A slightly more careful answer (short of explicit computation) goes as follows. Taking the approximation for \( \binom{N}{K} \) to the next order, we find:

\[
\binom{N}{N/2} \simeq 2^N \frac{1}{\sqrt{2\pi N/4}}.
\]

(1.40)

This approximation can be proved from an accurate version of Stirling’s approximation (1.12), or by considering the binomial distribution with \( p = 1/2 \) and noting

\[
1 = \sum_K \binom{N}{K} 2^{-N} \simeq 2^{-N} \sum_{r=-N/2}^{N/2} e^{-r^2/2\sigma^2} \simeq 2^{-N} \left( \frac{N}{N/2} \right) \sqrt{2\pi \sigma},
\]

(1.41)

where \( \sigma = \sqrt{N/4} \), from which equation (1.40) follows. The distinction between \( \lceil N/2 \rceil \) and \( N/2 \) is not important in this term since \( \binom{N}{K} \) has a maximum at \( K = N/2 \).

Then the probability of error (for odd \( N \)) is to leading order

\[
p_b \simeq \left( \frac{N}{(N+1)/2} \right) f^{(N+1)/2}(1-f)^{(N-1)/2}
\]

(1.42)

\[
\simeq 2^N \frac{1}{\sqrt{\pi N/2}} f \left( f(1-f) \right)^{(N-1)/2} \simeq \frac{1}{\sqrt{\pi N/8}} f \left( 4f(1-f) \right)^{(N-1)/2}.
\]

(1.43)

The equation \( p_b = 10^{-15} \) can be written

\[
\frac{(N-1)/2}{\log 10^{-15} + \log \frac{\sqrt{\pi N/8}}{f}} \simeq \frac{\log 4}{\log f(1-f)}
\]

(1.44)

which may be solved for \( N \) iteratively, the first iteration starting from \( \hat{N}_1 = 68 \):

\[
\frac{\hat{N}_2 - 1}{2} \simeq \frac{-15 + 1.7}{-0.44} = 29.9 \Rightarrow \hat{N}_2 \simeq 60.9.
\]

(1.45)

This answer is found to be stable, so \( N \simeq 61 \) is the blocklength at which \( p_b \simeq 10^{-15} \).

Solution to exercise 1.6 (p.13).

(a) The probability of block error of the Hamming code is a sum of six terms – the probabilities that 2, 3, 4, 5, 6, or 7 errors occur in one block.

\[
p_B = \sum_{r=2}^{7} \binom{T}{r} f^r(1-f)^{7-r}.
\]

(1.46)

To leading order, this goes as

\[
p_B \simeq \binom{7}{2} f^2 = 21f^2.
\]

(1.47)

(b) The probability of bit error of the Hamming code is smaller than the probability of block error because a block error rarely corrupts all bits in the decoded block. The leading-order behaviour is found by considering the outcome in the most probable case where the noise vector has weight two. The decoder will erroneously flip a third bit, so that the modified received vector (of length 7) differs in three bits from the transmitted vector. That means, if we average over all seven bits, the probability that a randomly chosen bit is flipped is 3/7 times the block error probability, to leading order. Now, what we really care about is the probability that
a source bit is flipped. Are parity bits or source bits more likely to be among these three flipped bits, or are all seven bits equally likely to be corrupted when the noise vector has weight two? The Hamming code is in fact completely symmetric in the protection it affords to the seven bits (assuming a binary symmetric channel). [This symmetry can be proved by showing that the role of a parity bit can be exchanged with a source bit and the resulting code is still a \((7, 4)\) Hamming code; see below.] The probability that any one bit ends up corrupted is the same for all seven bits. So the probability of bit error (for the source bits) is simply three sevenths of the probability of block error.

\[ p_b \approx \frac{3}{7} p_B \approx 9 f^2. \] (1.48)

**Symmetry of the Hamming \((7, 4)\) code**

To prove that the \((7, 4)\) code protects all bits equally, we start from the parity-check matrix

\[
H = \begin{bmatrix}
1 & 1 & 1 & 0 & 1 & 0 & 0 \\
0 & 1 & 1 & 1 & 0 & 1 & 0 \\
1 & 0 & 1 & 1 & 0 & 0 & 1
\end{bmatrix}.
\] (1.49)

The symmetry among the seven transmitted bits will be easiest to see if we reorder the seven bits using the permutation \((t_1 t_2 t_3 t_4 t_5 t_6 t_7) \rightarrow (t_5 t_2 t_3 t_4 t_1 t_6 t_7)\). Then we can rewrite \(H\) thus:

\[
H = \begin{bmatrix}
1 & 1 & 1 & 0 & 1 & 0 & 0 \\
0 & 1 & 1 & 1 & 0 & 1 & 0 \\
0 & 0 & 1 & 1 & 1 & 0 & 1
\end{bmatrix}.
\] (1.50)

