

CS 540: Introduction to Artificial Intelligence

Homework Assignment 3

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Section 1

0 Late Days Used

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1 Question 1: Probabilities

The following two tables provide the full joint distribution for three boolean variables X, Y and Z .

Table 1: Z		
	Y	\bar{Y}
X	0.2	0.12
\bar{X}	0.04	0.24

Table 2: \bar{Z}		
	Y	\bar{Y}
X	0.1	0.08
\bar{X}	0.06	0.16

1.1 a) What is $P(\bar{Y})$?

$$\begin{aligned}
 P(\bar{Y}) &= P\left(\sum X, \bar{Y}, \sum Z\right) \\
 &= P(X, \bar{Y}, Z) + P(X, \bar{Y}, \bar{Z}) + P(\bar{X}, \bar{Y}, Z) + P(\bar{X}, \bar{Y}, \bar{Z}) \\
 &= 0.12 + 0.08 + 0.24 + 0.16
 \end{aligned}$$

$$P(\bar{Y}) = 0.6$$

1.2 b) What is $P(X|\bar{Y})$?

$$\begin{aligned}
 P(X|\bar{Y}) &= \frac{P(X, \bar{Y})}{P(\bar{Y})} \\
 P(X, \bar{Y}) &= P(X, \bar{Y}, \sum Z) \\
 &= P(X, \bar{Y}, Z) + P(X, \bar{Y}, \bar{Z}) \\
 &= 0.12 + 0.08 = 0.2 \\
 P(\bar{Y}) &= P(X, \bar{Y}, Z) + P(X, \bar{Y}, \bar{Z}) + P(\bar{X}, \bar{Y}, Z) + P(\bar{X}, \bar{Y}, \bar{Z}) \\
 &= 0.12 + 0.08 + 0.24 + 0.16 = 0.6
 \end{aligned}$$

$$P(X|\bar{Y}) = \frac{0.2}{0.6}$$

$$P(X|\bar{Y}) = \frac{1}{3}$$

1.3 c) What is $P(\bar{Z}|\bar{X}, Y)$?

$$\begin{aligned}
 P(\bar{Z}|\bar{X}, Y) &= \frac{P(\bar{Z}, \bar{X}, Y)}{P(\bar{X}, Y)} \\
 P(\bar{Z}, \bar{X}, Y) &= 0.06 \\
 P(\bar{X}, Y) &= P(\bar{X}, Y, \sum Z) \\
 &= P(\bar{X}, Y, Z) + P(\bar{X}, Y, \bar{Z}) \\
 &= 0.04 + 0.06 = 0.1 \\
 P(\bar{Z}|\bar{X}, Y) &= \frac{0.06}{0.1} \\
 \boxed{P(\bar{Z}|\bar{X}, Y) = 0.6}
 \end{aligned}$$

Note on Independence

Two events A, B , are independent if:

$$P(A, B) = P(A) \times P(B) \quad (1)$$

$$P(A|B) = P(A) \quad (2)$$

$$P(B|A) = P(B) \quad (3)$$

1.4 d) Verify whether or not X and Y are independent.

Test the conditions in the section labeled “Independence”.

Condition 1

We attempt to verify (1).

$$P(X, Y) \stackrel{?}{=} P(X) \times P(Y)$$

First we simplify the left hand side:

$$\begin{aligned}
 P(X, Y) &= P(X) \times P(Y|X) \\
 &= P(X) * \frac{P(Y, X)}{P(X)} \\
 &= P(X, \sum Y, \sum Z) * \frac{P(X, Y, \sum Z)}{P(X, \sum Y, \sum Z)} \\
 &= (0.2 + 0.12 + 0.1 + 0.08) * \frac{0.2 + 0.1}{0.2 + 0.12 + 0.1 + 0.08} \\
 &= 0.2 + 0.1 = 0.3
 \end{aligned}$$

$$\boxed{P(X, Y) = 0.3}$$

Next we simplify the right hand side.

$$\begin{aligned} P(X) \times P(Y) &= P(X, \sum Y, \sum Z) \times P(Y, \sum X, \sum Z) \\ &= (0.2 + 0.12 + 0.1 + 0.08) \times (0.2 + 0.04 + 0.1 + 0.06) \\ &= 0.5 \times 0.4 \end{aligned}$$

$$\boxed{P(X) \times P(Y) = 0.2}$$

Now we can compare the two:

$$\begin{aligned} P(X, Y) &\stackrel{?}{=} P(X) \times P(Y) \\ \boxed{0.3 \neq 0.2} \end{aligned}$$

Since (1) is not true, we know that X and Y are not independent.

