

MATH473 HW6

Jiasen Zhang

Primal-Dual Algorithm

The problem is:

$$\min_x \frac{1}{2} \|Ax - b\|_2^2 + \alpha \|Wx\|_1 \quad A = PF$$

The corresponding primal-dual problem is:

$$\min_x \max_p \frac{1}{2} \|Ax - b\|_2^2 + \alpha \langle Wx, p \rangle$$

First solve the x-sub problem:

$$\begin{aligned} \min_x \frac{\tau}{2} \|Ax - b\|_2^2 + \alpha \tau \langle Wx, p \rangle + \frac{1}{2} \|x - x^k\|_2^2 \\ \alpha \tau \langle Wx, p \rangle = \alpha \tau \langle x, W^*p \rangle \end{aligned}$$

$$\begin{aligned} \tau A^*(Ax - b) + \alpha \tau W^*p + x - x^k &= 0 \\ (\tau A^*A + I)x &= \tau A^*b - \alpha \tau W^*p + x^k \\ F(\tau F^*P^*PF + I)F^*Fx &= \tau FF^*P^*b + F(-\alpha \tau W^*p + x^k) \\ (\tau P^*P + I)Fx &= \tau P^*b + F(-\alpha \tau W^*p + x^k) \end{aligned}$$

Then solve the p-sub problem:

$$\begin{aligned} \max_p \alpha \sigma \langle Wx, p \rangle - \frac{1}{2} \|p - p^k\|_2^2 \\ \alpha \sigma Wx - (p - p^k) &= 0 \\ p' &= p^k + \alpha \sigma Wx \\ p_{ij} &= \frac{p'_{ij}}{\max(1, |p'_{ij}|)} \end{aligned}$$

1 Radial sampling mask

I set $\tau = \frac{1e-3}{\alpha}$ and $\sigma = 10000 * \tau$ with which the result performs best. From Table 1 we can see $\alpha = 1e-5$ corresponds to the largest SNR and smallest relative error.

α	SNR	Relative Error
1e-3	39.3987	1.0717e-2
1e-4	41.3587	8.5519e-3
1e-5	41.3817	8.5293e-3
1e-6	41.3814	8.5296e-3
1e-7	41.3708	8.5400e-3

Table 1: α vs. SNR and relative error

The result with $\alpha = 1e-5$ is shown in Figure 1.

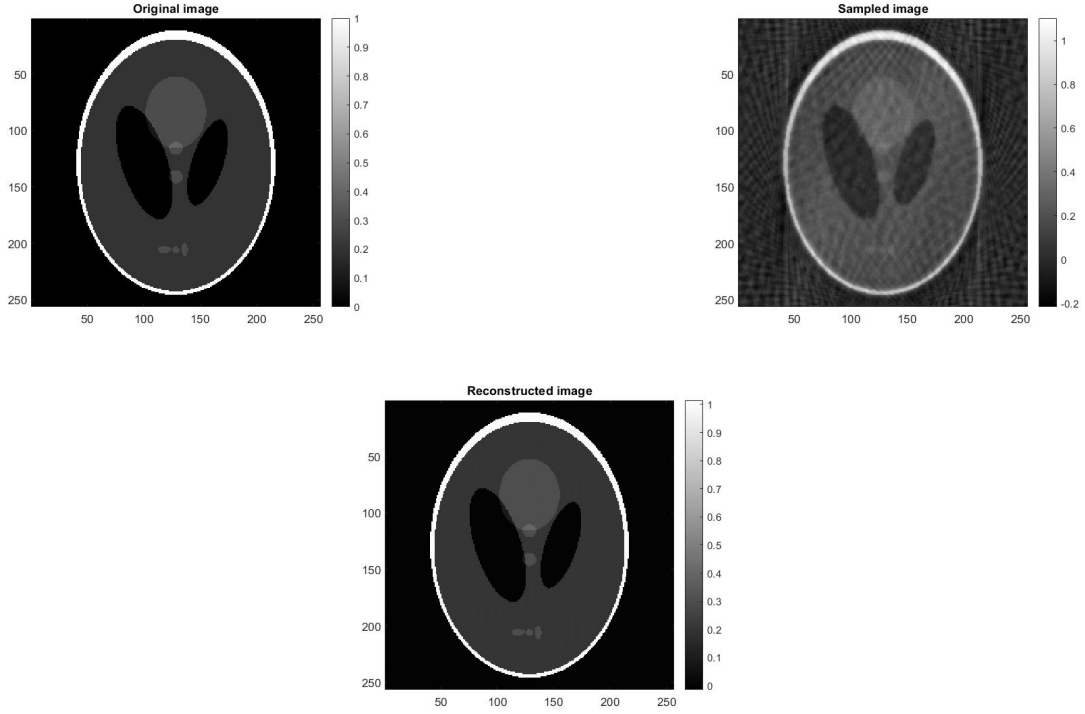


Figure 1: Reconstruction of radial sampling mask.

