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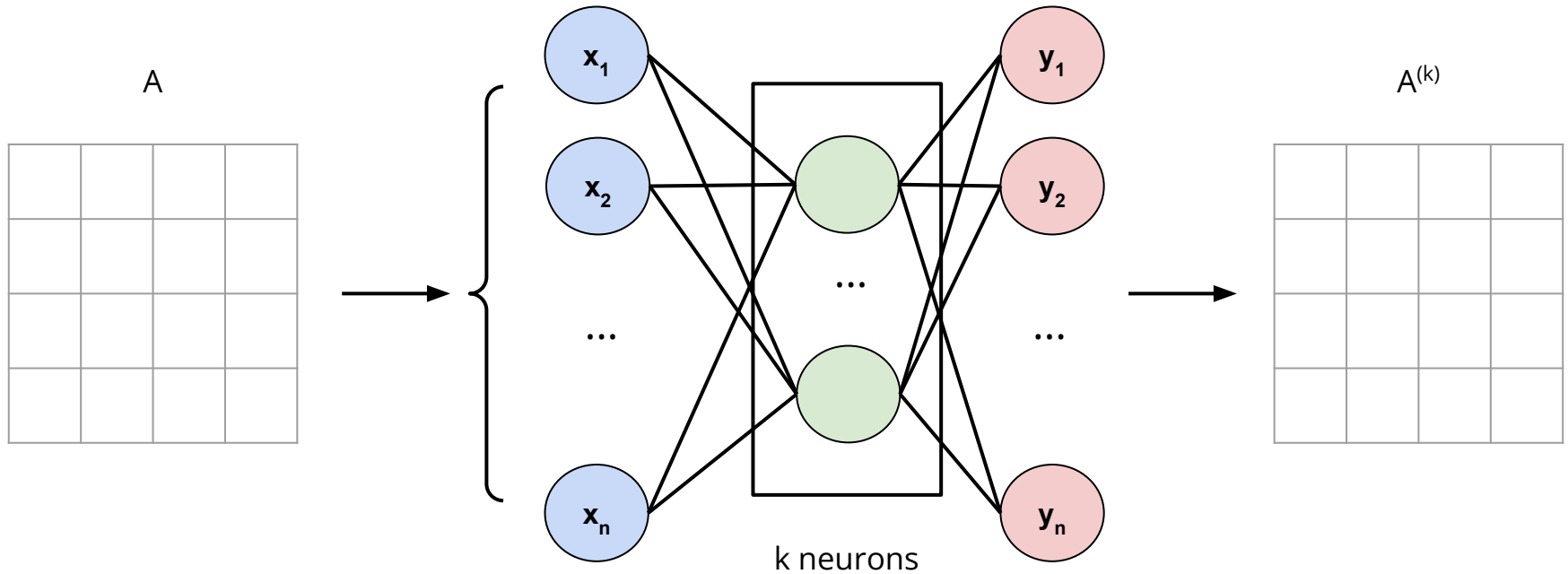
# Advanced Neural Networks

— Boston University CS 506 - Lance Galletti —

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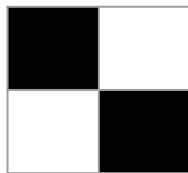
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# Neural Networks - Auto Encoders



# Logistic Regression Revisited

Given a 2 x 2 grid where each cell  $a_{ij}$  can take on one of two colors  $c_1$  and  $c_2$ , find a function that can identify the following diagonal pattern:



$= c_1 = -1$



$= c_2 = 1$

# Logistic Regression Revisited

That is, find  $\mathbf{f}$  such that

$$\mathbf{f} \left( \begin{array}{|c|c|} \hline a_{00} & a_{01} \\ \hline a_{10} & a_{11} \\ \hline \end{array} \right) = \begin{cases} \checkmark & \text{if} \\ \times & \text{otherwise} \end{cases}$$
$$\begin{array}{|c|c|} \hline a_{00} & a_{01} \\ \hline a_{10} & a_{11} \\ \hline \end{array} = \begin{array}{|c|c|} \hline c_1 & c_2 \\ \hline c_2 & c_1 \\ \hline \end{array} = \begin{array}{|c|c|} \hline \blacksquare & \square \\ \hline \square & \blacksquare \\ \hline \end{array}$$

We can define:  $\checkmark = 1$  and  $\times = 0$


# Logistic Regression Revisited

We can assign weights to each cell

$w_1$	$w_2$
$w_3$	$w_4$

# Logistic Regression Revisited

$w_1$	$w_2$
$w_3$	$w_4$




$a_{00}$	$a_{01}$
$a_{10}$	$a_{11}$

 $= w_1 a_{00}$

# Logistic Regression Revisited

$w_1$	$w_2$
$w_3$	$w_4$




$a_{00}$	$a_{01}$
$a_{10}$	$a_{11}$

$$= w_1 a_{00} + w_2 a_{01}$$

# Logistic Regression Revisited

$w_1$	$w_2$
$w_3$	$w_4$




$a_{00}$	$a_{01}$
$a_{10}$	$a_{11}$

$$= w_1 a_{00} + w_2 a_{01} + w_3 a_{10}$$



# Logistic Regression Revisited

$w_1$	$w_2$
$w_3$	$w_4$



$a_{00}$	$a_{01}$
$a_{10}$	$a_{11}$

$$= w_1 a_{00} + w_2 a_{01} + w_3 a_{10} + w_4 a_{11}$$

# Logistic Regression Revisited

We can assign weights to each cell  
such that:

$w_1$	$w_2$
$w_3$	$w_4$

$$\underbrace{w_1 a_{00} + w_2 a_{01} + w_3 a_{10} + w_4 a_{11}} = b \quad \text{if diagonal pattern found}$$

$w_1$	$w_2$
$w_3$	$w_4$



$a_{00}$	$a_{01}$
$a_{10}$	$a_{11}$

# Logistic Regression Revisited

For example:

$$\begin{array}{|c|c|} \hline w_1 & w_2 \\ \hline w_3 & w_4 \\ \hline \end{array} = \begin{array}{|c|c|} \hline -2 & 2 \\ \hline 2 & -2 \\ \hline \end{array}$$

What value  $b$  do we get when applied to the diagonal pattern?

$$\begin{array}{|c|c|} \hline -2 & 2 \\ \hline 2 & -2 \\ \hline \end{array} \otimes \begin{array}{|c|c|} \hline \text{black} & \text{white} \\ \hline \text{white} & \text{black} \\ \hline \end{array} = ?$$

# Logistic Regression Revisited

Any other pattern will have a value lower:

$$\begin{array}{|c|c|} \hline -2 & 2 \\ \hline 2 & -2 \\ \hline \end{array} \otimes \begin{array}{|c|c|} \hline -1 & -1 \\ \hline 1 & -1 \\ \hline \end{array} = ?$$

