

Deep Learning and Practice — Final Exam

Date: Thursday, September 2, 2021

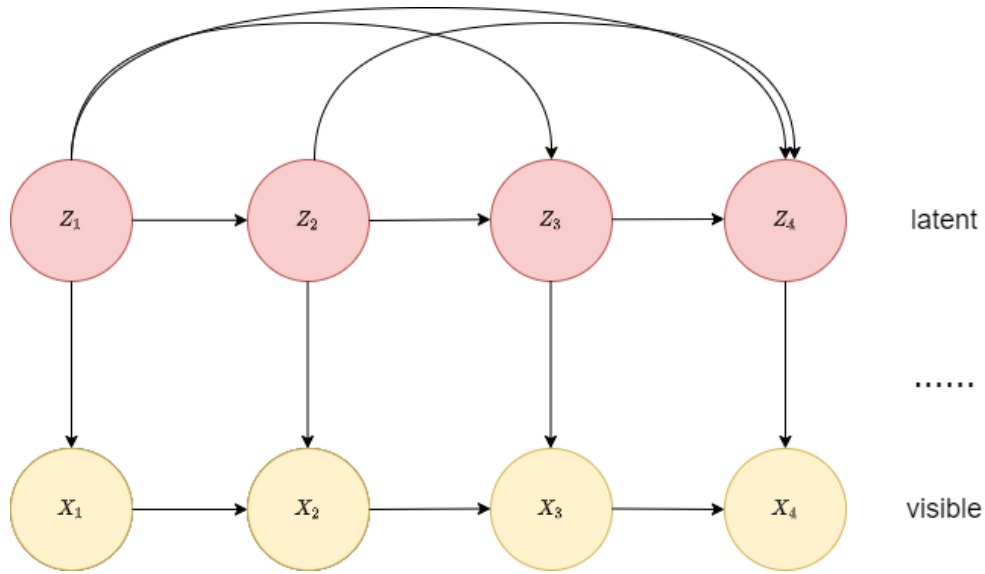
Time: 01:20pm – 04:20pm (180 minutes)

Format: Open book

Instructions:

- 1) You may give your answers in Chinese or English.
- 2) Please give your answers in succinct phrases or point form.
- 3) Please write your answers clearly (with explicit denotation of labels and symbols used).

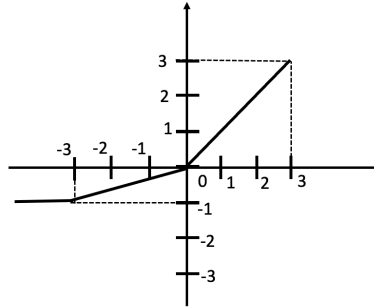
1. (25 pts) Consider the following latent factor model, where $Z_i, i = 1, 2, \dots, T$ are latent variables and $X_i, i = 1, 2, \dots, T$ are visible variables.



- (a) (8 pts) Design an inevitable mapping from X_i 's to their latent representations Z_i 's using the flow model idea. Use $T = 3$ as an example. Think about how you should convert each of the X_i 's into the corresponding Z_i 's and what the conditioning variable should be utilized in each of these conversions. The conversions may involve coupling layers as fundamental building blocks.
- (b) (2 pts) According to the graphical model, factorize the joint distribution $p(Z_{1-T})$.
- (c) (5 pts) Describe how you would evaluate the probability $p(X_1, X_2, X_3)$ of one specific sample X_1, X_2, X_3 once the model is trained.
- (d) (5 pts) How would you train the flow model in (a)?
- (e) (5 pts) Describe how you would sample X_i 's from Z_i 's using the designed flow model.

2. (20 pts) Convolution, Pooling, Activation function, CNN

- (a) (4 pts) What is the output feature map size for an 256×256 input image after convolution with kernel $(3,3)$, padding $(2,2)$, and stride $(2,2)$?
- (b) (4 pts) Describe the function of pooling layer.
- (c) (3 pts) Use the maxout unit to design the activation function below.



- (d) (6 pts) What may cause gradient vanish problem and how to solve it? Explain your answer.
- (e) (3 pts) What are the major reasons that contribute to the success of convolutional neural networks?

3. (15 pts) Training the VAE.

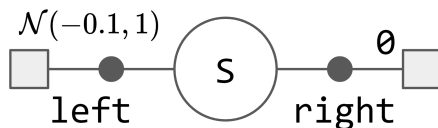
- (a) (3 pts) In training the VAE, we try to maximize a variational lower bound on the data log-likelihood. Explain the main idea and provide the exact objective function to be maximized.
- (b) (3 pts) What distribution does the approximate posterior $q(z|x)$ take for training VAE? Is this an assumption?
- (c) (3 pts) Explain the notion of the re-parameterization trick.
- (d) (3 pts) True or False: In maximizing the variational lower bound, the approximate posterior $q(z|x)$ should ideally be identical to the prior $p(z)$ when the variational lower bound is maximized. Explain your answer.
- (e) (3 pts) How would you evaluate the KL divergence $KL(q(z|x)||p(z))$ if the prior $p(z)$ is replaced with a Gaussian Mixture distribution?

4. (10 pts) Compare VAE, GAN, WGAN, and Flow Models in terms of (a) their generator/decoder outputs (stochastic or deterministic) and (b) training objectives. (c) Which models allow you to evaluate the probability of a specific data sample? (d) Which models allow you to generate the latent code of a specific data sample. (e) From their training losses, how do you evaluate the generator quality for these models? Which models cannot you decide the generator quality?

5. (14 pts)

- (a) (7 pts) Prove the following theorem: For any ε -greedy policy π , the ε -greedy policy π' with respect to q_π is an improvement, that is $v_{\pi'}(s) \geq v_\pi(s)$.
- (b) (7 pts) Show how ε -greedy policy improvement will converge to the optimal.

6. (18 pts) Consider the undiscounted MDP shown below.



Episodes start in S with a choice between two actions, left and right. The right action transitions immediately to the terminal state with a reward of zero. The left action also transitions to the terminal state but with a reward drawn from a normal distribution with mean -0.1 and variance 1.0 .

- (a) (2 pts) What is the optimal state-action value $Q^*(S, \text{left})$ and $Q^*(S, \text{right})$?
- (b) (2 pts) What is the probability of taking the left action using the ε -greedy policy ($\varepsilon = 0.1$) with respect to Q^* ?
- (c) (14 pts) What problem will you encounter when training with Deep Q-learning (DQN)? In order to avoid the problem, what method can we apply?