

# Decoding methods

In **coding theory**, **decoding** is the process of translating received messages into **codewords** of a given **code**. There have been many common methods of mapping messages to codewords. These are often used to recover messages sent over a **noisy channel**, such as a **binary symmetric channel**.

## 1 Notation

$C \subset \mathbb{F}_2^n$  is considered a **binary code** with the length  $n$ ;  $x, y$  shall be elements of  $\mathbb{F}_2^n$ ; and  $d(x, y)$  is the distance between those elements.

## 2 Ideal observer decoding

One may be given the message  $x \in \mathbb{F}_2^n$ , then **ideal observer decoding** generates the codeword  $y \in C$ . The process results in this solution:

$$\mathbb{P}(y \text{ sent} \mid x \text{ received})$$

For example, a person can choose the codeword  $y$  that is most likely to be received as the message  $x$  after transmission.

### 2.1 Decoding conventions

Each codeword does not have an expected possibility: there may be more than one codeword with an equal likelihood of mutating into the received message. In such a case, the sender and receiver(s) must agree ahead of time on a decoding convention. Popular conventions include:

1. Request that the codeword be resent -- **automatic repeat-request**
2. Choose any random codeword from the set of most likely codewords which is nearer to that.
3. If **another code follows**, mark the ambiguous bits of the codeword as erasures and hope that the outer code disambiguates them

## 3 Maximum likelihood decoding

Further information: **Maximum likelihood**

Given a received codeword  $x \in \mathbb{F}_2^n$  **maximum likelihood decoding** picks a codeword  $y \in C$  that **maximizes**

$$\mathbb{P}(x \text{ received} \mid y \text{ sent})$$

that is, the codeword  $y$  that maximizes the probability that  $x$  was received, **given that**  $y$  was sent. If all codewords are equally likely to be sent then this scheme is equivalent to ideal observer decoding. In fact, by **Bayes Theorem**,

$$\begin{aligned}\mathbb{P}(x \text{ received} \mid y \text{ sent}) &= \frac{\mathbb{P}(x \text{ received}, y \text{ sent})}{\mathbb{P}(y \text{ sent})} \\ &= \mathbb{P}(y \text{ sent} \mid x \text{ received}) \cdot \frac{\mathbb{P}(x \text{ received})}{\mathbb{P}(y \text{ sent})}.\end{aligned}$$

Upon fixing  $\mathbb{P}(x \text{ received})$ ,  $x$  is restructured and  $\mathbb{P}(y \text{ sent})$  is constant as all codewords are equally likely to be sent. Therefore  $\mathbb{P}(x \text{ received} \mid y \text{ sent})$  is maximised as a function of the variable  $y$  precisely when  $\mathbb{P}(y \text{ sent} \mid x \text{ received})$  is maximised, and the claim follows.

As with ideal observer decoding, a convention must be agreed to for non-unique decoding.

The maximum likelihood decoding problem can also be modeled as an **integer programming problem**.<sup>[1]</sup>

The maximum likelihood decoding algorithm is an instance of the “marginalize a product function” problem which is solved by applying the **generalized distributive law**.<sup>[2]</sup>

## 4 Minimum distance decoding

Given a received codeword  $x \in \mathbb{F}_2^n$ , **minimum distance decoding** picks a codeword  $y \in C$  to minimise the **Hamming distance**:

$$d(x, y) = \#\{i : x_i \neq y_i\}$$

i.e. choose the codeword  $y$  that is as close as possible to  $x$ .

Note that if the probability of error on a **discrete memory-less channel**  $p$  is strictly less than one half, then *minimum distance decoding* is equivalent to *maximum likelihood decoding*, since if

$$d(x, y) = d,$$

then:

$$\begin{aligned}\mathbb{P}(y \text{ received} \mid x \text{ sent}) &= (1 - p)^{n-d} \cdot p^d \\ &= (1 - p)^n \cdot \left(\frac{p}{1 - p}\right)^d\end{aligned}$$

which (since  $p$  is less than one half) is maximised by minimising  $d$ .