Now, if we take any two parity constraints that \(t\) satisfies and add them together, we get another parity constraint. For example, row 1 asserts \(t_5 + t_2 + t_3 + t_1 = \text{even}\), and row 2 asserts \(t_2 + t_3 + t_4 + t_6 = \text{even}\), and the sum of these two constraints is

\[ t_5 + 2t_2 + 2t_3 + t_1 + t_4 + t_6 = \text{even}; \] (1.51)

we can drop the terms \(2t_2\) and \(2t_3\), since they are even whatever \(t_2\) and \(t_3\) are; thus we have derived the parity constraint \(t_5 + t_1 + t_4 + t_6 = \text{even}\), which we can if we wish add into the parity-check matrix as a fourth row. [The set of vectors satisfying \(Ht = 0\) will not be changed.] We thus define

\[
H' = \begin{bmatrix}
1 & 1 & 1 & 0 & 1 & 0 & 0 \\
0 & 1 & 1 & 1 & 0 & 1 & 0 \\
0 & 0 & 1 & 1 & 1 & 0 & 1 \\
1 & 0 & 0 & 1 & 1 & 1 & 0
\end{bmatrix}.
\] (1.52)

The fourth row is the sum (modulo two) of the top two rows. Notice that the second, third, and fourth rows are all cyclic shifts of the top row. If, having added the fourth redundant constraint, we drop the first constraint, we obtain a new parity-check matrix \(H''\),

\[
H'' = \begin{bmatrix}
0 & 1 & 1 & 1 & 0 & 1 & 0 \\
0 & 0 & 1 & 1 & 1 & 0 & 1 \\
1 & 0 & 0 & 1 & 1 & 1 & 0
\end{bmatrix},
\] (1.53)

which still satisfies \(H''t = 0\) for all codewords, and which looks just like the starting \(H\) in (1.50), except that all the columns have shifted along one
1.6: Solutions
to the right, and the rightmost column has reappeared at the left (a cyclic permutation of the columns).

This establishes the symmetry among the seven bits. Iterating the above procedure five more times, we can make a total of seven different $H$ matrices for the same original code, each of which assigns each bit to a different role.

We may also construct the super-redundant seven-row parity-check matrix for the code,

$$
H'' = \begin{bmatrix}
1 & 1 & 1 & 0 & 1 & 0 & 0 \\
0 & 1 & 1 & 1 & 0 & 1 & 0 \\
0 & 0 & 1 & 1 & 1 & 0 & 1 \\
1 & 0 & 0 & 1 & 1 & 1 & 0 \\
0 & 1 & 0 & 0 & 1 & 1 & 1 \\
1 & 0 & 1 & 0 & 0 & 1 & 1 \\
1 & 1 & 0 & 1 & 0 & 0 & 1 \\
\end{bmatrix} \quad (1.54)
$$

This matrix is ‘redundant’ in the sense that the space spanned by its rows is only three-dimensional, not seven.

This matrix is also a cyclic matrix. Every row is a cyclic permutation of the top row.

**Cyclic codes:** if there is an ordering of the bits $t_1 \ldots t_N$ such that a linear code has a cyclic parity-check matrix, then the code is called a cyclic code.

The codewords of such a code also have cyclic properties: any cyclic permutation of a codeword is a codeword.

For example, the Hamming $(7, 4)$ code, with its bits ordered as above, consists of all seven cyclic shifts of the codewords 1110100 and 1011000, and the codewords 0000000 and 1111111.

Cyclic codes are a cornerstone of the algebraic approach to error-correcting codes. We won’t use them again in this book, however, as they have been superseded by sparse-graph codes (Part VI).

**Solution to exercise 1.7 (p.13).** There are fifteen non-zero noise vectors which give the all-zero syndrome; these are precisely the fifteen non-zero codewords of the Hamming code. Notice that because the Hamming code is linear, the sum of any two codewords is a codeword.

**Graphs corresponding to codes**

**Solution to exercise 1.9 (p.14).** When answering this question, you will probably find that it is easier to invent new codes than to find optimal decoders for them. There are many ways to design codes, and what follows is just one possible train of thought. We make a linear block code that is similar to the $(7, 4)$ Hamming code, but bigger.

Many codes can be conveniently expressed in terms of graphs. In figure 1.13, we introduced a pictorial representation of the $(7, 4)$ Hamming code. If we replace that figure’s big circles, each of which shows that the parity of four particular bits is even, by a ‘parity-check node’ that is connected to the four bits, then we obtain the representation of the $(7, 4)$ Hamming code by a bipartite graph as shown in figure 1.20. The 7 circles are the 7 transmitted bits. The 3 squares are the parity-check nodes (not to be confused with the 3 parity-check bits, which are the three most peripheral circles). The graph is a ‘bipartite’ graph because its nodes fall into two classes – bits and checks.
and there are edges only between nodes in different classes. The graph and the code’s parity-check matrix (1.30) are simply related to each other: each parity-check node corresponds to a row of \( \mathbf{H} \) and each bit node corresponds to a column of \( \mathbf{H} \); for every 1 in \( \mathbf{H} \), there is an edge between the corresponding pair of nodes.

Having noticed this connection between linear codes and graphs, one way to invent linear codes is simply to think of a bipartite graph. For example, a pretty bipartite graph can be obtained from a dodecahedron by calling the vertices of the dodecahedron the parity-check nodes, and putting a transmitted bit on each edge in the dodecahedron. This construction defines a parity-check matrix in which every column has weight 2 and every row has weight 3. [The weight of a binary vector is the number of 1s it contains.]

