



## CONVOLUTIONAL NEURAL NETWORKS FOR IMAGE PROCESSING

# Convolutions

Ariel Rokem

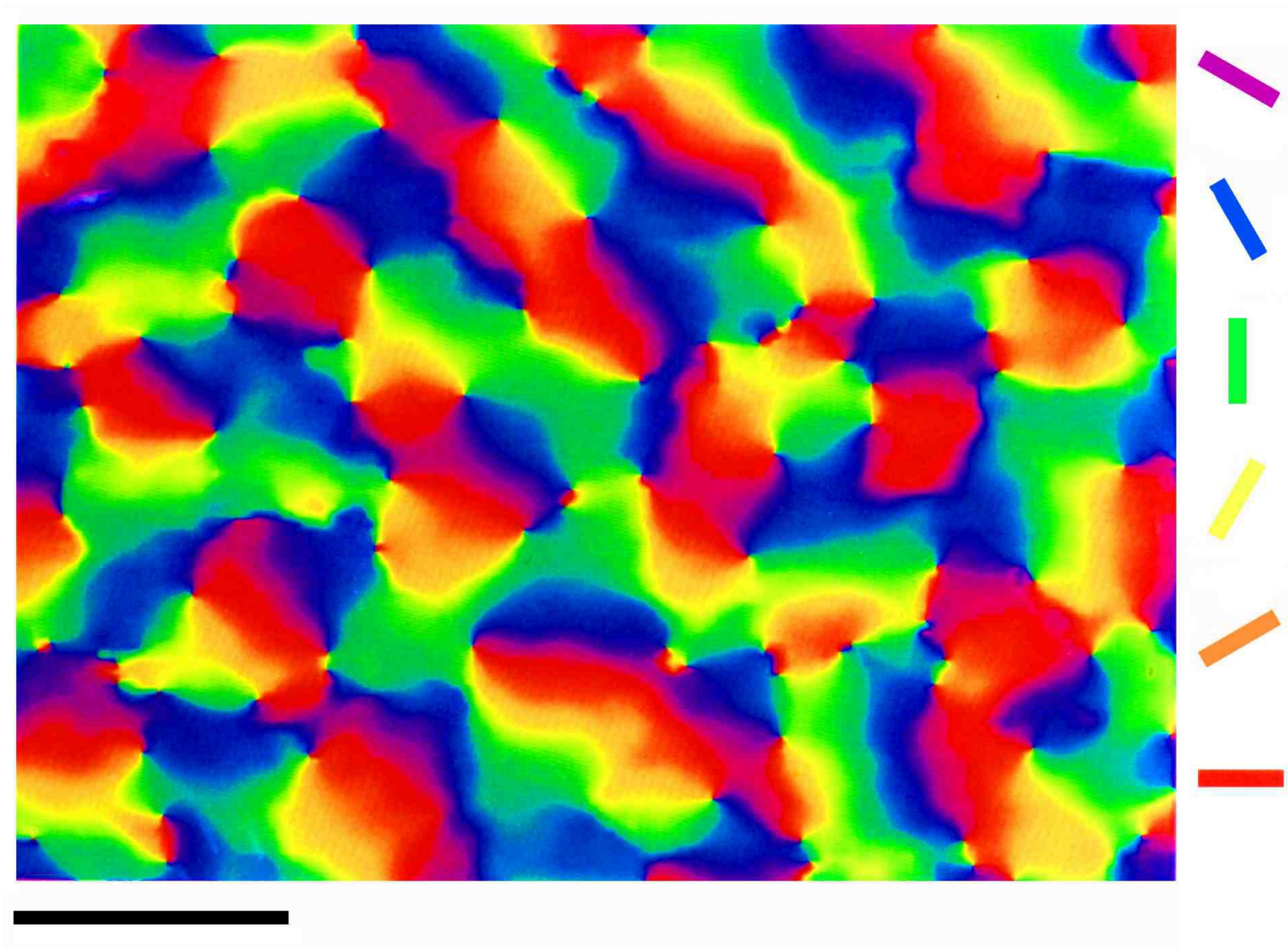
Senior Data Scientist, University of Washington



# Using correlations in images

- Natural images contain spatial correlations
- For example, pixels along a contour or edge
- How can we use these correlations?

