

Spatial Attentive Single-Image Deraining with a High Quality Real Rain Dataset

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Overview. In this supplemental, we first visualize the generated attention maps to demonstrate the effectiveness of the proposed SPANet in detecting the rain regions in Section 1. We then provide additional comparisons between our SPANet and state-of-the-art single-image derainers on real rain images on the proposed benchmark and from the Internet in Sections 2 and 3, respectively. In Section 4, we provide results of the state-of-the-art video deraining methods, mode filter and our proposed background generation method on synthetic rain videos.

1. Visualization of the Attention Maps

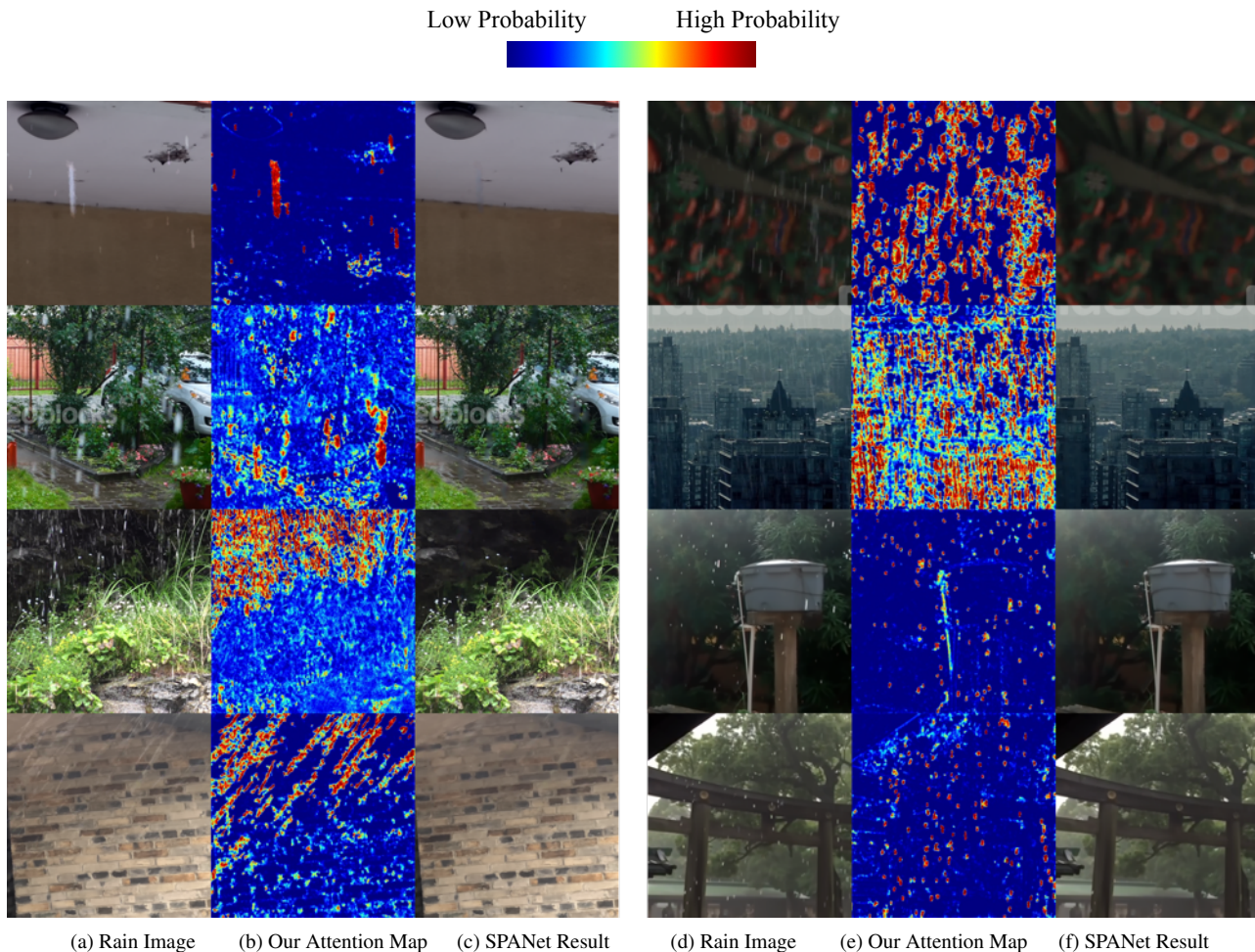


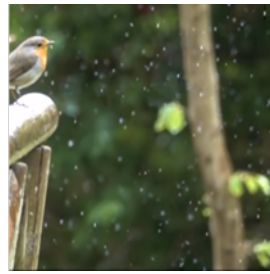
Figure 1. Visualization of the attention maps generated by our SPANet.

