

Question 3

- 1). If $C = c_1$ AND $B = b_1$ THEN NO (A)
- 2). If $C = c_1$ AND $B = b_2$ THEN YES (B)
- 3). If $C = c_2$ AND $A = a_1$ THEN YES (C)
- 4). If $C = c_2$ AND $A = a_2$ AND $B = b_1$ THEN YES (D)
- 5). If $C = c_2$ AND $A = a_2$ AND $B = b_2$ THEN NO (E)
- 6). If $C = c_2$ AND $A = a_3$ THEN NO (F)
- 7). If $C = c_3$ THEN NO (G)

2). T is consistent with S_1 because of A.

T is consistent with S_2 because of D.

T is not consistent with S_3 because of G.

T is not consistent with S_4 because of E.

Question 4

1). The maximum a posteriori hypothesis is $h_{MAP} = \underset{h \in H}{\operatorname{argmax}} \frac{\varphi(D|h) \varphi(h)}{\varphi(D)} = \underset{h \in H}{\operatorname{argmax}} \varphi(D|h) \varphi(h)$

If we assume that $\varphi(h_i) = \varphi(h_j)$ we can simplify and we obtain the maximum likelihood hypothesis $h_{ML} = \underset{h \in H}{\operatorname{argmax}} \varphi(D|h)$

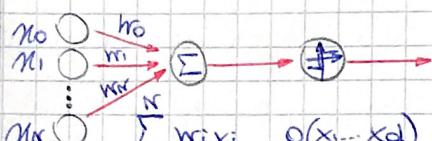
2). The Bayes Optimal Classifier is an optimal classifier, it returns always the optimal solution. Given a target function $f: X \rightarrow V$ with $V = \{V_1 \times V_2 \times \dots \times V_n\}$

$$V_{ML} = \underset{v \in V}{\operatorname{argmax}} \sum_{h \in H} \varphi(v|x, D|h) \varphi(h|D) = \underset{v \in V}{\operatorname{argmax}} \sum_{h \in H} \varphi(v|x, h) \varphi(h|D)$$

3). Bayes Optimal Classifier is an optimal classifier but can be used if the hypothesis space is not large or if we have analytical solutions, otherwise is not usable.

Question 3

1). The perceptron model is based on the following structure:



$$o(x_1 \dots x_d) = \begin{cases} 1 & \text{if } w_0 x_0 + \dots + w_d x_d > 0 \\ -1 & \text{otherwise} \end{cases}$$

$\sum_{i=0}^n w_i x_i = o(x_1 \dots x_d)$. For the moment we ignore this function.

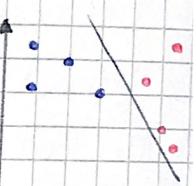
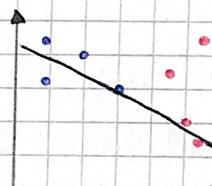
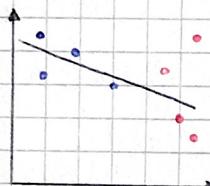
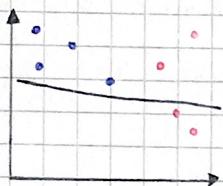
$$E(w) = \frac{1}{2} \sum_{n=1}^N (t_n - o_n)^2 = \frac{1}{2} \sum_{n=1}^N (t_n - w^T x_n)^2$$

$$\frac{\partial E(w)}{\partial w_i} = \sum_{n=1}^N (t_n - w^T x_n) (-x_{in}). \text{ To find } w^* \text{ we use a sequential algorithm:}$$

$$\hat{w} \leftarrow \hat{w} + \Delta w_i \text{ with } \Delta w_i = -\eta \sum_{n=1}^N (t_n - w^T x_n) (x_{in}) \quad \text{Now we consider } o(x_1 \dots x_d)$$

$$\hat{w} \leftarrow \hat{w} - \eta \sum_{n=1}^N (t_n - \text{sign}(w^T x_n)) (x_{in})$$

2).



Question 4

$$1). \dim(W_1) = 128 \times 10 \quad \dim(W_2) = 50 \times 10 = 500$$

$$2). h = g(W^T x + c) \text{ with } g(z) = \max(0, z) \text{ ReLU function}$$

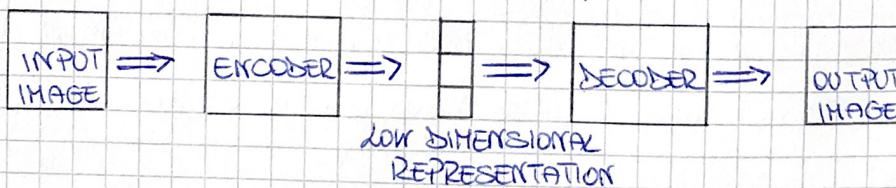
$$y(n, N) = w^T \max(0, W^T x + c) + b$$

Question 6

1). An autoencoder is simply a combination of two NNs: an encoder and a decoder.

The training is based on reconstruction loss and in an autoencoder we have hidden layers with reduced size (bottlenecks). Given a dataset $\{x_n\}$ an autoencoder is trained with x_n both in input and in output.

2).



LOW DIMENSIONAL
REPRESENTATION