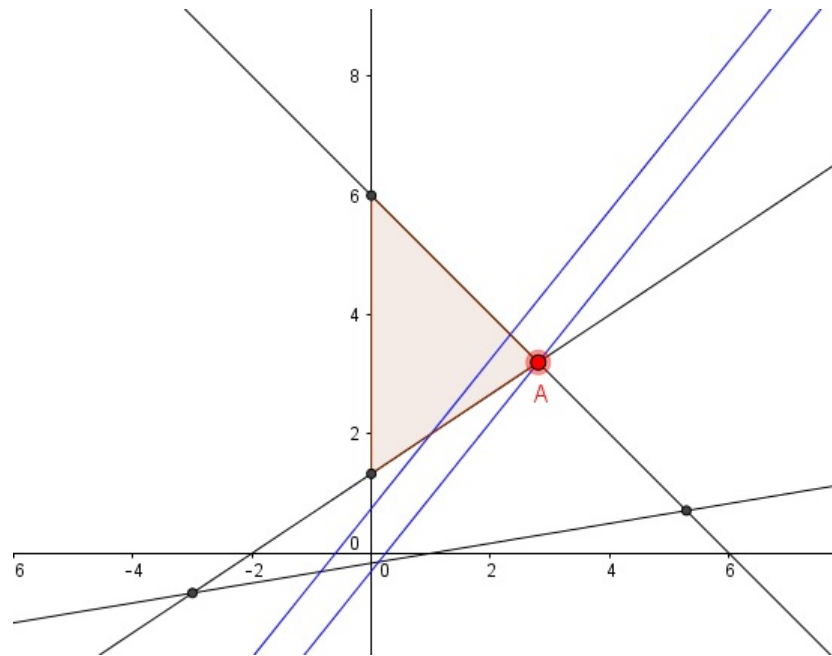
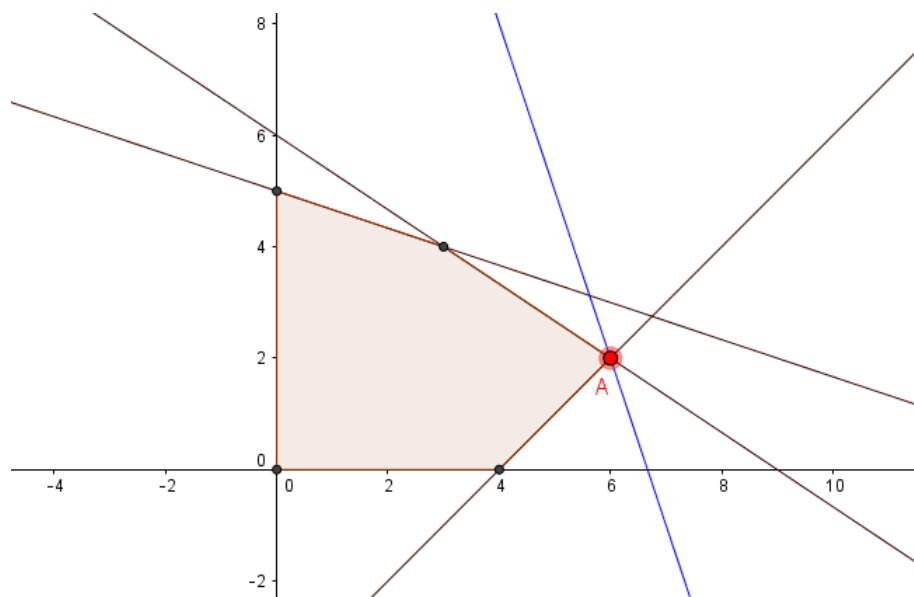


8.1.