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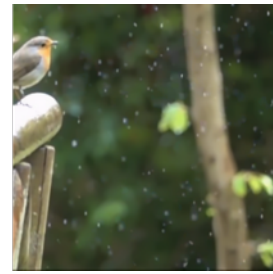
2. Comparison on Real Rain Images from the Proposed Benchmark



(a) Rain Image 29.17 / 0.9130



(b) DSC [8] 29.51 / 0.9108



(c) LP [7] 30.95 / 0.9286



(d) SILS [3] 30.44 / 0.9186



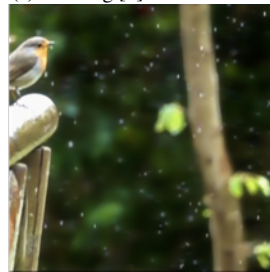
(e) Clearing [1] 28.64 / 0.9029



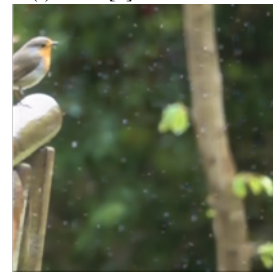
(f) DDN [2] 30.01 / 0.9207



(g) JORDER [10] 28.54 / 0.9054



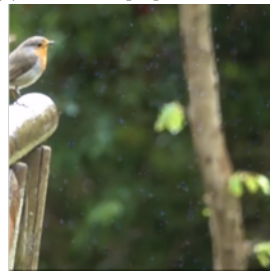
(h) DID-MDN [11] 21.27 / 0.7701



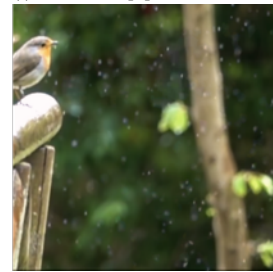
(i) RESCAN [6] 30.31 / 0.9201



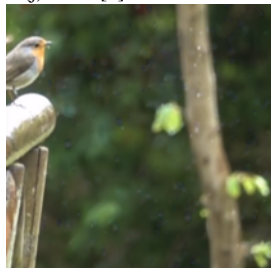
(j) DDN [2] 38.36 / 0.9668



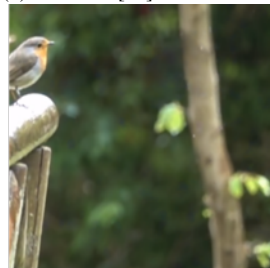
(k) JORDER [10] 33.13 / 0.9445



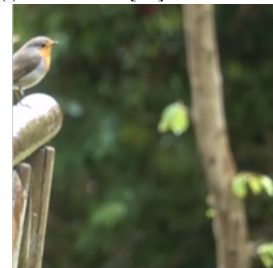
(l) DID-MDN [11] 34.57 / 0.9548



(m) RESCAN [6] 34.82 / 0.9605



(n) **Our SPANet 39.69 / 0.9874**



(o) Clean Image ∞ / 1

Figure 2. Comparison on real rain images. Methods in red were trained on the proposed dataset. PSNR/SSIM results are included for reference.



(a) Rain Image 34.51 / 0.9343



(b) DSC [8] 35.5 / 0.9366



(c) LP [7] 39.09 / 0.9755



(d) SILS [3] 39.46 / 0.9753



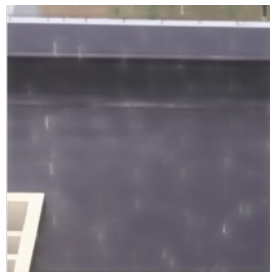
(e) Clearing [1] 35.82 / 0.9632



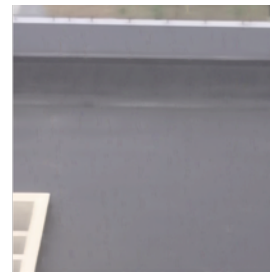
(f) DDN [2] 39.59 / 0.9747



(g) JORDER [10] 37.07 / 0.9689



(h) DID-MDN [11] 16.87 / 0.8267



(i) RESCAN [6] 38.74 / 0.9734



(j) DDN [2] 41.83 / 0.9770



(k) JORDER [10] 43.26 / 0.9835



(l) DID-MDN [11] 19.11 / 0.8844



(m) RESCAN [6] 35.14 / 0.9740



(n) **Our SPANet** 48.72 / **0.9956**



(o) Clean Image ∞ / 1

Figure 3. Comparison on real rain images. Methods in red were trained on the proposed dataset. PSNR/SSIM results are included for reference.



