

INT307 Multimedia Security System

Information Retrieval with Deep Learning

Shengchen Li

Xi'an Jiaotong-Liverpool University

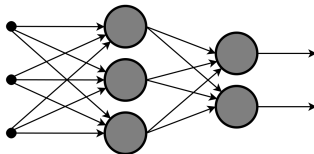
07th Oct 2022

Aims

- Master the working principle of deep learning
- Understand basic knowledge related to deep learning
- Master the framework of multimedia information retrieval via machine learning

Recall INT104

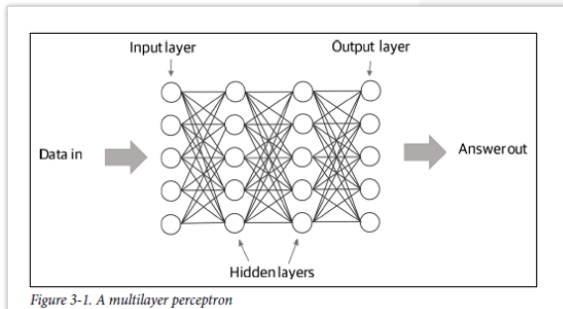
- The boundaries between classes are not necessary linear but can be approximate as a combination of linear functions.



- Could be single layer or multiple layer
- There is a threshold process after the output of each neuron, which is named as activation function

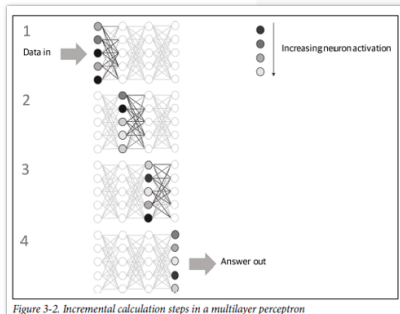
Artificial Neural Networks

- Data
 - Input Layer
 - Hidden Layer
 - Output Layer
-
- Feature Extraction
 - Classification



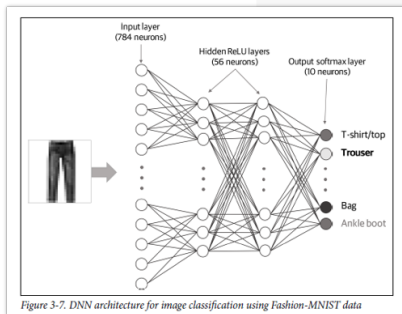
Forward Propagation

- Neurons effectively represent a mapping between feature spaces
- In neural networks, the mapping is represented as weighted sums with activation functions



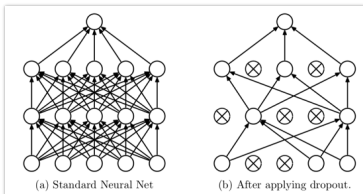
Forward Propagation

- Diagram 28×28
- 784 input neurons
- Two hidden layers with 56 neurons each
- RELU as activation functions



Common Tricks

■ Dropout



■ Batch Normalisation

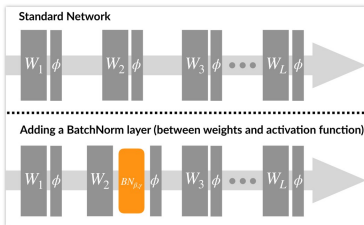


Image Processing with Deep Learning

- Scene classification
- Object detection and localisation
- Semantic segmentation
- Facial recognition



Figure 4-2. An example of image segmentation¹

Recall

How images are presented by computer systems?

