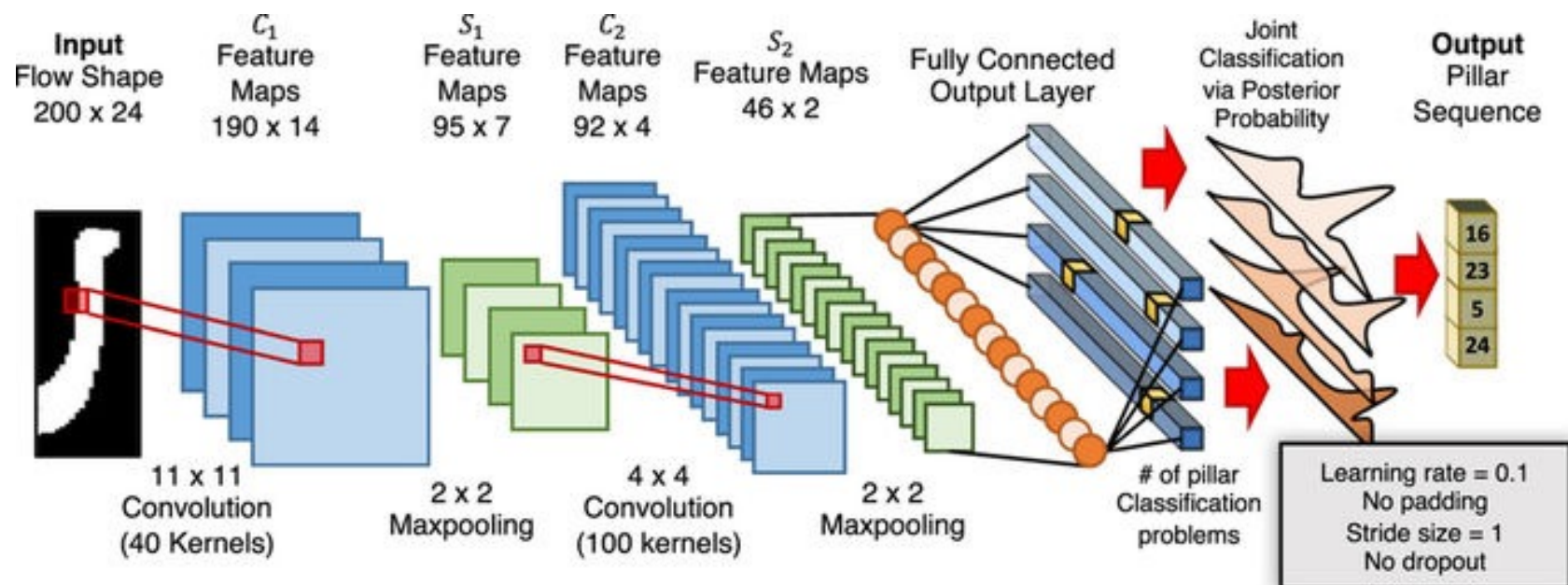


CONVOLUTIONAL NEURAL NETWORKS

Convolutional Neural Networks

Revision: What is a CNN?



Convolutional Neural Networks

Convolution Layers

- The CNN's primary distinguishing feature is the convolution layer.
 - Made up of kernels that convolve over the input and over each other.

1 _{x1}	1 _{x0}	1 _{x1}	0	0
0 _{x0}	1 _{x1}	1 _{x0}	1	0
0 _{x1}	0 _{x0}	1 _{x1}	1	1
0	0	1	1	0
0	1	1	0	0

