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d22180@students.iitmandi.ac.in ✓

NPTEL (<https://swayam.gov.in/explorer?ncCode=NPTEL>) » Deep Learning for Computer Vision (course)

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## Course outline

How does an  
NPTEL online  
course work? ()

Week 0 ()

Week 1 ()

Week 2 ()

Week 3 ()

Week 4 ()

○ Neural Networks:  
A Review - Part 1  
(unit?unit=49&  
lesson=50)

○ Neural Networks:  
A Review - Part 2  
(unit?unit=49&  
lesson=51)

○ Feedforward  
Neural Networks  
and  
Backpropagation -  
Part 1

# Week 4: Assignment 2

Assignment not submitted

Due date: 2023-08-23, 23:59 IST.

1) Match the following:

1 point

- 1) Dense registration through optical flow
- 2) Wide baseline spatial matching
- 3) Rotation in x – y plane
- 4) Bag-of-words

- i) Assumption of rigid transformations
- ii) Detect different instances of same object
- iii) Two degrees of freedom
- iv) Aperture problem
- v) One degree of freedom

- ☐ 1→iv, 2→i, 3→iii, 4→ii  
☐ 1→v, 2→ii, 3→iv, 4→i  
☐ 1→iv, 2→i, 3→v, 4→ii  
☐ 1→v, 2→i, 3→iii, 4→ii

2) Which one of the following statements is **false**?

1 point

- ☐ The difference between VLAD and BoW is that VLAD yields a residual vector  
☐ BoW allows partial matching  
☐ Aggregated Selective Match Kernel uses concepts from Hamming Embedding method and VLAD  
☐ A codebook or a visual vocabulary consists of visual words

3) Let us consider RANSAC applied on 100 datapoints. Suppose you fit 10 lines  $line1$  –  $line10$  at **1 point** each iteration time step  $t1$  –  $t10$ . Number of outliers for each line is given by

[ $line1$  : 80,  $line2$  : 90,  $line3$  : 55,  $line4$  : 60,  $line5$  : 70,  $line6$  : 40,  $line7$  : 50,  $line8$  : 80,  $line9$  : 90,  $line10$  : 60]

While applying the RANSAC algorithm, which line will be stored as solution at time step  $t5$  and  $t10$ ?

- ☐  $t5$  :  $line5$ ,  $t10$  :  $line7$   
☐  $t5$  :  $line5$ ,  $t10$  :  $line10$   
☐  $t5$  :  $line3$ ,  $t10$  :  $line6$   
☐  $t5$  :  $line4$ ,  $t10$  :  $line6$

4) Given that a geometric transformation requires 2 correspondences between an input image and the corresponding transformed image, the degrees of freedom for that respective geometric transformation will be: **1 point**

- ☐ 3  
☐ 4  
☐ 5

(unit?unit=49&lesson=52)

☐ Feedforward Neural Networks and Backpropagation - Part 2

(unit?unit=49&lesson=53)

☐ Gradient Descent and Variants - Part 1 (unit?unit=49&lesson=54)

☐ Gradient Descent and Variants - Part 2 (unit?unit=49&lesson=55)

☐ Regularization in Neural Networks - Part 1 (unit?unit=49&lesson=56)

☐ Regularization in Neural Networks - Part 2 (unit?unit=49&lesson=57)

☐ Improving Training of Neural Networks - Part 1 (unit?unit=49&lesson=58)

☐ Improving Training of Neural Networks - Part 2 (unit?unit=49&lesson=59)

☐ Lecture Materials (unit?unit=49&lesson=60)

☐ Quiz: Week 4: Assignment 2 (assessment?name=226)

☐ Quiz: Week 4: Programming Quiz 1 (assessment?name=227)

☐ Week 4 Feedback Form : Deep Learning for Computer Vision (unit?unit=49&lesson=210)

- ☐ 3 or 4  
☐ 4 or 5

5) Consider 2 images ( $X$  and  $Y$ ) where 2 binarized descriptors from each image are assigned to the same visual word. Given the following descriptors for each image, compute the score using Hamming embedding: **1 point**

- First Binarized descriptor for image  $X(b_{x1}) : [1, 1, 0, 0, 1]$
- Second Binarized descriptor for image  $X(b_{x2}) : [0, 1, 1, 1, 1]$
- First Binarized descriptor for image  $Y(b_{y1}) : [0, 0, 0, 1, 1]$
- Second Binarized descriptor for image  $Y(b_{y2}) : [0, 1, 1, 1, 1]$

Consider the threshold  $\tau = 3$ .

