

Computer Vision Group Prof. Daniel Cremers



Practical Course: Vision-based Navigation SS 2018

Lecture 1. Basics

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Contents

- Course contents and preliminary knowledges
- Framework and mathematic form of a SLAM problem
- 3D geometry

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General overview of computer vision tasks

Computer vision



Object detection Object recognition Object tracking Segmentation

SLAM

Real world cameras

CV tasks

Image and video sequences

What is SLAM?



Indoor/outdoor localization

Computer vision



Dense/semi-dense reconstruction

What is SLAM?

ElasticFusion: Dense SLAM Without A Pose Graph

Thomas Whelan, Stefan Leutenegger, Renato Salas-Moreno, Ben Glocker, Andrew Davison

Imperial College London

RGB-D dense reconstruction

SLAM applications



Hand-held devices

Autonomous driving

Augmented reality/VR

Computer vision



Harley and Zisserman, Multiple view geometry in computer vision

Tim Barfoot, State estimation for robotics

- Course Contents
- Lecture 1. Basic knowledge, 3D motion
- Lecture 2. Lie group/Lie algebra, Camera models
- Lecture 3. State estimation, Nonlinear optimization
- Lecture 4. Visual Odometry
- Lecture 5. Backend Optimization
- Our course takes place at Monday a.m.
- Programming Assignments and Final Project are required

- Preliminary knowledge
- Math: Calculus, Linear algebra, Probability theory
- Programing: C++/Linux
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- SLAM problem
 - Fundamental problems in intelligent robots
 - Where am I?
 -Localization
 - What is around me?
 Mapping
- Chicken and egg problem
 - Localization needs accurate map
 - Mapping needs accurate localization



- How to do SLAM? -Sensors
- Sensor is the way to measure the outside environment
- Interoseptive sensors: accelerometer, gyroscope ...
- Exteroceptive sensors: camera, laser rangefinder, GPS ...



(a)



(b)



(c)



(f)

Some sensors must be placed in a cooperative environment, other can be directly equipped in the robot itself

- Visual SLAM -SLAM (mainly) by cameras
- Cameras
 - Monocular
 - Stereo
 - RGB-D
 - Omnidirectional, Event camera, etc
- Cameras
 - Cheap, rich information
 - Record 2D projected image of the 3D world
 - The 3D-2D projection throws away one dimension: distance







RGB-D (depth) camera



Stereo camera



- Various kinds of cameras:
- Monocular: image only, need other methods to estimate the depth
- Stereo: disparity to depth
- RGB-D: physical depth measurements









Stereo vision estimates the depth from disparity



Moving stereo: disparity can be estimated in the motion

Ambiguity in mono vision: small + close or large + far away?

SLAM framework



- Visual odometry
 - Motion estimation betwee adjacent frames
 - Simplest: two-view geometry
- Method
 - Feature method
 - Direct method
- Backend
 - Long-term trajectory and map estimation
 - MAP: Maximum of a Posteri
 - Filters/Graph Optimization





- Loop closing
 - Correct the drift in estimation
 - Loop detection and correction



- Mapping
 - Generate globally consisten map for navigation/planning/comm nication/visualization etc
 - Grid/topological/hybrid maps
 - Pointcloud/Mesh/TSDF ...



2D grid map



2D topological map



Point cloud maps



TSDF models

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- Mathematical representation of visual SLAM
- Assume a camera is moving in 3D space
 - But measurements are taken at discrete times:

$$\begin{array}{l} \boldsymbol{x}_{k} = f\left(\boldsymbol{x}_{k-1}, \boldsymbol{u}_{k}, \boldsymbol{w}_{k}\right) \\ \boldsymbol{z}_{k,j} = h\left(\boldsymbol{y}_{j}, \boldsymbol{x}_{k}, \boldsymbol{v}_{k,j}\right) \end{array} \text{ Non-linear form} \begin{array}{l} \text{Motion model} \\ \text{Non-linear form} \end{array} \begin{array}{l} \text{Motion model} \\ \text{Motion model} \end{array} \begin{cases} \boldsymbol{x}_{k} = A_{k} \boldsymbol{x}_{k-1} + B_{k} \boldsymbol{u}_{k} + \boldsymbol{w}_{k} \\ \boldsymbol{z}_{k,j} = C_{j} \boldsymbol{y}_{j} + D_{k} \boldsymbol{x}_{k} + \boldsymbol{v}_{k,j} \end{cases}$$