(a) Rain Image 32.48 / 0.9380



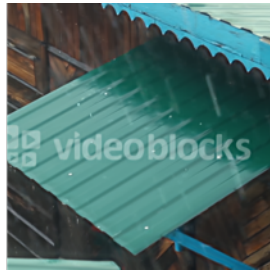
(b) DSC [8] 33.23 / 0.9374



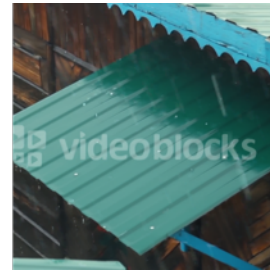
(c) LP [7] 33.44 / 0.9396



(d) SILS [3] 33.52 / 0.9359



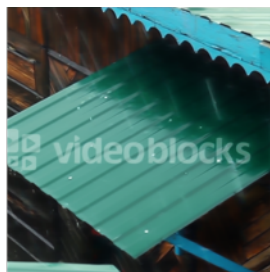
(e) Clearing [1] 31.38 / 0.9221



(f) DDN [2] 34.36 / 0.9455



(g) JORDER [10] 33.41 / 0.9265



(h) DID-MDN [11] 24.87 / 0.8372



(i) RESCAN [6] 30.75 / 0.9138



(j) DDN [2] 35.67 / 0.9541



(k) JORDER [10] 36.42 / 0.9597



(l) DID-MDN [11] 28.63 / 0.9218



(m) RESCAN [6] 34.97 / 0.9536



(n) **Our SPANet 37.78 / 0.9743**



(o) Clean Image ∞ / 1

Figure 4. Comparison on real rain images. Methods in red were trained on the proposed dataset. PSNR/SSIM results are included for reference.



(a) Rain Image 26.34 / 0.9324



(b) DSC [8] 27.17 / 0.9256



(c) LP [7] 27.02 / 0.9103



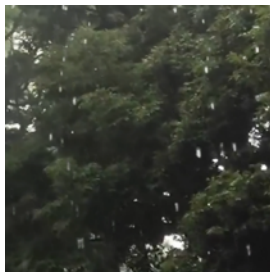
(d) SILS [3] 27.44 / 0.9060



(e) Clearing [1] 25.82 / 0.9112



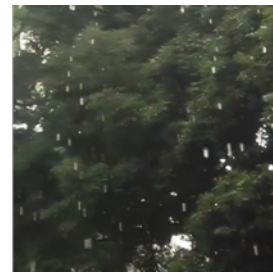
(f) DDN [2] 25.94 / 0.8965



(g) JORDER [10] 28.63 / 0.9196



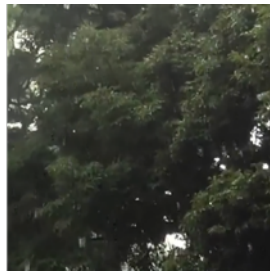
(h) DID-MDN [11] 20.27 / 0.6762



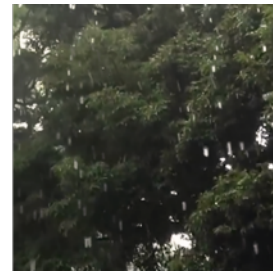
(i) RESCAN [6] 28.03 / 0.8943



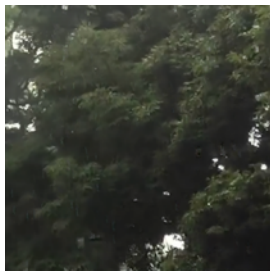
(j) DDN [2] 30.89 / 0.9272



(k) JORDER [10] 33.88 / 0.9573



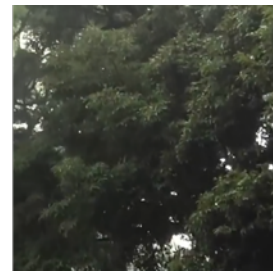
(l) DID-MDN [11] 28.21 / 0.9446



(m) RESCAN [6] 31.13 / 0.9333

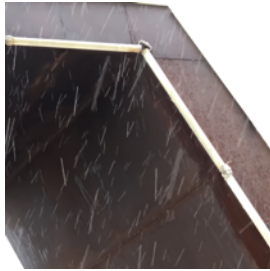


(n) **Our SPANet 38.11 / 0.9860**

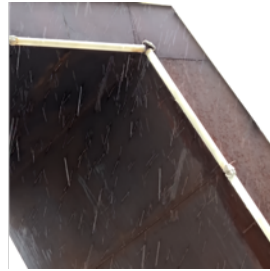


(o) Clean Image ∞ / 1

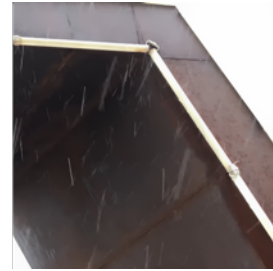
Figure 5. Comparison on real rain images. Methods in red were trained on the proposed dataset. PSNR/SSIM results are included for reference.



