Problem 2.4 worth 4 points

Linear function? The function $\phi: \mathbf{R}^3 \to \mathbf{R}$ satisfies

$$\phi(1,1,0) = -1$$
 $\phi(-1,1,1) = 1$ $\phi(1,-1,-1) = 1$

Choose one of the following, and justify your choice: ϕ must be linear; ϕ could be linear; ϕ cannot be linear

Solution:

$$\alpha \cdot \phi(\vec{x}) = \phi(\alpha \cdot \vec{x})$$
 (principle of superposition)

Let
$$\alpha = -1$$
, $\vec{x}_b = (-1, 1, 1)$, and $\vec{x}_c = (1, -1, -1)$

$$\alpha \cdot \phi(\vec{x}_b) = \phi(\alpha \cdot \vec{x}_b)$$

$$-1 \cdot \phi(-1, 1, 1) = \phi(-1 \cdot \begin{pmatrix} -1 \\ 1 \\ 1 \end{pmatrix})$$

$$-1 \cdot (1) = \phi(\begin{pmatrix} 1 \\ -1 \\ -1 \end{pmatrix})$$

$$-1 = \phi(1, -1, -1)$$

$$-1 \neq 1$$

Because we are told the function $\phi(\vec{x}_c) = 1$ when $\vec{x}_c = (1, -1, -1)$, and since $\alpha \cdot \vec{x}_b = (1, -1, -1) = \vec{x}_c$, we see that superposition does not hold, therefore ϕ cannot be linear.

Problem 2.4 worth 4 points

Affine function. Suppose $\psi: \mathbf{R}^2 \to \mathbf{R}$ is an affine function, with $\psi(1,0) = 1$, $\psi(1,-2) = 2$.

- (a) What can you say about $\psi(1,-1)$? Either give the value of $\psi(1,-1)$ or state that it cannot be determined.
- (b) What can you say about $\psi(2,-2)$? Either give the value of $\psi(2,-2)$ or state that it cannot be determined.

Solution:

- (a) Since we're given $\psi(1,0) = 1$ and $\psi(1,-2) = 2$ we can conclude that the value $1 < \psi(1,-1) < 2$ because we see \vec{x}_1 remains constant as \vec{x}_2 is changing which, in turn, causes $\psi(\vec{x})$ to change. Since an affine function must be linear, if we keep a value constant and change another the result will change linearly, and since $|\Delta \vec{x}_2| = 2$ and $|\Delta \psi(\vec{x})| = 1$, if we look at the change in result with respect to \vec{x}_2 we get $\frac{1}{2}$. Add this on to our given initial value and we get $\psi(1,-1) = \frac{3}{2}$.
- (b) We are not told how the function behaves as \vec{x}_1 changes, so the value of $\psi(2,-2)$ cannot be determined.

Problem 2.9 worth 6 points

Taylor approximation. Consider the function $f: \mathbf{R}^2 \to \mathbf{R}$ given by $f(x_1, x_2) = x_1 x_2$. Find the Taylor approximation \hat{f} at the point z = (1, 1). Compare f(x) and $\hat{f}(x)$ for the following values of x.

$$x = (1,1), \quad x = (1.05, 0.95), \quad x = (0.85, 1.25), \quad x = (-1,2).$$

Make a brief comment about the accuracy of the Taylor approximation in each case.

Solution: If z = (1, 1), then we see $f(z) = (1) \cdot (1) = 1$.

$$\hat{f}(x) = f(z) + \frac{\delta f(z)}{\delta x_1} (x_1 - z_1) + \frac{\delta f(z)}{\delta x_2} (x_2 - z_2)$$

Now we take a look at our partial derivatives for x_1 and x_2 when f(z) = 1

$$\frac{\delta f(z)}{\delta x_1} = x_2(1) \qquad \frac{\delta f(z)}{\delta x_2} = x_1(1) = (1)(1) = 1 = (1)(1) = 1$$