# Biological inspiration





# What is a convolution?

```
array = np.array([0, 0, 0, 0, 0, 1, 1, 1, 1, 1])
kernel = np.array([-1, 1])
conv = np.array([0, 0, 0, 0, 0, 0, 0, 0, 0])
conv[0] = (kernel * array[0:2]).sum()
conv[1] = (kernel * array[1:3]).sum()
conv[2] = (kernel * array[2:4]).sum()
```

...

```
for ii in range(8):
    conv[ii] = (kernel * array[ii:ii+2]).sum()

conv
array([0, 0, 0, 0, 1, 0, 0, 0, 0])
```



# Convolution in one dimension

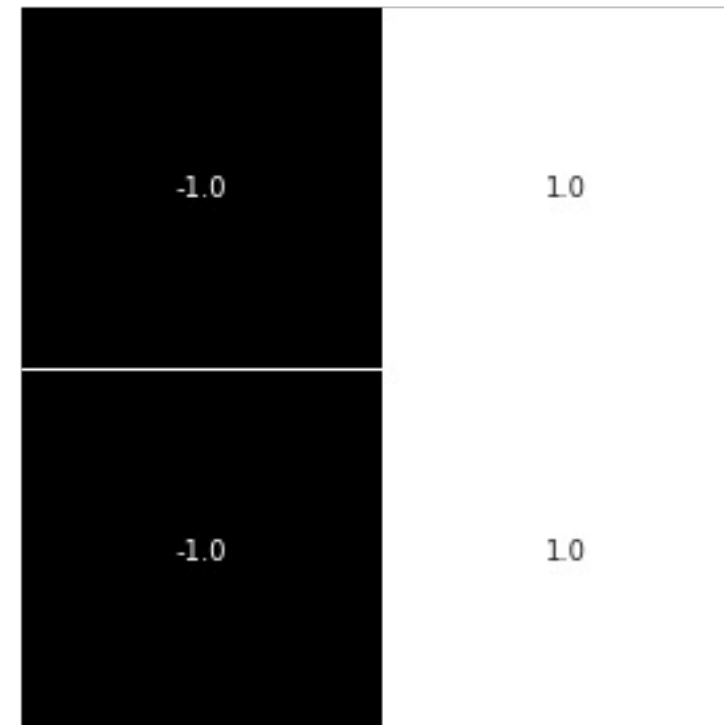
```
array = np.array([0, 0, 1, 1, 0, 0, 1, 1, 0, 0])
kernel = np.array([-1, 1])

conv = np.array([0, 0, 0, 0, 0, 0, 0, 0, 0])
for ii in range(8):
    conv[ii] = (kernel * array[ii:ii+2]).sum()

conv

array([ 0,  1,  0, -1,  0,  1,  0, -1,  0])
```