Image

4		

Convolved
Feature

Kernel/Filter, K =

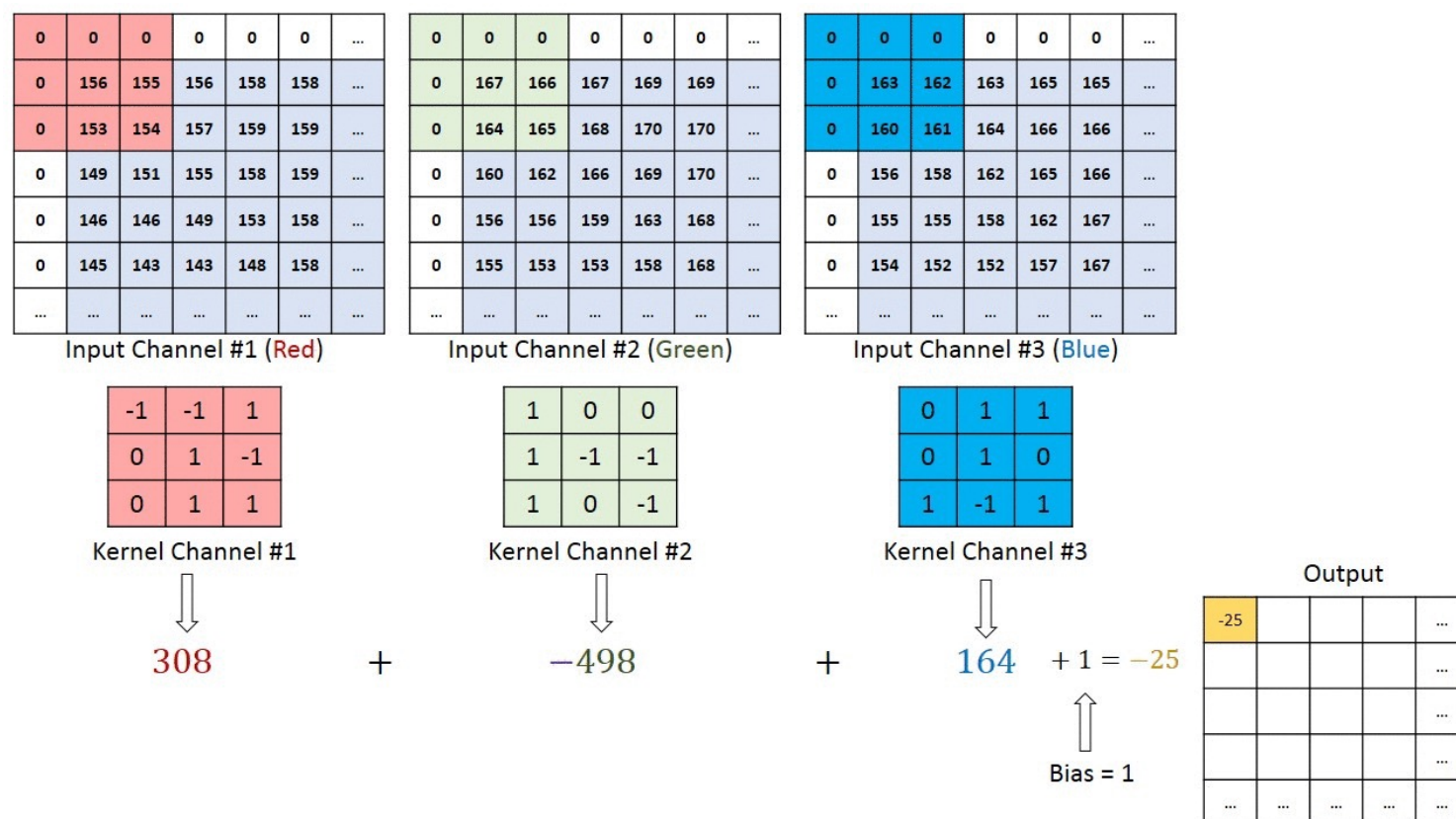
1	0	1
0	1	0
1	0	1

$$(I * K)_{xy} = \sum_{i=1}^h \sum_{j=1}^w K_{ij} \cdot I_{x+i-1,y+j-1}$$

Convolutional Neural Networks

Convolution Layers

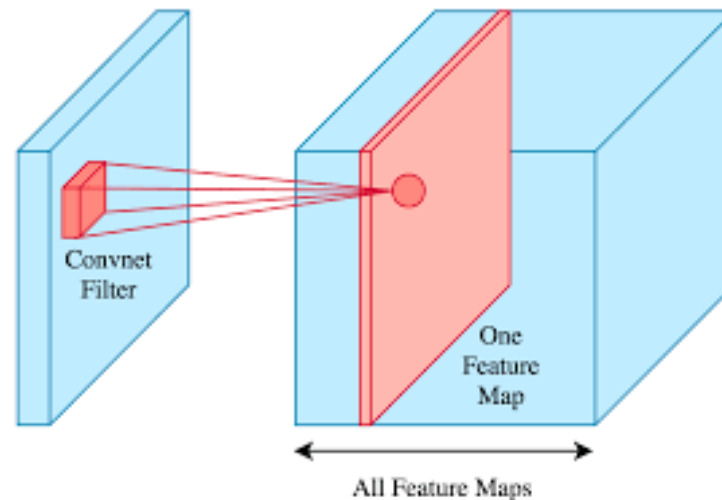
- The depth of the kernel is equal to the depth of the data. E.g. with color images there are 3 channels, and the kernel is 3 layers deep:



Convolutional Neural Networks

Convolution Layers

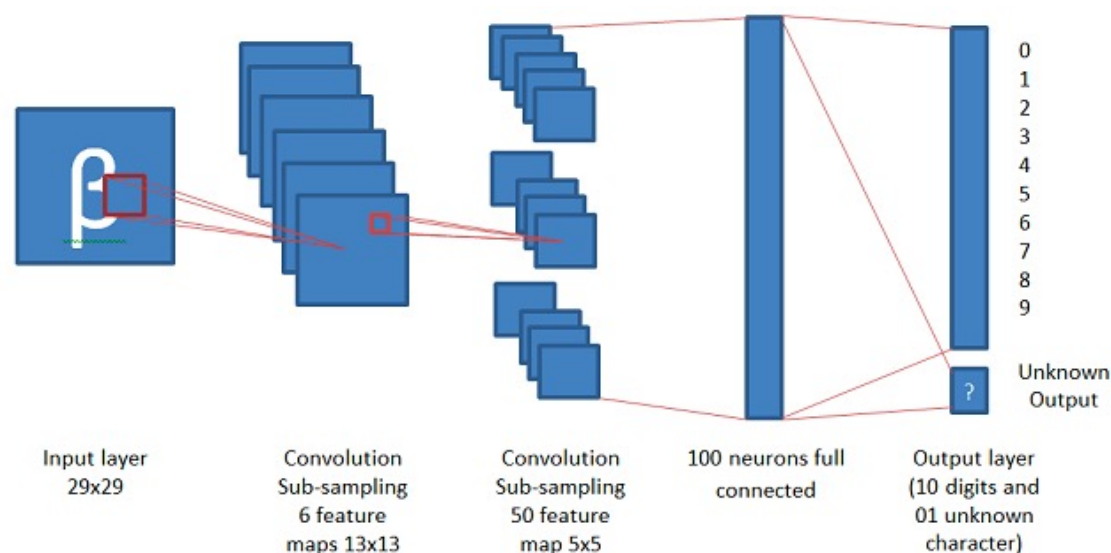
- **The kernels function as feature extractors and the output from the convolution operation is called a “feature map”**
 - As training progresses the kernels become optimized to extract key information like edges, repeating patterns, etc. in the input data.
- **Many kernels can be used on one layer.**
 - This will produce a volume of feature maps instead of a single feature map.
 - Due to the randomness of the initial kernel, each feature map ends up extracting a different feature of the input.



Convolutional Neural Networks

Convolution Layers

- **You can convolve kernels over earlier kernels.**
 - The feature maps generated will represent higher level features of the input.
 - E.g. the lower kernel might extract lines, the higher kernel might extract shapes.