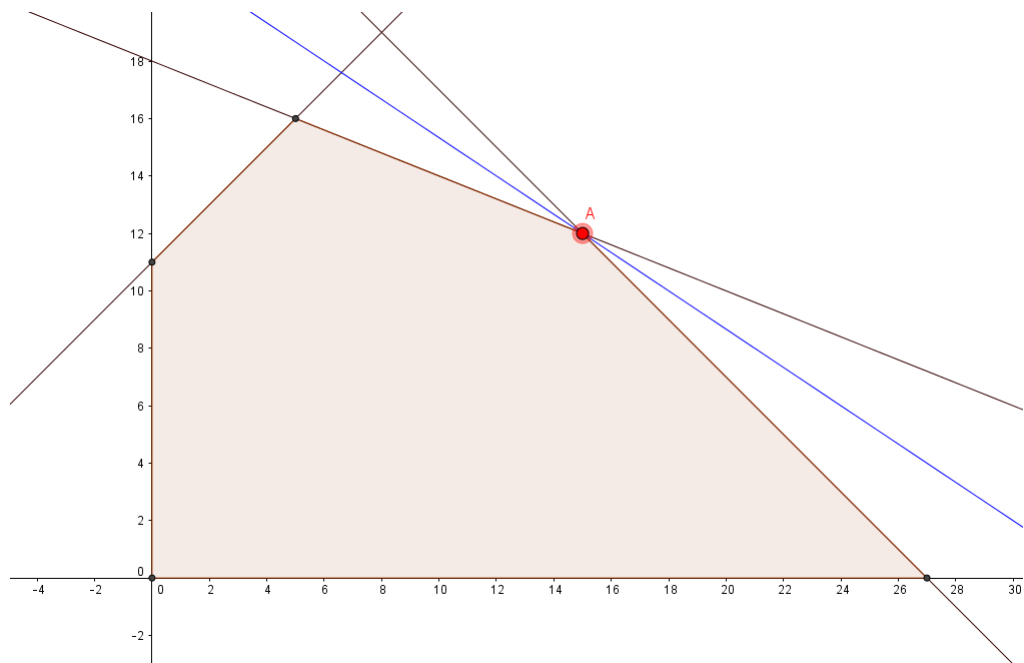
The optimal point is $A = (2.8, 3.2)$.

8.2.



(i)

The optimal point is $A = (6, 2)$ and the optimal value is 20.



(ii)

The optimal point is $A = (15, 12)$ and the optimal value is 132.

8.5.

(i)

$$\begin{array}{rcl}
 \xi & = & 3x_1 + x_2 \\
 \hline
 w_1 & = & 15 - x_1 - 3x_2 \\
 w_2 & = & 18 - 2x_1 - 3x_2 \\
 w_3 & = & 4 - x_1 + x_2 \\
 \hline
 \xi & = & 12 - 3w_3 + 4x_2 \\
 w_1 & = & 11 + w_3 - 4x_2 \\
 w_2 & = & 10 + 2w_3 - 5x_2 \\
 x_1 & = & 4 - w_3 + x_2 \\
 \hline
 \xi & = & 20 - \frac{7}{5}w_3 - \frac{4}{5}w_2 \\
 w_1 & = & 3 - \frac{2}{5}w_3 + \frac{4}{5}x_2 \\
 x_2 & = & 2 + \frac{2}{5}w_3 - \frac{1}{5}w_2 \\
 x_1 & = & 6 - \frac{3}{5}w_3 - \frac{1}{5}w_2
 \end{array}$$

The optimal point is $x_1 = 6, x_2 = 2$, the optimum value is 20.

(ii)

$$\begin{array}{rcl}
 \xi & = & 4x + 6y \\
 \hline
 z & = & 11 + x - y \\
 t & = & 27 - x - y \\
 w & = & 90 - 2x - 5y \\
 \hline
 \xi & = & 108 - 4t + 2y \\
 z & = & 38 - t - 2y \\
 x & = & 27 - t - y \\
 w & = & 36 + 2t - 3y \\
 \hline
 \xi & = & 132 - \frac{8}{3}t - \frac{2}{3}w \\
 z & = & 14 - \frac{7}{3}x + \frac{2}{3}w \\
 x & = & 15 - \frac{5}{3}t + \frac{1}{3}w \\
 y & = & 12 + \frac{2}{3}t - \frac{1}{3}w
 \end{array}$$

The optimal point is $x = 15, y = 12$ and the optimum value is 132.

8.7.

(i) Solve the auxiliary problem

$$\begin{array}{rcll}
 \xi & = & & -x_0 \\
 \hline
 w_1 & = & -8 & + 4x_1 + 2x_2 + x_0 \\
 w_2 & = & 6 & + 2x_1 - 3x_2 + x_0 \\
 w_3 & = & 3 & - x_1 + x_0 \\
 \hline
 \xi & = & -8 & + 4x_1 + 2x_2 - w_1 \\
 x_0 & = & 8 & - 4x_1 - 2x_2 + w_1 \\
 w_2 & = & 14 & - 2x_1 - 5x_2 + w_1 \\
 w_3 & = & 11 & - 5x_1 - 2x_2 + w_1 \\
 \hline
 \xi & = & & -x_0 \\
 x_1 & = & 2 & - \frac{1}{4}x_0 - \frac{1}{2}x_2 + \frac{1}{4}w_1 \\
 w_2 & = & 10 & + \frac{1}{2}x_0 - 4x_2 - \frac{3}{2}w_1 \\
 w_3 & = & 1 & + \frac{5}{4}x_0 + \frac{1}{2}x_2 - \frac{9}{4}w_1
 \end{array}$$

