

Coursera – Coding the Matrix notes

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Chapter 0

Week 0: The function and the field

0.1 The function and other preliminaries

0.1.1 Set terminology and notation

Definition 0.1.1 (Set). *A set is an unordered collection of objects.*

- \in : indicates that an object belongs to a set. (e.g. $a \in \{a, b, \dots\}$)
- $\mathbb{A} \subseteq \mathbb{B}$: “ \mathbb{A} is a **subset** of \mathbb{B} ”. Every element of \mathbb{A} is also an element of \mathbb{B}
- $\mathbb{A} = \mathbb{B}$: two sets are equal if they contain exactly the same elements.

Set expressions

$\{x \in \mathbb{R} : x \geq 0\}$ is the set of nonnegative numbers. First part specifies where the elements of the set comes from and introduces variables. The second part gives a rule that restricts which elements specified in the first part actually get to make it into the set.

Definition 0.1.2 (Cardinality). *If a set \mathbb{S} is not infinite, we use $|\mathbb{S}|$ to denote the number of elements or cardinality of the set.*

Definition 0.1.3. $\mathbb{A} \times \mathbb{B}$ is the set of all pairs (a, b) where $a \in \mathbb{A}$ and $b \in \mathbb{B}$

0.1.2 The function

Informally, for each input element in a set \mathbb{A} , a function assigns a single output element from another set \mathbb{B}

- \mathbb{A} is called the **domain** of the function
- \mathbb{B} is called the **co-domain**

Definition 0.1.4 (Function). *A function is a set of pairs (a, b) no two of which have the same first element.*

Definition 0.1.5 (Image). *The output of a given input is called the image of that input. The image of q under a function f is denoted $f(q)$*

If $f(q) = r$, we say q maps to r under f . In Mathese, we write this as $q \mapsto r$.

The set from which all the outputs are chosen is called the co-domain. We write:

$$f : \mathbb{D} \rightarrow \mathbb{F}$$

when we want to say that f is a function with domain \mathbb{D} and co-domain \mathbb{F} .

Definition 0.1.6 (Image of a function). *The image of a function is the set of all images of inputs. Mathese: $\text{Im } f$*

Example. $\cos : \mathbb{R} \rightarrow \mathbb{R}$, which means the domain is \mathbb{R} , and the co-domain is \mathbb{R} . The image of $\cos(x)$, $\text{Im } \cos$ is $\{x \in \mathbb{R} : -1 \leq x \leq 1\}$.

Definition 0.1.7. For sets \mathbb{F} and \mathbb{D} , $\mathbb{F}^{\mathbb{D}}$ denotes all functions from \mathbb{D} to \mathbb{F} .

Proposition 0.1.7.1. For finite sets, $|\mathbb{F}^{\mathbb{D}}| = |\mathbb{F}|^{|\mathbb{D}|}$.

Definition 0.1.8 (Identity function). For any domain \mathbb{D} . $\text{id}_{\mathbb{D}} : \mathbb{D} \rightarrow \mathbb{D}$ maps each domain element d to itself.

Definition 0.1.9 (Functional composition). For functions $f : \mathbb{A} \rightarrow \mathbb{B}$ and $g : \mathbb{B} \rightarrow \mathbb{C}$, the functional composition of f and g is the function $(g \circ f) : \mathbb{A} \rightarrow \mathbb{C}$ defined by $(g \circ f)(x) = g(f(x))$.

Proposition 0.1.9.1. $h \circ (g \circ f) = (h \circ g) \circ f$

Definition 0.1.10 (Functional inverses). Functions f and g are functional inverses if $f \circ g$ and $g \circ f$ are defined and are identity functions. A function that has an inverse is invertible.

Definition 0.1.11. $f : \mathbb{D} \rightarrow \mathbb{F}$ is **one-to-one** if $f(x) = f(y)$ implies $x = y$.

Definition 0.1.12. $f : \mathbb{D} \rightarrow \mathbb{F}$ is **ontox** if for every $z \in \mathbb{F}$ there exists an a such that $f(a) = z$.

Proposition 0.1.12.1. Invertible functions are one-to-one.

Theorem 0.1.1 (Function Invertibility Theorem). A function f is invertible if and only if it is one-to-one and onto.

0.2 The Field: Introduction to complex numbers

$i = \sqrt{-1}$ is an imaginary number, this is a solution to an equation such as $x^2 = -1$. For $(x - 1)^2 = 9$, the solution is $x = 1 + 3i$.

