

Supplementary File for Conjugate Momentum Quadratic Penalty Alternating Minimization for Total Variation Image Restoration

Ong Yin Ren
Faculty of Computing
Universiti Teknologi Malaysia
81310, Skudai, Johor, Malaysia
yinren-2000@graduate.utm.my

Tarmizi Adam
Faculty of Computing
Universiti Teknologi Malaysia
81310, Skudai, Johor, Malaysia
tarmizi.adam@utm.my

Nur Syarafina Mohamed
Department of Mathematical Sciences
Faculty of Sciences
Universiti Teknologi Malaysia
81310, Skudai, Johor, Malaysia
nursyarafina@utm.my

Mohd Fikree Hassan
School of Information Technology,
Monash University Malaysia,
47500 Subang Jaya, Selangor, Malaysia
mohdfikree.hassan@monash.edu

Pang Yee Yong
Faculty of Computing
Universiti Teknologi Malaysia
81310, Skudai, Johor, Malaysia
yeeyongpang@utm.my

I. TEST IMAGES

This section serves as an appendix, providing details and results that complement the main file. Figure 1 shows the tested images used for the experiments that discussed in the subsection V-A.



(a) *Foot* (512 × 512) (b) *Pepper* (512 × 512)



(c) *Cameraman* (256 × 256)

Fig. 1: Test images used

II. TABLES

Table I, provided here presents the comparative results between QPAM, AQPAM, and CG-QPAM under Gaussian and Poisson noise conditions, as discussed in subsection V-C of the main paper.

III. RESTORED IMAGES

Figure 2 shows the results of the SSIM between QPAM and CG-QPAM for BSNR 30 that discussed in V-C-1.

IV. FIGURES OF CG-QPAM CONVERGENCE

Figure 3 shows the effects of convergence of CG-QPAM if the parameter α are modified that discussed in the V-C-1.

BSNR	Image	Method	PSNR	SSIM	Time (s)	Itr
20	<i>Pepper</i>	QPAM	29.665	0.799	3.337	145
		AQPAM	29.282	0.789	2.812	116
		CG-QPAM	29.829	0.812	2.763	103
	<i>Foot</i>	QPAM	27.690	0.725	5.726	253
		AQPAM	27.132	0.701	3.966	174
		CG-QPAM	27.720	0.732	5.965	259
	<i>Cameraman</i>	QPAM	24.953	0.749	0.901	197
		AQPAM	24.734	0.751	0.619	137
		CG-QPAM	24.976	0.775	1.010	185
30	<i>Pepper</i>	QPAM	32.332	0.841	2.746	109
		AQPAM	31.951	0.838	3.469	98
		CG-QPAM	31.058	0.847	1.968	74
	<i>Foot</i>	QPAM	29.997	0.831	4.475	192
		AQPAM	29.439	0.829	3.223	144
		CG-QPAM	29.076	0.874	3.616	149
	<i>Cameraman</i>	QPAM	27.149	0.828	0.886	159
		AQPAM	27.131	0.828	0.501	119
		CG-QPAM	25.526	0.802	0.482	115
40	<i>Pepper</i>	QPAM	34.456	0.877	1.890	77
		AQPAM	34.090	0.875	1.766	73
		CG-QPAM	30.513	0.821	0.456	11
	<i>Foot</i>	QPAM	32.787	0.905	3.707	153
		AQPAM	32.155	0.904	2.532	112
		CG-QPAM	29.953	0.904	1.386	30
	<i>Cameraman</i>	QPAM	30.007	0.890	0.726	141
		AQPAM	29.992	0.890	0.387	98
		CG-QPAM	25.076	0.765	0.088	17

