

Supplementary Material for “Real-World Light Field Image Spatial Super-Resolution Via A Hybrid Imaging System” Part 2

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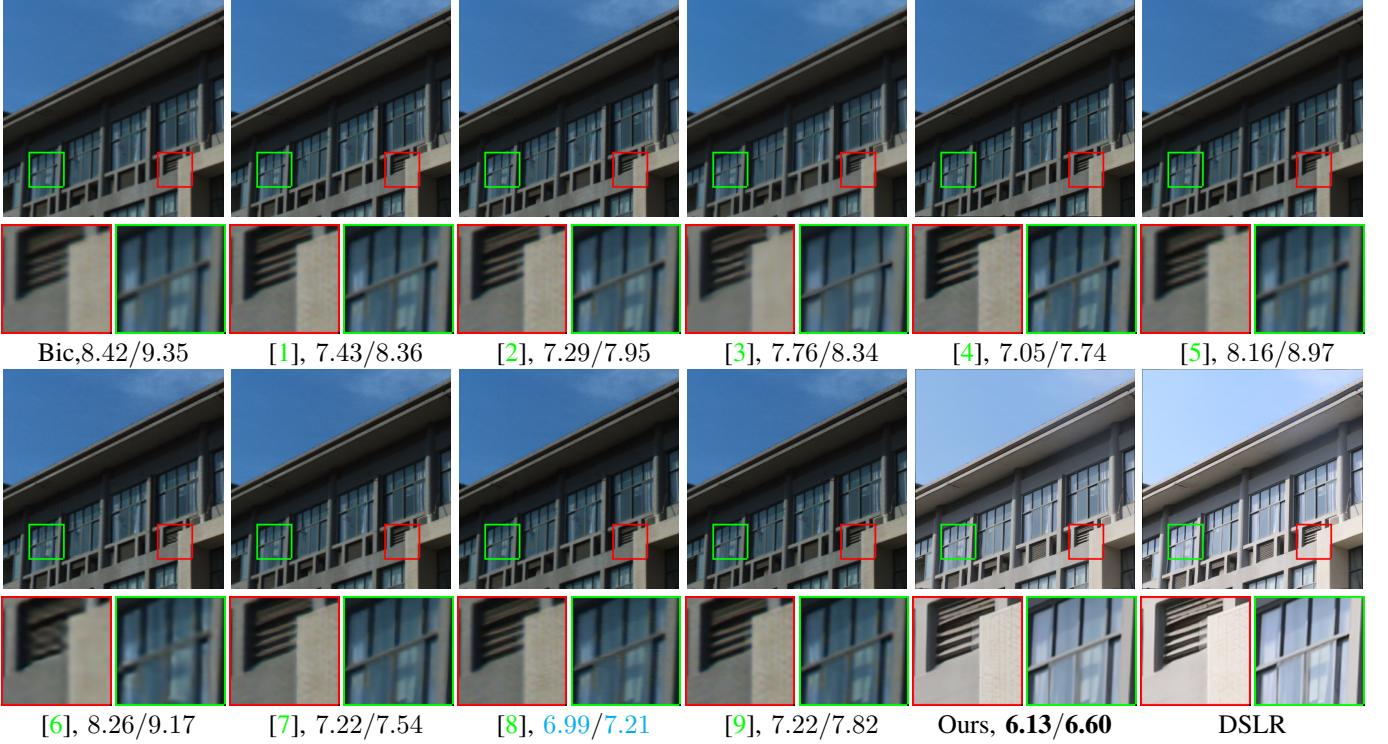


Fig. 1: Comparisons of different methods for spatially super-resolving LFIs. The super-resolution images for each method is marked with its PI/NIQE values. The best value is marked in **bold**, and the second best is marked in cyan.