Filter and Convolution

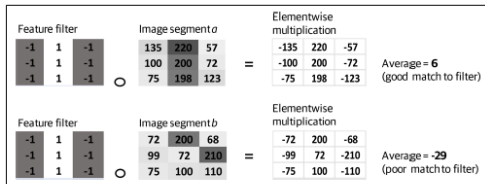


Figure 4-4. Application of a simple 3 x 3 filter to two different image segments

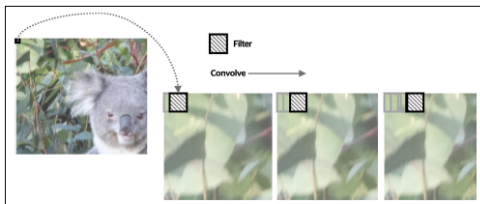


Figure 4-3. A convolutional filter is applied iteratively across an image

Convolutional Layers

- Kernel
 - Size
 - Padding
 - Hop
- Feature Maps

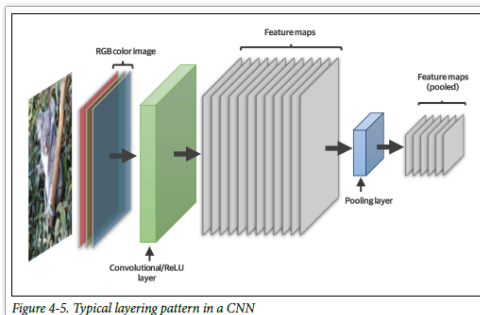


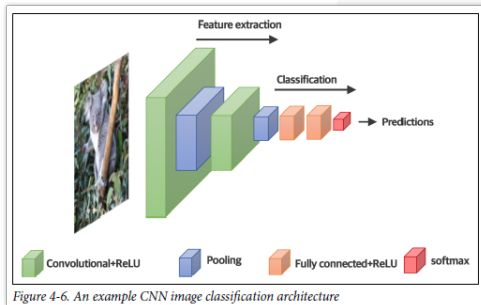
Figure 4-5. Typical layering pattern in a CNN

Question

- 1 Why convolution?
- 2 Why multiple feature maps?

Convolutional Neural Network

- Convolutional Layers
- Pooling Layers
- Fully Connected Layers
- Classifier



Common Image Processing Networks

■ VGG

- VGG-16
- VGG-19

■ Inception

- 22 Layers
- With many variations

Recall: Audio Representation

- 1 Waveform
- 2 Sampling
- 3 Quantisation
- 4 Linear Transform*

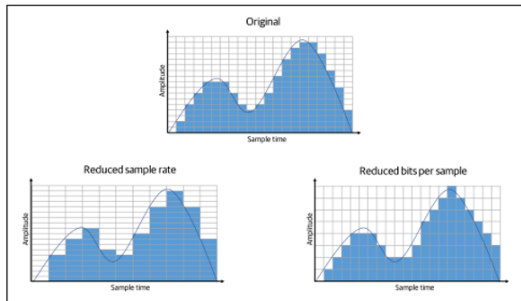


Figure 4-8. The effect of reduced sample rate and bits per sample in digital audio

Time – Frequency Transform

- Usually waveform is transformed to time-frequency domain before being processed
- Commonly used time-frequency transforms are:
 - DFT (FFT)
 - DCT
 - Mel-Spectrogram (MFCC)
 - Wavelet Transform

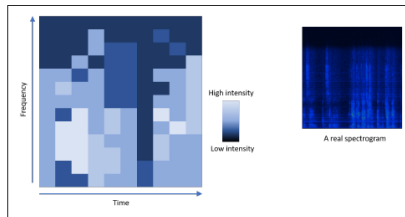


Figure 4-9. A spectrogram depicts changing intensities at different frequencies over time

Recurrent Neural Network

- Recurrent Neural Network is commonly used to process sequential media
- RNN features time slice analysis
- Commonly used time-frequency transforms are:
 - LSTM (Long Short Time Memory)
 - GRU (Gated Recurrent Unit)

