22. Greedy Algorithms

Fractional Knapsack Problem, Huffman Coding [Cormen et al, Kap. 16.1, 16.3]

The Fractional Knapsack Problem

set of $n \in \mathbb{N}$ items $\{1, \ldots, n\}$ Each item i has value $v_i \in \mathbb{N}$ and weight $w_i \in \mathbb{N}$. The maximum weight is given as $W \in \mathbb{N}$. Input is denoted as $E = (v_i, w_i)_{i=1,\ldots,n}$.

Wanted: Fractions $0 \le q_i \le 1$ ($1 \le i \le n$) that maximise the sum $\sum_{i=1}^{n} q_i \cdot v_i$ under $\sum_{i=1}^{n} q_i \cdot w_i \le W$.

Greedy heuristics

Sort the items decreasingly by value per weight v_i/w_i.

Assumption $v_i/w_i \geq v_{i+1}/w_{i+1}$

Let $j = \max\{0 \leq k \leq n : \sum_{i=1}^{k} w_i \leq W\}$. Set

- $q_i = 1$ for all $1 \leq i \leq j$.
- $q_{j+1} = \frac{W - \sum_{i=1}^{j} w_i}{w_{j+1}}$.
- $q_i = 0$ for all $i > j+1$.

That is fast: $\Theta(n \log n)$ for sorting and $\Theta(n)$ for the computation of the q_i.

Correctness

Assumption: optimal solution (r_i) $(1 \leq i \leq n)$.

The knapsack is full: $\sum_i r_i \cdot w_i = \sum_i q_i \cdot w_i = W$.

Consider k: smallest i with $r_i \neq q_i$ Definition of greedy: $q_k > r_k$. Let $x = q_k - r_k > 0$.

Construct a new solution (r_i'): $r_i' = r_i \forall i < k$. $r_k' = q_k$. Remove weight $\sum_{i=k+1}^{n} \delta_i = x \cdot w_k$ from items $k+1$ to n. This works because $\sum_{i=k}^{n} r_i \cdot w_i = \sum_{i=k}^{n} q_i \cdot w_i$.

Correctness

$$\sum_{i=k}^{n} r'_i v_i = r_k v_k + x w_k \frac{v_k}{w_k} + \sum_{i=k+1}^{n} (r_i w_i - \delta_i) \frac{v_i}{w_i}$$

$$\geq r_k v_k + x w_k \frac{v_k}{w_k} + \sum_{i=k+1}^{n} r_i w_i \frac{v_i}{w_i} - \delta_i \frac{v_k}{w_k}$$

$$= r_k v_k + x w_k \frac{v_k}{w_k} - x w_k \frac{v_k}{w_k} + \sum_{i=k+1}^{n} r_i w_i \frac{v_i}{w_i} = \sum_{i=k}^{n} r_i v_i.$$

Thus (r'_i) is also optimal. Iterative application of this idea generates the solution (q_i).

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File consisting of 100.000 characters from the alphabet $\{a, \ldots, f\}$.

	a	b	c	d	e	f
Frequency (Thousands)	45	13	12	16	9	5
Code word with fix length	000	001	010	011	100	101
Code word variable length	0	101	100	111	1101	1100

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File size (code with fix length): 300.000 bits.
File size (code with variable length): 224.000 bits.

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- Decoding simple because prefixcode
  $0110011001101 \rightarrow 0 \cdot 1100 \cdot 1100 \cdot 1101 \rightarrow affe$

Code trees
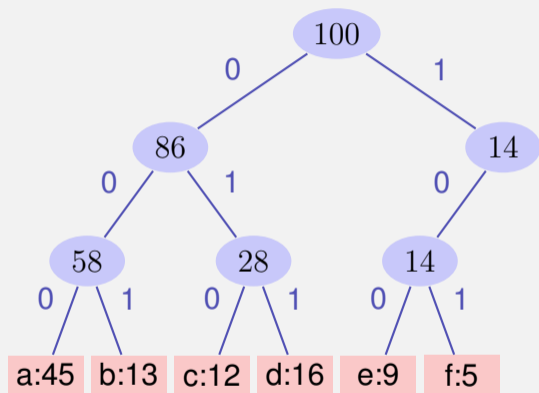
Code words with fixed length
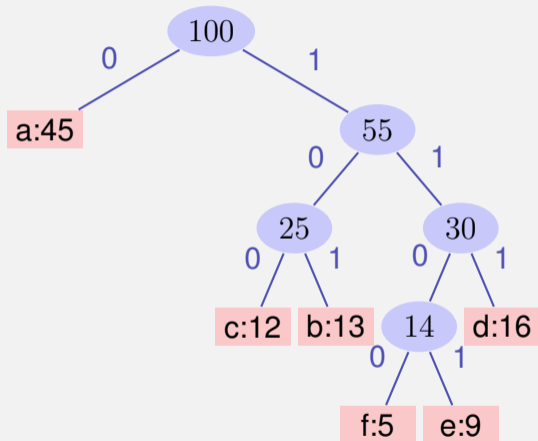
Code words with variable length

Properties of the Code Trees

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$$B(T) = \sum_{c \in C} f(c) \cdot d_T(c).$$

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In the following a code tree is called optimal when it minimizes the costs.

