

2.1

(a)

$$g'(w) = qw + r$$

$$g''(w) = q$$

(b)

$$g'(w) = \sin(2\pi w^2) \cdot 4\pi w + 2w$$

$$= 4\pi w \sin(2\pi w^2) + 2w$$

$$g''(w) = 4\pi \sin(2\pi w^2) + 4\pi w \cos(2\pi w^2) \cdot 4\pi w + 2$$

$$= 16\pi^2 w^2 \cos(2\pi w^2) + 4\pi \sin(2\pi w^2) + 2$$

(c)

$$g'(w) = \sum_{p=1}^P \frac{1}{1 + e^{-a_p w}} \cdot (-a_p e^{-a_p w})$$

$$= - \sum_{p=1}^P \frac{a_p}{1 + e^{a_p w}}$$

$$g''(w) = - \sum_{p=1}^P \frac{a_p}{1 + e^{a_p w}}$$

$$= - \sum_{p=1}^P \frac{a_p}{(1 + e^{a_p w})^2} \cdot (-a_p e^{a_p w})$$

$$= \sum_{p=1}^P \frac{a_p^2 e^{a_p w}}{(1 + e^{a_p w})^2}$$

2.2

(a)

$$\nabla g(\mathbf{w}) = \mathbf{Q}\mathbf{w} + \mathbf{r}.$$

$$\nabla^2 g(\mathbf{w}) = \frac{1}{2}(\mathbf{Q} + \mathbf{Q}^T)$$

$$= \mathbf{Q}$$

(b)

$$\nabla g(\mathbf{w}) = \sin\left(2\pi \mathbf{w}^T \mathbf{w}\right) 4\pi \mathbf{w} + 2\mathbf{w}.$$

$$\nabla^2 g(\mathbf{w}) = \cos\left(2\pi \mathbf{w}^T \mathbf{w}\right) (4\pi)^2 \mathbf{w} \mathbf{w}^T + \left[\sin\left(2\pi \mathbf{w}^T \mathbf{w}\right) + 2\right] \mathbf{I}_{N \times N},$$

where $\mathbf{I}_{N \times N}$ is the $N \times N$ identity matrix.

(c)

$$\nabla g(\mathbf{w}) = - \sum_{p=1}^P \frac{\mathbf{a}_p}{1 + e^{\mathbf{a}_p^T \mathbf{w}}}.$$

$$\nabla^2 g(\mathbf{w}) = \sum_{p=1}^P \frac{e^{\mathbf{a}_p^T \mathbf{w}}}{\left(1 + e^{\mathbf{a}_p^T \mathbf{w}}\right)^2} \mathbf{a}_p \mathbf{a}_p^T.$$

2.5

The tangent hyperplane generated by the first order Taylor series approximation centered at a point \mathbf{v} is given as

$$h(\mathbf{w}) = g(\mathbf{v}) + \nabla g(\mathbf{v})^T (\mathbf{w} - \mathbf{v}).$$

Any point $\mathbf{Q} = [h(\mathbf{w}) \ \mathbf{w}]^T$ that satisfies the above equation is in the plane. It is easy to see from the equation that point $\mathbf{P} = [g(\mathbf{v}) \ \mathbf{v}]^T$ is in the plane. To verify that $\mathbf{n} = [1 - \nabla g(\mathbf{v})]^T$ is the normal vector of the tangent hyperplane, we can prove that for any point \mathbf{Q} in the plane, the inner product of \mathbf{n} and \mathbf{PQ} is always zero. We have

$$\begin{aligned} \mathbf{PQ} &= \mathbf{Q} - \mathbf{P} \\ &= [h(\mathbf{w}) \ \mathbf{w}]^T - [g(\mathbf{v}) \ \mathbf{v}]^T \\ &= [h(\mathbf{w}) - g(\mathbf{v}) \ (\mathbf{w} - \mathbf{v})^T]^T. \end{aligned}$$

The inner product of \mathbf{n} and \mathbf{PQ} is

$$\mathbf{n}^T \mathbf{PQ} = h(\mathbf{w}) - g(\mathbf{v}) - \nabla g(\mathbf{v})^T (\mathbf{w} - \mathbf{v}) = 0$$

We see that \mathbf{n} is perpendicular to any vector within the plane, thus \mathbf{n} is the normal vector of the tangent hyperplane.

2.7

(a)

$$g'(w) = 2w,$$

$$g''(w) = 2 > 0.$$

(b)

$$g'(w) = 2we^{w^2},$$

$$g''(w) = 2e^{w^2} + (2w)^2 e^{w^2} > 0.$$

(c)

$$g'(w) = \frac{e^w}{1 + e^w},$$

$$\begin{aligned} g''(w) &= \frac{e^w(1 + e^w) - (e^w)^2}{(1 + e^w)^2} \\ &= \frac{e^w}{(1 + e^w)^2} > 0. \end{aligned}$$

(d)

$$g'(w) = -\frac{1}{w},$$

$$g''(w) = \frac{1}{w^2} > 0.$$

2.17

(a)

$$\nabla g(\mathbf{w}) = \frac{2e^{\mathbf{w}^T \mathbf{w}}}{1 + e^{\mathbf{w}^T \mathbf{w}}} \mathbf{w}$$

Since the scalar weight $\frac{2e^{\mathbf{w}^T \mathbf{w}}}{1 + e^{\mathbf{w}^T \mathbf{w}}} \geq 1$ the only way the equality can occur is when $\mathbf{w} = \mathbf{0}_{N \times 1}$.

(b)

$$\nabla^2 g(\mathbf{w}) = \frac{4e^{\mathbf{w}^T \mathbf{w}}}{(1 + e^{\mathbf{w}^T \mathbf{w}})^2} \mathbf{w} \mathbf{w}^T + \frac{2e^{\mathbf{w}^T \mathbf{w}}}{1 + e^{\mathbf{w}^T \mathbf{w}}} \mathbf{I}_{N \times N}.$$

(d) why faster convergence with initialization point [4,4] ?

Because \mathbf{w} is large, $g(\mathbf{w}) \approx \mathbf{w}^T \mathbf{w}$

And

$$\begin{aligned} h(\mathbf{w}) &= g(\mathbf{w}^0) + \nabla g(\mathbf{w}^0)^T (\mathbf{w} - \mathbf{w}^0) + \frac{1}{2} (\mathbf{w} - \mathbf{w}^0)^T \nabla^2 g(\mathbf{w}^0) (\mathbf{w} - \mathbf{w}^0) \\ &= \mathbf{w}^{0T} \mathbf{w}^0 + 2\mathbf{w}^{0T} (\mathbf{w} - \mathbf{w}^0) + \frac{1}{2} (\mathbf{w} - \mathbf{w}^0)^T \cdot 2(\mathbf{w} - \mathbf{w}^0) \\ &= \mathbf{w}^T \mathbf{w} \end{aligned}$$

Newton's method optimizes the quadratic function $h(\mathbf{w})$, which is the same as $g(\mathbf{w})$, when \mathbf{w} is large. So the optimal solution of Newton's method is the same as optimal point as $g(\mathbf{w})$.