Machine Learning and Artificial Intelligence

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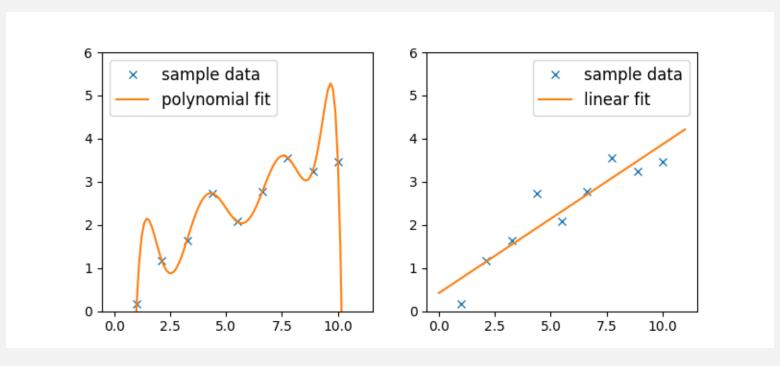




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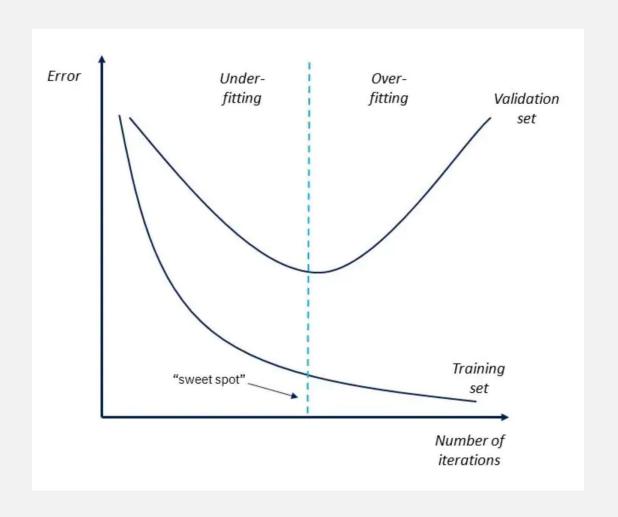
Overfitting



