

E2 205 Error-Control Coding

Lecture 6

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1 Linear Independence

Let $(V, +, \mathbb{F}, \cdot)$ be a vector space. Let $A = \{\underline{\alpha}_1, \underline{\alpha}_2, \dots\}$ be a (possibly infinite) set of vectors drawn from V .

By "a Linear Combination of Vectors from A " we mean terms of the form:

$$\sum_{j=0}^n c_{i_j} \underline{\alpha}_{i_j}, \quad c_{i_j} \in \mathbb{F}, \quad j = 1, 2, \dots, n, \quad n \geq 1 \text{ is an integer.}$$

We say that A is a linearly independent set if for any finite collection

$$\{\underline{\alpha}_{i_j} \in A \mid j = 1, 2, \dots, n\},$$

we have that

$$\sum_{i=0}^n c_{i_j} \underline{\alpha}_{i_j} = 0 \text{ iff } c_{i_j} = 0 \text{ for all } j = 1, 2, \dots, n.$$

We say that A spans V , if given a vector $\underline{v} \in V$, there exist $\{\underline{\alpha}_{i_j} \in A \mid j = 1, 2, \dots, n\}$ for some integer $n \geq 1$ such that

$$\underline{v} = \sum_{j=1}^n c_{i_j} \underline{\alpha}_{i_j}.$$

The space spanned by A is the set

$$\left\{ \sum_{j=0}^n c_{i_j} \underline{\alpha}_{i_j} \mid c_{i_j} \in \mathbb{F}, \quad j = 1, 2, \dots, n, \quad \underline{\alpha}_{i_j} \in A \right\}$$

Example 1 Consider the vector space $(\mathbb{F}[X], +, \mathbb{F}, \cdot)$ and set $A = \{1, x, x^2, \dots\}$ then

- the elements of set A are linearly independent,
- the elements of set A span $\mathbb{F}[X]$.

2 Basis

Definition 2 A basis B for a vector space $(V, +, \mathbb{F}, \cdot)$ is a collection of vectors $\{\alpha_1, \alpha_2, \dots\}$ such that:

1. the set is a linearly independent set,
2. the set spans the vector space V .

2.1 Does every vector space have a basis?

1. If V is finitely generated, i.e, V is of the form

$$V = \langle r_1, r_2, \dots, r_m \rangle,$$

then this is clearly a yes.

2. In the general case, the answer is still yes but the proof relies upon Zorn's Lemma which is equivalent to the Axiom of Choice.

Lemma 3 In our setting, Zorn's Lemma tells us that if T is a set, \mathcal{A} is a collection of subsets of T and if for every chain of subset

$$S_1 \subseteq S_2 \subseteq S_3 \dots \subseteq S_m \dots,$$

the union $\cup_{j=1}^{\infty} S_j \in \mathcal{A}$, then T contains a maximal subset that is not contained in any other subset.

Claim 4 Every vector space has a basis.

Proof: Let T be the vector space V , \mathcal{A} be the collection of all linearly independent subsets S_j of V . Then for every chain

$$S_1 \subseteq S_2 \subseteq S_3 \dots \subseteq S_m \dots,$$

the union $\cup_{j=1}^{\infty} S_j \in \mathcal{A}$. Thus T has a maximal linearly independent subset B .

Claim 5 B is a basis of V .

Proof: Clearly B is a linearly independent set. Now it remains to show that B spans V . Suppose it does not span V .

Let $\underline{x} \in V$ and $\underline{x} \notin \langle B \rangle$.

This implies that $B \cup \{\underline{x}\}$ set contradicts that B is the maximal linearly independent subset. Hence B is the basis for V .

However, it is hard to construct a basis in general.

Example 6 Vector space $(\mathbb{R}^\infty, +, \mathbb{R}, \cdot)$.

$\underline{x} \in \mathbb{R}^\infty \Rightarrow \underline{x} = (x_1, x_2, \dots, x_n, \dots)$. This is an ∞ -dimensional space.

3 Finite dimensional vector space

Definition 7 A vector space is said to be finite dimensional if it contains a basis consisting of a finite number of elements.

Theorem 8 Let $(V, +, \mathbb{F}, \cdot)$ be a finite dimensional vector space. Then any two basis for V must contain the same number of elements.

The proof will make use of following two lemmas.

Lemma 9 If a vector space V has a basis consisting of m elements, then any collection of $n > m$ elements is a linearly dependent set.

Lemma 10 If a vector space V has a basis consisting of n elements, then any collection of $m < n$ elements cannot span the space.

From these two lemmas, it follows that a basis is simultaneously:

1. a maximal linearly independent set
2. a minimal spanning set

Proof(Theorem 8): Let $\{\underline{\alpha}_1, \underline{\alpha}_2, \dots, \underline{\alpha}_m\} = \{\underline{\beta}_1, \underline{\beta}_2, \dots, \underline{\beta}_n\}$ be two basis for the vector space V .