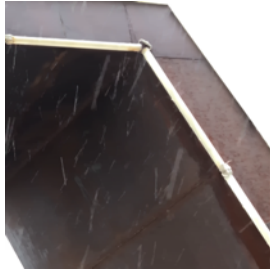
(a) Rain Image 28.87 / 0.8347



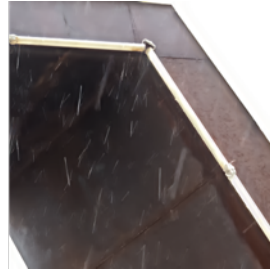
(b) DSC [8] 31.29 / 0.8714



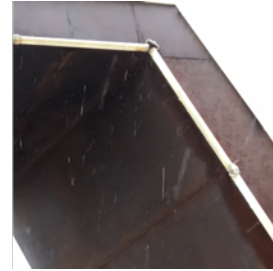
(c) LP [7] 32.83 / 0.9280



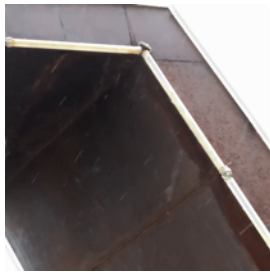
(d) SILS [3] 33.04 / 0.9251



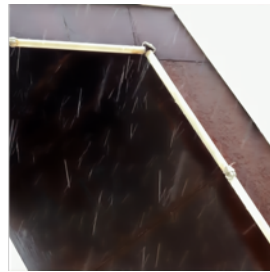
(e) Clearing [1] 30.00 / 0.8996



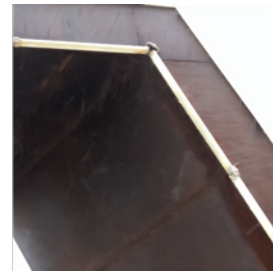
(f) DDN [2] 34.02 / 0.9341



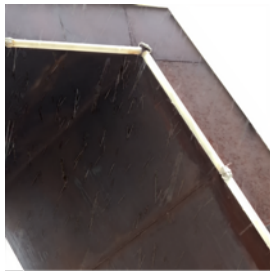
(g) JORDER [10] 33.96 / 0.9445



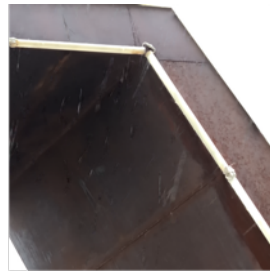
(h) DID-MDN [11] 21.97 / 0.7558



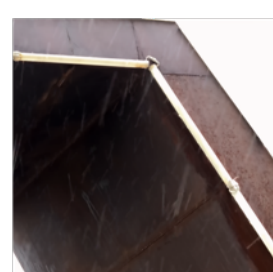
(i) RESCAN [6] 35.19 / 0.9361



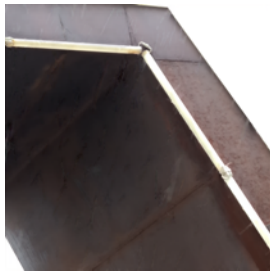
(j) DDN [2] 34.46 / 0.9239



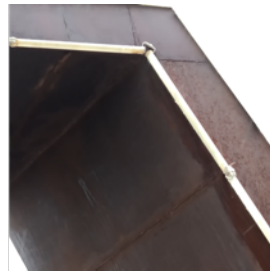
(k) JORDER [10] 36.29 / 0.9572



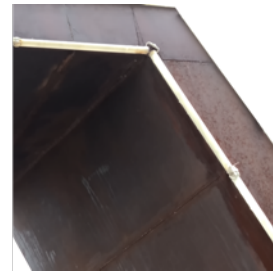
(l) DID-MDN [11] 25.02 / 0.8648



(m) RESCAN [6] 35.95 / 0.9547



(n) **Our SPANet 39.44 / 0.9866**



(o) Clean Image ∞ / 1

Figure 6. Comparison on real rain images. Methods in red were trained on the proposed dataset. PSNR/SSIM results are included for reference.



(a) Rain Image 31.03 / 0.8398



(b) DSC [8] 33.31 / 0.8636



(c) LP [7] 34.42 / 0.9186



(d) SILS [3] 35.02 / 0.9265



(e) Clearing [1] 32.36 / 0.9069



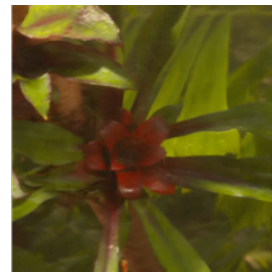
(f) DDN [2] 34.65 / 0.9135



(g) JORDER [10] 35.29 / 0.9319



(h) DID-MDN [11] 27.60 / 0.8378



(i) RESCAN [6] 30.24 / 0.8919



(j) DDN [2] 34.48 / 0.9050



(k) JORDER [10] 35.86 / 0.9384



(l) DID-MDN [11] 31.56 / 0.9128



(m) RESCAN [6] 35.73 / 0.9304



(n) **Our SPANet 38.63 / 0.9749**



(o) Clean Image ∞ / 1

Figure 7. Comparison on real rain images. Methods in red were trained on the proposed dataset. PSNR/SSIM results are included for reference.



