

# Supplementary Material for

## Absolute Pose Estimation from Line Correspondences using Direct Linear Transformation

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This supplementary material contains boxplots visualizing distributions of errors of individual PnL methods. Each distribution over 1000 trials is depicted by a box, where

- black dot inside a mark = median,
- box body = interquartile range (IQR),
- whiskers = minima and maxima in the interval of  $10 \times$  IQR, and
- isolated dots = outliers.

### 1 Robustness to image noise

Figures 1 – 15 depict errors in estimated camera pose as a function of the number of lines (from 3 to 10,000). The following methods were tested. Each method is assigned a different color and shape in the figures.

- ▶ **Ansar**, the method by Ansar and Daniilidis (2003), implementation from Xu et al. (2016).
- **Mirzaei**, the method by Mirzaei and Roumeliotis (2011).
- ◆ **RPnL**, the method by Zhang et al. (2013).
- ◆ **ASPnL**, the method by Xu et al. (2016).
- ★ **LPnL\_Bar\_LS**, the method by Xu et al. (2016).
- ★ **LPnL\_Bar\_ENull**, the method by Xu et al. (2016).
- ▲ **DLT-Lines**, the method by Hartley and Zisserman (2004, p. 180), our implementation.
- ▼ **DLT-Plücker-Lines**, the method by Příbyl et al. (2015), our implementation.
- **DLT-Combined-Lines**, the proposed method.

Figures 1 – 5 depict errors in estimated camera orientation  $\Delta\Theta$  [°] for increasing levels of image noise with standard deviation  $\sigma = 1, 2, 5, 10$  and  $20$  pixels. Accordingly, Figures 6 – 10 depict errors in estimated camera position  $\Delta T$  [m], and Figures 11 – 15 depict reprojection errors  $\Delta\pi$  [°].

### 2 Robustness to quasi-singular cases

The methods were evaluated in three quasi-singular scenarios:

- Limited number of line directions (2 random directions, 3 random directions, 3 orthogonal directions) – Figure 16.
- Near-planar lines – Figure 17.
- Near-concurrent lines – Figure 18.

### 3 Robustness to outliers

Figure 19 depicts errors in estimated camera pose as a function of the fraction of outliers, out of total 500 line correspondences. Legend is provided in the figure itself.



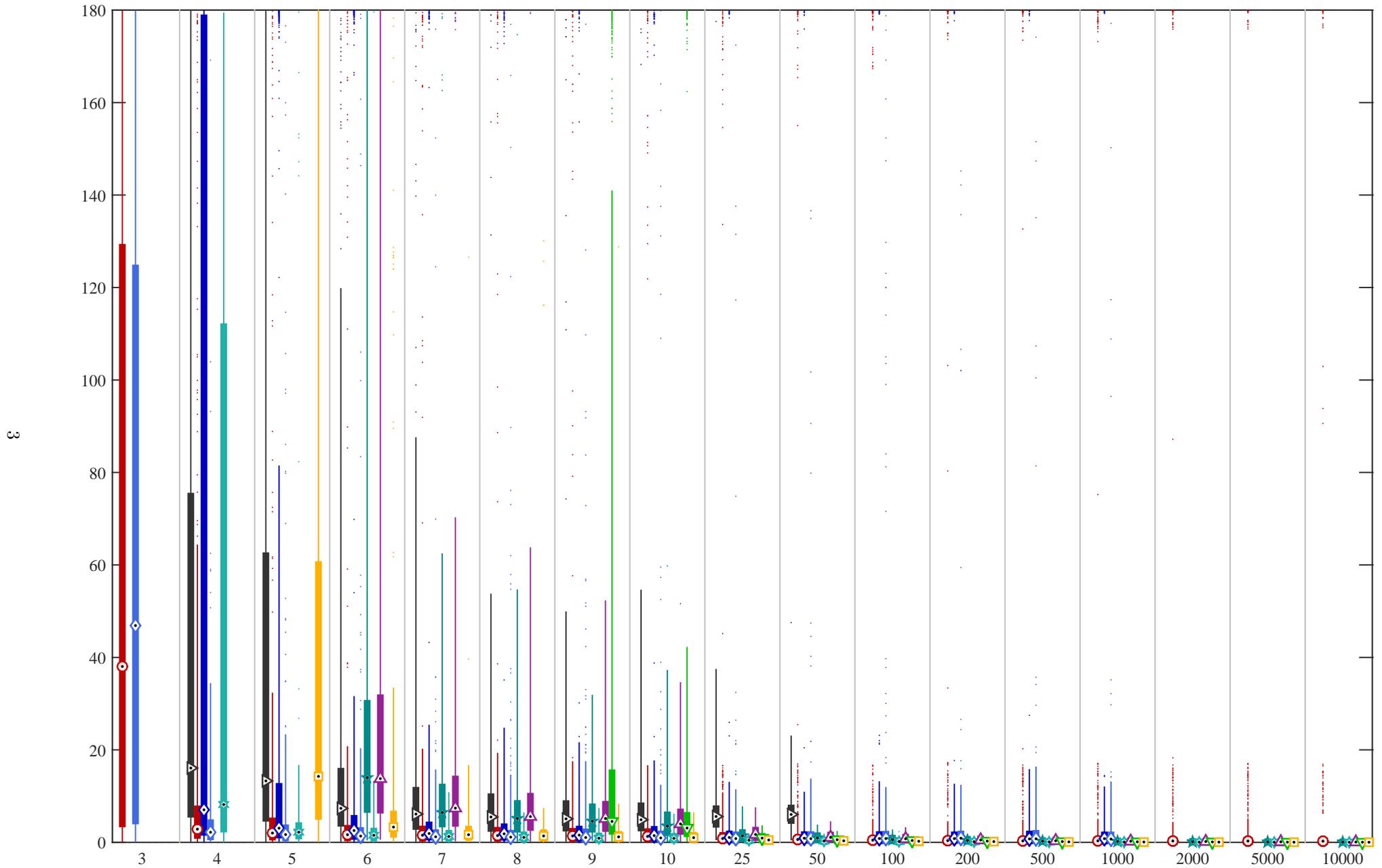


Figure 2: Errors in camera orientation  $\Delta\Theta$  [°] for image noise with standard deviation  $\sigma = 2$  pixels.

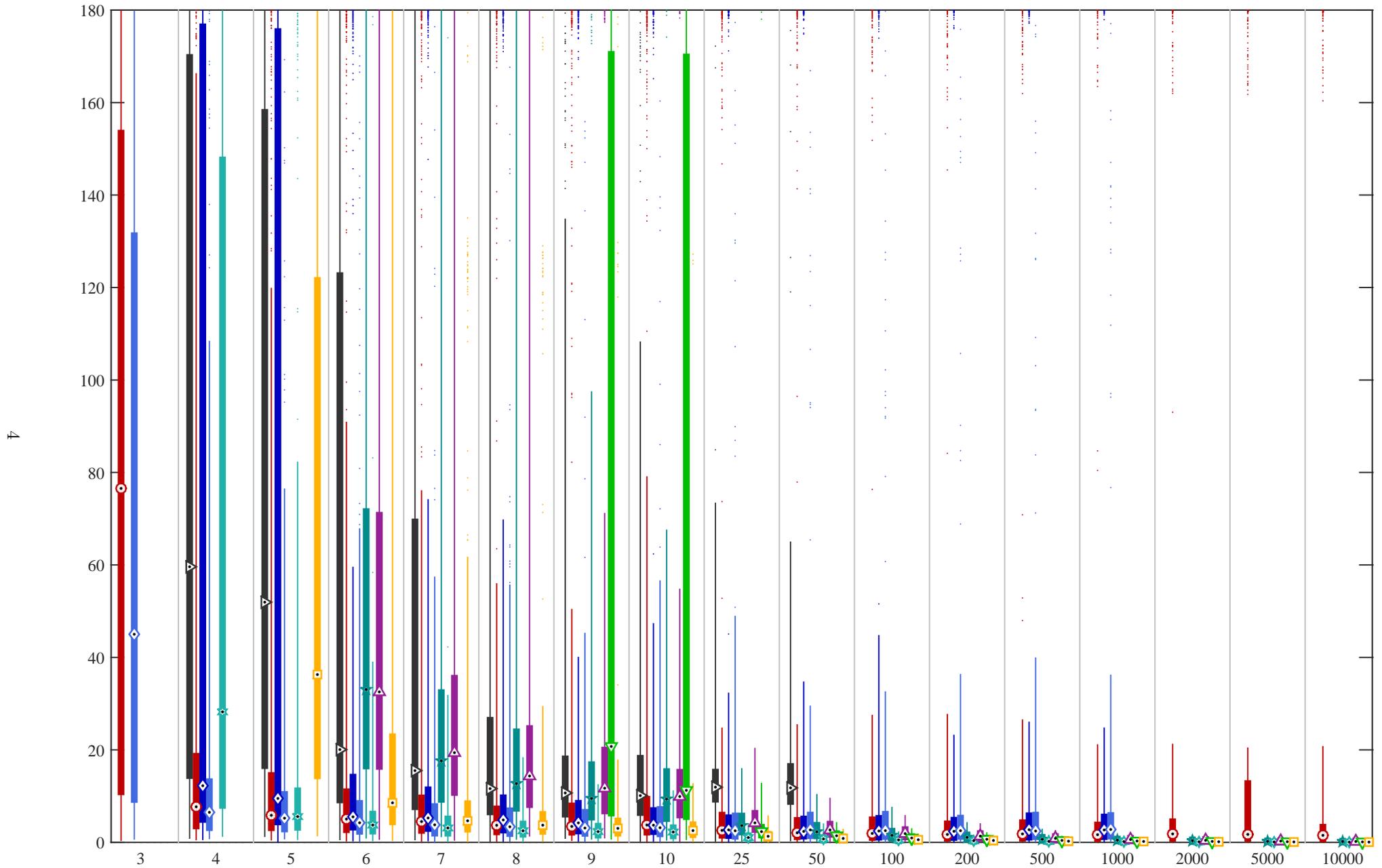


Figure 3: Errors in camera orientation  $\Delta\Theta$  [°] for image noise with standard deviation  $\sigma = 5$  pixels.