00	00	00	00	00	00	00	00	00	00	01	01	00	00	00	00	00	02	01	00	00	00	00	00	00	00	00	00	00
00	00	00	00	00	00	00	00	00	00	00	03	00	00	00	00	00	02	00	00	00	00	00	00	00	00	00	00	00
00	00	00	00	00	00	00	00	00	00	00	02	02	00	00	00	02	05	00	00	00	00	00	00	00	00	00	00	00
00	00	00	00	00	00	00	00	00	00	00	07	08	02	01	02	08	08	01	00	00	00	00	00	00	00	00	00	00
00	00	00	00	00	00	00	00	00	00	01	09	09	09	10	09	08	09	04	00	00	00	00	00	00	00	00	00	00
00	00	00	00	00	00	00	00	00	00	00	08	09	08	08	08	08	09	01	00	00	00	00	00	00	00	00	00	00
00	00	00	00	00	00	00	00	00	00	00	08	09	08	08	08	08	09	00	00	00	00	00	00	00	00	00	00	00
00	00	00	00	00	00	00	00	00	00	00	07	09	08	08	08	08	10	00	00	00	00	00	00	00	00	00	00	00
00	00	00	00	00	00	00	00	00	00	00	06	09	08	08	08	08	10	00	00	00	00	00	00	00	00	00	00	00
00	00	00	00	00	00	00	00	00	00	00	07	09	08	08	08	08	10	01	00	00	00	00	00	00	00	00	00	00
00	00	00	00	00	00	00	00	00	00	00	08	09	08	08	08	08	09	01	00	00	00	00	00	00	00	00	00	00
00	00	00	00	00	00	00	00	00	00	00	10	09	08	08	08	08	09	02	00	00	00	00	00	00	00	00	00	00
00	00	00	00	00	00	00	00	00	00	00	10	08	08	08	08	08	09	05	00	00	00	00	00	00	00	00	00	00
00	00	00	00	00	00	00	00	00	00	02	10	08	08	08	08	08	09	06	00	00	00	00	00	00	00	00	00	00
00	00	00	00	00	00	00	00	00	00	03	10	08	09	08	09	08	09	08	00	00	00	00	00	00	00	00	00	00
00	00	00	00	00	00	00	00	00	00	04	10	08	09	08	09	08	09	10	00	00	00	00	00	00	00	00	00	00
00	00	00	00	00	00	00	00	00	00	05	10	08	09	09	09	08	09	10	00	00	00	00	00	00	00	00	00	00
00	00	00	00	00	00	00	00	00	00	06	10	08	09	09	09	09	09	10	00	00	00	00	00	00	00	00	00	00
00	00	00	00	00	00	00	00	00	00	06	10	08	09	09	09	09	09	10	01	00	00	00	00	00	00	00	00	00
00	00	00	00	00	00	00	00	00	00	06	10	08	09	09	09	09	09	10	01	00	00	00</						





[illegible]

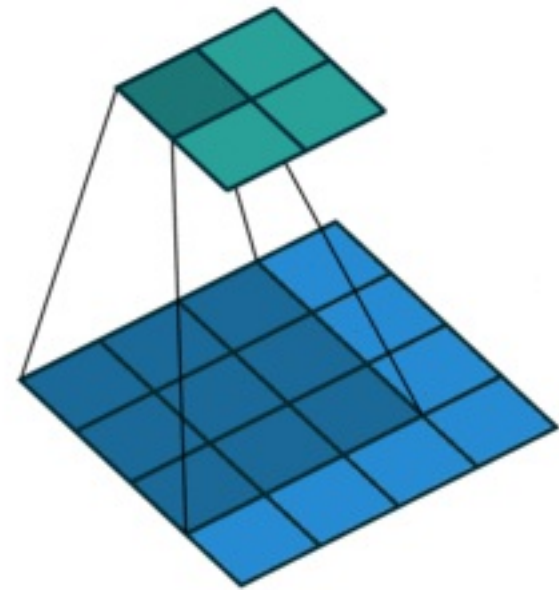


# Two-dimensional convolution

```
kernel = np.array([[ -1,  1],  
                   [ -1,  1]])  
  
conv = np.zeros((27, 27))  
  
for ii in range(27):  
    for jj in range(27):  
        window = image[ii:ii+2, jj:jj+2]  
        conv[ii, jj] = np.sum(window * kernel)
```



