CHAPTER 6.6 "LEAST-SQUARES LINE"

Given an experimental data

$$(x_1, y_1), (x_2, y_2), \dots, (x_n, y_n)$$

we want to find a line $y = \beta_0 + \beta_1 x$ that best fits the data. In particular, we want β_0 and β_1 such that

$$\beta_0 + \beta_1 x_1 = y_1$$

$$\vdots$$

$$\beta_0 + \beta_1 x_n = y_n$$

This is same as trying to solve the linear system

$$\underbrace{\begin{bmatrix} 1 & x_1 \\ 1 & x_2 \\ \vdots & \vdots \\ 1 & x_n \end{bmatrix}}_{=A} \underbrace{\begin{bmatrix} \beta_0 \\ \beta_1 \\ \vdots \\ =\beta \end{bmatrix}}_{=\beta} = \underbrace{\begin{bmatrix} y_1 \\ y_2 \\ \vdots \\ y_n \end{bmatrix}}_{=b}$$

In many real life applications, $A\beta = b$ is inconsistent. The least-squares solution $\begin{bmatrix} \beta_0 & \beta_1 \end{bmatrix}^T$ defines a line $y = \beta_0 + \beta_1 x$ which we call **least-squares line** that best fits the data point $(x_1, y_1), \dots, (x_n, y_n)$. To recall, $\begin{bmatrix} \beta_0 & \beta_1 \end{bmatrix}^T$ is the solution to the equation $A^T A\beta = A^T b$.

CHAPTER 7.1 DIAGONALIZATION OF SYMMETRIC MATRICES

Definition 1. A matrix *A* is **symmetric** if $A = A^T$. Equivalently, the matrix has arbitrary entries along the main diagonal, and its entries are symmetric with respect to the main diagonal.

Definition 2. *A* is **orthogonally diagonalizable** if there exists an orthogonal matrix $P(P^{-1} = P^D)$ and a diagonal matrix *D* such that $A = PDP^T$.

Theorem 3 (Spectral Theorem). Let *A* be an symmetric $n \times n$ -matrix. Then

- (a) *A* has *n* real eigenvalues, counting multiplicities.
- (b) The dimension of the eigenspace for each eigenvalue λ equals the multiplicity of λ as a root of the characteristic equation. (i.e. diagonalizable)
- (c) The eigenspaces are mutually orthogonal, in the sense that eigenvectors corresponding to different eigenvalues are orthogonal.

Furthermore, an $n \times n$ -matrix A is symmetric if and only if A is orthogonally diagonalizable.

How do we orthogonally diagonalize an $n \times n$ -matrix *A*? You can do this when you can find an orthonormal basis consisting $\{u_1, \ldots, u_n\}$ of eigenvectors of *A* (not always possible). Let

$$Q = \begin{bmatrix} u_1 & \cdots & u_n \end{bmatrix}$$

which is an orthogonal matrix. Then

$$A = QDQ^T$$

where *D* is the diagonal matrix with eigenvalues corresponding to $\{u_1, ..., u_n\}$. What is **amazing** about the spectral theorem is that it says that for a symmetric matrix *A*, you can always find an orthonormal basis of eigenvectors of *A*.