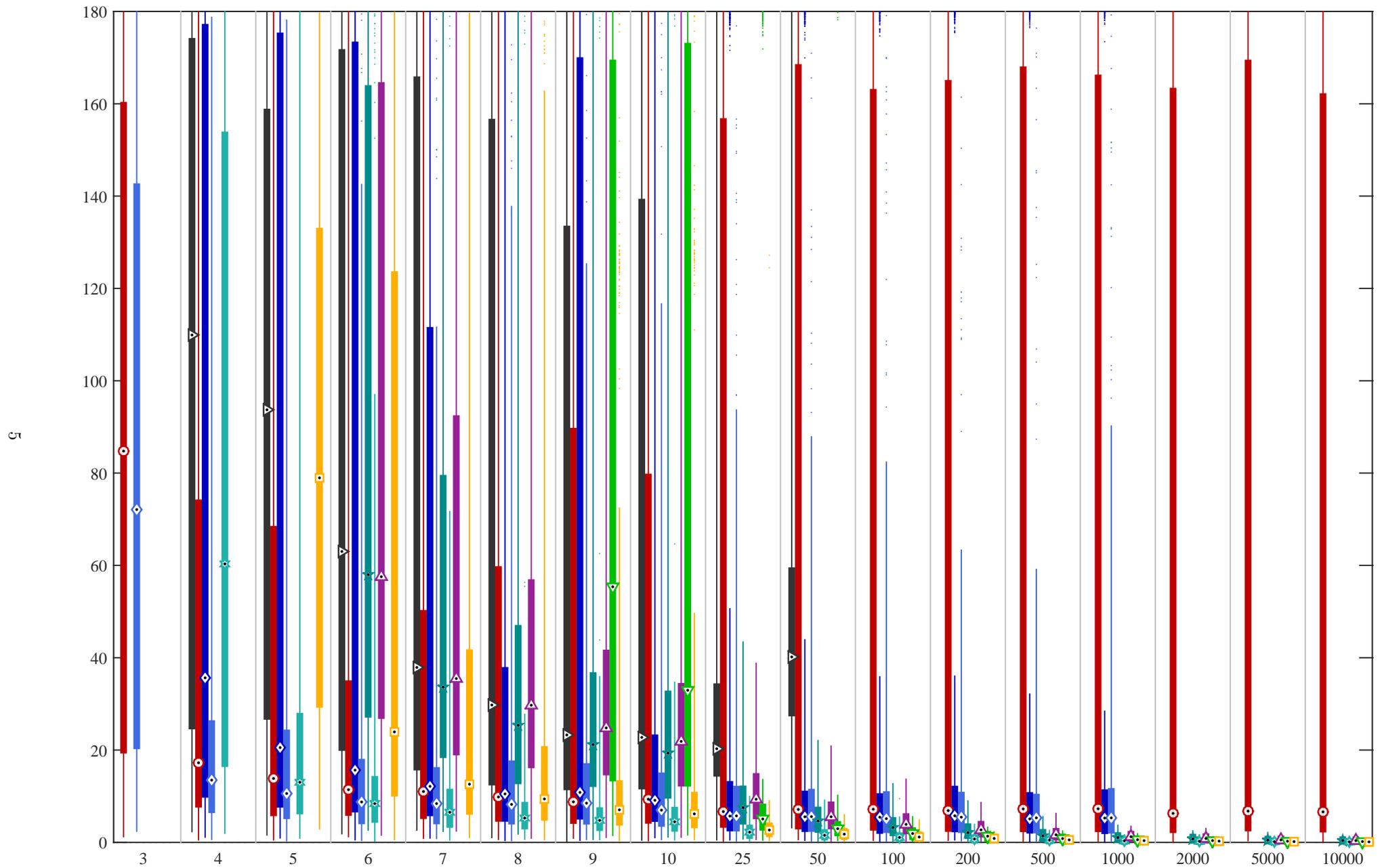


Figure 4: Errors in camera orientation  $\Delta\Theta$  [°] for image noise with standard deviation  $\sigma = 10$  pixels.

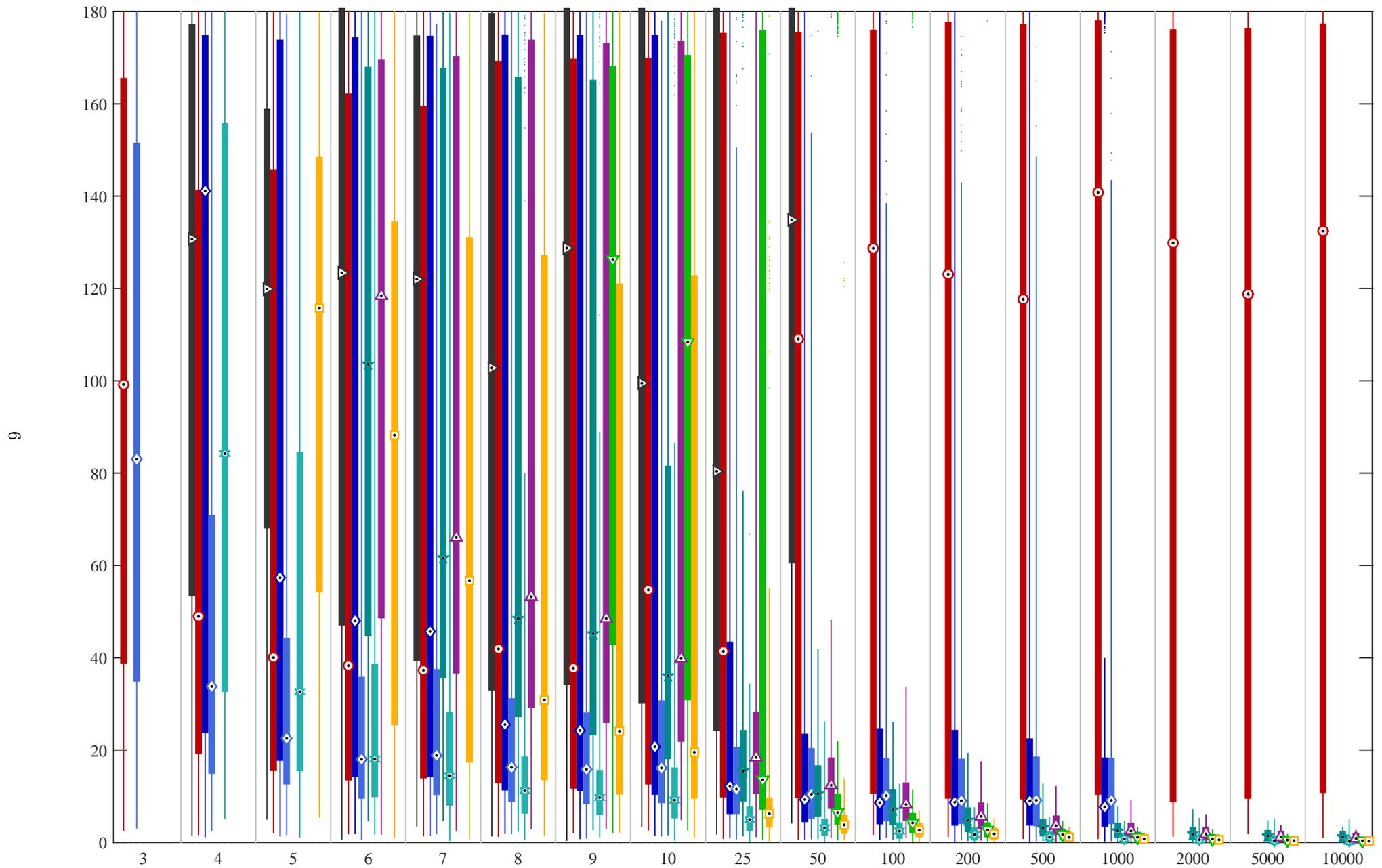


Figure 5: Errors in camera orientation  $\Delta\Theta$  [°] for image noise with standard deviation  $\sigma = 20$  pixels.