# Convolution





## CONVOLUTIONAL NEURAL NETWORKS FOR IMAGE PROCESSING

**Let's practice!**



CONVOLUTIONAL NEURAL NETWORKS FOR IMAGE PROCESSING

# Implementing convolutions in Keras

Ariel Rokem

Senior Data Scientist, University of Washington

# Keras 'Convolution' layer

```
from keras.layers import Conv2D
Conv2D(10, kernel_size=3, activation='relu')
```

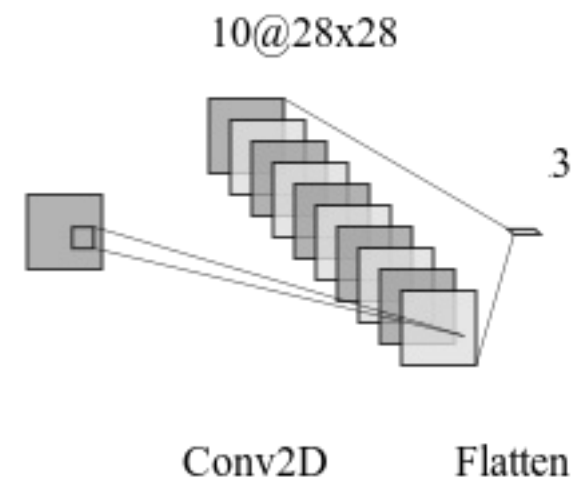
# Integrating convolution layers into a network

```
from keras.models import Sequential
from keras.layers import Dense, Conv2D, Flatten

model = Sequential()
model.add(Conv2D(10, kernel_size=3, activation='relu',
                 input_shape=(img_rows, img_cols, 1)))

model.add(Flatten())
model.add(Dense(3, activation='softmax'))
```

# Our CNN





# Fitting a CNN

```
model.compile(optimizer='adam',  
              loss='categorical_crossentropy',  
              metrics=['accuracy'])  
  
train_data.shape  
  
(50, 28, 28, 1)  
  
model.fit(train_data, train_labels, validation_split=0.2, epochs=3)  
  
model.evaluate(test_data, test_labels, epochs=3)
```





## CONVOLUTIONAL NEURAL NETWORKS FOR IMAGE PROCESSING

**Let's practice!**



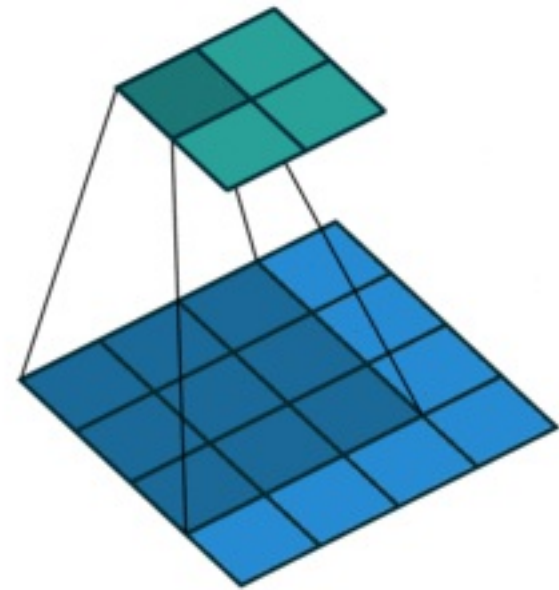
CONVOLUTIONAL NEURAL NETWORKS FOR IMAGE PROCESSING

