

# Assignment 5: Naive Bayes and Perceptrons

CS4811 — Artificial Intelligence

Due December 13th, 11:59pm

1. (15 points) *Perceptrons and Neural Networks*

- (a) (4 points) Consider a perceptron with two inputs,  $x_1$  and  $x_2$ . The perceptron uses a transfer function with a threshold value of 0.5 (any value  $\geq 0.5$  outputs 1, a value  $< 0.5$  outputs 0). The input and weight vectors are formatted as  $\vec{x} = \langle x_1, x_2 \rangle$  and  $\vec{w} = \langle w_1, w_2 \rangle$ . *Note that there is no intercept term for the weight vector.*

Assume you have the following data set:

Sample	$x_1$	$x_2$	$y$
1	0	0	1
2	0	1	0
3	1	0	0
4	1	1	0

Assume the weights are initialized to  $\vec{w} = \langle 0.7, 0.3 \rangle$ . Apply each data sample to the perceptron in turn (sample 1, sample 2, etc.). For each mis-classified sample, use the perceptron training rule, with a learning rate of 0.2 to update the weights.

Please provide a table, like below, showing how the weights change as each sample is applied. Include your work (calculations) to show how the new weight values were reached.

**Solution:**

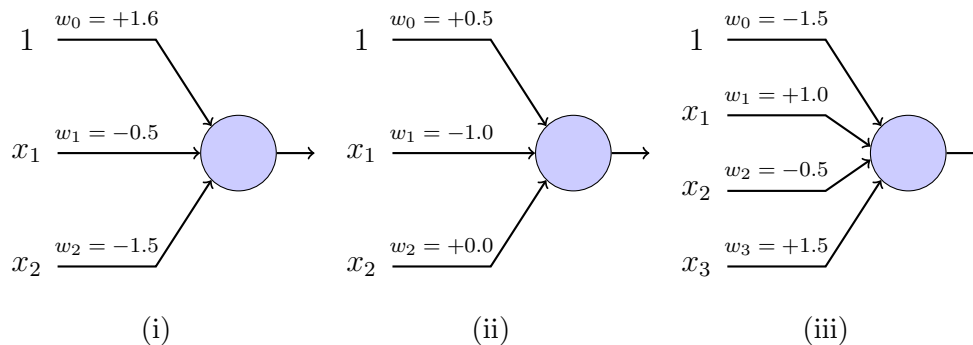
	$w_1$	$w_2$
Initial Weights	0.7	0.3
(1) Weights after observing $\vec{x} = \langle 0, 0 \rangle$	0.70	0.30
(2) Weights after observing $\vec{x} = \langle 0, 1 \rangle$	0.70	0.30
(3) Weights after observing $\vec{x} = \langle 1, 0 \rangle$	0.50	0.30
(4) Weights after observing $\vec{x} = \langle 1, 1 \rangle$	0.30	0.10

$x_1$	$x_2$	$y$								
0	0	1								
0	1	0				$f(w'$				
1	0	0				$f(w'$				
1	1	0								
$x_1$	$x_2$	$w_1$	$w_2$	$f(w*x)$	out	$y$	$\eta$	$w_1'$	$w_2'$	
1	0	0.7	0.3	=SUM((B8*D8)+(C8*E8))	0	1	0.2	=D8+(I8*((H8-G8)*B8))	=E8+(I8*((J8-H8)*C8))	
2	0	1	=J8	=K8	=SUM((B9*D9)+(C9*E9))	0	0	0.2	=D9+(I9*((H9-G9)*B9))	=E9+(I9*((H9-G9)*C9))
3	1	0	=J9	=K9	=SUM((B10*D10)+(C10*E10))	1	0	0.2	=D10+(I10*((H10-G10)*B10))	=E10+(I10*((H10-G10)*C10))
4	1	1	=J10	=K10	=SUM((B11*D11)+(C11*E11))	1	0	0.2	=D11+(I11*((H11-G11)*B11))	=E11+(I11*((H11-G11)*C11))

- (b) (1 point) If the application of the perceptron training rule continues in a above, would the next cycle through the data have any misclassified samples?

**Solution:** It should not have any misclassified as all samples have been processed and all possible binary combinations of  $x_1, x_2$  have been bracketed.

- (c) (3 points) Which Boolean functions do the following perceptron units (i), (ii), and (iii) represent?



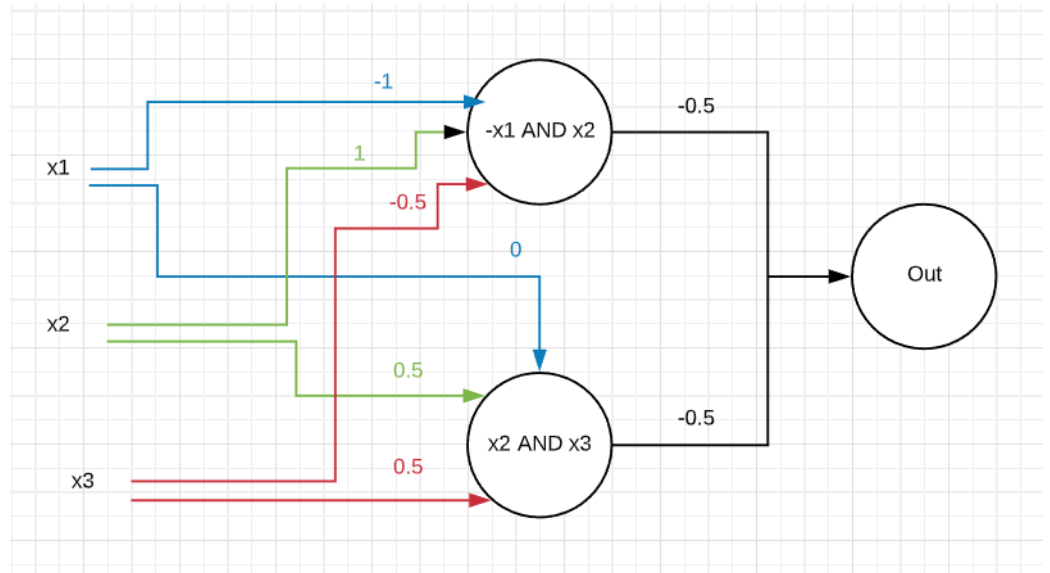
Assume all inputs take values  $\{0, 1\}$  (false, true) and the perceptrons the a transfer function with a threshold value of 0 ( $x \geq 0$  outputs 1,  $x < 0$  outputs 0).