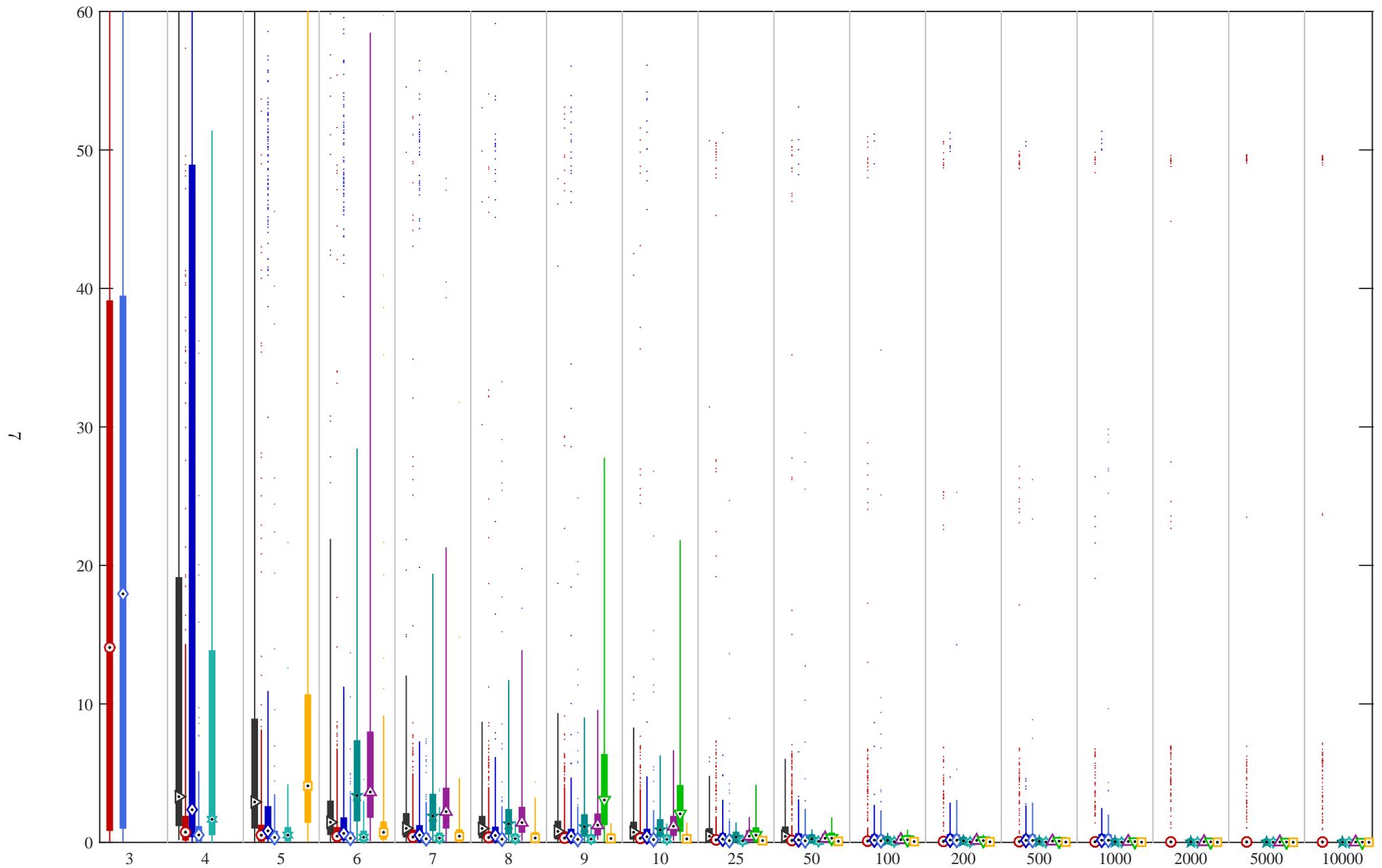


Figure 6: Errors in camera position  $\Delta T$  [m] for image noise with standard deviation  $\sigma = 1$  pixel.

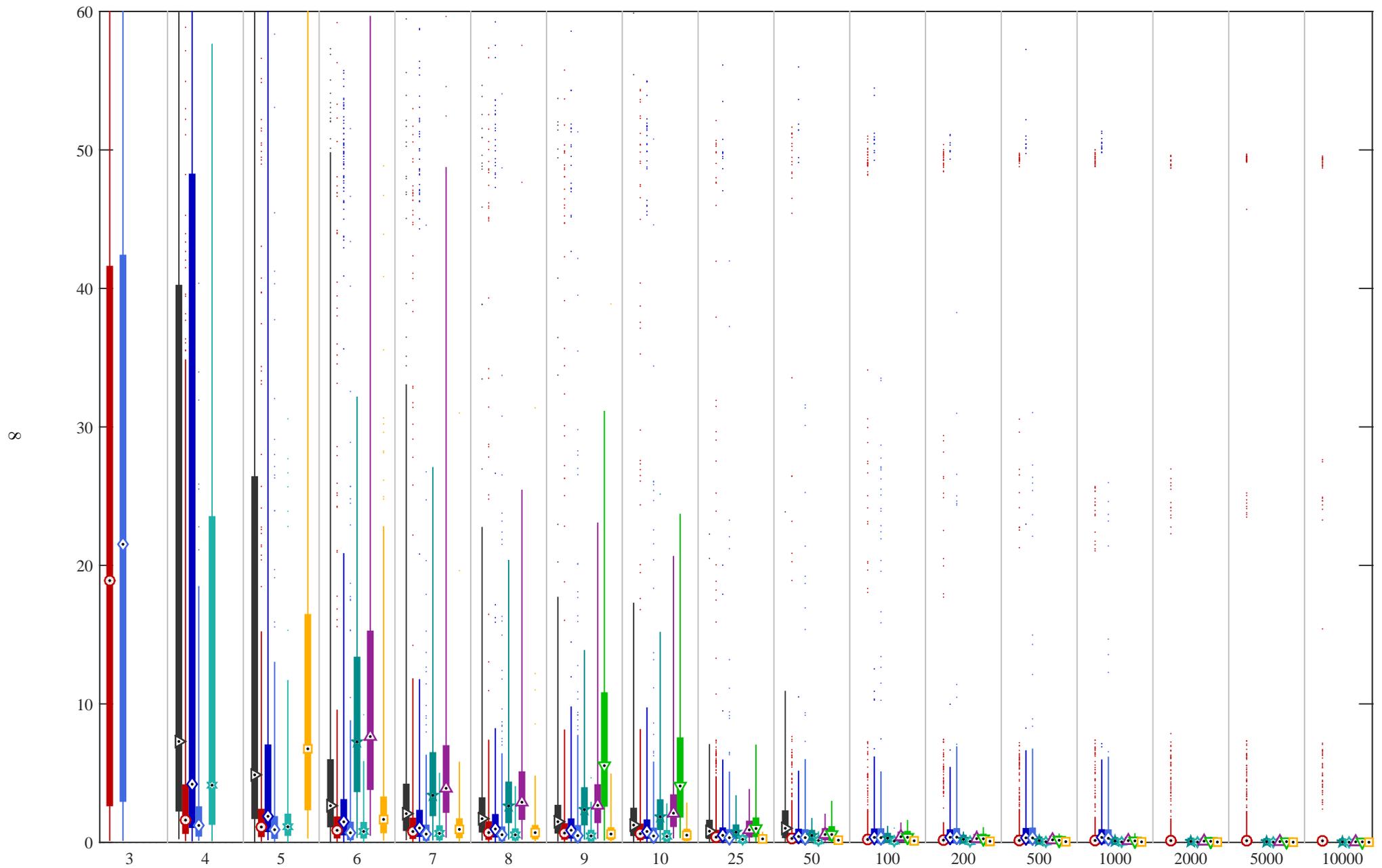


Figure 7: Errors in camera position  $\Delta T$  [m] for image noise with standard deviation  $\sigma = 2$  pixels.



