

A Crash Course on Coding Theory

Topic: Linear time decoding - Part II

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This lecture will focus on

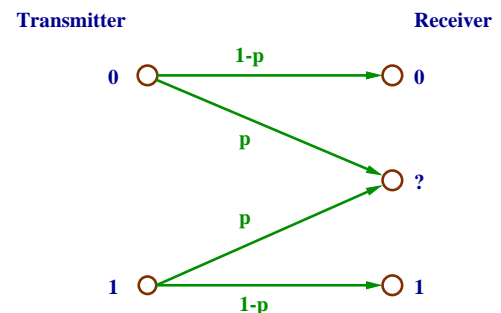
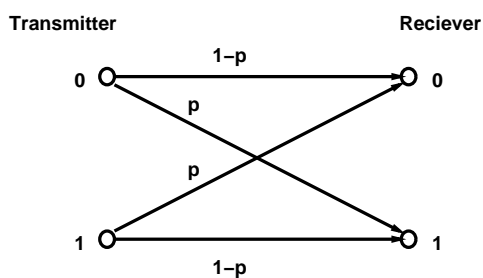
- Simple and fast decoding algorithms
- Rate of noise close to optimal
- But noise is random, not adversarial.

Recall Shannon Capacity

Other channels

Binary symmetric channel

Erasure channel



Capacity = $1 - H(p)$, i.e.,
 $\forall \epsilon > 0, \exists$ code of rate $1 - H(p) - \epsilon$,
s.t. if we transmit using code,
and decode from $\approx (p + \epsilon)$ -fraction errors,
then recover message w.p. $1 - \exp(-n)$.

Capacity = $1 - p$

How to decode so much?

AWGN Channel

(Additive White Gaussian Noise)

- Transmitted sequence in $\{-1, +1\}^n$.
- Received coordinates in \mathbb{R}^n
 i th rec'd element equals
 Transmitted number + e_i
 Where e_i is Gaussian r.v.
 with mean 0 and variance σ^2 .

Note: σ replaces the parameter p .

- Polynomial time encoding and decoding up to capacity on Binary Symmetric Channel. (Original motivation of [Forney].)
- Linear time encoding and decoding up to capacity on Binary Symmetric Channel. (Using [Spielman].)
- A simple linear time encoding and decoding algorithm for the erasure channel. (Due to [Luby, Mitzenmacher, Shokrollahi, Spielman, Stemann].)

Forney Codes

Fix BSC parameter p and $\epsilon > 0$.The code

- Let C_1 be $[n, (1 - \epsilon)n, \epsilon n]_n$ RS code.
- Let C_2 be $[\ell, (1 - H(p + \epsilon))\ell, (p + \epsilon)\ell]_2$ code with n messages. (i.e., $\ell \approx \log n$.)
- Let $C = C_1 \circ C_2$ be their concatenation.
- Transmit messages using C .

Its Parameters

- Block length $N = n\ell$.
- Rate = $(1 - \epsilon)(1 - H(p + \epsilon)) \approx (1 - H(p))$.
- Distance Rate = ϵp .

Decoding Forney Codes

Simple Decoding algorithm**Step 1** Decode each inner block using Brute Force in time 2^ℓ .**Step 2** Decode outer code using RS decoder.

- But distance is pathetic!
- Why is this any good?
- (Work in Inf. Th. \Rightarrow Know your probability.)

Analysis: Motivation

- Errors are random, so they are evenly distributed.
- Most blocks contain only p -fraction errors.
- Most errors caught by inner decoder.
- Outer decoder comes in to clean up.

Analysis: Formally

1. For any fixed inner block:
 $\Pr[\# \text{ errors} \geq (p + \epsilon)\ell] \leq 2^{-\delta\ell}$.
Call the bad event above a decoding failure.
2. $\Pr[\# \text{ decoding failures} \geq \epsilon n/2] \leq 2^{-\gamma N}$.
($\gamma > 0$ depends on ϵ and is called the error exponent.)
3. If event in (2) doesn't happen, then decode successfully!

Thm: \exists codes with rate $1 - H(p) - \epsilon$
with polytime encoders and decoders,
with decoding error prob. $\exp(-n)$
on the BSC with parameter p .

