a B

C

Considering a Singular Value Decomposition  $X = U S V^T$ , what are the special properties of matrices U, S and V?

Select one:

- $\bigcirc\,$  a. U, V are unitary and S is diagonal.
- Ob. U is orthogonal, V is upper triangular and S is symmetric.
- o. U, V are upper triangular, and S is diagonal.
- O d. U, V are symmetric and S is orthogonal.

## 选A

Consider these two probability distributions on the same space  $\Omega = \{A, B, C, D, E\}$ 

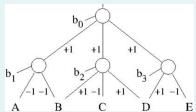
$$p = \langle \frac{1}{16}, \frac{1}{16}, \frac{1}{8}, \frac{1}{4}, \frac{1}{2} \rangle$$

$$q = \langle \frac{1}{4}, \frac{1}{16}, \frac{1}{16}, \frac{1}{2}, \frac{1}{8} \rangle$$

Compute (correct to at least two decimal places):

- \* The Entropy H(p):
- \* The KL-Divergence  $D_{KL}(p || q)$ :

Consider the following multi-layer perceptron, using the threshold activation function, and assume that TRUE is represented by 1; FALSE by 0.



For which values of the biases  $b_0$ ,  $b_1$ ,  $b_2$  and  $b_3$  would this network compute the logical function

(¬A ∨ ¬B) ∧ (B ∨ ¬C ∨ D) ∧ (D ∨ ¬E)

| * b <sub>0</sub> = | <b>+</b> |
|--------------------|----------|
| * b <sub>1</sub> = | <b>‡</b> |
| * b2=              | <b></b>  |
| * ho-              | _        |

-2.5, 1.5, 0.5, 0.5

Consider a Perceptron whose output is given by  $h(w_0+w_1x_1+w_2x_2)$ , where  $x_1$ ,  $x_2$  are inputs and h() is the Heaviside (step) function.

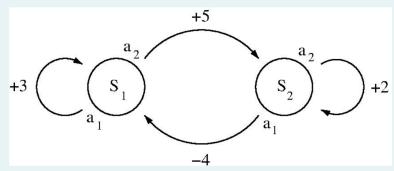
Assume this Perceptron is being trained on the data in the following table, and that the current values of the weights are  $w_0$ = -0.5,  $w_1$ = -1 and  $w_2$ = -2.

| Training Example | X1 | х2 | Class |
|------------------|----|----|-------|
| (a)              | -2 | 2  | Pos   |
| (b)              | 2  | 1  | Neg   |
| (c)              | -1 | -1 | Neg   |

If the Perceptron Learning Rule is applied to the current weights, using training item (a) and a learning rate of  $\eta$ =1.0, the new values for  $w_0$ ,  $w_1$  and  $w_2$  at the end of this training step will be:

| * w <sub>0:</sub>  | <b>\$</b> |
|--------------------|-----------|
| * w <sub>1</sub> : | <b>\$</b> |
| * w <sub>2</sub> : | <b>‡</b>  |

Consider an environment with two states  $S = \{S_1, S_2\}$  and two actions  $A = \{a_1, a_2\}$ , where the (deterministic) transitions  $\delta$  and reward R for each state and action are as follows:



Assuming a discount factor of y = 0.7, determine:

Again assuming y = 0.7, compute these values (correct to two decimal places):

- \* Q\*(S<sub>1</sub>,a<sub>1</sub>)
- \* Q\*(S<sub>1</sub>,a<sub>2</sub>)
- \* Q\*(S<sub>2</sub>,a<sub>1</sub>)
- \* Q\*(S<sub>2</sub>,a<sub>2</sub>)

If  $\gamma$  is allowed to vary between 0 and 1, for which range of values of  $\gamma$  is this policy optimal (correct to two decimal places)?

- \* Minimum value of γ:
- \* Maximum value of γ:

(1) a1 a2

(2)10 9.33 3 6.67

(3) 0.67 1

|             | For the Generative Adversarial Networks discussed in this course, the game between the Generator and Discriminator: |  |  |
|-------------|---|--|--|
|             | <ul><li>○ b.</li><li>○ c.</li></ul>   | can be either zero-sum or not, but the zero-sum version produces better images is never zero-sum can be either zero-sum or not, but the non-zero-sum version produces better images is always zero-sum         |  |
| Ε           |   |  |  |
|             | The Ac  | tor-Critic algorithm combines:   |  |
|             | <ul><li> b.</li><li> c.</li></ul>   | one: Q-Learning and Policy Gradients two different Q-Learners (one for action selection, the other for value estimation) adversarially trained Generator and Discriminator networks Q-Learning and TD-Learning |  |
| 选           | a   |  |  |
| F           |   |  |  |
|             | Which o   | Time le  |  |
|             | <ul><li>b.</li><li>c.</li></ul>   | Batch Normalization Conjugate Gradients Weight Initialization Rectified Linear Unit  |  |
| g<br>h<br>I |   |  |  |

