# Linear Algebra I - Midterm Review

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# The definition of Vector spaces

A vector space V over a field F is a set together with two binary operators: the addition from  $V \times V$  to V and the scalar multiplication from  $F \times V$  to V which satisfy the following properties.  $\forall \vec{x}, \vec{y}, \vec{z} \in V$  and  $\forall a, b \in F$ :

- 1.  $\vec{x} + \vec{y} \in V$ .
- $2. \ a\vec{x} \in V.$
- 3.  $\vec{x} + \vec{y} = \vec{y} + \vec{x}$ .
- 4.  $\vec{x} + (\vec{y} + \vec{z}) = (\vec{x} + \vec{y}) + \vec{z}$ .
- 5. There exists  $\vec{0} \in V$  (independent of  $\vec{x}$ ), such that  $\vec{x} + \vec{0} = \vec{x}$ .
- 6. There exists  $\vec{x}' \in V$ , such that  $\vec{x} + \vec{x}' = \vec{0}$ . (Such  $\vec{x}'$  is unique and it is denoted by  $-\vec{x}$ .)
- 7.  $1\vec{x} = \vec{x}$ .
- 8.  $a(b\vec{x}) = (ab)\vec{x}$ .
- 9.  $a(\vec{x} + \vec{y}) = a\vec{x} + a\vec{y}$ .
- 10.  $(a+b)\vec{x} = a\vec{x} + b\vec{x}$ .

# Subspaces

The subset W is a subspace of V if it inherits the vector space structure from V. More precisely, W is a subspace if the following conditions hold.

- 1. Show that  $\vec{0} \in W$ .
- 2. Show that for  $k \in F$ ,  $\vec{v_1}$ ,  $\vec{v_2} \in W$ , we have  $k\vec{v_1} + \vec{v_2} \in W$ .

# Verify the structure of vector spaces

A standard way to show a given set V together with addition and scalar multiplication is a vector space over F is the following.

- 1. Identify the set V as a subset of a known vector space U.
- 2. Show that the set V is a subspace of U.

# Spanning sets (generating set)

For a subset  $S = \{\vec{v}_1, \dots, \vec{v}_n\},\$ 

$$\operatorname{span}(S) = \left\{ \sum_{i=1}^{n} a_i \vec{v}_i | \vec{v}_i \in S, a_i \in F \right\}.$$

The following are equivalent.

- 1. S spans V, which means  $\operatorname{span}(S) = V$ .
- 2. For all  $\vec{b} \in V$ , the system  $x_1\vec{v}_1 + \cdots + x_n\vec{v}_n = \vec{b}$  always has a solution.

# Linearly independent sets

For a subset  $S = \{\vec{v}_1, \dots, \vec{v}_n\}$  of V, the following are equivalent.

- 1. S is linearly independent.
- 2. For all  $\vec{v} \in S$ ,  $\vec{v} \notin \text{span}(S \setminus \{\vec{v}\})$ .
- 3. Whenever  $\sum_{i=1}^{n} c_i \vec{v}_n = 0$ , we have  $c_1 = \cdots = c_n = 0$ .
- 4. The linear system  $x_1\vec{v_1} + \cdots + x_n\vec{v_n} = 0$  has the only trivial solution.

### **Basis**

A ordered subset  $S = \{\vec{v}_1, \dots, \vec{v}_n\}$  of V is a basis if it is linearly independent and it spans V. The cardinality of the basis is called the dimension of the vector space.

The following are equivalent.

- 1. S is a basis.
- 2. S is linearly independent and S spans V.
- 3. S is a maximal linearly independent set.
- 4. S is a minimal spanning set.
- 5. S is linearly independent and dim V = n.
- 6. S spans V and dim V = n.
- 7. For all  $\vec{b} \in V$ ,  $x_1\vec{v}_1 + \cdots + x_n\vec{v}_n = \vec{b}$  always has a unique solution.
- 8. For all  $\vec{v} \in V$ ,  $\vec{v}$  can be written as  $c_1\vec{v}_1 + \cdots + c_n\vec{v}_n$  for some unique  $c_i \in F$ . In this case, we also write it as

$$\operatorname{Rep}_S(\vec{v}) = \begin{pmatrix} c_1 \\ \vdots \\ c_n \end{pmatrix}.$$

The following are the basic property of basis.

