

CSE 575: Statistical Machine Learning (Spring 2021)

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Neural Networks & Deep Learning



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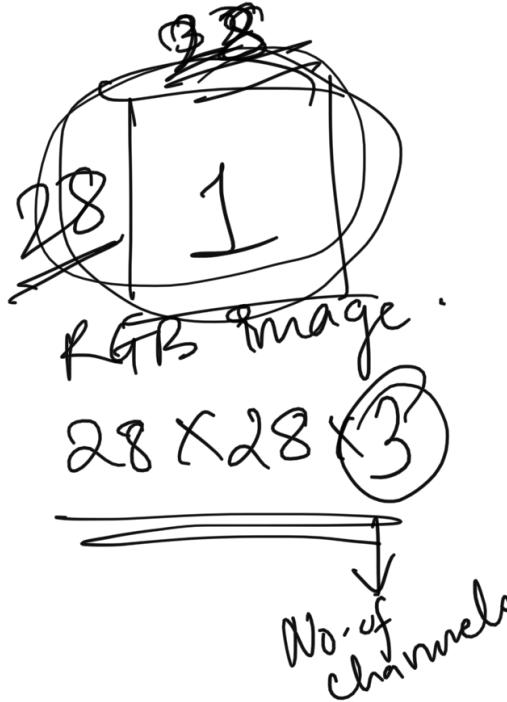
3. Batch normalization

Number of parameters in dense layers

FC (fully connected).

$$\underline{224 \times 224}$$

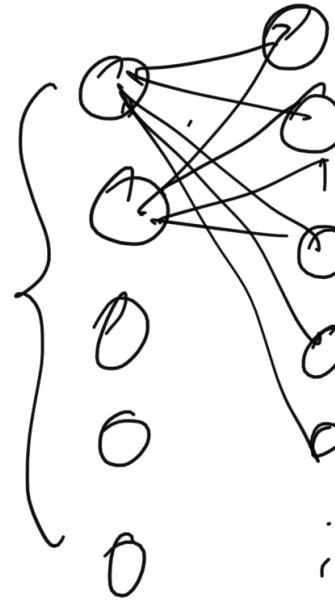
1000.



$$28 \times 28 \times 3 \Rightarrow 784 \times 3$$



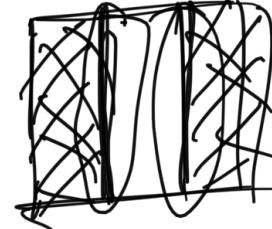
784 x 3
parameters



huge no. of
parameters

$$\underline{784 \times 3 \times 1000}$$

Convolutional Neural Network (CNN)



- Basic building block is the convolution for image processing
- Weight sharing - reduces parameters
- Learn the kernels based on the task.

