

Question 1:

a)

$$\begin{aligned}
 \frac{\sqrt{W_1}}{w_0} &= \sum_{n=1}^N (y_n - w_0 - w_1 x_n)^2 = \sum_{n=1}^N 2(y_n - w_0 - w_1 x_n) \cdot (0 - 1 - 0) \\
 \frac{\sqrt{W_1}}{w_0} &= 2 \sum_{n=1}^N -y_n + w_0 + w_1 x_n \\
 0 &= -2 \sum_{n=1}^N y_n + 2w_0 \sum_{n=1}^N + 2w_1 \sum_{n=1}^N x_n \\
 &\quad \cancel{+ 2w_0 \sum_{n=1}^N y_n - 2w_1 \sum_{n=1}^N x_n} \\
 2 \sum_{n=1}^N y_n - 2w_1 \sum_{n=1}^N x_n &= 2w_0 \sum_{n=1}^N \\
 \frac{2}{N} \sum_{n=1}^N y_n - w_1 \left(\frac{2}{N} \sum_{n=1}^N x_n \right) &= 2w_0 N \\
 \frac{2}{N} \sum_{n=1}^N y_n - w_1 \left(\frac{2}{N} \sum_{n=1}^N x_n \right) &= w_0 \\
 \frac{\sqrt{W_1}}{w_1} &= \sum_{n=1}^N 2(y_n - w_0 - w_1 x_n) \cdot (0 - 0 - x_n) \\
 0 &= \sum_{n=1}^N -2x_n (y_n - w_0 - w_1 x_n) \\
 0 &= \sum_{n=1}^N -2x_n y_n + 2x_n w_0 + 2w_1 x_n^2 \\
 -2w_1 \sum_{n=1}^N x_n^2 &= 2w_0 \sum_{n=1}^N x_n + 2 \sum_{n=1}^N x_n y_n \\
 w_1 &= \frac{-2(w_0 \sum_{n=1}^N x_n - \sum_{n=1}^N x_n y_n)}{2 \sum_{n=1}^N x_n^2} \Rightarrow w_1 = \frac{w_0 \sum_{n=1}^N x_n - w_0 \sum_{n=1}^N x_n}{\sum_{n=1}^N x_n^2}
 \end{aligned}$$

b)

$$\begin{aligned}
 & \text{If } \bar{x} = \frac{1}{N} \sum_{n=1}^N x_n \text{ and } \bar{y} = \frac{1}{N} \sum_{n=1}^N y_n \\
 & W_0^* = \left(\frac{1}{N} \sum_{n=1}^N y_n \right) - W_1 \left(\frac{1}{N} \sum_{n=1}^N x_n \right) \\
 & = \bar{y} - w_1 \bar{x} \\
 & W_1^* = \frac{\sum_{n=1}^N x_n y_n - N \left(\frac{1}{N} \sum_{n=1}^N x_n \right) \left(\frac{1}{N} \sum_{n=1}^N y_n \right)}{\sum_{n=1}^N x_n^2 - N \left(\frac{1}{N} \sum_{n=1}^N x_n \right)^2} \\
 & = \frac{x_n y_n - N(\bar{x})(\bar{y})}{x_n^2 - N(\bar{x})^2} - \frac{N(\bar{x})(\bar{y})}{x_n^2 - N(\bar{x})^2} \\
 & = \frac{\sum_{n=1}^N x_n y_n}{\sum_{n=1}^N x_n^2 - N(\bar{x})^2} - \frac{\cancel{N(\bar{x})}}{\cancel{N(\bar{x})}} \frac{\cancel{x_n \bar{y}}}{\cancel{x_n(x_n - \bar{x})}} \\
 & = \frac{\sum_{n=1}^N x_n y_n}{\sum_{n=1}^N x_n^2 - N \left(\frac{1}{N} \sum_{n=1}^N x_n \right) \left(\frac{1}{N} \sum_{n=1}^N x_n \right)} - \frac{\bar{y}}{(x_n - \bar{x})} \\
 & = \frac{\sum_{n=1}^N x_n y_n}{\sum_{n=1}^N x_n^2 - x_n \cdot \bar{x}} - \frac{\bar{y}}{(x_n - \bar{x})}
 \end{aligned}$$