A **complex number** has a real part and an imaginary part.

0.2.1 Field notation

When we want to refer to a field without specifying which field we will use the notation \mathcal{F} .

We study three fields;

- The field \mathbb{R} of real numbers.
- The field \mathbb{C} of complex numbers.
- The finite field $GF(2)$, which consists of 0 and 1 under mod 2 arithmetic.

0.3 The Field of playing with \mathbb{C}

We can interpret real and imaginary parts of a complex number as x and y coordinates. Assume that $z \in \mathbb{C}$.

- **Translation** $f(z) = z + z_0, z_0 \in \mathbb{C}$. A translation can “move” a picture anywhere in the complex plane.
- **Scaling** $f(z) = mz, m \in \mathbb{R}$.
- **Invert** $f(z) = (-1)z$.
- **Rotate counterclockwise by 90 degrees** $f(z) = iz$.
- **Rotating by an angle** $f(z) = z \cdot e^{\tau i}$, does rotation by angle τ .

0.4 The Field of playing with $GF(2)$

$GF(2)$ = Galois Field 2, has just two elements: 0 and 1.

- Addition is like exclusive-or. (e.g. $XOR(a, b) = a \neq b; a, b \in GF(2)$)
- Multiplication is just like normal multiplication.

Chapter 1

Week 1: The vector

1.1 What is a vector?

A vector is an array of d numbers, and also can be thought as functions that maps from $\{0, 1, \dots, d-1\}$ to \mathcal{F} with \mathcal{F}^d as the notation.

A vector most of whose values are zero is called a *sparse* vector. If no more than k of the entries are nonzero, we say the vector is k -sparse. A k -sparse vector can be represented using space proportional to k .

1.2 Vector addition and scalar-vector multiplication

Definition 1.2.1 (Vector addition).

$$[u_1, u_2, \dots, u_n] + [v_1, v_2, \dots, v_n] = [u_1 + v_1, u_2 + v_2, \dots, u_n + v_n]$$

Definition 1.2.2 (Zero vector). *The D -vector whose entries are all zero is the zero vector, written $\mathbf{0}_D$ or just $\mathbf{0}$.*

$$\mathbf{v} + \mathbf{0} = \mathbf{v}$$

Definition 1.2.3 (Associativity).

$$(\mathbf{x} + \mathbf{y}) + \mathbf{z} = \mathbf{x} + (\mathbf{y} + \mathbf{z})$$

Definition 1.2.4 (Commutativity).

$$\mathbf{x} + \mathbf{y} = \mathbf{y} + \mathbf{x}$$

For vectors, we refer to field elements as scalars, we use them to scale vectors: $\alpha \mathbf{v}$. Greek letters (e.g. α, β, γ) denote scalars.

Definition 1.2.5. *Multiplying a vector \mathbf{v} by a scalar α is defined as multiplying each entry of \mathbf{v} by α .*

$$\alpha[\mathbf{v}_1, \dots, \mathbf{v}_n] = [\alpha\mathbf{v}_1, \dots, \alpha\mathbf{v}_n]$$

The set of points $\{\alpha\mathbf{v} : \alpha \in \mathbb{R}\}$ forms the line through the origin and \mathbf{v} .

An expression of the form $\alpha\mathbf{u} + \beta\mathbf{v}$ where $0 \leq \alpha \leq 1, 0 \leq \beta \leq 1$, and $\alpha + \beta = 1$ is called a *convex combination* of \mathbf{u} and \mathbf{v} . An expression of the form $\alpha\mathbf{u} + \beta\mathbf{v}$ where $\alpha + \beta = 1$ is called an *affine combination* of \mathbf{u} and \mathbf{v} .

1.3 Dot-product

Definition 1.3.1 (Dot-product of two D -vectors). *Dot-product of two D -vectors is the sum of product of corresponding entries:*

$$\mathbf{u} \cdot \mathbf{v} = \sum_{k \in \{1, \dots, D\}} \mathbf{u}_k \mathbf{v}_k$$

1.3.1 Linear equations

Definition 1.3.2. *A linear equation is an equation of the form*

$$\mathbf{a} \cdot \mathbf{x} = \beta$$

where \mathbf{a} is a vector, β is a scalar, and \mathbf{x} is a vector of variables.

