Supplementary Material for Making Minimal Solvers for Absolute Pose Estimation Compact and Robust

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1. Elimination Ideal

To find the new constraints on the camera matrix,

$$p_{21}p_{31} + p_{22}p_{32} + p_{23}p_{33} = 0$$

$$p_{11}p_{31} + p_{12}p_{32} + p_{13}p_{33} = 0$$

$$p_{11}p_{21} + p_{12}p_{22} + p_{13}p_{23} = 0$$

$$p_{11}^{2} + p_{12}^{2} + p_{13}^{2} - p_{21}^{2} - p_{22}^{2} - p_{23}^{2} = 0$$

$$p_{13}^{2}p_{32} - p_{21}^{2}p_{32} - p_{22}^{2}p_{32} - p_{12}p_{13}p_{33} - p_{22}p_{23}p_{33} = 0$$

$$p_{12}p_{13}p_{32} + p_{22}p_{23}p_{32} - p_{12}^{2}p_{33} + p_{21}^{2}p_{33} + p_{23}^{2}p_{33} = 0$$

$$p_{11}p_{13}p_{32} + p_{21}p_{23}p_{32} - p_{11}p_{12}p_{33} - p_{21}p_{22}p_{33} = 0$$

$$p_{13}^{2}p_{31} - p_{22}^{2}p_{31} + p_{21}p_{22}p_{32} - p_{11}p_{13}p_{33} = 0$$

 $p_{12}p_{13}p_{31} + p_{22}p_{23}p_{31} - p_{11}p_{12}p_{33} - p_{21}p_{22}p_{33} = 0$

the following Macaulay2 [2] code was used:

```
KK = ZZ / 30097;
R = KK[p_{1,1}..p_{3,3},r_{1,1}..r_{3,3},f,s]
-- we can ignore translation since it poses
-- no extra constraint on P
```

P = map(R³, R³, (i, j) -> p_{i+1, j+1}); Q = map(R³, R³, (i, j) -> r_{i+1, j+1}); K = matrix({{f,0,0}, {0, f, 0}, {0, 0, 1}}); eye = matrix({{s,0,0}, {0, s, 0}, {0, 0, s}});

eqs = flatten(P-K*Q | transpose(Q)*Q-eye);

```
I = saturate(ideal eqs, s);
```

2. Additional Figures from the Experiments

2.1. Noise Experiment

Figure 1 shows the errors in the radial distortion for the noise experiment in Section 4.2.

2.2. Evaluation on Real Images

Figure 2 and Figure 3 shows histograms of the relative errors for the *Rotunda* and *Graffiti* datasets. Figure 4 and Figure 5 shows example images from the datasets.

References

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- [2] D. R. Grayson and M. E. Stillman. Macaulay 2, a software system for research in algebraic geometry, 2002. 1
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Figure 1. Comparison of the errors of radial distortion estimated by different solvers for varying levels of noise. The ground truth values were set to $f_{gt} = 1.5$ and $k_{gt} = -0.4$. Left: Median radial distortion errors. Right: Boxplot of radial distortion errors.



Figure 2. Errors (\log_{10}) for the *Rotunda* dataset. The measured errors are the relative focal length error $|f - f_{GT}|/f_{GT}$, the distortion error $|k - k_{GT}|$, the rotation error in degrees and the relative translation error $||t - t_{GT}||/||t_{GT}||$.



Figure 3. Errors (\log_{10}) for the *Graffiti* dataset. The measured errors are the relative focal length error $|f - f_{GT}|/f_{GT}$, the distortion error $|k - k_{GT}|$, the rotation error in degrees and the relative translation error $||t - t_{GT}|/|t_{GT}||$.



Figure 4. Some examples of the images in the Rotunda. Left: Original images. Right: Undistorted images.



Figure 5. Some examples of the images in the Graffiti. Left: Original images. Right: Undistorted images.