Solving this set of m+n equations in the m+n variables  $x^*$ ,  $\nu^*$  gives the optimal primal and dual variables for (5.50).

**Example 5.2** Water-filling. We consider the convex optimization problem

minimize 
$$-\sum_{i=1}^{n} \log(\alpha_i + x_i)$$
  
subject to  $x \succeq 0$ ,  $\mathbf{1}^T x = 1$ ,

where  $\alpha_i > 0$ . This problem arises in information theory, in allocating power to a set of n communication channels. The variable  $x_i$  represents the transmitter power allocated to the ith channel, and  $\log(\alpha_i + x_i)$  gives the capacity or communication rate of the channel, so the problem is to allocate a total power of one to the channels, in order to maximize the total communication rate.

Introducing Lagrange multipliers  $\lambda^* \in \mathbf{R}^n$  for the inequality constraints  $x^* \succeq 0$ , and a multiplier  $\nu^* \in \mathbf{R}$  for the equality constraint  $\mathbf{1}^T x = 1$ , we obtain the KKT conditions

$$x^* \succeq 0,$$
  $\mathbf{1}^T x^* = 1,$   $\lambda^* \succeq 0,$   $\lambda_i^* x_i^* = 0,$   $i = 1, \dots, n,$   $-1/(\alpha_i + x_i^*) - \lambda_i^* + \nu^* = 0,$   $i = 1, \dots, n.$ 

We can directly solve these equations to find  $x^*$ ,  $\lambda^*$ , and  $\nu^*$ . We start by noting that  $\lambda^*$  acts as a slack variable in the last equation, so it can be eliminated, leaving

$$x^* \succeq 0,$$
  $\mathbf{1}^T x^* = 1,$   $x_i^* (\nu^* - 1/(\alpha_i + x_i^*)) = 0,$   $i = 1, \dots, n,$   $\nu^* \ge 1/(\alpha_i + x_i^*),$   $i = 1, \dots, n.$ 

If  $\nu^* < 1/\alpha_i$ , this last condition can only hold if  $x_i^* > 0$ , which by the third condition implies that  $\nu^* = 1/(\alpha_i + x_i^*)$ . Solving for  $x_i^*$ , we conclude that  $x_i^* = 1/\nu^* - \alpha_i$  if  $\nu^* < 1/\alpha_i$ . If  $\nu^* \ge 1/\alpha_i$ , then  $x_i^* > 0$  is impossible, because it would imply  $\nu^* \ge 1/\alpha_i > 1/(\alpha_i + x_i^*)$ , which violates the complementary slackness condition. Therefore,  $x_i^* = 0$  if  $\nu^* \ge 1/\alpha_i$ . Thus we have

$$x_i^{\star} = \left\{ \begin{array}{ll} 1/\nu^{\star} - \alpha_i & \nu^{\star} < 1/\alpha_i \\ 0 & \nu^{\star} \geq 1/\alpha_i, \end{array} \right.$$

or, put more simply,  $x_i^* = \max\{0, 1/\nu^* - \alpha_i\}$ . Substituting this expression for  $x_i^*$  into the condition  $\mathbf{1}^T x^* = 1$  we obtain

$$\sum_{i=1}^{n} \max\{0, 1/\nu^{*} - \alpha_{i}\} = 1.$$

The lefthand side is a piecewise-linear increasing function of  $1/\nu^*$ , with breakpoints at  $\alpha_i$ , so the equation has a unique solution which is readily determined.

This solution method is called water-filling for the following reason. We think of  $\alpha_i$  as the ground level above patch i, and then flood the region with water to a depth  $1/\nu$ , as illustrated in figure 5.7. The total amount of water used is then  $\sum_{i=1}^{n} \max\{0, 1/\nu^* - \alpha_i\}$ . We then increase the flood level until we have used a total amount of water equal to one. The depth of water above patch i is then the optimal value  $x_i^*$ .