(a) Rain Image 28.21 / 0.9228



(b) DSC [8] 28.67 / 0.9228



(c) LP [7] 29.83 / 0.9384



(d) SILS [3] 30.09 / 0.9447



(e) Clearing [1] 28.44 / 0.9274



(f) DDN [2] 29.57 / 0.9459



(g) JORDER [10] 32.67 / 0.9565



(h) DID-MDN [11] 18.89 / 0.7420



(i) RESCAN [6] 32.52 / 0.9531



(j) DDN [2] 35.24 / 0.9574



(k) JORDER [10] 35.42 / 0.9672



(l) DID-MDN [11] 22.47 / 0.8680



(m) RESCAN [6] 35.45 / 0.9674



(n) **Our SPANet** 43.83 / 0.9926



(o) Clean Image ∞ / 1

Figure 8. Comparison on real rain images. Methods in red were trained on the proposed dataset. PSNR/SSIM results are included for reference.



(a) Rain Image 31.90 / 0.9427



(b) DSC [8] 33.42 / 0.9465



(c) LP [7] 33.54 / 0.9467



(d) SILS [3] 33.14 / 0.9541



(e) Clearing [1] 31.03 / 0.9409



(f) DDN [2] 33.97 / 0.9551



(g) JORDER [10] 31.30 / 0.9416



(h) DID-MDN [11] 25.07 / 0.9365



(i) RESCAN [6] 31.20 / 0.9461



(j) DDN [2] 35.70 / 0.9636



(k) JORDER [10] 35.42 / 0.9641



(l) DID-MDN [11] 32.58 / 0.9559



(m) RESCAN [6] 34.39 / 0.9627



(n) **Our SPANet 38.28 / 0.9803**



(o) Clean Image ∞ / 1

Figure 9. Comparison on real rain images. Methods in red were trained on the proposed dataset. PSNR/SSIM results are included for reference.