- 1. Every vector space has a basis.
- 2. Every linearly independent set can be extended to a basis.
- 3. Every spanning set contains a basis.
- 4. Any two basis have the same cardinality.

### Direct Sum

For subspace  $W_1, \dots, W_k$  of V. The following are equivalent.

- 1.  $V = W_1 \oplus W_2 \oplus \cdots \oplus W_k$ .
- 2. Every vector  $v \in V$  can be uniquely written as  $w_1 + \cdots + w_n$  with  $w_i \in W_i$ .
- 3. Let  $\alpha_i$  be a basis of  $W_i$ , then  $\alpha_1 \sqcup \alpha_2 \sqcup \cdots \sqcup \alpha_k$  forms a basis of V.
- 4.  $\dim V = \dim W_1 + \cdots + \dim W_k$  and  $W_1 + \cdots + W_k = V$ .

#### Row Rank and Column Rank

The following are some basic properties about row rank and column rank.

- 1. The row/column rank of a matrix is the dimension of its row/column space.
- 2. The row operation does not change the row space.
- 3. The row operation may change the column space but does not change the column rank.
- 4. The row rank is equal to the number of nonzero rows of the echelon form.
- 5. The column rank is equal to the number of leading variables.
- 6. The row rank is always equal to the column rank.

### Linear transformations

Let V and W be vector spaces over F. A map  $T: V \to W$  is a linear transformation if

- 1.  $T(\vec{u} + \vec{v}) = T(\vec{u}) + T(\vec{v})$  for all  $\vec{u}, \vec{v}$  in V.
- 2.  $T(k\vec{u}) = kT(\vec{u})$  for all  $\vec{u}$  in V and k in F.

Moreover, if T is also a bijection, then T is called a (linear) isomorphism.

# **Dimension Forumla**

The kernel of T is

$$\ker(T) = \{ \vec{v} \in V | T(\vec{v}) = \vec{0} \}.$$

The image of T is

$$Im(T) = T(V) = \{T(\vec{v}) | \vec{v} \in V\}.$$

Dimension formula:

$$\dim(V) = \dim \ker(T) + \dim \operatorname{Im}(T).$$

For a linear transform  $T: V \to W$ , the following are equivalent

- 1. T is an isomorphism.
- 2. T is onto.
- 3.  $\ker(T) = \{\vec{0}\}.$

# Matrix Representations

Let  $\alpha = {\vec{\alpha}_1, \dots, \vec{\alpha}_n}$  be a basis of V and  $\beta$  be a basis of W. Then the matrix representing T with respect to the bases  $\alpha$  and  $\beta$  is

$$\operatorname{Rep}_{\alpha,\beta}(T) = \left(\operatorname{Rep}_{\beta}(T(\vec{\alpha}_1)) \cdots \operatorname{Rep}_{\beta}(T(\vec{\alpha}_n))\right).$$

Moreover, we have

$$\operatorname{Rep}_{\beta}(T(\vec{v})) = \operatorname{Rep}_{\alpha,\beta}(T)\operatorname{Rep}_{\alpha}(\vec{v}).$$

When V = W and  $\alpha = \beta$ , we also denote  $\text{Rep}_{\alpha,\beta}(T)$  by  $\text{Rep}_{\alpha}(T)$  for short.