(a) Gaussian Noise

Pixel	Image	Method	PSNR	SSIM	Time (s)	Itr
255	<i>Pepper</i>	QPAM	27.584	0.761	9.394	170
		AQPAM	27.113	0.750	4.658	154
		CG-QPAM	24.968	0.734	0.301	9
	<i>Foot</i>	QPAM	27.123	0.876	7.434	279
		AQPAM	26.484	0.870	5.633	194
		CG-QPAM	24.833	0.836	0.251	5
	<i>Cameraman</i>	QPAM	23.849	0.720	1.079	211
		AQPAM	23.704	0.720	0.847	166
		CG-QPAM	24.833	0.836	0.251	5
400	<i>Pepper</i>	QPAM	24.075	0.670	4.881	174
		AQPAM	23.534	0.656	4.372	158
		CG-QPAM	21.212	0.637	0.199	7
	<i>Foot</i>	QPAM	23.815	0.865	7.738	285
		AQPAM	23.061	0.858	5.930	212
		CG-QPAM	20.945	0.810	0.147	5
	<i>Cameraman</i>	QPAM	20.358	0.663	1.106	224
		AQPAM	20.225	0.654	0.785	175
		CG-QPAM	17.648	0.588	0.020	4

(b) Poisson Noise

TABLE I: Comparison of methods for *Pepper*, *Foot*, and *Cameraman* images with two different noise types.

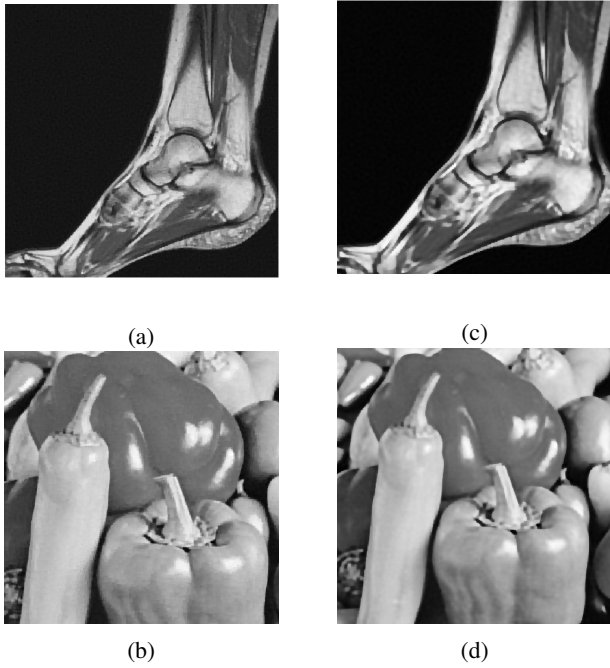
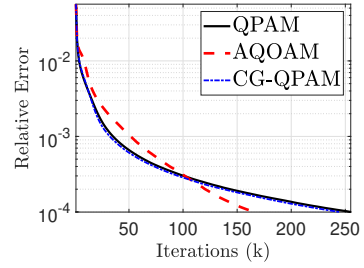
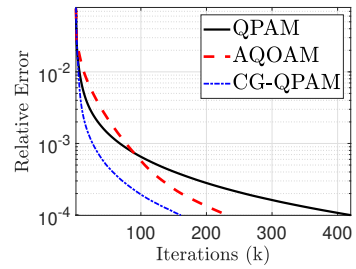


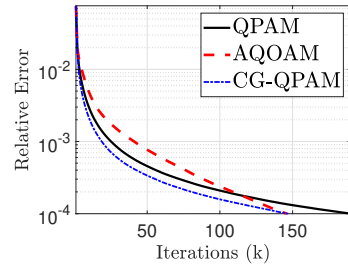
Fig. 2: Restored *Foot* and *Pepper* images at 30 BSNR value with 2 different methods. (a) and (b) are the QPAM restored images. (c) and (d) are the CG-QPAM restored images.



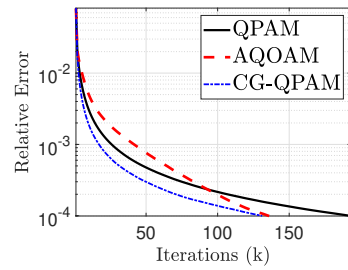
(a)



(b)



(c)



(d)

Fig. 3: (a) and (b) BSNR=20. (c) and (d) BSNR=30. The λ in (a) is 0.9 and (b) is 4.0. The λ in (c) is 5.2 and (d) is 7.0. The value of λ can improve the convergence of the CG-QPAM