

VIR 2019

Name: \_\_\_\_\_

Test 2-B

Time Limit:

Points: \_\_\_\_\_

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1. In all questions, assume stride denotes length of convolutional stride, pad denotes symmetric zero-padding, rate is dilatation rate of atrous convolution. conv stands for convolution or atrous convolution layer and max is max-pooling layer.

You are given input feature map (image)  $\mathbf{x}$  and kernel  $\mathbf{w}$ :

$$\mathbf{x} = \begin{array}{|c|c|c|} \hline 2 & -1 & 0 \\ \hline 1 & 0 & -1 \\ \hline 1 & 0 & 2 \\ \hline \end{array} \quad \mathbf{w} = \begin{array}{|c|c|} \hline 1 & 1 \\ \hline 0 & -2 \\ \hline \end{array}$$

Compute outputs of the following layers:

- $\text{conv}(\mathbf{x}, \mathbf{w}, \text{stride} = 1, \text{pad} = 0, \text{rate} = 2) =$

- $\text{conv}(\mathbf{x}, \mathbf{w}, \text{stride} = 3, \text{pad} = 1, \text{rate} = 1) =$

- $\text{max}(\mathbf{x}, 2 \times 2) =$

2. You are given network (without loss layer) which consists of the convolutional layer and the max-pooling layer. The structure is defined as follows:

$$f(\mathbf{x}, \mathbf{w}) = \max \left( \text{conv}(\mathbf{x}, \mathbf{w}), 1 \times 2 \right)$$

- Draw computational graph and compute the feed-forward pass for input feature map (image)  $\mathbf{x}$  and convolutional kernel  $\mathbf{w}$ :

$$\mathbf{x} = \begin{bmatrix} 2 & 1 & 2 \end{bmatrix} \quad \mathbf{w} = \begin{bmatrix} 1 & 0 \end{bmatrix}$$

- Estimate gradient wrt kernel  $\mathbf{w}$  (i.e. compute local gradients for the max-pooling layer and the convolutional layer and substitute edge-values computed in the feed-forward pass).

$$\frac{\partial f(\mathbf{x}, \mathbf{w})}{\partial \mathbf{w}} =$$



- Approximately how many operations does a feed forward pass take? Assume every addition and multiplication counts as a single operation.

4. You are given function  $\mathbf{y} = f(\mathbf{x}, \mathbf{w})$ , which maps N-dimensional input  $\mathbf{x}$  to N-dimensional output  $\mathbf{y}$  (preserves dimensionality) and it is parameterized using weight vector  $\mathbf{w}$ .

- Draw computational graph of two layer neural network, where each layer consists of the function  $f(\mathbf{x}, \mathbf{w})$  with additional the skip-connection.

- Derive gradient of the output from the second layer with respect to weights in the first layer.