# Supplementary Material: CrossNet: An End-to-end Reference-based Super Resolution Network using Cross-scale Warping

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No Institute Given

This supplementary material contains the visualization of the ×8 RefSR results evaluated on Stanford light field dataset (from Table 2 of the main paper) and Scene light field dataset (from Table 3 of the main paper).
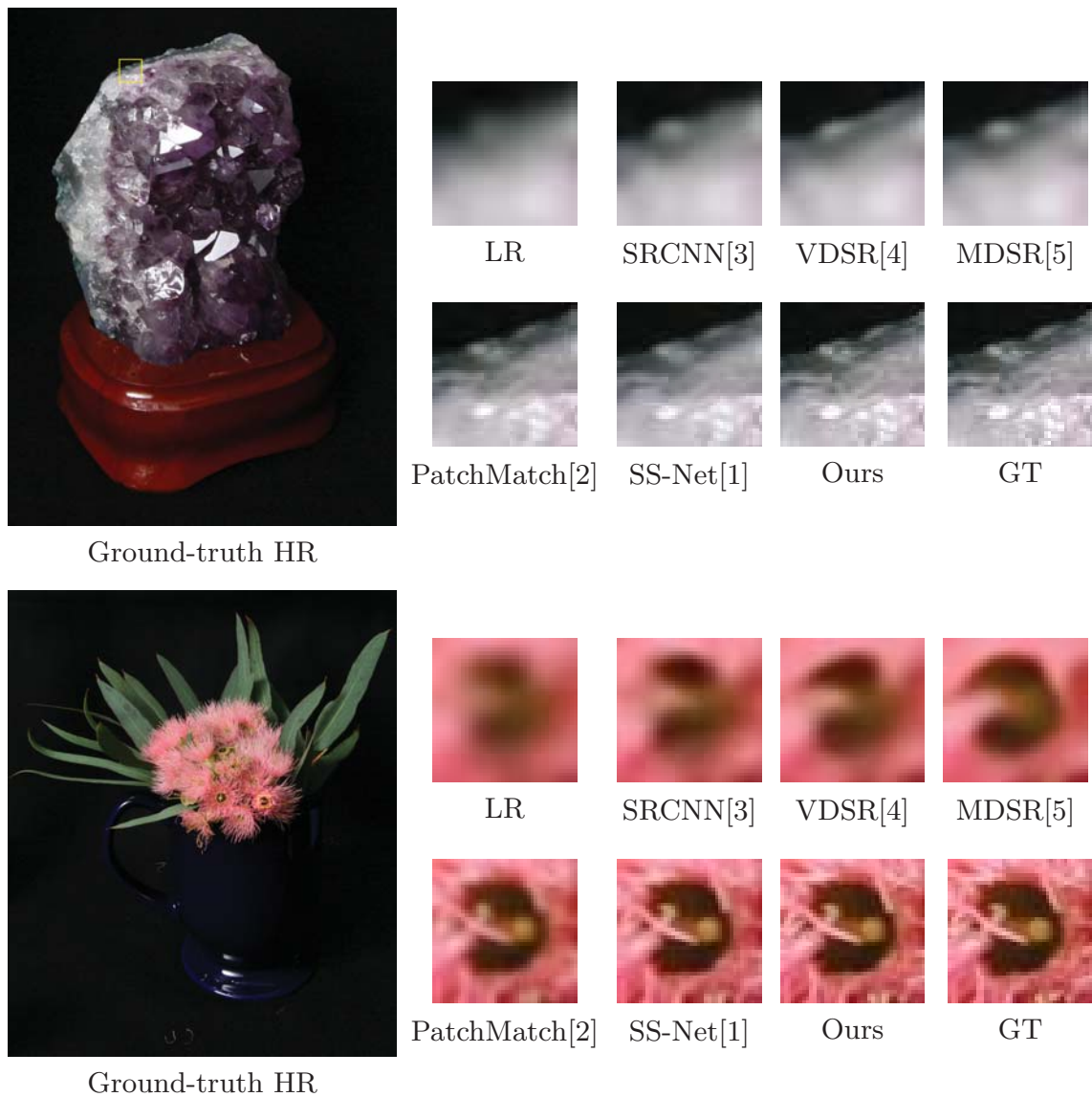
Fig. 1. Visual comparison for ×8 RefSR on Scene light field dataset. The reference images are sampled at location (0,0), where LR images are sampled at location (1,0).