- ☐ 1  
☐ 2  
☐ 3  
☐ 4

6) Consider two  $3 \times 3$  images  $x_1$  and  $x_2$  such that  $x_1 = \begin{bmatrix} 1 & 2 & 3 \\ 4 & 6 & 8 \\ 5 & 11 & 12 \end{bmatrix}$  and  $x_2 = \begin{bmatrix} 3 & 2 & 7 \\ 9 & 1 & 8 \\ 4 & 5 & 6 \end{bmatrix}$ . **1 point**

Their corresponding one-hot encoded label vectors are  $y_1 = [1, 0, 0]$  and  $y_2 = [0, 1, 0]$ . Perform mixup data augmentation between  $x_1$  and  $x_2$  given that  $\lambda = 0.6$ .

- ☐  $\tilde{x} = \begin{bmatrix} 1.8 & 2.0 & 4.6 \\ 6.0 & 4.0 & 8.0 \\ 4.6 & 8.6 & 9.6 \end{bmatrix}; \tilde{y} = [0, 0.6, 0.4]$
- ☐  $\tilde{x} = \begin{bmatrix} 1.8 & 2.0 & 4.6 \\ 6.0 & 3.0 & 8.0 \\ 4.6 & 8.4 & 10.8 \end{bmatrix}; \tilde{y} = [0.6, 0.4, 0]$
- ☐  $\tilde{x} = \begin{bmatrix} 1.8 & 2.0 & 4.6 \\ 6.0 & 4.0 & 8.0 \\ 4.6 & 8.6 & 9.6 \end{bmatrix}; \tilde{y} = [0.6, 0.4, 0]$
- ☐  $\tilde{x} = \begin{bmatrix} 5.2 & 4.2 & 3.2 \\ 5.6 & 9.0 & 10.4 \\ 1.4 & 5.4 & 6.0 \end{bmatrix}; \tilde{y} = [0, 0.4, 0.6]$

7) Which of the following statements are **true**?(select all that apply) **1 point**

- ☐ Batch Normalization increases the dependence on weight initialization.
- ☐ Any constant weight initialization will result in poor results.
- ☐ We cannot use higher learning rates when Batch Normalization is applied.
- ☐ Both very large and small weights can cause certain activation functions (sigmoid/tanh) to saturate, leading to small gradients.

8) Consider a simple perceptron  $f$  such that  $f : \mathbb{R}^4 \rightarrow \mathbb{R}$  which uses tanh as its activation function. **1 point**  
 The input  $X = [0.4, 0.3, 0.8, 0.2]$  and the corresponding weights  $W = [1.2, -0.8, 0.2, 0.5]$ . Compute the derivative of tanh of  $z$ , where  $z$  is a linear combination of weights vector  $W$  and input vector  $X$ . Ignore the bias term.

- ☐ 0.87  
☐ 0.82  
☐ 0.78  
☐ 0.35

9) Consider the following statements. **1 point**

(i) Adding Gaussian noise (with zero mean) to input is equivalent to L2 weight decay when loss function is MSE.

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(ii) At test time, batch normalization uses running average of training mini-batch mean and variance.

- ☐ statement (i) is true but statement (ii) is false.
- ☐ statement (ii) is true but statement (i) is false.
- ☐ Both statements are false.
- ☐ Both statements are true.

10) Which one of the following statements is **true**:

**1 point**

- ☐ Weight change criterion is a method of 'early stopping' that checks whether or not the error is dropping over epochs to decide whether to continue training or stop.
- ☐  $L_2$  norm tends to create more sparse weights than  $L_1$  norm.
- ☐ During the training phase, for each iteration, Dropout ignores a random fraction,  $p$ , of nodes, and accounts for it in the test phase by scaling down the activations by a factor of  $p$ .
- ☐ A single McCulloch-Pitts neuron is capable of modeling AND, OR, XOR, NOR, and NAND functions.

11) Using Adagrad-based gradient descent, find the new value of parameter  $\theta_{t+1}$ , given that the old value  $\theta_t = 0.9$ , aggregated gradient  $\Delta\theta_t = 0.8$ , gradient accumulation  $r_{t-1} = 0.5$ , learning rate  $\alpha = 0.2$  and small constant  $\delta = 10^{-8}$ . (Round decimal point till 3 places.)

**1 point**

12) Using RMSProp-based gradient descent, find the new value of parameter  $\theta_{t+1}$ , given that the old value  $\theta_t = 1.8$ , aggregated gradient  $\Delta\theta_t = 0.6$ , gradient accumulation  $r_{t-1} = 0.4$ , learning rate  $\alpha = 0.1$ , decay rate  $\rho = 0.6$  and small constant  $\delta = 10^{-9}$ . (Round decimal point till 3 places.)

**1 point**

You may submit any number of times before the due date. The final submission will be considered for grading.

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