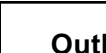




ECE 599/692 – Deep Learning
Lecture 3 – Convolutional Neural Network (CNN)



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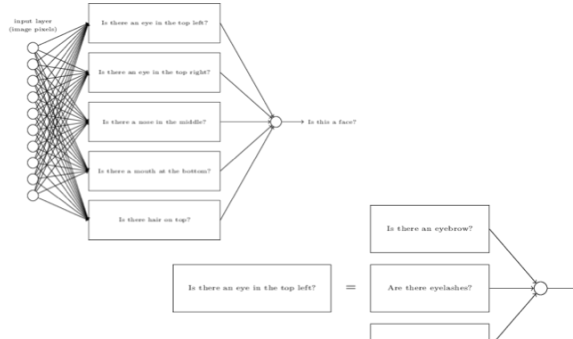
Outline


- Lecture 3: Core ideas of CNN
 - Receptive field
 - Pooling
 - Shared weight
 - Derivation of BP in CNN
- Lecture 4: Practical issues
 - Normalized input and initialization of hyperparameters
 - Cross validation
 - Momentum
 - Learning rate
 - Activation functions
 - Pooling strategies
 - Regularization
- Lecture 5: Variants of CNN
 - From LeNet to AlexNet to GoogleNet to VGG to ResNet
- Lecture 6: Implementation
- Lecture 7: Applications of CNN – Binary hashing
- Lecture 8: Applications of CNN – Person re-identification


2





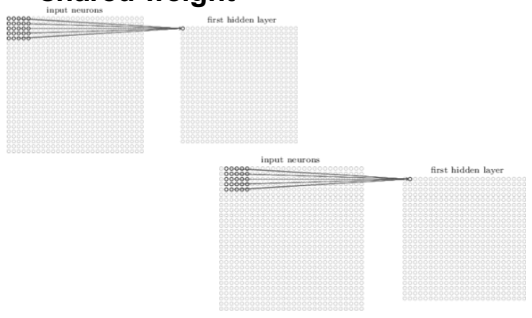
From NN to deep NN





3


Receptive field (RF) and shared weight

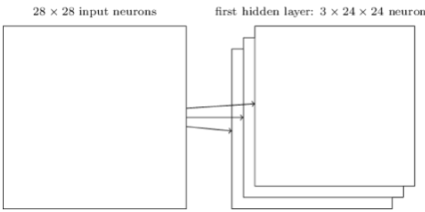






4


Feature maps

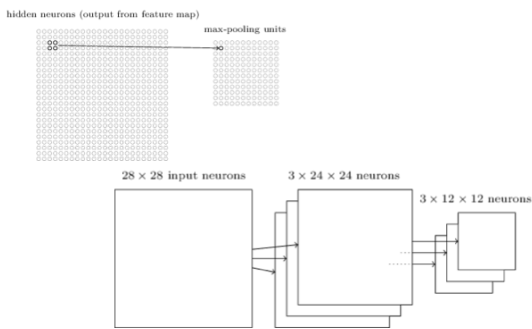






5

Max pooling



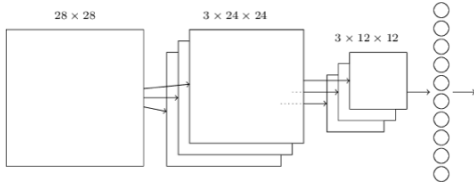



6

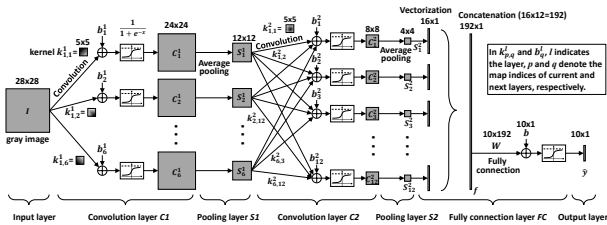
Output of max pooling invariant to shifts in the inputs

- Not very clear (from Le)
- Translational invariant systems
- Major source of distortions in natural data is typically translation

A simple CNN framework



A more practical framework



Acknowledgement

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All figures except the last one are from [Nielsen]
