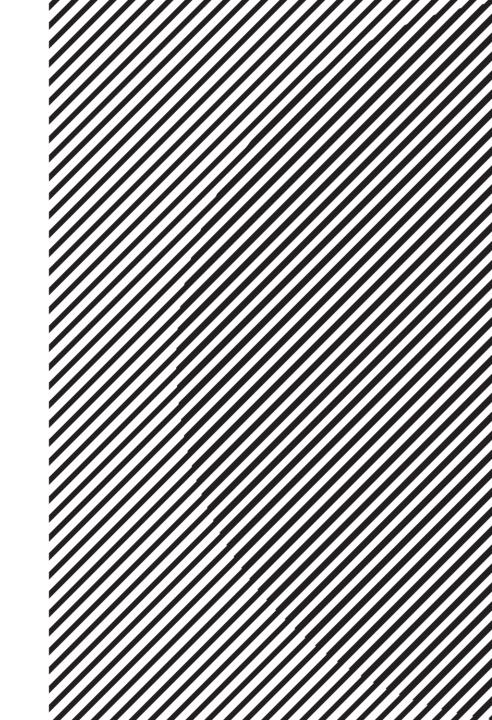
Linear Algebra

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Lecture Overview

- Elements in linear algebra
- Linear system
- Linear combination, vector equation, Four views of matrix multiplication
- Linear independence, span, and subspace
- Linear transformation
- Least squares
- Eigendecomposition
- Singular value decomposition

• Given vectors $[\mathbf{v}_1, \mathbf{v}_2] \cdots$, $[\mathbf{v}_p]$ in \mathbb{R}^n and given scalars

$$c_1, c_2, \cdots, c_p,$$

$$c_1 \mathbf{v}_1 + \cdots + c_p \mathbf{v}_p$$

is called a **linear combination** of v_1, \dots, v_p with weights or coefficients c_1, \dots, c_p .

 The weights in a linear combination can be any real numbers, including zero.



From Matrix Equation to Vector Equation

Recall the matrix equation of a linear system:

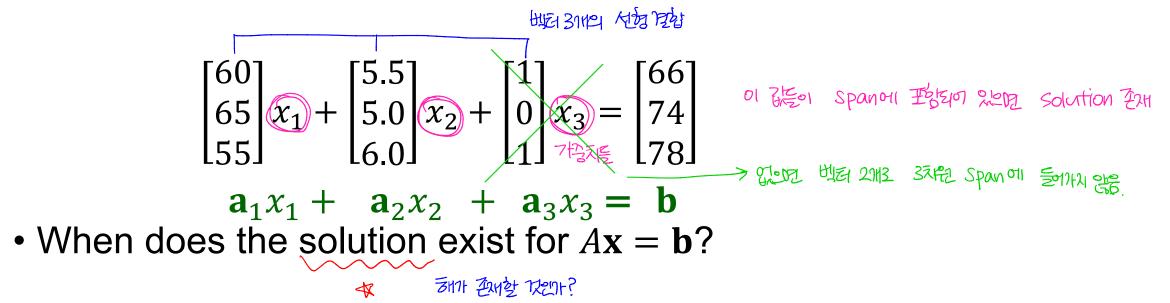
Person D	Weight	Height	ls_smoking	Life-span	[60	5.5	1]	$\begin{bmatrix} x_1 \end{bmatrix}$		[66
1	60kg	5.5ft	Yes (=1)	66	65	5.0	0	$ x_2 $	_	74
2	65kg	5.0ft	No (=0)	74	L55	6.0	1	$[x_3]$		L78
3	55kg	6.0ft	Yes (=1)	78		1		X		_
						$\boldsymbol{\Lambda}$		Λ		IJ

A matrix equation can be converted into a vector equation:



Existence of Solution for Ax = b

Consider its vector equation:







- **Definition**: Given a set of vectors $\mathbf{v}_1, \cdots, \mathbf{v}_p \in \mathbb{R}^n$, Span $\{\mathbf{v}_1, \cdots, \mathbf{v}_p\}$ is defined as the set of all linear combinations of $\mathbf{v}_1, \cdots, \mathbf{v}_p$.
- That is, Span $\{\mathbf{v}_1, \cdots, \mathbf{v}_p\}$ is the collection of all vectors that can be written in the form

$$c_1\mathbf{v}_1 + c_2\mathbf{v}_2 \cdots + c_p\mathbf{v}_p$$

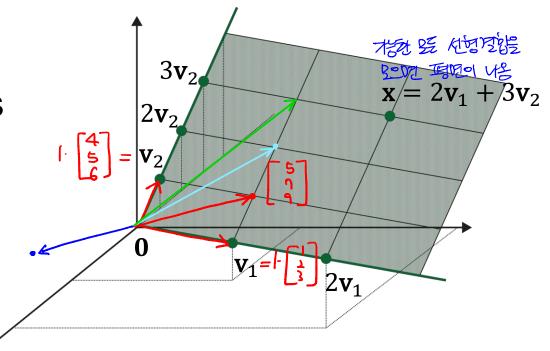
with arbitrary scalars c_1, \dots, c_p .

• Span $\{v_1, \dots, v_p\}$ is also called the subset of \mathbb{R}^n spanned (or generated) by v_1, \dots, v_p .



Geometric Description of Span

- If \mathbf{v}_1 are \mathbf{v}_2 nonzero vectors in \mathbb{R}^3 , with \mathbf{v}_2 not a multiple of \mathbf{v}_1 , then Span $\{\mathbf{v}_1, \mathbf{v}_2\}$ is the plane in \mathbb{R}^3 that contains \mathbf{v}_1 , \mathbf{v}_2 and $\mathbf{0}$.
- In particular, Span $\{v_1, v_2\}$ contains the line in \mathbb{R}^3 through v_1 and 0 and the line through v_2 and 0.





Geometric Interpretation of Vector Equation

Finding a linear combination of given vectors a₁, a₂, and a₃ to be equal to b:

$$\begin{bmatrix} 60 \\ 65 \\ 55 \end{bmatrix} x_1 + \begin{bmatrix} 5.5 \\ 5.0 \\ 6.0 \end{bmatrix} x_2 + \begin{bmatrix} 1 \\ 0 \\ 1 \end{bmatrix} x_3 = \begin{bmatrix} 66 \\ 74 \\ 78 \end{bmatrix}$$

$$\overset{\alpha}{\mathbf{a}}_1 x_1 + \overset{\alpha}{\mathbf{a}}_2 x_2 + \overset{\alpha}{\mathbf{a}}_3 x_3 = \mathbf{b}$$

• The solution exists only when $b \in \text{Span } \{a_1, a_2, a_3\}$.



Matrix Multiplications as Linear Combinations of Vectors

 Recall: we defined matrix-matrix multiplications as the inner product between the row on the left and the column on the right:

• e.g.,
$$\begin{bmatrix} 1 & 6 \\ 3 & 4 \\ 5 & 2 \end{bmatrix} \begin{bmatrix} 1 & -1 \\ 2 & 1 \end{bmatrix} = \begin{bmatrix} 13 & 5 \\ 11 & 1 \\ 9 & -3 \end{bmatrix}$$

• Inspired by the vector equation, we can view $A\mathbf{x}$ as a linear combination of columns of the left matrix:

•
$$\begin{bmatrix} 60 & 5.5 & 1 \\ 65 & 5.0 & 0 \\ 55 & 6.0 & 1 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} = A\mathbf{x} = \begin{bmatrix} \mathbf{a}_1 & \mathbf{a}_2 & \mathbf{a}_3 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} = \mathbf{a}_1 x_1 + \mathbf{a}_2 x_2 + \mathbf{a}_3 x_3$$



Matrix Multiplications as Column Combinations

- Linear combinations of columns
 - Left matrix: bases, right matrix: coefficients

