

## Math 355 Homework Problems #8

MATRIX ANALYSIS AND APPLIED LINEAR ALGEBRA, by C. Meyer

1. For the data in each of the files, “LeastSquaresDataLinear.mat” and “LeastSquaresDataQuadratic.mat”, you are to do the following:

- (a) compute the singular values
- (b) determine the relative error associated with approximating the data with only the first term in the outer expansion
- (c) use least-squares (line for “. . .Linear.mat”, and quadratic for “. . .Quadratic.mat”) to fit the data associated with the first term in the outer expansion - make sure you explicitly write down the coefficients
- (d) plot the least-squares curve for the data associated with the first term in the outer expansion, along with the original data, on one plot.

2. Let  $A \in \mathcal{M}_{m \times n}(\mathbb{F})$ , and suppose  $A$  has full rank. Show that the solution using the Moore-Penrose pseudo-inverse,  $\mathbf{x} = A^\dagger \mathbf{b}$ , to the linear system,  $A\mathbf{x} = \mathbf{b}$ , is the solution to the associated normal equations.

3. Consider the matrix  $A = \begin{pmatrix} -4 & -2 & -4 & -2 \\ 2 & -2 & 2 & 1 \\ -4 & 1 & -4 & -2 \end{pmatrix}$ .

- (a) What is  $r = \text{rank}(A)$ ?
- (b) Compute the SVD,  $A = U\Sigma V^H$ , where

$$U \in \mathcal{M}_{3 \times r}(\mathbb{F}), \quad \Sigma \in \mathcal{M}_r(\mathbb{F}), \quad V \in \mathcal{M}_{4 \times r}(\mathbb{F}).$$

- (c) Using the SVD of  $A$ , provide an orthonormal basis for  $\text{Col}(A)$ .
- (d) Using the SVD of  $A$ , provide an orthonormal basis for  $\text{Col}(A^H)$ .
- (e) Compute the Moore-Penrose pseudo-inverse,  $A^\dagger \in \mathcal{M}_{4 \times 3}(\mathbb{F})$ .

4. Let  $A \in \mathcal{M}_{m \times n}(\mathbb{F})$  have the SVD outer product expansion,

$$A = \sum_{j=1}^r \sigma_j \mathbf{u}_j \mathbf{v}_j^H, \quad \sigma_1 \geq \sigma_2 \geq \cdots \geq \sigma_r > 0.$$

Show that the Moore-Penrose pseudo-inverse of  $A$  has the outer expansion,

$$A^\dagger = \sum_{j=1}^r \frac{1}{\sigma_j} \mathbf{v}_j \mathbf{u}_j^H.$$

5. Let  $A \in \mathcal{M}_n(\mathbb{F})$  be invertible and have the SVD outer product expansion,

$$A = \sum_{j=1}^n \sigma_j \mathbf{u}_j \mathbf{v}_j^H, \quad \sigma_1 \geq \sigma_2 \geq \cdots \geq \sigma_n > 0.$$

For a given  $s < n$  let  $B$  be a lower rank approximation,

$$B = \sum_{j=1}^s \sigma_j \mathbf{u}_j \mathbf{v}_j^H \quad \rightsquigarrow \quad B^\dagger = \sum_{j=1}^s \frac{1}{\sigma_j} \mathbf{v}_j \mathbf{u}_j^H.$$

Using the norm induced by the standard inner product on  $\mathbb{F}^n$ , show that

$$\|A^{-1}\mathbf{b} - B^\dagger\mathbf{b}\| \geq \frac{1}{\sigma_{s+1}} \left( \sum_{j=s+1}^n |b_j|^2 \right)^{1/2}, \quad b_j = \langle \mathbf{u}_j, \mathbf{b} \rangle.$$

In other words, the exact solution,  $A^{-1}\mathbf{b}$ , to the linear system is not necessarily close to the normal equations solution,  $B^\dagger\mathbf{b}$ , of the approximate system. (*Hint*: use the Moore-Penrose pseudo-inverse for  $A^{-1}$ )

6. Take your favorite digital image of minimal size 2 MB. Using the provided MATLAB program `Data-CompressionColor.m` as a template, determine the size of your reduced image which has a relative error of 10% or less when compared to your original image. Provide supporting evidence for your conclusion. (*Hint*: use the MATLAB command `imwrite` to convert a data file to an image)