Embedded Machine Learning

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Objectives

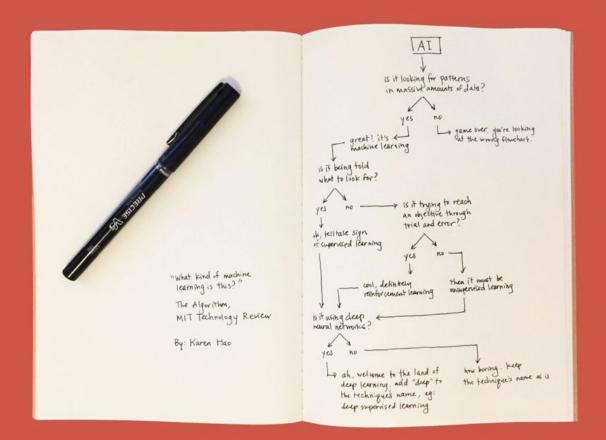
- 1. Introduction to ML on embedded platforms
- 2. Setting up learning machine
- 3. Setting up target platform
- 4. Training an example
- 5. Running the example

Sources for the talk

github.com/knud/CompSciAl202002

Assumptions

- Tensorflow 1.15, but discuss 2.x
- Linux for training
 - Ubuntu in general. Examples are Mint 19.3
 - Nvidia GPU
- Artemis platform, can extend to Raspberry Pi, Arduino, and others

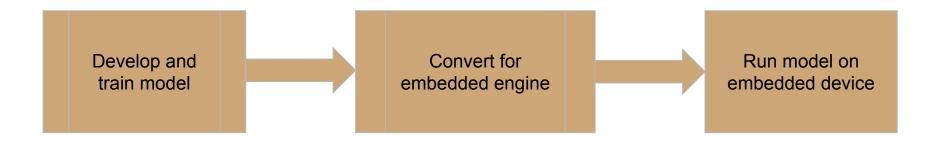


Karen Hao - technologyreview.com

Machine-learning algorithms use statistics to find patterns in massive* amounts of data. And data, here, encompasses a lot of things—numbers, words, images, clicks, what have you. If it can be digitally stored, it can be fed into a machine-learning algorithm.

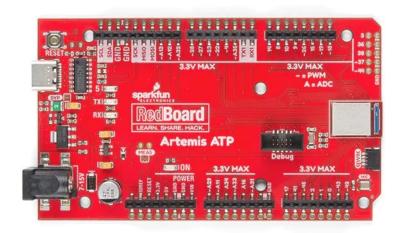
Wake-word Detection

Well-known problem with lots of examples.



Artemis ATP - Apollo3 Blue module

Low Power Machine Learning BLE Cortex-M4F





Ambiq Apollo3 Blue-based module

Low power - 6µA/MHz, RX/TX 3 mA

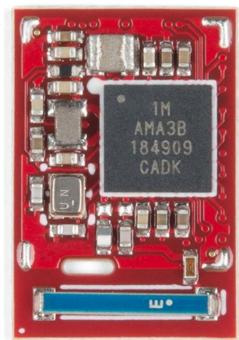
48 MHz, 96 MHz turbo mode

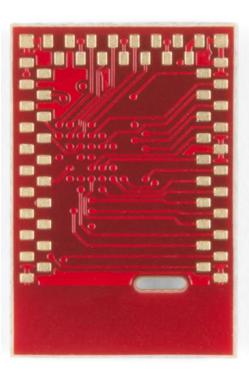
FPU, MPU

32 interrupts

Loads of peripherals, etc.

Loads of GPIO





TensorFlow 1.x Lite for Microcontrollers

Not TensorFlow 2...

Specially designed for very small footprint devices.

- Core is ~ 16 kB
- With enough operators for speech, ~ 22 kB

Leaves lots of room for model and other functions need for your application

https://www.tensorflow.org/lite/microcontrollers

Prep model dev platform; Anaconda, TF1.15, Jupyter, ...

