Machine Learning and Having It Deep and Structured HOMEWORK #1

Training Deep Neural Network 呂相弘 吳彥諶 r03942039@ntu.edu.tw b99901106@ntu.edu.tw

Outline

- DNN in Speech Recognition
- DNN
- Kaggle
- Dataset and Format
- Submission Requirements
- Grading
- GPU Acceleration
- Tips
- Baselines
- 語音導覽(103-2)

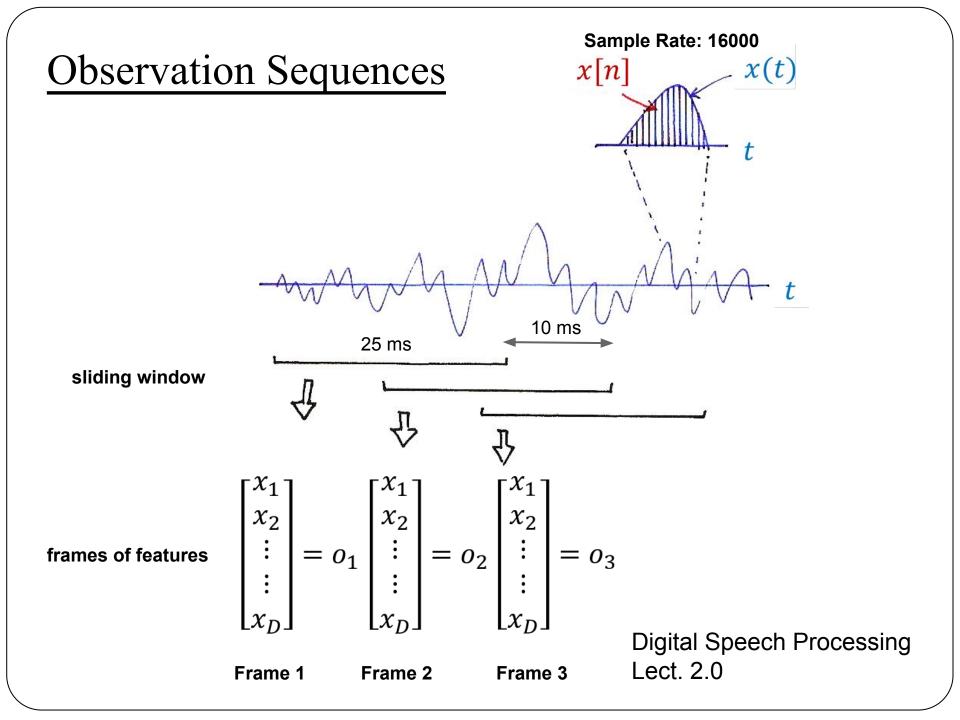
DNN IN SPEECH RECOGNITION

Speech Recognition

- In speech processing...
 - each word consists of syllables
 - each syllable consists of phonemes

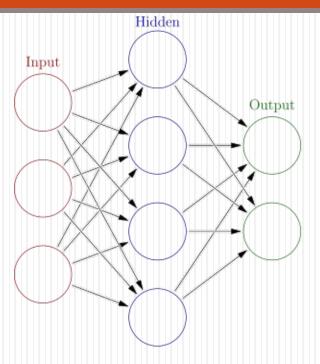
```
"青色" \rightarrow "青(く | \angle )色(ムさ、)" \rightarrow "く" (syllables) 青: TSI --I \rightarrow N (phonemes) 色: S--@ (phonemes)
```

 Each time frame, with an observance (vector) mapped to a phoneme.

