M19CSE483 Mid-sem exam

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TOTAL POINTS

25.5 / 30

QUESTION 1

1Q1 - Warm up 4/5

- √ + 1 pts Camera calibration
- √ + 1 pts Triangulation
- √ + 1 pts Stereo rectification
- √ + 1 pts PnP
 - + 1 pts Bundle adjustment
 - + O pts Wrong

QUESTION 2

Q2 - Pose transformations 5 pts

2.1 Q2.1 2 / 2

- + 0 pts Incorrect
- √ + 1 pts Rotation matrix
- √ + 1 pts Translation vector
 - 0.5 pts Proper explanation not given for R
 - **0.5 pts** Proper explanation not given for

translation vector

2.2 Q2.2.a 1/1

- + 0 pts Incorrect
- √ + 1 pts Correct
 - 0.5 pts No explanation
 - + 0.5 pts correct explanation

2.3 Q2.2.b 1/1

- + 0 pts Incorrect
- √ + 1 pts Correct
 - 0.5 pts Incorrect order of rotations

2.4 Q2.2.C 1/1

- + 0 pts Incorrect
- √ + 0.5 pts Rotation matrix
- √ + 0.5 pts Translation vector

QUESTION 3

Q3 5 pts

3.1 Q3.1 - Single-view geometry 2/2

- + 0 pts Incorrect
- √ + 1 pts Camera center
- √ + 1 pts Rotation matrix

3.2 Q3.2 - Reconstruction ambiguities 1/3

- + 0 pts Incorrect
- √ + 0.5 pts Requirements for (a)
 - + 0.5 pts Requirements for (b)
- √ + 0.5 pts Requirements for (c)
 - + 0.5 pts Justification for (a)
 - + 0.5 pts Justification for (b)
- + 0.5 pts Justification for (c)

QUESTION 4

4 Q4 - Special Essential matrix 5 / 5

- + 1 pts Equation
- + 1 pts Partial Image Points
- + 1 pts Partial Proof
- + 2 pts Complete Image Points
- + 2 pts Complete proof
- + 0 pts Not correct / Unattempted
- + 0.5 pts Additional Info
- √ + 5 pts Correct Derivation

QUESTION 5

Q5 - Homography from pure rotations 5

pts

5.1 Q5.a 2.5 / 2.5

- √ + 2.5 pts Complete
 - + 1 pts Right approach
 - + O pts Incorrect / Not attempted

5.2 Q5.b **2.5** / **2.5**

- √ + 2.5 pts Complete
 - + 1 pts Right approach
 - + 0 pts Incorrect / Not attempted

QUESTION 6

Q6 - Dense Visual Odometry 5 pts

6.1 Q6.a 3/3

- √ + 1.5 pts Found 3D point in C1 frame
- √ + 1.5 pts Re-project onto I2
 - + 0 pts Incorrect / Not attempted
 - Last line doesn't make sense though

6.2 Q6.b **0.5/2**

- + 1 pts Type
- + 1 pts How to solve
- √ + 0.5 pts Only name of solver is mentioned
 - + O pts Incorrect / Not attempted

CSE483-Mobile Robotics

Mid-semester exam Monsoon 2019 September 21st

Maximum points: 30

Duration: 90 minutes

Instructions

- This is an **open-book** exam. You are allowed to use any paper notes or textbooks that you have brought with you.
- Laptops, tablets, or smartphones are NOT allowed. You also cannot collaborate with other students.
- Your answers must be concise and to-the-point. Verbosity will NOT fetch you additional marks.
- Sufficient space has been provided for each question. Using additional sheets are discouraged, if you need them you're probably doing something wrong.
- You do NOT get credit for replicating whatever is present in the textbook or your notes. Please do not fill your answer scripts with excerpts from such sources.
- Use the last page for rough work or for any of your answers, if necessary.
- State your assumptions clearly if there is any ambiguity with the question(s).