Create or obtain TF model

Convert to TF Lite FlatBuffer

Convert FlatBuffer to C array

Prep MCU dev platform; Arduino IDE

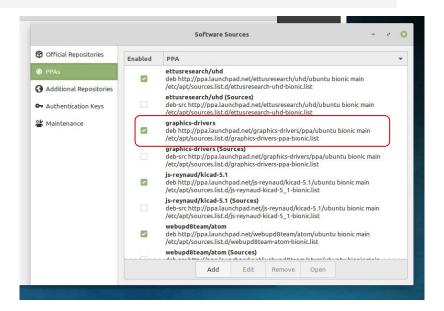
Integrate with TF Lite for MCU C++ library

Deploy to device

Model Development

Enable GPU support

\$sudo add-apt-repository ppa:graphics-drivers/ppa
\$sudo apt-get update



Enable GPU support cont'd

Subuntu-drivers devices

```
knud 20:33:28 $ubuntu-drivers devices
== /sys/devices/pci0000:00/0000:00:1c.0/0000:02:00.0 ==
modalias : pci:v00008086d00003165sv00008086sd00004010bc02sc80i00
vendor : Intel Corporation
model : Wireless 3165 (Dual Band Wireless AC 3165)
manual install: True
driver : backport-iwlwifi-dkms - distro free
== /sys/devices/pci0000:00/0000:00:01.0/0000:01:00.0 ==
modalias : pci:v000010DEd0000139Bsv00001462sd0000115Abc03sc02i00
vendor : NVIDIA Corporation
model : GM107M [GeForce GTX 960M]
driver : nvidia-driver-410 - third-party free
driver : nvidia-driver-435 - distro non-free
driver : nvidia-driver-440 - third-party free recommended
driver : nvidia-driver-430 - third-party free
driver : nvidia-driver-415 - third-party free
driver : nvidia-driver-390 - third-party free
driver : xserver-xorg-video-nouveau - distro free builtin
knud 21:57:20 $
```

\$sudo apt install nvidia-driver-440

Enable GPU support cont'd

Reboot

\$nvidia-smi

```
knud 10:01:34 $nvidia-smi
Tue Feb 4 10:01:37 2020
                         Driver Version: 440,48.02
                                                       CUDA Version: 10.2
 NVIDIA-SMI 440.48.02
                  Persistence-M| Bus-Id
                                                Disp.A |
                                                         Volatile Uncorr. ECC
 GPU
      Name
      Temp Perf Pwr:Usage/Capl
                                          Memory-Usage
                                                         GPU-Util Compute M.
      GeForce GTX 960M
                           0ff
                                  00000000:01:00.0 Off
                                                                          N/A
                                                              0%
                                                                      Default
 N/A
                    N/A / N/A
                                    211MiB / 2004MiB
                                                                   GPU Memory
 Processes:
  GPU
             PID
                   Type
                                                                   Usage
                          Process name
            3951
                          /usr/lib/xorg/Xorg
                                                                       147MiB
     Θ
                          ...quest-channel-token=3152215896715755542
            7664
                                                                        60MiB
knud 10:01:37 $
```

Anaconda Navigator, Jupyter

- For experiments and training
 - https://docs.anaconda.com/anaconda/install/linux/
 - https://docs.anaconda.com/anaconda/user-guide/tasks/tensorflow/
 - conda create -n tf15-gpu tensorflow-gpu=1.15
 - conda activate tf15-gpu
 - conda install anaconda-navigator
 - https://garywoodfine.com/set-up-anaconda-jupyter-notebook-tensorflow-for-deep-learning/



Model and training for microcontroller

- Tensorflow for Microcontrollers
 - https://www.tensorflow.org/lite/microcontrollers
 - Arduino Nano 33 BLE Sense
 - SparkFun Edge
 - STM32F746 Discovery kit
 - Adafruit EdgeBadge
 - Adafruit TensorFlow Lite for Microcontrollers Kit
 - Adafruit Circuit Playground Bluefruit
 - Espressif ESP32-DevKitC
 - Espressif ESP-EYE
 - https://github.com/tensorflow/tensorflow/tree/master/tensorflow/lite

Things to keep in mind...

