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# Panchromatic and Multispectral Image Fusion via Alternating Reverse Filtering Network (Supplementary Materials)

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## 1 Experimental Results

### 1.1 Quantitative comparison

We show the quantitative experiment comparisons of several measurement metrics, including the peak signal-to-noise ratio (PSNR), the structural similarity (SSIM), the spectral angle mapper (SAM), the relative dimensionless global error in synthesis (ERGAS), the correlation coefficient (SCC), Q-index, the three non-reference metrics of the spectral distortion index ( $D_\lambda$ ), the spatial distortion index ( $D_S$ ) and the quality without reference (QNR) in Table 1 and Table 2, Table 3 between our predictions and that of the baseline methods on WordView-II, GaoFen2 and WordView-III dataset, respectively. The best results are highlighted by bold. It can be clearly seen that our alternating reverse filtering network performs the best compared with other state-of-the-art methods in all the indexes, indicating the superiority of our proposed method.

### 1.2 Qualitative comparison

The qualitative visualization comparisons between our method and several state-of-the-art pan-sharpening methods are shown in Figure 1 and Figure 2 on the representative samples of the WordView-II and GaoFen2 datasets. Images in the last row are the MSE residues between the fused results and the ground truth. Compared with other competing methods, our model has minor spatial and spectral distortions. It can be easily concluded from the observation of MSE maps. As for the MSE residues, it's noticed that our proposed method is closest to the ground truth than other comparison methods. Therefore, It can be affirmed that our method achieves the best performance than other competitive pan-sharpening algorithms.

## 2 Visual Examples of Intermediate Results

The iterative procedure progressively recovers the structure details of the fused image according to the filtering effect. The intermediate visual results of alternating reverse filtering network with different iterations are shown in Figure 3, from which we can observe that more detailed information is recovered along with the greater number of iterations.

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Table 1: Quantitative comparison of nine metrics on the WordView-II dataset. Best results are highlighted by **bold**.  $\uparrow$  indicates that the larger the value, the better the performance, and  $\downarrow$  indicates that the smaller the value, the better the performance.

Methods	PSNR $\uparrow$	SSIM $\uparrow$	SAM $\downarrow$	ERGAS $\downarrow$	SCC $\uparrow$	Q $\uparrow$	$D_\lambda$ $\downarrow$	$D_S$ $\downarrow$	QNR $\uparrow$
SFIM	34.1297	0.8975	0.0439	2.3449	0.9079	0.6064	0.0915	0.1277	0.7942
GS	35.6376	0.9176	0.0423	1.8774	0.9225	0.6307	0.0607	0.1285	0.8195
Brovey	35.8646	0.9216	0.0403	1.8238	0.8913	0.6163	0.077	0.136	0.7977
IHS	35.2962	0.9027	0.0461	2.0278	0.8534	0.5704	0.0774	0.1578	0.777
GFPCA	34.5581	0.9038	0.0488	2.1411	0.8924	0.4665	0.1016	0.1656	0.7508
PNN	40.7550	0.9624	0.0259	1.0646	0.9677	0.7426	0.065	0.1186	0.825
PANNet	40.8176	0.9626	0.0257	1.0557	0.968	0.7437	0.0645	0.1189	0.8252
MSDCNN	41.3355	0.9664	0.0242	0.9940	0.9721	0.7577	0.0635	0.1172	0.8276
SRPPNN	41.4538	0.9679	0.0233	0.9899	0.9729	0.7691	0.0637	0.1164	0.8281
GPPNN	41.1622	0.9684	0.0244	1.0315	0.9722	0.7627	0.0642	0.1163	0.8278
Ours	<b>41.7587</b>	<b>0.9691</b>	<b>0.0229</b>	<b>0.9540</b>	<b>0.9749</b>	<b>0.7731</b>	<b>0.0631</b>	<b>0.1184</b>	<b>0.8285</b>

Table 2: Quantitative comparison of nine metrics on the GaoFen2 dataset. Best results are highlighted by **bold**.  $\uparrow$  indicates that the larger the value, the better the performance, and  $\downarrow$  indicates that the smaller the value, the better the performance.

