# **Assignment 3 – GAN Theory (Part 3)**

# Q1

Given: I=8, K=4, S=2, P=1, OP=1.

Formula:  $O = (I-1)\cdot S - 2P + K + OP$ 

Computation:  $O = (8-1)\cdot 2 - 2\cdot 1 + 4 + 1 = 7\cdot 2 - 2 + 4 + 1 = 14 - 2 + 4 + 1 = 17$ 

Answer: 17×17

# Q2

Change stride from S=2 to S=3.

$$O = (8-1)\cdot 3 - 2\cdot 1 + 4 + 1 = 7\cdot 3 - 2 + 4 + 1 = 21 - 2 + 4 + 1 = 24$$

Answer: Output becomes 24×24 (it increases roughly proportionally to the stride).

#### Q3

General formula (no dilation):  $O = (I-1)\cdot S - 2P + K + OP$ 

If dilation D is used:  $O = (I-1)\cdot S - 2P + D\cdot (K-1) + OP + 1$ 

When D=1, it reduces to the first formula since  $D \cdot (K-1)+1=K$ .

#### **Q4**

Upsample  $16\times16 \rightarrow 32\times32$  with no padding (P=0) and OP=0.

We want  $(16-1)\cdot S + K = 32$ . Pick  $S=2 \rightarrow (15)\cdot 2 + K = 30 + K = 32 \Rightarrow K=2$ .

One valid configuration: K=2, S=2, P=0, OP=0.

# **Q5**

Batch of values: [6, 8, 10, 6]. BatchNorm (without  $\gamma$ ,  $\beta$ ) standardizes to zero mean, unit variance.

Mean:  $\mu = (6 + 8 + 10 + 6)/4 = 7.5$ 

Variance:  $\sigma^2 = ((6-7.5)^2 + (8-7.5)^2 + (10-7.5)^2 + (6-7.5)^2)/4 = 2.75$ 

Standard deviation:  $\sigma = \sqrt{2.75} \approx 1.6583$ 

Normalized values: [-0.9045, 0.3015, 1.5076, -0.9045]

# Q6

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ReLU(x) = max(0, x) = 
{ 0, if x < 0 ; x, if x \ge 0 }
LeakyReLU(x; \alpha) = max(\alphax, x) = 
{ \alphax, if x < 0 ; x, if x \ge 0 }
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Key difference: ReLU has zero slope for x<0; LeakyReLU has a small positive slope  $\alpha$ .

# Q7

Why prefer LeakyReLU over ReLU in deep networks (and often in GANs)?

- Avoids 'dying ReLUs' ReLU can output 0 for many negatives, leading to zero gradients.
- Maintains gradient flow on negative inputs for better optimization stability.
- Empirically stabilizes GAN discriminator training and reduces mode collapse.