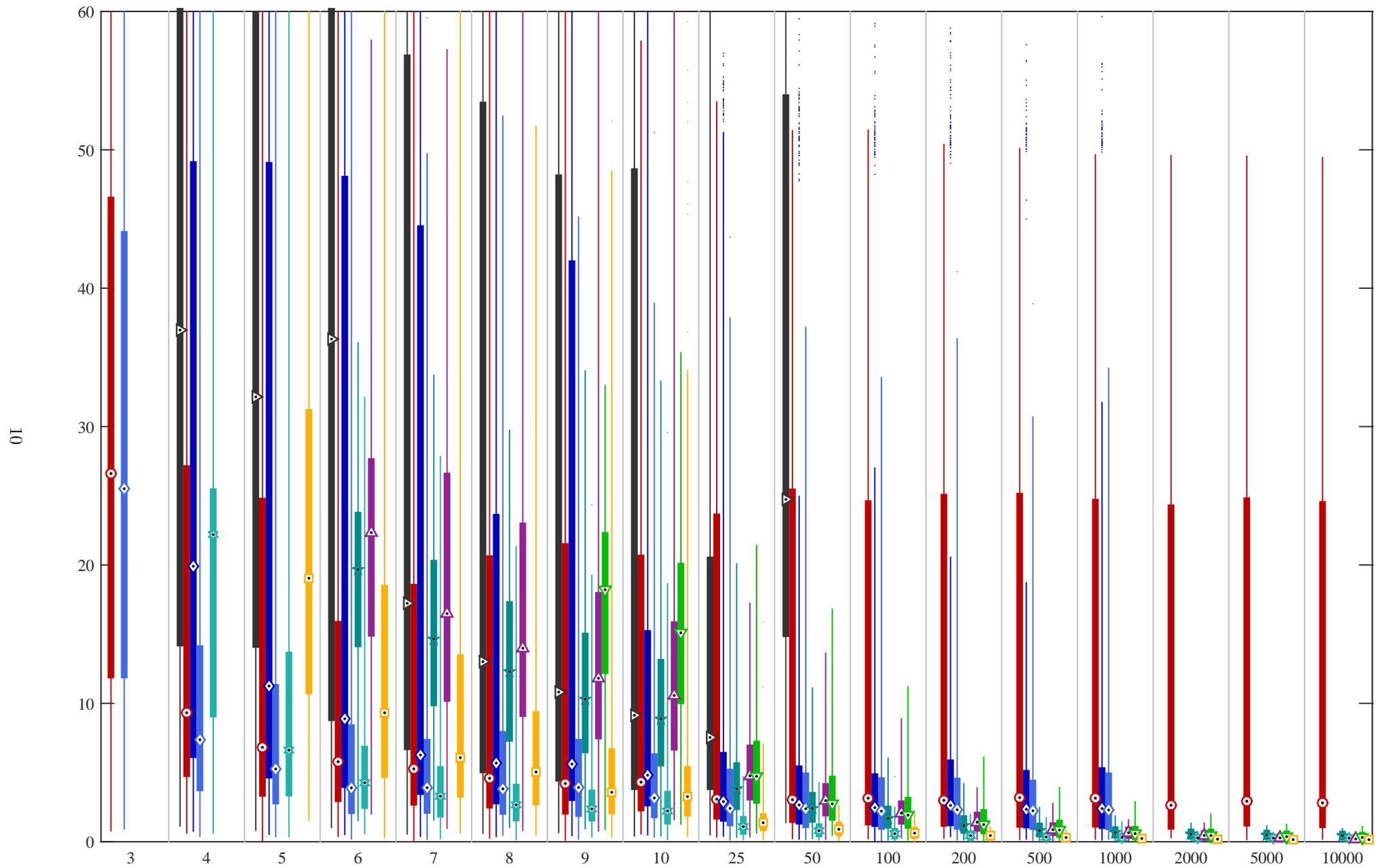


Figure 9: Errors in camera position  $\Delta T$  [m] for image noise with standard deviation  $\sigma = 10$  pixels.

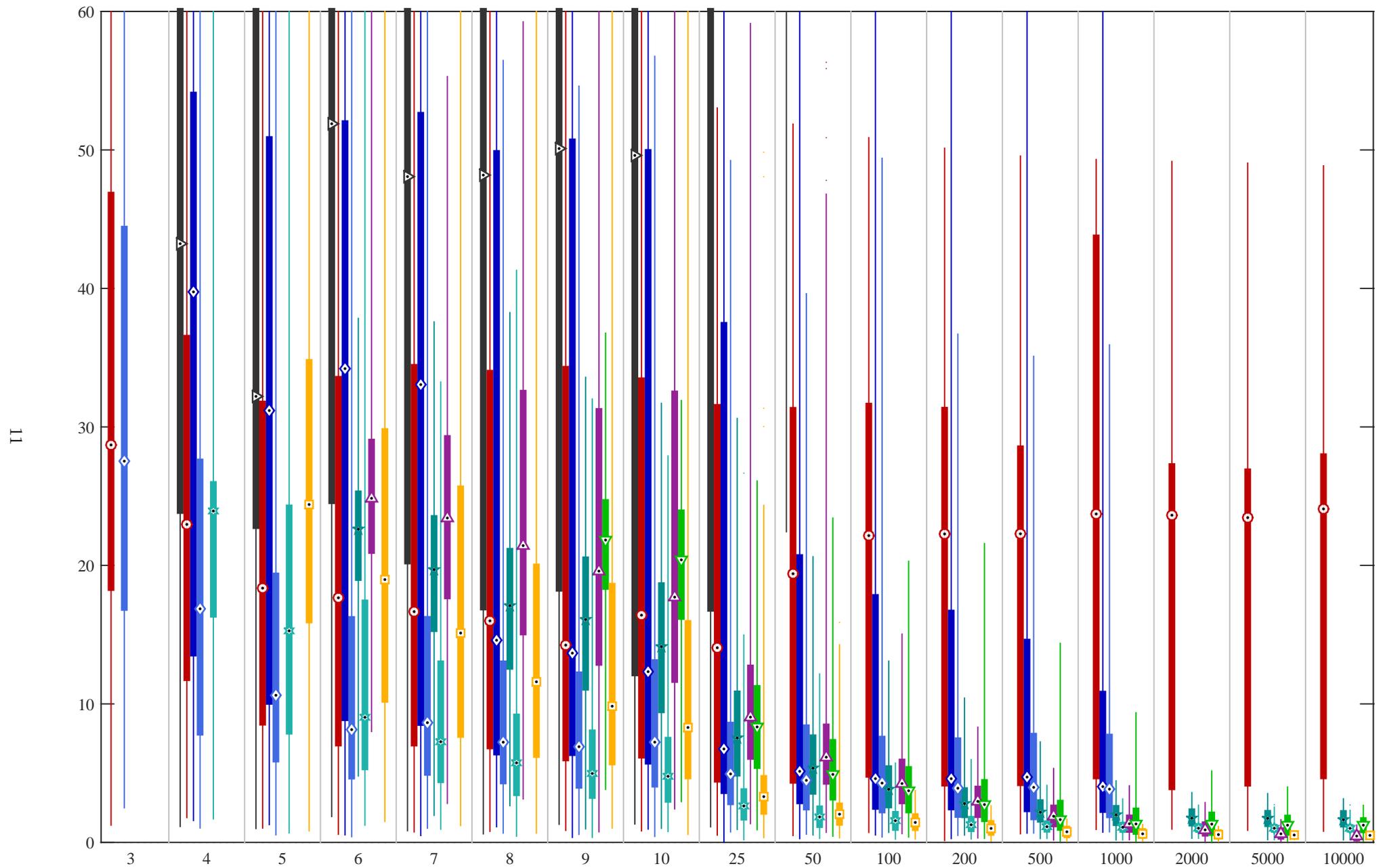


Figure 10: Errors in camera position  $\Delta T$  [m] for image noise with standard deviation  $\sigma = 20$  pixels.