Methods	PSNR $\uparrow$	SSIM $\uparrow$	SAM $\downarrow$	ERGAS $\downarrow$	SCC $\uparrow$	Q $\uparrow$	$D_\lambda$ $\downarrow$	$D_S$ $\downarrow$	QNR $\uparrow$
SFIM	36.9060	0.8882	0.0318	1.7398	0.8128	0.4349	0.0691	0.1312	0.8109
GS	37.2260	0.9034	0.0309	1.6736	0.7851	0.4211	0.0397	0.1214	0.8445
Brovey	37.7974	0.9026	0.0218	1.3720	0.6446	0.3857	0.0905	0.1443	0.779
IHS	38.1754	0.9100	0.0243	1.5336	0.6738	0.3682	0.0418	0.1345	0.8301
GFPCA	37.9443	0.9204	0.0314	1.5604	0.8032	0.3236	0.0898	0.1815	0.7445
PNN	43.1208	0.9704	0.0172	0.8528	0.9400	0.739	0.0387	0.1162	0.8494
PANNet	43.0659	0.9685	0.0178	0.8577	0.9402	0.7309	0.0369	0.1219	0.8455
MSDCNN	45.6874	0.9827	0.0135	0.6389	0.9526	0.7759	0.0368	0.1112	0.8560
SRPPNN	47.1998	0.9877	0.0106	0.5586	0.9564	0.7900	0.0364	0.1087	0.8588
GPPNN	44.2145	0.9815	0.0137	0.7361	0.9510	0.7721	0.0350	0.1078	0.8612
Ours	<b>47.2238</b>	<b>0.9892</b>	<b>0.0102</b>	<b>0.5495</b>	<b>0.9602</b>	<b>0.8026</b>	<b>0.0361</b>	<b>0.1012</b>	<b>0.8665</b>

Table 3: Quantitative comparison of nine metrics on the WordView-III dataset. Best results are highlighted by **bold**.  $\uparrow$  indicates that the larger the value, the better the performance, and  $\downarrow$  indicates that the smaller the value, the better the performance.

Methods	PSNR $\uparrow$	SSIM $\uparrow$	SAM $\downarrow$	ERGAS $\downarrow$	SCC $\downarrow$	Q $\uparrow$	$D_\lambda$ $\downarrow$	$D_S$ $\downarrow$	QNR $\uparrow$
SFIM	21.8212	0.5457	0.1208	8.9730	0.6952	0.4531	0.0448	0.1265	0.8347
GS	22.5608	0.5470	0.1217	8.2433	0.7131	0.4411	<b>0.0350</b>	0.2011	0.7695
Brovey	22.5060	0.5466	0.1159	8.2331	0.7033	0.4394	0.0481	0.2006	0.7603
IHS	22.5579	0.5354	0.1266	8.3616	0.6994	0.4301	0.0356	0.2073	0.7634
GFPCA	22.3344	0.4826	0.1294	8.3964	0.6987	0.3115	0.0528	0.1214	0.8325
PNN	29.9418	0.9121	0.0824	3.3206	0.954	0.8679	0.046	0.0933	0.8654
PANNet	29.6840	0.9072	0.0851	3.4263	0.9512	0.8631	0.0474	0.0942	0.8634
MSDCNN	30.3038	0.9184	0.0782	3.1884	0.9577	0.8763	0.0432	0.0877	0.8732
SRPPNN	30.4346	0.9202	0.0770	3.1553	0.9581	0.8776	0.0414	0.0909	0.8719
GPPNN	30.1785	0.9175	0.0776	3.2593	0.9569	0.8739	0.0438	0.0936	0.8671
Ours	<b>30.5425</b>	<b>0.9216</b>	<b>0.0768</b>	<b>3.1049</b>	<b>0.9567</b>	<b>0.8803</b>	0.0412	<b>0.0872</b>	<b>0.8757</b>

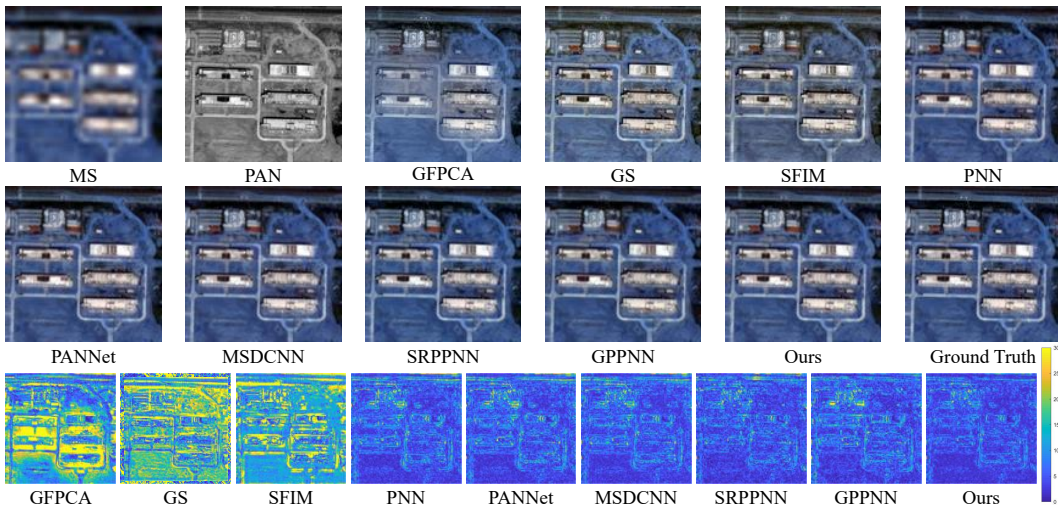


Figure 1: Qualitative visualization comparison of our method with other representative counterparts on a typical satellite image pair from the WordView-II dataset. Images in the last row visualizes the MSE between the fused images and the ground truth.

