

# Supplementary Material: Semi-sparsity Priors for Image Structure Analysis and Extraction

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## 1 ADDITIONAL EXPERIMENTAL RESULTS

We further provide some additional experimental results to show the benefits of our semi-sparse image decomposition. In Fig. 1, we compare the decomposed results in cases of piece-wise constant structures with large oscillating textures. In Fig. 2 and 3, we compare the decomposed results in cases of piece-wise constant and polynomial-smoothing structures with large oscillating textures. As shown in all cases, the proposed method produces similar comparable or superior results compared with the cutting-edge methods, which is consistent with the ones in the main paper. Additionally, we also show some results in Fig. 4, where the

images are selected from our new dataset. Clearly, our semi-sparsity method is capable of decomposing image structures from a variety of textural backgrounds.

## REFERENCES

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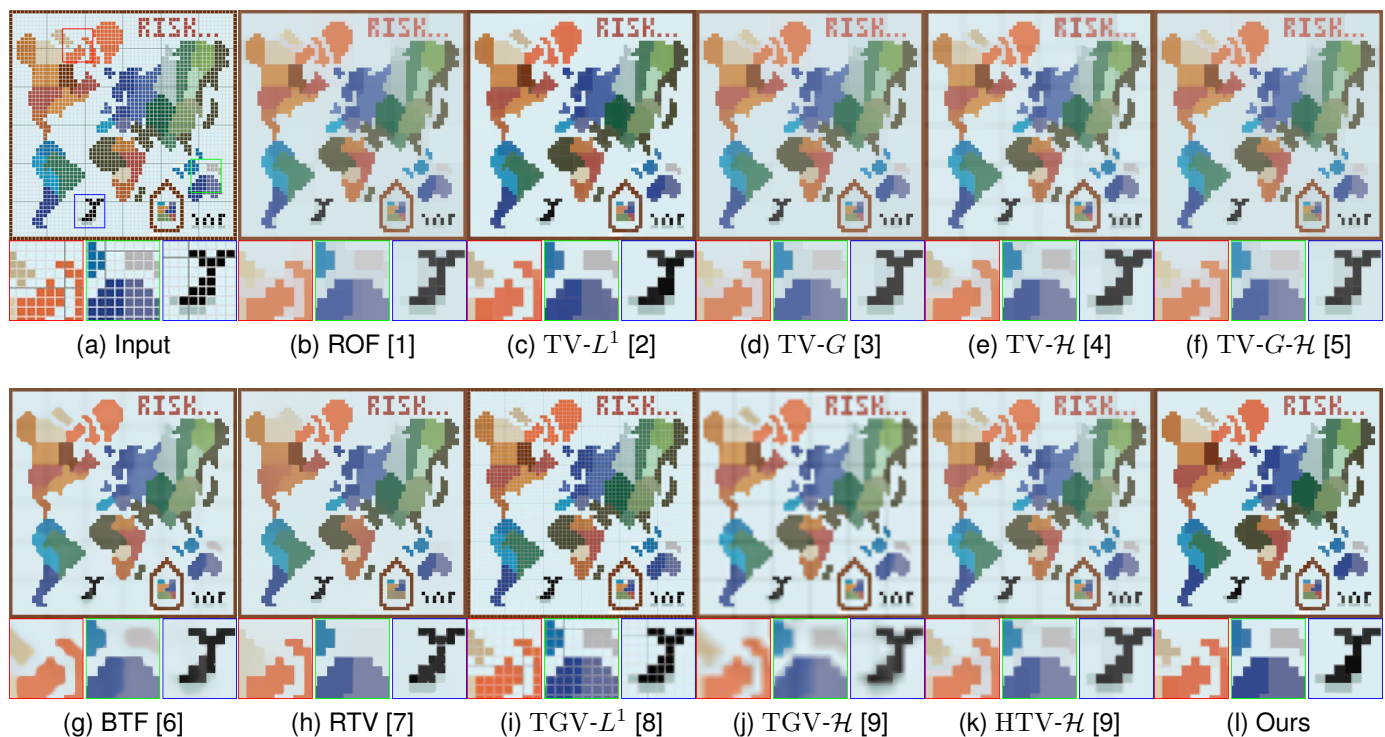


