

$$\frac{\partial \text{loss}}{\partial \text{kernel}_1} = \text{Convolution}(\text{Input } X, \text{Loss gradient } \frac{\partial \text{loss}}{\partial \text{result}_1})$$

$$\frac{\partial \text{loss}}{\partial X} = \text{Full Convolution}(\text{rotated Filter kernel}, \text{Loss gradient } \frac{\partial \text{loss}}{\partial \text{result}_1})$$

$$\frac{\partial \text{loss}}{\partial \text{kernel}_1} = \text{Con}(\text{result}_1, 2(\text{result}_2 - \text{reference})) \cdot \frac{\partial \text{loss}}{\partial \text{result}_2}$$

$$= \begin{bmatrix} 0 & 1 & 1 \\ -1 & 0 & 1 \\ -1 & -1 & 0 \end{bmatrix} \cdot 2 \cdot \begin{bmatrix} 0 & 2 \\ -2 & 0 \end{bmatrix} = 2 \cdot \begin{bmatrix} 4 & 2 \\ 2 & 4 \end{bmatrix} = \begin{bmatrix} 8 & 4 \\ 4 & 8 \end{bmatrix}$$

$$\frac{\partial \text{loss}}{\partial \text{kernel}_2} = \text{Con}(\text{Input } X, \frac{\partial \text{loss}}{\partial \text{result}_1})$$

$$\frac{\partial \text{loss}}{\partial \text{result}_1} = \text{Full Con}(\text{rotated kernel}_2, \frac{\partial \text{loss}}{\partial \text{result}_2})$$

$$\frac{\partial \text{loss}}{\partial \text{result}_2} = 2(\text{result}_2 - \text{reference})$$

$$\therefore \frac{\partial \text{loss}}{\partial \text{result}_1} = 2 \begin{bmatrix} 0 & 0 & 0 & 0 \\ 0 & 0 & 2 & 0 \\ 0 & -2 & 0 & 0 \\ 0 & 0 & 0 & 0 \end{bmatrix} \cdot \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix}_{180^\circ} = 2 \cdot \begin{bmatrix} 0 & 2 & 0 \\ -2 & 0 & 2 \\ 0 & -2 & 0 \end{bmatrix} = \begin{bmatrix} 0 & 4 & 0 \\ -4 & 0 & 4 \\ 0 & -4 & 0 \end{bmatrix}$$

$$\therefore \frac{\partial \text{loss}}{\partial \text{kernel}_2} = \begin{bmatrix} 1 & 0 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 & 0 \\ 0 & 0 & 1 & 0 & 0 \\ 0 & 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 0 & 1 \end{bmatrix} \cdot \begin{bmatrix} 0 & 4 & 0 \\ -4 & 0 & 4 \\ 0 & -4 & 0 \end{bmatrix} = \begin{bmatrix} 0 & -8 & 0 \\ 8 & 0 & -8 \\ 0 & 8 & 0 \end{bmatrix}$$