$$\begin{aligned}
 W_1^* &= \frac{\sum y_n}{\sum (x_n - \bar{x})} - \frac{\bar{y}}{\sum (x_n - \bar{x})} \\
 &= \frac{\sum_{n=1}^N y_n - \bar{y}}{\sum_{n=1}^N (x_n - \bar{x})} \\
 &= \frac{\sum_{n=1}^N y_n - \bar{y}}{\sum_{n=1}^N (x_n - \bar{x})} + \frac{(x_n - \bar{x})}{(x_n - \bar{x})} \\
 &= \frac{\sum_{n=1}^N (y_n - \bar{y})(x_n - \bar{x})}{\sum_{n=1}^N (x_n - \bar{x})^2}
 \end{aligned}$$

c) W_1^* represents the covariance of the variances of the input features "x" and "y". W_0^* represents the difference of the means of x and y associated with the D^{train} . W_1^* is used as a weight to influence the mean of x as adjusted for minimized prediction error.

Question 2:

- a) The gradient vector of J with respect to $x(0)$ is a first-order derivative while the Hessian matrix, which contains the second order derivatives, specifically the derivative with respect to $x(0)$. In accordance to the Taylor expansion to the second order, the function is approximated as follows:

$$\begin{aligned}
 f(x) &= f(x_0) + (\nabla f|_{x=x_0})^T (x - x_0) + \frac{1}{2} (x - x_0)^T H_f|_{x=x_0} (x - x_0) \\
 J(w) &\approx J(w(k)) + (\nabla J|_{w=w(k)})^T (w - w(k)) + \frac{1}{2} (w - w(k))^T \dots \\
 &\quad \cdot H_f|_{w=w(k)} (w - w(k))
 \end{aligned}$$

b)

$$\begin{aligned}
 J(w(k+1)) &\approx J(w(k)) + (\nabla J|_{w=w(k+1)})^T \cdot (w - w(k+1)) \\
 &+ \frac{1}{2} (\nabla J|_{w=w(k+1)})^T \cdot H_j|_{w=w(k)} \cdot (w - w(k))
 \end{aligned}$$

$$\begin{aligned}
 &\approx J(w(k)) + \|(\nabla J|_{w=w(k)})\|_2^2 \cdot ((w(k) - \hat{w}(k) - a(k)) \cdot \nabla J|_{w=w(k)}) \\
 &+ \frac{1}{2} ((w(k) - \hat{w}(k) - a(k)) \cdot J|_{w=w(k)})^T \cdot H_j|_{w=w(k)} \cdot \\
 &\quad \dots \cdot (w(k) - \hat{w}(k) - a(k)) \cdot \nabla J|_{w=w(k)} \\
 &\approx J(w(k)) + \|(\nabla J|_{w=w(k)})\|_2^2 \cdot a(k) + \dots \\
 &\quad + \frac{1}{2} ((\nabla J|_{w=w(k)})^T (H_j|_{w=w(k)}) (\nabla J|_{w=w(k)} \cdot a(k)))
 \end{aligned}$$

c)

$$J(w(k)) = \|\nabla J\|$$

$$J(w(k)) = \|\nabla J|_{w=w(k)}\|_2^2 + 1 + \frac{1}{2}(\alpha)$$

$$J(w) \approx J(w(k)) - \|\nabla J|_{w=w(k)}\|_2^2 \cdot \alpha(k) + \frac{1}{2}(\alpha) \cdot \alpha^2(k)$$

$$\text{Let } A = ((\nabla J|_{w=w(k)})^T H_j|_{w=w(k)} (\nabla J|_{w=w(k)}))$$

~~$$J(w) \approx -\|\nabla J|_{w=w(k)}\|_2^2 + 1 + \frac{1}{2}(\alpha) \cdot \alpha^2(k)$$~~

