

QUES No 1

INPUT LAYER:-

Input Shape: $(16, 16, 4)$

Activation size: $16 \times 16 \times 4 = 1024$

Trainable Parameters = 0

Convolution Layer 1:-

no Filters = 10

filter size = 5

stride = 1

valid padding

width, height = $(16 - 5 + 2 \times 0) / 1 + 1 = 12$

Activation shape: $(12, 12, 10)$

Activation size = $12 \times 12 \times 10 = 1440$

Trainable Parameters = $(5 \times 5 \times 4 + 1) \times 10$
 $= 1010$

Pooling Layer 1:-

filter size: 2

stride: 2

width, height = $(12 - 2 + 2 \times 0) / 2 + 1 = 6$

Activation shape: $6 \times 6 \times 10$

Activation size = 360

Trainable Parameters = 0

Convolution Layer 2

no Filters = 16

Filter size = 3

Stride = 1

Activation shape = (6, 6, 16)

Activation size = 576

→ Depth wise : $(3 \times 3 \times 1) + 1$

→ Point wise : $1 \times 1 \times 16 \times 16 + 16$

Trainable Parameters = $(3 \times 3 \times 1 + 1) \times 16 + (1 \times 1 \times 16 \times 16 + 16)$
= 4544

Pooling Layer 2:-

Filter size = 2

Stride = 2

width, height : $(6 - 2 + 2 \times 0) / 2 + 1 = 3$

Activation shape = (3, 3, 16)

Activation size = $3 \times 3 \times 16 = 144$

Fully Connected Layer 3:-

no Neuron = 100

Activation shape = (100, 1)

Activation size = 100

Trainable Parameters = $(3 \times 3 \times 16 \times 100) + 100$
= 43300

Fully Connected Layer 4:

no of neurons = 10

Activation shape = (10, 1)

Activation size = 10

$$\begin{aligned}\text{Trainable Parameters} &= (10 \times 100) + 10 \\ &= 1010\end{aligned}$$

Output Layer:

no of neurons = 5

Activation shape = (5, 1)

Activation size = 5

$$\begin{aligned}\text{Trainable Parameters} &= (10 \times 5) + 5 \\ &= 55\end{aligned}$$

QUES No 2

Layer	Activation Shape	Activation Size	Trainable Para
Input	(8, 8, 3)	192	0
Conv1	(8, 8, 2)	1812	19
Pool1	(3, 3, 2)	18	0
Conv2	(3, 3, 1)	9	10
Pool2	(1, 1, 1)	1	0
FC3	(2, 1)	2	4
Output	(1, 1)	1	2

Cost / Error

$$J(\theta) = -[1 \times \log(0.5) + (1-1) \times \log(1-0.5)]$$

$$= 0.693$$

Backpropagate:-

$$\Delta J(w_{out}) = (a - r) \times a_{in}$$

$$a_{in} = f(\sum(w_{jk} \times a_{conv}))$$

$$a_{in} = \max(0, \sum(w_{jk} \times a_{conv}))$$

$$\Delta J(w_{out}) = (0.5 - 1) \times \max(0, \sum(w_{jk} \times a_{conv}))$$

Convolution Layer 2

③

Fully Connected:

$$\Delta J(w_{fc}) = (w_{out}^T \times \Delta J(w_{out})) \times a_{conv}$$

Convolution Layers:

$$\Delta J(w_{conv}) = \sum (\Delta J(w_{fc}) \times w_{fc}^T \times f'(a_{conv}) \times a_{prev})$$

$$\Delta J(w_{conv}) = \sum (\Delta J(w_{fc}) \times w_{fc}^T) \times 1(a_{conv} > 0) \times a_{prev}$$

$$w = w - \alpha \times \Delta J(w)$$

for

$$w_{out} = 0.5$$

$$w_{fc} = [[0.1, 0.2], [0.3, 0.4]]$$

$$a_{conv} = 0.6, 0.7$$

$$\alpha = 0.1$$

$$\Delta J(w_{out}) = -0.25$$

$$\Delta J(w_{fc}) = [-0.15, -0.175]$$

$$\Delta J(w_{conv}) = [0.33, 0.397]$$

$$w_{out} = [0.5] - 0.1 \times -0.25 = 0.525$$

$$w_{fc} = [0.1, 0.2], [0.3, 0.4] \times [-0.15, -0.175]$$
$$= [[0.115, 0.225], [0.35, 0.425]]$$

$$w_{conv} = [0.8, 0.9] - 0.1 \times [0.33, 0.397]$$
$$= [0.47, 0.51]$$

The table for activation shape, size is the same.

$$a_{out} = \sigma(\sum(w_{out} \times a_{in}) + b)$$

$$= \sigma(\sum(w_{out} \times \sigma(\sum(w_{in} \times \sigma(\sum(w_{conv} \times a_{prev})))$$

$$a_{out} = \sigma(0.525 \times \sigma(0.115 a_{prev1} + 0.225 a_{prev2} + 0.375 a_{prev3} + 0.25 a_{prev4}))$$

The output will be closer to 1 than 0.5.

5) Calculate cost/error value using the cost function

$$\text{for } a_{out} = 0.9$$

$$\begin{aligned} J(d) &= -\log(a_{out}) \\ &= -\log(0.9) \\ &= 0.105 \end{aligned}$$

6) Compare Error

Error has reduced from 0.693 to 0.105.

7) What is predicted output.

0.9. Predicted output is 0.9.