- 2. Framework of SLAM
 - Questions:

$$\begin{cases} \boldsymbol{x}_{k} = f\left(\boldsymbol{x}_{k-1}, \boldsymbol{u}_{k}, \boldsymbol{w}_{k}\right) \\ \boldsymbol{z}_{k,j} = h\left(\boldsymbol{y}_{j}, \boldsymbol{x}_{k}, \boldsymbol{v}_{k,j}\right) \end{cases}$$

Motion model

Observation model

- How to represent state variables?
 - 3D geometry, Lie group and Lie algebra
- Exact form of motion/observation model?
 - Camera intrinsic and extrinsics
- How to estimate the state given measurement data?
 - State estimation problem
 - Filters and optimization

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- Point and Coordinate system
- 2D: (x,y) and angle
- 3D?

- 3D coordinate system
- Vectors and their coordinates



- Vector operations
 - Addition/subtraction
 - Dot product

$$\boldsymbol{a} \cdot \boldsymbol{b} = \boldsymbol{a}^T \boldsymbol{b} = \sum_{i=1}^3 a_i b_i = |\boldsymbol{a}| |\boldsymbol{b}| \cos \langle \boldsymbol{a}, \boldsymbol{b} \rangle.$$

Cross product

$$m{a} imes m{b} = \left[egin{array}{cccc} m{i} & m{j} & m{k} \\ a_1 & a_2 & a_3 \\ b_1 & b_2 & b_3 \end{array}
ight] = \left[egin{array}{ccccc} a_2b_3 - a_3b_2 \\ a_3b_1 - a_1b_3 \\ a_1b_2 - a_2b_1 \end{array}
ight] = \left[egin{array}{cccccc} 0 & -a_3 & a_2 \\ a_3 & 0 & -a_1 \\ -a_2 & a_1 & 0 \end{array}
ight] m{b} \stackrel{\Delta}{=} m{a}^{\wedge} m{b}.$$

Skew-symmetric operator

- Questions
 - Compute the coordinates in different systems?

- In SLAM:
 - Fixed world frame
 - Moving camera frame
 - Other sensor frames



• 3D rigid body motion can be described with rotation and translation



- Translation is just a vector addition
- How to represent rotations?

- Rotation
 - Consider coordinate system (e_1, e_2, e_3) is rotated and become $(e_1^{'}, e_2^{'}, e_3^{'})$
 - Vector *a* is fixed, then how are its coordinates changed?

$$\begin{bmatrix} \boldsymbol{e}_1, \boldsymbol{e}_2, \boldsymbol{e}_3 \end{bmatrix} \begin{bmatrix} a_1 \\ a_2 \\ a_3 \end{bmatrix} = \begin{bmatrix} \boldsymbol{e}_1', \boldsymbol{e}_2', \boldsymbol{e}_3' \end{bmatrix} \begin{bmatrix} a_1' \\ a_2' \\ a_3' \end{bmatrix}$$

• Left multiplied by $\left[e_1^T, e_2^T, e_3^T\right]^T$

$$\begin{bmatrix} a_1 \\ a_2 \\ a_3 \end{bmatrix} = \begin{bmatrix} e_1^T e_1' & e_1^T e_2' & e_1^T e_3' \\ e_2^T e_1' & e_2^T e_2' & e_2^T e_3' \\ e_3^T e_1' & e_3^T e_2' & e_3^T e_3' \end{bmatrix} \begin{bmatrix} a_1' \\ a_2' \\ a_3' \end{bmatrix} \stackrel{\Delta}{=} \mathbf{R} \mathbf{a}'.$$
 Rotation

matrix

$$\begin{bmatrix} a_1 \\ a_2 \\ a_3 \end{bmatrix} = \begin{bmatrix} e_1^T e_1' & e_1^T e_2' & e_1^T e_3' \\ e_2^T e_1' & e_2^T e_2' & e_2^T e_3' \\ e_3^T e_1' & e_3^T e_2' & e_3^T e_3' \end{bmatrix} \begin{bmatrix} a_1' \\ a_2' \\ a_3' \end{bmatrix} \stackrel{\Delta}{=} \mathbf{R} \mathbf{a}'.$$

