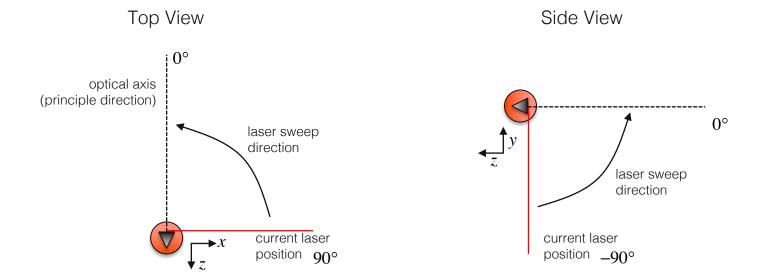
 Teensy usually runs at 48 MHz, so 48,000,000 clock ticks per second

- at time of respective sync pulse, laser is at 90° horizontally and -90° vertically
- each laser rotates 360° in 1/60 sec

Side View

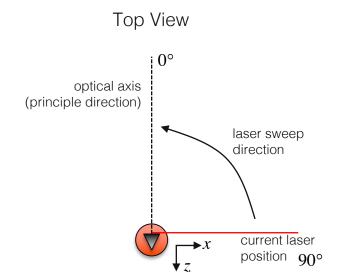


- at time of respective sync pulse, laser is at 90° horizontally and -90° vertically
- each laser rotates 360° in 1/60 sec

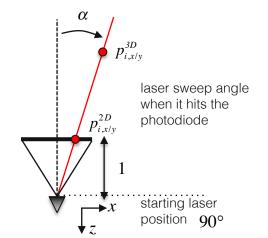


- at time of respective sync pulse, laser is at 90° horizontally and -90° vertically
- each laser rotates 360° in 1/60 sec

 convert from ticks to angle first and then to relative position on plane at unit distance



Top View

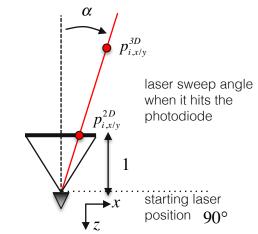


raw number of ticks from photodiode

$$\Delta t [\sec] = \frac{\# ticks}{48,000,000 \left[\frac{ticks}{\sec}\right]} \leftarrow \text{CPU speed}$$

convert from ticks to angle first and then to relative position on plane at unit distance

Top View



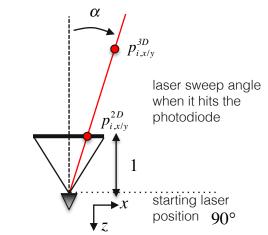
raw number of ticks from photodiode
$$\frac{\downarrow}{\Delta t [\sec]} = \frac{\# ticks}{48,000,000 \left[\frac{ticks}{\sec}\right]} \leftarrow \text{CPU speed}$$

offset from sync pulse

time per 1 revolution

convert from ticks to angle first and then to relative position on plane at unit distance

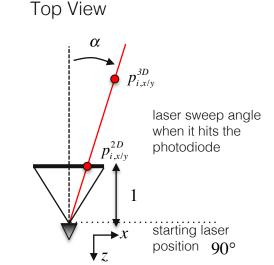
Top View



 convert from ticks to angle first and then to relative position on plane at unit distance

$$p_{i,x/y}^{2D} = \tan\left(\frac{\alpha}{360[^{\circ}]} \cdot 2\pi\right)$$

offset from sync pulse time per 1 revolution



Horizontal Sweep

Vertical Sweep

$$\alpha[\circ] = -\frac{\Delta t[\sec]}{\frac{1}{60} \left[\frac{\sec}{\circ}\right]} + \frac{360}{4} [\circ]$$

$$\alpha[^{\circ}] = \frac{\Delta t[\sec]}{\frac{1}{60} \left[\frac{\sec}{^{\circ}}\right]} - \frac{360}{4} [^{\circ}]$$

