

Given

i/p image size $i \times i = 224 \times 224$

No. of i/p channels $C = 3$

o/p size: $0 \times 0 = 55 \times 55$

No. of o/p feature maps (filters): $N = 96$.

Kernel size: $k \times k = 11 \times 11$.

Multiplications in a conv. layer $= 0 \times 0 \times N \times C \times k \times k$.

per o/p feature map

For each of the $N = 96$ o/p feat. maps, the conv. op. involves sliding the kernel over the i/p volume (which has $C = 3$ channels).

For each of posⁿ in the o/p (55×55), the kernel (11×11) is applied to each of 3 i/p channels, resulting in $11 \times 11 \times 3$ multiplications per o/p pixel.

Total multiplications:

Multiply the no. of o/p pixels (55×55) by the no. of feat maps (96) and the multiplications per pixel ($11 \times 11 \times 3$):

$$\underline{55 \times 55 \times 96 \times 3 \times 11 \times 11}$$

Result:

$$\text{No. of multiplications} = 105,019,200.$$

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Q2

Step 1: Compute Mean(μ) of the mini batch

$$\mu = (2.1 + 4.1 + 0.2 + 0.5 + 3.4) / 5 = 10.3 / 5 = \underline{2.06}$$

Step 2: Compute the Variance (σ^2) using the mean μ)

$$\begin{aligned}\sigma^2 &= \frac{(2.1 - 2.06)^2 + (4.1 - 2.06)^2 + (0.2 - 2.06)^2 + (0.5 - 2.06)^2 + (3.4 - 2.06)^2}{5} \\ &= (0.0016 + 4.1616 + 3.4596 + 2.4336 + 1.7956) / 5 \\ &= (11.852) / 5 \\ &= 2.3704\end{aligned}$$

Step 3: Normalize each value

Batch Norm formula: $\hat{x}_i = \frac{x_i - \mu}{\sqrt{\sigma^2 + \epsilon}}$

(Assuming $\epsilon = 10^{-5}$ for numerical stability, but here $\sqrt{2.3704} \approx 1.54$)

$$x_i = \frac{x_i - 2.06}{1.54}$$

$$\hat{x}_1 \approx 0.03$$

$$\hat{x}_2 \approx 1.32$$

$$\hat{x}_3 \approx -1.21$$

$$\hat{x}_4 \approx -1.01$$

$$\hat{x}_5 \approx 0.87$$

Apply scale(γ) and shift(β) $y_i = \gamma x_i + \beta$

given $\gamma = 1.0$ and $\beta = 0.9$

$$y_1 = \cancel{1.0} \times 0.03 + \cancel{0.9} = 0.03 + 0.9 = 1.0$$

$$y_2 = \cancel{1.0} \times 1.32 + \cancel{0.9} = 1.32 + 0.9 = 2.22$$

$$y_3 = \cancel{1.0} \times (-1.21) + \cancel{0.9} = -1.21 + 0.9 = -0.31$$

$$y_4 = \cancel{1.0} \times (-1.01) + \cancel{0.9} = -1.01 + 0.9 = -0.11$$

$$y_5 = \cancel{1.0} \times 0.87 + \cancel{0.9} = 0.87 + 0.9 = 1.77$$

Final O/p (to 1 Dec. place)

$$y = [0.9, 2.2, -0.3, -0.1, 1.8]$$

$$y = [1.0, 2.2, -0.1, 0.1, 1.8]$$

$$y = [1.0, 2.2, -0.1, 0.1, 1.8]$$

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Confusion Matrix:

$$\begin{bmatrix} 21 & 10 & 3 \\ 2 & 31 & 5 \\ 4 & 3 & 28 \end{bmatrix}$$

Step 1 (Totals)

$$\text{Total Samples} = 21 + 10 + 3 + 2 + 31 + 5 + 4 + 3 + 28 = 107$$

Step 2 (Per class metrics)

Class A.

$$TP = 21 \text{ (correctly predicted A)}$$

$$FP = 2 + 4 = 6. \text{ (others predicted as A)}$$

$$FN = 10 + 3 = 13 \text{ (A predicted as others)}$$

$$TN = \text{Total} - TP - FP - FN = 107 - 21 - 6 - 13 = 67$$

$$\text{Accuracy: } \frac{TP + TN}{\text{Total}} = \frac{21 + 67}{107} = \frac{88}{107} \approx 0.82.$$

$$\text{Precision: } \frac{TP}{TP + FP} = \frac{21}{21 + 6} = \frac{21}{27} \approx 0.78.$$

$$\text{Recall: } \frac{TP}{TP + FN} = \frac{21}{21 + 13} = \frac{21}{34} \approx 0.62.$$

Class B.

$$TP = 31$$

$$FP = 10 + 3 = 13.$$

$$FN = 2 + 5 = 7$$

$$TN = 107 - 31 - 13 - 7 = 56.$$

Accuracy:

$$\frac{31 + 56}{107} = \frac{87}{107} \approx 0.81$$

Precision:

$$\frac{31}{31 + 13} = \frac{31}{44} \approx 0.70.$$

$$\text{Recall: } \frac{31}{31 + 7} = \frac{31}{38} \approx 0.82$$

Class C

$$TP = 28$$

$$FP = 3 + 5 = 8$$

$$FN = 4 + 3 = 7$$

$$TN = 107 - 28 - 8 - 7 = 64$$

Accuracy:

$$\frac{28+64}{107} = \frac{92}{107} \approx 0.86.$$

Precision:

$$\frac{28}{28+8} = \frac{28}{36} \approx 0.78$$

Recall:

$$\frac{28}{28+7} = \frac{28}{35} \approx 0.80.$$

Final 3×3 matrix (rounded to 2 decimal places).

$$\begin{bmatrix} 0.82 & 0.78 & 0.62 \\ 0.81 & 0.70 & 0.82 \\ 0.86 & 0.78 & 0.80 \end{bmatrix}$$

Each row:

[Accuracy, Precision, Recall] for A, B, C respectively

Given:

Output probabilities (softmax output's):

$$P_c = [0.19, 0.24, 0.23, 0.17, 0.17]$$

Ground truth class labels

$$Y_c = 3 \text{ [classes are indexed from 0 to 4].}$$

Step 1 One-hot encoding of ground truth.

For $Y_c = 3$, the one-hot encoded vector $\mathbf{y} = [0, 0, 0, 1, 0]$

Step 2 Cross Entropy loss (L)

$$L = - \sum_{c=0}^4 Y_c \log(P_c)$$

Since only $Y_3 = 1$, this simplifies to:

$$L = -\log(P_3)$$

Step 3 Plug in the probability

From P_c , the probability of the true class ($c=3$) is $P_3 = 0.17$.

$$L = -\log(0.17)$$

Step 4 Compute the logarithm

Using natural logarithm (base e):

$$L = -\log(0.17) \approx -(-1.77196) = \underline{\underline{1.772}}$$