Going back to our function \hat{f}

$$\hat{f}(x) = f(z) + \frac{\delta f(z)}{\delta x_1} (x_1 - z_1) + \frac{\delta f(z)}{\delta x_2} (x_2 - z_2)$$

$$\hat{f}(x) = (1) + (1)(x_1 - 1) + (1)(x_2 - 1)$$

$$\hat{f}(x) = (1) + (x_1 - 1) + (x_2 - 1)$$

Now let's evaluate for each of our given vectors:

$$\hat{f}(1,1) = (1) + (1-1) + (1-1)$$

$$\hat{f}(1,1) = 1 + 0 + 0$$

$$\hat{f}(1,1) = 1$$

$$\hat{f}(1.05, 0.95) = (1) + (1.05 - 1) + (.95 - 1)$$

$$\hat{f}(1.05, 0.95) = (1) + (.05) - (.05)$$

$$\hat{f}(1.05, 0.95) = 1$$

$$\hat{f}(0.85, 1.25) = (1) + (.85 - 1) + (1.25 - 1)$$

$$\hat{f}(0.85, 1.25) = (1) - (.15) + (.25)$$

$$\hat{f}(0.85, 1.25) = 1.1$$

$$\hat{f}(-1,2) = (1) + (-1 - 1) + (2 - 1)$$

$$\hat{f}(-1,2) = (1) - 2 + 1$$

$$\hat{f}(-1,2) = 0$$

What each of these numbers says about the point is an estimation for the value of f(x), based on the value of f(z) and point x's proximity to z. For vectors $x_1 = (1,1)$ and $x_2 = (1.05, 0.95)$ we can see that our values should be close to one $(f(x_1) = 1)$ and $f(x_2) = .9975$ in these cases) so we get a value of 1. These are very accurate because if we look at our points as edges of a triangle then using Pythagorean formula $(a^2 + b^2) = c^2$ where a is given by Δx_1 and b is given by Δx_2 we see the following values for each of our points:

$$(1,1) = \sqrt{(1-1)^2 + (1-1)^2}$$

$$(1,1) = \sqrt{0}$$

$$(1,1) = 0$$

$$(0 because we're at the same point)$$

$$(1.05,0.95) = \sqrt{(1-1.05)^2 + (1-0.95)^2}$$

$$(1.05,0.95) = \sqrt{.0025 + .0025}$$

$$(1.05,0.95) = .0707$$

$$(.07 says we're very close)$$

$$(0.85,1.25) = \sqrt{(1-0.85)^2 + (1-1.25)^2}$$

$$(0.85,1.25) = \sqrt{.0225 + .0625}$$

$$(0.85,1.25) = .29$$

$$(.29 says we're further)$$

$$(-1,2) = \sqrt{(1-(-1))^2 + (1-2)^2}$$

$$(-1,2) = \sqrt{4+1}$$

$$(-1,2) = 2.236$$

$$(2.236 says we're very far from our point)$$

For vector $x_3 = (0.85, 1.25)$, our function should return the value 1.0625 for x_3 , and we see a rounded version of this value as our approximation. For $x_4 = (-1, 2)$, we see our point is much further away than our other 3 as they compare to z. Our function $f(x_4)$ would return the value -2, but since we are approximating a value based on our current known value for z, we see the accuracy decreases the further away we are, and we get 0 as our answer.

Problem 2.10 worth 6 points

Regression model. Consider the regression model $\hat{y} = x^T \beta + v$, where \hat{y} is the predicted response, x is an 8-vector of features, β is an 8-vector of coefficients, and v is the offset term. Determine whether each of the following statements is true or false.

- (a) If $\beta_3 > 0$, and $x_3 > 0$, then $\hat{y} \ge 0$.
- (b) If $\beta_2 = 0$ then the prediction \hat{y} does not depend on the second feature x_2 .
- (c) If $\beta_6 = -0.8$, then increasing x_6 (keeping all other x_i s the same) will decrease \hat{y} .

Solution:

- (a) False. We can't conclude the value of \hat{y} from just a singular x_i when we have a vector of 8 features.
- (b) True. If $\beta_2 = 0$ then it "eliminates" the value at x_2 meaning the final product does not depend on that value.
- (c) True. Increasing the value of x_6 will create a larger negative value. Since $x^T\beta$ involves summing all our elements after multiplying them with each corresponding coefficient, and since we are leaving all other x_i s the same, then increasing our value x_6 will create a larger negative value in the summation, creating a smaller scalar of \hat{y} .