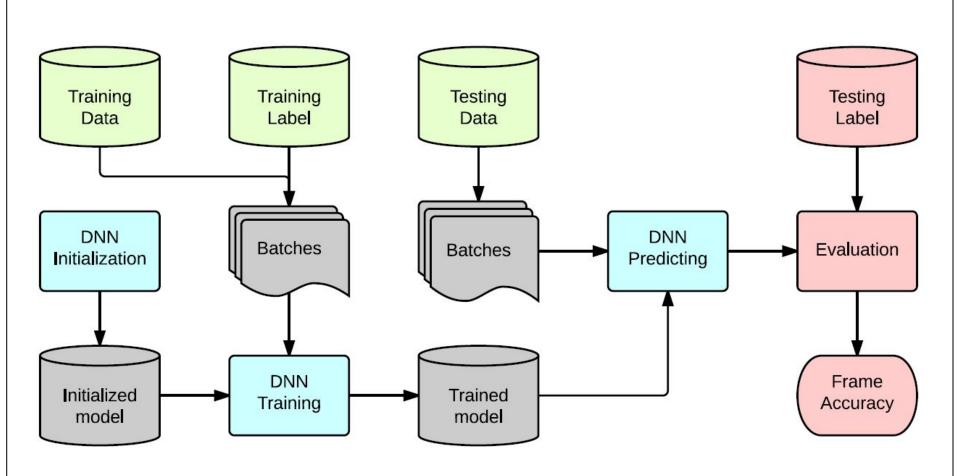


DNN in Speech Recognition

- Goal: predict phoneme given feature in each time frame.
 - Frame-wise prediction
- Input: acoustic features
 - Raw waveform, MFCC, FBANK or...
- Output: pronunciation units
 - Phonemes or...
- To know more about Automatic Speech Recognition(ASR), please refer to <u>DSP</u>



Homework 1: Training Deep Neural Network



Main Problems in HW1

- Model initialize
- Feedforward
- Backpropagate
- Update
- Predict

Model Initialize

- DNN sometimes fails at local optimum problem, so **initialization matters**.
- Practically, there exists unsupervised pretraining technique on initialization.

Local Maximum

 However, in this homework, we recommend you initialize them randomly for the simplicity and efficiency.

Feedforward

 \mathbf{W}^1, b^1

$$y = f(x) = \sigma(\mathbf{W}^L \dots \sigma(\mathbf{W}^2 \sigma(\mathbf{W}^1 x + b^1) + b^1) \dots + b^L)$$

$$\sigma(\mathbf{W}^L a^{L-1} + b^L) = a^L = y$$
Input Layer 1 Layer 2 Layer L Output
$$x_1 \xrightarrow{x_1} \xrightarrow{x_2} \xrightarrow{a^1} \xrightarrow{a^2} \dots \xrightarrow{y_2} \text{vector}$$

$$\mathbf{y}$$
vector
$$\mathbf{x}$$

 $\mathbf{W}^{\mathbf{L}}, b^{\mathbf{L}}$

 W^2, b^2

Backpropagate

Layer L
Layer L-1 (Output layer)

$$\frac{\partial C^r}{\partial w_{ij}^L} = \frac{\partial C^r}{\partial y_i^r} \frac{\partial y_i^r}{\partial z_i^L} \frac{\partial z_i^L}{\partial w_{ij}^L}$$

$$= a_j^{L-1} \sigma' \left(z_i^L \right) \left(y_i^r - \hat{y}_i^r \right)$$

Example: (x^r, \hat{y}^r)

$$Compute \ \nabla C(\theta^0)$$

$$Update$$

$$\theta^1 = \theta^0 - \eta \nabla C(\theta^0)$$

$$Compute \ \nabla C(\theta^1)$$

$$Starting$$

$$\theta^2 = \theta^1 - \eta \nabla C(\theta^1)$$

$$Parameters$$

$$Compute \ \nabla C(\theta^1)$$

η is called "*learning rate*" η should be small enough, but should not be too small.

