

An introduction to coding theory

Adrish Banerjee

Department of Electrical Engineering
Indian Institute of Technology Kanpur
Kanpur, Uttar Pradesh
India

Feb. 6, 2017



Lecture #7A: Bounds on the size of a code



Outline of the lecture

- Hamming bound



Outline of the lecture

- Hamming bound
 - Perfect codes



Outline of the lecture

- Hamming bound
 - Perfect codes
- Singleton bound



Outline of the lecture

- Hamming bound
 - Perfect codes
- Singleton bound
 - Maximum distance separable codes



Outline of the lecture

- Hamming bound
 - Perfect codes
- Singleton bound
 - Maximum distance separable codes
- Plotkin Bound



Outline of the lecture

- Hamming bound
 - Perfect codes
- Singleton bound
 - Maximum distance separable codes
- Plotkin Bound
- Gilbert-Varshamov bound



Bounds on the size of a code

- The basic problem is to find the largest code of a given length, n and minimum distance, d .

Bounds on the size of a code

- The basic problem is to find the largest code of a given length, n and minimum distance, d .
- Let $A(n, d)$ be the maximum number of codewords in any binary code of length n and minimum distance d between the codewords.

Bounds on the size of a code

- The basic problem is to find the largest code of a given length, n and minimum distance, d .
- Let $A(n, d)$ be the maximum number of codewords in any binary code of length n and minimum distance d between the codewords.
- We are interested in finding the maximum number of binary codewords $A(n, d)$ from the n -dimensional vector space.



Bounds on the size of a code

- The basic problem is to find the largest code of a given length, n and minimum distance, d .
- Let $A(n, d)$ be the maximum number of codewords in any binary code of length n and minimum distance d between the codewords.
- We are interested in finding the maximum number of binary codewords $A(n, d)$ from the n -dimensional vector space.
- In other words, we are interested in finding the minimum parity bits $(n - k)$ required for a t -correcting binary code of length n .



Hamming Bound

- For any binary (n, k) linear code with minimum distance $2t + 1$ or greater, the number of parity-check bits satisfies the following inequality:

$$n - k \geq \log_2 \left[1 + \binom{n}{1} + \binom{n}{2} + \cdots + \binom{n}{t} \right]$$

Proof:



Hamming Bound

- For any binary (n, k) linear code with minimum distance $2t + 1$ or greater, the number of parity-check bits satisfies the following inequality:

$$n - k \geq \log_2 \left[1 + \binom{n}{1} + \binom{n}{2} + \cdots + \binom{n}{t} \right]$$

Proof:

- Recall that all the vectors of weight t or less can be used as coset leaders.



Hamming Bound

- For any binary (n, k) linear code with minimum distance $2t + 1$ or greater, the number of parity-check bits satisfies the following inequality:

$$n - k \geq \log_2 \left[1 + \binom{n}{1} + \binom{n}{2} + \cdots + \binom{n}{t} \right]$$

Proof:

- Recall that all the vectors of weight t or less can be used as coset leaders.
- Number of vectors (n -tuple) of weight t or less are:

$$\binom{n}{0} + \binom{n}{1} + \cdots + \binom{n}{t}$$



Hamming Bound

- Total number of coset leaders are 2^{n-k} .



Hamming Bound

- Total number of coset leaders are 2^{n-k} .
- Therefore, we have

$$2^{n-k} \geq \binom{n}{0} + \binom{n}{1} + \cdots + \binom{n}{t}$$



Hamming Bound

- Total number of coset leaders are 2^{n-k} .
- Therefore, we have

$$2^{n-k} \geq \binom{n}{0} + \binom{n}{1} + \cdots + \binom{n}{t}$$

- Taking logarithm on both sides of the inequality, we get

$$n - k \geq \log_2 \left[1 + \binom{n}{1} + \cdots + \binom{n}{t} \right]$$



Perfect code

- A t –error correcting (n, k) block code is called a perfect code, if its standard array has all the error patterns of t or fewer errors and no other error pattern as their coset leaders.



Perfect code

- A t –error correcting (n, k) block code is called a perfect code, if its standard array has all the error patterns of t or fewer errors and no other error pattern as their coset leaders.
- Perfect code satisfies the Hamming bound with equality.



Perfect code

- A t –error correcting (n, k) block code is called a perfect code, if its standard array has all the error patterns of t or fewer errors and no other error pattern as their coset leaders.
- Perfect code satisfies the Hamming bound with equality.
- Examples of perfect codes: single error correcting Hamming code, triple error correcting $(23,12)$ Golay code.



