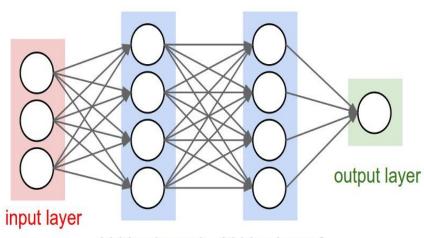
Machine-learning system tackles speech and object recognition, all at once

Devyan Biswas and Rishab Sukumar

General Overview

- Correlate segments of spoken caption with objects within an associated image.
 - Highlighting relevant regions of images in real time
- Demo: https://vimeo.com/290377425

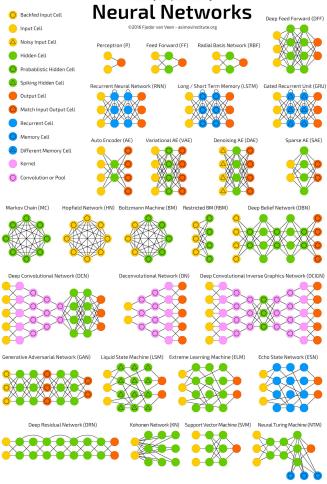
Object Recognition Today: What are CNN's?



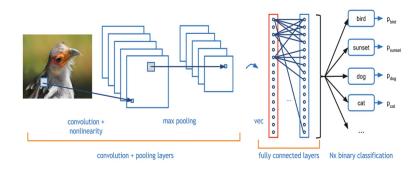
hidden layer 1 hidden layer 2

- Convolutional Neural Networks
 - So...what are neural networks?
- Fundamentals: Linear Algebra, Calculus (depending on model)
- Neural networks are models based on the human brain, have nodes that act like neurons
 - Like how humans learn: training and association
 - Interconnected nodes in layers, feed forward;
 not necessarily 1:1
 - Uses weighting with incoming connections to get value; sends or doesn't send data.
 - Weighting is adjusted as it is training

A mostly complete chart of



What are CNN's? (cont...)

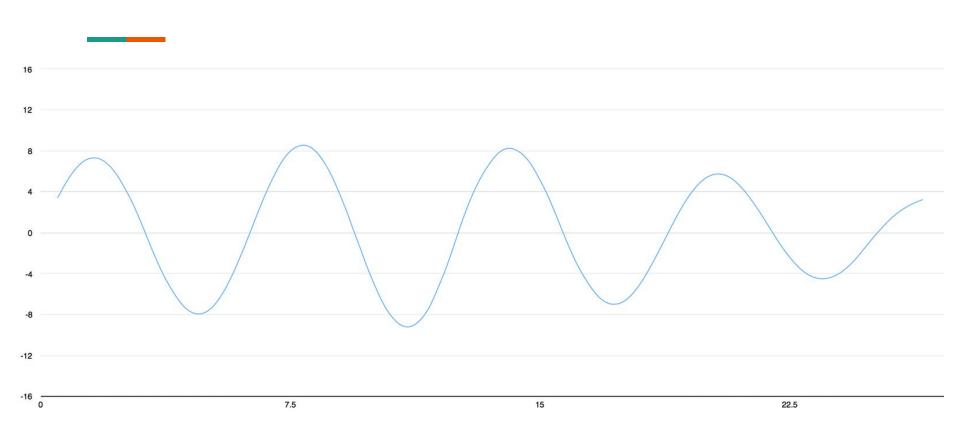


https://adeshpande3.github.io/A-Beginner's-Guide-To-Understanding-Convolutional-Neural-Networks/

- Used in image processing and object recognition
- CNN's are useful in 2 regards
 - Reduce total units (generally) because they translate many:1, so quicker learning (generally)
 - Contextual
- What does "convolution" mean?
 - Using a filter (feature identifier/neuron), convolving over input image
 - Performs element-wise multiplication
 - Sums and returns to one point
 - Gives you activation/feature map
- Last output: Fully Connected Layer
- Training with backpropagation

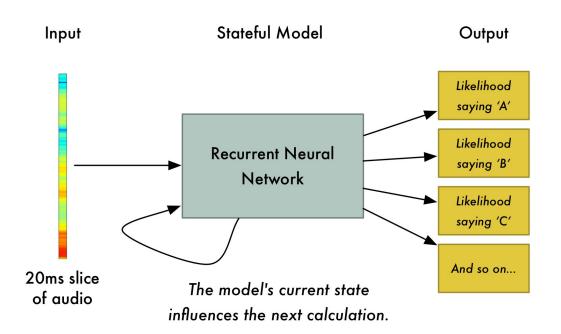
Current Speech-Recognition Technologies

- Require manual transcriptions and annotations of training set images.
- Systems such as Siri require transcriptions of many thousands of hours of speech recordings.
- Uses spectrograms of audio recordings.
- The spectrogram is converted into a series of values based on the height of the sound wave at fixed intervals.



Current Speech-Recognition Technologies (Cont...)

- Uses short samples of spectrograms as input for a Recurrent Neural Network (i.e. a neural network that has a memory that influences future predictions.)
- Output of RNN is an N-dimensional vector where each element denotes the likelihood of related classes.



How is MIT's model different?

- Eliminates need for manual transcriptions and annotations for training.
- Focuses on a more natural method of learning where knowledge of written text is not necessary.
- Training set consists of numerous audio recordings describing the contents of an image.
- Uses two CNN's. One to process images and the other to process spectrograms.
- Matches words to specific patches of pixels in the image.
- The model has **no access** to any true alignment information between the speech and the image.

The MatchMap Concept

- What is the matchmap?
 - Correlates sections of audio data(spectrogram) with image data
 - Able to directly learn what corresponds between speech frame and group of pixels
 - No need for annotated training data
 - "Reversible"; Unpool localized information analysis from pooled representation of input
- Results (from Academic Paper):
 - Speech-Prompted Localization
 - Image-Caption Retrieval
 - Image-Word Dictionary

Possible Future Applications

- Speech-object recognition in videos.
- Learning translations between different languages, without need of a bilingual annotator, possibly in real-time.
- Generate images given a spoken description.
- Generate artificial speech describing a visual scene.

Questions?

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