

Figure 12.4: Illustration of a simple hierarchy of word categories, with 8 words w_0, \ldots, w_7 organized into a three level hierarchy. The leaves of the tree represent actual specific words. Internal nodes represent groups of words. Any node can be indexed by the sequence of binary decisions (0=left, 1=right) to reach the node from the root. Super-class (0) contains the classes (0,0) and (0,1), which respectively contain the sets of words $\{w_0,w_1\}$ and $\{w_2, w_3\}$, and similarly super-class (1) contains the classes (1,0) and (1,1), which respectively contain the words (w_4, w_5) and (w_6, w_7) . If the tree is sufficiently balanced, the maximum depth (number of binary decisions) is on the order of the logarithm of the number of words |V|: the choice of one out of |V| words can be obtained by doing $O(\log |V|)$ operations (one for each of the nodes on the path from the root). In this example, computing the probability of a word y can be done by multiplying three probabilities, associated with the binary decisions to move left or right at each node on the path from the root to a node y. Let $b_i(y)$ be the i-th binary decision when traversing the tree towards the value y. The probability of sampling an output y decomposes into a product of conditional probabilities, using the chain rule for conditional probabilities, with each node indexed by the prefix of these bits. For example, node (1,0) corresponds to the prefix $(b_0(w_4) = 1, b_1(w_4) = 0)$, and the probability of w_4 can be decomposed as follows:

$$P(y = w_4) = P(b_0 = 1, b_1 = 0, b_2 = 0)$$
(12.11)

$$= P(\mathbf{b}_0 = 1)P(\mathbf{b}_1 = 0 \mid \mathbf{b}_0 = 1)P(\mathbf{b}_2 = 0 \mid \mathbf{b}_0 = 1, \mathbf{b}_1 = 0). \tag{12.12}$$