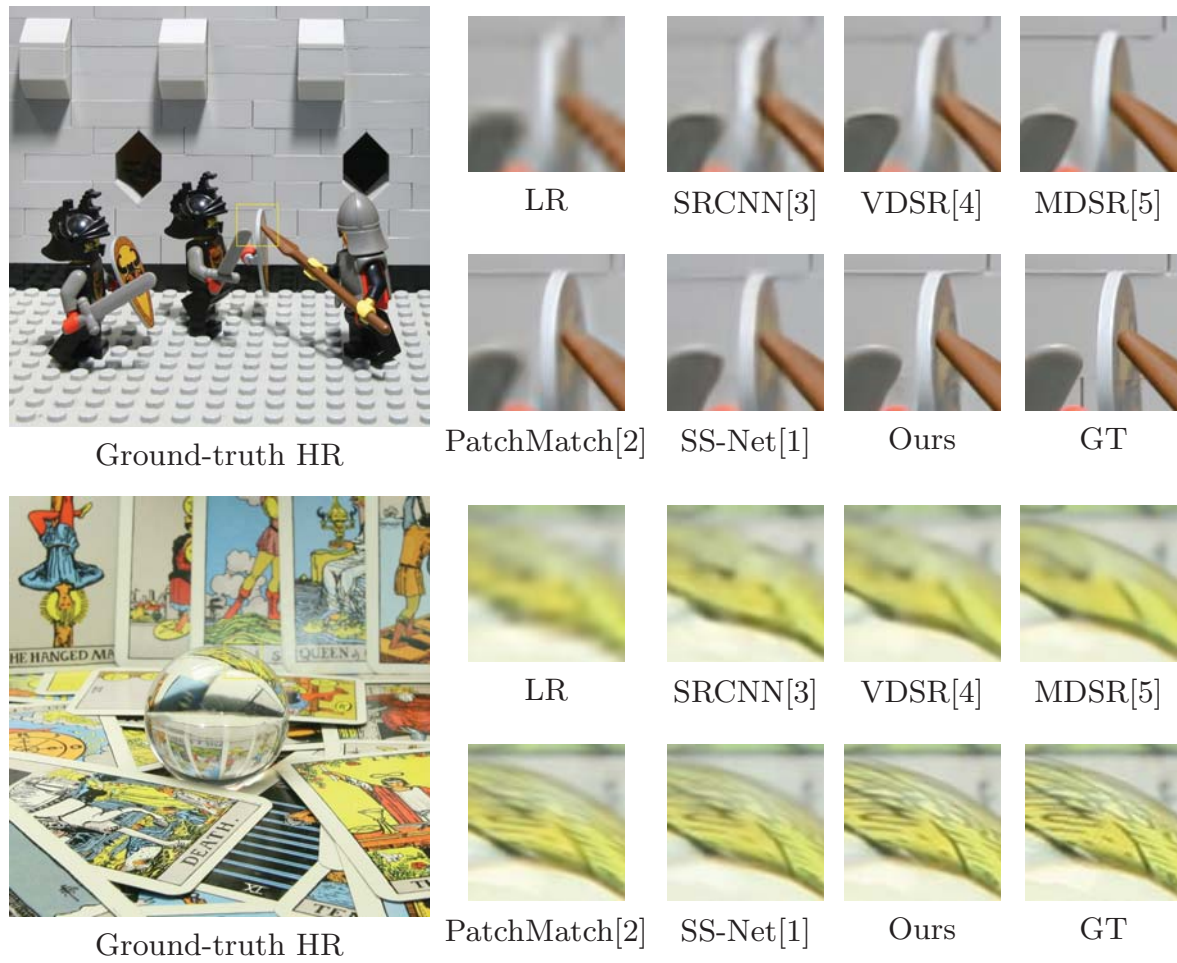
Fig. 2. Visual comparison for ×8 RefSR on Scene light field dataset. The reference images are sampled at location (0,0), where LR images are sampled at location (3,0).
