



南方科技大学
SOUTHERN UNIVERSITY OF SCIENCE AND TECHNOLOGY

Course Name: Machine Learning Exam Duration: 2 hours

Dept.: Department of Computer Science and Engineering

Exam Paper Setter(Signature): _____ Reviewer(Signature): _____

Question No.	1	2	3	4	5	6	7	8	9	10
Score	20	5	5	10	10	10	20	20	10	

This exam paper contains 9 questions and the score is 110 in total. (Please hand in your exam paper, answer sheet, and your scrap paper to the proctor when the exam ends.)

Problem I Multiple Choice (20 Points)

(only one correct answer for each question)

A

1. (2 points) The most suitable loss function for linear regression is _____
- A. the sum of squared errors
 - B. the entropy function
 - C. the cross entropy function
 - D. the number of mistakes

B

2. (2 points) The logistic regression is a _____ regression technique that is used to model data having a _____ outcome.
- A. linear, numeric
 - B. linear, binary
 - C. nonlinear, numeric
 - D. nonlinear, binary

B

3. (2 points) The most suitable loss functions for neural networks are _____
- A. the entropy function and KL divergence
 - B. the squared error function and cross-entropy function
 - C. the cross-entropy function and KL divergence
 - D. the number of mistakes and entropy function

C

4. (2 points) The most suitable loss functions for SVM are _____?
- A. the entropy function and KL divergence

- B. the squared error function and cross-entropy function
- C. the hinge error function and ϵ -insensitive error function
- D. the number of mistakes and entropy function

A

5. (2 points) The most suitable loss functions for GMM are _____.

- A. the maximum likelihood function and maximum a posterior function
- B. the squared error function and cross-entropy function
- C. the cross-entropy function and KL divergence
- D. the number of mistakes and entropy function

B

6. (2 points) The three most important problems for HMM are _____.

- A. message propagation, expectation and maximization
- B. learning, evaluation and decoding
- C. belief propagation, parameter learning and state estimation
- D. ML learning, MAP learning, and fully Bayesian learning

D

7. (2 points) The reinforcement learning problem can be solved through _____.

- A. dynamic programming if the rewards and transition probabilities are known
- B. the Monte Carlo method if only reward functions are known
- C. the temporal difference method if the online learning is preferred
- D. all of the above

C

8. (2 points) Which activation function has the least computational complexity?

- A. tanh
- B. sigmoid
- C. ReLu
- D. Leaky ReLu

C

9. (2 points) Which of the following is NOT a way to reduce the model under-fitting?

- A. increase the amount of training data
- B. increase the model complexity
- C. decrease the number the model parameters with prior distributions
- D. decrease the amount of data augmentation

D

10. (2 points) Which of the following is NOT true for a machine learning system?

- A. It has three main components: model, error function and optimization algorithm.
- B. It reduces the KL divergence between the distributions of data and the model.
- C. It involves the procedure of EM for incomplete data problems.
- D. It will achieve the **global optimum** if the training data is sufficient enough.

Problem II Numerical Calculation (40 Points)

- (1) **Linear Regression (5 points).** For three points $\{(1, 0), (3, 3), (5, 4)\}$, what is the linear regression function for the least squared errors (*assuming $y = ax + b$, using pseudo-inverse*)?
- (2) **Maximum margin classifier (5 points).** For one class of two points $\{(1, 2), (2, 2)\}$ and another class of two points $\{(4, 4), (5, 6)\}$, what are the support vectors and what is the decision boundary's function (*plot your answer*) ?
- (3) **Clustering (10 points).** For four points with two classes, $\{(1, 2), (2, 2), (4, 4), (5, 6)\}$, how to achieve two cluster centers using the K-means algorithm?

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- (4) **Factor Graph (10 points).** How to design a factor graph to solve the following linear equation $[2 \ 4 \ 3]^T = [1 \ 0 \ 1; 1 \ 2 \ 1; 1 \ 1 \ 1][x_1 \ x_2 \ x_3]^T$? Assuming the initial value of X is $[0 \ 1 \ 1]$, show the computation procedure of one iteration.
- (5) **Hidden Markov Model (10 points).** For a HMM, the hidden states are $\{\text{bull}, \text{bear}\}$, the observation variables are $\{\text{rise}, \text{fall}\}$, the initial state probability distribution π is $[0.5 \ 0.5]^T$, the transition probability distribution A is $[0.4 \ 0.7; 0.6 \ 0.3]$, and the observation probability distribution B is $[0.8 \ 0.1; 0.2 \ 0.9]$. If the observation sequence is $\{\text{fall fall rise}\}$, please show the computation procedure for estimating the most likely state sequence?

Problem III Theoretical Analysis (40 Points)

- (1) **Density Mixture Model (20 points).** For a random variable X distributed in a mixture of probability densities, the joint distribution of X and its latent variable Z with the model θ is given by

$$p(X, Z | \theta) = \prod_{i=1}^K [\pi_i p(X | \theta_i)]^{z_i}$$

- a) Summarize the general EM scheme for DMM (**E**-step and **M**-step).
- b) Assuming each probability density is Bernoulli, i.e., $p(X | \theta_i) = \theta_i^x (1 - \theta_i)^{1-x}$, please derive the corresponding model learning procedure for $\{\pi_i, \theta_i\}$ under the EM scheme.
- (2) **Hidden Markov Model (20 Points).** For a finite-state random sequence $\{Z_t\}$ with the model of $\{\pi, A\}$ and its observation sequence is $\{X_t\}$, the joint distribution of X and Z with the model θ is given by $p(X, Z | \theta)$.
- a) Summarize the general EM scheme for HMM (**E**-step and **M**-step).
- b) Assuming each observation probability density is Bernoulli, i.e.

$$p(X, Z | \theta) = \prod_{i=1}^K [p(z_i) p(X | \theta_i)]^{z_i}$$

please derive the corresponding model learning procedure under the EM scheme.

Problem IV Expectation and Maximization (Bonus 10 Points)

- (1) What is the EM procedure? When do we need the EM procedure for machine learning?
- (2) What is the EM procedure in terms of the Q function?
- (3) What is the EM procedure in terms of likelihood and KL divergence?
- (4) What is the EM procedure in terms of optimization of non-convex function?
- (5) What is the EM procedure for the factor graph network model?