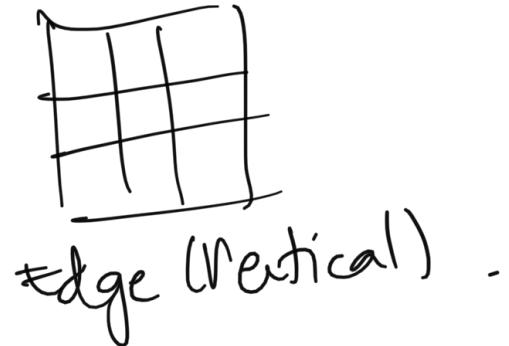
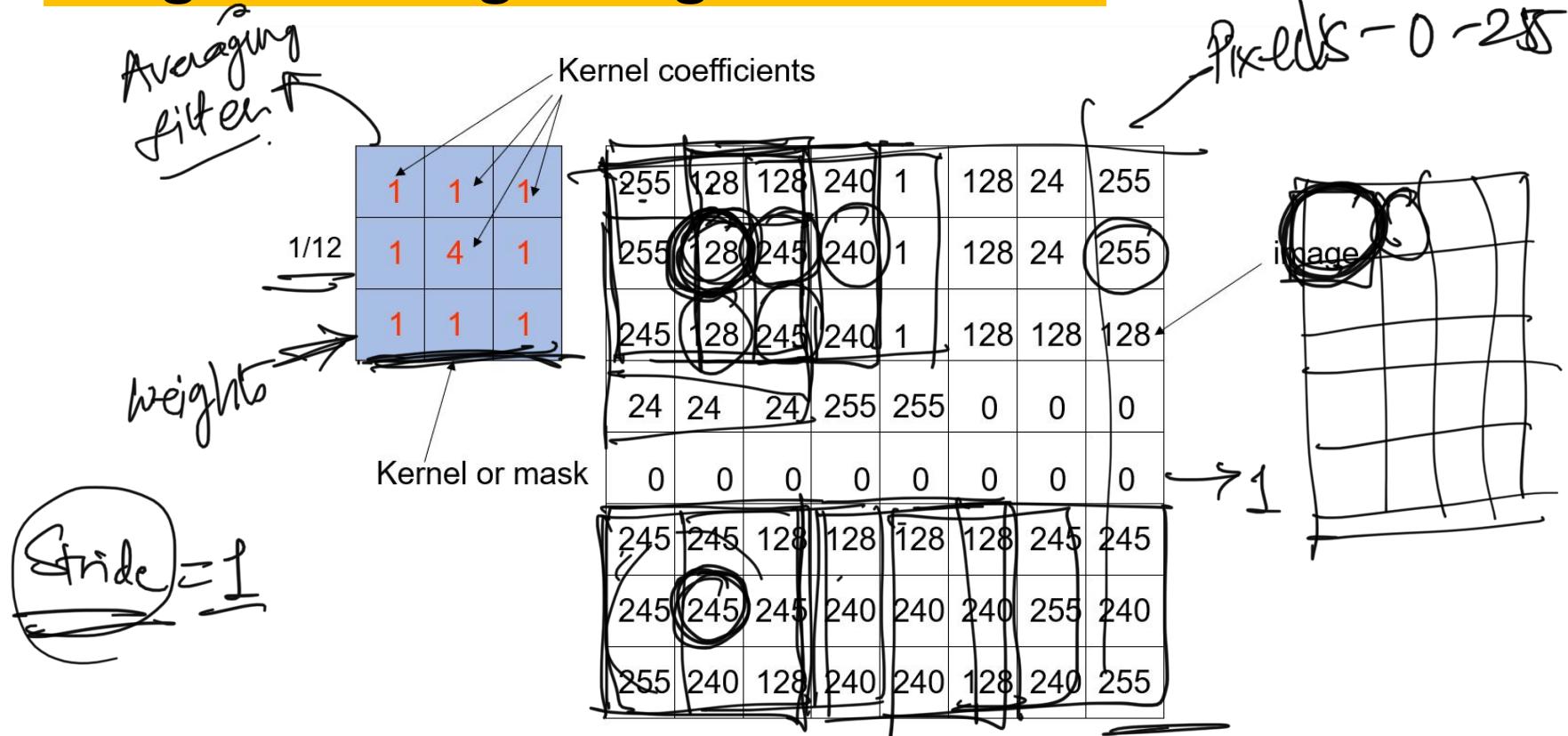