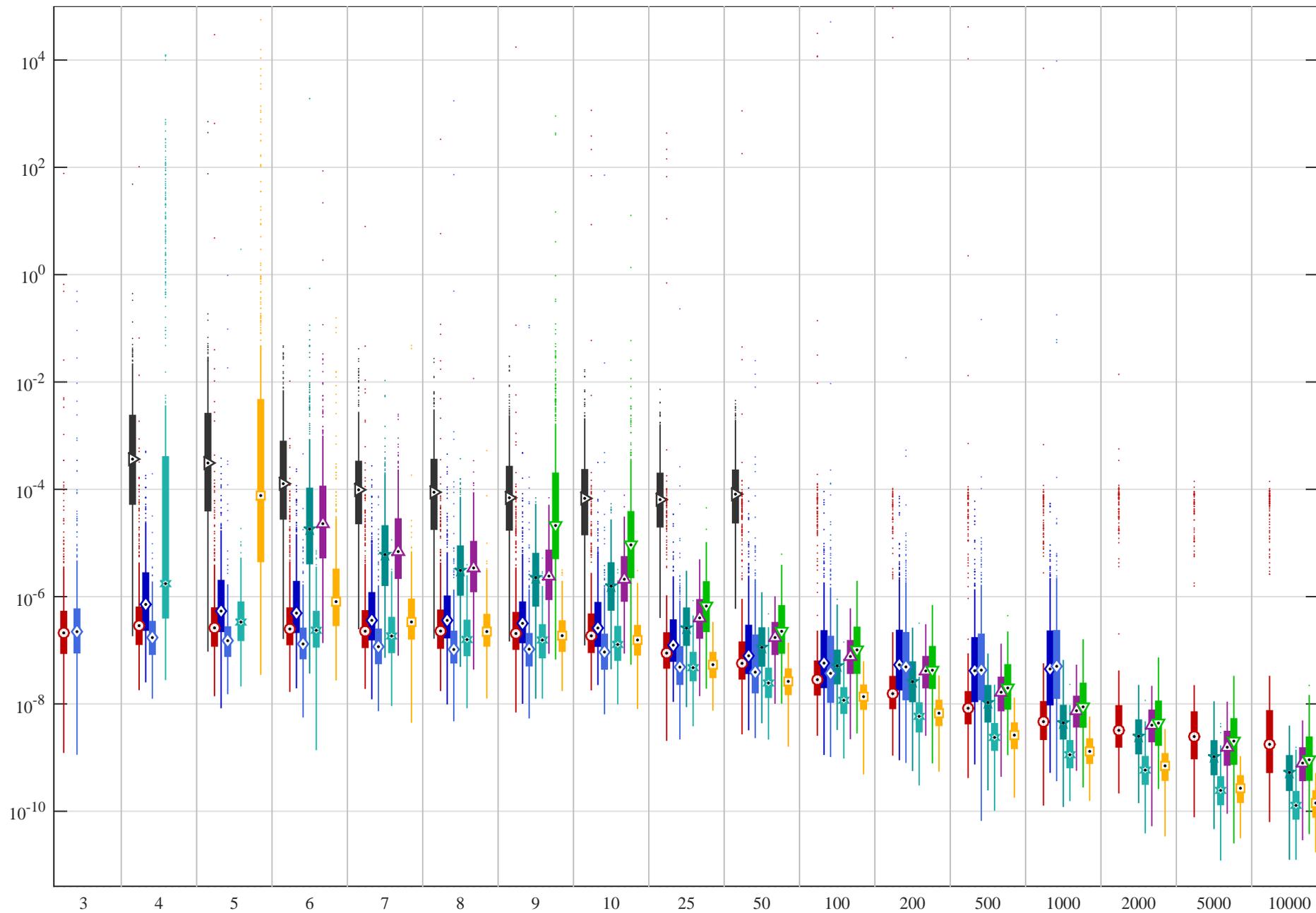


Figure 11: Reprojection errors  $\Delta\pi$  [] for image noise with standard deviation  $\sigma = 1$  pixel.

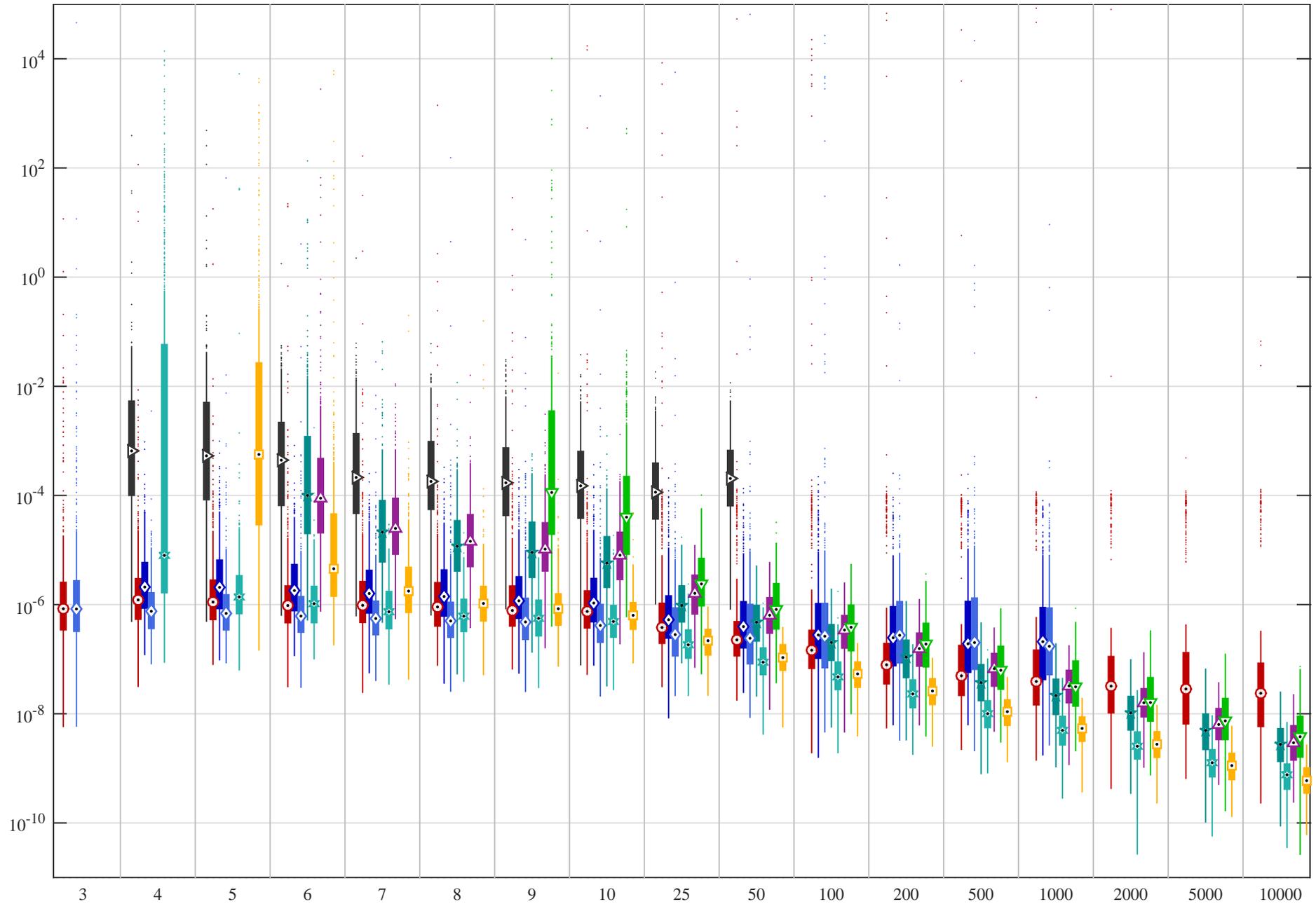


Figure 12: Reprojection errors  $\Delta\pi$  [ ] for image noise with standard deviation  $\sigma = 2$  pixels.

