Lesson 20: Eigenvalues and eigenvectors

- Review of main linear algebra problems and their solution by a particular matrix factorization
- The eigenvalue problem
- Characteristic polynomial
- Eigenproblem in matrix form
- Eigendecomposition applications

Solving the main problems of linear algebra by matrix factorizations

- Compute the coordinates in a new basis, also known as solving a linear system Ax = Ib when $A \in \mathbb{R}^{m \times m}$ square, non-singular
 - 1. Compute LU factorization, LU = A
 - 2. Solve Ly = b by forward substitution
 - 3. Solve Ux = y by backward substition
- Compute the closest approximation to a high dimensional vector $b \in \mathbb{R}^m$ by linear combination of n vectors $a_1, ..., a_n \in \mathbb{R}^m$, $A = (a_1 ... a_n)$, also known as the *least squares problem*
 - 1. Compute ${m Q}{m R}$ factorization, ${m Q}{m R}={m A}$. The projector onto $C({m A})=C({m Q})$ is ${m P}_{m Q}={m Q}{m Q}^T$
 - 2. Projection of b onto C(A) is QQ^Tb . Set this equal to a linear combination of columns of A, $Ax = QQ^Tb$, Since A = QR, solve the triangular system $Rx = Q^Tb$ to find x
- For a square matrix $A \in \mathbb{C}^{m \times m}$ find those non-zero vectors whose directions are not changed by multiplication by A, $Ax = \lambda x$, known as the *eigenvalue problem*.

Characteristic polynomial of a matrix

• Consider the eigenproblem $Ax = \lambda x$ for $A \in \mathbb{C}^{m \times m}$ with $\lambda \in \mathbb{C}$, and $x \in \mathbb{C}^m$. Rewrite as

$$Ax = \lambda x \Rightarrow (A - \lambda I)x = 0.$$

Since $x \neq 0$, a solution to eigenproblem exists only if $A - \lambda I$ is singular or $\det(A - \lambda I) = 0$

• Investigate form of $\det(\boldsymbol{A} - \lambda \boldsymbol{I}) = 0$

$$\det(\mathbf{A} - \lambda \mathbf{I}) = \begin{vmatrix} a_{11} - \lambda & a_{12} & a_{13} & \dots & a_{1m} \\ a_{21} & a_{22} - \lambda & a_{23} & \dots & a_{2m} \\ a_{31} & a_{32} & a_{33} - \lambda & \dots & a_{3m} \\ \vdots & \vdots & \vdots & \ddots & \vdots \\ a_{m1} & a_{m2} & a_{m3} & \dots & a_{mm} - \lambda \end{vmatrix}$$

• Recall algebraic definition of a determinant as sum of products of index permutations to deduce that $\det(\boldsymbol{A}-\lambda\boldsymbol{I})=0$ is an m^{th} degree polynomial in λ , defined as the characteristic polynomial of $\boldsymbol{A}\in\mathbb{C}^{m\times m}$

$$p_m(\lambda) = \det(\boldsymbol{A} - \lambda \boldsymbol{I})$$

Eigenproblem in matrix form

• For $A \in \mathbb{R}^{m \times m}$, the eigenvalue problem $Ax = \lambda x \ (x \neq 0)$ can be written in matrix form as

$$AX = X\Lambda, X = (x_1 \dots x_m)$$
 eigenvector, $\Lambda = \text{diag}(\lambda_1, ..., \lambda_m)$ eigenvalue matrices

ullet If the column vectors of $oldsymbol{X}$ are linearly independent, then $oldsymbol{X}$ is invertible and $oldsymbol{A}$ can be represented as

$$A = X \Lambda X^{-1}$$

ullet The above form can also be used to reduce A to diagonal form

$$\mathbf{\Lambda} = \mathbf{X}^{-1} \mathbf{A} \mathbf{X}$$

• Link to determinants: recall "determinant of product = product of determinants"

$$\det(\mathbf{A}\mathbf{X}) = \det(\mathbf{X}\mathbf{\Lambda}) \Rightarrow \det(\mathbf{A}) = \det(\mathbf{\Lambda}) \text{ (for } \det(X) \neq 0)$$

Eigendecomposition applications

• Diagonal forms are useful in solving linear ODE systems

$$y' = Ay \Leftrightarrow (X^{-1}y) = \Lambda(X^{-1}y)$$

Also useful in repeatedly applying A

$$\boldsymbol{u}_k = \boldsymbol{A}^k \boldsymbol{u}_0 = \boldsymbol{A} \boldsymbol{A} ... \boldsymbol{A} \boldsymbol{u}_0 = (\boldsymbol{X} \boldsymbol{\Lambda} \boldsymbol{X}^{-1}) (\boldsymbol{X} \boldsymbol{\Lambda} \boldsymbol{X}^{-1}) ... (\boldsymbol{X} \boldsymbol{\Lambda} \boldsymbol{X}^{-1}) \boldsymbol{u}_0 = \boldsymbol{X} \boldsymbol{\Lambda}^k \boldsymbol{X}^{-1} \boldsymbol{u}_0$$