



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)

- (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
- (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
- (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
- (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  $\epsilon$ -insensitive error function
  - D. the number of mistakes and entropy function
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
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
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
8. (2 points) Which activation function has the least computational complexity?
- A. tanh
  - B. sigmoid
  - C. ReLu
  - D. Leaky ReLu
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
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?
- (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  $\pi$  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  $\theta$  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  $\theta$  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?