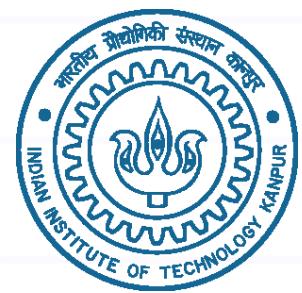


Convolutional Neural Networks

EE698V - Machine Learning for Signal Processing

Vipul Arora



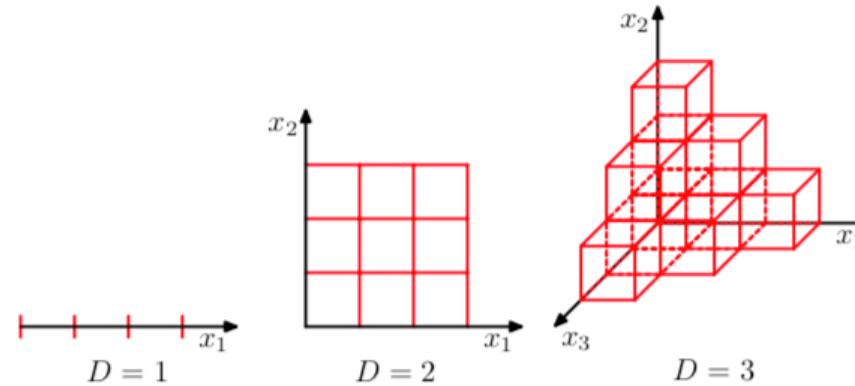
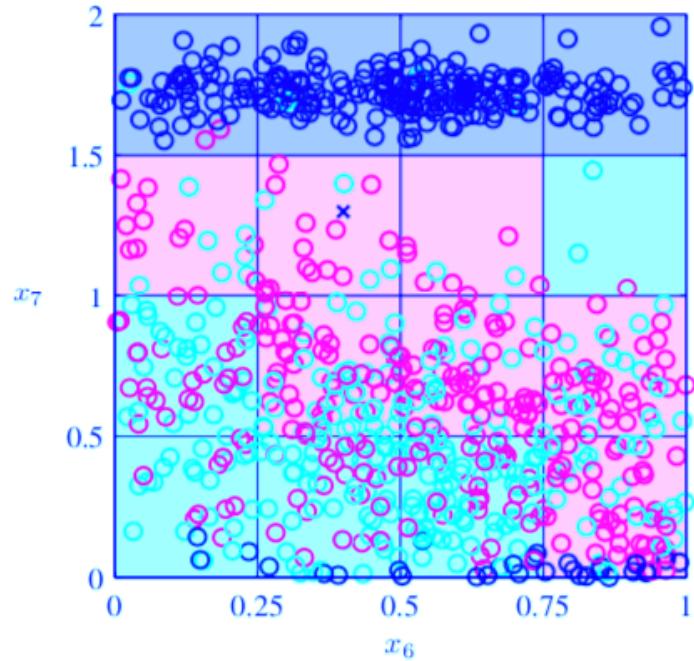
Announcements

- Theory Quiz on Oct 31 (Thurs) or Nov 1 (Fri)
- Coding Quiz on Nov 3 (Sun)

Curse of Dimensionality

- We have been studying about dimensionality reduction
- Let us try to appreciate the scale of the problem of large dimensionality
- **Reference:** PRML Section 1.4

For Classification



$$\frac{D}{\begin{array}{c} 2 \\ 3 \end{array}} \mid N = 4^D \text{ (no. of cells)}$$

16
14

For Images



Template Matching

- We can find blobs/regions that are likely to classify the image



- Cross-correlation of these templates with the test image will give a matching score

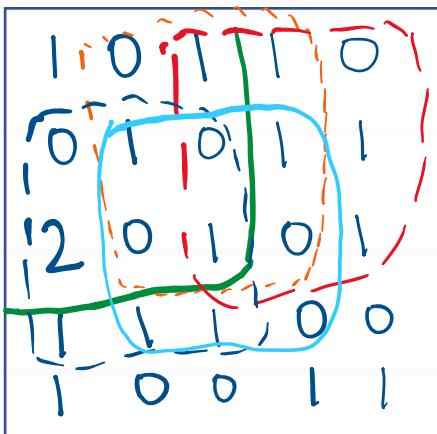
Finding optimal templates

- We can find these templates optimally from the data discriminatively
- Recall, cross-correlation
- $H = X \otimes W$