- Microcontrollers have limited resources
 - Code space (FLASH)
 - RAM
 - Computational power
 - Other things may be running... e.g., Bluetooth stack
- Cannot process large data samples
 - For speech, keep the amount of data low by processing short time intervals
 - 1 second snippets typically used

May think of the microcontroller as a preprocessor for some applications.

Wake word training

- Based on the micro speech example
 - https://github.com/tensorflow/tensorflow/tree/master/tensorflow/lite/micro/examples
- The Jupyter notebook, train_speech_model.ipynb, is the starting point, but is set up to do too much if you already have things downloaded
 - Assumes access to root-level folders, so moved them
- Can select one or more words to train. Example shipped is "yes" & "no"
- Changed to train for "stop"

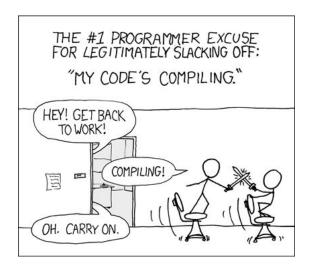
Training for "stop"

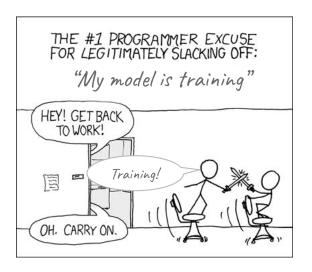
- In the Jupyter notebook CompSciAITF115 provided in github.com/knud/CompSciAI202002, change the words to train near the top from "yes" "no" to "stop"
- Run the notebook through to the end

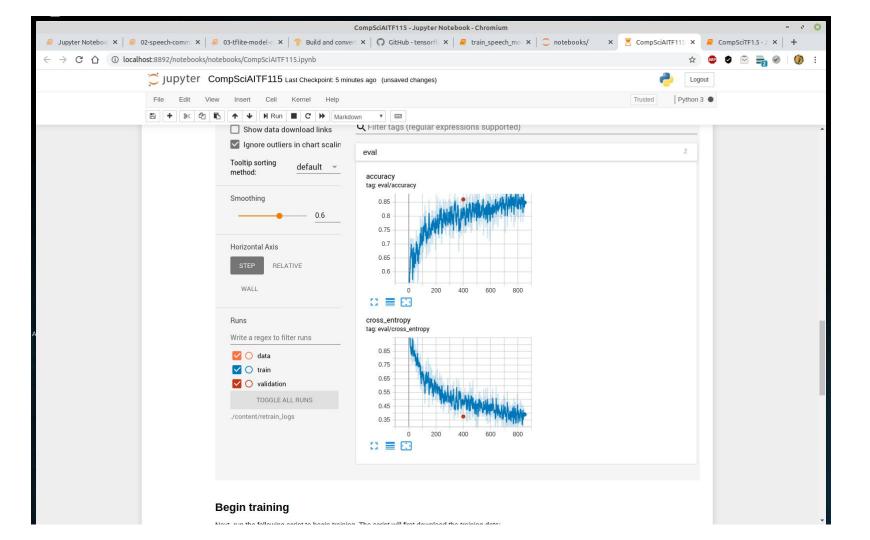
The main result at the end is a TensorFlow Lite model that has been converted into a C array suitable for deployment.

tiny_conv.cc

Training...







Modify the Artemis source for "stop"

- Open the micro speech sketch in the Arduino IDE
- Copy the C array for the trained model from tiny_conv.cc into
 micro_features_tiny_conv_micro_features_model_data.cpp,
 replacing the data in the array
 g_tiny_conv_micro_features_model_data.
- Just past the array, set the size to that shown at the bottom of tiny conv.cc.