# Logistic Regression Revisited

Equivalently we can decide to move the value  $b$  to the left of the equation in order for the weighted sum to reveal a diagonal pattern at 0:

$$w_1 a_{00} + w_2 a_{01} + w_3 a_{10} + w_4 a_{11} + b = 0 \quad \text{if diagonal pattern found}$$

# Logistic Regression Revisited

We could then find a function  $\sigma$  to apply to the result of this sum in order to get probabilities of being diagonal:

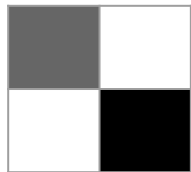
$$\sigma(w_1 a_{100} + w_2 a_{201} + w_3 a_{310} + w_4 a_{411} + b) > \frac{1}{2} \text{ if } w_1 a_{100} + w_2 a_{201} + w_3 a_{310} + w_4 a_{411} + b > 0$$

# Logistic Regression Revisited

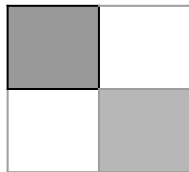
$$\sigma(x) = \frac{1}{1 + e^{-x}}$$

When  $\sigma$  is the logit<sup>-1</sup> (also called sigmoid) function, this is Logistic Regression.

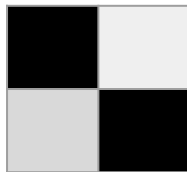
So for each cell we're looking to learn a weight  $w_i$  that makes  $\sigma$  larger for diagonal patterns. The bias term  $b$  lets us account for systemic dimming or brightening of cells (i.e. when the data is not normalized).