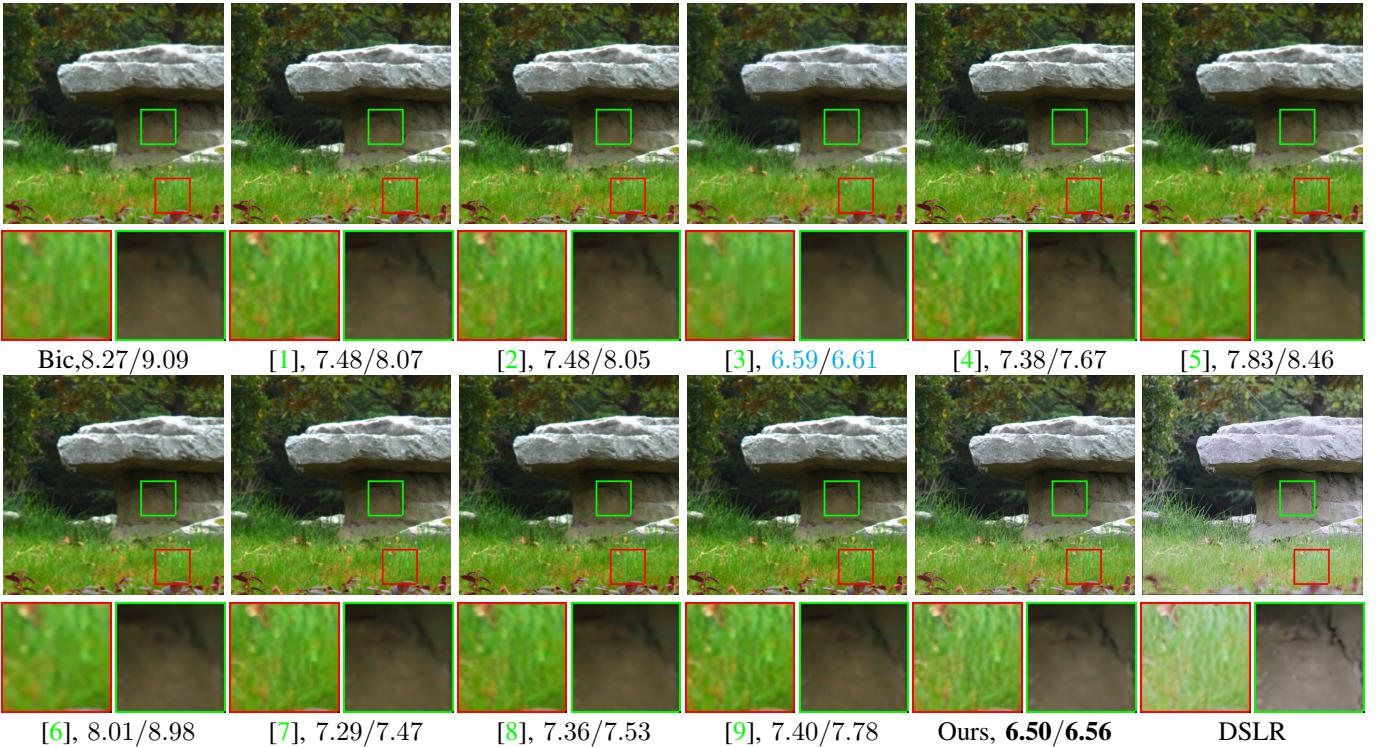


Fig. 2: Comparisons of different methods for spatially super-resolving LFIs. The super-resolution images for each method is marked with its PI/NIQE values. The best value is marked in **bold**, and the second best is marked in cyan.



Fig. 3: Comparisons of different methods for spatially super-resolving LFIs. The super-resolution images for each method is marked with its PI/NIQE values. The best value is marked in **bold**, and the second best is marked in **cyan**.