Minimum distance decoding is also known as *nearest neighbour decoding*. It can be assisted or automated by using a **standard array**. Minimum distance decoding is a reasonable decoding method when the following conditions are met:

1. The probability  $p$  that an error occurs is independent of the position of the symbol
2. Errors are independent events - an error at one position in the message does not affect other positions

These assumptions may be reasonable for transmissions over a **binary symmetric channel**. They may be unreasonable for other media, such as a DVD, where a single scratch on the disk can cause an error in many neighbouring symbols or codewords.

As with other decoding methods, a convention must be agreed to for non-unique decoding.

## 5 Syndrome decoding

**Syndrome decoding** is a highly efficient method of decoding a **linear code** over a *noisy channel*, i.e. one on which errors are made. In essence, syndrome decoding is *minimum distance decoding* using a reduced lookup table. This is allowed by the linearity of the code.<sup>[3]</sup>

Suppose that  $C \subset \mathbb{F}_2^n$  is a linear code of length  $n$  and minimum distance  $d$  with **parity-check matrix**  $H$ . Then clearly  $C$  is capable of correcting up to

$$t = \left\lfloor \frac{d-1}{2} \right\rfloor$$

errors made by the channel (since if no more than  $t$  errors are made then minimum distance decoding will still correctly decode the incorrectly transmitted codeword).

Now suppose that a codeword  $x \in \mathbb{F}_2^n$  is sent over the channel and the error pattern  $e \in \mathbb{F}_2^n$  occurs. Then  $z = x + e$  is received. Ordinary minimum distance decoding would lookup the vector  $z$  in a table of size  $|C|$  for the nearest match - i.e. an element (not necessarily unique)  $c \in C$  with

$$d(c, z) \leq d(y, z)$$

for all  $y \in C$ . Syndrome decoding takes advantage of the property of the parity matrix that:

$$Hx = 0$$

for all  $x \in C$ . The *syndrome* of the received  $z = x + e$  is defined to be:

$$Hz = H(x + e) = Hx + He = 0 + He = He$$

To perform ML decoding in a **Binary symmetric channel**, one has to look-up a precomputed table of size  $2^{n-k}$ , mapping  $He$  to  $e$ .

Note that this is already of significantly less complexity than that of a **Standard array decoding**.

However, under the assumption that no more than  $t$  errors were made during transmission, the receiver can look up the value  $He$  in a further reduced table of size

$$\sum_{i=0}^t \binom{n}{i} < |C|$$

only (for a binary code). The table is against pre-computed values of  $He$  for all possible error patterns  $e \in \mathbb{F}_2^n$ .

Knowing what  $e$  is, it is then trivial to decode  $x$  as:

$$x = z - e$$

## 6 Partial response maximum likelihood

Main article: **PRML**

Partial response maximum likelihood (PRML) is a method for converting the weak analog signal from the head of a magnetic disk or tape drive into a digital signal.

## 7 Viterbi decoder

Main article: **Viterbi decoder**

A Viterbi decoder uses the Viterbi algorithm for decoding a bitstream that has been encoded using **forward error correction** based on a convolutional code. The **Hamming distance** is used as a metric for hard decision Viterbi decoders. The *squared Euclidean distance* is used as a metric for soft decision decoders.

## 8 See also

- Error detection and correction

## 9 Sources

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## 10 References

- [1] Feldman, Jon; Wainwright, Martin J.; Karger, David R. (March 2005). "Using Linear Programming to Decode Binary Linear Codes". *IEEE Transactions on Information Theory*. **51** (3). pp. 954–972. doi:10.1109/TIT.2004.842696.
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- [3] Albrecht Beutelspacher & Ute Rosenbaum (1998) *Projective Geometry*, page 190, Cambridge University Press ISBN 0-521-48277-1

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