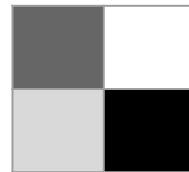
.92



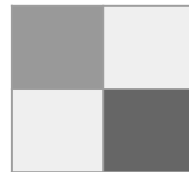
.81



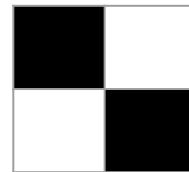
.95



.73

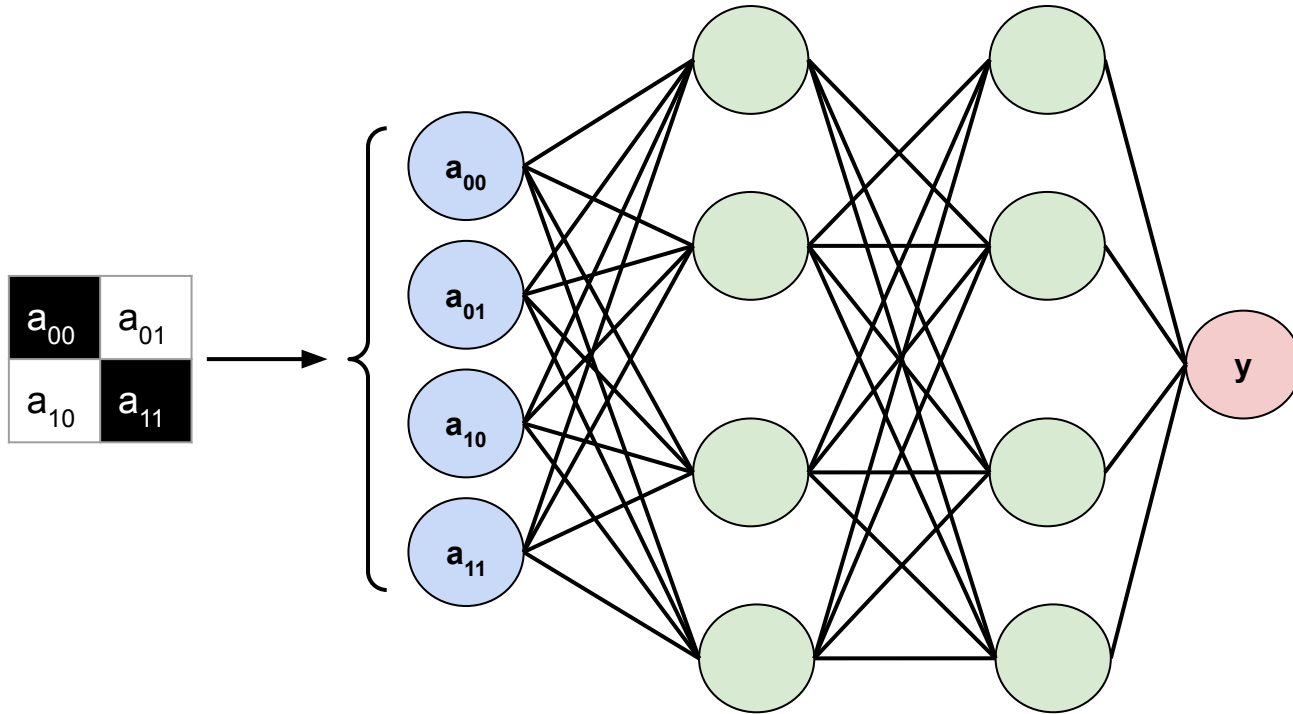


.68



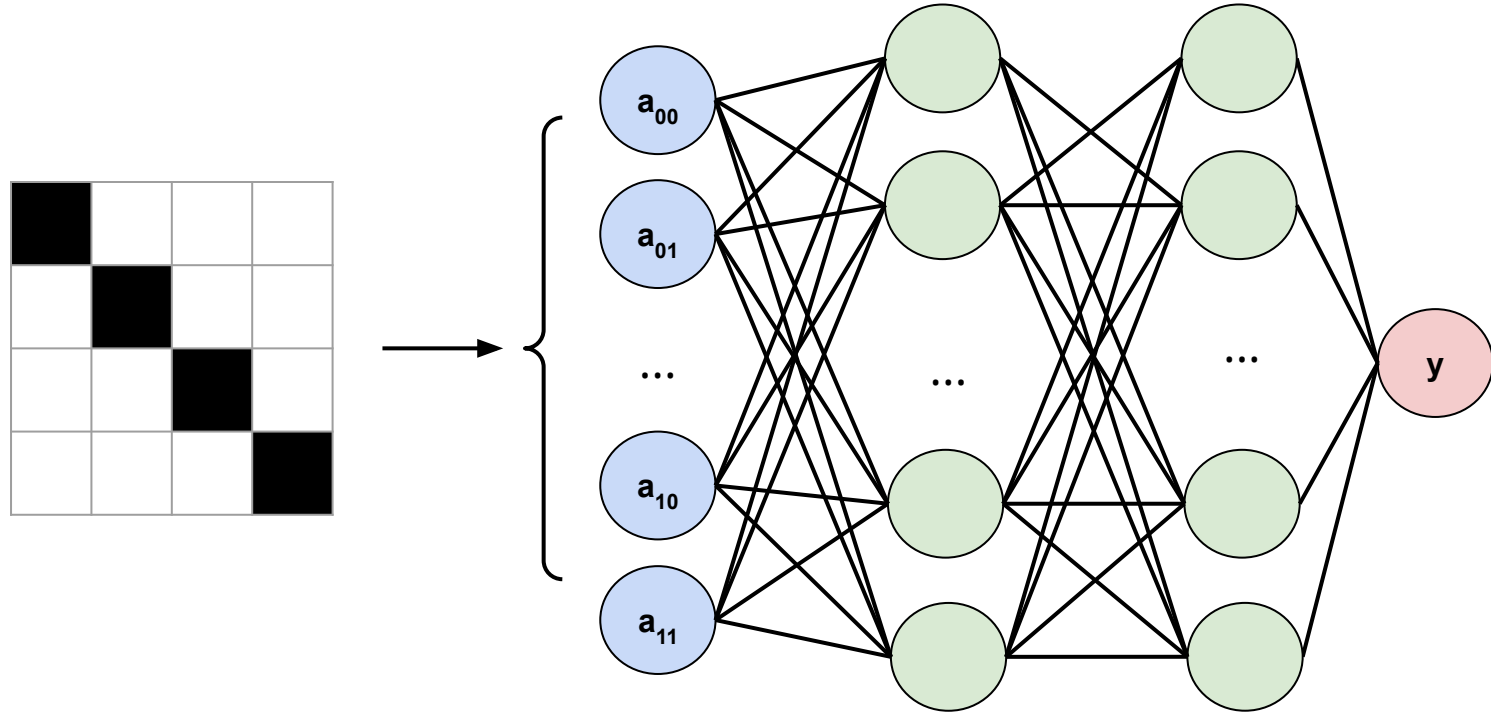
.99

# Neural Networks - Convolutional Neural Networks

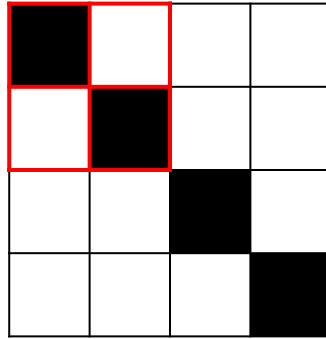




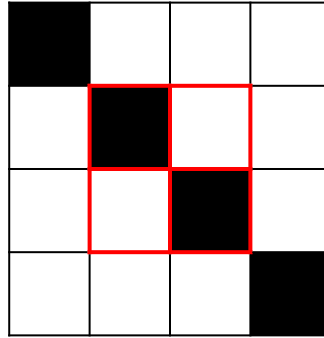
# Neural Networks - Convolutional Neural Networks



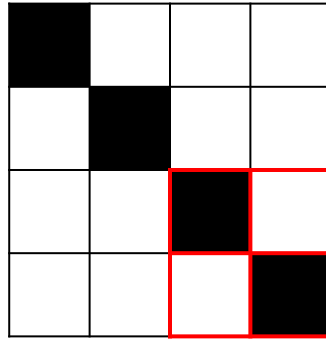
# Neural Networks - Convolutional Neural Networks



# Neural Networks - Convolutional Neural Networks



# Neural Networks - Convolutional Neural Networks




# Neural Networks - Convolutional Neural Networks

Recall: Our network learns weights for each cell

$w_1$	$w_2$
$w_3$	$w_4$

# Neural Networks - Convolutional Neural Networks

$w_1$	$w_2$
$w_3$	$w_4$



$a_{00}$	$a_{01}$
$a_{10}$	$a_{11}$

 $= w_1 a_{00}$

# Neural Networks - Convolutional Neural Networks

The diagram illustrates a 2D convolution operation. On the left, a 2x2 kernel matrix is shown with weights  $w_1$ ,  $w_2$ ,  $w_3$ , and  $w_4$ . The element  $w_2$  is highlighted in orange. This kernel is multiplied (indicated by a circle with an 'X') by a 2x2 input matrix with values  $a_{00}$ ,  $a_{01}$ ,  $a_{10}$ , and  $a_{11}$ . The element  $a_{01}$  is highlighted in green. The result of the operation is the equation  $w_1 a_{00} + w_2 a_{01}$ .

$$\begin{bmatrix} w_1 & w_2 \\ w_3 & w_4 \end{bmatrix} \otimes \begin{bmatrix} a_{00} & a_{01} \\ a_{10} & a_{11} \end{bmatrix} = w_1 a_{00} + w_2 a_{01}$$