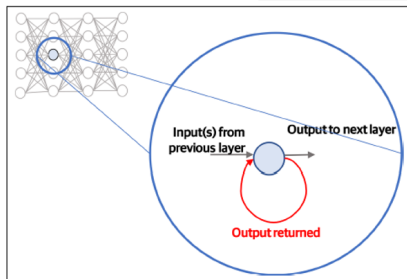


Figure 4-10. Basic concept underpinning RNN architectures

Typical Processing Chain for Audio

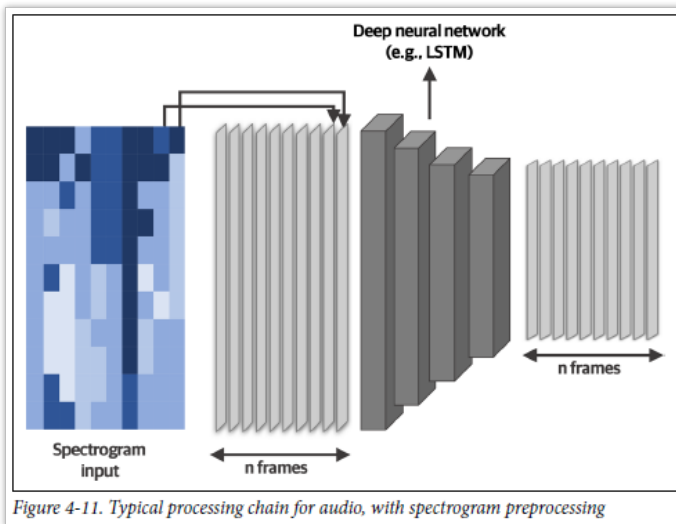
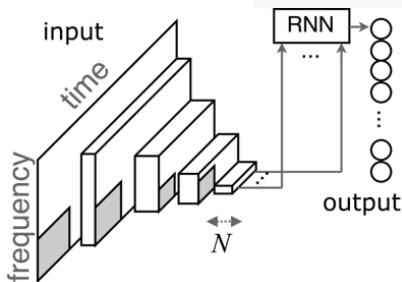


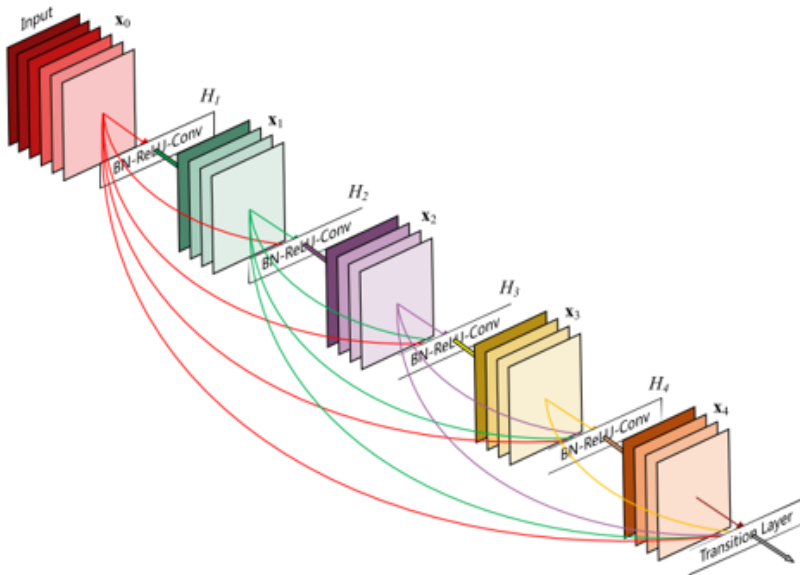
Figure 4-11. Typical processing chain for audio, with spectrogram preprocessing