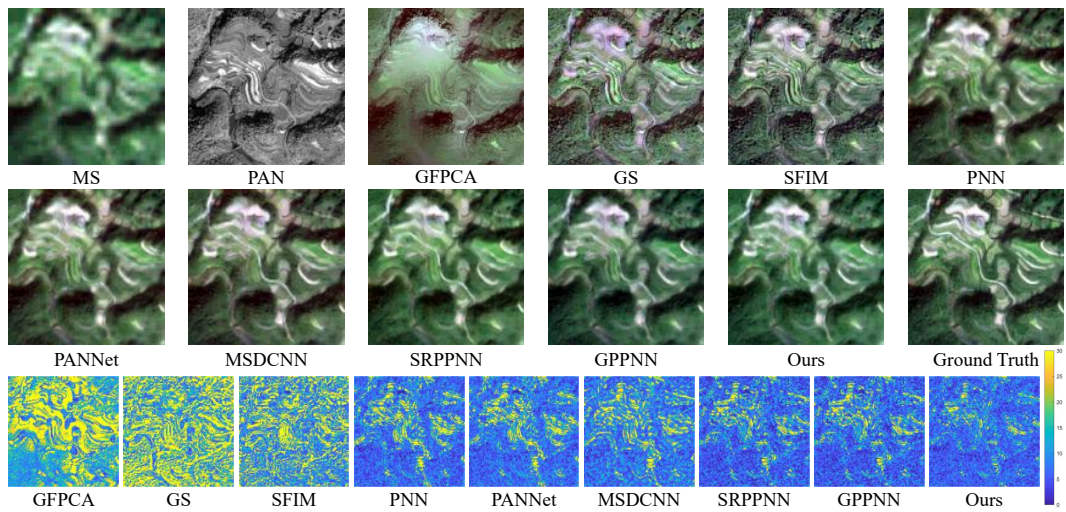


Figure 2: Qualitative visualization comparison of our method with other representative counterparts on a typical satellite image pair from the GaoFen2 dataset. Images in the last row visualizes the MSE between the fused images and the ground truth.

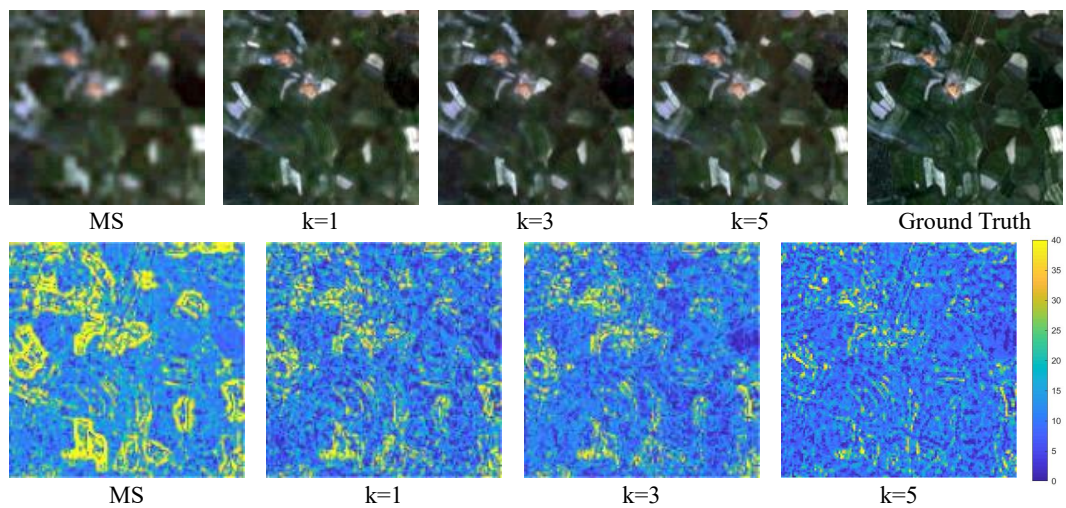


Figure 3: Intermediate visual results of different alternate iterations of ARFNet on WorldView-II. The last row visualizes the MSE residues between the fused results and the ground truth.