Training for "stop" cont'd

- In micro_features_micro_model_settings.h, change kCategoryCount to the number of words + 2 (e.g., 3 for just "stop")
- In micro_features_micro_model_settings.cpp, change the words to recognize (e.g., replace "yes" and "no" by "stop")
- Compile and upload to the Artemis board
- Test!!!

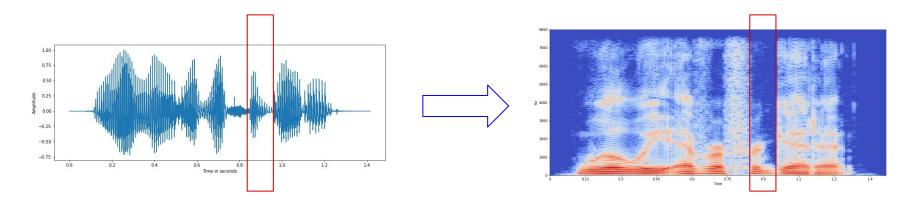
Training for "stop" optional

- In arduino_command_responder.cpp, can edit the RespondToCommand method to
 - Look for the reported word "stop"
 - Do something other than turn on an LED. E.g., turn off a motor, gather sensor data, etc.

So what is actually happening?

Speech as "image recognition"

Process 1-dimensional speech signal to get a 2-dimensional representation, aka images, that can be used for CNN network training



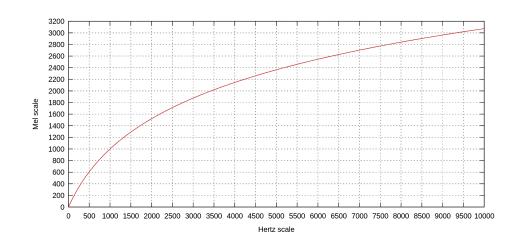
from https://towardsdatascience.com/beginners-guide-to-speech-analysis-4690ca7a7c05

Mel Frequency Cepstral Coefficients (MFCC)

Found that some processing is needed to aid recognition.

Mel Scale - perceptual scale of pitches judged to be equidistant

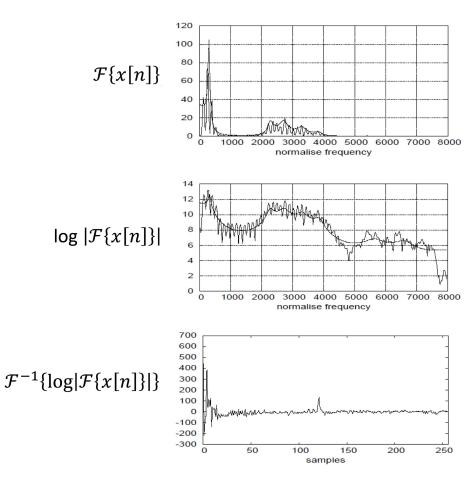
$$m = 2595 \log_{10} \left(1 + \frac{f}{700} \right)$$



MFCC cont'd

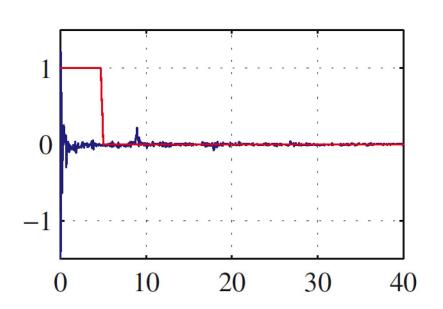
Cepstral comes from cepstrum, which is defined as the inverse DFT of the log magnitude of the DFT of a signal

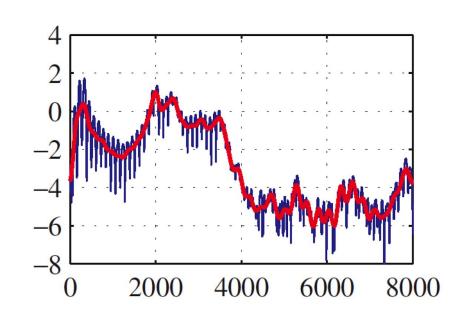
$$c[n] = IDFT \{log_{10} | DFT \{x[n]\}|\}$$