# **Tweaking your convolutions**

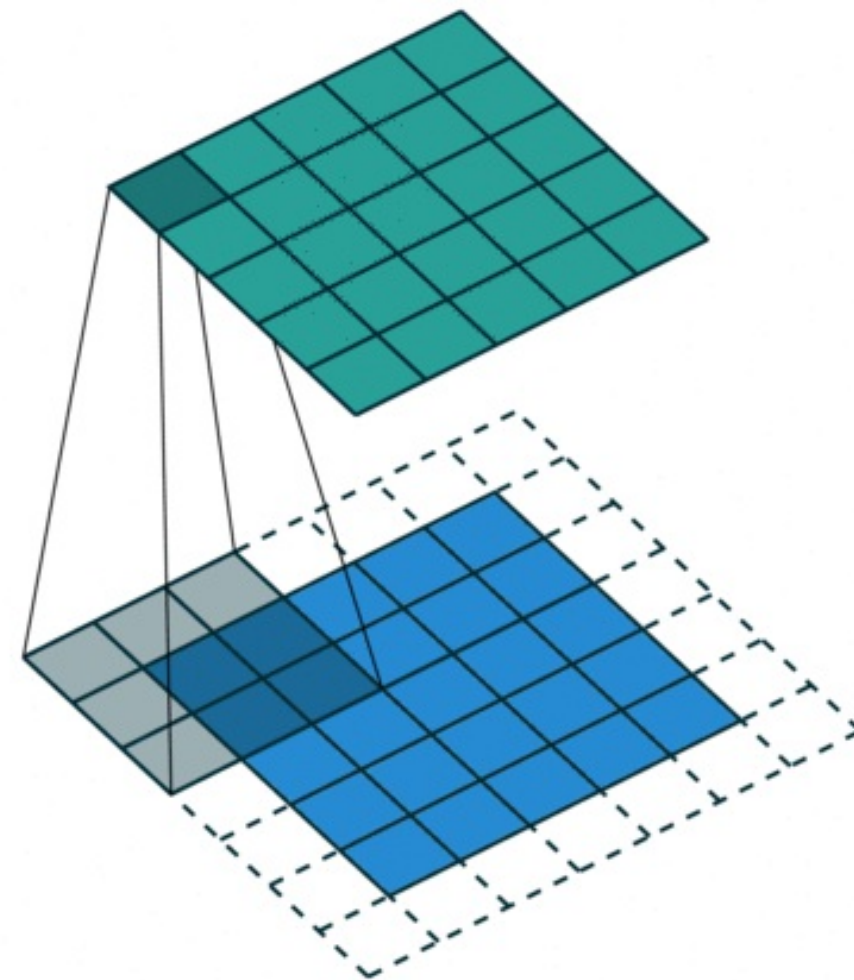
[Ariel Rokem](#)

Senior Data Scientist, University of Washington

# Convolution



# Convolution with zero padding





# Zero padding in Keras

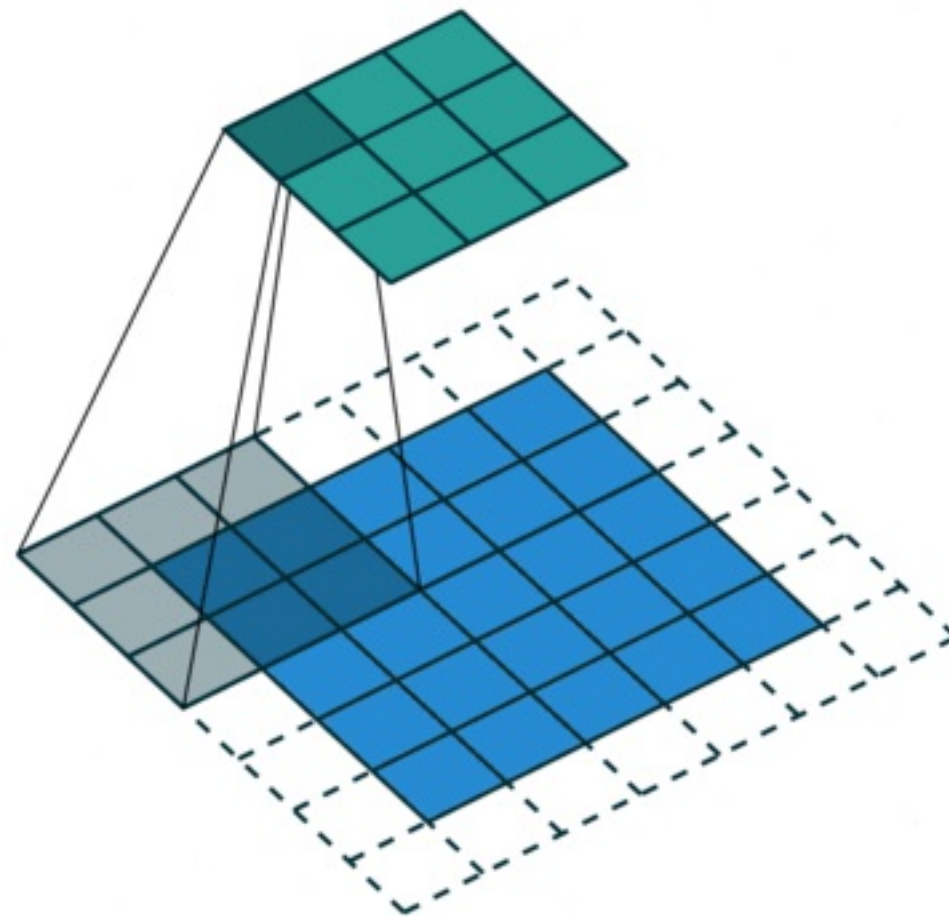
```
model.add(Conv2D(10, kernel_size=3, activation='relu',  
                 input_shape=(img_rows, img_cols, 1)),  
           padding='valid')
```