Perfect code

- A t –error correcting (n, k) block code is called a perfect code, if its standard array has all the error patterns of t or fewer errors and no other error pattern as their coset leaders.
- Perfect code satisfies the Hamming bound with equality.
- Examples of perfect codes: single error correcting Hamming code, triple error correcting $(23,12)$ Golay code.
- Note perfect codes are not the best error correcting codes.



Singleton Bound

- The minimum distance d_{\min} of an (n, k) linear code satisfies the following inequality

$$d_{\min} \leq n - k + 1$$

Proof:



Singleton Bound

- The minimum distance d_{\min} of an (n, k) linear code satisfies the following inequality

$$d_{\min} \leq n - k + 1$$

Proof:

- For an (n, k) code that an $(n - k) \times n$ parity check matrix, \mathbf{H} , the row rank of any \mathbf{H} is $(n-k)$.



Singleton Bound

- The minimum distance d_{\min} of an (n, k) linear code satisfies the following inequality

$$d_{\min} \leq n - k + 1$$

Proof:

- For an (n, k) code that an $(n - k) \times n$ parity check matrix, \mathbf{H} , the row rank of any \mathbf{H} is $(n-k)$.
- Hence, the column rank of any \mathbf{H} is $(n-k)$. Any combinations of $(n-k+1)$ columns of \mathbf{H} must be linearly dependent.



Singleton Bound

- The minimum distance d_{\min} of an (n, k) linear code satisfies the following inequality

$$d_{\min} \leq n - k + 1$$

Proof:

- For an (n, k) code that an $(n - k) \times n$ parity check matrix, \mathbf{H} , the row rank of any \mathbf{H} is $(n-k)$.
- Hence, the column rank of any \mathbf{H} is $(n-k)$. Any combinations of $(n-k+1)$ columns of \mathbf{H} must be linearly dependent.
- Recall, that the minimum distance of a code is equal to the minimum number of nonzero columns in \mathbf{H} that are linearly dependent.



Singleton Bound

- The minimum distance d_{\min} of an (n, k) linear code satisfies the following inequality

$$d_{\min} \leq n - k + 1$$

Proof:

- For an (n, k) code that an $(n - k) \times n$ parity check matrix, \mathbf{H} , the row rank of any \mathbf{H} is $(n-k)$.
- Hence, the column rank of any \mathbf{H} is $(n-k)$. Any combinations of $(n-k+1)$ columns of \mathbf{H} must be linearly dependent.
- Recall, that the minimum distance of a code is equal to the minimum number of nonzero columns in \mathbf{H} that are linearly dependent.
- Hence,

$$d_{\min} \leq n - k + 1$$



Singleton Bound

Another proof:

- Any nonzero codeword with only one information weight can at most have $n - k + 1$ codeword weight.



Singleton Bound

Another proof:

- Any nonzero codeword with only one information weight can at most have $n - k + 1$ codeword weight.
- Since, minimum distance of a code is equal to the minimum weight of a nonzero codeword.

$$d_{\min} \leq n - k + 1$$

Maximum Distance Separable (MDS) Codes

- MDS codes satisfies the Singleton bound with equality, i.e.

$$n - k = d - 1$$

Maximum Distance Separable (MDS) Codes

- MDS codes satisfies the Singleton bound with equality, i.e.

$$n - k = d - 1$$

- MDS code has the maximum possible distance between the codewords.



Maximum Distance Separable (MDS) Codes

- MDS codes satisfies the Singleton bound with equality, i.e.

$$n - k = d - 1$$

- MDS code has the maximum possible distance between the codewords.
- Example: Repetition codes, Single parity check codes, Reed-Solomon codes.



Plotkin Bound

- The minimum distance d_{\min} of an (n, k) linear code satisfies the following inequality

$$d_{\min} \leq \frac{n \cdot 2^{k-1}}{2^k - 1}$$

Proof:



Plotkin Bound

- The minimum distance d_{\min} of an (n, k) linear code satisfies the following inequality

$$d_{\min} \leq \frac{n \cdot 2^{k-1}}{2^k - 1}$$

Proof:

- Consider an (n, k) linear code C with generator matrix \mathbf{G} . Arrange the 2^k codewords of C as a $2^k \times n$ array.