Understanding Pose Estimation

- how to get projected
 2D coordinates?
- 2. image formation
- estimate pose with linear homography method
- 4. estimate pose with nonlinear Levenberg-Marquardt method (next class)

- how 3D points project into 2D coordinates in a camera (or a Lighthouse base station)
- very similar to graphics pipeline

Image Formation

 image formation is model for mapping 3D points in local "object" coordinate system to 2D points in "window" coordinates

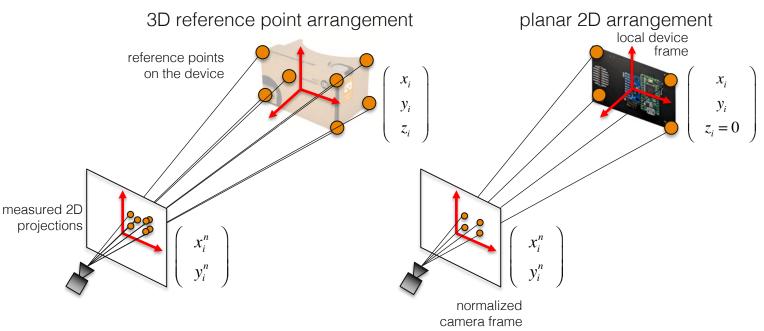


Image Formation – 3D Arrangement

1. transform 3D point into view space:

$$\begin{pmatrix} x_i^c \\ y_i^c \\ \hline w_i^c \end{pmatrix} = \begin{pmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & -1 \end{pmatrix} \cdot \begin{pmatrix} r_{11} & r_{12} & r_{13} & t_x \\ r_{21} & r_{22} & r_{23} & t_y \\ r_{31} & r_{32} & r_{33} & t_z \end{pmatrix} \cdot \begin{pmatrix} x_i \\ y_i \\ z_i \\ 1 \end{pmatrix}$$

"modelview matrix" "projection matrix" 3x3 rotation matrix and translation 3x1 vector

This is the homogeneous

translation 3x1 vector the homogeneous coordinate
$$\frac{\text{erspective divide}}{\text{erspective divide}} : \qquad \begin{pmatrix} x_i^n \\ y_i^n \end{pmatrix} = \begin{pmatrix} \frac{x_i^c}{w_i^c} \\ \frac{y_i^c}{w_i^c} \end{pmatrix}$$

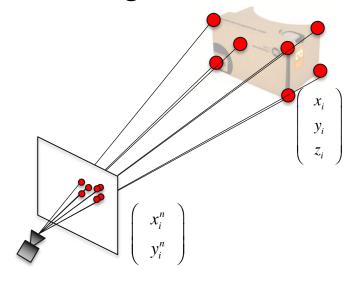


Image Formation – 2D Arrangement

1. transform 3D point into view space:

$$\begin{pmatrix} x_i^c \\ y_i^c \\ w_i^c \end{pmatrix} = \begin{pmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & -1 \end{pmatrix} \cdot \begin{pmatrix} r_{11} & r_{12} & r_{13} & t_x \\ r_{21} & r_{22} & r_{23} & t_y \\ r_{31} & r_{32} & r_{33} & t_z \end{pmatrix} \cdot \begin{pmatrix} x_i \\ y_i \\ 0 \\ 1 \end{pmatrix}$$

$$\begin{bmatrix} y_i^c \\ w_i^c \end{bmatrix} = \begin{bmatrix} 0 & 1 & 0 \\ 0 & 0 & -1 \end{bmatrix} \cdot \begin{bmatrix} r_{21} & r_{22} & r_{23} & t_y \\ r_{31} & r_{32} & r_{33} & t_z \end{bmatrix} \cdot \begin{bmatrix} 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & -1 & 0 \end{bmatrix} \cdot \begin{bmatrix} r_{11} & r_{12} & t_x \\ r_{21} & r_{22} & t_y \\ r_{31} & r_{32} & t & 0 \end{bmatrix} \cdot \begin{bmatrix} x_i & x_i & x_i \\ y_i & y_i & y_i \\ 1 & 1 & 0 \end{bmatrix}$$