In practice, η can simply set to be constant

Predict

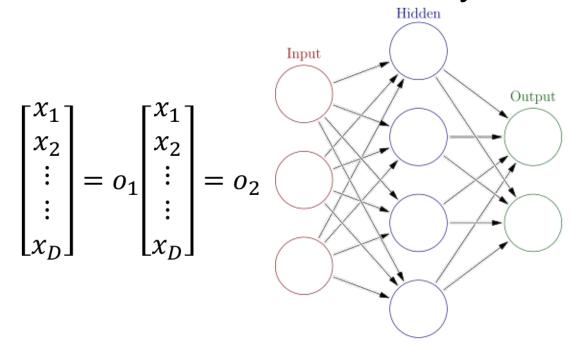
- DNN classifier usually output the 1-best label as prediction.
- Practically, the DNN may preserve certain number of good candidates for further processing due to the **structure of speech**.
 - That is, the best in one frame would be unacceptable when considering the context frames.
- The sequence "fixing" procedure is actually your HW2, comparable among the world as announced.
- In HW1, you should only predict the framewise 1-best output.

More About Input...

- We provide **3 types** of feature for the DNN input w.r.t each frame.
 - Raw Waveform(16k sampling rate)
 - MFCC(dimension = 39)
 - FBANK(dimension = 69)
- MFCC and FBANK are 2 types of feature extracted from the wave. However, the details lie beyond the scope of this course.
- There does exist debates over which feature is better. So the choice is yours.

More About Input...

- You could generate your own features based on the WAV/MFCC/FBANK.
- However, be sure your task can be evaluated in a frame-wise way.



More About Output...

- We provide **3 types** of training target for the DNN output w.r.t each frame.
 - states(index from 0 to 1942)
 - 48 phonemes
 - 39 phonemes
- The difference also lies beyond the scope of this course. You could simply view them as more **delicate pronunciation** units at different level.

More About Output...

- Actually, the framewise labels are usually absent. There are only human annotation w.r.t each sentence.
- Practically, we forcibly align the labels to each frame by generative models other than DNN.
- The alignment results are then prepared for DNN as training targets.
- In this homework, you do not need to bother. The alignments are transformed into framewise labels for you.

Validation

- There are two scores displayed when you upload your results on the site.
- <u>Public score</u> is based on <u>validation set</u> and you can compare yours with others'.
- <u>Private score</u> is based on <u>test set</u> and should determine your final place.
- To better control your training procedure, you may want to test your model on a validation set after each epoch.
 - However, submitting the results onto the scoreboard after each epoch can be tedious,
 - In such case, it is recommended to **split your own validation set** from the training set.

Evaluation

- Framewise phoneme prediction
- Frame Accuracy

$$Frame\ Accuracy = \frac{\#\ of\ correct\ frames}{\#\ of\ total\ frames}$$

Evaluation

- Despite the type of training target you used, you have to map all of them to 39 phoneme set for evaluation.
 - Some training targets would be merged into the same phoneme.
- The frame accuracy is calculated according to the merged prediction.