Plotkin Bound

- The minimum distance d_{\min} of an (n, k) linear code satisfies the following inequality

$$d_{\min} \leq \frac{n \cdot 2^{k-1}}{2^k - 1}$$

Proof:

- Consider an (n, k) linear code C with generator matrix \mathbf{G} . Arrange the 2^k codewords of C as a $2^k \times n$ array.
- Each column of this array has 2^{k-1} zeros and 2^{k-1} ones.



Plotkin Bound

- The minimum distance d_{\min} of an (n, k) linear code satisfies the following inequality

$$d_{\min} \leq \frac{n \cdot 2^{k-1}}{2^k - 1}$$

Proof:

- Consider an (n, k) linear code C with generator matrix \mathbf{G} . Arrange the 2^k codewords of C as a $2^k \times n$ array.
- Each column of this array has 2^{k-1} zeros and 2^{k-1} ones.
 - Show that the number of codewords that “1” at the l -th position is same as number of codewords that have “0” at the l -th position.



Plotkin Bound

Proof (contd.):

- In the code array, each column contains at least one nonzero entry.



Plotkin Bound

Proof (contd.):

- In the code array, each column contains at least one nonzero entry.
- Consider the l -th column of the code array. Let S_0 be the codewords with a “0” at the l -th position and S_1 be the codewords with a “1” at the l -th position. Let \mathbf{x} be a codeword from S_1 .



Plotkin Bound

Proof (contd.):

- In the code array, each column contains at least one nonzero entry.
- Consider the l -th column of the code array. Let S_0 be the codewords with a “0” at the l -th position and S_1 be the codewords with a “1” at the l -th position. Let \mathbf{x} be a codeword from S_1 .
- Adding \mathbf{x} to each vector in S_0 , we obtain a set S'_1 of codewords with a “1” at the l -th position.

$$|S'_1| = |S_0| \quad \text{and} \quad S'_1 \subseteq S_1$$



Plotkin Bound

Proof (contd.):

- In the code array, each column contains at least one nonzero entry.
- Consider the l -th column of the code array. Let S_0 be the codewords with a “0” at the l -th position and S_1 be the codewords with a “1” at the l -th position. Let \mathbf{x} be a codeword from S_1 .
- Adding \mathbf{x} to each vector in S_0 , we obtain a set S'_1 of codewords with a “1” at the l -th position.

$$|S'_1| = |S_0| \quad \text{and} \quad S'_1 \subseteq S_1$$

- The above condition implies that

$$|S_0| \leq |S_1| \tag{1}$$



Plotkin Bound

Proof (contd.):

- Adding \mathbf{x} to each vector in S_1 , we obtain a set S'_0 of codewords with a “0” at the l —th position.

$$|S'_0| = |S_1| \quad \text{and} \quad S'_0 \subseteq S_0$$

Plotkin Bound

Proof (contd.):

- Adding \mathbf{x} to each vector in S_1 , we obtain a set S'_0 of codewords with a “0” at the l —th position.

$$|S'_0| = |S_1| \quad \text{and} \quad S'_0 \subseteq S_0$$

- - The above condition implies that

$$|S_1| \leq |S_0| \tag{2}$$

Plotkin Bound

Proof (contd.):

- Adding \mathbf{x} to each vector in S_1 , we obtain a set S'_0 of codewords with a “0” at the l -th position.

$$|S'_0| = |S_1| \quad \text{and} \quad S'_0 \subseteq S_0$$

- - The above condition implies that

$$|S_1| \leq |S_0| \tag{2}$$

- From (1) and (2), we get $|S_0| = |S_1|$. Therefore l -th column contains 2^{k-1} zeros and 2^{k-1} ones.



Plotkin Bound

Proof (contd.):

- Adding \mathbf{x} to each vector in S_1 , we obtain a set S'_0 of codewords with a “0” at the l -th position.

$$|S'_0| = |S_1| \quad \text{and} \quad S'_0 \subseteq S_0$$

- - The above condition implies that

$$|S_1| \leq |S_0| \tag{2}$$

- From (1) and (2), we get $|S_0| = |S_1|$. Therefore l -th column contains 2^{k-1} zeros and 2^{k-1} ones.

- Thus the total number of ones in the array is $n \cdot 2^{k-1}$.