Hence we have found a feasible point $x_1 = 2, x_2 = 0$. Now we solve for the optimal value

$$\begin{array}{rcll}
 \xi & = & 2 & + \frac{1}{4}w_1 + \frac{3}{2}x_2 \\
 x_1 & = & 2 & + \frac{1}{4}w_1 - \frac{1}{2}x_2 \\
 w_2 & = & 10 & - \frac{3}{2}w_1 - 4x_2 \\
 w_3 & = & 1 & - \frac{9}{4}w_1 + \frac{1}{2}x_2
 \end{array}$$

8.13. Suppose we have n decision variables. We start at the origin since it is feasible. Suppose $\mathbf{x} = 0$ is not an optimum point, then there exists a nonbasic variable x_i whose coefficient c_i is strictly positive. Let $w_j = -\sum_{k=1}^n a_{jk}x_k$ be the most binding basic variable. Since we know that we can improve x_i , it follows that $a_{ji} < 0$, otherwise improving x_i will make w_j negative or unchanged. However, then it follows that w_j is not a binding variable anymore, because improve x_i as large as possible won't make w_j negative. This shows that either there is not binding variable, which means the problem is unbounded, or $\mathbf{x} = 0$ is an optimum point.

8.17. Suppose we have a primal problem

$$\begin{array}{ll}
 \text{maximize} & \mathbf{c}^T \mathbf{x} \\
 \text{subject to} & A\mathbf{x} \preceq \mathbf{b} \\
 & \mathbf{x} \succeq \mathbf{0}
 \end{array}$$

Its corresponding dual is

$$\begin{array}{ll}
 \text{minimize} & \mathbf{b}^T \mathbf{y} \\
 \text{subject to} & A^T \mathbf{y} \succeq \mathbf{c} \\
 & \mathbf{y} \succeq \mathbf{0}
 \end{array}$$

which is equivalent to

$$\begin{array}{ll}
 \text{maximize} & -\mathbf{b}^T \mathbf{y} \\
 \text{subject to} & -A^T \mathbf{y} \preceq -\mathbf{c} \\
 & \mathbf{y} \succeq \mathbf{0}
 \end{array}$$

The dual of the dual is therefore

$$\begin{array}{ll}
 \text{minimize} & -\mathbf{c}^T \mathbf{x} \\
 \text{subject to} & -A\mathbf{x} \succeq -\mathbf{b} \\
 & \mathbf{x} \succeq \mathbf{0}
 \end{array}$$

or equivalently,

$$\begin{array}{ll}\text{maximize} & \mathbf{c}^T \mathbf{x} \\ \text{subject to} & A\mathbf{x} \preceq \mathbf{b} \\ & \mathbf{x} \succeq \mathbf{0}\end{array}$$

which is exactly the primal problem.

9.10. We have

$$Df(\mathbf{x}) = \frac{1}{2} \mathbf{x}^T (Q^T + Q) - \mathbf{b}^T = \mathbf{x}^T Q - \mathbf{b}^T = (Q\mathbf{x} - \mathbf{b})^T.$$

and

$$D^2 f(\mathbf{x}) = Q.$$

The unique minimizer of f is the root of $Q\mathbf{x} = \mathbf{b}$. Since Q is positive definite, this means that the unique minimizer is $\mathbf{x} = Q^{-1}\mathbf{b}$. Suppose we start at any initial point \mathbf{x}_0 , then one iteration of Newton's method yields

$$\begin{aligned}\mathbf{x}_1 &= \mathbf{x}_0 - D^2 f(\mathbf{x}_0)^{-1} Df(\mathbf{x}_0)^T \\ &= \mathbf{x}_0 - Q^{-1}(Q\mathbf{x}_0 - \mathbf{b}) \\ &= Q^{-1}\mathbf{b} + \mathbf{x}_0 - Q^{-1}Q\mathbf{x}_0 \\ &= Q^{-1}\mathbf{b}.\end{aligned}$$

This shows that \mathbf{x}_1 is indeed the unique minimizer of f .