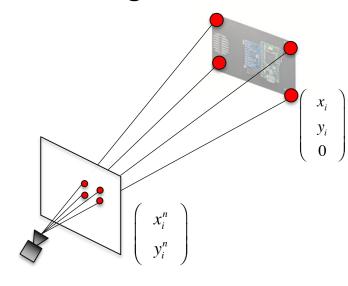


Image Formation – 2D Arrangement

all rotation matrices are orthonormal, i.e. $\frac{\sqrt{r_{11}^2 + r_{21}^2 + r_{31}^2} = 1}{\sqrt{r_{12}^2 + r_{22}^2 + r_{32}^2} = 1}$

$$\sqrt{r_{11}^2 + r_{21}^2 + r_{31}^2} = 1$$
$$\sqrt{r_{12}^2 + r_{22}^2 + r_{32}^2} = 1$$

rotation R translation T
$$\begin{pmatrix}
1 & 0 & 0 \\
0 & 1 & 0 \\
0 & 0 & -1
\end{pmatrix}
\cdot
\begin{pmatrix}
r_{11} & r_{12} \\
r_{21} & r_{22} \\
r_{31} & r_{32}
\end{pmatrix}
\begin{pmatrix}
t_x \\
t_y \\
t_z
\end{pmatrix}$$

The Homography Matrix

all rotation matrices are orthonormal, i.e. $\sqrt{r_{11}^2 + r_{21}^2 + r_{31}^2} = 1$

$$\sqrt{r_{11}^2 + r_{21}^2 + r_{31}^2} = 1$$
$$\sqrt{r_{12}^2 + r_{22}^2 + r_{32}^2} = 1$$

translation T rotation R

$$= \begin{pmatrix} h_1 & h_2 & h_3 \\ h_4 & h_5 & h_6 \\ h_7 & h_8 & h_9 \end{pmatrix}$$

H

Understanding Pose Estimation

- how to get projected
 2D coordinates?
- 2. image formation
- 3. estimate pose with linear homography method
- 4. estimate pose with nonlinear Levenberg-Marquardt method (next class)

- how to compute the homography matrix
- how to get position and rotation from that matrix

The Homography Matrix

Turns out that: any homography matrix has only 8 degrees of freedom – can scale matrix by s and get the same 3D-to-2D mapping

• image formation with scaled homography matrix sH

$$\begin{pmatrix} x_i^n \\ y_i^n \end{pmatrix} = \begin{pmatrix} \frac{x_i^c}{w_i^c} \\ \frac{y_i^c}{w_i^c} \end{pmatrix} = \begin{pmatrix} \frac{sh_1 x_i + sh_2 y_i + sh_3}{sh_7 x_i + sh_8 y_i + sh_9} \\ \frac{sh_4 x_i + sh_5 y_i + sh_6}{sh_7 x_i + sh_8 y_i + sh_9} \end{pmatrix} = \begin{pmatrix} \frac{x(h_1 x_i + h_2 y_i + h_3)}{x(h_7 x_i + h_8 y_i + h_9)} \\ \frac{x(h_2 x_i + h_3 y_i + h_4)}{x(h_3 x_i + h_3 y_i + h_4)} \end{pmatrix}$$

The Homography Matrix

- common approach: estimate a scaled version of the homography matrix, where $h_0 = 1$
- we will see later how we can get scale factor s

image formation changes to

$$\begin{pmatrix} x_i^c \\ y_i^c \\ w_i^c \end{pmatrix} = s \begin{pmatrix} h_1 & h_2 & h_3 \\ h_4 & h_5 & h_6 \\ h_7 & h_8 & 1 \end{pmatrix} \cdot \begin{pmatrix} x_i \\ y_i \\ 1 \end{pmatrix}$$

1

homography matrix with

8 unknowns!