- R is rotation matrix, which satisfies:
 - R is orthogonal
 - Det(R) = +1 (if Det(R)=-1 then it's improper rotation)
- Special orthogonal group:

$$SO(n) = \{ \boldsymbol{R} \in \mathbb{R}^{n \times n} | \boldsymbol{R} \boldsymbol{R}^T = \boldsymbol{I}, \det(\boldsymbol{R}) = 1 \}.$$

Rotation from frame 2 to 1 can be written as:

$$a_1 = R_{12}a_2$$
 and vise vesa: $a_2 = R_{21}a_1$
 $R_{21} = R_{12}^{-1} = R_{12}^T$

Rotation plus translation:

$$a' = Ra + t.$$



Compounding rotation and translation:

$$\bullet \quad b = \mathbf{R}_1 \mathbf{a} + \mathbf{t}_1, \quad \mathbf{c} = \mathbf{R}_2 \mathbf{b} + \mathbf{t}_2. \quad \blacksquare \quad \bullet \quad \mathbf{c} = \mathbf{R}_2 \left(\mathbf{R}_1 \mathbf{a} + \mathbf{t}_1 \right) + \mathbf{t}_2.$$

Homogeneous form:

$$\begin{bmatrix} a'\\1 \end{bmatrix} = \begin{bmatrix} R & t\\0^T & 1 \end{bmatrix} \begin{bmatrix} a\\1 \end{bmatrix} \stackrel{\Delta}{=} T \begin{bmatrix} a\\1 \end{bmatrix} \cdot \qquad \tilde{b} = T_1 \tilde{a}, \ \tilde{c} = T_2 \tilde{b} \quad \Rightarrow \tilde{c} = T_2 T_1 \tilde{a}.$$
Inverse:
$$T^{-1} = \begin{bmatrix} R^T & -R^T t\\0^T & 1 \end{bmatrix}.$$

Homogenous coordinates:

$$\tilde{a} = \begin{bmatrix} a \\ 1 \end{bmatrix} \qquad \qquad \tilde{a} = \begin{bmatrix} a \\ 1 \end{bmatrix} = k \begin{bmatrix} a \\ 1 \end{bmatrix}$$

Still keeps equal when multiplying any non-zero factors

Transform matrix forms Special Euclidean Group

$$SE(3) = \left\{ \boldsymbol{T} = \left[\begin{array}{cc} \boldsymbol{R} & \boldsymbol{t} \\ \boldsymbol{0}^T & \boldsymbol{1} \end{array} \right] \in \mathbb{R}^{4 \times 4} | \boldsymbol{R} \in SO(3), \boldsymbol{t} \in \mathbb{R}^3 \right\}$$

- Alternative rotation representations
 - Rotation vectors
 - Euler angles
 - Quaternions
- Rotation vectors
 - Angle + axis: θn
 - Rotation angle θ
 - Rotation axis n
- Rotation vector to rotation matrix: Rodrigues' formula

$$\boldsymbol{R} = \cos \theta \boldsymbol{I} + (1 - \cos \theta) \boldsymbol{n} \boldsymbol{n}^{T} + \sin \theta \boldsymbol{n}^{\wedge}$$

Rn = n.