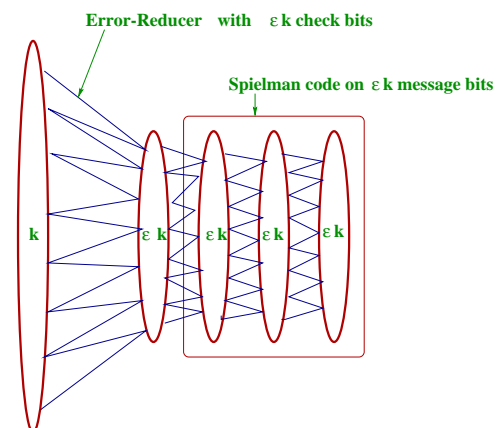
Moving on

- [Forney]'s work is from 1966.
- Introduced all the above ideas (concatenation, decoding, error analysis) and more.
- In fact also introduced GMD - why? To improve the error exponent!
- Very careful analysis needed to see why GMD helps!

Next: Linear time encoding and decoding.

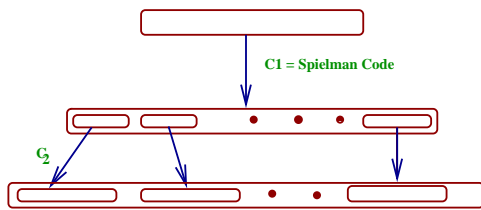
Aside: High-rate Spielman codes

As described [Spielman] codes had rate $1/4$.
But can also get codes of high rates.



Thm: $\forall \epsilon > 0, \exists \delta > 0$ and codes of rate $1 - \epsilon$
that are linear-time encodable and decodable
up to δ fraction errors.

Encoding:



- Given k message bits.
- Encode using Spielman codes of rate $1 - \epsilon$.
- Partition into blocks with $\approx \frac{1}{\epsilon^2}$ bits.
- Encode blocks using random inner code of rate $\approx 1 - H(p)$.

- As usual decode inner blocks and then decode outer block.
- Prob. of inner decoding failure is small constant.
- Prob. that # of inner decoding failures is twice the expectation is exponentially small.

Thm: \exists codes with rate $1 - H(p) - \epsilon$ with linear-time encoders and decoders, with decoding error prob. $\exp(-n)$ on the BSC with parameter p .

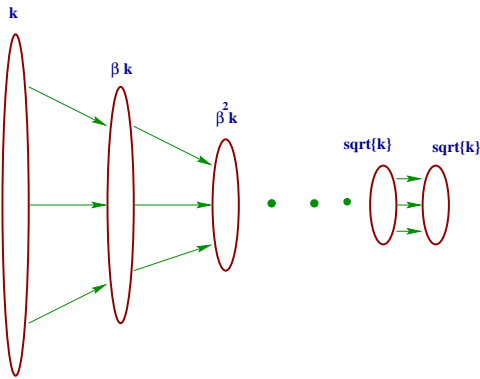
Towards Practice

- Last theorem seems best possible theoretically.
- Not so good for practice.
- Needs large block lengths!
- Running time is actually $O(2^{\frac{1}{\epsilon^2}} \cdot n)$.
- While, can hope for running time of $O(n \cdot \text{poly} \log \frac{1}{\epsilon})!$

Recent developments

- Turbo codes + decoding:
 - Simple codes + decoding algorithms, giving good results in simulations. [Benedetto, Montorsi, Thitijima].
 - But no analysis?
- Low-density Parity Check Codes:
 - Provably good performance.
 - Reach capacity on erasure channel [LMSSS].
 - Come close on error channel. [LMSS, Richardson+Urbanke, ...].

Binary erasure channel with parameter β .



Main idea:

- Cascade seq. of "Error-reduction" codes.
- This helps correct the check bits first.
- Then correct message bits.

The construction

- Fix seq. of bipartite graphs G_1, G_2, \dots
- G_i has $\beta^{i-1}k$ left nodes and $\beta^i k$ right nodes.
- Identify right vertices of G_{i-1} with left vertices of G_i .
- Terminate when $\#$ vertices $\approx \sqrt{k}$
- Truncate with $O(n^2)$ -time decodable code.

Encoding

- Message sets values of k left nodes of G_1 .
- Encode left to right setting vertices to parity of their left neighbours.

Decoding

- Decode right to left.
- First decode final layer.
- Then, assuming all checkbits known for G_i , decode for "message" bits of G_i .
- Claim: Each layer fails with exponentially small probability.

Unspecified

- How are the graphs G_1, G_2, \dots , picked?
- How to decode them?

Will explain for G_1 .