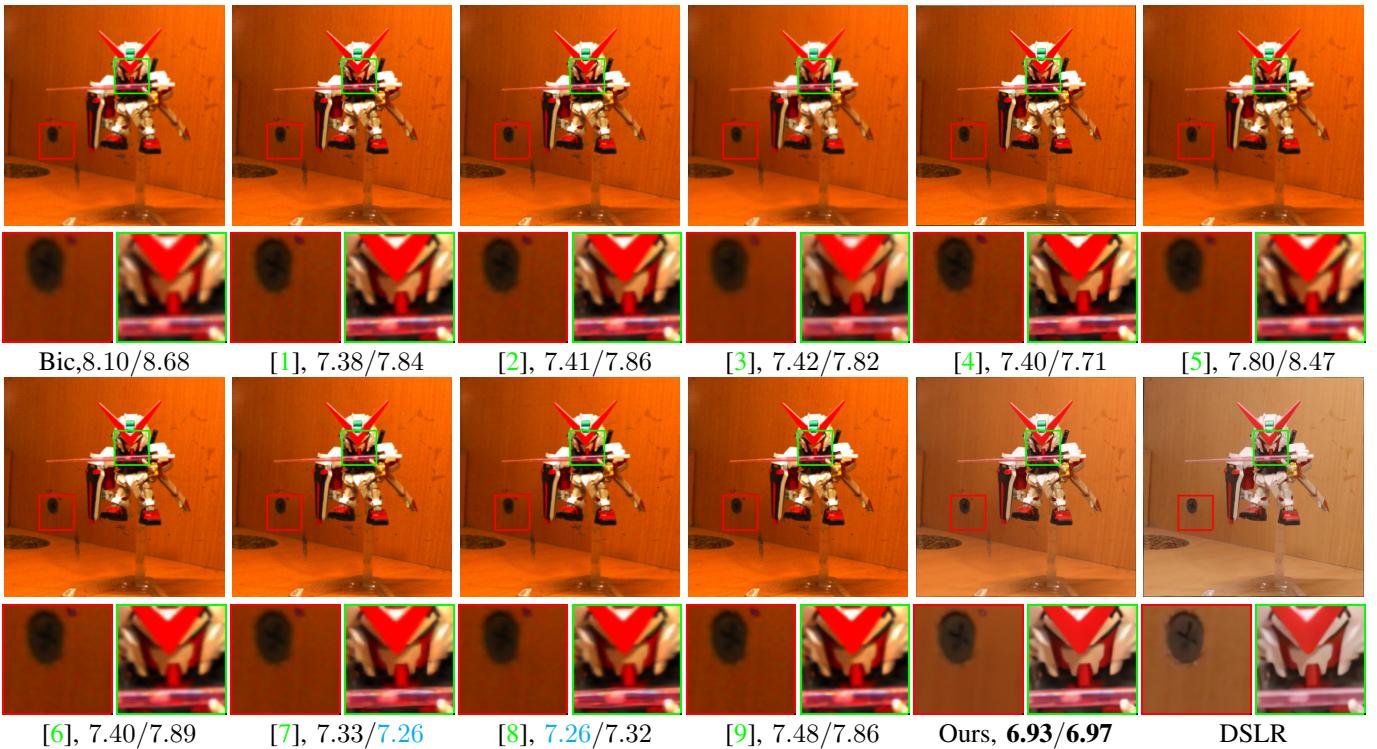


Fig. 4: Comparisons of different methods for spatially super-resolving LFIs. The super-resolution images for each method is marked with its PI/NIQE values. The best value is marked in **bold**, and the second best is marked in **cyan**.