0 loss, poor generalization

Good generalization

Overfitting



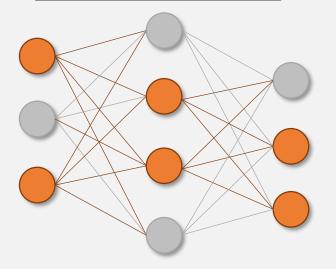
Overfitting

Remedies

Regularization

$$L \mapsto L + \lambda \sum_{ij\ell} |\omega_{ij}^{\ell}|$$

Dropout

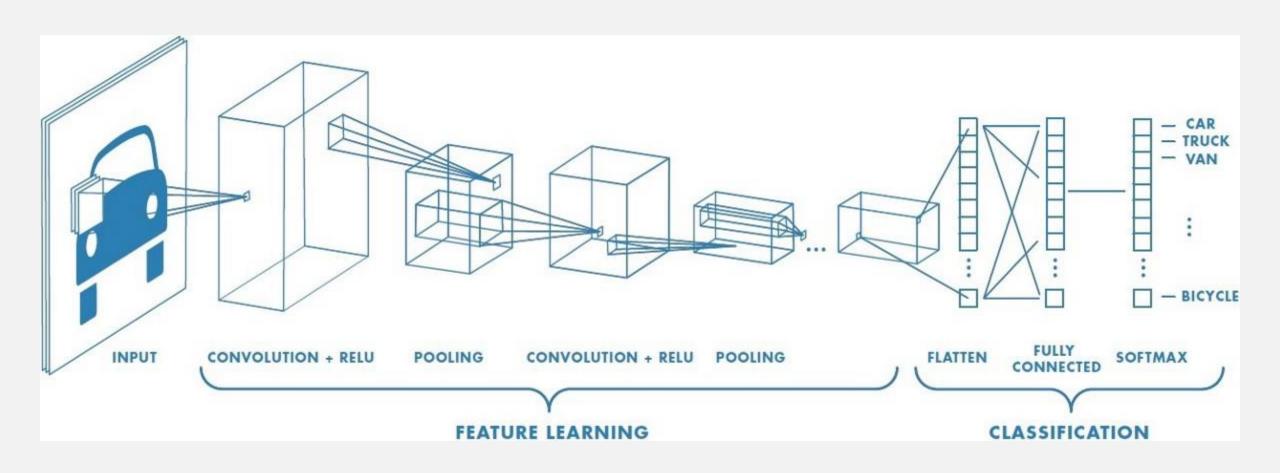


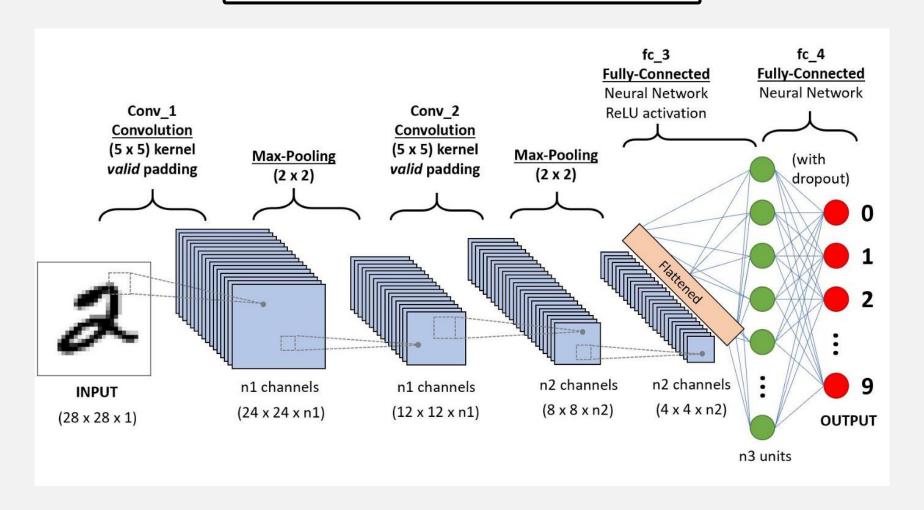
Data augmentation

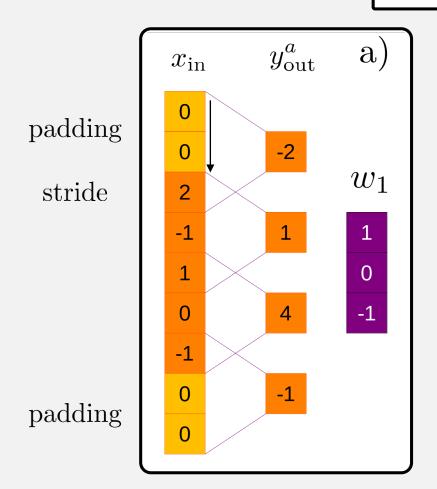


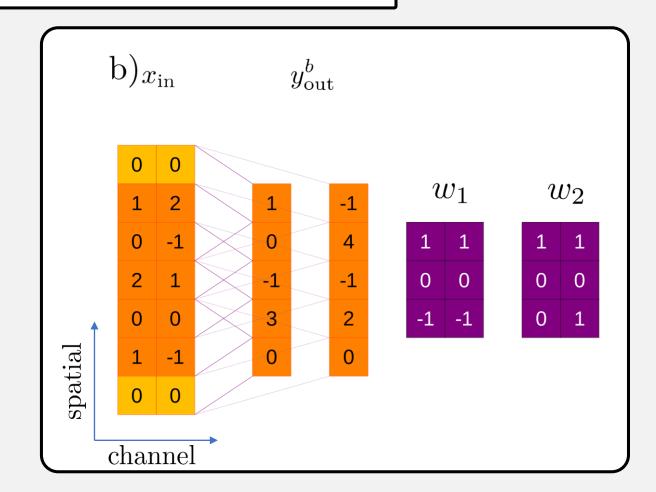


Part II: Convolutional Layers



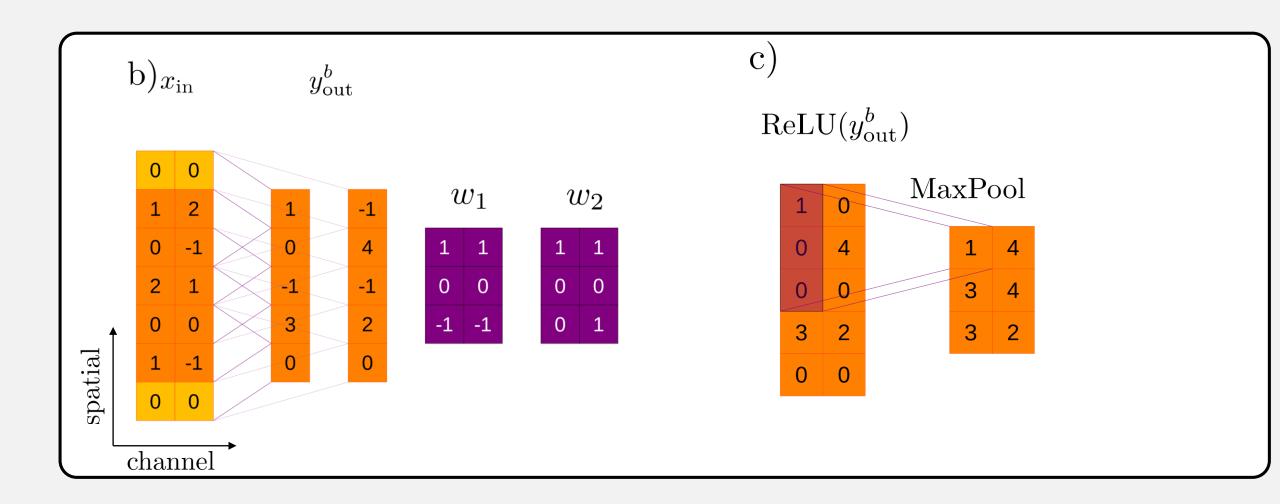




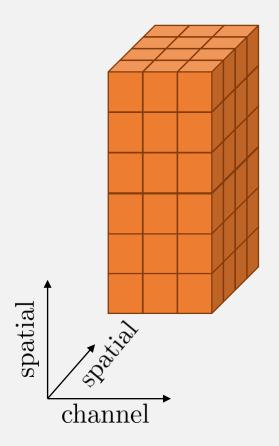


$$|y| = (|x| - |w| + 2\text{pad})/\text{stride} + 1$$

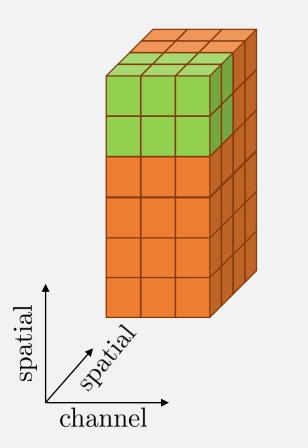
(5 - 3 + 2*2)/2 + 1 = 4



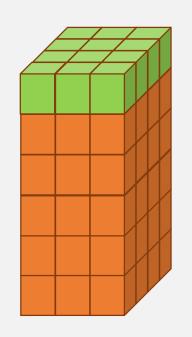
Define (d_1, d_2) and n_{filters}



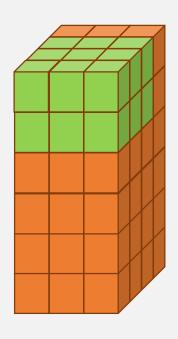
Action of (2,2) filter



Action of (4,1) filter



Action of (3,2) filter



Ex: Given color images of 32x32 (|x| = (32, 32, 3)) and the following convolutional neural network, what is are the input and output dimensions for each layer?

```
class NeuralNetwork(nn.Module):
    def init (self):
        super(NeuralNetwork, self). init ()
        self.flatten = nn.Flatten()
        self.conv block = nn.Sequential(
            nn.Conv2d(in channels=1, out channels=16, kernel size=3, padding="same"), # (28 x 28 x 1) -> (28 x 28 x 16)
            nn.ReLU(),
                                                      # (28 x 28 x 16) -> (14 x 14 x 16)
            nn.MaxPool2d(2),