2 Random sampling mask

2.1 $c=20$

Now use a random sampling mask choosing $c\%$ of data outside a small square. Here $c = 20$. From Table 2 we can see $\alpha = 1e - 6$ corresponds to the largest SNR and smallest relative error.

α	SNR	Relative Error
1e-3	27.8095	4.0694e-2
1e-4	45.7048	5.1851e-3
1e-5	46.0436	4.9868e-3
1e-6	46.1541	4.9237e-3
1e-7	45.7131	5.1802e-3

Table 2: α vs. SNR and relative error

The result with $\alpha = 1e - 6$ is shown in Figure 2.

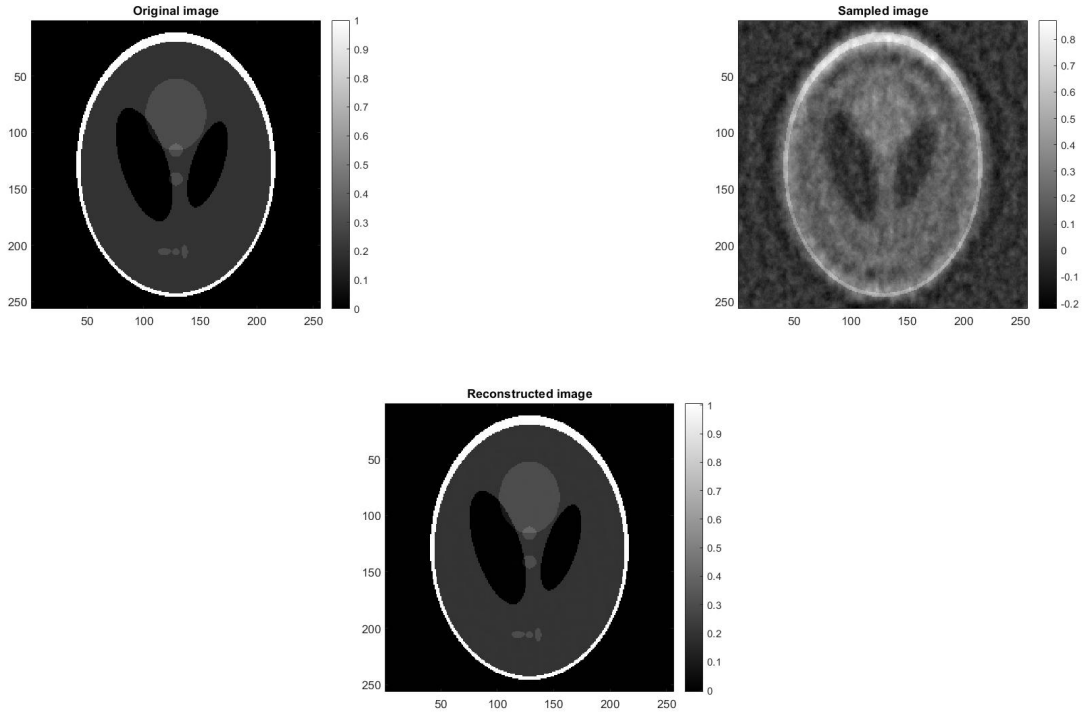


Figure 2: Reconstruction of random sampling mask with $c = 20$.

2.2 $c=10$

When $c = 10$, from Table 3 we can see $\alpha = 1e - 6$ corresponds to the largest SNR and smallest relative error.

α	SNR	Relative Error
1e-3	11.6472	2.6160e-1
1e-4	11.8358	2.5598e-1
1e-5	13.0707	2.2206e-1
1e-6	13.4722	2.1203e-1
1e-7	13.2431	2.1769e-1

Table 3: α vs. SNR and relative error

The result with $\alpha = 1e - 6$ is shown in Figure 3.