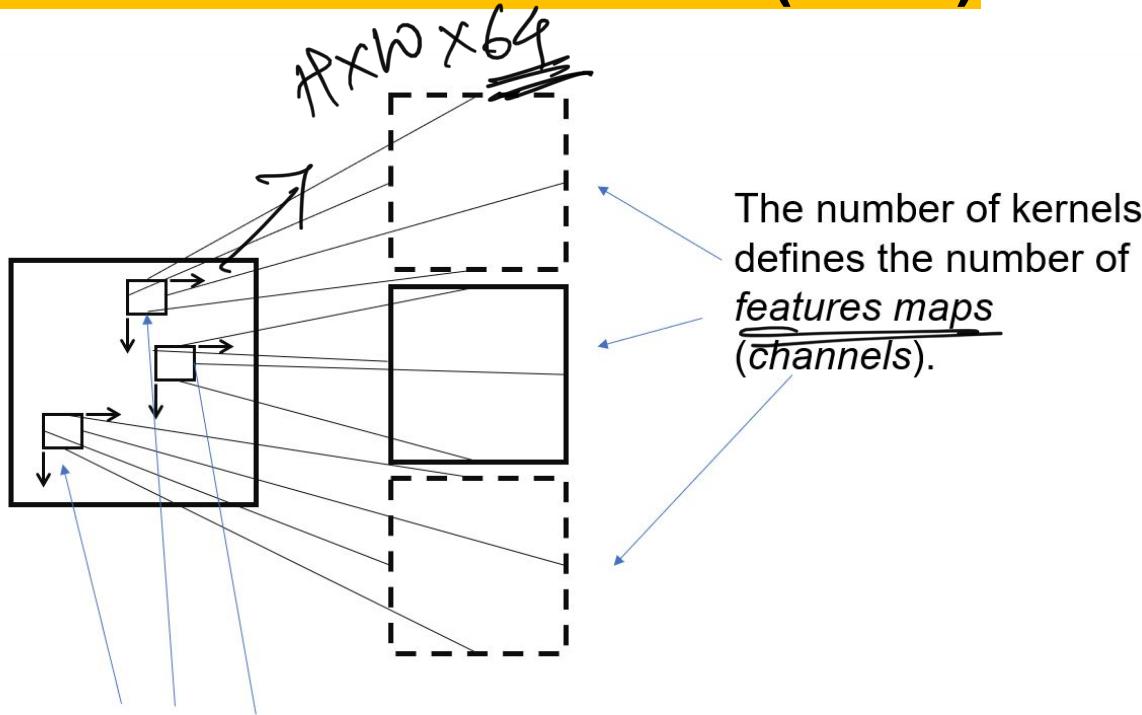