# Zero padding in Keras

```
model.add(Conv2D(10, kernel_size=3, activation='relu',  
                 input_shape=(img_rows, img_cols, 1)),  
            padding='same')
```

# Strides







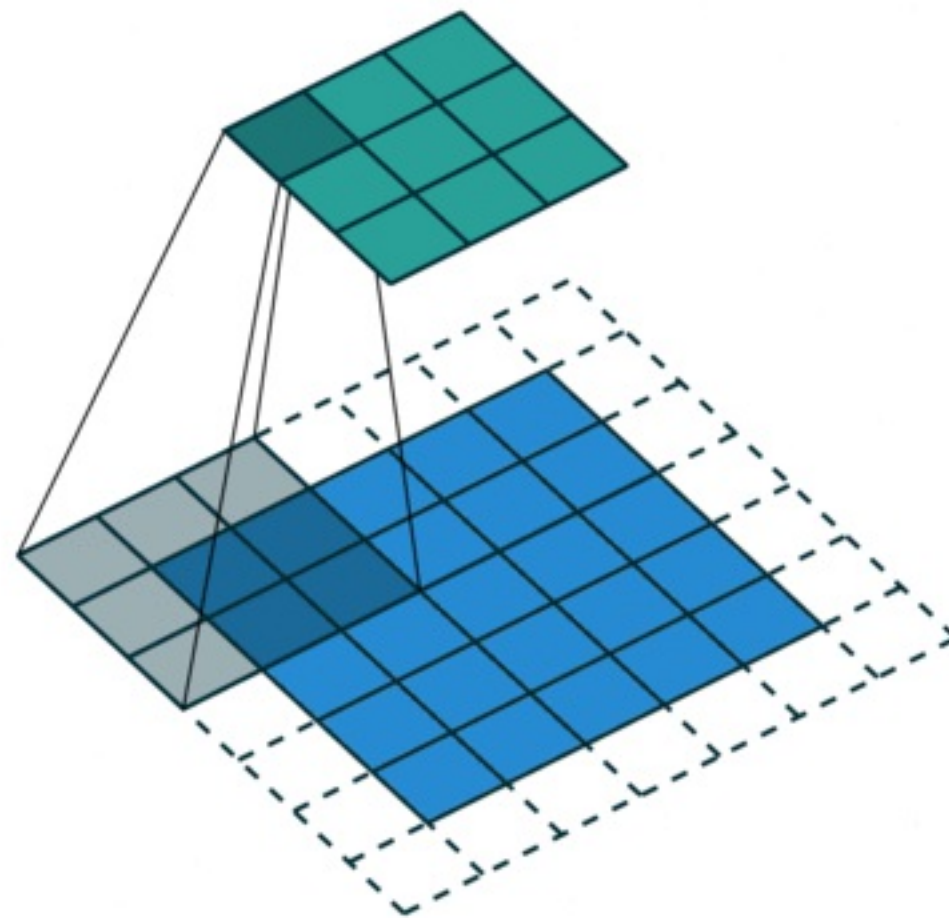
# Strides in Keras

```
model.add(Conv2D(10, kernel_size=3, activation='relu',  
                 input_shape=(img_rows, img_cols, 1)),  
            strides=1)
```