2D Convolution: actually cross-correlation

- $H[i_1, j_1] = \sum_{k=0}^K \sum_{l=0}^K X[i_1s + k, j_1s + l] W[k, l]$
- s is the hop size (stride)
- K is the kernel size (template size)
- $H[i_1, j_1]$ will be maximum when match occurs

2D Convolution



⊗

$$\begin{matrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{matrix}$$

=

$$\begin{matrix} 3 & 0 & 3 \\ 1 & 2 & 0 \\ 3 & 2 & 2 \end{matrix}$$

hop size, $h=1$

$$\boxed{\begin{matrix} 3 & 3 \\ 3 & 2 \end{matrix}}$$

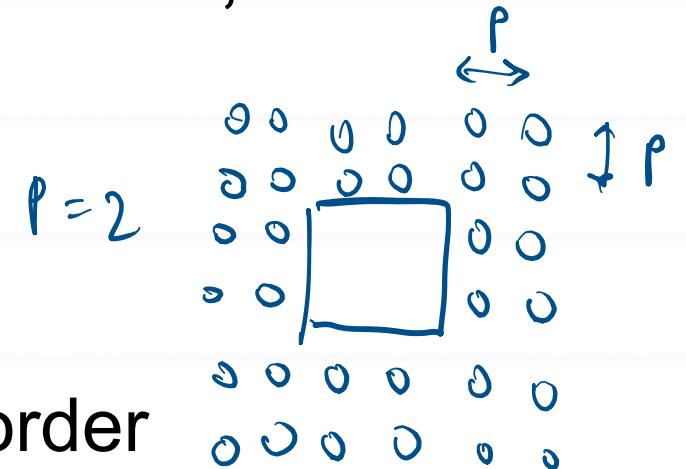
$h=2$

- Use hop size = 1 or 2

Question

- Given an image of size $I \times J$ and a kernel of size $K \times L$, with a hop of s and s' along the two dimensions, what is the size of the output:

- without any zero padding
- with a zero padding of width P along each border



How to optimize kernel for classification

- Define multiple kernels, say N
- Find H for each kernel
- Give these as input to a neural network classifier
- Train end to end !!

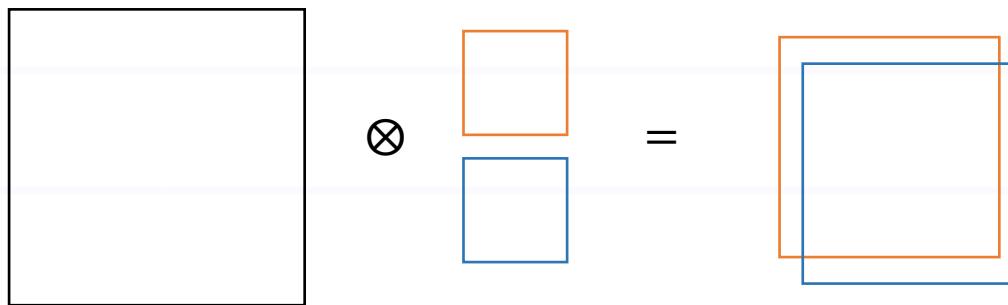
How to optimize kernel for classification

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$$\begin{array}{c} \boxed{} \\ \otimes \end{array} \quad \boxed{} = \quad \boxed{}$$