Image Filtering using convolution



Convolutional Neural Network (CNN)



The sizes of the kernels define the *receptive fields*.

Padding in convolutional layer

- Zero-padding

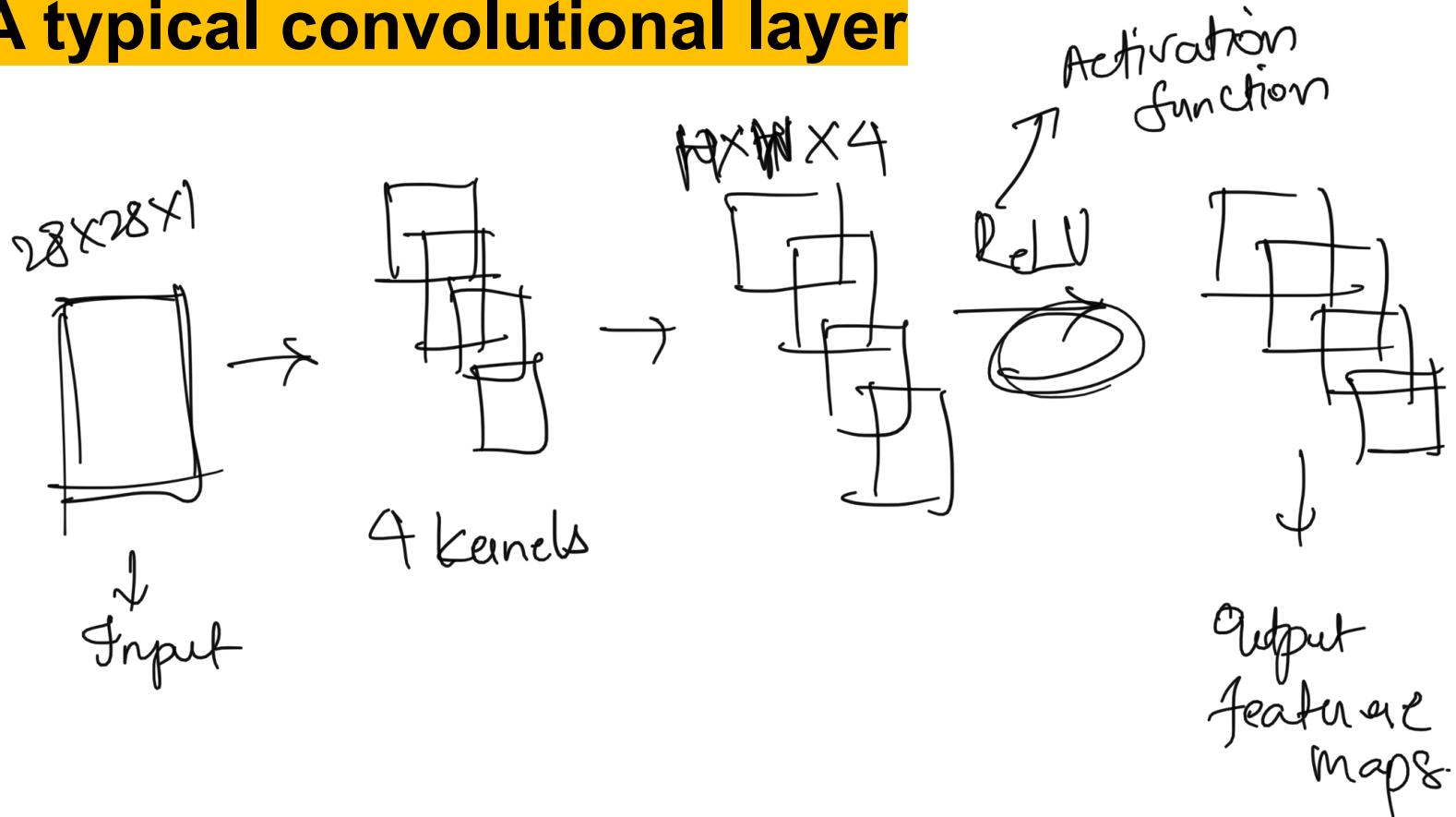
Equally value the boundary pixels.

- Same / less reduced

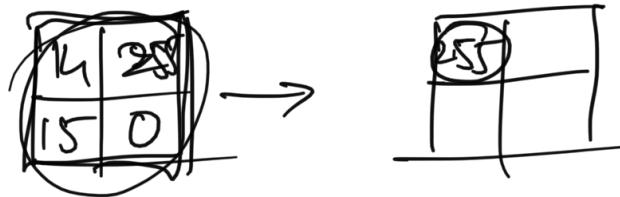
size of the output.

0	0	0	0	0	0	0	0	0	0
0	255	128	128	240	1	128	24	255	0
0	255	128	245	240	1	128	24	255	0
0	245	128	245	240	1	128	128	128	0
0	24	24	24	255	255	0	0	0	0
0	0	0	0	0	0	0	0	0	0
/	245	245	128	128	128	128	245	245	0
/	245	245	245	240	240	240	255	240	0
/	255	240	128	240	240	128	240	255	0

A typical convolutional layer



Pooling in CNN

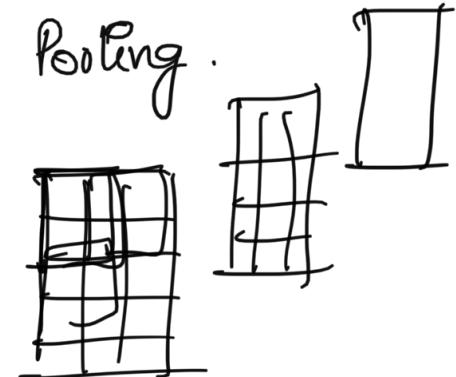


- Used to reduce the size of the output, thereby reducing the number of parameters and preventing overfitting.
- Done on every channel along the height and width of the output. So, the number of channels remain the same.