Roll number: 2019 7010 13

Seat: (6)

Invigilator sign: Pulgly

Q1	Q2	Q3	Q4	Q5	Q6	Total

Q1) Warm-up: Fill up the following table by indicating the quantities that are known, to be estimated, or unknown, and the type of measurements that are needed. (5 points)

Problem	Structure (Scene geometry)	Motion (Camera parameters)	Measurements
F-matrix estimation	Unknown	Estimate	2D - 2D features
Camera calibration	Known	Estinate	3D-2D corresponden
Triangulation	Estimate	Known	SD-JD carestrapy
Stereo rectification	Unknown	new ones	R, t1, K, K2
PnP	Known	Pose testimate	30-20 (Oso spondue
Bundle adjustment	Known	Esmete P	y ==

Q2) Transformations:

Derive the expression for
$$T_W^C$$
 if $T_C^W = \begin{bmatrix} R_C^W & P_{CORG}^W \\ 0_{3\times 1} & 1 \end{bmatrix}$. (2 points)

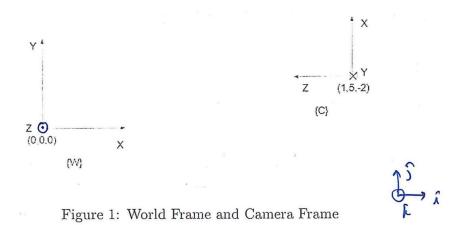
$$T_{M}^{c} = \left(\begin{array}{cc} k_{M}^{c} & k_{M}^{c} \\ k_{M}^{c} & k_{M}^{c} \end{array}\right)$$

Now, Pwing is simply - re of Pcor, after rotating it to comera's from world from.

i.c.
$$l_{w}^{c} = -\left(l_{w}^{c} l_{con}^{u} \right)$$

Thurfor,
$$T_{W} = \begin{bmatrix} R_{c} & -R_{W} & P_{corr}^{W} \\ O_{JN} & & \end{bmatrix}$$

(ii) Consider the following figure and answer questions (a) to (c).



{W} represents the world frame and {C} represents the camera frame. The Z axis of {W} is coming out of the plane. Whereas the Y axis of {C} is going into the plane.

Find
$$R_C^W$$
. (2 points)
$$R_C^W = \begin{pmatrix}
\hat{x}_C & \hat{x}_W & \hat{y}_C & \hat{y}_W & \hat{y}_C & \hat{x}_W \\
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(b) Find the YXZ-Euler angles representation for R_C^W . (2 points)

Notate d Dy First about
$$\hat{Y}_B$$
 by \hat{P}_A , then \hat{X}_B by \hat{P}_A hally

 \hat{Z}_B by \hat{X}_A .

 $\hat{P} = -90^\circ$, $\hat{Y} = 0$, $\hat{X}_A = -90^\circ$

A \hat{P}_A = \hat{P}_A (\hat{P}_A) $\hat{P$

(c) Find P_{WORG}^C and T_W^C . (1 points)

Q3.1) Single-view geometry: Given a camera matrix P, detail how you can obtain the camera center and the rotation matrix R without knowing the intrinsic parameter matrix K. (2 points).

We need to decompose
$$P$$
 into X_0 A A A without knowing K .

$$\hat{P} = \hat{K} \hat{R} \left(T_3 \right) - \hat{X}_0 \right) = \left[\hat{H}_0 \right] \hat{k}$$

$$\hat{H}_0 : \hat{K} \hat{R} \right) \hat{h} = -\hat{K} \hat{K} \hat{X}_0$$

Now, X_0 is simple: $\left[\hat{X}_0 = -\hat{H}_0^{-1} \hat{k} \right] \rightarrow formalis.$

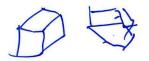
(original all decomposition of $\hat{H}_0 = (\hat{K} \hat{R})^{-1} = \hat{R}^{-1} \hat{K}^{-1} = \hat{R}^{-1} \hat{K}^{-1}$

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Q3.2) Reconstruction: State and justify the cases when the 3D reconstruction obtained from two views is (a) Unambiguous (b) Up to an unknown scaling factor (c) Up to an unknown projective transformation. (3 points)