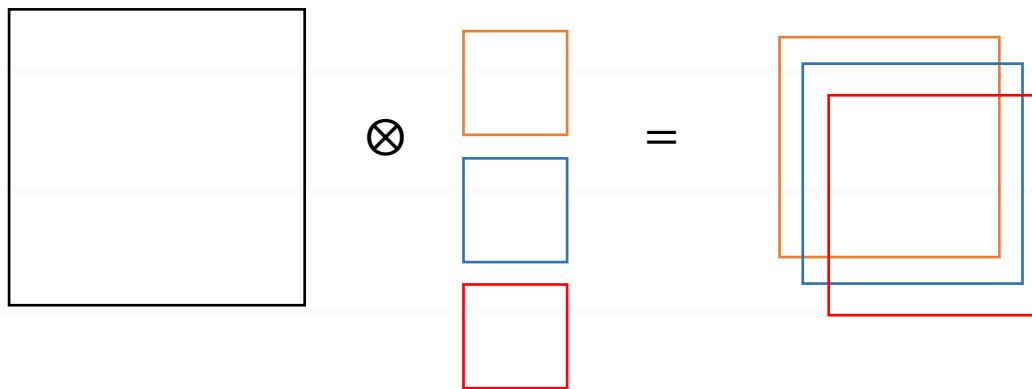
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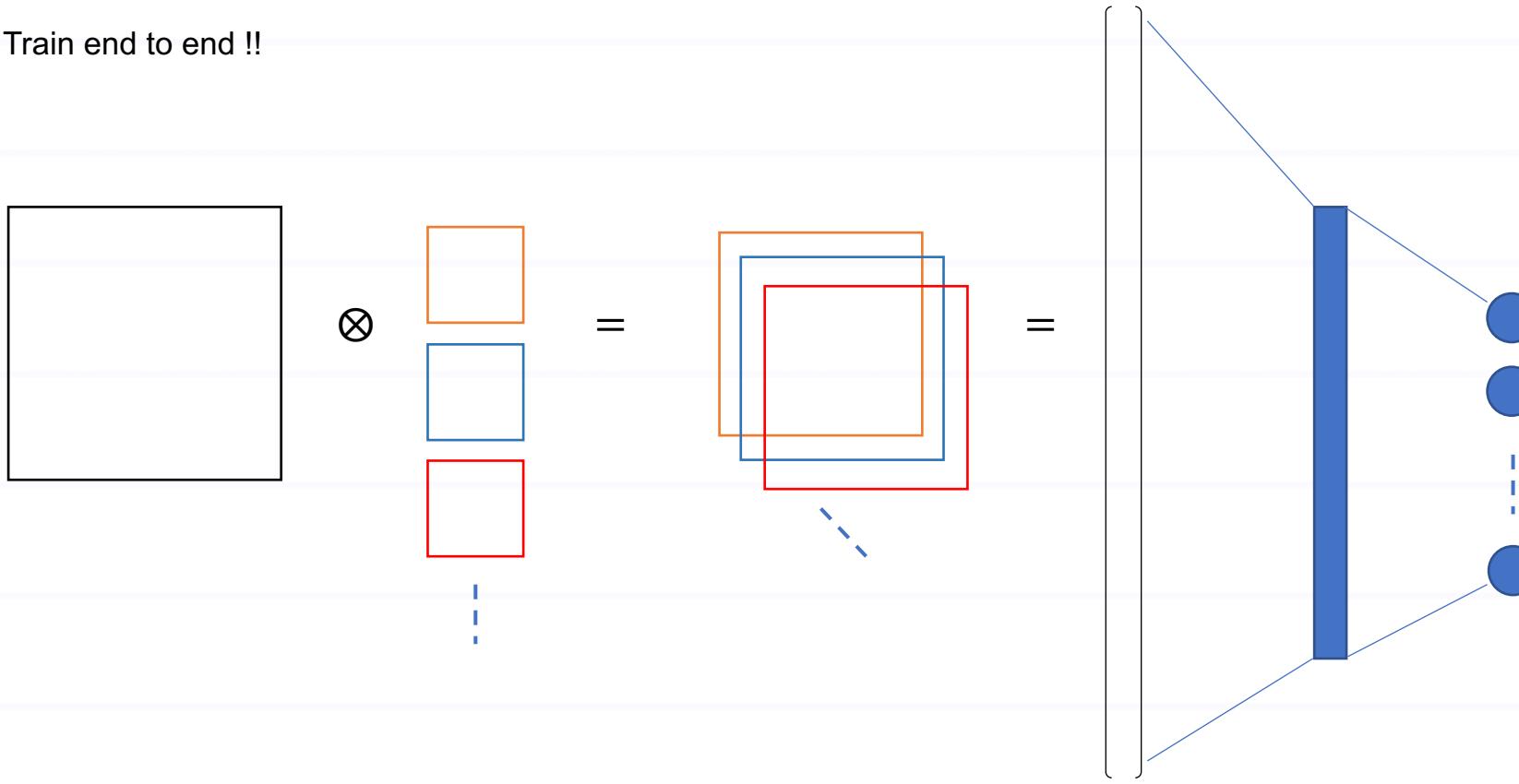
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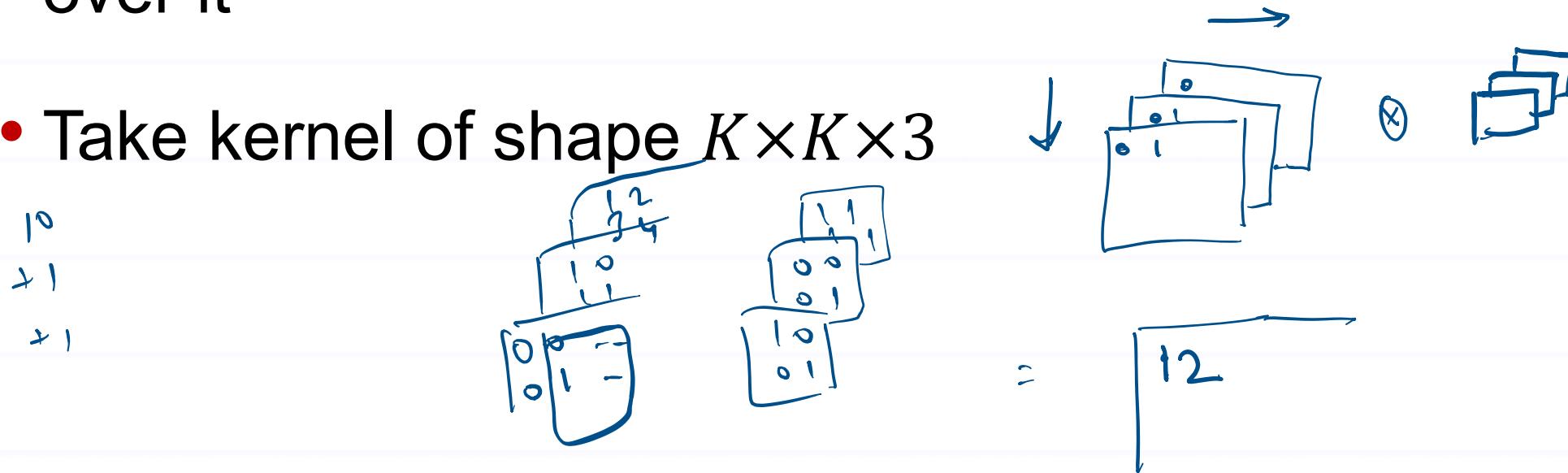
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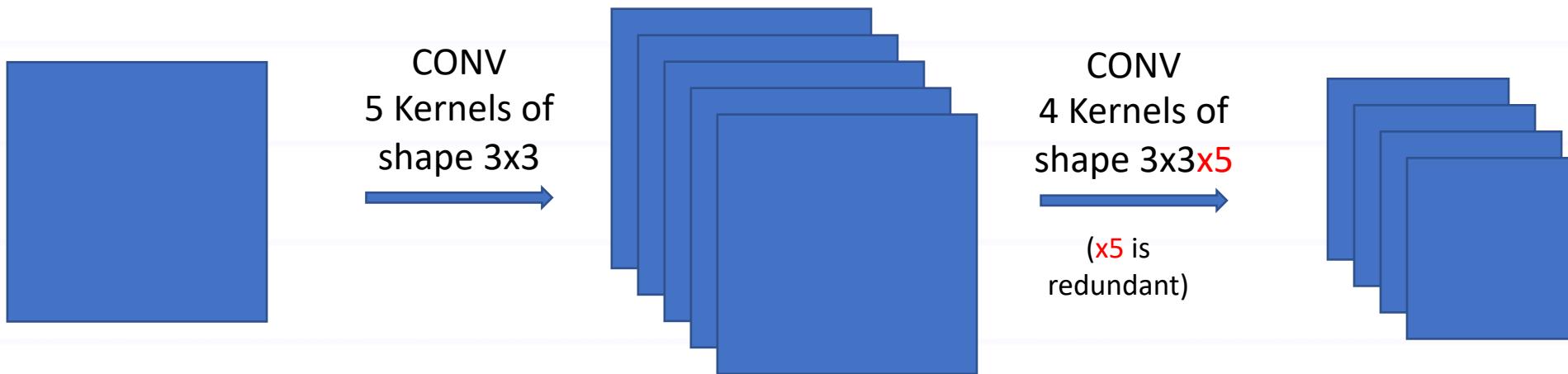


How to deal with RGB images?

- Input is of size $I \times J \times 3$
- The 3 channels are correlated, so no point in convolving over it
- Take kernel of shape $K \times K \times 3$



We can stack layers of kernels



- This is known as
 - Convolutional Neural Network
 - ConvNet
 - CNN

CNNs

References: (highly recommended)

- <https://youtu.be/j41cU9ad7SU>
- https://youtu.be/P5OgPz_h9CE

Standard concepts

Learn the following concepts from the reference videos. No need to memorize the architectures of the popular CNNs.

- Pooling
- Drop out
- Flattening
- Find the number of parameters in each layer
- Find the computational complexity of each layer
- The problem of vanishing gradients

Standard practices

- Generally odd sized kernel are used. E.g. $K=3$ or $K=5$ or $K=7$
- Use padding $P=(K-1)/2$ and $s=1$ for output size to be same as input size
- ReLU activation is used to take care of vanishing gradients