— Max Pooling

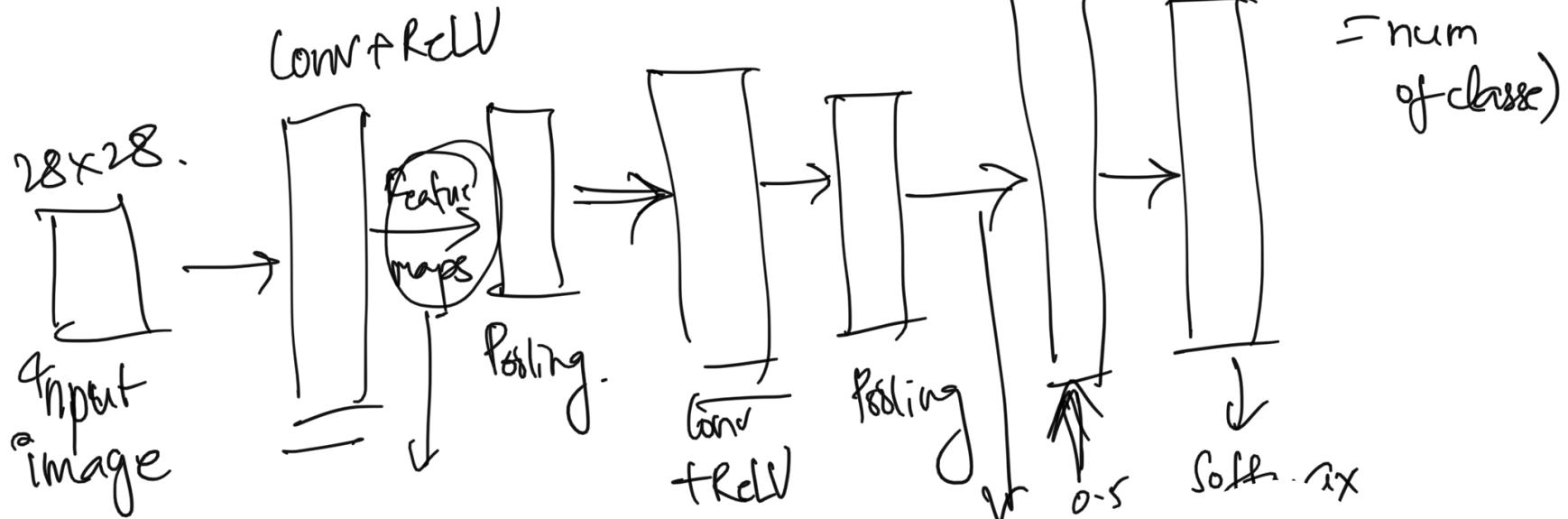
— Average Pooling .

— Kernel size , type of pooling , stride .



$$28 \times 28 \Rightarrow 784$$

A typical CNN



$$224 \times 224$$

$$299 \times 299$$

Batch size \times ~~$B \times H \times W \times C$~~

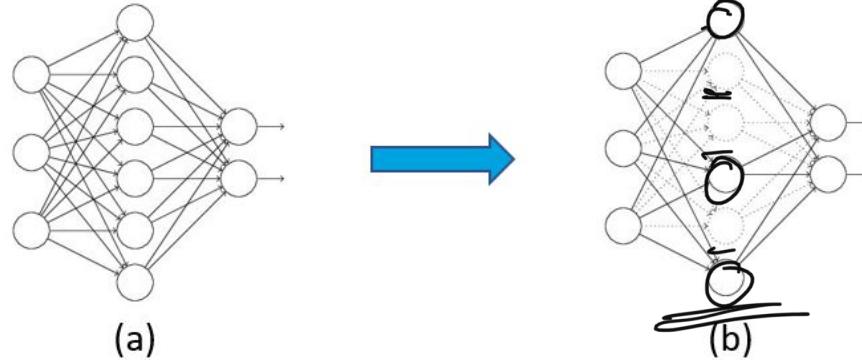
Why is regularization necessary?

- Huge parameter space - poor solutions
- Convergence to local minimum which leads to good performance on training data only.
- To overcome -
 - L2 norm of weights
 - Dropout
 - Batch normalization .