- Assume all checkbits known.
- Delete vertices corr. to message bits that are not erased, and incident edges.
- Iterate the following steps:
 - If \exists edge (m, c) in residual graph, with c having degree one, then
 - Set m to be parity of c with ngbrs of c (in original graph).
 - Delete m and c from residual graph.
- Stop when no such vertex exists.

Properties of Cascade codes

- Rate = $1 - \beta$.
- If graphs have linear number of edges, then encodable and decodable in linear time.
- Correct from β -fraction erasures, with all but exponentially small error probability, assuming the bipartite graphs can be constructed.

The bipartite graphs

- Option 1: Go the [Sipser+Spielman] route. (c, d) -regular graph with expansion $> c/2$.
This is good to correct small # fraction of errors, but not close to capacity.
- Regular graphs seem to be no good!
- Irregular degree graphs work!
Key innovation of [LMSSS].

The bipartite graphs (contd.)

- Pick a degree sequence $\{\lambda_i\}_i, \{\rho_i\}_i$, where λ_i (resp. ρ_i) denotes fraction of edges of left (resp. right) degree i .
- Let G_i 's be random graphs with this degree pattern on appropriate # of edges.
- Rate condition: Degree seq. must satisfy

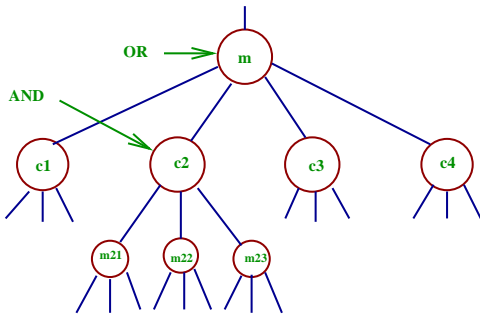
$$\frac{\sum_{i \geq 1} \lambda_i / i}{\sum_{i \geq 1} \rho_i / i} = \beta$$

- Analyze as a function of the degree sequences.

Analysis via And-Or trees

- Say, decode in rounds: Delete all degree 1 edges simultaneously etc.
- Fix edge m, c . What is the prob. that this edge is not deleted by the ℓ th round?
 1. m must be an erasure. AND
 2. \exists check bit c_j s.t. for all m_{jk} adjacent to c_j (other than m), m_{jk} not deleted by round $\ell - 1$.
- Analysis leads to an "And-Or Tree" [LMS]. (assume no short cycles in graph).

And-Or trees



Let $q_\ell = \text{Prob. of failure after } \ell \text{ rounds. Then}$

$$q_\ell \approx \beta \left(1 - \sum_i \lambda_i \left(1 - \sum_j \rho_j q_{\ell-1}^{j-1} \right)^{i-1} \right)$$

(Above informal: formal analysis hairier.)

Analysis (contd.)

Some compact notation:

Represent degree sequences by polynomials

$$\lambda(x) = \sum_{i \geq 1} \lambda_i x^{i-1}$$

$$\text{and } \rho(x) = \sum_{i \geq 1} \rho_i x^{i-1}.$$

Then $q_\ell = \beta(1 - \lambda(1 - \rho(q_{\ell-1})))$

When is decoding going to be successful?

If $q_\ell < q_{\ell-1}$.

Happens if

$$\beta(1 - \lambda(1 - \rho(x))) < x, \quad \forall x \in (0, \beta).$$

Degree sequences?

- Given a degree sequence, can tell if it is good enough by previous analysis.
- How to find one?
 - [LMSSS] give good sequence:
 - λ_i proportional to $1/i$, up to max degree D .
 - ρ_i 's give Poisson distribution, with mean adjusted so as to satisfy rate condition.
- Note: Analysis only works for constant # of rounds. To finish off, add a Sipser-Spielman like analysis.

Theorem: Have linear time ($O(n \ln \frac{1}{\epsilon})$) encodable and decodable codes achieving capacity on binary erasure channel.

Extending to BSC

- For the Binary Symmetric Channel, decoding algorithm has to change:
- Use a "Belief-Propagation" algorithm.
 - Maintain estimate (on edges) of prob. that incident message bit is 1.
 - On even rounds average the edges at the message end.
 - On odd rounds update the probability on the edges based on check bits.
- [LMSS], Richardson+Urbanke] prove that some degree sequences do very well.
- No analytic forms known on degrees. Numerically results come close to capacity (but not arbitrarily close.)

Conclusion

- Linear time decoding is an important feature in practice.
- Theoretically good analysis has resulted in good influence on practice.