

# Indian Institute of Technology Kharagpur

## Department of Electrical Engineering

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Subject No.: EE60020      Subject: Machine Learning for Signal Processing  
Date of Assignment: 15 April 2024      Semester: Spring 2023-24  
Assignment Number: 4 Solution      Duration: 1 hour 50 mins      Full points: 308

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Name: \_\_\_\_\_ Roll No: \_\_\_\_\_

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1. You are provided with a convolutional neural network defined as follows

$$\begin{aligned} \text{net}_E(\cdot) \mapsto & (1 : \text{Conv2D})16c5w1s0p \rightarrow (2 : \text{MaxPool2D})2w2s \\ & \rightarrow (3 : \text{Conv2D})64c3w1s1p \rightarrow (4 : \text{MaxPool2D})2w2s \end{aligned} \quad (1)$$

$$\begin{aligned} \text{net}_D(\cdot) \mapsto & (1 : \text{Conv2D})16c3w1s1p \rightarrow (2 : \text{MaxUnPool2D})2w2s \\ & \rightarrow (3 : \text{Conv2D})3c5w1s2p \rightarrow (4 : \text{MaxUnPool2D})2w2s \end{aligned} \quad (2)$$

$$\text{net}_C(\cdot) \mapsto (1 : \text{FC})160 \rightarrow (2 : \text{FC})84 \rightarrow (3 : \text{FC})10 \quad (3)$$

such that it operates as follows when provided with an input  $\mathbf{x} \in \mathbb{R}^{3 \times 32 \times 32}$

$$\mathbf{z} = \text{net}_E(\mathbf{x}) \quad (4)$$

$$\hat{\mathbf{x}} = \text{net}_D(\mathbf{z}) \quad (5)$$

$$\hat{\mathbf{y}} = \text{net}_C(\mathbf{z}) \quad (6)$$

- (a) (16 points) Find the following associated with feedforward operation in  $\text{net}_E(\cdot)$

	$a$ in input $\in \mathbb{R}^a$	$b$ in output $\in \mathbb{R}^b$	$c$ in weight $\in \mathbb{R}^c$	$d$ in bias $\in \mathbb{R}^d$
$\text{net}_E : 1$	$3 \times 32 \times 32$	$16 \times 28 \times 28$	$16 \times 3 \times 5 \times 5$	$16 \times 1$
$\text{net}_E : 2$	$16 \times 28 \times 28$	$16 \times 14 \times 14$	-	-
$\text{net}_E : 3$	$16 \times 14 \times 14$	$64 \times 14 \times 14$	$64 \times 16 \times 3 \times 3$	$64 \times 1$
$\text{net}_E : 4$	$64 \times 14 \times 14$	$64 \times 7 \times 7$	-	-

- (b) (16 points) Find the following associated with feedforward operation in  $\mathbf{net}_D(\cdot)$

	$a$ in input $\in \mathbb{R}^a$	$b$ in output $\in \mathbb{R}^b$	$c$ in weight $\in \mathbb{R}^c$	$d$ in bias $\in \mathbb{R}^d$
$\mathbf{net}_D : 1$	$64 \times 7 \times 7$	$16 \times 7 \times 7$	$16 \times 64 \times 3 \times 3$	$16 \times 1$
$\mathbf{net}_D : 2$	$16 \times 7 \times 7$	$16 \times 14 \times 14$	-	-
$\mathbf{net}_D : 3$	$16 \times 14 \times 14$	$3 \times 14 \times 14$	$3 \times 16 \times 5 \times 5$	$3 \times 1$
$\mathbf{net}_D : 4$	$3 \times 14 \times 14$	$3 \times 28 \times 28$	-	-

- (c) (12 points) Find the following associated with feedforward operation in  $\mathbf{net}_C(\cdot)$

	$a$ in input $\in \mathbb{R}^a$	$b$ in output $\in \mathbb{R}^b$	$c$ in weight $\in \mathbb{R}^c$	$d$ in bias $\in \mathbb{R}^d$
$\mathbf{net}_C : 1$	$3, 136 \times 1$	$160 \times 1$	$160 \times 3, 136$	$160 \times 1$
$\mathbf{net}_C : 2$	$160 \times 1$	$84 \times 1$	$84 \times 160$	$84 \times 1$
$\mathbf{net}_C : 3$	$84 \times 1$	$10 \times 1$	$10 \times 84$	$10 \times 1$

