

Least Squares - Orthogonal Diagonalization - Spectral Decomposition - Singular Value Decomposition

1. Given,

$$\mathbf{A} = \begin{bmatrix} 2 & 4 \\ -5 & -1 \\ 1 & 2 \end{bmatrix}.$$

Determine an orthonormal basis for the column space of  $\mathbf{A}$ .

2. Given the linear system of equations,

$$x_1 + x_2 = 2$$

$$x_1 + x_2 = 4$$

- (a) Determine the least-squares solution to the linear system.
- (b) Determine the least-squares error associated with the linear system.
- (c) Graph the linear system, the least-squares solution, and the least-squares error in  $\mathbb{R}^2$ .

3. Given,

$$\mathbf{A} = \begin{bmatrix} 2 & 3 \\ 2 & 4 \\ 1 & 1 \end{bmatrix}, \quad \mathbf{b} = \begin{bmatrix} 7 \\ 3 \\ 1 \end{bmatrix}.$$

- (a) Show that the columns of  $\mathbf{A}$  are linearly independent.
- (b) Determine the  $\mathbf{QR}$  factorization of  $\mathbf{A}$ .
- (c) Using this factorization calculate the unique least-squares solution  $\hat{\mathbf{x}} = \mathbf{R}^{-1}\mathbf{Q}^T\mathbf{b}$ .<sup>1</sup>

4. Recall the Pauli Spin Matrix from a previous homework,

$$\sigma_2 = \sigma_y = \begin{bmatrix} 0 & -i \\ i & 0 \end{bmatrix}.$$

- (a) Show that  $\sigma_y$  is self-adjoint.<sup>2</sup>
- (b) Find the orthogonal diagonalization of  $\sigma_y$ .<sup>3</sup>
- (c) Show that  $\sigma_y = \lambda_1\mathbf{x}_1\mathbf{x}_1^H + \lambda_2\mathbf{x}_2\mathbf{x}_2^H$ , where  $\mathbf{x}_1$  and  $\mathbf{x}_2$  are the normalized eigenvectors from part (b).<sup>4</sup>

5. Given,

$$\mathbf{A} = \begin{bmatrix} 7 & 1 \\ 0 & 0 \\ 5 & 5 \end{bmatrix}.$$

Find a singular value decomposition of  $\mathbf{A}$ .

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<sup>1</sup>See theorem 6.5.15 on page 414.

<sup>2</sup>If a matrix is equal to its transpose then we say that the matrix is symmetric. If the matrix has complex numbers then we have a more general definition. For  $\mathbf{A} \in \mathbb{C}^{n \times n}$  we say that  $\mathbf{A}$  is self-adjoint if  $\mathbf{A}^H = \bar{\mathbf{A}}^T = \mathbf{A}$ . That is, a matrix is self-adjoint if it is equal to its own complex-conjugate transpose. Self-adjoint matrices are the analogue of symmetric matrices for the complex number field. Also, notice that this definition recovers the definition of symmetric if the matrix has only real entries.

<sup>3</sup>You should find eigenvectors with complex entries. If you use the standard definition of inner-product then you will get zero length, which is sensible since part of the direction of these vectors is into the complex number system. However, this will lead you to a division by zero when trying to normalize the eigenvector. In the case where vectors have complex entries the inner-product is generalised to  $\mathbf{x} \cdot \mathbf{y} = \mathbf{x}^H\mathbf{y} = \bar{\mathbf{x}}^T\mathbf{y}$ . Notice, again that the standard definition of inner-product is recovered when the vectors are real.

<sup>4</sup>This is called the spectral decomposition of a self-adjoint matrix. It is interesting to note that  $\mathbf{x}_i^H\mathbf{x}_i = 1$ , which implies that the 'matrix' has the same structure as the unitary matrices of chapter 6. Since,  $\mathbf{x}_i\mathbf{x}_i^H$  is not the identity matrix we have that  $\mathbf{x}_i\mathbf{x}_i^H$  has the same structure as unitary matrices from problem 3 in homework 8. Consequently, we conclude that the self-adjoint matrix has been decomposed into projection matrices, which project an arbitrary vector into eigen-subspaces of the original matrix.