# Neural Networks - Convolutional Neural Networks

$w_1$	$w_2$
$w_3$	$w_4$



$a_{00}$	$a_{01}$
$a_{10}$	$a_{11}$

$$= w_1 a_{00} + w_2 a_{01} + w_3 a_{10}$$



# Neural Networks - Convolutional Neural Networks

$w_1$	$w_2$
$w_3$	$w_4$

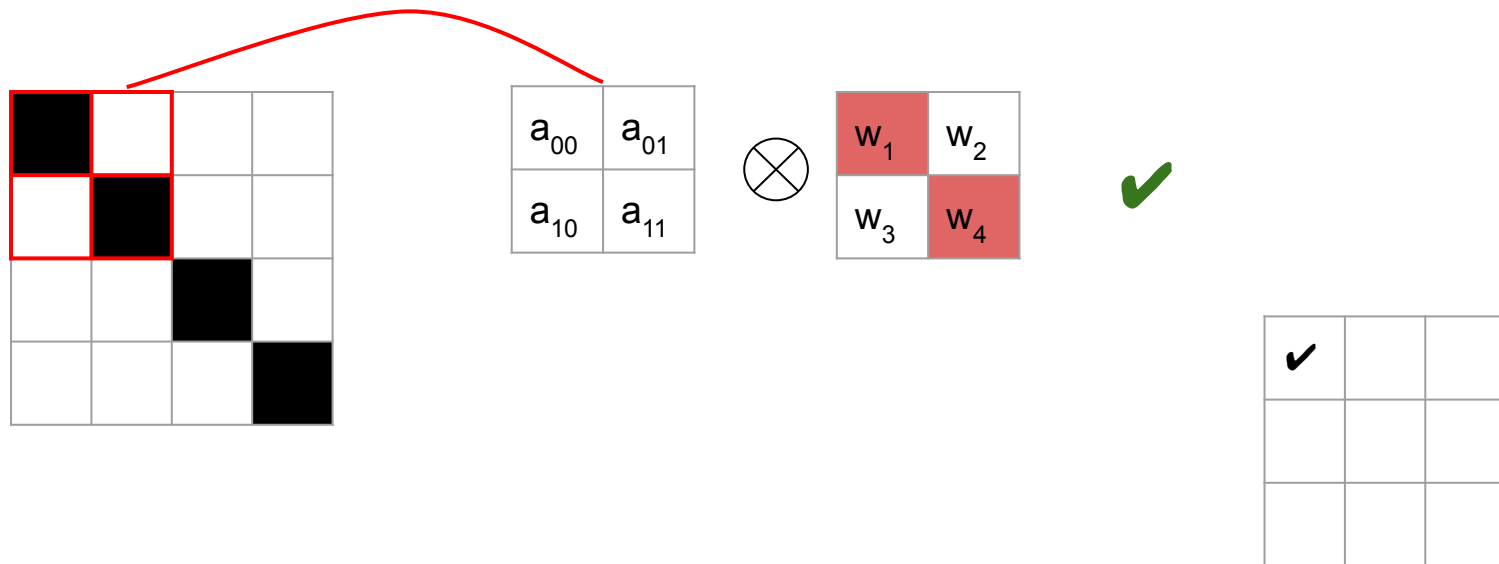


$a_{00}$	$a_{01}$
$a_{10}$	$a_{11}$

$$= w_1 a_{00} + w_2 a_{01} + w_3 a_{10} + w_4 a_{11}$$

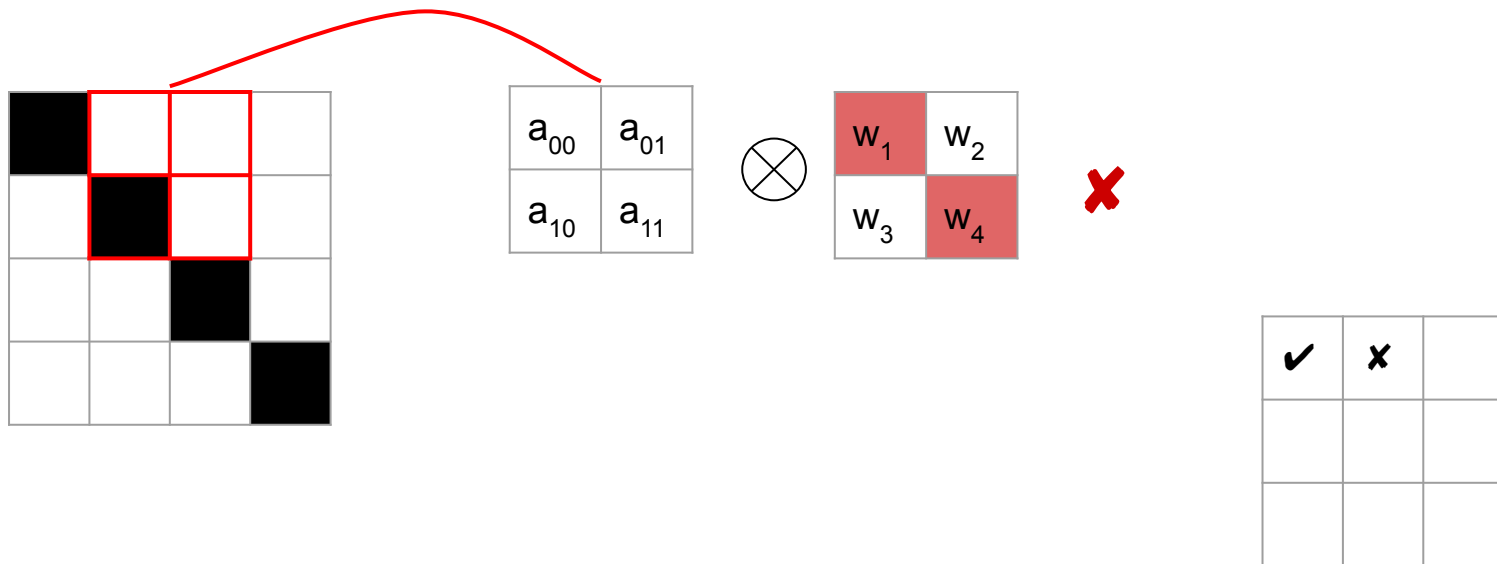
# Neural Networks - Convolutional Neural Networks

Knowing this, what happens if we slide this filter across the larger diagonal?



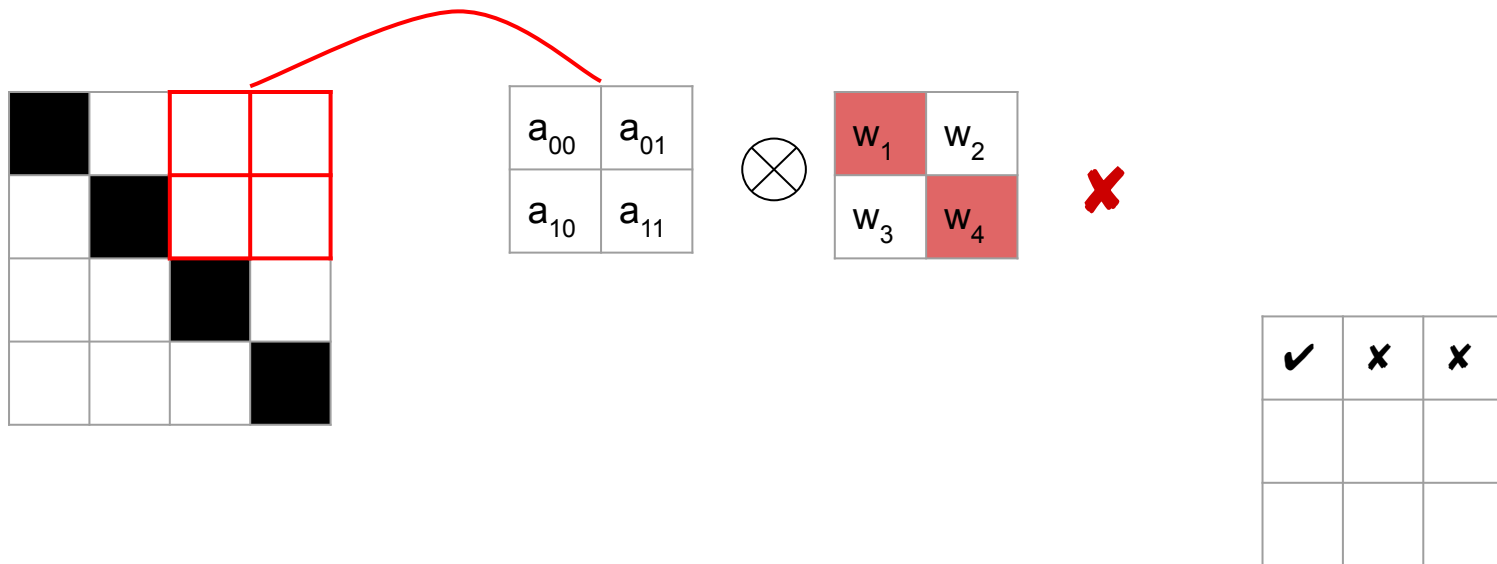
# Neural Networks - Convolutional Neural Networks

Knowing this, what happens if we slide this filter across the larger diagonal?