**Solution:** (i) NOT  $x_1$  or  $x_2$  (ii) NOT  $x_1$  (iii)  $x_1$  AND  $x_3$

- (d) (5 points) Construct a 2-layer network (1 hidden layer and 1 output layer) which recognizes the Boolean expression

$$(\text{not } x_1 \text{ and } x_2) \text{ nor } (x_2 \text{ and } x_3)$$

using binary input values  $\{0, 1\}$ . Label each edge with its weight. You may include a bias ( $x_0 = 1$ ) if you wish. You are free to use as many perceptrons in the hidden layer as you wish, but 2 hidden perceptrons are sufficient.



**Solution:**

(e) (2 points) Can the Boolean expression

$$(x_2 \text{ nand } x_3) \text{ or } (x_1 \text{ and not } x_2) \text{ or } (\text{not } x_1 \text{ and } x_2)$$

be learned with a single perceptron? If so, construct the perceptron, labeling each edge with its weight. Otherwise, explain why not.

**Solution:** No, this is not possible with a single perceptron. This requires at least 2 perceptrons because it calls for both 1 and 0 for  $x_1$  out in a single node.

2. (10 points) *Naive Bayes*

Consider the following data set of three Boolean variables: *Weather*, *Roads*, and *Temperature*, and a label *Transport*. This data set describes instances in which your friend drove or walked to campus depending on the conditions outside.

<i>Weather</i>	<i>Roads</i>	<i>Temperature</i>	<i>Transport</i>
<i>Clear</i>	<i>Safe</i>	<i>Cold</i>	<i>Drive</i>
<i>Clear</i>	<i>Slippery</i>	<i>Cold</i>	<i>Walk</i>
<i>Clear</i>	<i>Slippery</i>	<i>Frigid</i>	<i>Drive</i>
<i>Snowing</i>	<i>Safe</i>	<i>Cold</i>	<i>Walk</i>
<i>Snowing</i>	<i>Slippery</i>	<i>Frigid</i>	<i>Drive</i>

(a) (6 points) Calculate the conditional probability distributions for each variable (e.g. *Weather*, *Roads*, *Temperature*) given the label *Transport*:

1.  $P(\text{Weather} \mid \text{Transport})$
2.  $P(\text{Roads} \mid \text{Transport})$

3.  $P(\text{Temperature} \mid \text{Transport})$

**Solution:**

1.  $P(\text{Weather} \mid \text{Transport})$

$$P(\text{Clear} \mid \text{Walk}) = 1/2 = 0.5$$

$$P(\text{Snowing} \mid \text{Walk}) = 1/2 = 0.5$$

$$P(\text{Clear} \mid \text{Drive}) = 2/3 = .666$$

$$P(\text{Snowing} \mid \text{Drive}) = 1/3 = .333$$

2.  $P(\text{Roads} \mid \text{Transport})$

$$P(\text{Safe} \mid \text{Walk}) = 1/2 = 0.5$$

$$P(\text{Slippery} \mid \text{Walk}) = 1/2 = 0.5$$

$$P(\text{Safe} \mid \text{Drive}) = 1/3 = .333$$

$$P(\text{Slippery} \mid \text{Drive}) = 2/3 = .666$$

3.  $P(\text{Temp} \mid \text{Transport})$

$$P(\text{Cold} \mid \text{Walk}) = 2/2 = 1.0$$

$$P(\text{Frigid} \mid \text{Walk}) = 0/2 = 0$$

$$P(\text{Cold} \mid \text{Drive}) = 1/3 = .333$$

$$P(\text{Frigid} \mid \text{Drive}) = 2/3 = .666$$

- (b) (4 points) Suppose today that it is snowing ( $\text{Weather} = \text{Snowing}$ ), the roads are ( $\text{Roads} = \text{Safe}$ ), and the temperature is frigid ( $\text{Temperature} = \text{Frigid}$ ). What are the probabilities associated with the two labels ( $\text{Drive}$  and  $\text{Walk}$ )? Which classification is preferred? (As the probabilities are simple, you are required to carry out the calculations by hand and show your work)

**Solution:**

$$P(\text{Drive} \mid \text{Snowing, Safe, Frigid}) = 1/3 \times 1/3 \times 2/3 = 2/27 = 0.0741$$

$$P(\text{Walk} \mid \text{Snowing, Safe, Frigid}) = 1/2 \times 1/2 \times 0/2 = 0$$

Drive is the preferred classification. This is due to  $P(\text{Frigid} \mid \text{Drive}) = 2/3 = .666$  and  $P(\text{Frigid} \mid \text{Walk}) = 0/2 = 0$ .