1.5 e) Verify whether or not Y and Z are independent.

Condition 1

We attempt to verify (1).

$$P(Y, Z) \stackrel{?}{=} P(Y) \times P(Z)$$

First we simplify the left hand side:

$$\begin{aligned} P(Y, Z) &= P(Y) \times P(Z|Y) \\ &= P(Y, \sum X, \sum Z) \times \frac{P(Z, Y, \sum X)}{P(Y, \sum X, \sum Z)} \\ &= P(Y, Z, \sum X) = P(X, Y, Z) + P(\bar{X}, Y, Z) = 0.2 + 0.04 \end{aligned}$$

$$\boxed{P(Y, Z) = 0.24}$$

Now we simplify the right hand side:

$$\begin{aligned} P(Y) \times P(Z) &= P(Y, \sum X, \sum Z) \times P(Z, \sum X, \sum Y) \\ &= (0.2 + 0.04 + 0.1 + 0.06) \times (0.2 + 0.04 + 0.12 + 0.24) \\ &= 0.4 \times 0.6 \end{aligned}$$

$$\boxed{P(Y) \times P(Z) = 0.24}$$

Now we compare the two:

$$\begin{aligned} P(Y, Z) &\stackrel{?}{=} P(Y) \times P(Z) \\ \boxed{0.24 = 0.24} \end{aligned}$$

We have satisfied (1), and now must move on to the next condition.

Condition 2

We attempt to verify (2).

$$P(Y|Z) = P(Y)$$

First we simplify the left hand side:

$$\begin{aligned} P(Y|Z) &= \frac{P(Y, Z)}{P(Z)} \\ &= \frac{P(Y, Z, \sum X)}{P(Z, \sum X, \sum Y)} \\ &= \frac{0.2 + 0.04}{0.2 + 0.12 + 0.04 + 0.24} = \frac{0.24}{0.6} \\ \boxed{P(Y|Z) = 0.4} \end{aligned}$$

Next we simplify the right hand side:

$$\begin{aligned} P(Y) &= P(Y, \sum X, \sum Z) \\ &= 0.2 + 0.4 + 0.1 + 0.06 \\ \boxed{P(Y) = 0.4} \end{aligned}$$

Now we compare the two sides:

$$\begin{aligned} P(Y|Z) &\stackrel{?}{=} P(Y) \\ \boxed{0.4} &= \boxed{0.4} \end{aligned}$$

We have successfully verified (2). One more condition to verify!

Condition 3

We attempt to verify (3).

$$P(Z|Y) = P(Z)$$

First we simplify the left hand side:

$$\begin{aligned} P(Z|Y) &= \frac{P(Z, Y)}{P(Y)} \\ &= \frac{P(Z, Y, \sum X)}{P(Y, \sum X, \sum Z)} \\ &= \frac{0.2 + 0.04}{0.2 + 0.04 + 0.1 + 0.06} = \frac{0.24}{0.4} \\ \boxed{P(Z|Y) = 0.6} \end{aligned}$$

Next we simplify the right hand side:

$$\begin{aligned} P(Z) &= P(Z, \sum X, \sum Y) \\ &= 0.2 + 0.12 + 0.04 + 0.24 \\ \boxed{P(Z) = 0.6} \end{aligned}$$

Now we compare the two sides:

$$P(Z|Y) \stackrel{?}{=} P(Z)$$

$$\boxed{0.6 = 0.6}$$

We have satisfied (3).

Conclusion

We have satisfied (1), (2), and (3). Therefore, we can conclude that Y and Z are independent.

2 Question 2: Neural Networks

The following problems were answered using the provided neural network on page 2 of “hw3v2.pdf”. Note that all bias nodes b have an output of 1.

Definition of linear perceptron output

$$a = \sum_{d=0..D} w_d \times x_d \quad (4)$$

$$\sigma(x) = \frac{1}{1 + e^{(-w'x)}} \quad (5)$$

Output of all nodes with inputs x_i and weights w_i are calculated with (4), and the classification is designated by the sigmoid function (5).