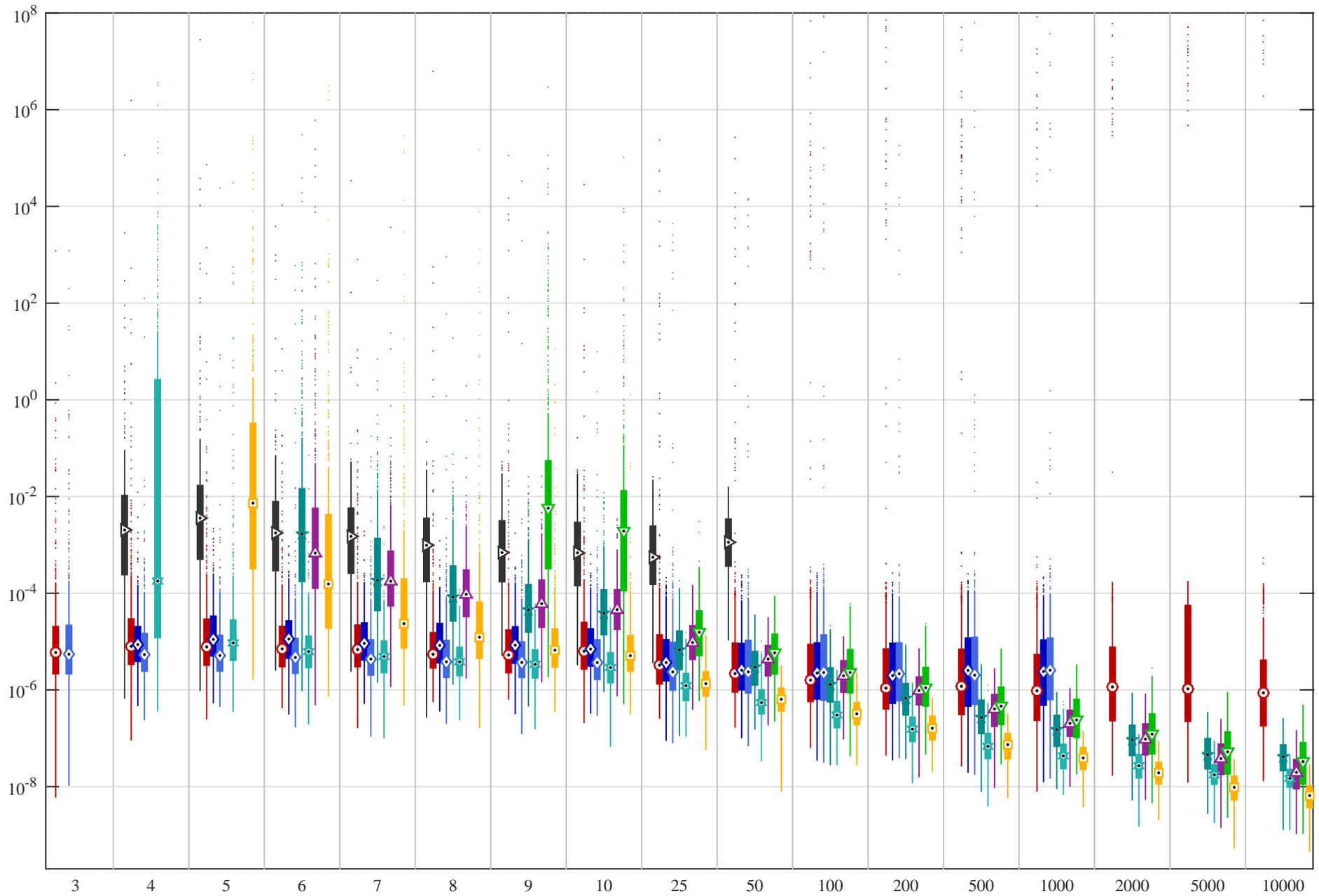


Figure 13: Reprojection errors  $\Delta\pi$  [ ] for image noise with standard deviation  $\sigma = 5$  pixels.

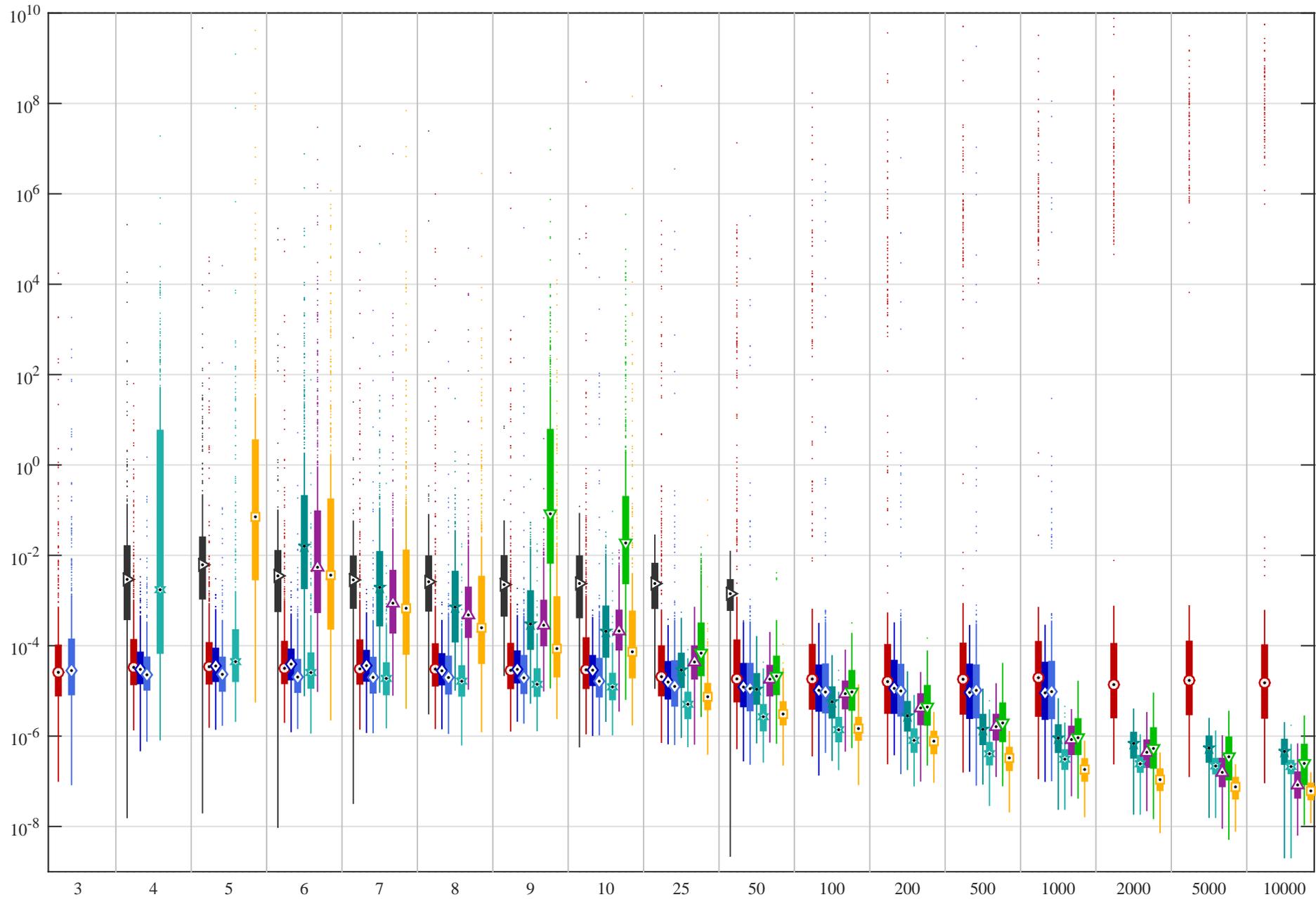


Figure 14: Reprojection errors  $\Delta\pi$  [ ] for image noise with standard deviation  $\sigma = 10$  pixels.

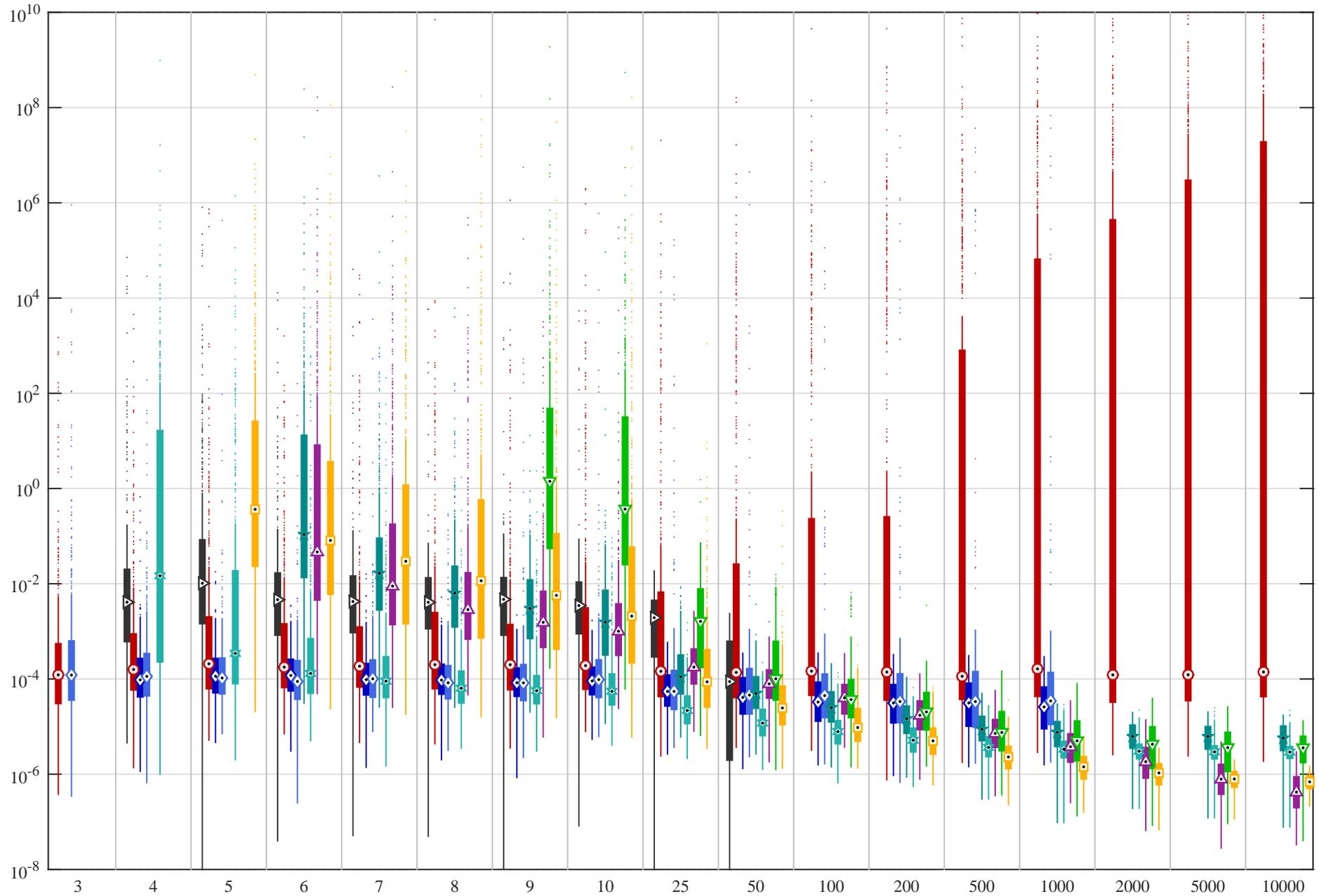


Figure 15: Reprojection errors  $\Delta\pi$  [ ] for image noise with standard deviation  $\sigma = 20$  pixels.