Inverse: $\theta = \arccos(\frac{\operatorname{tr}(\boldsymbol{R}) - 1}{2}).$



Rotation vectors Only three parameters

- Euler angles
 - Any rotation can be decomposed into three principal rotations



- However the order of axis can be defined very differently:
- Roll-pitch-yaw (in navigation)
 Spin-nutation-precession in mechanics



XYZ order

3-1-3 order

- Gimbal lock
 - Singularity always exist if we want to use 3 parameters to describe rotation
 - Degree-of-Freedom is reduced in singular case
 - In yaw-pitch-roll order, when pitch=90 degrees

normal





singular

- Quaternions
 - In 2D case, we can use (unit) complex numbers to denote rotations

$$z = x + iy = \Gamma e^{iq}$$

- How about 3D case?
- (Unit) Quaternions
 - Extended from complex numbers
 - Three imaginary and one real part:

$$\boldsymbol{q} = q_0 + q_1 \boldsymbol{i} + q_2 \boldsymbol{j} + q_3 \boldsymbol{k},$$

Multiply i to rotate 90 degrees

The imaginary parts satisfy:

$$\begin{cases} i^{2} = j^{2} = k^{2} = -1 \\ ij = k, ji = -k \\ jk = i, kj = -i \\ ki = j, ik = -j \end{cases}$$

i,j,k look like complex numbers when multiplying with themselves And look like cross product when multiply with others

Ouaternions

$$\boldsymbol{q} = q_0 + q_1 i + q_2 j + q_3 k, \qquad \boldsymbol{q} = \lfloor$$

$$\boldsymbol{q} = [s, \boldsymbol{v}], \quad s = q_0 \in \mathbb{R}, \boldsymbol{v} = [q_1, q_2, q_3]^T \in \mathbb{R}^3,$$

Operations

$$oldsymbol{q}_a\pmoldsymbol{q}_b=[s_a\pm s_b,oldsymbol{v}_a\pmoldsymbol{v}_b]\,. \qquad \qquad oldsymbol{q}_a^*=s_a-x_ai-y_aj-z_ak=[s_a,-oldsymbol{v}_a].$$

$$\begin{aligned} q_a q_b &= s_a s_b - x_a x_b - y_a y_b - z_a z_b \\ &+ (s_a x_b + x_a s_b + y_a z_b - z_a y_b) i \\ &+ (s_a y_b - x_a z_b + y_a s_b + z_a x_b) j \\ &+ (s_a z_b + x_a y_b - y_b x_a + z_a s_b) k. \end{aligned}$$

$$oldsymbol{q}_aoldsymbol{q}_b = \left[s_as_b - oldsymbol{v}_a^Toldsymbol{v}_b, s_aoldsymbol{v}_b + s_boldsymbol{v}_a + oldsymbol{v}_a imesoldsymbol{v}_b
ight].$$

$$\boldsymbol{q}_a^* = s_a - x_a i - y_a j - z_a k = [s_a, -\boldsymbol{v}_a].$$

$$\|\boldsymbol{q}_a\| = \sqrt{s_a^2 + x_a^2 + y_a^2 + z_a^2}.$$

$$q^{-1} = q^* / ||q||^2.$$

$$k\boldsymbol{q} = [ks, k\boldsymbol{v}].$$

$$\boldsymbol{q}_a \cdot \boldsymbol{q}_b = s_a s_b + x_a x_b i + y_a y_b j + z_a z_b k.$$

From quaternions to angle-axis:

$$\boldsymbol{q} = \left[\cos\frac{\theta}{2}, n_x \sin\frac{\theta}{2}, n_y \sin\frac{\theta}{2}, n_z \sin\frac{\theta}{2}\right]^T$$

$$\begin{cases} \theta = 2 \arccos q_0 \\ \left[n_x, n_y, n_z \right]^T = \left[q_1, q_2, q_3 \right]^T / \sin \frac{\theta}{2} \end{cases}$$

- Rotate a vector by quaternions:
 - Vector p is rotated by q and become p', how to write their relationships?
 - Write p as quaternion (pure imaginary): p = [0, x, y, z] = [0, v].
 - Then: $m{p}' = m{q} m{p} m{q}^{-1}$. Also pure imaginary

Notes on programing assignments

- Use cmake to manage your C++ project in Linux
- Use Eigen to handle matrix and geometry computations