2.1 a) Calculate the output of this network for the input $\{x_1 = 0, x_2 = 1\}$.

h_1

First we calculate the value of $w'x$ for h_1 with (4):

$$w'x_{h_1} = -1 * b + 2 * x_1 + 1 * x_2$$

$$w'x_{h_1} = -1 * 1 + 2 * 0 + 1 * 1$$

$$w'x_{h_1} = -1 + 1$$

$$\boxed{w'x_{h_1} = 0}$$

Now we apply (5) to find h_1 :

$$\begin{aligned}h_1 &= \sigma(w'x_{h_1}) \\h_1 &= \frac{1}{1 + e^{(-w'x_{h_1})}} \\h_1 &= \frac{1}{1 + e^{(0)}} \\\boxed{h_1 &= \frac{1}{2}}\end{aligned}$$

h_2

Now we calculate the value of $w'x$ for h_2 with (4):

$$\begin{aligned}w'x_{h_2} &= 0.5 * b - 3 * x_1 - 2 * x_2 \\w'x_{h_2} &= 0.5 * 1 - 3 * 0 - 2 * 1 \\w'x_{h_2} &= 0.5 - 2 \\\boxed{w'x_{h_2} &= -1.5}\end{aligned}$$

Next we apply (5) to find h_2 :

$$\begin{aligned}h_2 &= \sigma(w'x_{h_2}) \\h_2 &= \frac{1}{1 + e^{(-w'x_{h_2})}} \\h_2 &= \frac{1}{1 + e^{(1.5)}} \\h_2 &= 0.1824255238 \\\boxed{h_2 &\approx 0.18}\end{aligned}$$

k

Finally, we can calculate $w'x$ for k with (4):

$$\begin{aligned}w'x_k &= -0.5 * b - 1 * h_1 + 0.8 * h_2 \\w'x_k &= -0.5 * 1 - 1 * 0 + 0.8 * (-1.5) \\w'x_k &= -0.5 - 1.2 \\\boxed{w'x_k &= -1.7}\end{aligned}$$

Again, we apply (5) to find k :

$$\begin{aligned}
 k &= \sigma(w'x_k) \\
 k &= \frac{1}{1 + e^{(-w'x_k)}} \\
 k &= \frac{1}{1 + e^{(1.7)}} \\
 k &= 0.1544652651 \\
 \boxed{k \approx 0.15}
 \end{aligned}$$

Thus the output of the network will be $y \approx 0.15$.

2.2 b) Now you are going to compute one step of the backpropagation algorithm. The weights of the output node k (red node) are fixed. The input for the training instance is $\{x_1 = 0, x_2 = 1\}$ and the output of this training instance is $y = 1$. Please compute the updated weights for the hidden layer (the two blue nodes) by performing ONE step of gradient descent. Let the step size α equal 0.1.

For output unit k , compute error term δ_k :

$$\begin{aligned}
 \delta_k &\leftarrow (o_k - y_k) o_k (1 - o_k) \\
 \delta_k &\leftarrow (0.15 - 1.) * 0.15 * (1 - 0.15) \\
 \delta_k &\leftarrow -0.85 * 0.15 * 0.85 \\
 \boxed{\delta_k \leftarrow 0.11}
 \end{aligned}$$

Now compute error term for hidden node h_1 :

$$\begin{aligned}
 \delta_{h_1} &\leftarrow \left(\sum_{i \in \text{succ}(h_1)} w_{ih_1} \delta_i \right) o_{h_1} (1 - o_{h_1}) \\
 \delta_{h_1} &\leftarrow (w_{kh_1} \delta_k) o_{h_1} (1 - o_{h_1}) \\
 \delta_{h_1} &\leftarrow (-1 * 0.11) * \frac{1}{2} * \left(1 - \frac{1}{2} \right) \\
 \boxed{\delta_{h_1} \leftarrow -0.0275}
 \end{aligned}$$

The error term for node h_2 is similar:

$$\begin{aligned}
 \delta_{h_2} &\leftarrow (w_{kh_2} \delta_k) o_{h_2} (1 - o_{h_2}) \\
 \delta_{h_2} &\leftarrow (0.8 * 0.11) * 0.18 * (1 - 0.18) \\
 \boxed{\delta_{h_2} \leftarrow 0.131}
 \end{aligned}$$

Now we need to update all non-fixed weights with (6).

$$w_{ji} \leftarrow w_{ji} - \alpha \delta_j x_{ji} \quad (6)$$

- $w_{h_i b}$
- $w_{h_i x_1}$
- $w_{h_i x_2}$

w_{bh_i}

$$\begin{aligned} w_{h_1 b} &\leftarrow w_{h_1 b} - \alpha \delta_{h_1} b \\ w_{h_1 b} &\leftarrow -1 - 0.1 * -0.0275 * 1 \\ w_{h_1 b} &\leftarrow -0.99725 \end{aligned}$$

$$\begin{aligned} w_{h_2 b} &\leftarrow w_{h_2 b} - \alpha \delta_{h_2} b \\ w_{h_2 b} &\leftarrow 0.5 - 0.1 * 0.131 * 1 \\ w_{h_2 b} &\leftarrow 0.4869 \end{aligned}$$