$$\approx -\|\nabla J|_{w=w(k)}\|_2^2 + \cancel{\frac{1}{2}(\alpha) \cdot \cancel{\alpha^2(k)}}$$

$$\approx -\|\nabla J|_{w=w(k)}\|_2^2 + A \cdot \alpha(k)$$

$$\|\nabla J|_{w=w(k)}\|_2^2 = A \cdot \alpha(k)$$

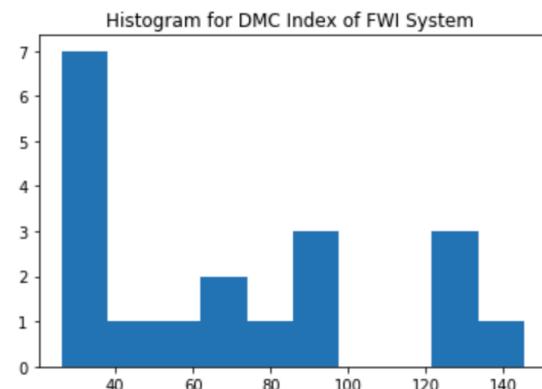
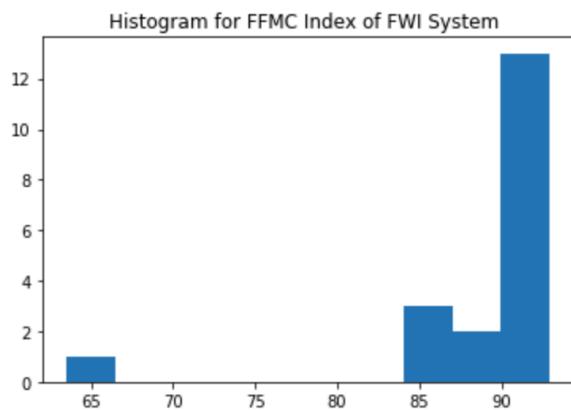
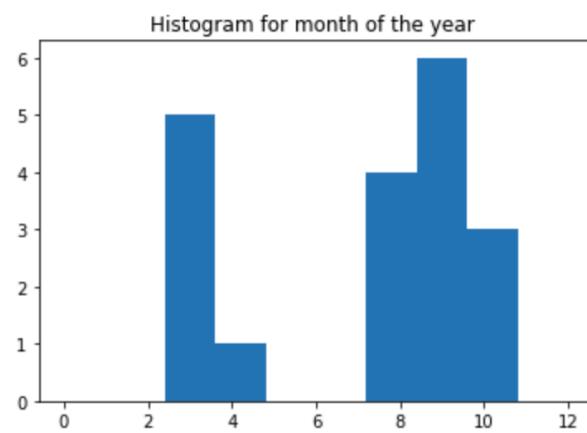
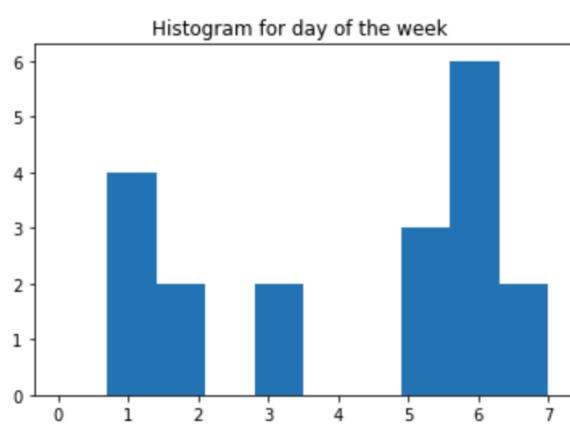
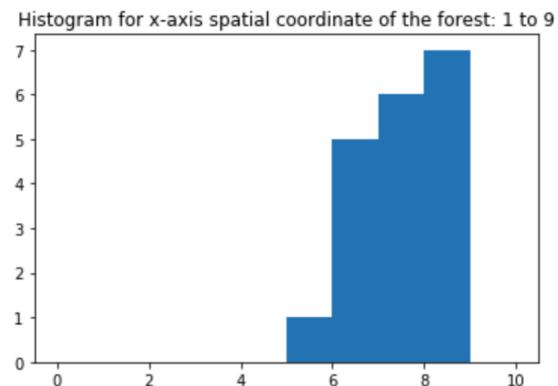
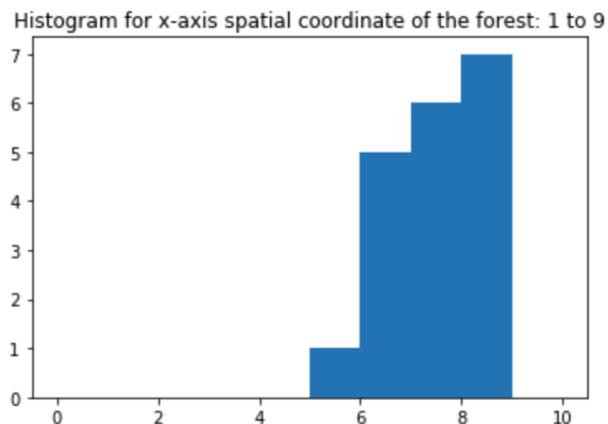
$$\frac{\|\nabla J|_{w=w(k)}\|_2^2}{A} = \alpha(k)$$

