

CS205 Homework #3

Problem 1

Consider an $n \times n$ matrix \mathbf{A} .

1. Show that if \mathbf{A} has distinct eigenvalues all the corresponding eigenvectors are linearly independent.
2. Show that if \mathbf{A} has a full set of eigenvectors (i.e. any eigenvalue λ with multiplicity k has k corresponding linearly independent eigenvectors), it can be written as $\mathbf{A} = \mathbf{Q}\mathbf{\Lambda}\mathbf{Q}^{-1}$ where $\mathbf{\Lambda}$ is a diagonal matrix of \mathbf{A} 's eigenvalues and \mathbf{Q} 's columns are \mathbf{A} 's eigenvectors. Hint: show that $\mathbf{A}\mathbf{Q} = \mathbf{Q}\mathbf{\Lambda}$ and that \mathbf{Q} is invertible.
3. If \mathbf{A} is symmetric show that any two eigenvectors corresponding to different eigenvalues are orthogonal.
4. If \mathbf{A} is symmetric show that it has a full set of eigenvectors. Hint: If (λ, \mathbf{q}) is an eigenvalue, eigenvector (\mathbf{q} normalized) pair and λ is of multiplicity $k > 1$, show that $\mathbf{A} - \lambda\mathbf{q}\mathbf{q}^T$ has an eigenvalue of λ with multiplicity $k - 1$. To show that consider the Householder matrix \mathbf{H} such that $\mathbf{H}\mathbf{q} = \mathbf{e}_1$ and note that $\mathbf{H}\mathbf{A}\mathbf{H}^{-1} = \mathbf{H}\mathbf{A}\mathbf{H}$ and \mathbf{A} are similar.
5. If \mathbf{A} is symmetric show that it can be written as $\mathbf{A} = \mathbf{Q}\mathbf{\Lambda}\mathbf{Q}^T$ for an orthogonal matrix \mathbf{Q} . (You may use the result of (4) even if you didn't prove it)

Problem 2

[Adapted from Heath 4.23] Let \mathbf{A} be an $n \times n$ matrix with eigenvalues $\lambda_1, \dots, \lambda_n$. Recall that these are the roots of the characteristic polynomial of \mathbf{A} , defined as $f(\lambda) \equiv \det(\mathbf{A} - \lambda\mathbf{I})$. Also we define the multiplicity of an eigenvalue to be the degree of it as a root of the characteristic polynomial.

1. Show that the determinant of \mathbf{A} is equal to the product of its eigenvalues, i.e. $\det(\mathbf{A}) = \prod_{j=1}^n \lambda_j$.
2. The *trace* of a matrix is defined to be the sum of its diagonal entries, i.e., $\text{trace}(\mathbf{A}) = \sum_{j=1}^n a_{jj}$. Show that the trace of \mathbf{A} is equal to the sum of its eigenvalues, i.e. $\text{trace}(\mathbf{A}) = \sum_{j=1}^n \lambda_j$.
3. Recall a matrix \mathbf{B} is similar to \mathbf{A} if $\mathbf{B} = \mathbf{T}^{-1}\mathbf{A}\mathbf{T}$ for a non-singular matrix \mathbf{T} . Show that two similar matrices have the same trace and determinant.

4. Consider the matrix

$$\mathbf{A} = \begin{pmatrix} -\frac{a_{m-1}}{a_m} & -\frac{a_{m-2}}{a_m} & \dots & -\frac{a_1}{a_m} & -\frac{a_0}{a_m} \\ 1 & 0 & \dots & 0 & 0 \\ 0 & 1 & \dots & 0 & 0 \\ \vdots & & \ddots & & \vdots \\ 0 & 0 & \dots & 1 & 0 \end{pmatrix}$$

What is \mathbf{A} 's characteristic polynomial? Describe how you can use the power method find the largest root (in magnitude) of an arbitrary polynomial.

Problem 3

Let \mathbf{A} be a $m \times n$ matrix and $\mathbf{A} = \mathbf{U}\mathbf{\Sigma}\mathbf{V}^T$ its singular value decomposition.

1. Show that $\|\mathbf{A}\|_2 = \|\mathbf{\Sigma}\|_2$
2. Show that if $m \geq n$ then for all $\mathbf{x} \in R^n$

$$\sigma_{\min} \leq \frac{\|\mathbf{Ax}\|_2}{\|\mathbf{x}\|_2} \leq \sigma_{\max}$$

where $\sigma_{\min}, \sigma_{\max}$ are the smallest and largest singular values, respectively.

3. The Frobenius norm of a matrix \mathbf{A} is defined as $\|\mathbf{A}\|_F = \sqrt{\sum_{i,j} a_{ij}^2}$. Show that $\|\mathbf{A}\|_F = \sqrt{\sum_{i=1}^p \sigma_i^2}$ where $p = \min\{m, n\}$.
4. If \mathbf{A} is an $m \times n$ matrix and \mathbf{b} is an m -vector prove that the solution \mathbf{x} of minimum Euclidean norm to the least squares problem $\mathbf{Ax} \cong \mathbf{b}$ is given by

$$\mathbf{x} = \sum_{\sigma_i \neq 0} \frac{\mathbf{u}_i^T \mathbf{b}}{\sigma_i} \mathbf{v}_i$$

where $\mathbf{u}_i, \mathbf{v}_i$ are the columns of \mathbf{U} and \mathbf{V} , respectively.

5. Show that the columns of \mathbf{U} corresponding to non-zero singular values form an orthogonal basis of the column space of \mathbf{A} . What space do the columns of \mathbf{U} corresponding to zero singular values span?