$w_{h_i x_1}$

$$\begin{aligned} w_{h_1 x_1} &\leftarrow w_{h_1 x_1} - \alpha \delta_{h_1} x_1 \\ w_{h_1 x_1} &\leftarrow 2 - 0.1 * -0.0275 * 0 \\ w_{h_1 x_1} &\leftarrow 2 \end{aligned}$$

$$\begin{aligned} w_{h_2 x_1} &\leftarrow w_{h_2 x_1} - \alpha \delta_{h_2} x_1 \\ w_{h_2 x_1} &\leftarrow -3 - 0 \\ w_{h_2 x_1} &\leftarrow -3 \end{aligned}$$

$w_{h_i x_2}$

$$\begin{aligned} w_{h_1 x_2} &\leftarrow w_{h_1 x_2} - \alpha \delta_{h_1} x_2 \\ w_{h_1 x_2} &\leftarrow 1 - 0.1 * -0.0275 * 1 \\ w_{h_1 x_2} &\leftarrow -0.99725 \end{aligned}$$

$$\begin{aligned} w_{h_2 x_2} &\leftarrow w_{h_2 x_2} - \alpha \delta_{h_2} x_2 \\ w_{h_2 x_2} &\leftarrow -2 - 0.1 * 0.131 * 1 \\ w_{h_2 x_2} &\leftarrow -2.0131 \end{aligned}$$

2.2.1 Answers

All weights have been updated. To maintain good organization, all final weight values have been listed below.¹

- $w_{h_1b} \leftarrow -0.99725$
- $w_{h_2b} \leftarrow 0.4869$
- $w_{h_1x_1} \leftarrow 2$
- $w_{h_2x_1} \leftarrow -3$
- $w_{h_1x_2} \leftarrow -0.99725$
- $w_{h_2x_2} \leftarrow -2.0131$

3 Question 3: Bayes Nets

All questions in this section were addressed using the Bayes Net provided in “hw3v2.pdf” on page 3.

Note the following abbreviations that will be employed:

- Storm $\rightarrow S$
- Camp Fire $\rightarrow CF$
- Lightning $\rightarrow L$
- Thunder $\rightarrow T$
- Forest Fire $\rightarrow FF$

¹Note that w_{ji} is the weight from unit i to unit j .

3.1 a) What is the probability of a forest fire?

The probability of a forest fire is the sum of probabilities of a forest fire joint with all other possible sets of events.

$$\Pr(FF) = \Pr\left(FF, \sum L, \sum CF\right)$$

$$\Pr(FF) = \Pr(FF, L, CF) + \Pr(FF, L, \bar{C}F) + \Pr(FF, \bar{L}, CF) + \Pr(FF, \bar{L}, \bar{C}F)$$

$$\Pr(FF) = [\Pr(FF|L, CF)\Pr(L)\Pr(CF)] + [\Pr(FF|L, \bar{C}F)\Pr(L)\Pr(\bar{C}F)] \\ + [\Pr(FF|\bar{L}, CF)\Pr(\bar{L})\Pr(CF)] + [\Pr(FF|\bar{L}, \bar{C}F)\Pr(\bar{L})\Pr(\bar{C}F)]$$

$$\Pr(L) = \Pr\left(L, \sum S\right) = \Pr(L|S)\Pr(S) + \Pr(L|\bar{S})\Pr(\bar{S})$$

$$\Pr(L) = 0.5 * 0.1 + 0.05 * 0.9$$

$$\Pr(L) = 0.05 + 0.045 = 0.095$$

$$\Pr(\bar{L}) = \Pr\left(\bar{L}, \sum S\right) = \Pr(\bar{L}|S)\Pr(S) + \Pr(\bar{L}|\bar{S})\Pr(\bar{S})$$

$$\Pr(\bar{L}) = 0.5 * 0.1 + 0.95 * 0.9$$

$$\Pr(\bar{L}) = 0.05 + 0.855 = 0.905$$

$$\Pr(FF) = [0.5 * 0.095 * 0.75] + [0.4 * 0.095 * 0.25] \\ + [0.1 * 0.905 * 0.75] + [0.01 * 0.905 * 0.25]$$

$$\Pr(FF) = [0.035625] + [0.0095] \\ + [0.067875] + [0.0022625]$$

$$\Pr(FF) = 0.1152625$$

$$\boxed{\Pr(FF) \approx 0.115}$$

3.2 b) What is the probability of a forest fire given thunder?