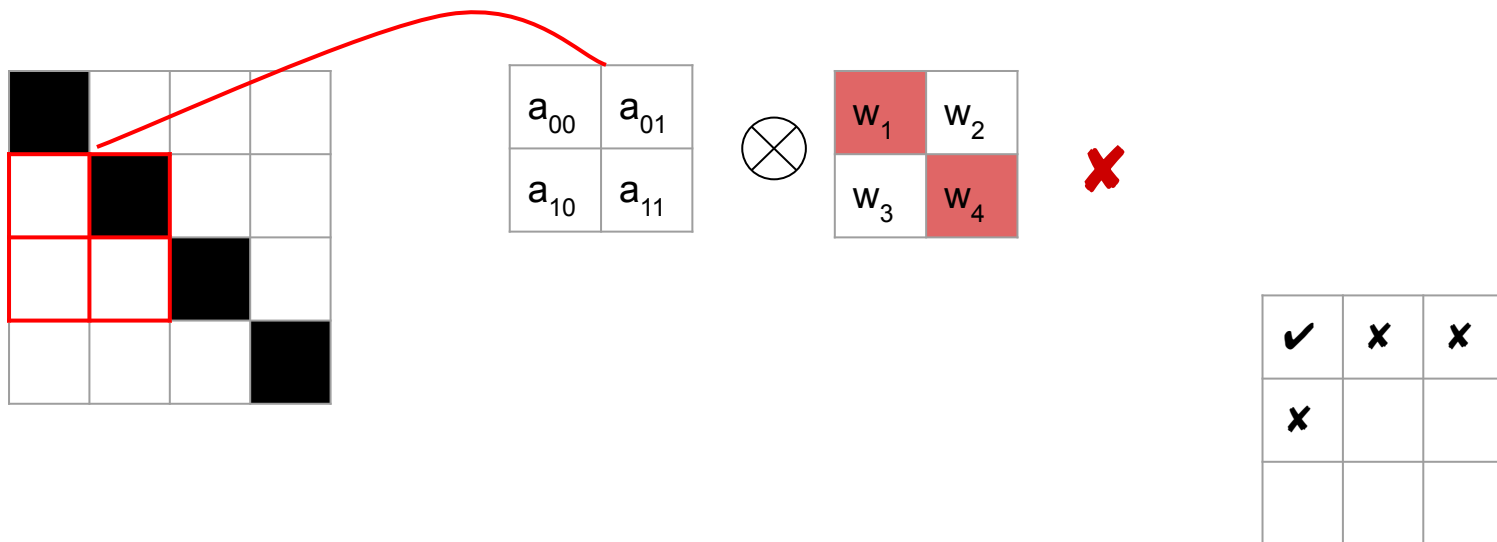
# Neural Networks - Convolutional Neural Networks

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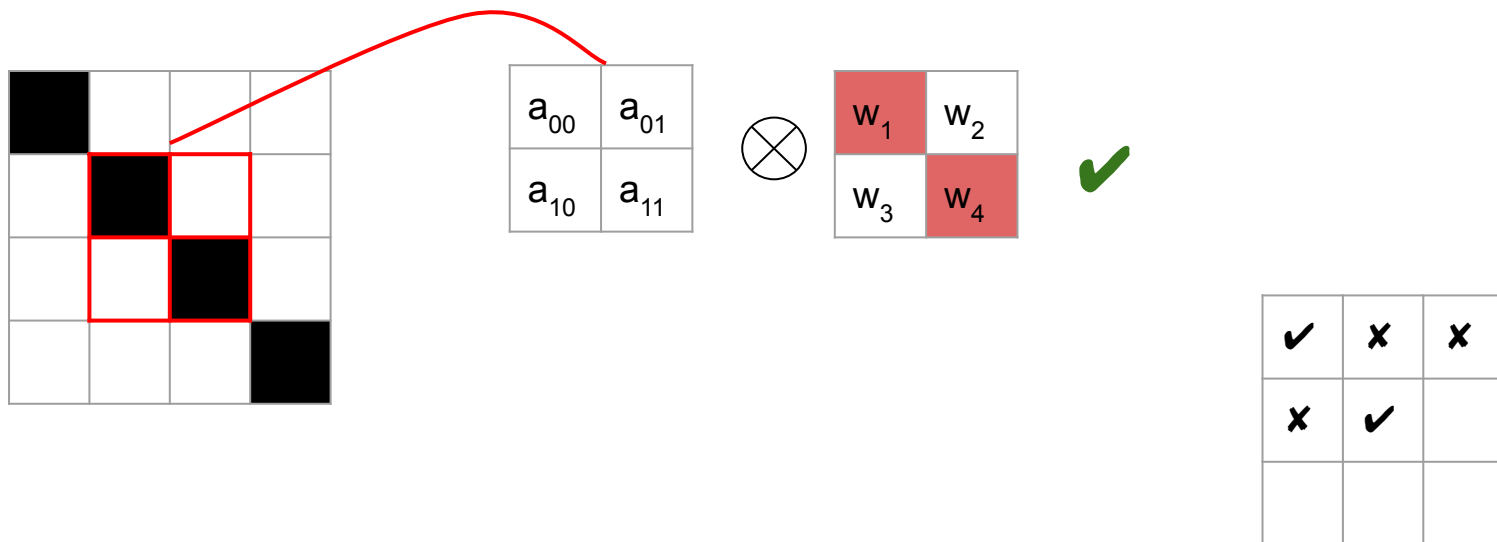
# Neural Networks - Convolutional Neural Networks

