

VMMC Lab Evaluation

To carry out this exam, which takes place during three sessions, develop your code in separate 'Surname_Name_X.m' script files, where 'X' stands for the section number. Once you have finished, compress the generated '.m' files in a single file 'Surname_Name.zip', connect to the Moodle delivery system and submit it to the link enabled for it.

To answer the sections/subsections of this exam (both textual or numerical answers and sample images), do it in a file 'Surname_Name.pdf', clearly indicating the section you are responding to and what image is represented (do not use subplot). This is the file that will be evaluated. The script files will only be evaluated in case of doubt, and only those that do not throw errors when executed individually (make sure of this fact putting `clear all` at the beginning of all and including all the required functions in the zip file).

ASSIGNMENT OVERVIEW

The objective of this evaluation assignment is to check the degree of assimilation of the practical procedures presented during the laboratory sessions. For this purpose, we propose to resolve step by step a practical problem of similar difficulty to that developed during these sessions.

The final aim of the proposed problem is to obtain a 3D reconstruction of a 3D object/scene of your choice. Which requires to select and calibrate a camera, to select an adequate object/scene according to some indications, to select an adequate number of views from the object, to extract and match feature points between views, to compute the fundamental matrix between views, to obtain a 3D points cloud reconstruction, and finally to represent object geometric elements over this points cloud.

Section 1: Obtention of the intrinsic parameters of a camera

The objective of this section is to calibrate a camera of your choice. As this will be the camera used throughout the rest of the evaluation, please be sure that your results are, at least, coherent.

First, select the camera you are going to work with. Remember that its internal parameters should be fixed, i.e., no changes in the focal distance nor changes in the aspect ratio should be performed after calibration.

In order to use a flexible method to calibrate a camera, we will replace the use of a physical checkerboard with the use of your laptop or desktop screen. We provide two square checkerboard images, a large 1080 pixels size and a smaller 720 pixels one, which you should open to simulate a physical checkerboard. Use the large one for the first exercise (enlarge it to cover all the screen) and the small one for the second exercise (in this case, do not resize it).

1. Use the code developed during the assignments of VMMC-Part I to calibrate your camera, using the large checkerboard image provided. Include in your exam report the following data:



- Size, in millimeters, of the checkerboard in your screen.
- The set of images of the screen checkerboard that you have used for calibration.
- The resolution of these captured images (in pixels).
- The obtained matrix of internal parameters, \mathbf{A} .

Additionally, in the light of the computed \mathbf{A} matrix, include a detailed and reasoned analysis of

the following three aspects:

- Are the pixels of your camera square?
 - Which is the degree of coincidence between the principal point and the center of the image plane?
 - Are the axes of the image plane orthogonal?
2. Finally, repeat the previous exercise for the small checkerboard image, including in your report the corresponding captured images, the obtained A' matrix, and the analysis of the obtained parameters, and comment on the theoretical and practical relationship between A and A' :

Note 1: For Sections 2 and 3 (point matches and reconstruction), you may use the same scene and images. Figure 1 shows several viewpoints of an example scene.



Figure 1 Different viewpoints of a scene for point-matching and reconstruction

Note 2: For computational reasons, you may need to reduce the size of your images. If you do so, you must modify the camera parameters obtained in Section 1 accordingly. For example: if you have an 800x600 camera and you reduce your images to a 400x300 resolution (1/2-scaling), the coefficients in your intrinsic matrix must be scaled accordingly (except the scale coefficient of the matrix: $K_{3,3}$).

Section 2: Finding local matches between several views of an object.

In this section, you are asked to use the camera that you have calibrated in Section 1 to capture several views of an object or a scene. Then, for pairs of views, you are asked to detect, describe and match feature points using several of the methods explained in class. Finally, you are asked to select a detector-descriptor couple and a pair of views according to qualitative and quantitative indicators. To achieve this goal, you will use the strategies that you have developed during the lab sessions of the 2nd Unit of the course. We suggest adopting the following strategy:

1. **Object/Scene capture.** Prepare a scene composed of one or more objects. Move around the scene and capture different views of it with your—already calibrated, camera. Note that the more views you capture, the better will be the 3D point cloud reconstruction that can be obtained from them, but also the harder will be to effectively match feature points. The captured images will compose your dataset.

Suggestions: Include textured 3D-objects (to ease feature detection) with sharp and rectilinear contours (to ease tasks in section 3). We also suggest capturing several views of the scene varying the angle and the distance of the camera with respect to the object/scene. Moreover, a nice experiment could be challenging the detectors/descriptors, by altering additional scene capture conditions (e.g. light).