Plotkin Bound

- Each nonzero codeword has weight atleast d_{\min} . Hence,

$$(2^k - 1) \cdot d_{\min} \leq n \cdot 2^{k-1}$$



Plotkin Bound

- Each nonzero codeword has weight atleast d_{\min} . Hence,

$$(2^k - 1) \cdot d_{\min} \leq n \cdot 2^{k-1}$$

- This implies that

$$d_{\min} \leq \frac{n \cdot 2^{k-1}}{2^k - 1}$$



Gilbert-Varshamov Bound

- There exists an (n, k) linear code with a minimum distance of at least d that satisfies the following inequality

$$1 + \binom{n-1}{1} + \dots + \binom{n-1}{d-2} < 2^{n-k}$$

Proof:



Gilbert-Varshamov Bound

- There exists an (n, k) linear code with a minimum distance of at least d that satisfies the following inequality

$$1 + \binom{n-1}{1} + \dots + \binom{n-1}{d-2} < 2^{n-k}$$

Proof:

- We shall construct an $n - k \times n$ parity check matrix, \mathbf{H} with the property that no $d - 1$ columns are linearly dependent.



Gilbert-Varshamov Bound

- There exists an (n, k) linear code with a minimum distance of at least d that satisfies the following inequality

$$1 + \binom{n-1}{1} + \dots + \binom{n-1}{d-2} < 2^{n-k}$$

Proof:

- We shall construct an $n - k \times n$ parity check matrix, \mathbf{H} with the property that no $d - 1$ columns are linearly dependent.
- Recall, that this will ensure a minimum distance of d .



Gilbert-Varshamov Bound

- There exists an (n, k) linear code with a minimum distance of at least d that satisfies the following inequality

$$1 + \binom{n-1}{1} + \dots + \binom{n-1}{d-2} < 2^{n-k}$$

Proof:

- We shall construct an $n - k \times n$ parity check matrix, \mathbf{H} with the property that no $d - 1$ columns are linearly dependent.
- Recall, that this will ensure a minimum distance of d .
- The first column could be any nonzero $n - k$ -tuple.



Gilbert-Varshamov Bound

- There exists an (n, k) linear code with a minimum distance of at least d that satisfies the following inequality

$$1 + \binom{n-1}{1} + \dots + \binom{n-1}{d-2} < 2^{n-k}$$

Proof:

- We shall construct an $n - k \times n$ parity check matrix, \mathbf{H} with the property that no $d - 1$ columns are linearly dependent.
- Recall, that this will ensure a minimum distance of d .
- The first column could be any nonzero $n - k$ -tuple.
- Suppose we have chosen i columns so that no $d - 1$ columns are linearly dependent.

Gilbert-Varshamov Bound

- Maximum number of distinct linear combinations of these i columns taken $d - 2$ or fewer at a time is given by N_i

$$N_i = \binom{i}{1} + \cdots + \binom{i}{d-2}$$

Gilbert-Varshamov Bound

- Maximum number of distinct linear combinations of these i columns taken $d - 2$ or fewer at a time is given by N_i

$$N_i = \binom{i}{1} + \cdots + \binom{i}{d-2}$$

- If this number, N_i is less than all possible nonzero $n - k$ -tuple, i.e. $2^{n-k} - 1$, we can add another column different from these linear combinations, and keep the property that any $d - 1$ columns of the new $(n - k) \times (i + 1)$ array are linearly independent.



Gilbert-Varshamov Bound

- Maximum number of distinct linear combinations of these i columns taken $d - 2$ or fewer at a time is given by N_i

$$N_i = \binom{i}{1} + \cdots + \binom{i}{d-2}$$

- If this number, N_i is less than all possible nonzero $n - k$ -tuple, i.e. $2^{n-k} - 1$, we can add another column different from these linear combinations, and keep the property that any $d - 1$ columns of the new $(n - k) \times (i + 1)$ array are linearly independent.
- We continue doing this as long as the following condition is satisfied.

$$\binom{i}{1} + \cdots + \binom{i}{d-2} < 2^{n-k} - 1$$



Gilbert-Varshamov Bound

- Maximum number of distinct linear combinations of these i columns taken $d - 2$ or fewer at a time is given by N_i

$$N_i = \binom{i}{1} + \dots + \binom{i}{d-2}$$

- If this number, N_i is less than all possible nonzero $n - k$ -tuple, i.e. $2^{n-k} - 1$, we can add another column different from these linear combinations, and keep the property that any $d - 1$ columns of the new $(n - k) \times (i + 1)$ array are linearly independent.
- We continue doing this as long as the following condition is satisfied.

$$\binom{i}{1} + \cdots + \binom{i}{d-2} < 2^{n-k} - 1$$

- The above condition should hold for all n columns of the parity check matrix, \mathbf{H} .