Since $\{\underline{\alpha}_1, \underline{\alpha}_2, \dots, \underline{\alpha}_m\}$ form a basis and the set $\{\underline{\beta}_1, \underline{\beta}_2, \dots, \underline{\beta}_n\}$ is a linearly independent set, it follows that $n \leq m$.

Also, since $\{\underline{\alpha}_1, \underline{\alpha}_2, \dots, \underline{\alpha}_m\}$ form a basis and the set $\{\underline{\beta}_1, \underline{\beta}_2, \dots, \underline{\beta}_n\}$ span the vector space V , it follows that $n \geq m$.

$\therefore n = m$ if both are basis.

3.1 Dimension

Definition 11 *The dimension k of a finite dimensional vector space $(V, +, \mathbb{F}, \cdot)$ is the number of elements in any basis for the vector space.*

3.2 Dimension of a Linear Code

Definition 12 *The dimension of a binary linear code \mathcal{C} of block length n is its dimension when viewed as a subspace of $(\mathbb{F}_2^n, +, \mathbb{F}, \cdot)$.*

4 Notation of Linear Code

A linear code is characterised by three parameters $[n, k, d_{min}]$ where :

- * n is the block length,
- * k is the dimension,
- * d_{min} is the minimum Hamming distance between any two code words.

Therefore,

Size of a linear code = 2^k ,

Rate of a linear code = $\frac{\log_2 2^k}{n} = \frac{k}{n}$.

Example 13 *Single Parity Check Code, $n=7$*

$$[n, k, d_{min}] = [7, 6, 2]$$

Example 14 *Repetition Code, $n=7$*

$$[n, k, d_{min}] = [7, 1, 7]$$

Example 15 *Hamming Code, $n=7$*

$$[n, k, d_{min}] = [7, 4, 3]$$

5 Generator Matrix

Definition 16 Let \mathcal{C} be an $[n,k]$ binary code. Then a generator matrix G to \mathcal{C} is any $(k \times n)$ matrix whose rows form a basis for \mathcal{C} .

$$G = \begin{bmatrix} \underline{g}_1^t \\ \underline{g}_2^t \\ \cdot \\ \cdot \\ \underline{g}_k^t \end{bmatrix}$$

where $\{\underline{g}_1, \underline{g}_2, \cdot, \underline{g}_k\}$ are a basis.

Note

1. A code can in general, have more than one generator matrix.
2. \mathcal{C} is the rowspace of G .

Example 17 For single parity check code, generator matrix can be given by:

$$G = \begin{bmatrix} 1 & 0 & 0 & 0 & 0 & 0 & 1 \\ 0 & 1 & 0 & 0 & 0 & 0 & 1 \\ 0 & 0 & 1 & 0 & 0 & 0 & 1 \\ 0 & 0 & 0 & 1 & 0 & 0 & 1 \\ 0 & 0 & 0 & 0 & 1 & 0 & 1 \\ 0 & 0 & 0 & 0 & 0 & 1 & 1 \end{bmatrix}.$$

Example 18 For repetition code, generator matrix can be given by:

$$G = [1 \ 1 \ 1 \ 1 \ 1 \ 1 \ 1].$$

Example 19 For Hamming code, generator matrix can be given by:

$$G = \begin{bmatrix} 1 & 0 & 0 & 0 & 1 & 1 & 0 \\ 0 & 1 & 0 & 0 & 0 & 1 & 1 \\ 0 & 0 & 1 & 0 & 1 & 0 & 1 \\ 0 & 0 & 0 & 1 & 1 & 1 & 1 \end{bmatrix}.$$

6 Dual Code

Definition 20 The dual \mathcal{C}^\perp of an $[n,k]$ binary code \mathcal{C} is the set :

$$\mathcal{C}^\perp = \{\underline{y} \mid \underline{x}^t \underline{y} = 0, \quad \text{all } \underline{x} \in \mathcal{C}\}$$

Theorem 21 If $G =$

$$\begin{bmatrix} \underline{g}_1^t \\ \underline{g}_2^t \\ \cdot \\ \cdot \\ \underline{g}_k^t \end{bmatrix}$$

is any $(k \times n)$ generator matrix for \mathcal{C} , then the nullspace of generator matrix is precisely the dual code, i.e, $\mathcal{C}^\perp = \eta(G)$.

