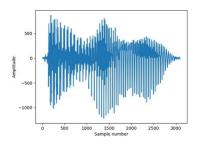
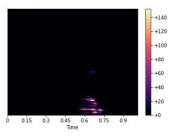
CNN Image Classification For Verbal Commands

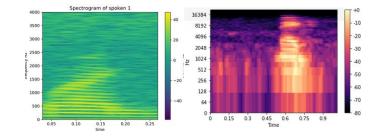
Chris Hickey 논문연구 1

Introduction

Various ways to represent sound visually:







Question?

Can a network be trained to recognise meaning from these images?

Why is this important?

- Provide alternative to RNN classifiers for simple speech recognition
- Leverages advances in computer vision/image recognition in the domain of NLP
- May be particularly useful for command recognition

Dataset: 65,000 1 second utterances of 30 words spoken in noisy environments.

Roughly 2,000 samples for each class

Relatively new dataset (2 years old)



The latest news from Google Al

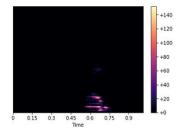
Launching the Speech Commands Dataset

Thursday, August 24, 2017

Results

Which visual representation of data performed best?

Initial experimentation found that the mel-spectrogram gave the image most trainable by NN.



What is a mel-spectrogram?

<u>Spectrogram:</u> Visual representation of a (audio) signal over time.

<u>Mel Scale:</u> A perceptual scale of pitches judged by listeners to be equal in distance. This a non-linear transformation of Hz scale, where the difference between 500 and 1000 Hz is obvious, whereas the difference between 7500 and 8000 Hz is barely noticeable.

Network Configurations

- 5-layer convolutional neural network.
- Batch normalization after 1st layer
- 2 x 2 **Max Pooling** after each layer
- ELU activatation after each layer
- Dropout 0.6 after each layer
- **Softmax** activation for output
- Bath-size: 32
- Loss function: categorical crossentrophy
- Optimizer: adadelta

Results

Data Split

- 65,000 audio files split 70/15/15 for Train/Val/Test
- Classes were represented equally in each data subset

Results (100 epochs)

- Performance on combined Train/Val set: 90% accuracy
- Performance on Test set: 89% accuracy

Real Life Testing

- Tested using samples recorded on my iPhone
- Correct over 90% of time, often with high certainty

```
In [18]: filename = os.path.join('../SelfRecordedSamples', 'bSayingBed.wav')
         print prediction(filename)
                          : 0.95234715938568115234375000000000
                             0.00552873499691486358642578125000
         bird
         cat
                             0.00161573465447872877120971679688
                             0.00029072977486066520214080810547
         down
                             0.00002668317392817698419094085693
         eight
                             0.00104765326250344514846801757812
         five
                             0.00118114787619560956954956054688
         four
                             0.00015520499437116086483001708984
                             0.00082924577873200178146362304688
```

 Mistakes are usually for words that sound similar with high uncertainty

```
In [17]: filename = os.path.join('../SelfRecordedSamples', 'meSayingNo.way')
         print prediction(filename)
                             0.00000228116914513520896434783936
         bird
                             0.00000020492534247296134708449244
         cat
                            0.00000102521119060838827863335609
                             0.00722104031592607498168945312506
                             0.01036341022700071334838867187500
                            0.00000125785084037488559260964394
                            0.00000002493662876190683164168149
                           : 0.00000288039177576138172298669815
                            0.51891952753067016601562500000000
                            0.00000010753013413022927124984562
                            0.00228480435907840728759765625006
                            0.00000020597867944616155000403523
                             0.00000003590669095387966081034392
                            0.00002505920383555348962545394897
                           : 0.46082231402397155761718750000000
```