Intelligent Systems: Exam Block 2

ETSINF, Universitat Politècnica de València, January 13, 2016

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Tick only one choice among the given options.

1 D Which of the following expressions is **CORRECT**?

A)
$$P(x | y) = \frac{1}{P(z)} \sum_{x} P(x, y, z)$$

B)
$$P(x | y) = \frac{1}{P(z)} \sum_{z} P(x, y, z)$$

C)
$$P(x | y) = \frac{1}{P(y)} \sum_{x} P(x, y, z)$$

D)
$$P(x \mid y) = \frac{1}{P(y)} \sum_{z} P(x, y, z)$$

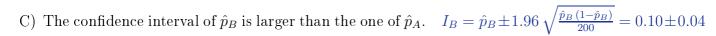
2 A physician knows that:

- The meningitis disease causes neck stiffness in the 70% of the cases.
- The prior probability that a patient suffers from meningitis is 1 / 100 000.
- The prior probability that a patient has neck stiffness is 1 %.

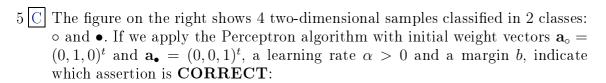
Based on the above knowledge, the probability P that a patient who has neck stiffness suffers from meningitis is:

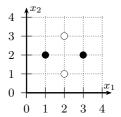
A)
$$0.000 \le P < 0.001$$
 $P = P(m \mid r) = \frac{P(m) P(r \mid m)}{P(r)} = \frac{1/100 000 \cdot 70/100}{1/100} = 0.0007$

- B) $0.001 \le P < 0.002$
- C) $0.002 \le P < 0.003$
- D) $0.003 \le P$
- 3 D Let's consider a typical classification problem in C classes and objects represented through D-dimensional real feature vectors. In general, we can say that it it more difficult to find an accurate classifier when
 - A) the values of C and D are smaller
 - B) the value of C is smaller and the value of D is larger
 - C) the value of C is larger and the value of D is smaller
 - D) the values of C and D are larger
- 4 B We have learnt two different classifiers, c_A and c_B , for a classification problem. The probability of error of c_A has been empirically estimated from 100 test samples, obtaining an empirical estimate of error $\hat{p}_A = 0.10 \ (10 \%)$. Similarly, the probability of error of c_B has been empirically estimated but with a set of 200 test samples, obtaining an empirical estimate of error of 10 %, too ($\hat{p}_B = 0.10$). Based on these estimations, we can affirm with a 95 % of confidence that:
 - A) The confidence intervals of \hat{p}_A y \hat{p}_B are identical.
 - B) The confidence interval of \hat{p}_A is larger than the one of \hat{p}_B . $I_A = \hat{p}_A \pm 1.96 \sqrt{\frac{\hat{p}_A (1-\hat{p}_A)}{100}} = 0.10 \pm 0.06$



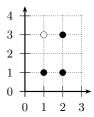
D) In this case, the confidence intervals of \hat{p}_A and \hat{p}_B are irrelevant because the estimate of error is the same.





- A) The algorithm will converge for some b > 0
- B) The algorithm only converges if $b \leq 0$
- C) If b > 0 there is no convergence but, by adjusting α , we can obtain good solutions after a finite number of iterations with respect to the probability of classification error (with 25 % of misclassification error)
- D) The algorithm is not applicable in this case because the classes are non-linearly separable.
- 6 B Which is the number of errors of a minimum-error linear classifier for the training samples of the above question?
 - A) 0
 - B) 1
 - C) 2
 - D) 3
- 7 B Given a linear classifier of two classes \circ and \bullet with weight vectors $\mathbf{a}_{\circ} = (3, 1, 1)^t$ and $\mathbf{a}_{\bullet} = (1, 2, 1)^t$, respectively (the first component is the threshold or independent term of the linear function), which assertion is **CORRECT**?
 - A) There are four decision regions because there are two weight vectors and it is a two-dimensional representation space
 - B) The weight vectors $\mathbf{a}_{\circ} = (2, -2, -2)^t$ and $\mathbf{a}_{\bullet} = (-2, 0, -2)^t$ define the same decision boundary than the weight vectors given in the question statement. The decision boundary equation is: $\mathbf{a}_{\circ}^t \mathbf{y} = \mathbf{a}_{\bullet}^t \mathbf{y}$. In both cases, we have: $y_1 = 2$.
 - C) The weight vectors $\mathbf{a}_{\circ} = (1, 2, 1)^t$ y $\mathbf{a}_{\bullet} = (3, 1, 1)^t$ define an equivalent classifier to the one given in the statement Opposed decision regions.
 - D) The decision boundary is defined as a plane in \mathbb{R}^3 because the weight vectors are three-dimensional
- 8 D Let's assume we apply the Decision Classification Tree (DCT) algorithm for a two-class problem, A and B. The DCT algorithm reaches a node t which includes two data: one sample that belongs to class A and the other belongs to class B. The entropy impurity of t, $\mathcal{I}(t)$, is:
 - A) $\mathcal{I}(t) < 0.0$
 - B) $0.0 \le \mathcal{I}(t) < 0.5$
 - C) $0.5 \le \mathcal{I}(t) < 1.0$
 - $\mathcal{I}(t) = -\hat{P}(A \mid t) \, \log_2 \hat{P}(A \mid t) \hat{P}(B \mid t) \, \log_2 \hat{P}(B \mid t) = -\frac{1}{2} \log_2 \frac{1}{2} \frac{1}{2} \log_2 \frac{1}{2} = 1$