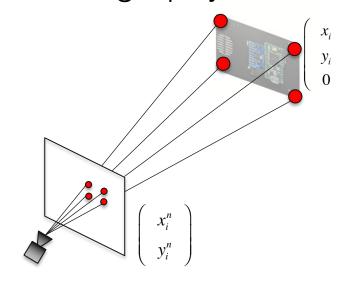
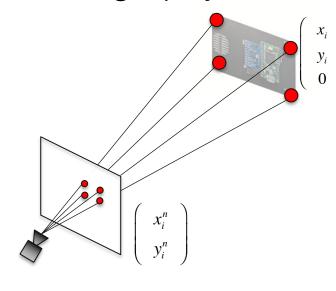


image formation changes to

$$\begin{pmatrix} x_i^c \\ y_i^c \\ w_i^c \end{pmatrix} = s \begin{pmatrix} h_1 & h_2 & h_3 \\ h_4 & h_5 & h_6 \\ h_7 & h_8 & 1 \end{pmatrix} \cdot \begin{pmatrix} x_i \\ y_i \\ 1 \end{pmatrix}$$

$$\begin{pmatrix} x_{i}^{n} \\ y_{i}^{n} \end{pmatrix} = \begin{pmatrix} \frac{x_{i}^{c}}{w_{i}^{c}} \\ \frac{y_{i}^{c}}{w_{i}^{c}} \end{pmatrix} = \begin{pmatrix} \frac{h_{1}x_{i} + h_{2}y_{i} + h_{3}}{h_{7}x_{i} + h_{8}y_{i} + 1} \\ \frac{h_{4}x_{i} + h_{5}y_{i} + h_{6}}{h_{7}x_{i} + h_{8}y_{i} + 1} \end{pmatrix}$$



multiply by denominator

$$\begin{pmatrix} x_{i}^{n} \\ y_{i}^{n} \end{pmatrix} = \begin{pmatrix} \frac{x_{i}^{c}}{w_{i}^{c}} \\ \frac{y_{i}^{c}}{w_{i}^{c}} \end{pmatrix} = \begin{pmatrix} \frac{h_{1}x_{i} + h_{2}y_{i} + h_{3}}{h_{7}x_{i} + h_{8}y_{i} + 1} \\ \frac{h_{4}x_{i} + h_{5}y_{i} + h_{6}}{h_{7}x_{i} + h_{8}y_{i} + 1} \end{pmatrix}$$



$$x_i^n (h_7 x_i + h_8 y_i + 1) = h_1 x_i + h_2 y_i + h_3$$

 $y_i^n (h_7 x_i + h_8 y_i + 1) = h_4 x_i + h_5 y_i + h_6$

reorder equations

$$h_1 x_i + h_2 y_i + h_3 - h_7 x_i x_i^n - h_8 y_i x_i^n = x_i^n$$

$$h_4 x_i + h_5 y_i + h_6 - h_7 x_i y_i^n - h_8 y_i y_i^n = y_i^n$$



$$x_i^n (h_7 x_i + h_8 y_i + 1) = h_1 x_i + h_2 y_i + h_3$$

 $y_i^n (h_7 x_i + h_8 y_i + 1) = h_4 x_i + h_5 y_i + h_6$

• 8 unknowns (red) but only 2 measurements (blue) per 3D-to-2D point correspondence

$$h_{1} x_{i} + h_{2} y_{i} + h_{3} - h_{7} x_{i} x_{i}^{n} - h_{8} y_{i} x_{i}^{n} = x_{i}^{n}$$

$$h_{4} x_{i} + h_{5} y_{i} + h_{6} - h_{7} x_{i} y_{i}^{n} - h_{8} y_{i} y_{i}^{n} = y_{i}^{n}$$