Complex Networks

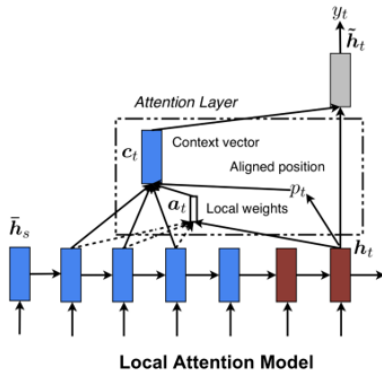
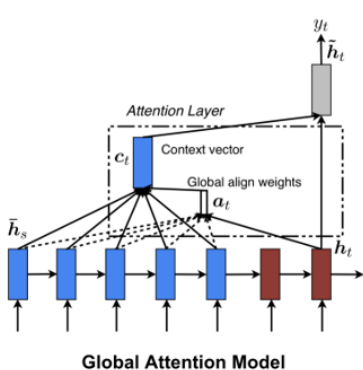
- A deep learning neural network can combine multiple types of structures
 - CNN = CNN + DNN
 - CRNN = CNN + RNN + DNN
- Discussion: Why CRNN can be considered as a way to analyse audio in multi-scale?



Residue Network



Attention



Transformer

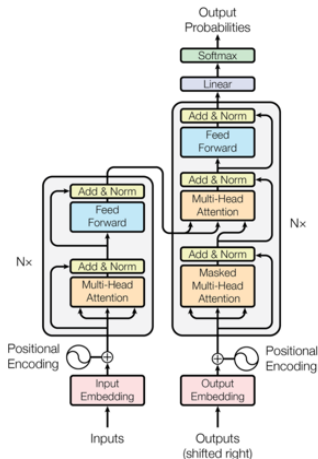
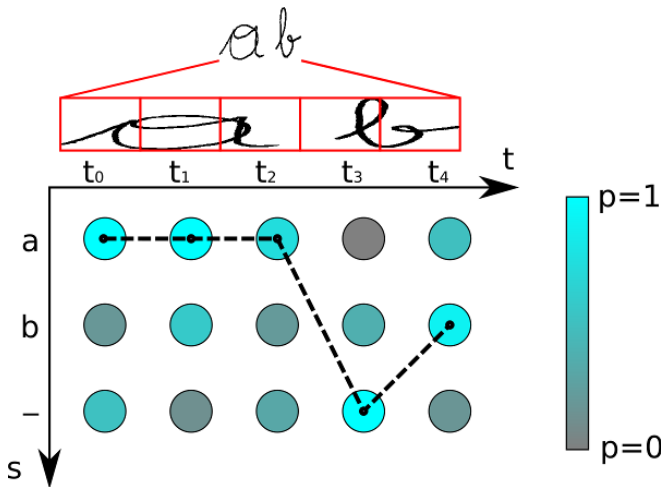
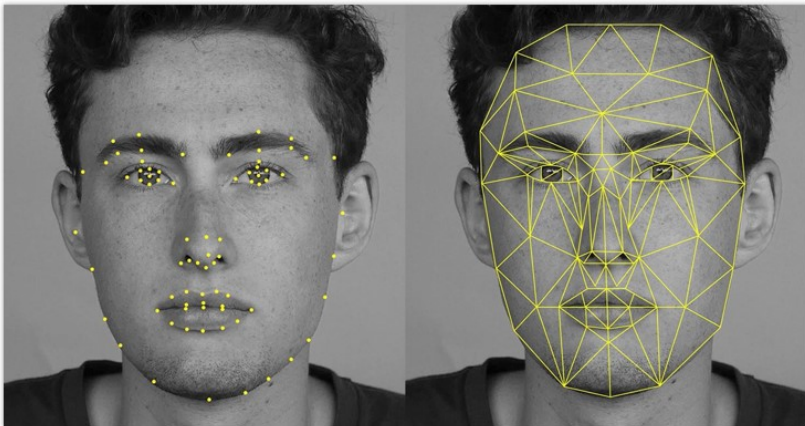


Figure 1: The Transformer - model architecture.

