Recap: Singleton bound

Theorem: For every (n, k, d) _q code, $n \geq (k + d - 1)$

Another way to look at this: $d \leq (n - k + 1)$

Codes that meet Singleton bound with equality are called **Maximum Distance Separable (MDS)**

Recap:

Maximum Distance Separable (MDS)

Only two binary MDS codes!

Q: What are they?

- 1. Repetition codes $(k = 1)$
- 2. Single-parity check codes $(n-k = 1)$

Need to go beyond the binary alphabet. Finite fields!

Reed-Solmon (RS) codes

One of the most widely codes

- Storage systems, communication systems
- Bar codes (2-dimensional Reed-Solomon bar codes)

Aztec code DataMatrix code

To make the $\alpha_{\rm i}$'s in P($\alpha_{\rm i}$) distinct, need field size q ≥ n

That is, need sufficiently large field size for desired codeword length.

Minimum distance of an (n, k) RS code

Theorem: RS codes have minimum distance $d = (n - k + 1)$ **Proof:** Any ideas?

Hint: Is it a linear code?

- *1. RS is a linear code:* if we add two codewords corresponding to $P(x)$ and $Q(x)$, we get a codeword corresponding to the polynomial $P(x) + Q(x)$. Similarly any linear combination..
- *2. So look at the least weight codeword*. It is the evaluation of a polynomial of degree k-1 at some n points. So it can be zero on only k-1 points. Hence non-zero on at most (n-(k-1)) points. This means distance at least n-k+1

Apply Singleton bound

Meets Singleton bound: RS codes are MDS

Generator matrix of RS code

"Vandermonde matrix"

Special property of Vandermonde matrices: Full rank (columns linearly independent)

Very useful in constructing codes.

Polynomials and their degrees

- **Fundamental theorem of Algebra:** Any non-zero polynomial of degree m has at most m roots (over any field).
- **Corollary 1:** If two degree-**k** polynomials P, Q agree on m+1 locations (i.e., if $P(x_i) = Q(x_i)$ for $x_0, x_1, ..., x_m$), then $P = Q$.
- **Corollary 2:** Given any $m+1$ points (x_i, y_i) , there is at most one degree-m polynomial that has $P(x_i) = y_i$ for all these i.
- **Theorem:** Given any $m+1$ points (x_i, y_i) , there is exactly one degree-m polynomial that has $P(x_i) = y_i$ for all these i. Proof: e.g., use Lagrange interpolation.

Decoding: Recovering Erasures

Recovering from at most (d-1) erasures:

 $d = n - k + 1$

Received codeword:

 $[P(\alpha_1),$, *, $P(\alpha_2), \ldots,$, *, $P(\alpha_n)]$: at most (d-1) symbols erased (where $* =$ erased)

Ideas?

- 1. At most n-k symbols erased
- 2. So have P($\alpha_{\rm i}$) for at least k evaluations
- 3. Interpolation to recover the polynomial

Matrix viewpoint: Solving system of linear equations

Decoding: Correcting Errors

Correcting s errors: (d ≥ 2s+1)

Naïve algo:

- Find κ +s symbols that agree on a degree (k-1) poly $P(x)$.
	- There must exist one: since originally $k + 2s$ symbols agreed and at most s are in error

 $d - 1 = 25$

(i.e., "guess" the n-s uncorrupted locations)

- Can we go wrong? Are there k+s symbols that agree on the wrong degree (k-1) polynomial P'(x)? No.
	- Any subset of k symbols will define $P'(x)$
	- Since at most s out of the k+s symbols are in error, $P'(x) = p(x)$

 $\frac{d-1}{2}$

Decoding: Correcting Errors

Correcting s errors: (d ≥ 2s+1)

Naïve algo:

- Find k+s symbols that agree on a degree (k-1) poly P(x).
	- There must exist one: since originally k + 2s symbols agreed and at most s are in error (i.e., "guess" the n-s uncorrupted locations)

This suggests a brute-force approach, very inefficient. "guess" = "enumerate", so time is (n choose s) \sim n^{λ}s.

More efficient algorithms exist:

"The Berlekamp Welch Algorithm" (results in solving a system of n linear equations)

Codes based on graphs

• Optimized for fast (de)coding

- Based on graphical constructions
- Constructions based on properties of expander graphs

Properties

- $-$ **Expansion:** every small subset ($k \leq \alpha n$) has many \geq β k) neighbors
- **Low degree –** not technically part of the definition, but typically assumed

(α, β) Expander Graphs (bipartite)

Properties

- $-$ **Expansion:** every small subset ($k \leq \alpha n$) on left has many ($\geq \beta k$) neighbors on right
- **Low degree –** not technically part of the definition, but typically assumed

d-regular graphs

An undirected graph is **d-regular** if every vertex has d neighbors.

A **bipartite** graph is **d-left-regular** if every vertex on the left has d neighbors on the right.

We consider only d-left-regular constructions. (And call it d-regular with abuse of notation.)

Expander Graphs: Constructions

Important parameters: size (n) , degree (d) , expansion (β)

Randomized constructions

- A random d-regular graph is an expander with a high probability
- Time consuming and cannot be stored compactly

Explicit constructions

- Cayley graphs, Ramanujan graphs etc
- Typical technique start with a small expander, apply operations to increase its size

Expander Graphs: Constructions

Theorem: For every constant $0 < c < 1$, can construct bipartite graphs with

n nodes on left, cn on right, (α, β) d-regular (left), that are $\sqrt{(\alpha, 3d/4)}$ expanders, for constants α and d that are functions of c alone.

"Any set containing at most alpha fraction of the left has (3d/4) times as many neighbors on the right"

Luby Mitzenmacher Shokrollahi Spielman 2001

Goal: low (linear-time) complexity encoding and decoding

We will focus on **erasure** recovery

- Each bit either reaches intact, or is lost.
- We know the positions of the lost bits.

The random erasure model

Random erasure model:

- Each bit is erased with some probability p (say $\frac{1}{2}$ here)
- Known: a random linear code with rate < 1-p works (why?)

For simplicity.

Can be extended to worst-case error, and bit corruption with extra effort.

[e.g., Spielman1996]

Tornado codes

• Have d-left-regular bipartite graphs with k nodes on the left and pk on the right. Hedgyskd

• Let's again assume 3d/4-expansion.

Tornado codes: Encoding

Why is it linear time? (Hint: Look at the number of edges)

Computes the sum modulo 2 of its neighbors

Number of edges = kd

First, assume that all the parity bits are intact

Find a parity bit such that only one of its neighbors is erased (an *unshared neighbor*)

Fix the erased bit, and repeat.

Tornado codes: Decoding

Intuition:

Want to always find such a parity bit with "Unshared neighbor" property.

Consider the set of corrupted message bit and their neighbors. (Suppose this set is small.)

=> at least one message bit has an unshared neighbor.

Tornado codes: Decoding

Can we always find unshared neighbors?

Expander graphs give us this property if expansion > d/2 (similar argument to one above)

Also, [Luby et al] show that if we construct the graph from a specific kind of degree distribution, then we can always find unshared neighbors.

What if parity bits are lost?

Cascading

- Use another bipartite graph to construct another level of parity bits for the parity bits
- **Final level is encoded using RS or some other code**

Tornado codes enc/dec complexity

Encoding time?

- for the first t stages : $|E| = d \times |V| = O(k)$
- for the last stage: poly(last size) = $O(k)$ by design.

Decoding time?

- start from the last stage and move left
- Last stage is O(k) by design
- Rest proportional to $|E| = O(k)$

So get very fast (linear-time) coding and decoding. 100s-10,000 times faster than RS