### **Determinants**

For  $A = (a_{ij}) \in M_n(F)$ , let  $M_{ij}(A)$  be the (i, j)-th minor of A obtained by removing the i-th row and the j-th column of A. Then for any  $1 \le i \le n$ , the determinant of A is given by

$$\det(A) = \sum_{i=1}^{n} a_{ij} c_{ij} = \sum_{i=1}^{n} a_{ji} c_{ji}.$$

Here

$$c_{ij} = (-1)^{i+j} \det(M_{ij}(A)),$$

which is the (i, j)-th cofactor of A.

Cramer's Ruler:

Let  $C = (c_{ij})$  be the cofactor matrix and let  $Adj(A) = C^T$ , called the adjugate matrix of A. Then

$$\operatorname{Adj}(A)A = \det(A)I_n.$$

Especially, when  $det(A) \neq 0$ ,

$$A^{-1} = \frac{1}{\det(A)} \mathrm{Adj}(A).$$

**Theorem 1.** The following are the properties of determinants.

- 1. A square matrix A is non-singular if and only if  $\det(A) \neq 0$ . In this case,  $\det(A^{-1}) = \det(A)^{-1}$ .
- 2. A matrix A is of rank greater than or equal to r if and only if A contains a square submatrix B of size r with  $det(B) \neq 0$ .
- 3. For two square matrices A and B of the same size,  $\det(AB) = \det(A)\det(B)$ . If B is invertible, then  $\det(BAB)^{-1} = \det(A)$ .

# Change of Basis

Now suppose  $T: V \to V$  is a linear transform and  $\alpha = \{\vec{\alpha}_1, \dots, \vec{\alpha}_n\}$  and  $\beta = \{\vec{\beta}_1, \dots, \vec{\beta}_n\}$  are two bases of V. Then

$$\operatorname{Rep}_{\alpha}(T) = \operatorname{Rep}_{\beta,\alpha}(id)\operatorname{Rep}_{\beta}(T)\operatorname{Rep}_{\alpha,\beta}(id).$$

Especially, when  $T = L_A$  (the multiplication by A from left) and  $\alpha$  is the standard basis of  $F^n$ , we have  $\text{Rep}_{\alpha}(T) = A$ . In this case, let

$$P = \operatorname{Rep}_{\beta,\alpha}(id) = \left(\vec{\beta}_1, \cdots, \vec{\beta}_n\right)$$

be the change of basis matrix. Then

$$A = P \cdot \operatorname{Rep}_{\beta}(T) \cdot P^{-1}.$$

Especially, when  $\vec{\beta}_1, \dots, \vec{\beta}_n$  are all eigenvectors, then  $\text{Rep}_{\beta}(T)$  is a diagonal matrix.

### Projection onto a line

Let  $V = \mathbb{R}^n$  and  $\langle \cdot, \cdot \rangle$  be the standard inner product. For two nonzero vectors  $\vec{v}, \vec{w}$  in V, set

$$\operatorname{proj}_{\vec{w}}(\vec{v}) = \frac{\langle \vec{w}, \vec{v} \rangle}{\langle \vec{w}, \vec{w} \rangle} \vec{w}.$$

#### Orthogonal and Orthonormal

A set of vectors  $\{\vec{v}_1, \dots, \vec{v}_m\}$  in V is orthogonal if

$$\langle \vec{v}_i, \vec{v}_j \rangle = 0$$

for all  $i \neq j$ . If it also satisfies the extra condition:  $\langle \vec{v}_i, \vec{v}_i \rangle = 1$  for all i, it is orthonormal.

**Theorem 2.** Every orthogonal  $\alpha$  set of nonzero vectors is linearly independent. Especially,  $\alpha$  forms a basis of V if  $|\alpha| = \dim V$ .