Detection, description and matching of feature points. Select some pairs of views in your dataset, you do not have to be exhaustive, i.e. not every possible pair needs to be studied, but a representative set should be selected. Extract and describe feature points for each view of the pair, and match points between the views. Estimate the homography transformation between the

views for each combination and each selected pair of views and evaluate its correctness according to the correspondences.

Suggestions: We suggest performing this stage together with the next one and testing at least the following detector + descriptor combinations: DoH + SIFT, SURF + SURF, KAZE + KAZE, SIFT + DSP-SIFT.

2. **Qualitative and quantitative evaluation.** Estimate the fundamental matrix between pair of views for every selected pair of views and every detector + descriptor combination. To this aim, we suggest the use of MatLab's **estimateFundamentalMatrix** function.

In order to select the best pair of views and detector-descriptor combination you may qualitatively evaluate the quality of the estimated fundamental matrix through the [vgg_gui F.m](#) function of the vgg-mvg toolbox. Furthermore, you can also evaluate quantitatively the quality of the estimated fundamental matrix by accounting for the number of inliers correspondences (i.e. those that agree with the estimated fundamental matrix) that is returned by **estimateFundamentalMatrix**.

3. **Selection.** Choose the best pair of views and the best detector-descriptor combination according to the results obtained in the previous stages.

Include your reflections, numerical and graphical results in your report—use at most three pages for this section, exceeding this limit has a negative effect in the section score.

Your report may contain in three separate subsections:

- a. A mosaic representation of your scenario, including all the captured views (see MatLab's **montage** function). Include also a brief description of the challenging factors that it contains.
- b. For all the explored pair of views and detector-descriptor combination:
 - A brief argumentation on the selection of the triplet (pair of views, detector and descriptor).
 - Numerical data in the shape of tables or graphs comparing the quantitative indicators.
 - A discussion on the performance of the explored detector and descriptor combinations according to the complexity and the challenging factors of the captured scene, the individual capabilities of the selected methods and their setup.
 - A discussion on the quality of the estimated homography and fundamental matrix according to the correspondences and the scene nature.
- c. For the selected combination:
 - The pair of images with the correspondences overlaid on them.
 - The estimated homography.
 - The warped images.
 - The estimated Fundamental matrix.
 - Screen captures of the [vgg_gui F.m](#) GUI.

Section 3: 3D reconstruction and calibration

In this section, you will use the intrinsic parameters of the camera, obtained in Section 1 and the feature point matches between images, obtained in Section 2, to obtain a 3D reconstruction of your object/scene (a 3D point cloud). To achieve this goal, you will use the strategies that you have used and developed during the lab sessions of the 3rd Unit of the course. We suggest that you follow this strategy:

1. Compute consistent point matches among N views. To that end, you can use the **n_view_matching** function that is provided (you are also allowed to modify this function or use other strategy to match feature points among all views). As an input to this function, you must provide interest points and descriptors in the same format that was used in the lab sessions of the 2nd Unit of the course. Use the results in Section 2 to decide which detector/descriptor to use and the separation between consecutive images of the scene. In the report:
 - a. Provide the images that you have used for the N-view point matching, indicating the detected interest points in each of them.
2. Compute the Fundamental matrix and an initial projective reconstruction from 2 of the cameras. In the report:
 - a. Provide the images that you have used for the estimation of the Fundamental matrix, indicating the detected interest points and point matches.
 - b. Provide the mean re-projection error and the reprojection error histogram.
3. Improve this initial reconstruction by means of a Projective Bundle Adjustment, using a higher number (maybe all) of your images. In the report:
 - a. Provide the mean re-projection error and the reprojection error histogram at two points: (i) after the resectioning step, and (ii) after the Projective Bundle Adjustment step.
 - b. Comment on the justification of the different re-projection error values in 2.b and the two steps of 3.a.
4. Re-compute the Fundamental matrix between two of the cameras, using the projection matrices obtained after the Projective Bundle Adjustment step. For this purpose, you can use the **vgg_F_from_P** function from the vgg-mvg toolbox.
5. Use the properties of the Essential matrix (between two cameras) to obtain a Euclidean reconstruction of the scene (use the re-projected points obtained after the Projective Bundle Adjustment step). Remember that you have already computed the intrinsic parameters of your camera from Step 1. In the report:
 - a. Provide the mean re-projection error and the reprojection error histogram (for these two cameras)
 - b. Provide illustrative results (several viewpoints and the 3D Matlab figure) of your 3D point cloud reconstruction.
 - c. **Extra:** Provide illustrative results (several viewpoints and the 3D Matlab figure) of an “improved” 3D point cloud. You can improve your point cloud with strategies such as: line segments that connect points that are joined by straight lines in your object/scene, “painting” your point cloud with RGB values of the pixels, cluster points from different objects and “paint” them with different colors, etc.