3. Comparison on Real Rain Images Obtained from the Internet

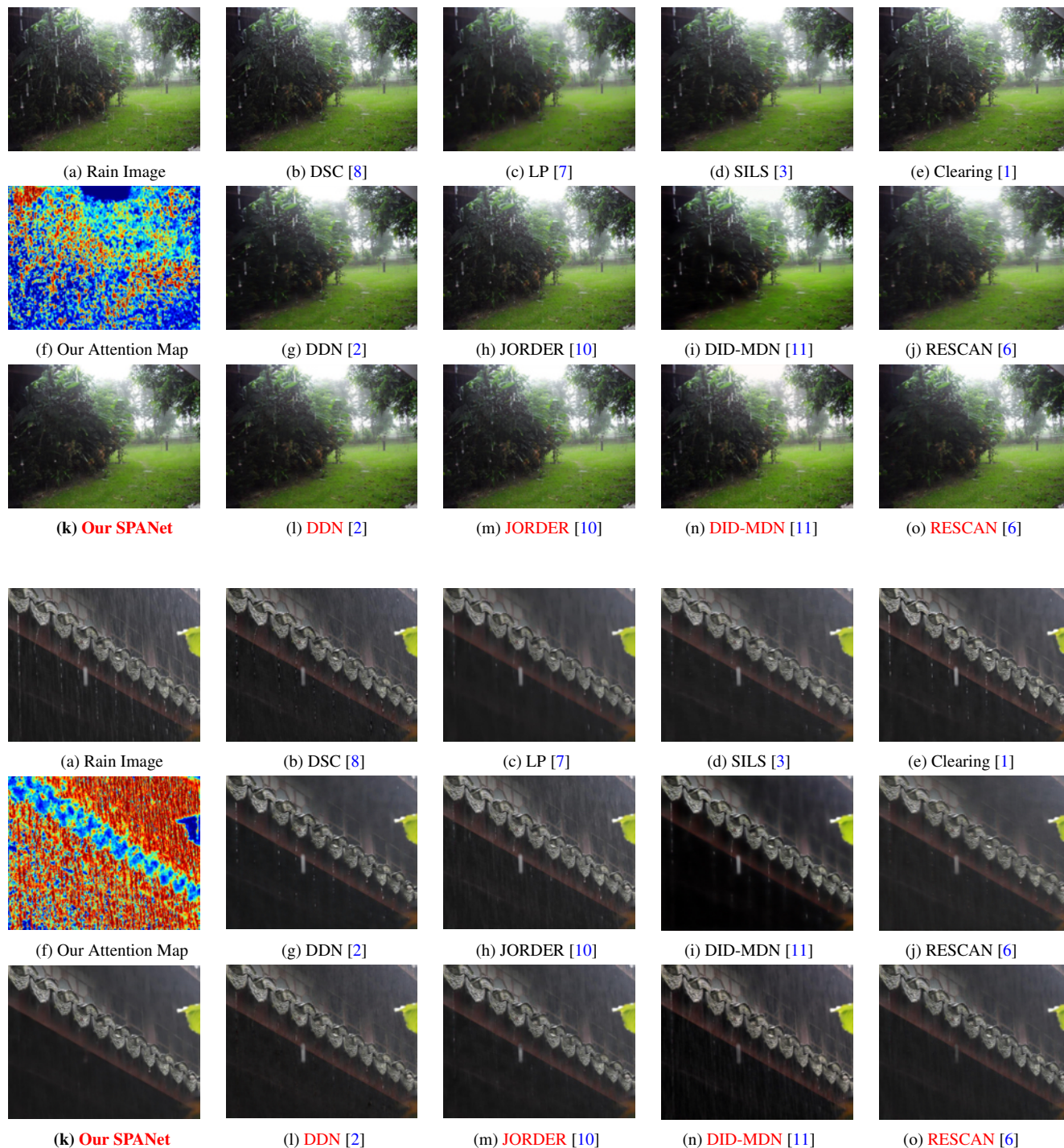


Figure 10. Comparison on real rain images obtained from the Internet. Methods in red were trained on the proposed dataset.

