CS 725: Foundations of Machine Learning

Overview of linear algebra for ML

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Vectors

An n-dimensional vector is simply n real numbers, written as a column:

$$\mathbf{u} = \begin{bmatrix} 1 \\ 2 \\ -3 \end{bmatrix}$$

The set of all *n*-dimensional vectors is denoted by \mathbb{R}^n

Note: Can consider more general vector spaces over "fields." We'll stick to \mathbb{R} .

Vector Operations

Vector operations: scalar multiplication and vector addition

Linear Combination: Combining a set of vectors in \mathbb{R}^n using these operations yields another vector in \mathbb{R}^n

E.g.,

$$\mathbf{u} = \begin{bmatrix} 1 \\ 2 \\ -3 \end{bmatrix} \qquad \mathbf{v} = \begin{bmatrix} -1 \\ -1 \\ 0 \end{bmatrix} \qquad \Rightarrow \qquad 2\mathbf{u} + \mathbf{v} = \begin{bmatrix} 1 \\ 3 \\ -6 \end{bmatrix}$$

Basis Vectors

Any vector in \mathbb{R}^n can be written as a linear combination of the (standard) basis vectors:

$$\mathbf{u} = \mathbf{e}_1 + 2\mathbf{e}_2 - 3\mathbf{e}_3$$

where

$$\mathbf{e}_1 = egin{bmatrix} 1 \ 0 \ 0 \end{bmatrix} \qquad \mathbf{e}_2 = egin{bmatrix} 0 \ 1 \ 0 \end{bmatrix} \qquad \mathbf{e}_3 = egin{bmatrix} 0 \ 0 \ 1 \end{bmatrix}$$

Linear Transformations

Linear algebra deals with "linear" functions (or linear transformations) that map vectors in \mathbb{R}^n to vectors in \mathbb{R}^m

f is said to be linear if

$$f(\mathbf{u} + \mathbf{v}) = f(\mathbf{u}) + f(\mathbf{v})$$

 $f(c \mathbf{u}) = c f(\mathbf{u})$

Linear Transformations

A linear transformation is fully specified by how it acts on the basis vectors:

$$\mathbf{u} = u_1 \mathbf{e}_1 + \ldots + u_n \mathbf{e}_n \implies f(\mathbf{u}) = u_1 f(\mathbf{e}_1) + \ldots + u_n f(\mathbf{e}_n)$$

Thus $(f(\mathbf{e}_1), \dots, f(\mathbf{e}_n))$ fully determines f

Can denote such a transformation by a matrix, $\mathbf{A} = \begin{bmatrix} \mathbf{a}_1 & \mathbf{a}_2 & \dots & \mathbf{a}_n \end{bmatrix}$ where each $\mathbf{a}_i = f(\mathbf{e}_i)$ is a column vector.

6

Linear Transformation using Matrix Multiplication

Let
$$\mathbf{A} = \begin{bmatrix} \mathbf{a}_1 & \mathbf{a}_2 & \dots & \mathbf{a}_n \end{bmatrix}$$
 be an $m \times n$ matrix.

Multiplication of a vector \mathbf{u} by the matrix \mathbf{A} is defined so that $f(\mathbf{u}) = \mathbf{A}\mathbf{u}$:

$$\begin{bmatrix} \mathbf{a}_1 & \mathbf{a}_2 & \dots & \mathbf{a}_n \end{bmatrix} \begin{bmatrix} u_1 \\ u_2 \\ \vdots \\ u_n \end{bmatrix} = \sum_i u_i \mathbf{a}_i =: \mathbf{A}\mathbf{u}$$

Multiplying two matrices:
$$\mathbf{B}\mathbf{A} = \begin{bmatrix} \mathbf{B}\mathbf{a}_1 & \mathbf{B}\mathbf{a}_2 & \dots & \mathbf{B}\mathbf{a}_n \end{bmatrix}$$
 where $\mathbf{B}\mathbf{a}_i = f_{\mathbf{B}}(\mathbf{a}_i) = f_{\mathbf{B}}f_{\mathbf{A}}(\mathbf{e}_i)$.

Matrix Multiplication

If **V** is a matrix of size $n \times m$ and **W** is an $m \times p$ matrix, then the matrix product **U** = **VW** is of size $n \times p$ and can be computed as follows:

$$\mathbf{V} = \begin{bmatrix} v_{1,1} & \dots & v_{1,m} \\ \vdots & \vdots & \vdots \\ v_{n,1} & \dots & v_{n,m} \end{bmatrix}, \mathbf{W} = \begin{bmatrix} w_{1,1} & \dots & w_{1,p} \\ \vdots & \vdots & \vdots \\ w_{m,1} & \dots & w_{m,p} \end{bmatrix}, \mathbf{U} = \begin{bmatrix} u_{1,1} & \dots & u_{1,p} \\ \vdots & \vdots & \vdots \\ u_{n,1} & \dots & u_{n,p} \end{bmatrix}$$

where
$$u_{i,j} = \sum_{k=1}^{m} v_{i,k} w_{k,j}$$

Matrix multiplication is not necessarily commutative ($AB \neq BA$), but it is associative (A(BC) = (AB)C).

8

Determinants

Determinants are defined for square matrices $\mathbf{A} \in \mathbb{R}^{n \times n}$, and denoted by $\det(\mathbf{A})$ or $|\mathbf{A}|$.

$$\begin{vmatrix} a_{11} \end{vmatrix} = a_{11}$$
 $\begin{vmatrix} a_{11} & a_{12} \ a_{21} & a_{22} \end{vmatrix} = a_{11}a_{22} - a_{12}a_{21}$
 $\begin{vmatrix} a_{11} & a_{12} & a_{13} \ a_{21} & a_{22} & a_{23} \ a_{31} & a_{32} & a_{33} \end{vmatrix} = a_{11}a_{22}a_{33} - a_{11}a_{23}a_{32} - a_{12}a_{21}a_{33} + a_{12}a_{23}a_{31} + a_{13}a_{21}a_{32} - a_{13}a_{22}a_{31}$