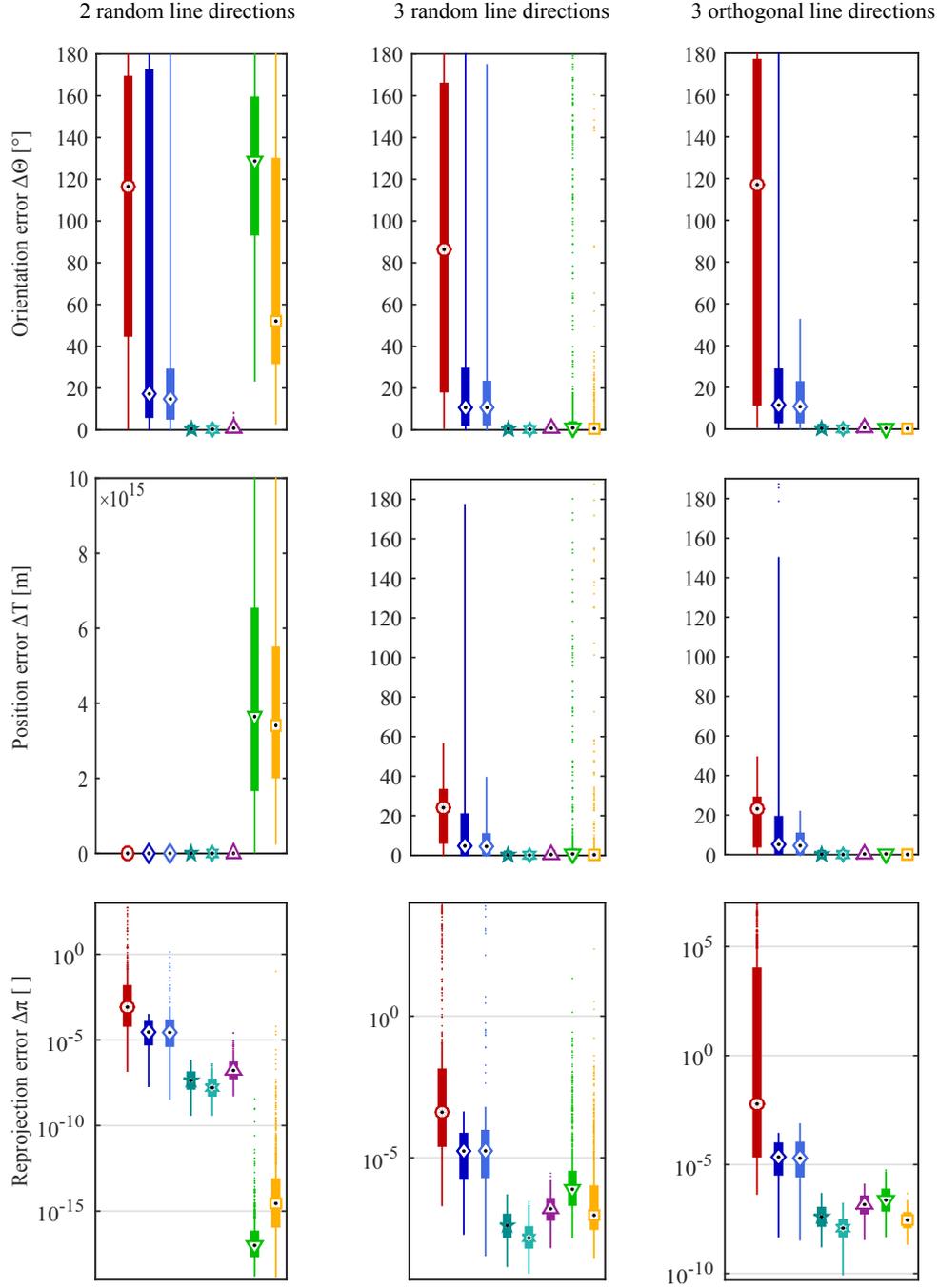


Figure 16: Robustness to quasi-singular cases. The distribution of orientation errors ( $\Delta\Theta$ , *top*), position errors ( $\Delta T$ , *middle*) and reprojection errors ( $\Delta\pi$ , *bottom*) for the case with 2 random line directions (*left*), 3 random line directions (*center*) and 3 orthogonal line directions (*right*). Number of lines was  $m = 200$  and standard deviation of image noise was  $\sigma = 2$  pixels.

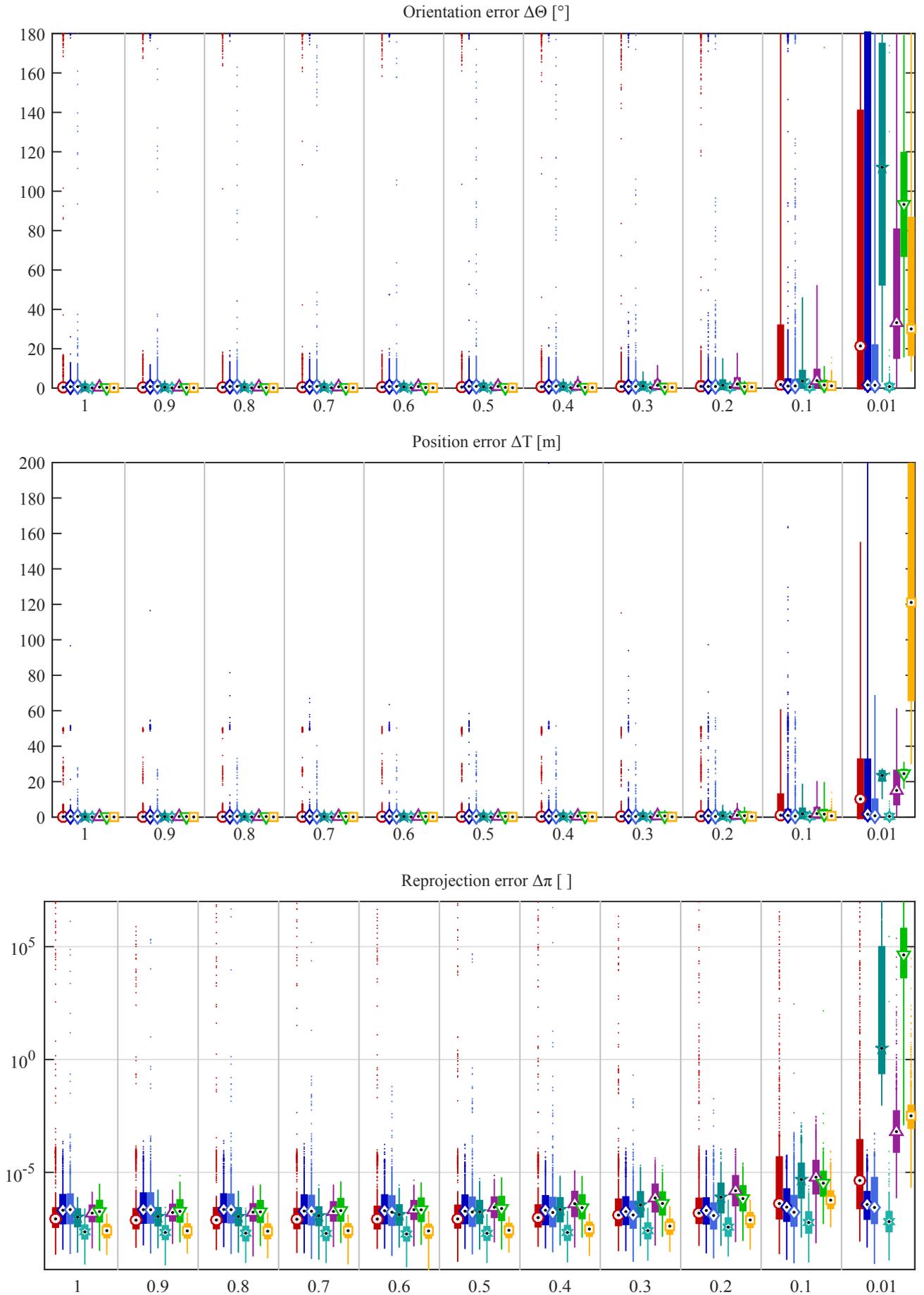


Figure 17: Robustness to near-planar line distribution. The distribution of orientation errors ( $\Delta\Theta$ , *top*), position errors ( $\Delta T$ , *middle*) and reprojection errors ( $\Delta\pi$ , *bottom*) as a function of ‘flatness’ (the ratio of height of a volume containing 3D lines w.r.t. to its remaining dimensions). Number of lines was  $m = 200$  and standard deviation of image noise was  $\sigma = 2$  pixels.