            nn.Conv2d(16, 16, 3, padding="same"), # (14 x 14 x 16) -> (14 x 14 x 16)
            nn.ReLU(),
                                                      \# (14 \times 14 \times 16) \rightarrow (7 \times 7 \times 16)
            nn.MaxPool2d(2),
            nn.Conv2d(16, 16, 3, padding="same"), # (7 x 7 x 16) -> (7 x 7 x 16)
            nn.ReLU(),
            nn.Flatten())
                                                        \# (7 \times 7 \times 16) \rightarrow (7*7*16)
        self.dense block = nn.Sequential(
            nn.Linear(7*7*16, 512), # input dimension, hidden1 dimension
            nn.ReLU(),  # Non-linear activation function
nn.Linear(512, 512),  # hidden1 dimension, hidden2 dimension
                         # Non-linear activation function
            nn.ReLU(),
            nn.Linear(512, out dim), # hidden2 dimension, output dimension
               17 10 1
```

Exercises

- Ex: Write (by hand using e.g. numpy) a fully connected layer
- Ex: Connect multiple fully connected layers by hand.
- Ex: Write (by hand using e.g. numpy) a 1D convolution function conv1D(x, filter, stride, padding) to process an input vector x (of arbitrary size).
 - The filter can be arbitrary but fixed. Try writing it such that it works for any stride, padding and filter size.
- Extra (difficult): Write all the above functions in such a way that they can process batches of inputs (so instead of a single input x, a collection of batch_size inputs x)