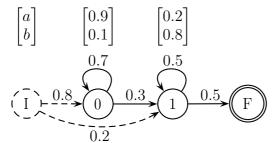
9 D The figure on the right shows a two-cluster partition of four two-dimensional data (represented by the symbols • and ∘). The Sum of Square Errors (SSE) of this partition is $J = \frac{30}{9}$. The transfer of the point $(2,3)^t$ from cluster \bullet to \circ leads to an increase in the SSE, ΔJ , such that:

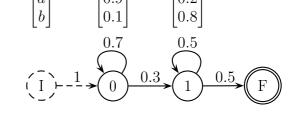


- A) $\Delta J > 0$
- B) $0 \ge \Delta J > -1$
- C) $-1 \ge \Delta J > -2$
- D) $-2 \ge \Delta J$ $\Delta J = -\frac{21}{9} = -2.33$ $(J = \frac{30}{9} \to J = 1)$
- 10 B Two well-known versions of the K-means algorithm are the Duda and Hart (DH) version and the "popular" version. Assuming both versions are applied in the same initial partition, indicate which of the following assertions is TRUE:
 - A) Both versions will get the same optimized partition
 - B) The DH version will get a partition which cannot be further improved with the "popular" version
 - C) The "popular" version will get a partition which cannot be further improved with the DH version
 - D) The final partition obtained with DH would could be further improved with the "popular" version and viceversa
- 11 A Given the Markov model M_A of the question 12, the approximated probability of the string "bba" calculated with Viterbi is:
 - A) $0.003200 \ P(bba, q_1q_2q_3 = 111 \mid M_A) = 0.2 \cdot 0.8 \cdot 0.5 \cdot 0.8 \cdot 0.5 \cdot 0.2 \cdot 0.5 = 0.0032$
 - B) 0.004328
 - C) 0.006400
 - D) None of the above options are correct.
- 12 B We have two equiprobable classes, A and B, for classifying strings of symbols in the alphabet $\Sigma = \{a, b\}$. The conditional probabilities of the classes are characterized by the Markov models

$$Model M_A: P(x \mid A) = P(x \mid M_A)$$

$$Model M_B: P(x \mid B) = P(x \mid M_B)$$





Indicate the CORRECT option if we want to classify the string "bba" by minimum classification error:

- A) Either A or B because both classes are equiprobable
- B) Class A. $\hat{c} = \arg \max_{c} P(c \mid "bba") = \arg \max_{c} P(c) P("bba" \mid c) = \arg \max_{c} P("bba" \mid c)$
- C) Class B. $P("bba" | A) \approx \tilde{P}("bba" | A) = 0.0032 \gg P("bba" | B) \approx \tilde{P}("bba" | B) = 0.0012 \rightarrow \hat{c} = A$
- D) It cannot be determined because M_B does not satisfy the normalization conditions.

13 \square Given the Markov model M_A of the question 12, if we apply the Forward algorithm to the string "bba", mark the **CORRECT** expression:

A)
$$\alpha(q=1, t=3) = \alpha(q=0, t=2) \cdot A_{01} \cdot B_{1a}$$

B)
$$\alpha(q=1, t=3) = \alpha(q=1, t=2) \cdot A_{11} \cdot B_{1a}$$

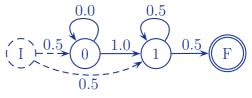
C)
$$\alpha(q=1,t=3) = \alpha(q=0,t=2) \cdot A_{01} \cdot B_{1a} + \alpha(q=1,t=2) \cdot A_{11} \cdot B_{1a}$$

D)
$$\alpha(q=1,t=3) = \alpha(q=0,t=2) \cdot A_{01} \cdot B_{1a} \cdot \alpha(q=1,t=2) \cdot A_{11} \cdot B_{1a}$$

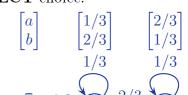
14 \square Given the Markov model M_A of the question 12, if we apply ONE iteration of Viterbi re-estimation algorithm with the strings "bba" and "ab", which assertion is **CORRECT**?:

A)
$$\pi_0 = 1$$

- B) No changes are produced in the model.
- C) All the transition probabilities change their value.
- D) Some of the transition and emission probabilities of state 0 are null.



- 15 B Given a Markov model with states $Q = \{0, 1, F\}$ and alphabet $\Sigma = \{a, b\}$ initialized through a linear segmentation with the strings "bbaa" y "ab", indicate the **CORRECT** choice:
 - A) Some emission probabilities are null.
 - B) It holds that $A_{00} = A_{11}$ and $A_{01} = A_{1F}$
 - C) It holds that $\pi_0 = \pi_1$
 - D) It holds that $B_{0a} = B_{1a}$



0.0