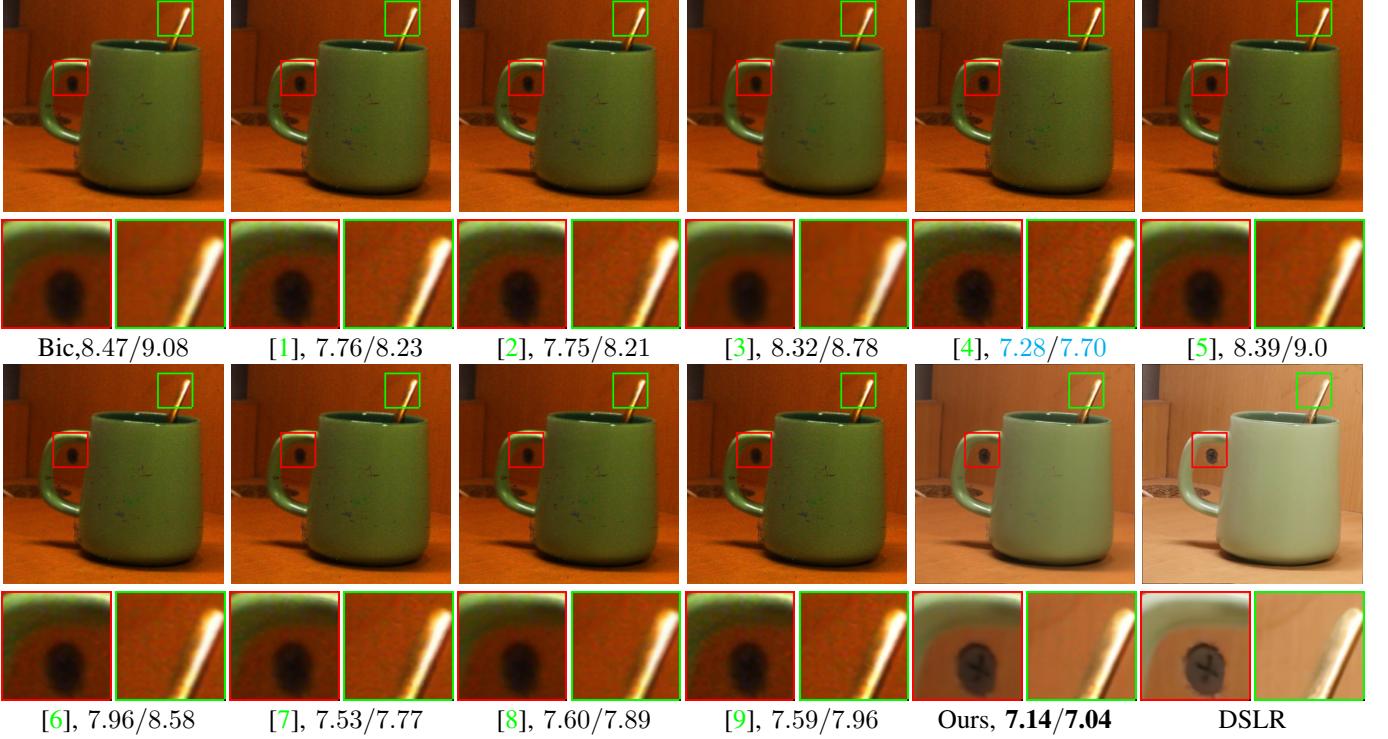


Fig. 5: Comparisons of different methods for spatially super-resolving LFIs. The super-resolution images for each method is marked with its PI/NIQE values. The best value is marked in **bold**, and the second best is marked in **cyan**.