Algorithm Idea

Tree construction bottom up

- Start with the set C of code words
- Replace iteriatively the two nodes with smallest frequency by a new parent node.
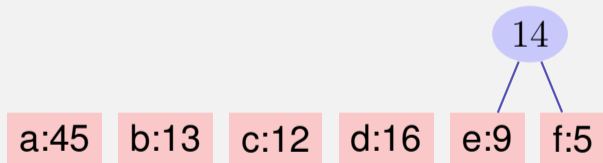
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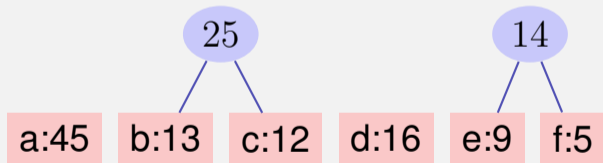
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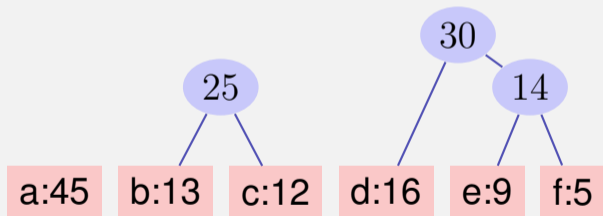
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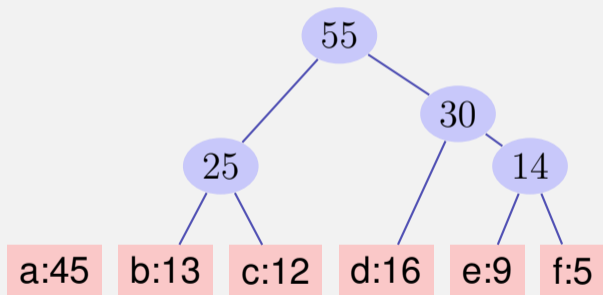Tree construction bottom up

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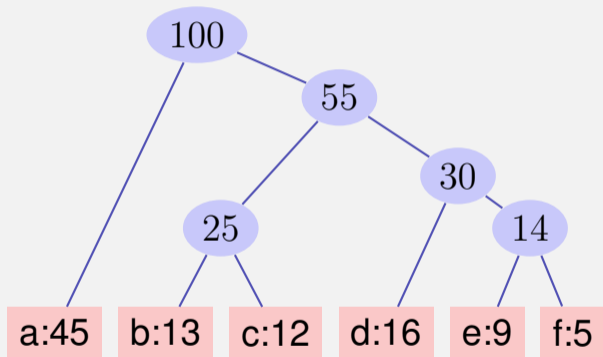
Algorithm Huffman(C)

Input: code words $c \in C$
Output: Root of an optimal code tree

$n \leftarrow |C|$
$Q \leftarrow C$
for $i = 1$ **to** $n - 1$ **do**
    allocate a new node z
    $z.\text{left} \leftarrow \text{ExtractMin}(Q)$ // extract word with minimal frequency.
    $z.\text{right} \leftarrow \text{ExtractMin}(Q)$
    $z.\text{freq} \leftarrow z.\text{left.freq} + z.\text{right.freq}$
    $\text{Insert}(Q, z)$
return $\text{ExtractMin}(Q)$

Analyse

Use a heap: build Heap in $\mathcal{O}(n)$. Extract-Min in $O(\log n)$ for n Elements. Yields a runtime of $O(n \log n)$.

The greedy approach is correct

Theorem

Let x, y be two symbols with smallest frequencies in C and let $T'(C')$ be an optimal code tree to the alphabet $C' = C - \{x, y\} + \{z\}$ with a new symbol z with $f(z) = f(x) + f(y)$. Then the tree $T(C)$ that is constructed from $T'(C')$ by replacing the node z by an inner node with children x and y is an optimal code tree for the alphabet C.

Proof

It holds that $f(x) \cdot d_T(x) + f(y) \cdot d_T(y) =$
$(f(x) + f(y)) \cdot (d_{T'}(z) + 1) = f(z) \cdot d_{T'}(x) + f(x) + f(y)$. Thus
$B(T') = B(T) - f(x) - f(y)$.

Assumption: T is not optimal. Then there is an optimal tree T'' with
$B(T'') < B(T)$. We assume that x and y are brothers in T''. Let T'''
be the tree where the inner node with children x and y is replaced by
z. Then it holds that
$B(T''') = B(T'') - f(x) - f(y) < B(T) - f(x) - f(y) = B(T')$.
Contradiction to the optimality of T'.

The assumption that x and y are brothers in T'' can be justified
because a swap of elements with smallest frequency to the lowest
level of the tree can at most decrease the value of B.