(a) When K, R, t all are known, it is unambijuous

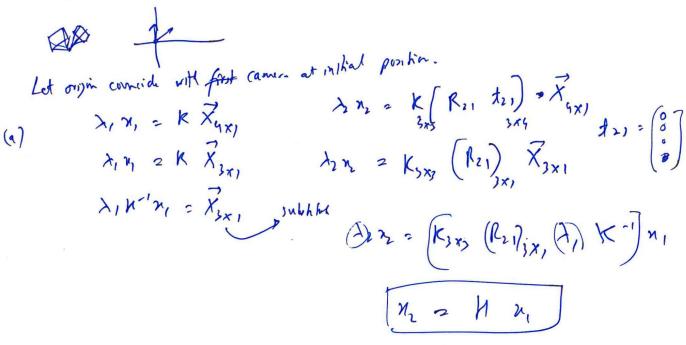
(c) If K is Musum best R,t are unknown it is upto an unknown projection transformation.



v e

Q4) Essential matrix: Two cameras fixate on a point P in 3D space such that their optical axes intersect at this point. Show that the E₃₃ element of their associated Essential matrix E is zero. (5 points) 4 or in an optical axis, nometized considered pay would be his of 4 his (of $\begin{pmatrix} 0 & 0 & f \end{pmatrix} \begin{pmatrix} \varepsilon_{11} & \varepsilon_{12} & \varepsilon_{13} \\ \varepsilon_{21} & \varepsilon_{12} & \varepsilon_{23} \\ \varepsilon_{31} & \varepsilon_{32} & \varepsilon_{13} \end{pmatrix} \begin{pmatrix} 0 \\ 0 \\ f \end{pmatrix} = 0$ (0 0 f) (f F13 f F33) 2 0

Homography: Suppose a camera, with intrinsic matrix K, rotates about its optical centre by a rotation matrix R. (a) Show that its two views are related by a homography H such that $x_2 = Hx_1$. (2.5 points) (b) Also show that if θ is the rotation between the two views then the angle 2θ corresponds to the homography H^2 . (2.5 points)



(b)
$$N^2 = K_{3\times3} R_{21} \times K R_{21} \times K$$

Rotatory by 0 once, again by 0 \Rightarrow 0 + 0 = 20.

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Q6) Dense-VO: Dense-VO is one other type of visual odometry where the camera motion is estimated by aligning consecutive image frames and then finding the transformation that best minimizes the photometric error between them. Suppose there is a camera C with known intrinsics K, and it captures two images I_1, I_2 from two views separated by a rotation R and translation t. The photometric error between these two views is given as $\sum_{\mathbf{x} \in I_1} \|I_1(\mathbf{x}) - I_2(w(\mathbf{x}, (\mathbf{R}|\mathbf{t})))\|^2$ where $w(\mathbf{x}, (\mathbf{R}|\mathbf{t}))$ is a function that maps a point \mathbf{x} in the first image to a point in the second image given the camera motion R, t. (a) Assuming d is the depth of the point \mathbf{x} in I_1 , describe the steps involved to map this point to the second image, and hence provide a mathematical expression for $w(\mathbf{x}, (\mathbf{R}|\mathbf{t}))$. (3 points) (b) What is the nature of this photometric error? Very briefly in words mention how it can be solved for to find the best camera motion. (2 points)

(a) $\lambda_{1} = k \vec{x}$ $\lambda_{2} \vec{R}_{1} = k (R_{21} t_{11}) \vec{x}$ $\lambda_{2} \vec{R}_{1} \cdot k (R_{21} t_{11}) \vec{x}$ $\vec{x} = d k^{-1} \vec{P}_{1} \text{ or } \vec{X} \cdot d k^{-1} \vec{x}$ (b) $\vec{x} = k (R_{11} t_{11}) \vec{R}_{1} \cdot k (R_{11} t_{11}) \vec{R}_{1} \cdot k (R_{21} t_{11}) \vec{R}_{2} \cdot k (R_{$

(b) Least squares minimi) along can be solved using hauss-Newton.

Extra space

Extra space

No. of the Section of