# Strides in Keras

```
model.add(Conv2D(10, kernel_size=3, activation='relu',  
                 input_shape=(img_rows, img_cols, 1)),  
            strides=2)
```





# Calculating the size of the output

$$O = ((I - K + 2P) / S) + 1$$

where

- $I$  = size of the input
- $K$  = size of the kernel
- $P$  = size of the zero padding
- $S$  = strides

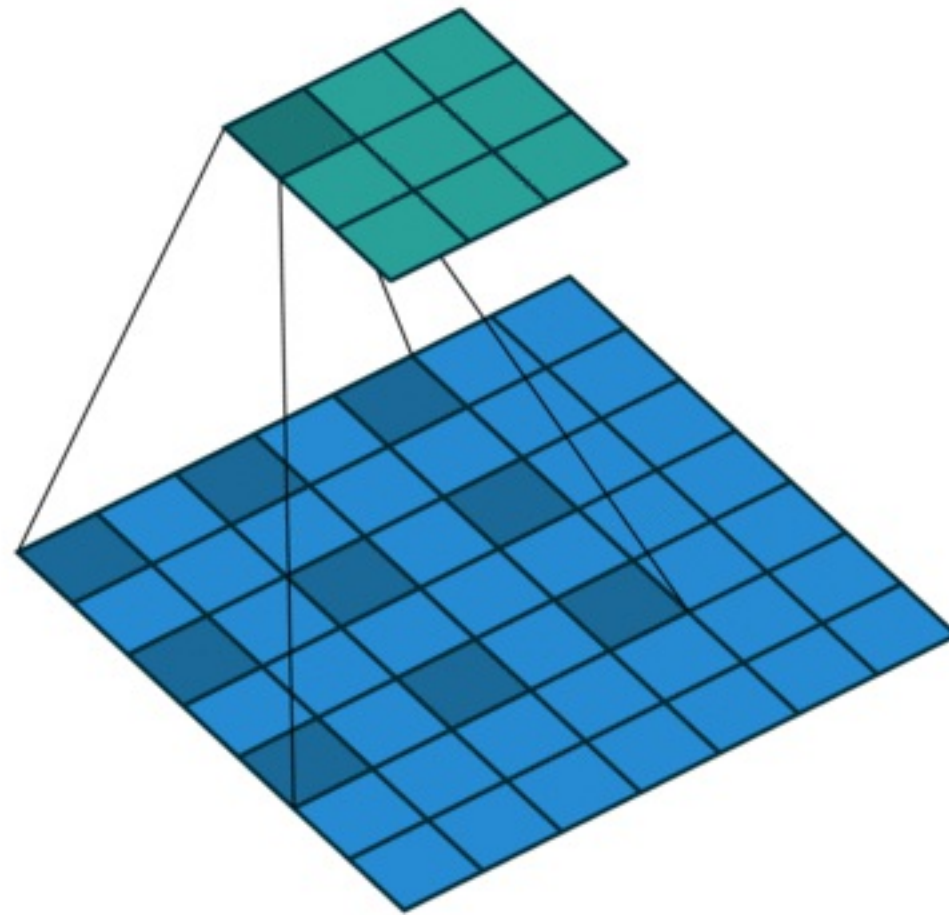


# Calculating the size of the output

$$28 = ((28 - 3 + 2)/1) + 1$$

$$10 = ((28 - 3 + 2)/3) + 1$$

# Dilated convolutions





# Dilation in Keras

```
model.add(Conv2D(10, kernel_size=3, activation='relu',  
                 input_shape=(img_rows, img_cols, 1)),  
            dilation_rate=2)
```





## CONVOLUTIONAL NEURAL NETWORKS FOR IMAGE PROCESSING

**Let's practice!**