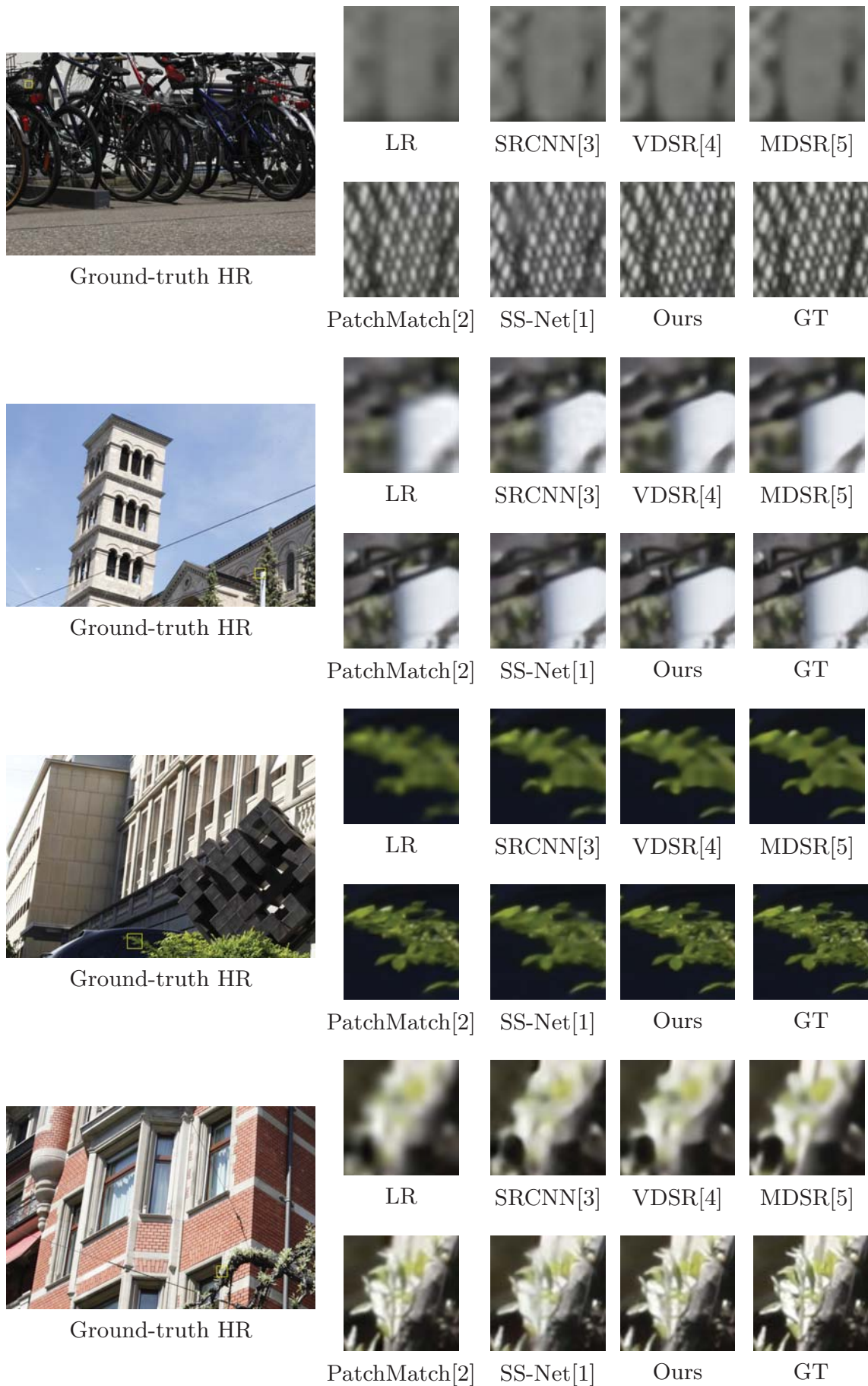
Fig. 3. Visual comparison for ×8 RefSR on Scene light field dataset. The reference images are sampled at location (0,0), where LR images are sampled at location (1,0).

References

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