Let us reformulate the problem as

minimize 
$$f(y)$$
  
subject to  $Ax + b = y$ ,

where we introduced the new variable  $y \in \mathbb{R}^m$  and the equality constraint Ax + b = y. The two problems are obviously equivalent. The Lagrangian of the reformulated problem is

$$L(x, y, \mu) = f(y) + \mu^{\top} (Ax + b - y)$$

where  $\mu \in \mathbb{R}^m$ . To find the dual function  $G(\mu)$  we minimize  $L(x, y, \mu)$  over x and y. Minimizing over x we see that  $G(\mu) = -\infty$  unless  $A^{\top}\mu = 0$ , in which case we are left with

$$G(\mu) = b^{\top} \mu + \inf_{y} (f(y) - \mu^{\top} y) = b^{\top} \mu - \inf_{y} (\mu^{\top} y - f(y)) = b^{\top} \mu - f^{*}(\mu),$$

where  $f^*$  is the conjugate of f. It follows that the dual program can be expressed as

maximize 
$$b^{\top}\mu - f^*(\mu)$$
  
subject to  $A^{\top}\mu = 0$ .

This formulation of the dual is much more useful than the dual of the original program.

**Example 50.12.** As a concrete example, consider the following unconstrained program:

minimize 
$$f(x) = \log \left( \sum_{i=1}^{n} e^{(A^i)^{\top} x + b_i} \right)$$

where  $A^i$  is a column vector in  $\mathbb{R}^n$ . We reformulate the problem by introducing new variables and equality constraints as follows:

minimize 
$$f(y) = \log \left( \sum_{i=1}^{n} e^{y_i} \right)$$
  
subject to  $Ax + b = y$ ,

where A is the  $n \times n$  matrix whose columns are the vectors  $A^i$  and  $b = (b_1, \ldots, b_n)$ . Since by Example 50.8(8), the conjugate of the log-sum-exp function  $f(y) = \log \left(\sum_{i=1}^n e^{y_i}\right)$  is

$$f^*(\mu) = \begin{cases} \sum_{i=1}^n \mu_i \log \mu_i & \text{if } \mathbf{1}^\top \mu = 1 \text{ and } \mu \ge 0\\ \infty & \text{otherwise,} \end{cases}$$