Convolutional Neural Networks

Convolution Layers

- **Convolution Kernel Hyper-Parameters:**
 - **Size:**
 - ✓ E.g. 3x3, 5x5, 9x9
 - ✓ Typically smaller kernels extract more refined features at the expense of higher computational costs. Can be more sensitive to noise.
 - ✓ Larger kernels use less computational power but produce features of lower resolution.
 - **Stride:**
 - ✓ The # of steps a kernel moves to the right or down each time.
 - ✓ Smaller strides capture more data but also produce larger feature maps.
 - ✓ Larger strides produce smaller feature maps but lose more data.
 - **Padding:**
 - ✓ What do we do when we reach the end of the input data and part of the kernel “hangs out” over the edge?
 - ✓ Keras supports a “same” padding that preserves the size of the input. Other choices may increase or decrease the size of the input.

Convolutional Neural Networks

Convolution Layers in Keras

- **Keras offers many types of convolution layers:**

<https://keras.io/layers/convolutional/>

- padding:

- ✓‘valid’: Convolution stops when the right/bottom edge of the filter hits the right/bottom edge of the input data. Will create a feature map smaller than the input

- ✓‘same’: Input is padded with ‘0’ to allow the feature map to be the same size as the input.

- activation:

- ✓If None, Keras will use an identity activation. I.e. $\text{id}(x) = x$, where x is the feature map computed by the convolution operation.

- ✓Otherwise can specify other activations like ReLU, tanh etc.

- kernel_size:

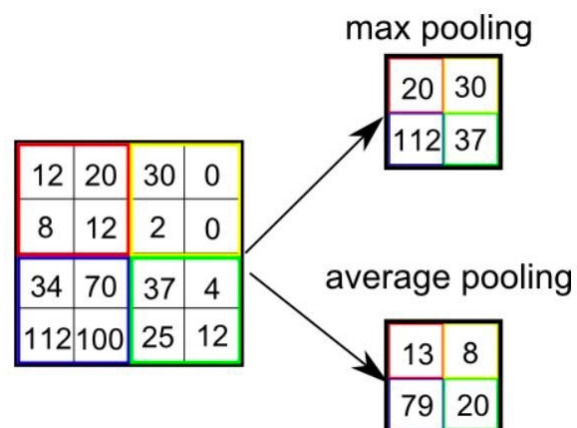
- ✓Size of the kernel window. E.g. (3, 3) makes a 3x3 filter

- filters: Number of kernels to make.

Convolutional Neural Networks

Pooling Layers

- The pooling layer takes a $n \times n$ region of the feature map and either picks the maximum or takes an average:



- **Pooling layers:**
 - Reduce dimensionality and hence computing power.
 - Extracts the most dominant features (max pooling):
 - ✓ Makes the model shift invariant and less noise sensitive
 - averages the features (average pooling)
 - ✓ Reduces the effect of noise by averaging it out – white noise has an average of 0.

Convolutional Neural Networks

Pooling Layers

- **Pooling layer hyperparameters:**
 - **stride:** As with the convolution layer, this determines the number of steps the pool moves each time to the right or down.
 - **size:** Controls the dimensionality reduction.
 - ✓ **Generally if stride = size, the feature map is reduced by 1/size in each dimension.**
 - **Keras pooling API:** <https://keras.io/layers/pooling/>
 - ✓ **pool_size:** The length or dimensions of the pool.
 - ✓ **strides:** As described above.
- **Generally the first few (or many?) layers of the CNN consist of alternating convolution and pooling layers:**
 - However stacking too many pooling layers can result in losing all data.

Convolutional Neural Networks

Flatten Layer / Example

- **The Flatten layer turns the feature maps into a 1D vector for feeding into the Dense layers.**
- **Example:**

```
model = Sequential()
model.add(Conv2D(32, kernel_size=(5,5),
activation='relu',
input_shape=(28, 28, 1), padding='same'))

model.add(MaxPooling2D(pool_size=(2,2), strides=2))
model.add(Conv2D(64, kernel_size=(5,5), activation='relu'))
model.add(Conv2D(128, kernel_size=(5,5), activation='relu'))
model.add(Conv2D(64, kernel_size=(5,5), activation='relu'))
model.add(MaxPooling2D(pool_size=(2,2), strides=2))
model.add(Flatten())
model.add(Dense(1024, activation='relu'))
model.add(Dropout(0.1))
model.add(Dense(10, activation='softmax'))
```