Time left 1:58:15 If 0=FALSE and 1=TRUE, which of these networks (with threshold activations at both the hidden and output layer) correctly computes the XOR function of two inputs? Select one +0.5 O b. +0.5 +0.5 O d. Consider a neural network trained using softmax for a classification task with three classes 1, 2, 3. Suppose a particular input is presented, producing outputs z<sub>1</sub>= 1.3, z<sub>2</sub>= 2.4, z<sub>3</sub>= 3.1 Assuming the correct class for this input is Class 2, and that Prob(2) is the softmax probability of the network choosing Class 2, compute the following, to two decimal places: \* d(log Prob(2))/dz<sub>1</sub>= \* d(log Prob(2))/dz2= \* d(log Prob(2))/dz<sub>3</sub>= -0.10 0.70 -0.60 -0.10 0.70 -0.60 Consider a convolutional neural network which takes as input a 65-by-77 color image (i.e. with three channels R, G, B). The first convolutional layer has 18 filters that are 5-by-5, with stride 4 and no zero-padding. Compute the number of: \* weights per neuron in this layer (including bias): \* neurons in this layer: \* independent parameters in this layer:

| Consider a convolutional neural network which takes as input a 65-by-77 color image (i.e. with three channels R, G, B). The first convolutional layer has 18 filters that are 3-by-3, with stride 2 and no zero-padding. |
|--|
| Compute the number of:   |
| * weights per neuron in this layer (including bias): 28  |
| * neurons in this layer: 21888   |
| * connections into the neurons in this layer: 612864   |
| * independent parameters in this layer: 504  |

Consider a Hopfield Network with the following weight matrix W:

| 0 0 -1 0 0 |
| 0 0 0 0 -1 |
| -1 0 0 +1 0 |
| 0 0 +1 0 +1 |
| 0 0 +1 0 +1 |
| 0 -1 0 +1 0 |

| 0 -1 0 +1 0 +1 |
| 0 -1 0 +1 0 |

| 0 +1 0 +1 0 +1 0 |

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| 0 -1 0 +1 0 +1 0 |

| 0 -1 0 +1 0 +1 0 |

| 0 -1

## Unstable stable unstable

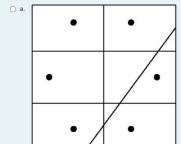
Consider a fully connected feedforward neural network with 6 inputs, 2 hidden units and 4 outputs, using tanh activation at the hidden units and sigmoid at the outputs. Suppose this network is trained on the following data, and that the training is successful.

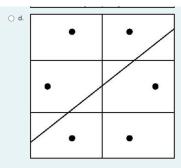
| Item | Inputs | Outputs |
|------|--------|---------|
|      | 123456 | 1234    |
| 1.   | 100000 | 0001    |
| 2.   | 010000 | 0011    |
| 3.   | 001000 | 0101    |
| 4.   | 000100 | 1000    |
| 5.   | 000010 | 1011    |
| 6.   | 000001 | 1100    |

Which of these diagrams correctly shows a point in hidden unit space corresponding to each input, and, for each output, a line dividing the hidden unit space into regions for which the value of that output is greater/less than one half?

Select one:

○ b.





J

I m n O

|        | aly 40% of the population have been vaccinated against a certain disease. Among those who are vaccinated, only 1% of them have the disease. But, among those who are not vaccinated, 2% of am have the disease. |  |  |
|--------|---|--|--|
|        | a random person is found to have the disease, what is the probability that they have been vaccinated?   |  |  |
| ()     | (You can give your answer either as a percentage, or as a number between 0 and 1)   |  |  |
| A      | swer:   |  |  |
|        | 0.405   |  |  |
|        | 0.125<br>0.25   |  |  |
| n      | 0.25  |  |  |
| p      |   |  |  |
| q<br>R |   |  |  |
|        |   |  |  |
|        | nforcement Learning is when an agent is: ect one:   |  |  |
|        | a. not presented with target outputs, but instead given a reward signal that it aims to maximize  |  |  |
|        | <ul> <li>b. presented once with examples of inputs and their target outputs</li> <li>c. presented multiple times (over time) with the same examples of inputs and their target outputs</li> </ul>               |  |  |
|        | d. only presented with the inputs and not target outputs, so it aims to find structure in these inputs  |  |  |
|        | 选A  |  |  |
|        | 近A  |  |  |
|        |   |  |  |
| s<br>T |   |  |  |
| •      | <u> </u>  |  |  |
|        | The principle "The most likely hypothesis is the simplest one consistent with the data." is called:   |  |  |
|        | The principle. The most interly hypothesis is the simplest one consistent with the data. Is called.   |  |  |
|        | Select one:   |  |  |
|        | ○ a. Ockham's Razor   |  |  |
|        | ○ b. Bayes' Rule  |  |  |
|        | ○ c. Perceptron Learning  |  |  |
|        | ○ d. Maximum Likelihood   |  |  |
|        |   |  |  |
|        |   |  |  |
|        |   |  |  |
|        | 选a  |  |  |
| Т      | 选a<br>ne Actor-Critic algorithm combines:   |  |  |
|        |   |  |  |