- (d) (16 points) Find the number of unitary arithmetic operations associated with feedforward operation in  $\mathbf{net}_E(\cdot)$

	#add	#mul	#logic	Total
$\mathbf{net}_E : 1$	940,800	940,800	0	1,881,600
$\mathbf{net}_E : 2$	0	0	9,408	9,408
$\mathbf{net}_E : 3$	1,806,336	1,806,336	0	3,612,672
$\mathbf{net}_E : 4$	0	0	9,408	9,408

- (e) (16 points) Find the number of unitary arithmetic operations associated with feedforward operation in  $\mathbf{net}_D(\cdot)$

	#add	#mul	#logic	Total
$\mathbf{net}_D : 1$	451,584	451,584	0	903,168
$\mathbf{net}_D : 2$	0	0	0	0
$\mathbf{net}_D : 3$	235,200	235,200	0	470,400
$\mathbf{net}_D : 4$	0	0	0	0

- (f) (12 points) Find the number of unitary arithmetic operations associated with feed-forward operation in  $\text{net}_c(\cdot)$

	#add	#mul	#logic	Total
$\text{net}_c : 1$	501,760	501,760	0	1,003,520
$\text{net}_c : 2$	13,440	13,440	0	26,880
$\text{net}_c : 3$	840	840	0	1,680

2. Consider that the neural networks in Q. 1 are trained with Mean Squared Error (MSE) as the loss function. Let  $J_1(\hat{\mathbf{x}}, \tilde{\mathbf{x}})$  and  $J_2(\hat{\mathbf{y}}, \mathbf{y})$  be evaluated with  $\tilde{\mathbf{x}} \in \mathbb{R}^{3 \times 28 \times 28}$  and  $\mathbf{y} \in \mathbb{R}^{10}$  representing the true states of the variables in the loss function. Let  $\nabla_1$  denote the derivative of  $J_1(\cdot)$  with respect to the output of a given layer, and  $\delta_1$  denote the derivative of  $J_1(\cdot)$  with respect to the input to that layer, such that for any layer represented mathematically as  $\mathbf{Q} = \mathbf{RS} + \mathbf{T}$ , where  $\mathbf{Q}$  is the output of the neural layer we have  $\nabla = \frac{\partial J(\cdot)}{\partial \mathbf{Q}}$ ,  $\mathbf{S}$  is the input to the neural layer we have  $\delta = \frac{\partial J(\cdot)}{\partial \mathbf{S}}$ ,  $\mathbf{R}$  denotes the weights and  $\mathbf{T}$  denotes the biases in the layer respectively. We have  $\frac{\partial J(\cdot)}{\partial \mathbf{R}} = \nabla \mathbf{S}^\top$ ,  $\frac{\partial J(\cdot)}{\partial \mathbf{T}} = \nabla$ ,  $\delta = \frac{\partial J(\cdot)}{\partial \mathbf{S}} = \mathbf{R}^\top \nabla$ . Similarly,  $\nabla_2$  and  $\delta_2$  correspond to these set of operations associated with  $J_2(\cdot)$ .

- (a) (16 points) Find the following associated with error backpropagation operation in  $\text{net}_E(\cdot)$  when  $\mathbf{w}$  and  $\mathbf{b}$  represent the weights and biases in a layer respectively.