- need at least 4 point correspondences to get to invertible system with 8 equations & 8 unknowns!
- VRduino has 4 photodiodes → need all 4 to compute pose

solve Ah=b on Arduino using Matrix Math Library via
 MatrixInversion function (details in lab)

$$\begin{pmatrix} x_1 & y_1 & 1 & 0 & 0 & 0 & -x_1 x_1^n & -y_1 x_1^n \\ 0 & 0 & 0 & x_1 & y_1 & 1 & -x_1 y_1^n & -y_1 y_1^n \\ x_2 & y_2 & 1 & 0 & 0 & 0 & -x_2 x_2^n & -y_2 x_2^n \\ 0 & 0 & 0 & x_2 & y_2 & 1 & -x_2 y_2^n & -y_2 y_2^n \\ x_3 & y_3 & 1 & 0 & 0 & 0 & -x_3 x_3^n & -y_3 x_3^n \\ 0 & 0 & 0 & x_3 & y_3 & 1 & -x_3 y_3^n & -y_3 y_3^n \\ x_4 & y_4 & 1 & 0 & 0 & 0 & -x_4 x_4^n & -y_4 x_4^n \\ 0 & 0 & 0 & x_4 & y_4 & 1 & -x_4 y_4^n & -y_4 y_4^n \end{pmatrix} \begin{pmatrix} h_1 \\ h_2 \\ h_3 \\ h_4 \\ h_5 \\ h_6 \\ h_7 \\ h_8 \end{pmatrix} = \begin{pmatrix} x_1^n \\ y_1^n \\ x_2^n \\ x_3^n \\ y_3^n \\ x_4^n \\ y_4^n \end{pmatrix}$$

 \boldsymbol{A}

h

b

just computed this

still need scale factor s to get position!

$$\begin{pmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & -1 \end{pmatrix} \cdot \begin{pmatrix} r_{11} & r_{12} & t_x \\ r_{21} & r_{22} & t_y \\ r_{31} & r_{32} & t_z \end{pmatrix} = s \begin{pmatrix} h_1 & h_2 & h_3 \\ h_4 & h_5 & h_6 \\ h_7 & h_8 & 1 \end{pmatrix}$$

• normalize homography to have approx. unit-length columns for the rotation part, such that $\sqrt{r_{11}^2 + r_{21}^2 + r_{31}^2} \approx 1$, $\sqrt{r_{12}^2 + r_{22}^2 + r_{32}^2} \approx 1$

$$s = \frac{2}{\sqrt{h_1^2 + h_2^2 + h_2^2 + \sqrt{h_2^2 + h_2^2 + h_2^2}}}$$

$$\begin{pmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & -1 \end{pmatrix} \cdot \begin{pmatrix} r_{11} & r_{12} & t_x \\ r_{21} & r_{22} & t_y \\ r_{31} & r_{32} & t_z \end{pmatrix} = s \begin{pmatrix} h_1 & h_2 & h_3 \\ h_4 & h_5 & h_6 \\ h_7 & h_8 & 1 \end{pmatrix}$$

· this gives us the position as

$$t_x = sh_3, \quad t_y = sh_6, \quad t_z = -s$$

$$\begin{pmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & -1 \end{pmatrix} \cdot \begin{pmatrix} r_{11} & r_{12} & t_x \\ r_{21} & r_{22} & t_y \\ r_{31} & r_{32} & t_z \end{pmatrix} = s \begin{pmatrix} h_1 & h_2 & h_3 \\ h_4 & h_5 & h_6 \\ h_7 & h_8 & 1 \end{pmatrix}$$

- 1. get normalized 1st column of 3x3 rotation matrix
- 2. get normalized 2nd column via orthogonalization
- 3. get missing 3rd column with cross product

$$\begin{pmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & -1 \end{pmatrix} \cdot \begin{pmatrix} r_{11} & r_{12} & t_x \\ r_{21} & r_{22} & t_y \\ r_{31} & r_{32} & t_z \end{pmatrix} = s \begin{pmatrix} h_1 & h_2 & h_3 \\ h_4 & h_5 & h_6 \\ h_7 & h_8 & 1 \end{pmatrix}$$