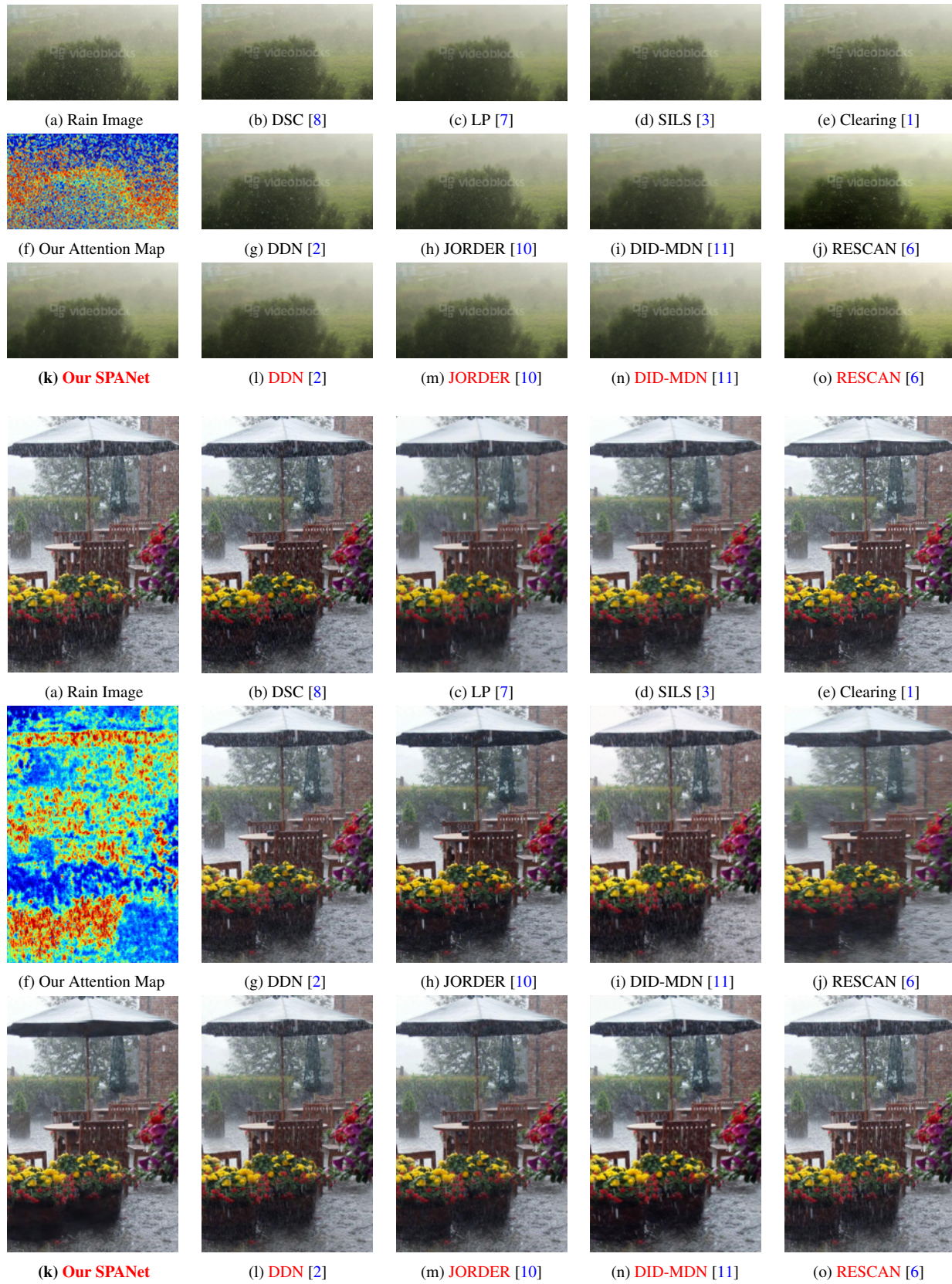


Figure 11. Comparison on real rain images obtained from the Internet. Methods in red were trained on the proposed dataset.