O b. Q-Learning and Policy Gradients

Od. Q-Learning and TD-Learning

O c. adversarially trained Generator and Discriminator networks

|             | Two common methods for unsupervised pre-training of neural networks are:  Select one:  a. Autoencoder and Deep Boltzmann Machine  b. Bayesian Inference and Weight Initialization  c. Weight Initialization and Autoencoder  d. Deep Boltzmann Machine and Bayesian Inference   |  |  |
|-------------|---|--|--|
|             | 选A  |  |  |
|             | The Conte   | ext Layer in a Simple Recurrent Network:   |  |
|             | Select one  | e:   |  |
|             | O a. is   | s comprised of the inputs in a sliding window around the current timestep  |  |
|             | Ob. is  | computed from the current input and the previous hidden layer  |  |
|             | O c. is   | a copy of the hidden layer from the previous timestep  |  |
|             | O d. is   | s computed from the current input and the previous output  |  |
| u<br>v<br>W | $\prime$  |  |  |
|             | Not yet answered Select Marked out of 1.00  F Flag question  O d  | n of these is NOT a method for dealing with the problem of vanishing or exploding gradients?  t one: Conjugate Gradients Rectified Linear Unit Batch Normalization Weight Initialization |  |
|             | When training on linear   | ly separable data using the Perceptron Learning Rule, what will happen if both the learning rate and the initial weights are scaled up by a large factor?                                |  |
|             | <ul> <li>a. The data will be learned successfully, in a smaller number of epochs</li> <li>b. The data will be learned successfully, in about the same number of epochs</li> <li>c. Learning may become unstable and fail to converge</li> <li>d. The data will be learned successfully, but in a larger number of epochs</li> </ul> |  |  |
|             | Which statement about word2vec is FALSE?  |  |  |
|             | Select one:  a. It aims to maximise the log probability of a word, based on the surrounding words  b. Representations for the same word at the input and output layers are different  c. The tanh activation function is used at the hidden nodes  d. Performance improves if frequent words are sampled less often                 |  |  |
|             | Which architecture would have the best chance of learning an Embedded Reber Grammar?  Select one:   |  |  |
|             | <ul><li>a. Gated Recurren</li><li>b. NetTalk system</li><li>c. Jordan Network</li><li>d. Elman Network</li></ul>  |  |  |

| When using Batch Normalization, in the Testing phase, the Mean and Variance of the activations at each node are typically:  Select one: |
|---|
| a. pre-computed from the training set      b. estimated using running averages  |
| c. either of the above  |
| d. none of the above  |
|   |
| 选C When comparing a Hopfield Network with a Boltzmann Machine, which statement is FALSE?  |
| Select one:   |
| ○ a. The updates are deterministic for one model, and stochastic for the other  |
| O b. The formula for the energy function is different for the two models  |
| c. The range of activations is {-1,1} for one model and {0,1} for the other   |
| ○ d. One model is used for retrieval, the other for generation  |
| Which of these statements about Dropout is FALSE:   |
| Select one:  (a. Dropout encourages redundancy)   |
| ) b. Dropout encourages the weight values to be small   |
| ○ c. Dropout helps to prevent overfitting   |
| Od. Dropout simulates an ensemble of network architectures  |
| 选B  |
| Whit type of Autoencoder explicitly forces the hidden features not to change much when the inputs are slightly altered?                 |
| Select one:   |
| ○ a. Variational Autoencoder  |
| O b. Contractive Autoencoder  |
| C. Denoising Autoencoder  |
| ○ d. Sparse Autoencoder   |
| 选B  |
| When using Batch Normalization, in the Testing phase, the Mean and Variance of the activations at each node are typically:              |
| Select one:   |
| a. pre-computed from the training set   |
|   |
| b. estimated using running averages   |
| o c. either of the above  |
| ○ d. none of the above  |
| Clear my choice   |
|   |
| Which architecture would have the best chance of learning an Embedded Reber Grammar?  |
| Select one:   |
| ○ a. Jordan Network   |
| ○ b. NetTalk system   |
|   |
| ○ c. Elman Network  |
| <ul><li>○ d. Gated Recurrent Unit</li></ul>   |
|   |

Which statement about word2vec is FALSE?

## Select one:

- a. The tanh activation function is used at the hidden nodes
- O b. Performance improves if frequent words are sampled less often
- oc. Representations for the same word at the input and output layers are different
- O d. It aims to maximise the log probability of a word, based on the surrounding words

When comparing a Hopfield Network with a Boltzmann Machine, which statement is FALSE?

## Select one:

- $\bigcirc$  a. The range of activations is  $\{-1,1\}$  for one model and  $\{0,1\}$  for the other
- One model is used for retrieval, the other for generation
- c. The formula for the energy function is different for the two models
- od. The updates are deterministic for one model, and stochastic for the other