

Return your written solutions either in person or by email
to ves.kaarnioja@fu-berlin.de by Monday 16 May, 2022, 12:15

1. Let $A \in \mathbb{R}^{m \times n}$ and $y \in \mathbb{R}^m$, and consider the equation $Ax = y$. Prove that the corresponding Landweber–Fridman iterates $\{x_k\}_{k=0}^{\infty}$ can be written explicitly as

$$x_k = \beta \sum_{j=0}^{k-1} (I - \beta A^T A)^j A^T y, \quad k = 1, 2, \dots$$

Moreover, with the help of a singular system (λ_j, v_j, u_j) of A , show that this is equal to

$$x_k = \sum_{j=1}^p \frac{1}{\lambda_j} (1 - (1 - \beta \lambda_j^2)^k) (u_j^T y) v_j, \quad k = 0, 1, \dots,$$

where $p = \text{rank}(A)$.

2. Let us consider a simple X-ray tomography problem with limited angle data. On the course's homepage at

http://numerik.mi.fu-berlin.de/wiki/SS_2022/InverseProblems.php

you can download the file `week3.mat`, which can be imported into MATLAB with the command

```
load week3 A S N
```

The file contains a sparse tomography matrix $A \in \mathbb{R}^{4900 \times 10000}$, a noisy sinogram $S \in \mathbb{R}^{70 \times 70}$, and the dimension of the original object $N = 100$.

Left-multiplying a (vectorized) square image $B \in \mathbb{R}^{10000}$ by matrix A computes 70 discretized X-ray projections of B from a limited angle view of $\pm 60^\circ$ with respect to the positive horizontal axis. For example, the code

```
B = phantom(100); % Shepp-Logan phantom with size 100x100
figure;
imagesc(B), axis square, colormap gray
y = A*B(:);
Z = reshape(y,70,70); % The corresponding sinogram
figure;
imagesc(Z), axis square, colormap gray
```

produces the sinogram corresponding to object B . The first column of the resulting Z is the X-ray projection of the phantom from the angle of view of -60° with respect to the positive horizontal axis, the last column of Z corresponds to the angle of view of $+60^\circ$ with respect to the positive horizontal

axis, and the remaining columns of Z are the X-ray projections of the phantom from equidistant angles of view between these two extremes.

Your task is to reconstruct the object corresponding to the given sinogram S . First, form the vectorized sinogram by $y = \mathbf{S}(:)$, then use Landweber–Fridman iteration to solve the equation

$$Ax = y.$$

Use $\beta = 3$ and the Morozov discrepancy principle with

$$\varepsilon = \sqrt{0.01^2 \cdot 70^2} = 0.70,$$

which is the square root of the expected value for the squared norm of the (vectorized) noise, under the assumption that each pixel of the sinogram S is contaminated with normally distributed additive noise with zero mean and standard deviation 0.01. Visualize the resulting reconstruction after using $\mathbf{X} = \text{reshape}(\mathbf{x}, 100, 100)$ to reshape the reconstruction into an image, and plot the value of the residual

$$f(k) := \|Ax_k - y\|$$

as a function of k . Here, $\{x_k\}$ denote the Landweber–Fridman iterates. Why is $\beta = 10$ a bad choice? How about $\beta = 0.01$?