Figure 1. Visual results of image structures in case of **piece-wise constant structures** and **large oscillating textures**. (a) Input, (b) ROF ( $\lambda = 0.0045, \alpha = 0.001$ ) [1], (c)  $TV-L^1$  ( $\lambda = 0.001, \alpha = 0.002$ ) [2], (d)  $TV-G$  ( $\lambda = 0.003, \alpha = 0.0006, \gamma = 0.0004$ ) [3], (e)  $TV-H$  ( $\lambda = 0.001, \alpha = 0.01$ ) [4], (f)  $TV-G-H$  ( $\lambda = 0.004, \alpha = 0.008, \gamma = 0.005$ ) [5], (g) BTF ( $\sigma = 7.0, iter = 3$ ) [6], (h) RTV ( $\lambda = 0.012, \sigma = 3.0$ ) [7], (i)  $TGV-L^1$  ( $\lambda = 0.001, \alpha = 0.002, \beta = 0.002$ ) [8], (j)  $TGV-H$  ( $\lambda = 0.002, \alpha = 0.045, \beta = 0.025$ ) [9], (k)  $HTV-H$  ( $\lambda = 0.004, \alpha = 0.006, \beta = 0.0015$ ) [9], and (l) Ours ( $\lambda = 0.002, \alpha = 0.005, \beta = 0.001$ ). Quantitative results with the  $STR(C_0/C_1)$  metrics, (b)~(j): 22.84 (0.1908/0.0918), 22.83 (0.0550/0.1161), 22.89 (0.1771/0.0954), 22.78 (0.0392/0.0424), 22.81 (0.0416/0.0326), 22.88 (0.0271/0.0526), 22.85 (0.1371/0.0447), 22.84 (0.0077/0.1087), 22.78 (0.0370/0.1994), 22.85 (0.0507/0.0869), 23.01 (0.0774/0.0644). (Zoom in for better view.)



Figure 2. Visual results of image structures in case of **hybrid and complicated textural and structural components**. (a) Input, (b) ROF ( $\lambda = 0.005, \alpha = 0.001$ ) [1], (c)  $TV-L^1$  ( $\lambda = 0.004, \alpha = 0.009$ ) [2], (d)  $TV-G$  ( $\lambda = 0.003, \alpha = 0.0006, \gamma = 0.0004$ ) [3], (e)  $TV-H$  ( $\lambda = 0.001, \alpha = 0.01$ ) [4], (f)  $TV-G-H$  ( $\lambda = 0.004, \alpha = 0.008, \gamma = 0.005$ ) [5], (g) BTF ( $\sigma = 7.0, iter = 3$ ) [6], (h) RTV ( $\lambda = 0.015, \sigma = 4.0$ ) [7], (i)  $TGV-L^1$  ( $\lambda = 0.001, \alpha = 0.002, \beta = 0.002$ ) [8], (j)  $TGV-H$  ( $\lambda = 0.002, \alpha = 0.05, \beta = 0.02$ ) [9], (k)  $HTV-H$  ( $\lambda = 0.004, \alpha = 0.002, \beta = 0.001$ ) [9], and (l) Ours ( $\lambda = 0.006, \alpha = 0.004, \beta = 0.003$ ). Quantitative results with the  $STR(C_0/C_1)$  metrics, (b)~(j): 18.54(0.2691/0.1004), 18.35(0.0409/0.0288), 18.60(0.2406/0.0222), 18.24(0.0897/0.0354), 18.32(0.0985/0.0235), 18.48(0.0126/0.0597), 18.72(0.0118/0.0406), 18.45(0.0908/0.1693), 18.26(0.0586/0.1220), 18.30(0.0935/0.1126), 18.51(0.0103/0.0217). (Zoom in for better view.)

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Figure 3. Visual results of image structures in case of **hybrid and complicated textural and structural components**. (a) Input, (b) ROF ( $\lambda = 0.004, \alpha = 0.001$ ) [1], (c)  $TV-L^1$  ( $\lambda = 0.002, \alpha = 0.003$ ) [2], (d)  $TV-G$  ( $\lambda = 0.003, \alpha = 0.0006, \gamma = 0.0003$ ) [3], (e)  $TV-H$  ( $\lambda = 0.004, \alpha = 0.03$ ) [4], (f)  $TV-G-H$  ( $\lambda = 0.005, \alpha = 0.004, \gamma = 0.005$ ) [5], (g) BTF ( $\sigma = 5.0, iter = 6$ ) [6], (h) RTV ( $\lambda = 0.015, \sigma = 3.0$ ) [7], (i)  $TGV-L^1$  ( $\lambda = 0.004, \alpha = 0.007, \beta = 0.003$ ) [8], (j)  $TGV-H$  ( $\lambda = 0.006, \alpha = 0.05, \beta = 0.04$ ) [9], (k)  $HTV-H$  ( $\lambda = 0.005, \alpha = 0.015, \beta = 0.01$ ) [9], and (l) Ours ( $\lambda = 0.0007, \alpha = 0.0008, \beta = 0.0005$ ). Quantitative results with the  $STR(C_0/C_1)$  metrics, (b)~(j): 18.22(0.1650/0.0251), 18.15(0.0295/0.0446), 18.19(0.1473/0.0120), 18.14(0.0696/0.0432), 18.21(0.0690/0.0453), 18.19(0.0350/0.0835), 18.24(0.0959/0.0207), 18.23(0.0106/0.1038), 18.18(0.0493/0.2023), 18.23(0.0659/0.1370), 18.17(0.0321/0.0065). (Zoom in for better view.)



Figure 4. Visual results of image structures in case of an image consisting of complicated textural and structural components. (Zoom in for better view).