1. get normalized 1st column of 3x3 rotation matrix

$$ilde{r_1} = \left(egin{array}{c} h_1 \\ h_4 \\ -h_7 \end{array}
ight), \quad r_1 = rac{ ilde{r_1}}{\left|\left| ilde{r_1}
ight|\right|_2}$$

$$\begin{pmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & -1 \end{pmatrix} \cdot \begin{pmatrix} r_{11} & r_{12} & t_x \\ r_{21} & r_{22} & t_y \\ r_{31} & r_{32} & t_z \end{pmatrix} = s \begin{pmatrix} h_1 & h_2 & h_3 \\ h_4 & h_5 & h_6 \\ h_7 & h_8 & 1 \end{pmatrix}$$

2. get normalized 2nd column via orthogonalization

$$ilde{r}_2 = \left(egin{array}{c} h_2 \\ h_5 \\ -h_8 \end{array}
ight) - \left(egin{array}{c} r_1 ullet \left(egin{array}{c} h_2 \\ h_5 \\ -h_8 \end{array}
ight)
ight) r_1, \quad r_2 = rac{ ilde{r}_2}{\left| \left| ilde{r}_2 \right| \right|_2} \end{array}$$

$$\begin{pmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & -1 \end{pmatrix} \cdot \begin{pmatrix} r_{11} & r_{12} & t_x \\ r_{21} & r_{22} & t_y \\ r_{31} & r_{32} & t_z \end{pmatrix} = s \begin{pmatrix} h_1 & h_2 & h_3 \\ h_4 & h_5 & h_6 \\ h_7 & h_8 & 1 \end{pmatrix}$$

3. get missing 3rd column with cross product: $r_3 = r_1 \times r_2$

• r_3 this is guaranteed to be orthogonal to the other two columns

$$\begin{pmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & -1 \end{pmatrix} \cdot \begin{pmatrix} r_{11} & r_{12} & t_x \\ r_{21} & r_{22} & t_y \\ r_{31} & r_{32} & t_z \end{pmatrix} = s \begin{pmatrix} h_1 & h_2 & h_3 \\ h_4 & h_5 & h_6 \\ h_7 & h_8 & 1 \end{pmatrix}$$

• make 3x3 rotation matrix $R = (r_1 r_2 r_3) = \begin{pmatrix} r_{11} & r_{12} & r_{13} \\ r_{21} & r_{22} & r_{23} \\ r_{31} & r_{32} & r_{33} \end{pmatrix}$

convert to quaternion or Euler angles

remember Euler angles (with yaw-pitch-roll order):

$$\underbrace{\begin{pmatrix} r_{11} & r_{12} & r_{13} \\ r_{21} & r_{22} & r_{23} \\ r_{31} & r_{32} & r_{33} \end{pmatrix}}_{\mathbf{R}} = \underbrace{\begin{pmatrix} \cos\left(\theta_z\right) & -\sin\left(\theta_z\right) & 0 \\ \sin\left(\theta_z\right) & \cos\left(\theta_z\right) & 0 \\ 0 & 0 & 1 \end{pmatrix}}_{\mathbf{R}_z(\theta_z)} \underbrace{\begin{pmatrix} 1 & 0 & 0 \\ 0 & \cos\left(\theta_x\right) & -\sin\left(\theta_x\right) \\ 0 & \sin\left(\theta_x\right) & \cos\left(\theta_x\right) \end{pmatrix}}_{\mathbf{R}_x(\theta_x)} \underbrace{\begin{pmatrix} \cos\left(\theta_y\right) & 0 & \sin\left(\theta_y\right) \\ 0 & 1 & 0 \\ -\sin\left(\theta_y\right) & 0 & \cos\left(\theta_y\right) \end{pmatrix}}_{\mathbf{R}_y(\theta_y)}$$