$$\Pr(FF|T) = \frac{\Pr(FF, T)}{\Pr(T)}$$

$$\Pr(FF, T) = \Pr\left(FF, T, \sum CF, \sum L, \sum S\right)$$

$$\begin{aligned}\Pr(FF, T) &= \Pr(FF, T, CF, L, S) + \Pr(FF, T, CF, L, \bar{S}) \\ &\quad + \Pr(FF, T, CF, \bar{L}, S) + \Pr(FF, T, CF, \bar{L}, \bar{S}) \\ &\quad + \Pr(FF, T, \bar{C}F, L, S) + \Pr(FF, T, \bar{C}F, L, \bar{S}) \\ &\quad + \Pr(FF, T, \bar{C}F, \bar{L}, S) + \Pr(FF, T, \bar{C}F, \bar{L}, \bar{S})\end{aligned}$$

$$\begin{aligned}\Pr(FF, T, CF, L, S) &= \Pr(FF|L, C) \Pr(T|L) \Pr(CF) \Pr(L|S) \Pr(S) \\ &= 0.5 * 0.95 * 0.75 * 0.5 * 0.1 = 0.0178125\end{aligned}$$

$$\begin{aligned}\Pr(FF, T, CF, L, \bar{S}) &= \Pr(FF|L, C) \Pr(T|L) \Pr(CF) \Pr(L|\bar{S}) \Pr(\bar{S}) \\ &= 0.5 * 0.95 * 0.75 * 0.05 * 0.9 = 0.01603125\end{aligned}$$

$$\begin{aligned}\Pr(FF, T, CF, \bar{L}, S) &= \Pr(FF|\bar{L}, C) \Pr(T|\bar{L}) \Pr(CF) \Pr(\bar{L}|S) \Pr(S) \\ &= 0.1 * 0.2 * 0.75 * 0.5 * 0.1 = 0.00075\end{aligned}$$

$$\begin{aligned}\Pr(FF, T, CF, \bar{L}, \bar{S}) &= \Pr(FF|\bar{L}, C) \Pr(T|\bar{L}) \Pr(CF) \Pr(\bar{L}|\bar{S}) \Pr(\bar{S}) \\ &= 0.1 * 0.2 * 0.75 * 0.95 * 0.9 = 0.012825\end{aligned}$$

$$\begin{aligned}\Pr(FF, T, \bar{C}F, L, S) &= \Pr(FF|L, \bar{C}F) \Pr(T|L) \Pr(\bar{C}F) \Pr(L|S) \Pr(S) \\ &= 0.4 * 0.95 * 0.25 * 0.5 * 0.1 = 0.00475\end{aligned}$$

$$\begin{aligned}\Pr(FF, T, \bar{C}F, L, \bar{S}) &= \Pr(FF|L, \bar{C}F) \Pr(T|L) \Pr(\bar{C}F) \Pr(L|\bar{S}) \Pr(\bar{S}) \\ &= 0.4 * 0.95 * 0.25 * 0.05 * 0.9 = 0.004275\end{aligned}$$

$$\begin{aligned}\Pr(FF, T, \bar{C}F, \bar{L}, S) &= \Pr(FF|\bar{L}, \bar{C}F) \Pr(T|\bar{L}) \Pr(\bar{C}F) \Pr(\bar{L}|S) \Pr(S) \\ &= 0.01 * 0.2 * 0.25 * 0.5 * 0.1 = 0.000025\end{aligned}$$

$$\begin{aligned}\Pr(FF, T, \bar{C}F, \bar{L}, \bar{S}) &= \Pr(FF|\bar{L}, \bar{C}F) \Pr(T|\bar{L}) \Pr(\bar{C}F) \Pr(\bar{L}|\bar{S}) \Pr(\bar{S}) \\ &= 0.01 * 0.2 * 0.25 * 0.95 * 0.9 = 0.0004275\end{aligned}$$

$$\begin{aligned}\Pr(FF, T) &= 0.0178125 + 0.01603125 + 0.00075 + 0.012825 \\ &\quad + 0.00475 + 0.004275 + 0.000025 + 0.0004275\end{aligned}$$