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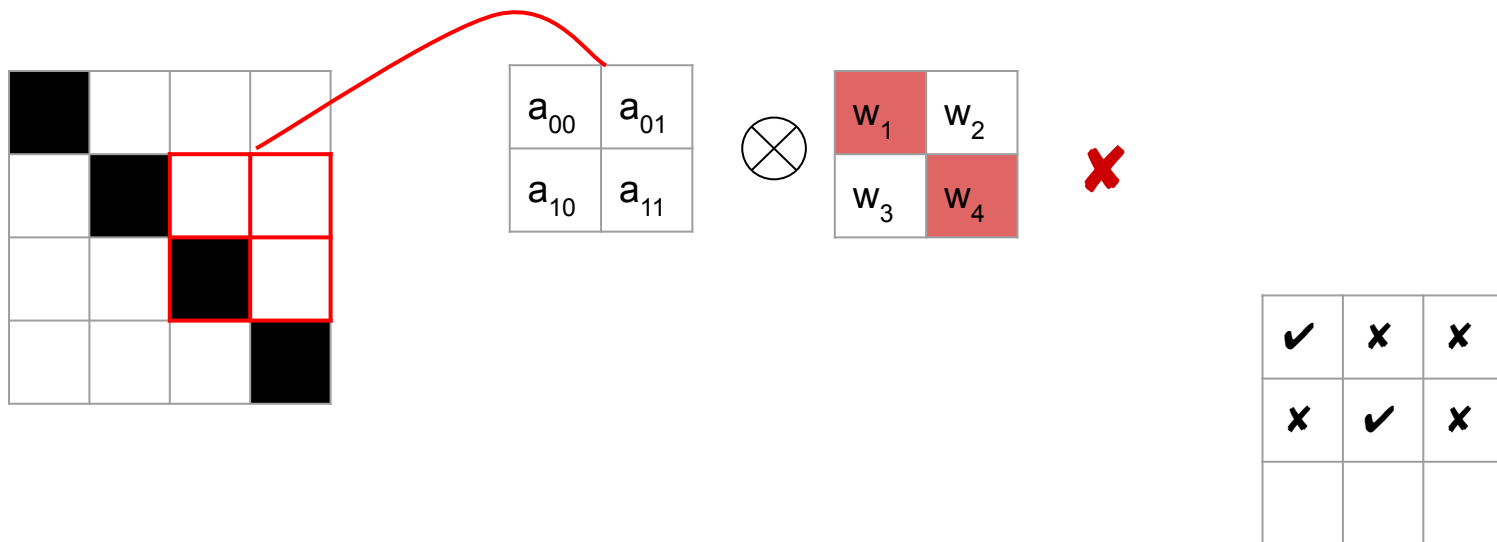
# Neural Networks - Convolutional Neural Networks

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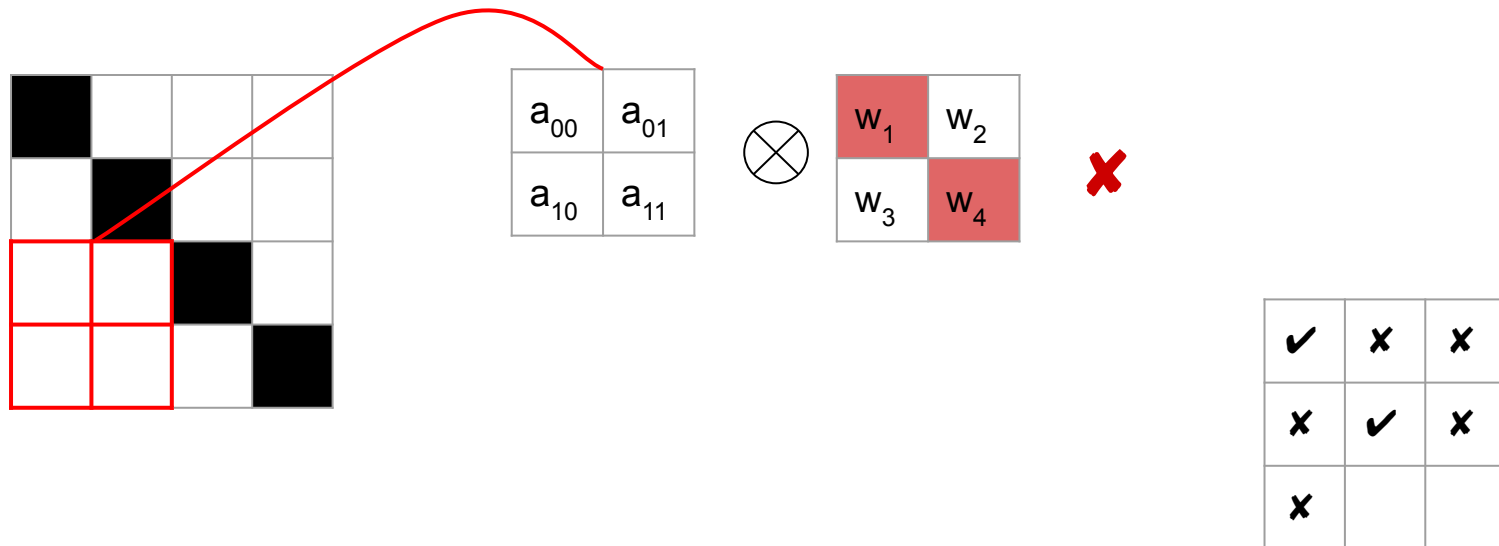
# Neural Networks - Convolutional Neural Networks

Knowing this, what happens if we slide this filter across the larger diagonal?



# Neural Networks - Convolutional Neural Networks

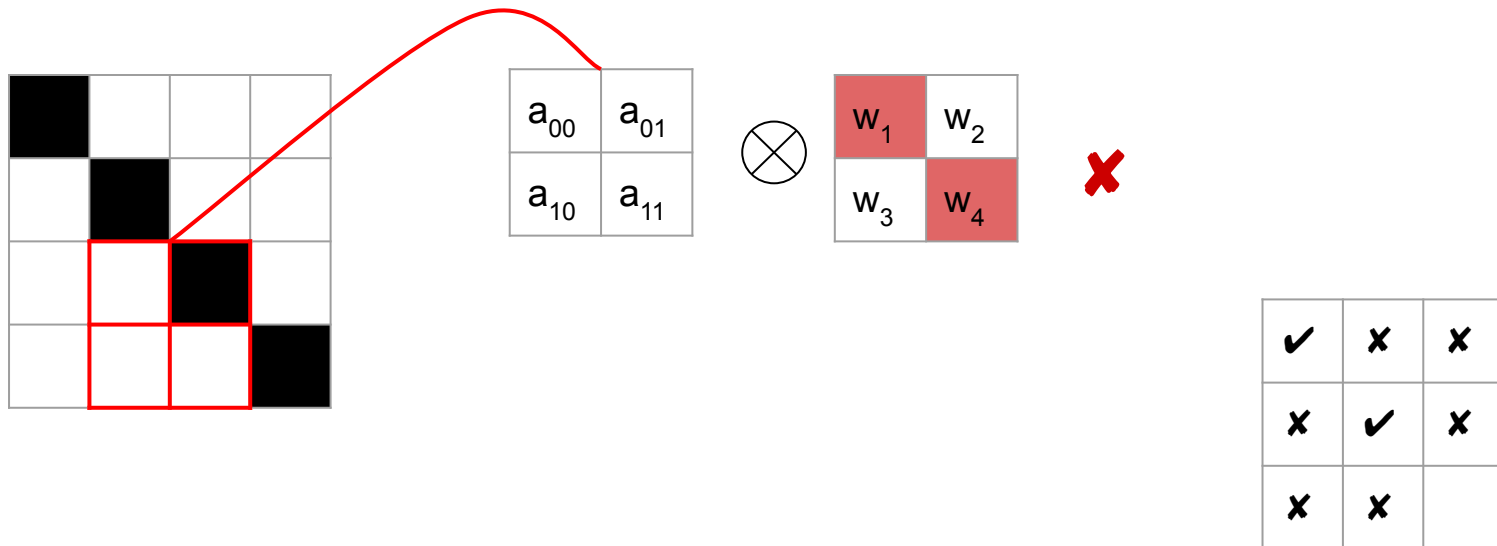
Knowing this, what happens if we slide this filter across the larger diagonal?