Dropout

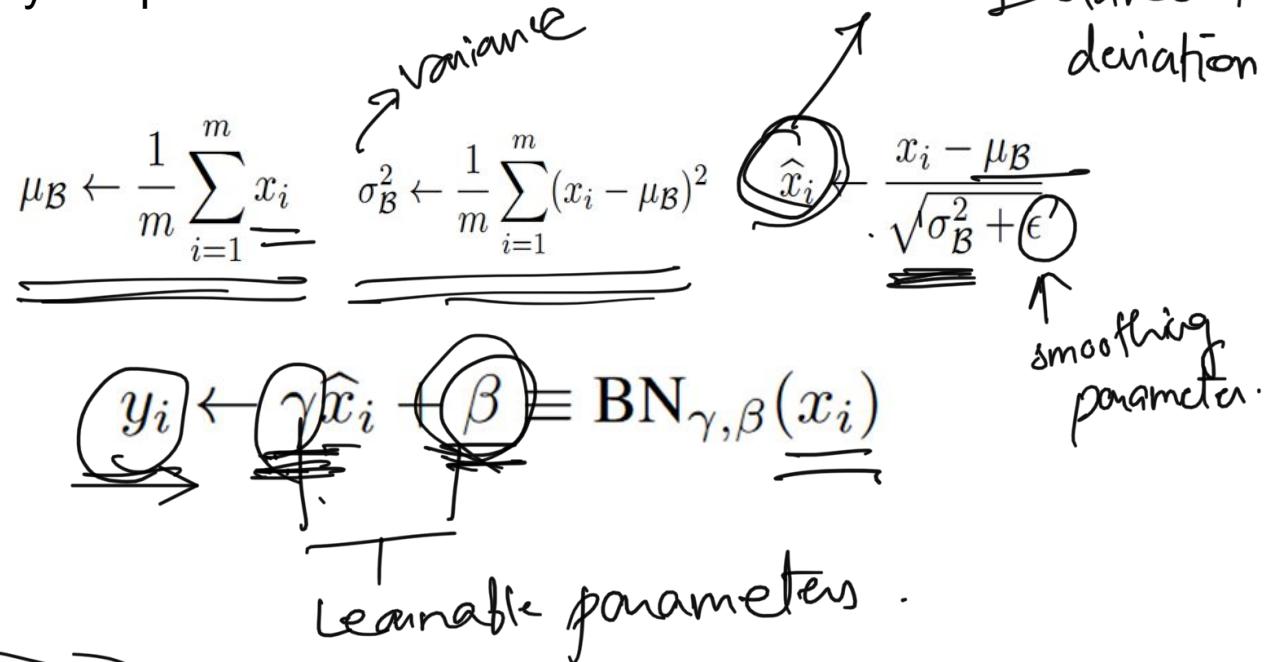
- Obtain (b) by randomly deactivate some hidden nodes in (a).
- Reducing co-adaptation of neurons

Probability of dropping a neuron. $(0 - 1) \xrightarrow{\underline{p}}$



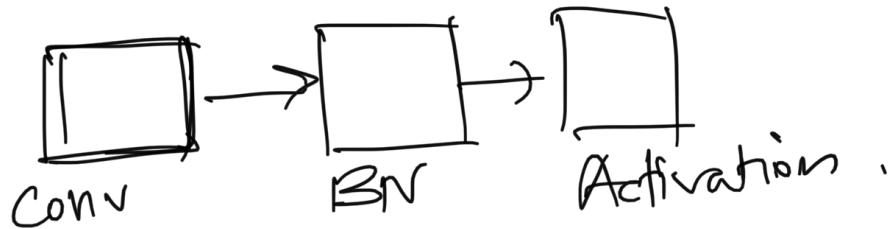
Batch Normalization

- Normalizes layer inputs of a batch



Batch Normalization - Advantages

- Use of higher learning rates
- Covariate shift is reduced
- Avoids vanishing gradients
- Makes the network less sensitive to weight initialization.



Questions?