$$\Pr(FF, T) = 0.05689625$$

$$\Pr(T) = \Pr\left(T, \sum L\right) = \Pr(T, L) + \Pr(T, \bar{L})$$

$$\Pr(T) = \Pr(T|L) \Pr(L) + \Pr(T|\bar{L}) \Pr(\bar{L})$$

$$\Pr(T) = 0.95 * 0.095 + 0.2 * 0.905 = 0.27125$$

$$\Pr(FF|T) = \frac{0.05689625}{0.27125}$$

$$\Pr(FF|T) = 0.20975576036866359447004608294931$$

$$\boxed{\Pr(FF|T) \approx 0.21}$$

3.3 c) What is the probability that there is a storm given that there is a forest fire?

$$\begin{aligned}
 \Pr(S|FF) &= \frac{\Pr(S, FF)}{\Pr(FF)} \\
 \Pr(S, FF) &= \Pr\left(S, FF, \sum L, \sum CF\right) \\
 &= \Pr(S, FF, L, CF) + \Pr(S, FF, L, \bar{C}F) + \Pr(S, FF, \bar{L}, CF) + \Pr(S, FF, \bar{L}, \bar{C}F) \\
 \Pr(S, FF, L, CF) &= \Pr(S) \Pr(FF|L, CF) \Pr(L|S) \Pr(CF) \\
 &= 0.1 * 0.5 * 0.5 * 0.75 = 0.01875 \\
 \Pr(S, FF, L, \bar{C}F) &= \Pr(S) \Pr(FF|L, \bar{C}F) \Pr(L|S) \Pr(\bar{C}F) \\
 &= 0.1 * 0.4 * 0.5 * 0.25 = 0.005 \\
 \Pr(S, FF, \bar{L}, CF) &= \Pr(S) \Pr(FF|\bar{L}, CF) \Pr(\bar{L}|S) \Pr(CF) \\
 &= 0.1 * 0.1 * 0.5 * 0.75 = 0.00375 \\
 \Pr(S, FF, \bar{L}, \bar{C}F) &= \Pr(S) \Pr(FF|\bar{L}, \bar{C}F) \Pr(\bar{L}|S) \Pr(\bar{C}F) \\
 &= 0.1 * 0.01 * 0.5 * 0.25 = 0.000125 \\
 \Pr(S, FF) &= 0.01875 + 0.005 + 0.00375 + 0.000125 = 0.027625 \\
 \Pr(FF) &= 0.1152625 \\
 \Pr(S|FF) &= 0.23967031775295521093156924411669 \\
 \boxed{\Pr(S|FF) \approx 0.24}
 \end{aligned}$$

3.4 d) What is the probability of thunder given that there is no storm?

$$\begin{aligned}
 \Pr(T|\bar{S}) &= \frac{\Pr(T, \bar{S})}{\Pr(\bar{S})} \\
 \Pr(T, \bar{S}) &= \Pr(T, \bar{S}, \sum L) \\
 &= \Pr(T, \bar{S}, L) + \Pr(T, \bar{S}, \bar{L}) \\
 &= \Pr(T|L) \Pr(\bar{S}) \Pr(L|\bar{S}) + \Pr(T|\bar{L}) \Pr(\bar{S}) \Pr(\bar{L}|\bar{S}) \\
 &= 0.95 * 0.9 * 0.05 + 0.2 * 0.9 * 0.95 = 0.04275 + 0.171 \\
 &= 0.21375 \\
 \Pr(\bar{S}) &= 0.9 \\
 \Pr(T|\bar{S}) &= \frac{0.21375}{0.9} \\
 \boxed{\Pr(T|\bar{S}) = 0.2375}
 \end{aligned}$$

3.5 e) What is the probability of a camp fire and a forest fire?

$$\begin{aligned}\Pr(CF, FF) &= \Pr(CF, FF, \sum L) \\ &= \Pr(CF, FF, L) + \Pr(CF, FF, \bar{L}) \\ &= \Pr(CF) \Pr(FF|CF, L) \Pr(L) + \Pr(CF) \Pr(FF|CF, \bar{L}) \Pr(\bar{L}) \\ &= 0.75 * 0.5 * 0.095 + 0.75 * 0.1 * 0.905 = 0.035625 + 0.067875\end{aligned}$$

$\Pr(CF, FF) = 0.1035$

Question 4: Spam Classification with a Naïve Bayes Classifier

This work has been completed separately and handed in electronically.