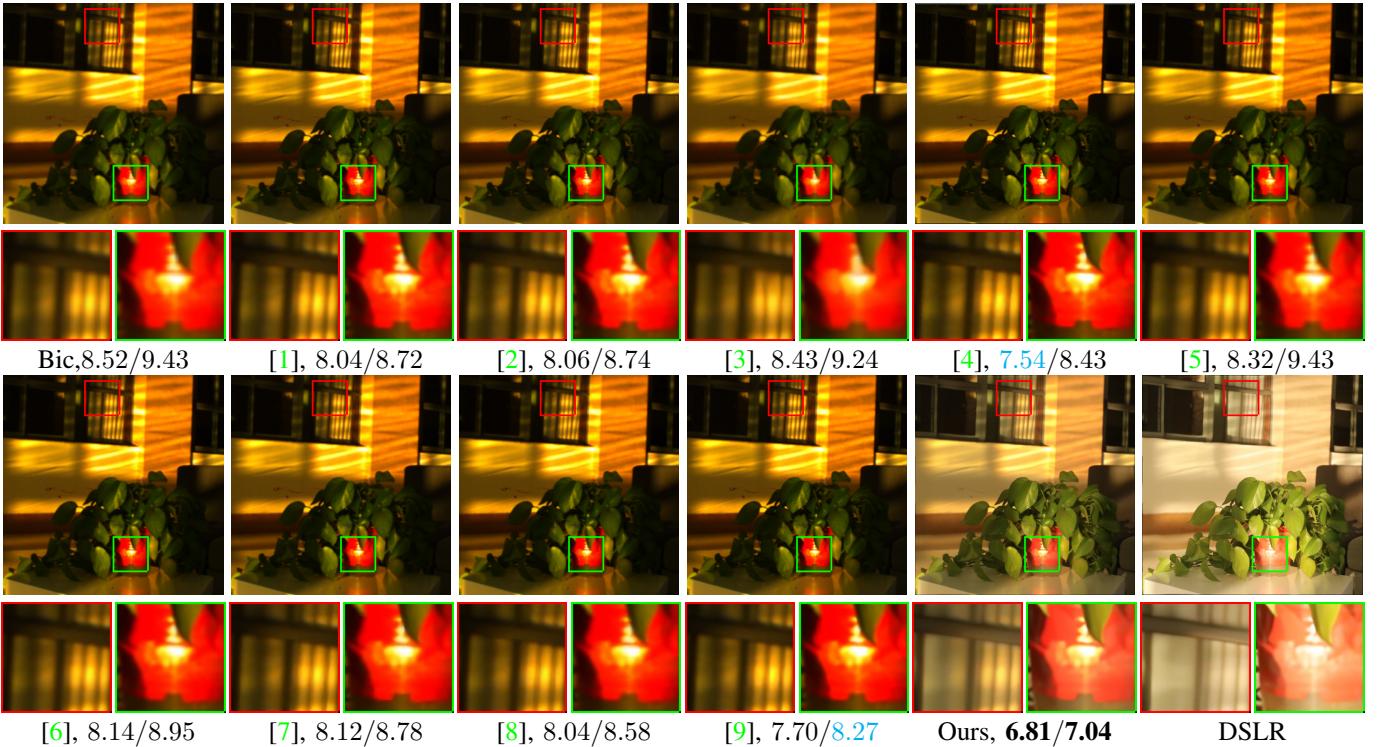


Fig. 6: Comparisons of different methods for spatially super-resolving LFIs. The super-resolution images for each method is marked with its PI/NIQE values. The best value is marked in **bold**, and the second best is marked in **cyan**.