Algebraic properties of dot-product:

- Commutativity $\mathbf{v} \cdot \mathbf{x} = \mathbf{x} \cdot \mathbf{v}$
- Homogeneity $(\alpha\mathbf{u}) \cdot \mathbf{v} = \alpha(\mathbf{u} \cdot \mathbf{v})$
- Distributive law $(\mathbf{v}_1 + \mathbf{v}_2) \cdot \mathbf{x} = \mathbf{v}_1 \cdot \mathbf{x} + \mathbf{v}_2 \cdot \mathbf{x}$

Chapter 2

Week 2: The vector space

2.1 Linear combinations

Definition 2.1.1. *An expression*

$$\alpha_1 \mathbf{v}_1 + \cdots + \alpha_n \mathbf{v}_n$$

is a linear combination of the vectors $\mathbf{v}_1, \dots, \mathbf{v}_n$. The scalars $\alpha_1, \dots, \alpha_n$ are the coefficients of the linear combination.

2.2 Span

Definition 2.2.1 (Span). *The set of all linear combinations of some vectors $\mathbf{v}_1, \dots, \mathbf{v}_n$ is called the span of these vectors. Written $\text{Span}\{\mathbf{v}_1, \dots, \mathbf{v}_n\}$.*

Definition 2.2.2. *Let \mathbb{V} be a set of vectors. If $\mathbf{v}_1, \dots, \mathbf{v}_n$ are vectors such that $\mathbb{V} = \text{Span}\{\mathbf{v}_1, \dots, \mathbf{v}_n\}$, then*

- *We say $\{\mathbf{v}_1, \dots, \mathbf{v}_n\}$ is a generating set for \mathbb{V} .*
- *We refer to the vectors $\mathbf{v}_1, \dots, \mathbf{v}_n$ as generators for \mathbb{V} .*

If we use the vectors $[1, 0, 0]$, $[0, 1, 0]$, and $[0, 0, 1]$:

$$[x, y, z] = x[1, 0, 0] + y[0, 1, 0] + z[0, 0, 1]$$

These are called *standard generators* for \mathbb{R}^3 .

2.3 Geometry of sets of vectors

Span of a single nonzero vector \mathbf{v} :

$$\text{Span}\{\mathbf{v}\} = \{\alpha\mathbf{v} : \alpha \in \mathbb{R}\}$$

This is the line through the origin and \mathbf{v} , One-dimensional. The span of the empty set: just the origin, Zero-dimensional.

One way to specify a plane can be with:

$$\{(x, y, z) : ax + by + cz = 0\}$$

Using dot-product, we could rewrite as

$$\{[x, y, z] : [a, b, c] \cdot [x, y, z] = 0\}$$

We can specify a line in three dimensions:

$$\{[x, y, z] : \mathbf{a}_1 \cdot [x, y, z] = 0, \mathbf{a}_2 \cdot [x, y, z] = 0\}$$

Two ways to represent a geometric object (line, plane, etc.) containing the origin:

- Span of some vectors.
- Solution set of some system of linear equations with zero right hand sides.

What is common with these two representations? R: Subset \mathcal{F}^D satisfies three properties:

1. Subset contains the zero vector $\mathbf{0}$.
2. If subset contains \mathbf{v} then it contains $\alpha\mathbf{v}$ for every scalar α .
3. If subset contains \mathbf{u} and \mathbf{v} then it contains $\mathbf{u} + \mathbf{v}$.

Definition 2.3.1 (Vector space). Any subset \mathbb{V} of \mathcal{F}^D satisfying the three properties is called a vector space.

Definition 2.3.2. If \mathbb{U} is also a vector space and \mathbb{U} is a subset of \mathbb{V} then \mathbb{U} is called a subspace of \mathbb{V} .

2.3.1 Convex hull

Definition 2.3.3. For vectors $\mathbf{v}_1, \dots, \mathbf{v}_n$ over \mathbb{R} , a linear combination

$$\alpha_1 \mathbf{v}_1, \dots, \alpha_n \mathbf{v}_n$$

is a convex combination if the coefficients are all nonnegative and they sum to 1.

2.4 Vector spaces

To represent an object that doesn't contain the origin we sum a vector \mathbf{c} and redefine the definition as:

$$\{\mathbf{c} + \mathbf{v} : \mathbf{v} \in \mathbb{V}\}$$

It can also be abbreviated as $\{\mathbf{c} + \mathbb{V}\}$.