Proof: Clearly, from the definition of the dual code, it follows that if $\underline{y} \in \mathcal{C}^\perp$

$$\begin{aligned} \Rightarrow \underline{g}_j^t \underline{y} &= \underline{0} \quad , \quad \text{all } 1 \leq j \leq k \\ \therefore \underline{y} &\in \eta(G) \\ \Rightarrow \mathcal{C}^\perp &\subseteq \eta(G) \end{aligned} \tag{1}$$

Next suppose $\underline{y} \in \eta(G)$

$$\Rightarrow \underline{g}_j^t \underline{y} = \underline{0} \quad , \quad \text{all } 1 \leq j \leq k$$

If \underline{x} is a codeword of \mathcal{C} , then $\underline{x} = \sum_{j=1}^k m_j \underline{g}_j$, $m_j \in \mathbb{F}_2$.

$$\begin{aligned} \Rightarrow \left(\sum_{j=1}^k m_j \underline{g}_j^t \right) \underline{y} &= \underline{0} \\ \Rightarrow \underline{x}^t \underline{y} &= \underline{0} \quad , \quad \text{for all } \underline{x} \in \mathcal{C} \\ \Rightarrow \underline{y} &\in \mathcal{C}^\perp \\ \Rightarrow \eta(G) &\subseteq \mathcal{C}^\perp \end{aligned} \tag{2}$$

Therefore from equation (1) and (2) $\eta(G) = \mathcal{C}^\perp$

7 Parity Check Matrix

Definition 22 A parity check matrix for an $[n, k]$ linear code \mathcal{C} is any generator matrix for \mathcal{C}^\perp

Theorem 23 Let H be a parity check matrix for \mathcal{C} , then $\mathcal{C} = \eta(H)$.

Proof:

Let $H =$

$$\begin{bmatrix} h_1^t \\ h_2^t \\ \cdot \\ \cdot \\ h_{n-k}^t \end{bmatrix}$$

Let $\underline{y} \in \mathcal{C}$, then clearly from the definition of parity check matrix

$$\Rightarrow h_j^t \underline{y} = 0$$

$$\therefore \underline{y} \in \eta(H)$$

$$\Rightarrow \mathcal{C} \subseteq \eta(H)$$

Also

$$\dim(\mathcal{C}) = k$$

$$\dim(\eta(H)) = n - (n - k) = k$$

$$\therefore \mathcal{C} = \eta(H)$$

Aside: Let A be any matrix, U is the row reduced echelon form of A . For example,

$$U = \begin{bmatrix} 1 & 0 & 2 & 1 & 3 \\ 0 & 0 & 6 & 1 & 7 \\ 0 & 0 & 0 & 0 & 5 \\ 0 & 0 & 0 & 0 & 0 \end{bmatrix}.$$

It follows that dimension of row space of A is equal to the dimension of column space of A which is further equal to the number of pivots in the row reduced echelon form U of A.

$\dim(\text{Row Space of A}) = \dim(\text{Column Space of A}) = \text{Number of pivots in U}$
 The common dimension is called the rank of A.

Theorem 24 Fundamental Theorem of Linear Algebra *If A is an $(m \times n)$ matrix, then $\text{rank}(A) + \dim(\text{nullspace}(A)) = n$, where n is the number of columns of A.*

Corollary 25 *The dual of the dual is the code itself, i.e., $(\mathcal{C}^\perp)^\perp = \mathcal{C}$.*

1. $(\text{Repetition code})^\perp = (\text{Single Parity Check Code})$
2. $(\text{Single Parity Check Code})^\perp = (\text{Repetition code})$

Theorem 26 *Let H be a $(n - k \times n)$ matrix such that*

1. $\text{rank}(H) = n - k$,
2. $\eta(H) = \mathcal{C}$,

then H is a parity check matrix for \mathcal{C} .

Proof: Consider the equation

$$\begin{bmatrix} \underline{h}_1^t \\ \underline{h}_2^t \\ \cdot \\ \cdot \\ \underline{h}_{n-k}^t \end{bmatrix} [\underline{x}] = \underline{0}$$

We know that

$$\underline{h}_j^t \underline{x} = \underline{0} \Rightarrow \underline{y}_j^t \underline{x} = \underline{0}, \quad \text{all } x \in \mathcal{C}, \quad \underline{y} \in \text{RowSpace}(H)$$

Clearly by the definition of the dual code

$$\Rightarrow \text{RowSpace}(H) \subseteq \mathcal{C}^\perp$$

But on the other hand, since the matrix H has rank = $n-k$, it follows that these vectors actually span the dual code and therefore the row space of H matrix is dual code.

$$\dim(\text{RowSpace}(H)) = n - k$$

$$\dim(\mathcal{C}^\perp) = \dim(\eta(G)) = n - k$$

$$\therefore \text{RowSpace}(H) = \mathcal{C}^\perp$$

So, it follows that H is a generator matrix for \mathcal{C}^\perp $\therefore H$ is a parity check matrix for \mathcal{C} .