[Taylor, 2009]

<u>Lift</u>ering in the <u>ceps</u>tral domain





[Rabiner & Schafer, 2007]

MFCC cont'd

Main steps

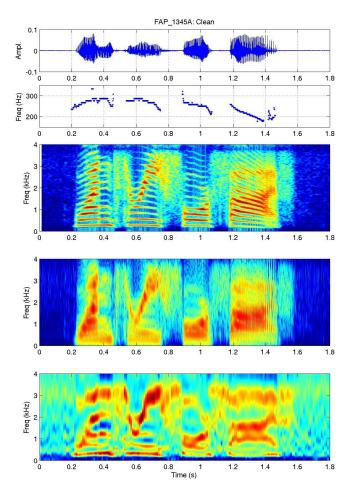
- Divide signal into "frames"
- 2. Calculate power spectrum using DFT
- Map to Mel frequencies (using a filter bank)
- 4. Compute the log of the spectral coefficients
- 5. Computer the discrete cosine transform (DCT)

DCT helps decorrelate spectral coefficients and allows pruning of high Mel frequency values that turn out not to help with recognition.

from http://practicalcryptography.com/miscellaneous/machine-learning/guide-mel-frequency-cepstral-coefficients-mfccs

MFCC cont'd

MFCC-based waveform and spectrogram for the utterance "one-three-four-five" by a female speaker.



from http://personal.ee.surrey.ac.uk/Personal/P.Jackson/eem.ssr/lab2.html

More details for the interested reader...

The math behind all this is nicely explained here

https://haythamfayek.com/2016/04/21/speech-processing-for-machine-learning.html

...with wee prizes...

Who created the first artificial neural network and when?

- Warren McCulloch and Walter Pitts
- 1943
- Perceptron
- Modelled using electrical circuits

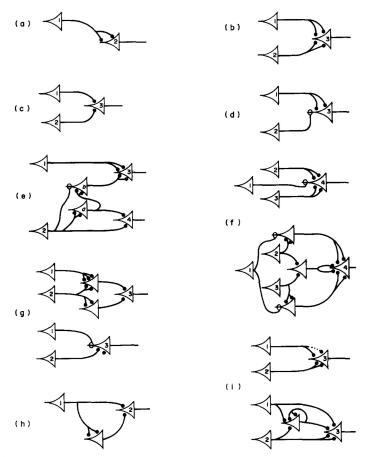


Figure 1. The neuron c_i is always marked with the numeral i upon the body of the cell, and the corresponding action is denoted by "N" with i s subscript, as in the text:

Who is the "father" of Java and where was he born?

What is an equivalent to the U of A's main frame computer from the 80s?

Amdahl 470 v/6

Date Introduced: 1975

Dimensions overall: 63" x 70" x 26"

Keywords: Clones; Plug compat; IBM

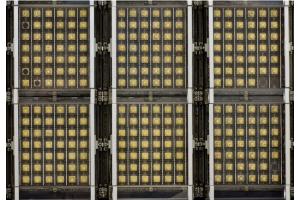
Speed: 3.5 MIPS

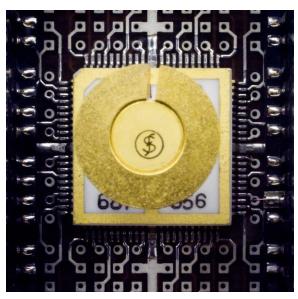
Memory Size: up to 8MB

Memory Width: 32-bit

Cost: \$3,750,000 (2020 \$17,981,250)







Arduino

Date Introduced: 2010

Dimensions overall: 2.7" x 2.1" x 0.6"

Speed: 8 - 11 MIPS

Memory Size: 32k FLASH 2k SRAM

Cost: ~\$20

Cost to make: < \$5



What was the most dreaded language in the 2019 Stackoverflow survey