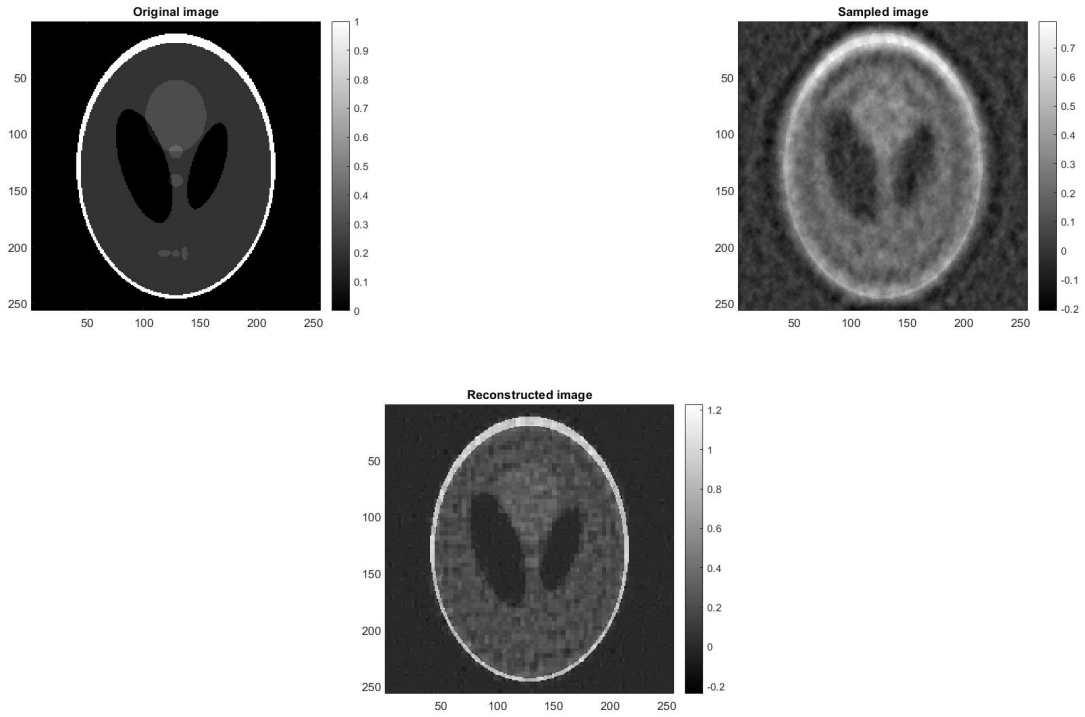


Figure 3: Reconstruction of random sampling mask with $c = 10$.

2.3 $c=5$

When $c = 5$, from Table 4 we can see $\alpha = 1e - 7$ corresponds to the largest SNR and smallest relative error.

α	SNR	Relative Error
1e-4	6.3579	4.8096e-1
1e-5	6.3940	4.7896e-1
1e-6	6.4531	4.7571e-1
1e-7	6.6115	4.6712e-1
1e-8	6.3627	4.8069e-1

Table 4: α vs. SNR and relative error

The result with $\alpha = 1e - 7$ is shown in Figure 4.