	$a$ in $\frac{\partial J_1(\cdot)}{\partial \mathbf{w}} \in \mathbb{R}^a$	$b$ in $\frac{\partial J_1(\cdot)}{\partial \mathbf{b}} \in \mathbb{R}^b$	$c$ in $\frac{\partial J_2(\cdot)}{\partial \mathbf{w}} \in \mathbb{R}^c$	$d$ in $\frac{\partial J_2(\cdot)}{\partial \mathbf{b}} \in \mathbb{R}^d$
$\text{net}_E : 1$	$16 \times 3 \times 5 \times 5$	$16 \times 1$	$16 \times 3 \times 5 \times 5$	$16 \times 1$
$\text{net}_E : 2$	-	-	-	-
$\text{net}_E : 3$	$64 \times 16 \times 3 \times 3$	$64 \times 1$	$64 \times 16 \times 3 \times 3$	$64 \times 1$
$\text{net}_E : 4$	-	-	-	-

- (b) (16 points) Find the following associated with error backpropagation operation in  $\text{net}_D(\cdot)$  when  $\mathbf{w}$  and  $\mathbf{b}$  represent the weights and biases in a layer respectively.

	$a$ in $\frac{\partial J_1(\cdot)}{\partial \mathbf{w}} \in \mathbb{R}^a$	$b$ in $\frac{\partial J_1(\cdot)}{\partial \mathbf{b}} \in \mathbb{R}^b$	$c$ in $\frac{\partial J_2(\cdot)}{\partial \mathbf{w}} \in \mathbb{R}^c$	$d$ in $\frac{\partial J_2(\cdot)}{\partial \mathbf{b}} \in \mathbb{R}^d$
$\text{net}_D : 1$	$16 \times 64 \times 3 \times 3$	$16 \times 1$	-	-
$\text{net}_D : 2$	-	-	-	-
$\text{net}_D : 3$	$3 \times 16 \times 5 \times 5$	$3 \times 1$	-	-
$\text{net}_D : 4$	-	-	-	-

- (c) (12 points) Find the following associated with error backpropagation operation in  $\text{net}_C(\cdot)$  when  $\mathbf{w}$  and  $\mathbf{b}$  represent the weights and biases in a layer respectively.

	$a$ in $\frac{\partial J_1(\cdot)}{\partial \mathbf{w}} \in \mathbb{R}^a$	$b$ in $\frac{\partial J_1(\cdot)}{\partial \mathbf{b}} \in \mathbb{R}^b$	$c$ in $\frac{\partial J_2(\cdot)}{\partial \mathbf{w}} \in \mathbb{R}^c$	$d$ in $\frac{\partial J_2(\cdot)}{\partial \mathbf{b}} \in \mathbb{R}^d$
$\text{net}_C : 1$	-	-	$160 \times 3,136$	$160 \times 1$
$\text{net}_C : 2$	-	-	$84 \times 160$	$84 \times 1$
$\text{net}_C : 3$	-	-	$10 \times 84$	$10 \times 1$

- (d) (64 points) Find the number of unitary arithmetic operations associated with error backpropagation operation in  $\text{net}_E(\cdot)$

		#add	#mul	#logic	Total
$\text{net}_E : 1$	$\frac{\partial J_1(\cdot)}{\partial \mathbf{w}}$	939,600	940,800	0	1,880,400
	$\frac{\partial J_1(\cdot)}{\partial \mathbf{b}}$	12,528	16	0	12,544
	$\frac{\partial J_2(\cdot)}{\partial \mathbf{w}}$	939,600	940,800	0	1,880,400
	$\frac{\partial J_2(\cdot)}{\partial \mathbf{b}}$	12,528	16	0	12,544
$\text{net}_E : 2$	$\frac{\partial J_1(\cdot)}{\partial \mathbf{w}}$	0	0	0	0
	$\frac{\partial J_1(\cdot)}{\partial \mathbf{b}}$	0	0	0	0
	$\frac{\partial J_2(\cdot)}{\partial \mathbf{w}}$	0	0	0	0
	$\frac{\partial J_2(\cdot)}{\partial \mathbf{b}}$	0	0	0	0
$\text{net}_E : 3$	$\frac{\partial J_1(\cdot)}{\partial \mathbf{w}}$	1,797,120	1,806,336	0	3,603,456
	$\frac{\partial J_1(\cdot)}{\partial \mathbf{b}}$	12,480	64	0	12,544
	$\frac{\partial J_2(\cdot)}{\partial \mathbf{w}}$	1,797,120	1,806,336	0	3,603,456
	$\frac{\partial J_2(\cdot)}{\partial \mathbf{b}}$	12,480	64	0	12,544
$\text{net}_E : 4$	$\frac{\partial J_1(\cdot)}{\partial \mathbf{w}}$	0	0	0	0
	$\frac{\partial J_1(\cdot)}{\partial \mathbf{b}}$	0	0	0	0
	$\frac{\partial J_2(\cdot)}{\partial \mathbf{w}}$	0	0	0	0
	$\frac{\partial J_2(\cdot)}{\partial \mathbf{b}}$	0	0	0	0