**Theorem 3.** Let  $\alpha = {\vec{\alpha}_1, \dots, \vec{\alpha}_n}$  be an orthogonal basis of V. For  $\vec{v} \in V$ ,

$$\vec{v} = \sum_{i=1}^{n} \operatorname{proj}_{\alpha_i}(\vec{v}) = \sum_{i=1}^{n} \frac{\langle \vec{\alpha}_i, \vec{v} \rangle}{\langle \vec{\alpha}_i, \vec{\alpha}_i \rangle} \vec{\alpha}_i$$

# Gram-Schmidt method

For a (linearly independent) set of vectors  $\{\vec{v}_1, \dots, \vec{v}_m\}$ , set

$$\vec{w}_1 = \vec{v}_1, \quad \text{and} \quad \vec{w}_i = \vec{v}_i - \sum_{k=1}^{i-1} \mathrm{proj}_{\vec{w}_k}(\vec{v}_i) \quad \text{for all } i \geq 2.$$

Then

- 1.  $\{\vec{w}_1, \dots, \vec{w}_m\}$  is orthogonal.
- 2.  $\{\vec{v}_1, \dots, \vec{v}_k\}$  and  $\{\vec{w}_1, \dots, \vec{w}_k\}$  span the same space for all k.

### Invariant subspaces

Let  $T:V\mapsto V$  be a linear transform. A subspace W is called a T-invariant subspace if  $T(W)\subseteq W$ . In other words, the restriction  $T\big|_W$  is a linear transform from W to W.

In this case, if  $\alpha$  is a basis of W and  $\alpha \sqcup \beta$  is a basis of V, then  $\operatorname{Rep}_{\alpha \sqcup \beta}(T)$  is a block upper triangular matrix of the form  $\begin{pmatrix} A & C \\ 0 & B \end{pmatrix}$ . Here  $A = \operatorname{Rep}_{\alpha}(T|_{W})$ . In this case,  $\det(T) = \det(A)\det(B)$ .

# Direc Sum of Invariant subspaces

Suppose  $V = W_1 \oplus W_2$  and  $W_1$  and  $W_2$  are both T- invariant subspace. Let  $\alpha$  and  $\beta$  be bases of  $W_1$  and  $W_2$  respectively. Then  $\operatorname{Rep}_{\alpha \sqcup \beta}(T)$  is a block diagonal matrix of the fomr  $\begin{pmatrix} A & 0 \\ 0 & B \end{pmatrix}$ . Here  $A = \operatorname{Rep}_{\alpha}(T|_{W_1})$  and  $B = \operatorname{Rep}_{\beta}(T|_{W_2})$ . In this case,  $\det(T) = \det(A) \det(B)$ .

#### Eigenspace, Eigenvalues and Eigenvectors

For  $\lambda \in F$ , the set

$$E(\lambda) = \{ \vec{v} \in V | T(\vec{v}) = \lambda \}$$

is always a T-invariant subspace. It is called the eigenspace corresponding to  $\lambda$  if it contains a non-zero element. In this case,  $\lambda$  is called an eigenvalue and non-zero elements of  $E(\lambda)$  are called eigenvectors corresponding to  $\lambda$ .

Eigenvalues  $\lambda$  are zeros of the characteristic polynomial

$$f_T(x) = \det(xI - T).$$

which is of degree  $n = \dim(V)$ . We say that the polynomial  $f_T(x)$  splits over F, if  $f_T(x)$  has n zeros (counting multiplicity) in F.

When W is a T-invariant subspace of V, we also have  $f_{T|_{W}}(x)$  divides  $f_{T}(x)$ .

# Diagonalization

**Theorem 4.** Let  $\lambda_1, \dots, \lambda_k$  be the all distinct eigenvalues of T. The following are equivalent.

- ${\it 1.} \ \, T \ \, is \, \, diagonalizable.$
- 2. There exists a set of eigenvectors of T which forms a basis of V.
- 3.  $V = E(\lambda_1) \oplus \cdots \oplus E(\lambda_n)$ .
- 4. The polynomial  $f_T(x)$  splits and dim  $E(\lambda_i)$  is equal to the multiplicity of  $\lambda_i$  as a zero of  $f_T(x)$  for all i.