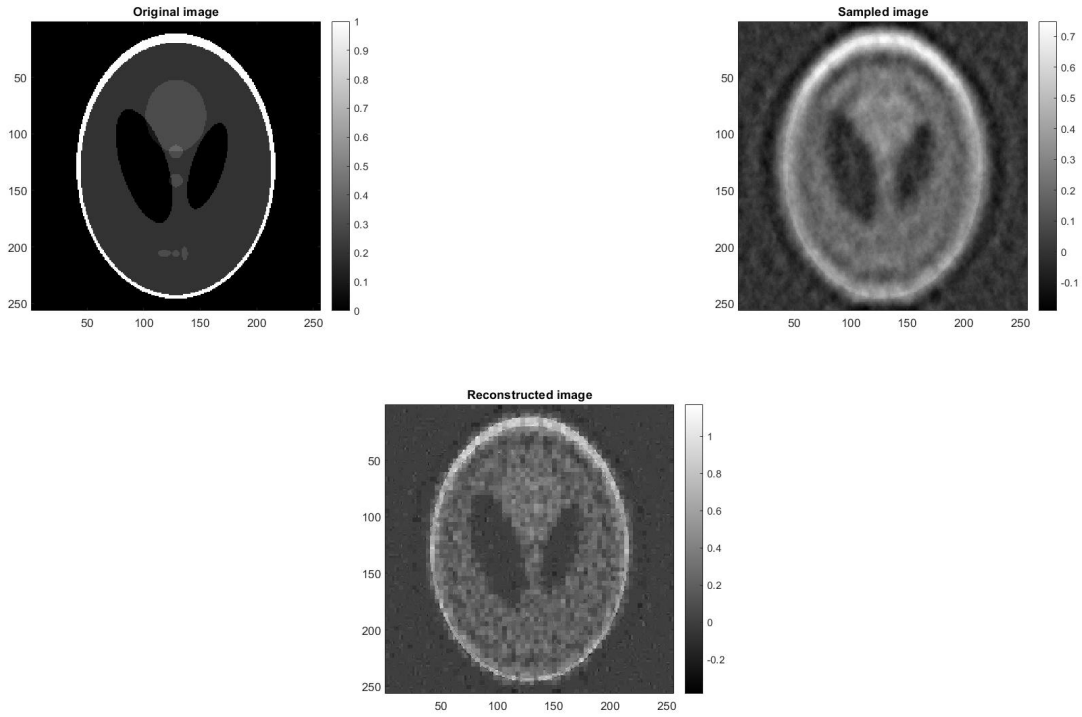


Figure 4: Reconstruction of random sampling mask with $c = 5$.

3 Matlab Code

```
1 clear all;clc;
2 tic
3 X = phantom(256);
4
5 % %%%%%%%%%Mask 1
6 % n = size(X,1); % X is a square matrix with dimension nxn.
7 % Ls = 50;
8 % sigma = 0.05;
9 % SpMsk= fftshift(MRImask(n,Ls));
10 % idx = find(SpMsk~=0);
11 % len = nnz(SpMsk); % number of nonzeros in SpMsk;
12 % K = fft2(X).*SpMsk; % simulate partial Fourier data K
13 % % add complex valued noise to the locations with data sampled
14 % K(idx) = K(idx)+sigma*(randn(len,1)+sqrt(-1)*randn(len,1));
15 % P = zeros(n,n);
16 % P(SpMsk~=0)=1;
17
18 % %%%%%%%%%Mask 2
19 n = size(X,1);
20 Mask = zeros(n,n);
21 c=20; % select c% of pixels in Mask
22 num = round(n*n*c/100);
23 idxAll = randperm(n^2);
24 idxChoose = idxAll(1:num);
25 Mask(idxChoose) = 1;
26 center = round(n/2);
27 h = 10;
28 Mask(center-h+1:center+h,center-h+1:center+h) = 1;
29 Mask = fftshift(Mask);
30 K = fft2(X).*Mask;
31 P = zeros(n,n);
32 P(Mask~=0)=1;
33
34 Psi = @(x) Wavedb1Phi(x,0);
35 PsiT = @(x) Wavedb1Phi(x,1);
36
37 % initialization
38 x = zeros(n,n);
39 p = zeros(1,n^2);
40
41 % parameters
42 alpha = 1e-6; % to be tuned
43 tau = 10;
44 sigma = 10000*tau;
45 snr = 0;
46
47 temp = sum(sum(X.^2))/(n*n);
48 % Primal-Dual Algorithm
49 for k=1:100000
50     xold=x;
51     snr0 = snr;
52
53     p0 = p + sigma*alpha*Psi(x);
54     p=p0./max(1,abs(p0));
55
56     rhs = tau*K + fft2(-alpha*tau*PsiT(p) + x);
57     x = ifft2( rhs./(tau*(P)+1) );
58     x=real(x);
59
60
61     mse = sum(sum((x-X).^2))/(n*n);
```

```

62     snr = 10*log10(temp/mse);
63     if mod(k,100)==0
64         imagesc(x);colormap(gray);axis square;drawnow;
65         fprintf('%d \t %f \n',k,snr);
66     end
67     if snr<snr0
68         break;
69     end
70 end
71
72 % compute MSE and relative error
73 mse = sum(sum((x-X).^2))/(n*n);
74 temp = sum(sum(X.^2))/(n*n);
75 snr = 10*log10(temp/mse);
76 fprintf('SNR = %f \n',snr);
77 re = norm(x-X,'fro')/norm(X,'fro');
78 fprintf('Relative Error = %e \n',re);
79
80 % plot
81 B = real(ifft2(K));
82 figure(1);imagesc(X);colormap(gray);colorbar;
83 title('Original image');axis square;
84 figure(2);imagesc(B);colormap(gray);colorbar;
85 title('Sampled image');axis square;
86 figure(3);imagesc(x);colormap(gray);colorbar;
87 title('Reconstructed image');axis square;
88 toc

```