This code has \( N = 30 \) bits, and it appears to have \( M_{\text{apparent}} = 20 \) parity-check constraints. Actually, there are only \( M = 19 \) independent constraints; the 20th constraint is redundant (that is, if 19 constraints are satisfied, then the 20th is automatically satisfied); so the number of source bits is \( K = N - M = 11 \). The code is a (30, 11) code.

It is hard to find a decoding algorithm for this code, but we can estimate its probability of error by finding its lowest-weight codewords. If we flip all the bits surrounding one face of the original dodecahedron, then all the parity checks will be satisfied; so the code has 12 codewords of weight 5, one for each face. Since the lowest-weight codewords have weight 5, we say that the code has distance \( d = 5 \); the (7, 4) Hamming code had distance 3 and could correct all single bit-flip errors. A code with distance 5 can correct all double bit-flip errors, but there are some triple bit-flip errors that it cannot correct. So the error probability of this code, assuming a binary symmetric channel, will be dominated, at least for low noise levels \( f \), by a term of order \( f^3 \), perhaps something like

\[
12 \binom{5}{3} f^3 (1 - f)^{27}. \tag{1.55}
\]

Of course, there is no obligation to make codes whose graphs can be represented on a plane, as this one can; the best linear codes, which have simple graphical descriptions, have graphs that are more tangled, as illustrated by the tiny (16, 4) code of figure 1.22.

Furthermore, there is no reason for sticking to linear codes; indeed some nonlinear codes — codes whose codewords cannot be defined by a linear equation like \( \mathbf{H} \mathbf{t} = \mathbf{0} \) — have very good properties. But the encoding and decoding of a nonlinear code are even trickier tasks.

Solution to exercise 1.10 (p.14). First let’s assume we are making a linear code and decoding it with syndrome decoding. If there are \( N \) transmitted bits, then the number of possible error patterns of weight up to two is

\[
\binom{N}{2} + \binom{N}{1} + \binom{N}{0}. \tag{1.56}
\]

For \( N = 14 \), that’s 91 + 14 + 1 = 106 patterns. Now, every distinguishable error pattern must give rise to a distinct syndrome; and the syndrome is a list of \( M \) bits, so the maximum possible number of syndromes is \( 2^M \). For a (14, 8) code, \( M = 6 \), so there are at most \( 2^6 = 64 \) syndromes. The number of possible error patterns of weight up to two, 106, is bigger than the number of syndromes, 64, so we can immediately rule out the possibility that there is a (14, 8) code that is 2-error-correcting.
The same counting argument works fine for nonlinear codes too. When
the decoder receives \( r = t + n \), his aim is to deduce both \( t \) and \( n \) from \( r \). If
it is the case that the sender can select any transmission \( t \) from a code of size
\( S_t \), and the channel can select any noise vector from a set of size \( S_n \), and those
two selections can be recovered from the received bit string \( r \), which is one of
at most \( 2^N \) possible strings, then it must be the case that
\[
S_t S_n \leq 2^N. \tag{1.57}
\]
So, for a \((N, K)\) two-error-correcting code, whether linear or nonlinear,
\[
2^K \left[ \binom{N}{2} + \binom{N}{1} + \binom{N}{0} \right] \leq 2^N. \tag{1.58}
\]

**Solution to exercise 1.11** (p.14). There are various strategies for making codes
that can correct multiple errors, and I strongly recommend you think out one
or two of them for yourself.

If your approach uses a linear code, e.g., one with a collection of \( M \) parity
checks, it is helpful to bear in mind the counting argument given in the previous
exercise, in order to anticipate how many parity checks, \( M \), you might need.

Examples of codes that can correct any two errors are the \((30, 11)\) dodeca-
hedron code on page 20, and the \((15, 6)\) pentagonful code to be introduced on
p.221. Further simple ideas for making codes that can correct multiple errors
from codes that can correct only one error are discussed in section 13.7.

**Solution to exercise 1.12** (p.16). The probability of error of \( R_2^3 \) is, to leading
order,
\[
p_h(R_2^3) \simeq 3 [p_h(R_3)]^2 = 3(3 f^2)^2 + \cdots = 27 f^4 + \cdots, \tag{1.59}
\]
whereas the probability of error of \( R_9 \) is dominated by the probability of five
flips,
\[
p_h(R_9) \simeq \binom{9}{5} f^5 (1 - f)^4 \simeq 126 f^5 + \cdots. \tag{1.60}
\]
The \( R_2^3 \) decoding procedure is therefore suboptimal, since there are noise vec-
tors of weight four that cause it to make a decoding error.

It has the advantage, however, of requiring smaller computational re-
sources: only memorization of three bits, and counting up to three, rather
than counting up to nine.

This simple code illustrates an important concept. Concatenated codes
are widely used in practice because concatenation allows large codes to be
implemented using simple encoding and decoding hardware. Some of the best
known practical codes are concatenated codes.