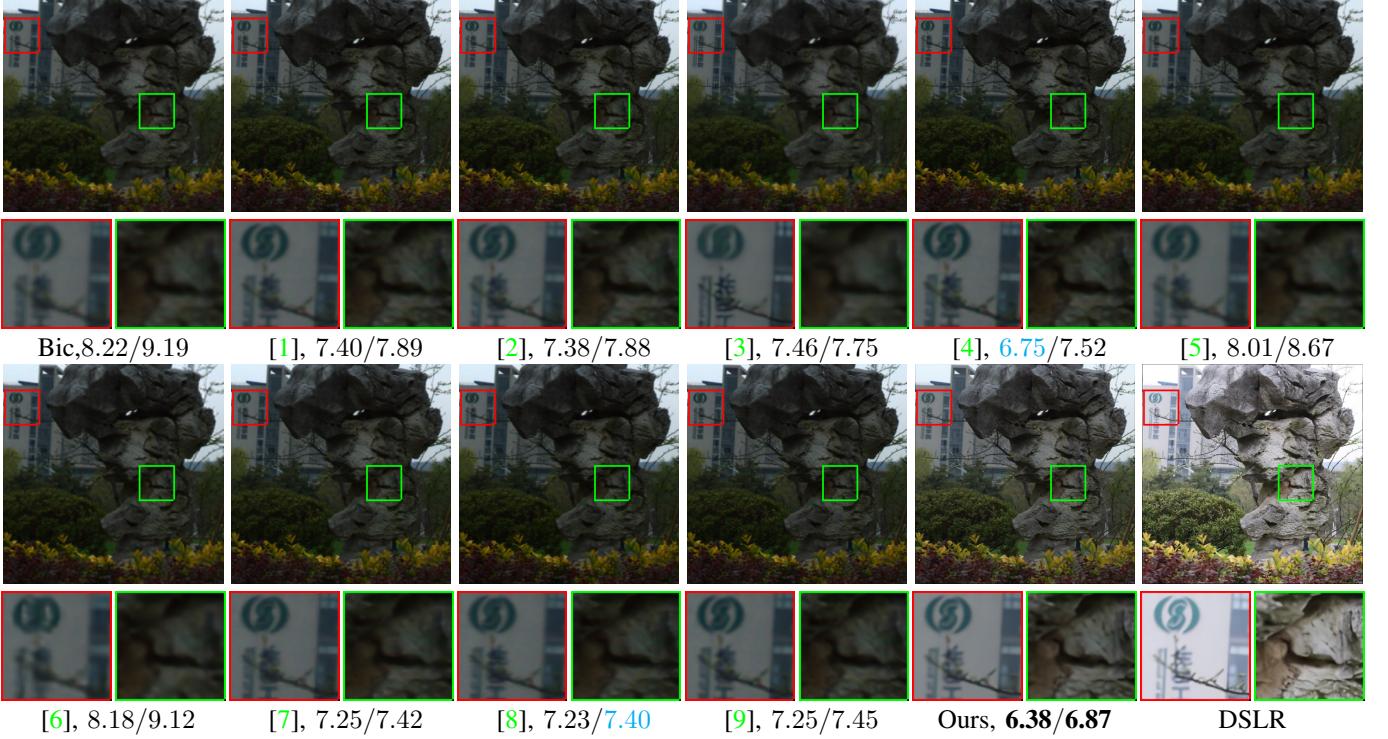


Fig. 7: Comparisons of different methods for spatially super-resolving LFI s. The super-resolution images for each method is marked with its PI/NIQE values. The best value is marked in **bold**, and the second best is marked in **cyan**.

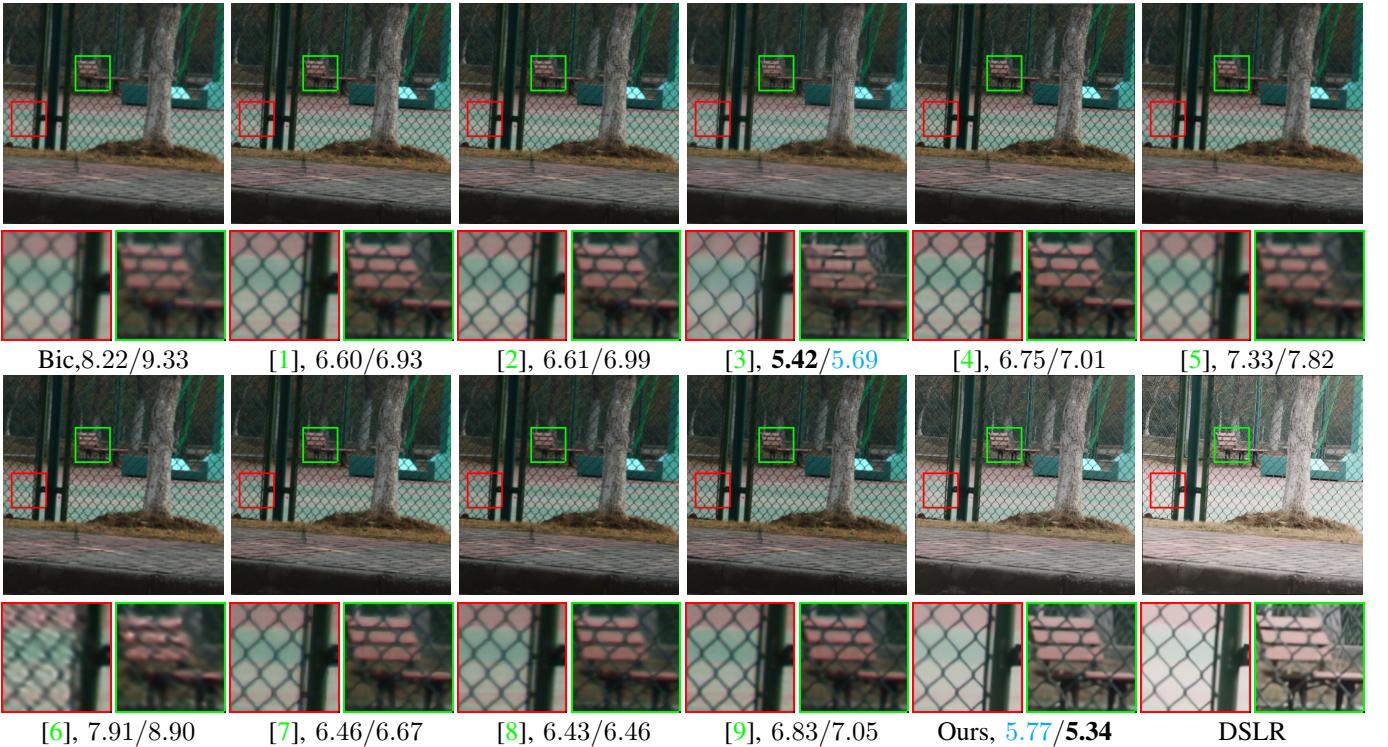


Fig. 8: Comparisons of different methods for spatially super-resolving LFI s. The super-resolution images for each method is marked with its PI/NIQE values. The best value is marked in **bold**, and the second best is marked in **cyan**.