Definition 2.4.1. A linear combination

$$\alpha_1 \mathbf{u}_1 + \alpha_2 \mathbf{u}_2 + \dots + \alpha_n \mathbf{u}_n$$

where

$$\alpha_1 + \alpha_2 + \dots + \alpha_n = 1$$

is an affine combination.

Definition 2.4.2. The set of all affine combinations of vectors $\mathbf{u}_1, \mathbf{u}_2, \dots, \mathbf{u}_n$ is called the affine hull of those vectors.

Affine hull of $\mathbf{u}_1, \mathbf{u}_2, \dots, \mathbf{u}_n = \{\mathbf{u}_1 + \text{Span}\{\mathbf{u}_2 - \mathbf{u}_1, \dots, \mathbf{u}_n - \mathbf{u}_1\}\}$. This shows that the affine hull is an affine space.

In general, a geometric object can be expressed as the solution set of a system of linear equations.

$$\{\mathbf{x} : \mathbf{a}_1 \cdot \mathbf{x} = \beta_1, \dots, \mathbf{a}_m \cdot \mathbf{x} = \beta_m\}$$

Conversely, is the solution set an affine space? Consider solution set of a contradictory system of equations, e.g. $1x = 1, 2x = 1$:

- Solution set is empty
- but a vector space \mathbb{V} always contains the zero vector,
- so an affine space $\mathbf{u}_1 + \mathbb{V}$ always contains at least one vector.

Turns out this is the only exception:

Theorem 2.4.1. *The solution set of a linear system is either empty or an affine space.*

Definition 2.4.3. *A linear equation $\mathbf{a} \cdot \mathbf{x} = 0$ with zero right-hand side is a homogeneous linear equation. A system of homogeneous linear equations is called a homogeneous linear system.*

Lemma 2.4.2. *Let \mathbf{u}_1 be a solution to a linear system. Then, for any other vector \mathbf{u}_2 , \mathbf{u}_2 is also a solution if and only if $\mathbf{u}_2 - \mathbf{u}_1$ is a solution to the corresponding homogeneous linear system.*

Chapter 3

Week 3: The matrix

3.1 What is a matrix

The traditional notion of a matrix: a two dimensional array

$$\mathbf{A} = \begin{bmatrix} a & b & c \\ d & e & f \end{bmatrix}$$

where \mathbf{A} is a 2×3 matrix.

For a matrix \mathbf{A} , the i, j element of \mathbf{A} is the element in row i , column j . It is traditionally written as $\mathbf{A}_{i,j}$

Definition 3.1.1. For finite sets \mathbb{R} and \mathbb{C} , and $\mathbb{R} \times \mathbb{C}$ matrix over field \mathcal{F} is a function from $\mathbb{R} \times \mathbb{C}$ to \mathcal{F} .

Definition 3.1.2. $D \times D$ identity matrix is the matrix $\mathbf{1}_D$ such that $\mathbf{1}_D[k, k] = 1$ for all $k \in \{1, \dots, D\}$ and zero elsewhere.

Usually we omit the subscript when D is clear from the context. Often letter \mathbf{I} is used instead of $\mathbf{1}$

3.1.1 Column space and row space

One simple role for a matrix: packing together a bunch of columns or rows.

Definition 3.1.3. Two vectors associated with a matrix \mathbf{M} :

- Column space of $\mathbf{M} = \text{Span}\{\text{columns of } \mathbf{M}\}$. Written $\text{Col}\mathbf{M}$.
- Row space of $\mathbf{M} = \text{Span}\{\text{rows of } \mathbf{M}\}$. Written $\text{Row}\mathbf{M}$.

3.1.2 Transpose

Transpose swaps rows and columns. It is written as \mathbf{A}^\top .

$$\mathbf{A} = \begin{bmatrix} a & b & c \\ d & e & f \end{bmatrix}$$

$$\mathbf{A}^\top = \begin{bmatrix} a & d \\ b & e \\ c & f \end{bmatrix}$$

3.1.3 Matrices as vectors

A matrix can be interpreted as a vector:

- an $\mathbb{R} \times \mathbb{S}$ matrix is a function from $\mathbb{R} \times \mathbb{S}$ to \mathcal{F} ,
- it can be interpreted as an $\mathbb{R} \times \mathbb{S}$ -vector:
 - scalar-vector multiplication
 - vector addition