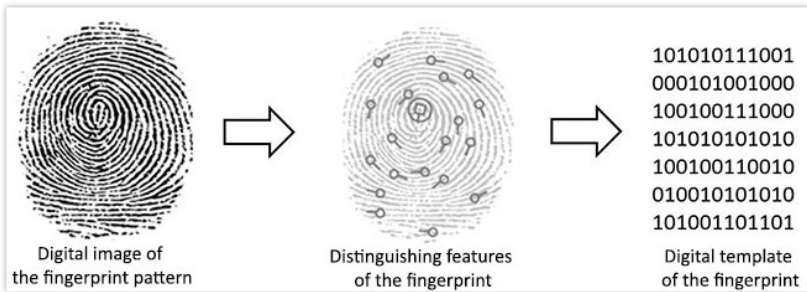
Connectionist Temporal Classification



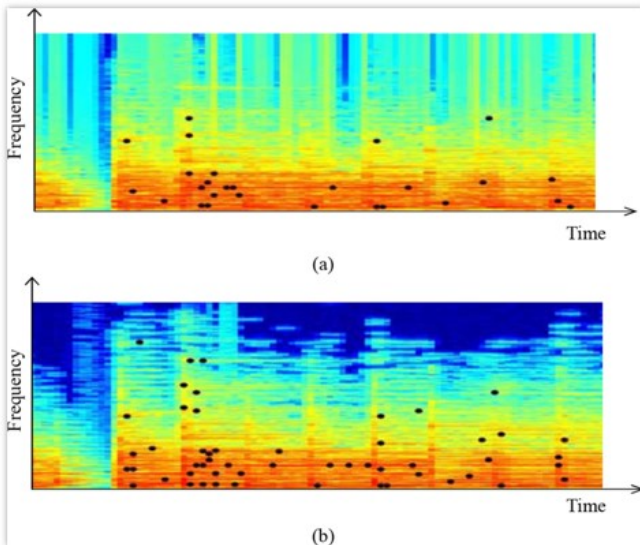
Face Recognition



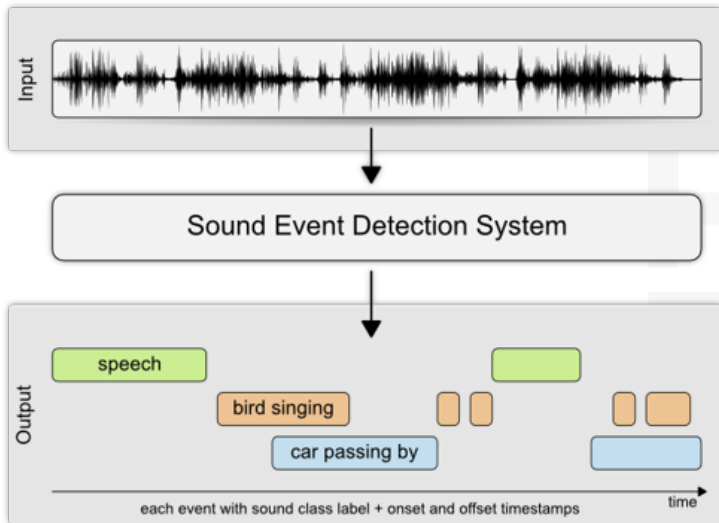
Fingerprint Recognition



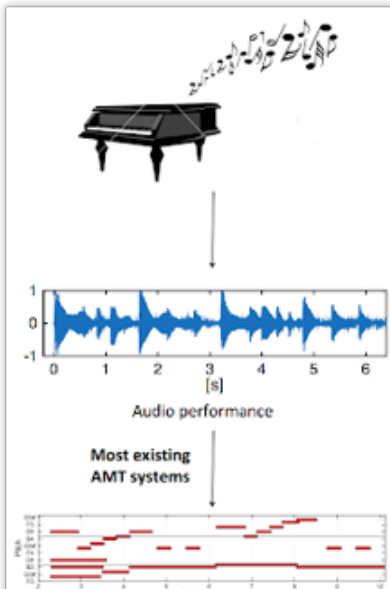
Audio Fingerprint



Audio Event Detection



Audio Event Detection



Translation