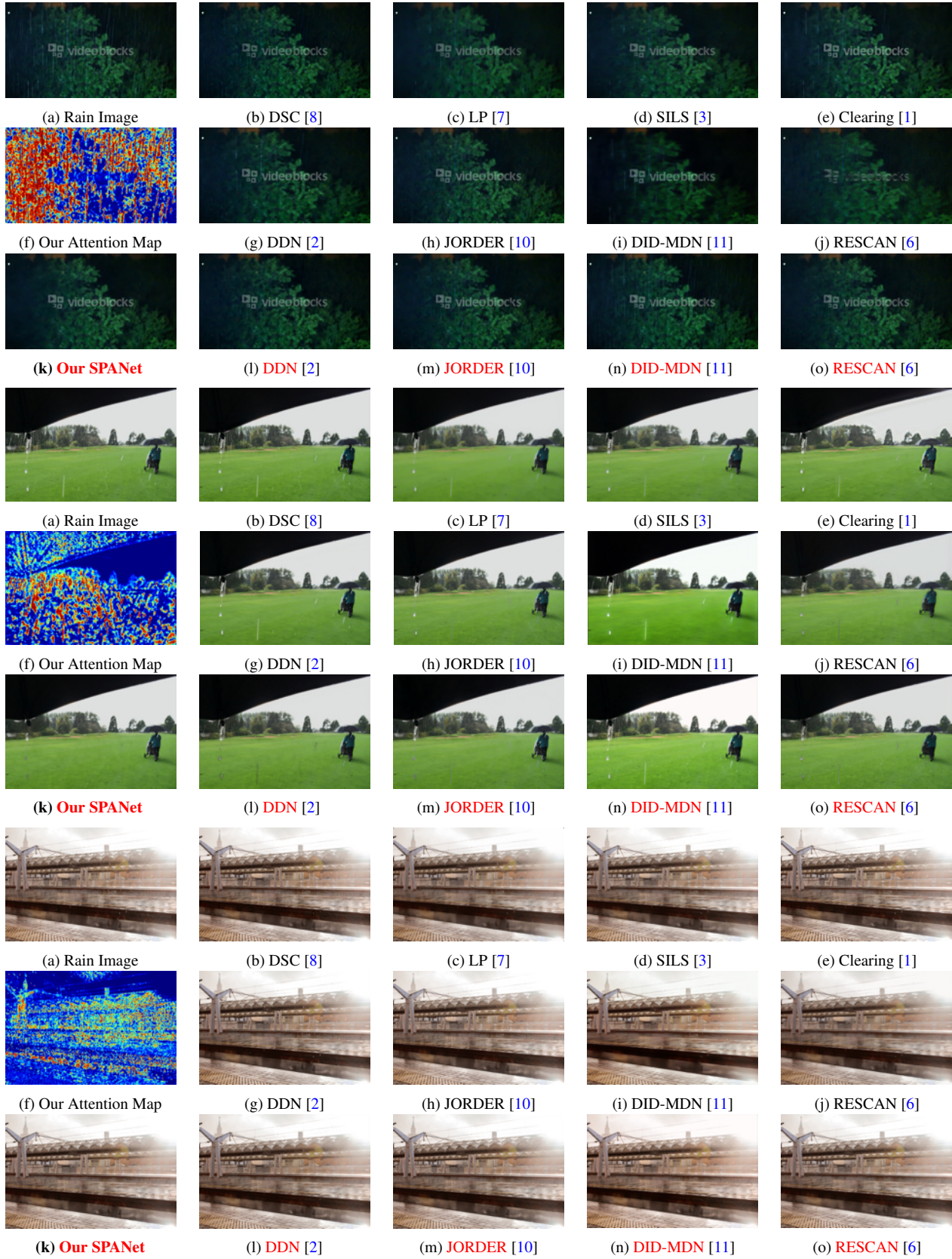
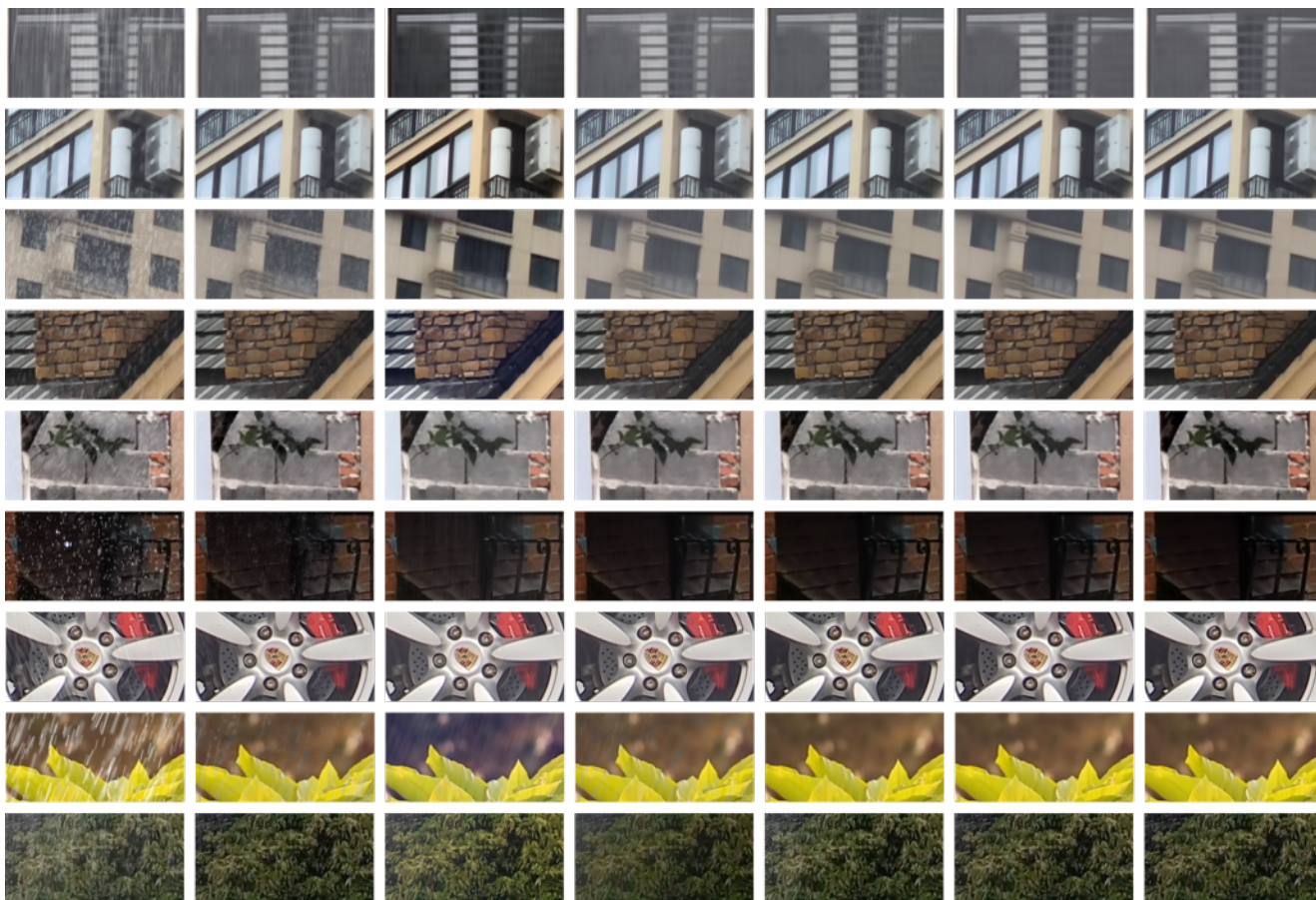


Figure 12. Comparison on real rain images obtained from the Internet. Methods in red were trained on the proposed dataset.

4. Comparison on Synthetic Rain Videos for Dataset Generation Methods



(a) Input (b) Jiang [4] (c) Wei [9] (d) Li [5] (e) Mode (f) **Ours** (g) Ground Truth

Figure 13. Comparison on synthetic rain videos of 100 frames. Our dataset generation method can get clearer background than other state-of-the-art video derain methods and mode filter.

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- [11] He Zhang and Vishal M. Patel. Density-aware single image de-raining using a multi-stream dense network. In *CVPR*, 2018. [2](#), [3](#), [4](#), [5](#), [6](#), [7](#), [8](#), [9](#), [10](#), [11](#), [12](#)