- (e) (64 points) Find the number of unitary arithmetic operations associated with error backpropagation operation in  $\text{net}_D(\cdot)$

		#add	#mul	#logic	Total
$\text{net}_D : 1$	$\frac{\partial J_1(\cdot)}{\partial \mathbf{w}}$	435,456	451,584	0	887,040
	$\frac{\partial J_1(\cdot)}{\partial \mathbf{b}}$	768	16	0	784
	$\frac{\partial J_2(\cdot)}{\partial \mathbf{w}}$	0	0	0	0
	$\frac{\partial J_2(\cdot)}{\partial \mathbf{b}}$	0	0	0	0
$\text{net}_D : 2$	$\frac{\partial J_1(\cdot)}{\partial \mathbf{w}}$	0	0	0	0
	$\frac{\partial J_1(\cdot)}{\partial \mathbf{b}}$	0	0	0	0
	$\frac{\partial J_2(\cdot)}{\partial \mathbf{w}}$	0	0	0	0
	$\frac{\partial J_2(\cdot)}{\partial \mathbf{b}}$	0	0	0	0
$\text{net}_D : 3$	$\frac{\partial J_1(\cdot)}{\partial \mathbf{w}}$	23,400	235,200	0	469,200
	$\frac{\partial J_1(\cdot)}{\partial \mathbf{b}}$	585	3	0	588
	$\frac{\partial J_2(\cdot)}{\partial \mathbf{w}}$	0	0	0	0
	$\frac{\partial J_2(\cdot)}{\partial \mathbf{b}}$	0	0	0	0
$\text{net}_D : 4$	$\frac{\partial J_1(\cdot)}{\partial \mathbf{w}}$	0	0	0	0
	$\frac{\partial J_1(\cdot)}{\partial \mathbf{b}}$	0	0	0	0
	$\frac{\partial J_2(\cdot)}{\partial \mathbf{w}}$	0	0	0	0
	$\frac{\partial J_2(\cdot)}{\partial \mathbf{b}}$	0	0	0	0

- (f) (48 points) Find the number of unitary arithmetic operations associated with error backpropagation operation in  $\text{net}_c(\cdot)$

		#add	#mul	#logic	Total
$\text{net}_c : 1$	$\frac{\partial J_1(\cdot)}{\partial \mathbf{w}}$	0	0	0	0
	$\frac{\partial J_1(\cdot)}{\partial \mathbf{b}}$	0	0	0	0
	$\frac{\partial J_2(\cdot)}{\partial \mathbf{w}}$	501,760	501,760	0	1,003,520
	$\frac{\partial J_2(\cdot)}{\partial \mathbf{b}}$	160	160	0	320
$\text{net}_c : 2$	$\frac{\partial J_1(\cdot)}{\partial \mathbf{w}}$	0	0	0	0
	$\frac{\partial J_1(\cdot)}{\partial \mathbf{b}}$	0	0	0	0
	$\frac{\partial J_2(\cdot)}{\partial \mathbf{w}}$	13,440	13,440	0	26,880
	$\frac{\partial J_2(\cdot)}{\partial \mathbf{b}}$	13,440	13,440	0	26,880
$\text{net}_c : 3$	$\frac{\partial J_1(\cdot)}{\partial \mathbf{w}}$	0	0	0	0
	$\frac{\partial J_1(\cdot)}{\partial \mathbf{b}}$	0	0	0	0
	$\frac{\partial J_2(\cdot)}{\partial \mathbf{w}}$	840	840	0	1,680
	$\frac{\partial J_2(\cdot)}{\partial \mathbf{b}}$	840	840	0	1,680

————— End of questions. —————