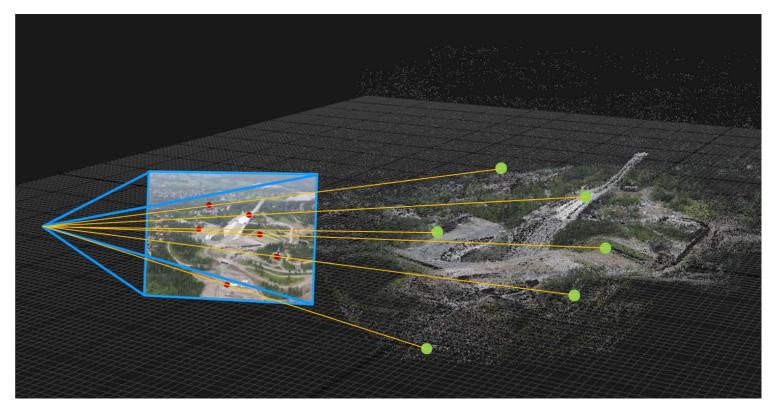


Lab 5 – Pose estimation and Augmented Reality

# **Topics**

- Camera calibration
- 3D-2D pose estimation
- AR and 3D visualization





#### Lab 5 in short

- Calibrate the camera
- Create a planar 3D world model with point descriptors
- Find 3D-2D correspondences between the world model and the current frame
- Estimate the camera pose from the 3D-2D correspondences
- Visualize the 3D world in the camera frame (AR)
- Visualize the camera and world model in 3D



# **Step 1: Camera calibration**

- Use the OpenCV application <u>opencv\_interactive-calibration</u> to calibrate the camera
  - Use tape to fasten the chessboard to the desk
  - Open the terminal and go to the project directory
  - Run opencv\_interactive-calibration:

opencv\_interactive-calibration -ci=0 -t=chessboard -sz=30 -w=8 -h=5 -pf=calibSettings.xml

- Make sure to measure the chessboard from different orientations and positions
- When you are happy with the calibration, store the results by pressing s
- Quit by pressing Esc
- Set calibration parameters in setupCameraModel() in lab\_5.cpp



```
// TODO 2-1: Compute M and M_bar.
// Compute the matrix M
// and extract M_bar (the two first columns of M).
Eigen::Matrix3d M;
Eigen::MatrixXd M_bar;
         \boldsymbol{M} = \boldsymbol{K}^{-1} \boldsymbol{H}_{W}^{C}
         \bar{\boldsymbol{M}} = [\boldsymbol{m}_1, \boldsymbol{m}_2]
```

```
// TODO 2-2: Compute Singular Value Decomposition.
// Perform SVD on M_bar.
auto svd = M_bar.jacobiSvd(Eigen::ComputeThinU | Eigen::ComputeThinV);
// TODO 2-3: Compute R bar.
// Compute R_bar (the two first columns of R)
// from the result of the SVD.
       \bar{R} = UV^T
       \bar{\boldsymbol{R}} = [r_1, r_2]
```

```
// TODO 2-4: Construct R.
// Construct R by inserting R_bar and
// computing the third column of R from the two first.
// Remember to check det(R)!
Eigen::Matrix3d R = Eigen::Matrix3d::Identity();
```

```
// TODO 2-5: Compute the scale.
// Compute the scale factor lambda.
double lambda = 0;
```

$$\lambda = \frac{\operatorname{trace}(\bar{\boldsymbol{R}}^T \bar{\boldsymbol{M}})}{\operatorname{trace}(\bar{\boldsymbol{M}}^T \bar{\boldsymbol{M}})} = \frac{\sum_{i=1}^{3} \sum_{j=1}^{2} R_{ij} M_{ij}}{\sum_{i=1}^{3} \sum_{j=1}^{2} M_{ij}^{2}}$$

```
// TODO 2-6: Compute t.
// Extract the translation t.
Eigen::Vector3d t = M.col(2) * lambda;
```

$$[r_1,r_2,t]=\pm\lambda M$$

```
// TODO 2-7: Find correct solution.
// Check that this is the correct solution
// by testing the last element of t.
```

$$[r_1,r_2,t]=\pm\lambda M$$

# **Step 3: Try the other pose estimators**

- PnPPoseEstimator pose\_estimator(camera\_model.K);
- auto init\_estimator = std::make\_shared<HomographyPoseEstimator>(camera\_model.K);
   GtsamPoseEstimator pose\_estimator(init\_estimator, camera\_model.K);



# Step 4: Play!

- Try other feature detectors and descriptors
  - What about <u>cv::xfeatures2d::AffineFeature2D</u>?
- Add cool new 3D elements to the AR viewer
- Visualize the 3D track over the last n frame