One column on the right $(A\alpha)^T = \alpha^T A^T$ $\begin{bmatrix} 1 & 2 & 3 \end{bmatrix} \begin{bmatrix} 1 & 0 & 1 \\ 1 & 0 & -1 \end{bmatrix} = \begin{bmatrix} 1 & 1 \\ 1 & 0 \end{bmatrix}$



Matrix Multiplications as Row Combinations

- Linear combinations of rows of the right matrix
 - Right matrix: bases, left matrix: coefficients

One row on the left

$$\begin{bmatrix} 1 & 2 & 3 \end{bmatrix} \begin{bmatrix} 1 & 1 & 0 \\ 1 & 0 & 1 \\ 1 & -1 & 1 \end{bmatrix} = \begin{array}{c} 1 \times \begin{bmatrix} 1 & 1 & 0 \\ +2 \times \begin{bmatrix} 1 & 0 & 1 \\ 1 & -1 & 1 \end{bmatrix} & \text{how vector-} \end{bmatrix}$$

Multiple rows on the left

$$\begin{bmatrix} 1 & 2 & 3 \\ 1 & 0 & -1 \end{bmatrix} \begin{bmatrix} 1 & 1 & 0 \\ 1 & 0 & 1 \\ 1 & -1 & 1 \end{bmatrix} = \begin{bmatrix} x_1 & x_2 & x_3 \\ y_1 & y_2 & y_3 \end{bmatrix} = \begin{bmatrix} \mathbf{x}^T \\ \mathbf{y}^T \end{bmatrix}$$

$$\mathbf{x}^T = \begin{bmatrix} x_1 & x_2 & x_3 \end{bmatrix} = \mathbf{1} \begin{bmatrix} 1 & 1 & 0 \end{bmatrix} + \mathbf{2} \begin{bmatrix} 1 & 0 & 1 \end{bmatrix} + \mathbf{3} \begin{bmatrix} 1 & -1 & 1 \end{bmatrix}$$

$$\mathbf{y}^T = \begin{bmatrix} y_1 & y_2 & y_3 \end{bmatrix} = \mathbf{1} \begin{bmatrix} 1 & 1 & 0 \end{bmatrix} + \mathbf{0} \begin{bmatrix} 1 & 0 & 1 \end{bmatrix} + (-1) \begin{bmatrix} 1 & -1 & 1 \end{bmatrix}$$



Matrix Multiplications as Sum of (Rank-1) Outer Products

• (Rank-1) outer product

$$\begin{bmatrix} 1 \\ 1 \\ 1 \end{bmatrix} \begin{bmatrix} 1 & 2 & 3 \\ 1 & 2 & 3 \\ 1 & 2 & 3 \end{bmatrix} = \begin{bmatrix} 1 & 2 & 3 \\ 1 & 2 & 3 \\ 1 & 2 & 3 \end{bmatrix}$$

Sum of (Rank-1) outer products

$$\begin{bmatrix} 1 \\ 1 \\ -1 \\ 1 \end{bmatrix} \begin{bmatrix} 1 & 2 & 3 \\ 4 & 5 & 6 \end{bmatrix} = \begin{bmatrix} 1 \\ 1 \\ 1 \end{bmatrix} \begin{bmatrix} 1 & 2 & 3 \\ 1 & 2 & 3 \end{bmatrix} + \begin{bmatrix} 4 & 5 & 6 \\ -4 & -5 & -6 \\ 4 & 5 & 6 \end{bmatrix}$$

$$= \begin{bmatrix} 1 & 2 & 3 \\ 1 & 2 & 3 \\ 1 & 2 & 3 \end{bmatrix} + \begin{bmatrix} 4 & 5 & 6 \\ -4 & -5 & -6 \\ 4 & 5 & 6 \end{bmatrix}$$



Matrix Multiplications as Sum of (Rank-1) Outer Products

- Sum of (Rank-1) outer products is widely used in machine learning
 - Covariance matrix in multivariate Gaussian
 - Gram matrix in style transfer