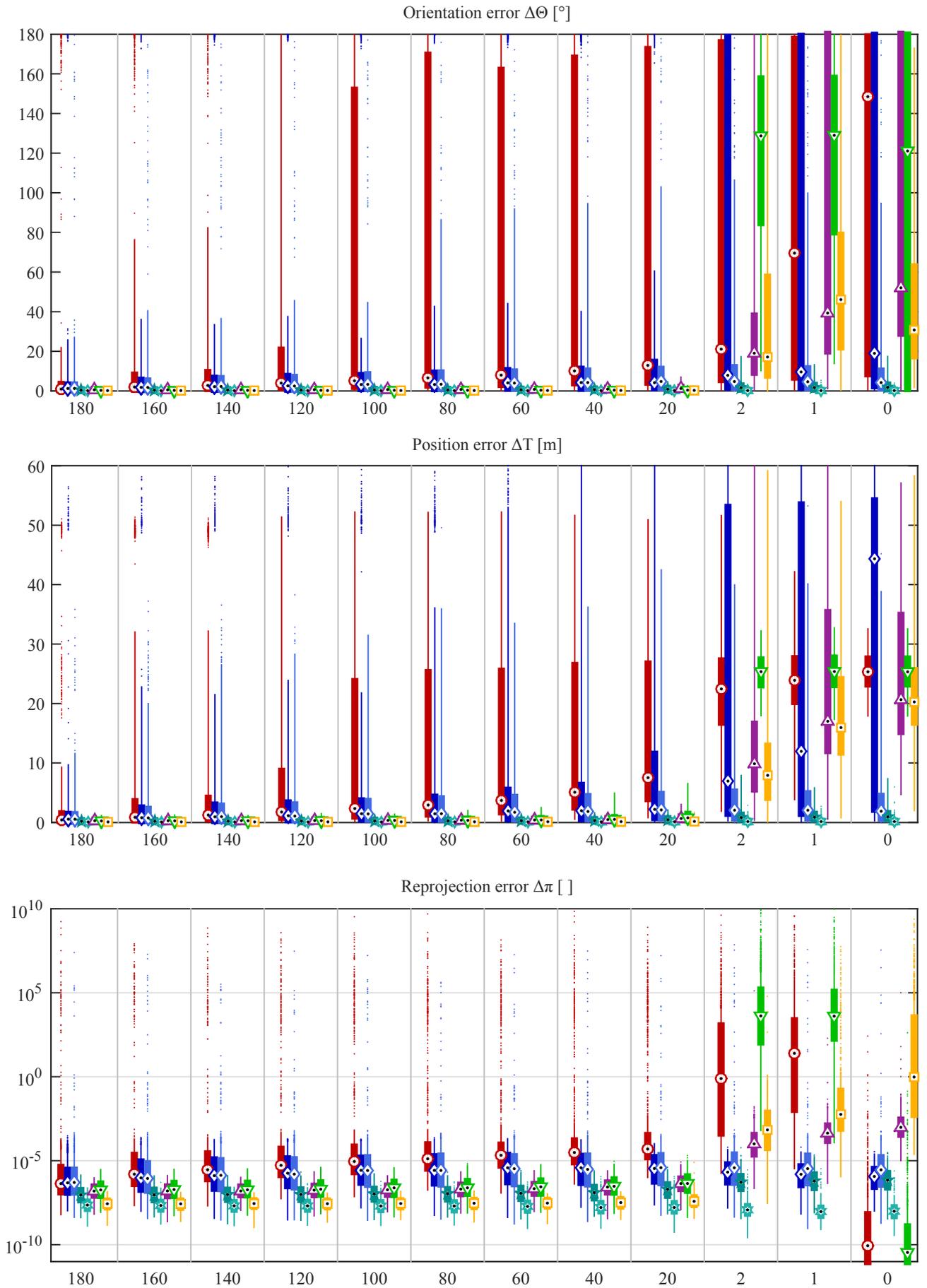


Figure 18: Robustness to near-concurrent line distribution. The distribution of orientation errors ( $\Delta\Theta$ , *top*), position errors ( $\Delta T$ , *middle*) and reprojection errors ( $\Delta\pi$ , *bottom*) as a function of the number of lines, which are not concurrent, out of all  $m = 200$  lines. Standard deviation of image noise was  $\sigma = 2$  pixels.

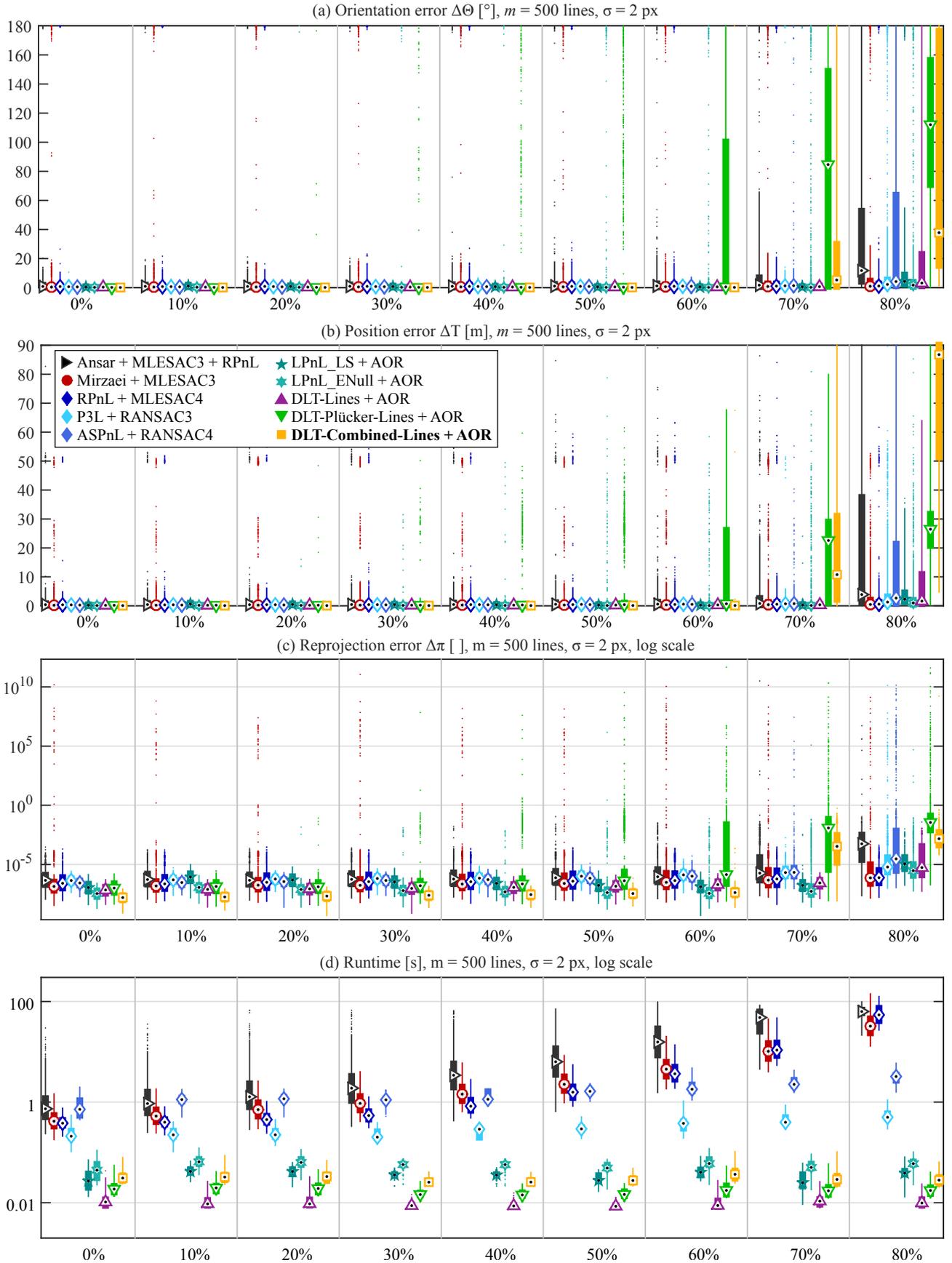


Figure 19: Robustness to outliers. The distribution of orientation errors ( $\Delta\Theta$ , *a*), position errors ( $\Delta T$ , *b*), reprojection errors ( $\Delta\pi$ , *c*) and runtimes (*d*) as a function of the fraction of outliers, out of total 500 line correspondences.

## References

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