$$= \begin{pmatrix} \cos\left(\theta_y\right)\cos\left(\theta_z\right) - \sin\left(\theta_x\right)\sin\left(\theta_y\right)\sin\left(\theta_z\right) & -\cos\left(\theta_x\right)\sin\left(\theta_z\right) & \sin\left(\theta_y\right)\cos\left(\theta_z\right) + \sin\left(\theta_x\right)\cos\left(\theta_y\right)\sin\left(\theta_z\right) \\ \cos\left(\theta_y\right)\sin\left(\theta_z\right) + \sin\left(\theta_x\right)\sin\left(\theta_y\right)\cos\left(\theta_z\right) & \cos\left(\theta_x\right)\cos\left(\theta_z\right) & \sin\left(\theta_y\right)\cos\left(\theta_z\right) + \sin\left(\theta_x\right)\cos\left(\theta_y\right)\cos\left(\theta_z\right) \\ -\cos\left(\theta_x\right)\sin\left(\theta_y\right) & \sin\left(\theta_y\right) & \sin\left(\theta_x\right) & \cos\left(\theta_x\right)\cos\left(\theta_y\right) \end{pmatrix}$$

get angles from 3x3 rotation matrix:

$$\begin{aligned} r_{32} &= \sin\left(\theta_x\right) & \Rightarrow \theta_x &= \sin^{-1}\left(r_{32}\right) = a\sin\left(r_{32}\right) \\ \frac{r_{31}}{r_{33}} &= -\frac{\cos\left(\theta_x\right)\sin\left(\theta_y\right)}{\cos\left(\theta_x\right)\cos\left(\theta_y\right)} = -\tan\left(\theta_y\right) & \Rightarrow \theta_y &= \tan^{-1}\left(-\frac{r_{31}}{r_{33}}\right) = a\tan^2\left(-r_{31}, r_{33}\right) \\ \frac{r_{12}}{r_{22}} &= -\frac{\cos\left(\theta_x\right)\sin\left(\theta_z\right)}{\cos\left(\theta_x\right)\cos\left(\theta_z\right)} = -\tan\left(\theta_z\right) & \Rightarrow \theta_z &= \tan^{-1}\left(-\frac{r_{12}}{r_{22}}\right) = a\tan^2\left(-r_{12}, r_{22}\right) \end{aligned}$$

Temporal Filter to Smooth Noise

• pose estimation is <u>very</u> sensitive to noise in the measured 2D coordinates!

→ estimated position and especially rotation may be noisy

• apply a simple temporal filter with weight α to smooth the pose at time step k:

$$\left(\theta_{x}, \theta_{y}, \theta_{z}, t_{x}, t_{y}, t_{z}\right)_{\text{filtered}}^{(k)} = \alpha \left(\theta_{x}, \theta_{y}, \theta_{z}, t_{x}, t_{y}, t_{z}\right)_{\text{filtered}}^{(k-1)} + \left(1 - \alpha\right) \left(\theta_{x}, \theta_{y}, \theta_{z}, t_{x}, t_{y}, t_{z}\right)_{\text{unfiltered}}^{(k)}$$

• smaller $\alpha \rightarrow$ less filtering, larger $\alpha \rightarrow$ more smoothing

Pose Estimation via Homographies – Step-by-Step

in each loop() call of the VRduino:

- 1. get timings from all 4 photodiodes in "ticks"
- 2. convert "ticks" to degrees and then to 2D coordinates on plane at unit distance (i.e. get x_i^n, y_i^n)
- 3. populate matrix A using the 2D and 3D point coordinates
- 4. estimate homography as h=A⁻¹b
- 5. get position t_x , t_y , t_z and rotation, e.g. in Euler angles from the estimated homography
- 6. apply temporal filter to smooth out noise

Must read: course notes on tracking!

Understanding Pose Estimation

- how to get projected
 2D coordinates?
- 2. image formation
- 3. estimate pose with linear homography method
- estimate pose with nonlinear Levenberg-Marquardt method (next class)

- advanced topic
- all details of this are also derived in course notes