$$\alpha(k) = \frac{\|\nabla J|_{w=w(k)}\|_2^2}{((\nabla J|_{w=w(k)})^T H_j|_{w=w(k)} (\nabla J|_{w=w(k)}))}$$

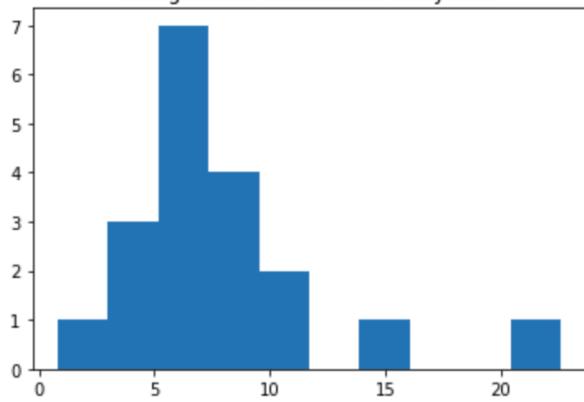
Question 3:

- a) The following features are categorical: 'X', 'Y', 'month', and 'day'. The following features are continuous: 'FFMC', 'DMC', 'DC', 'ISI', 'temp', 'RH', 'wind', and 'rain'.

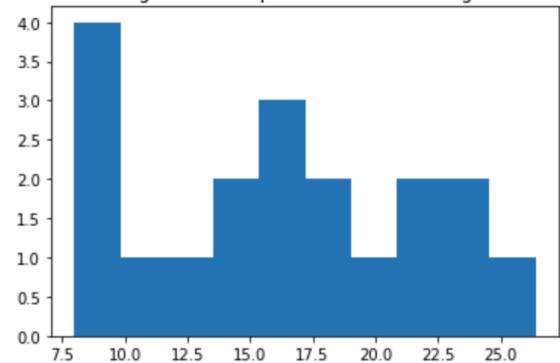
Histograms:



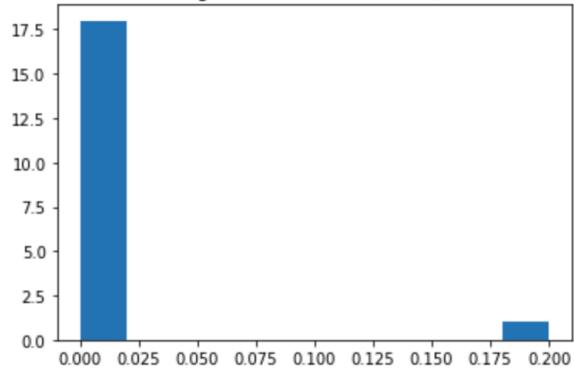
Histogram for ISI Index of FWI System



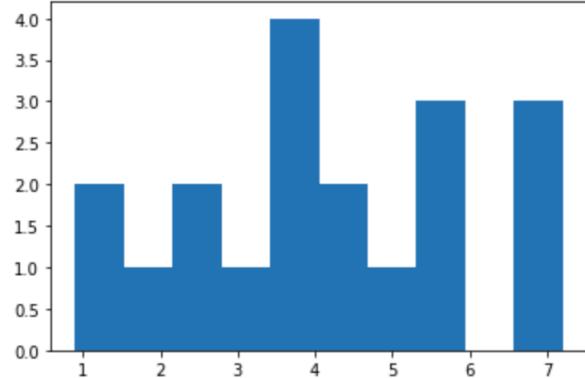
Histogram for temperature in Celsius degrees



Histogram for outside rain mm/m²



Histogram for wind speed km/h



b-i)

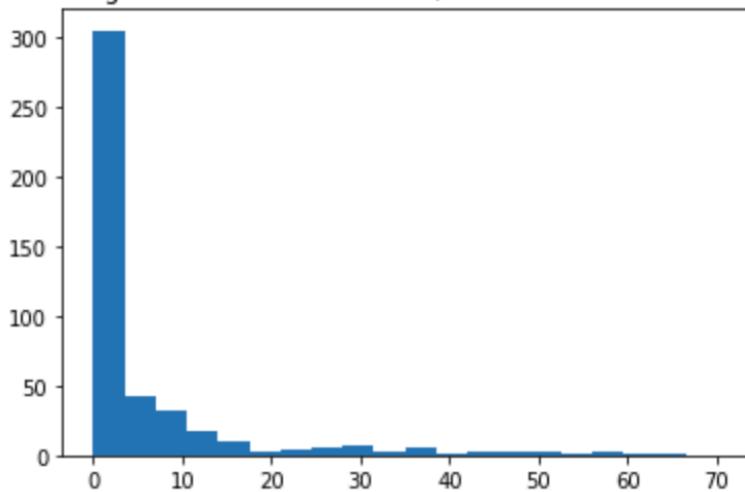
b-ii)

b-iii)

c)

Histogram of outcome variables:

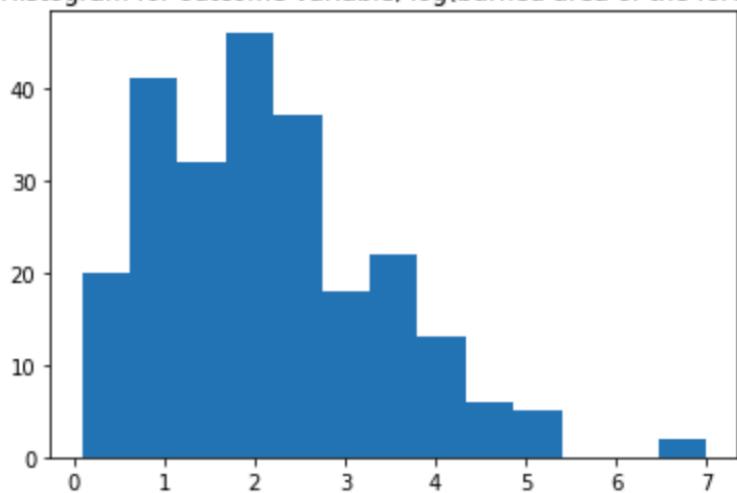
Histogram for outcome variable, burned area of the forest



Outcome values appear to be concentrated between 10-17 area units.

Histogram of logarithm of outcome variables:

Histogram for outcome variable, log(burned area of the forest)



Outcome values appear to be concentrated between 1-7 area units. Frequencies appeared to have decreased as they were distributed in the space.