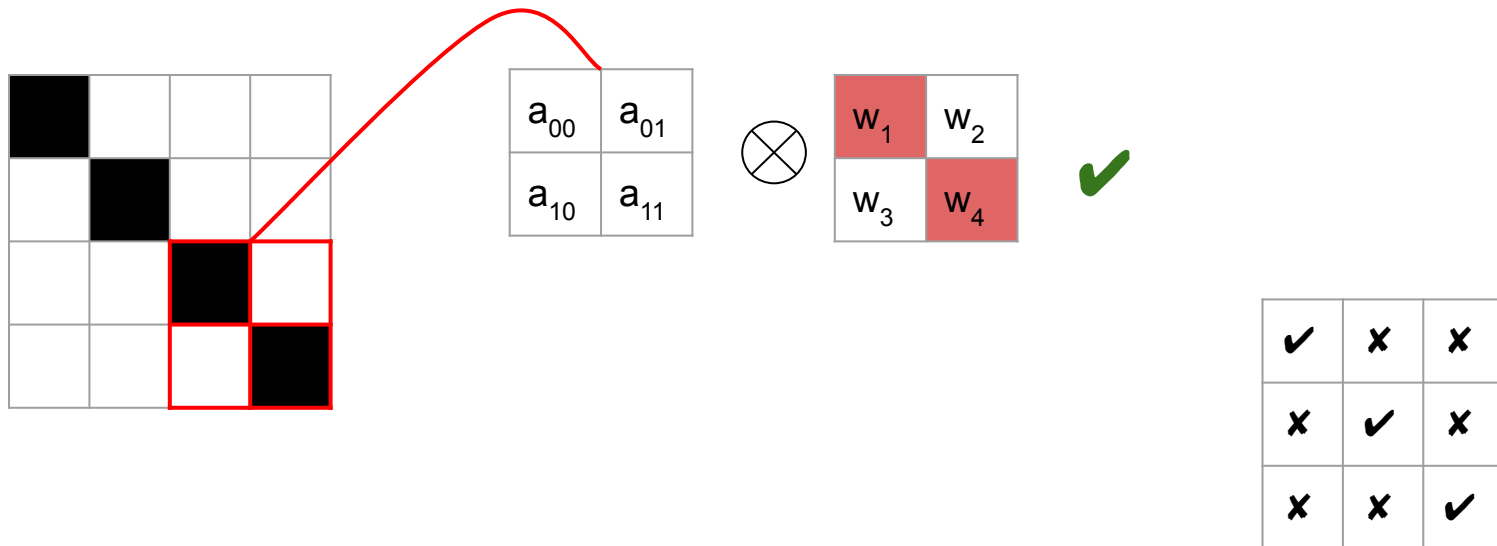
# Neural Networks - Convolutional Neural Networks

Knowing this, what happens if we slide this filter across the larger diagonal?



# Neural Networks - Convolutional Neural Networks

Knowing this, what happens if we slide this filter across the larger diagonal?



# Neural Networks - Convolutional Neural Networks

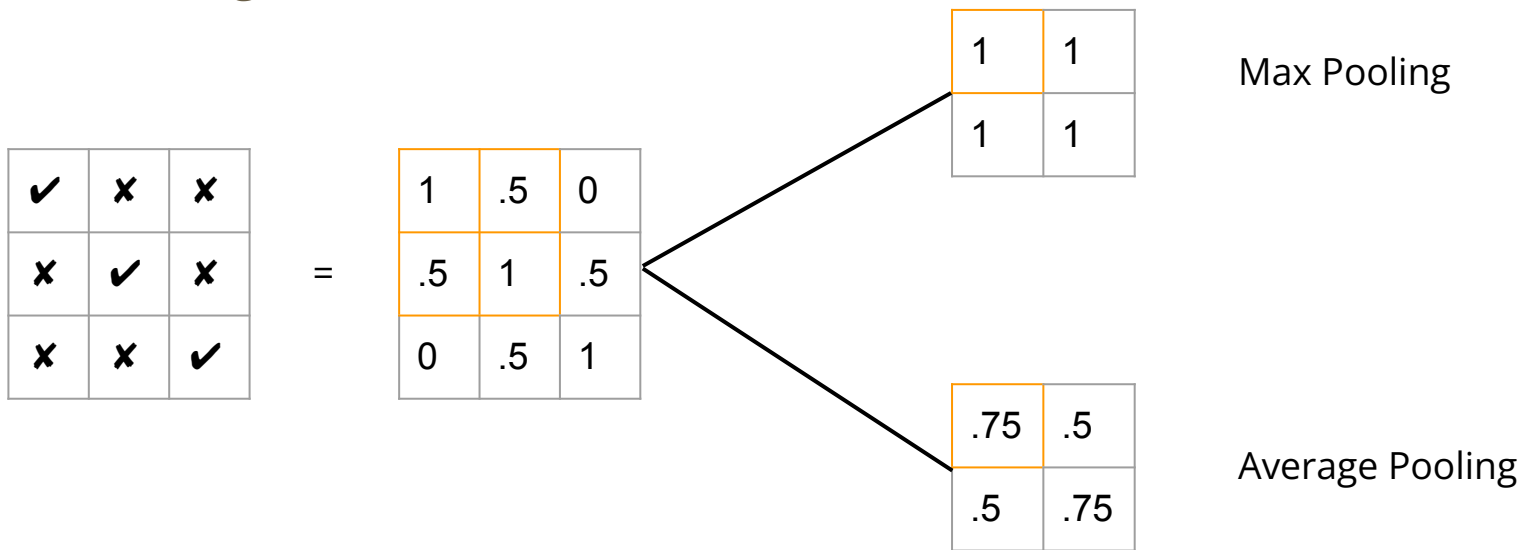
Creating such a filter allows us to:

1. Reduce the number of weights
2. Capture features all over the image

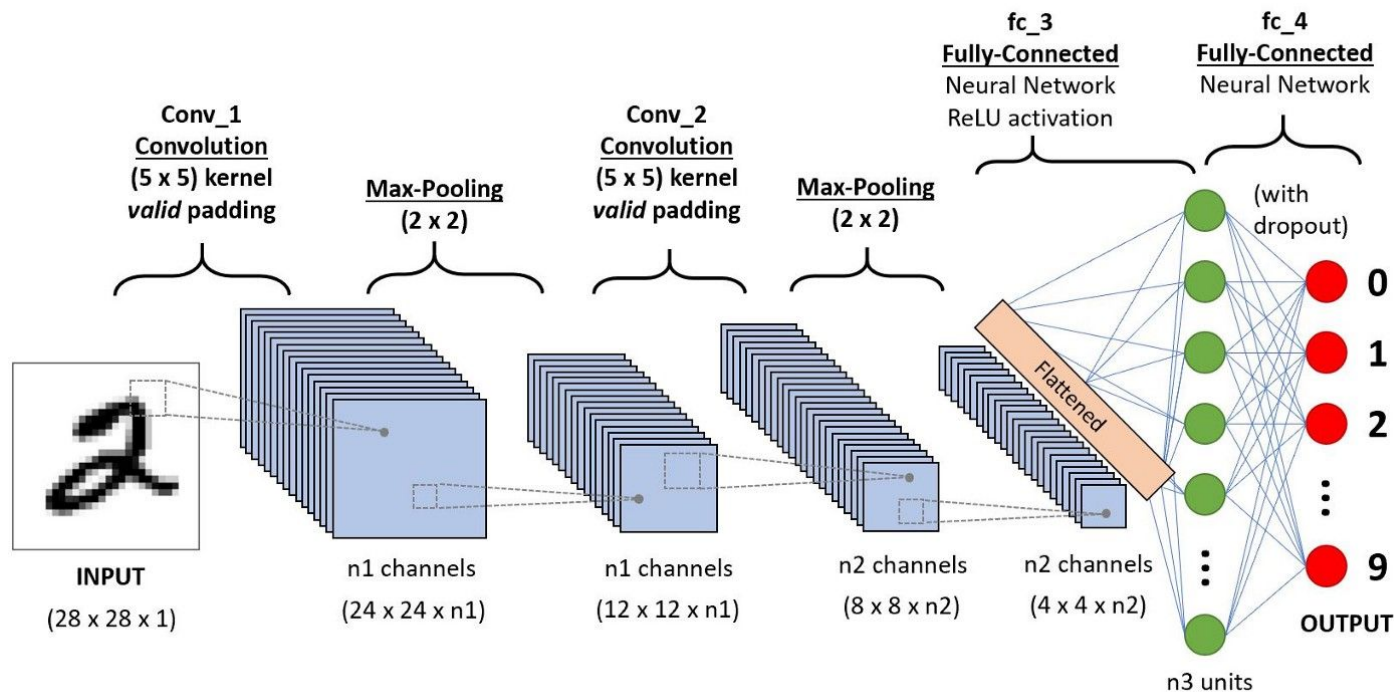
The process of applying a filter (or kernel) is called a convolution

# Neural Networks - Convolutional Neural Networks

To reduce the weights even further, another phase is done after convolution called Pooling:



# Neural Networks - Convolutional Neural Networks



# Neural Networks - Convolutional Neural Networks

Main application: Computer vision

# Recurrent Neural Networks

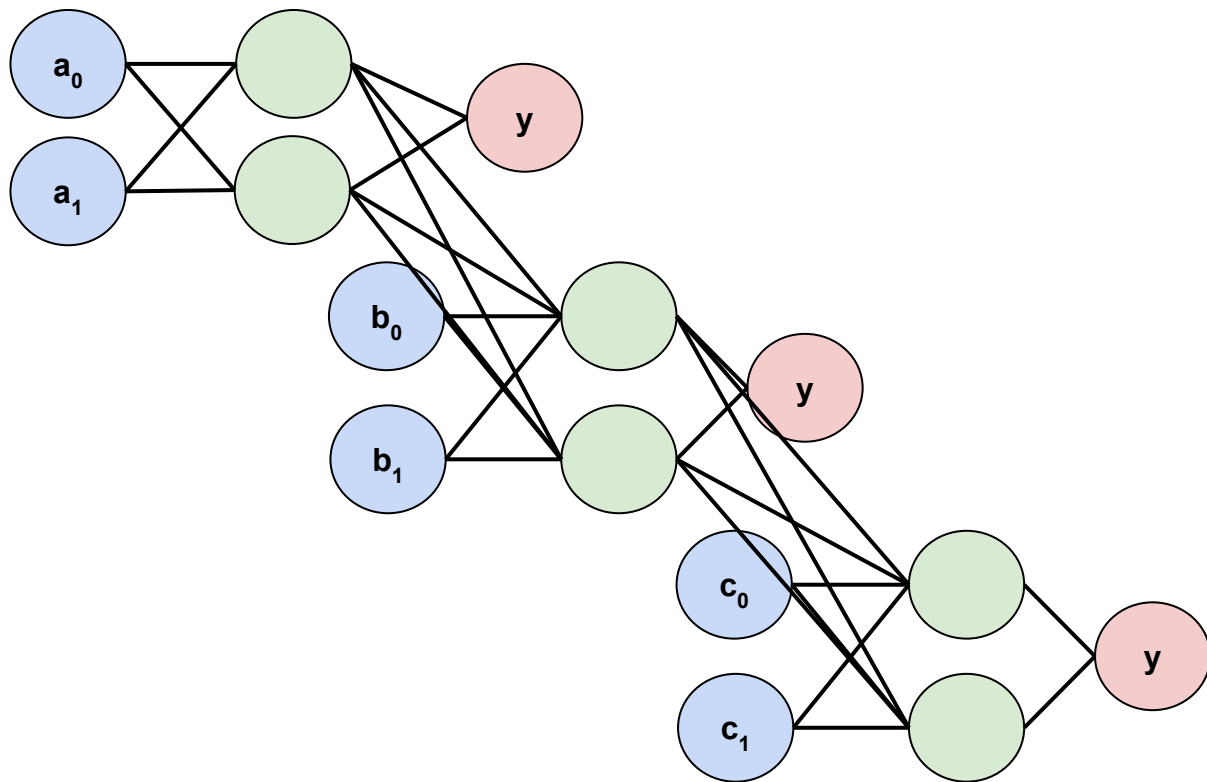
Handling sequences of input.

Intuition: What a word is / might be in a sentence is easier to figure out if you know the words around it.

Applications:

1. Predicting the next word
2. Translation
3. Speech Recognition
4. Video Tagging

# Recurrent Neural Networks





# Intro to Neural Networks

<https://medium.com/@gallettilance/list/introducing-neural-networks-d74f0dc25400>