Baseline

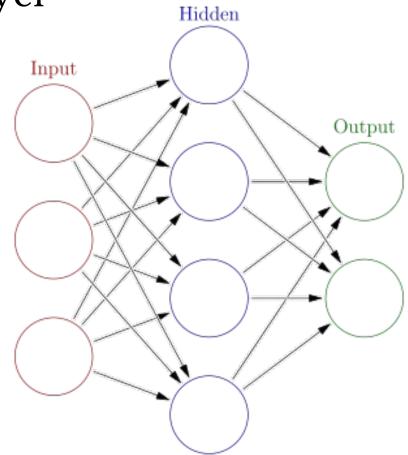
Only one hidden layer

• 69x128x48

• 69: fbank features

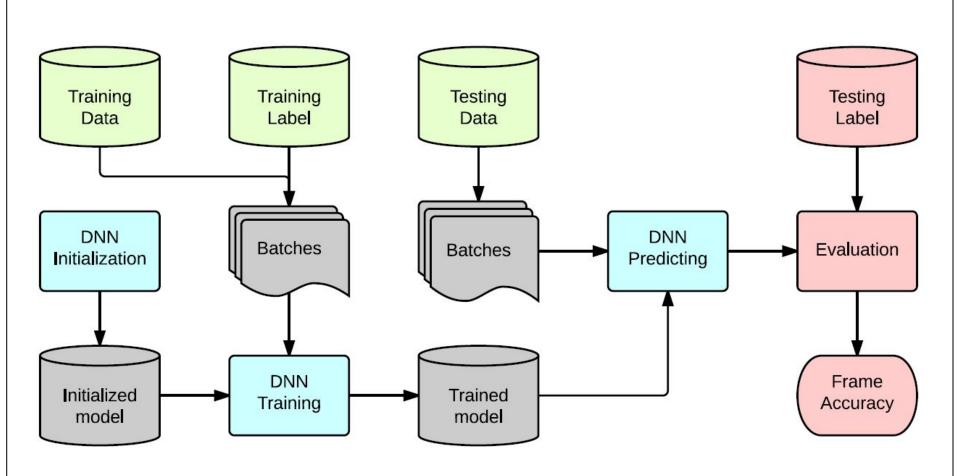
• 128: 1 hidden layer

• 48: 48 phoneme set



Pseudo Code

```
dnn = DNN( STRUCTURE , LEARNING_PARAMETERS )
train = Dataset( TRAIN_FILE , LEARNING_PARAMETERS
valid = Dataset( DEV_FILE , LERANING_PARAMETERS
test = Dataset( TEST_FILE , LEARNING_PARAMETERS
epoch =
while epoch < MAX_EPOCH</pre>
     while ( batch = train.load_batch() )
           dnn.forward(batch)
           dnn.calculate_error( ERROR_FUNCTION );
           dnn.backpropagate();
           dnn.update()
      end
           Ein = dnn.report_error_rate( train );
           Eval = dnn.report_error_rate( valid )
      epoch = epoch +
end
while( batch = test.load_batch() )
      dnn.forward( batch )
      dnn.output_predict()
end
```





Kaggle

- http://www.kaggle.com/
- A competition platform used by many academic institutes.
- Evaluate your result on this website.
 - With Scoreboard that can show off your improvement.
- You could upload your predictions only twice a day.

Kagg Eclass

• Be careful!

Registration

kaggle

Sign up ogin

The Home of Data Science

COMPETITIONS - CUSTOMER SOLUTIONS - JOBS BOARD

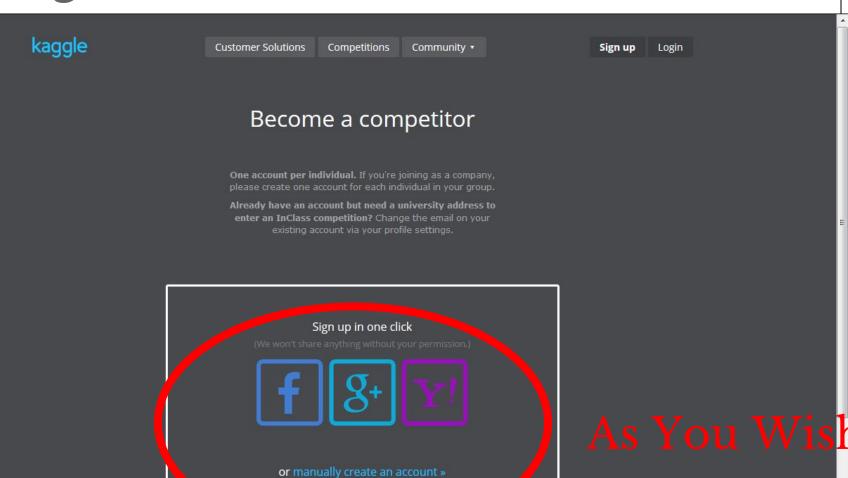
Get started »







Registration



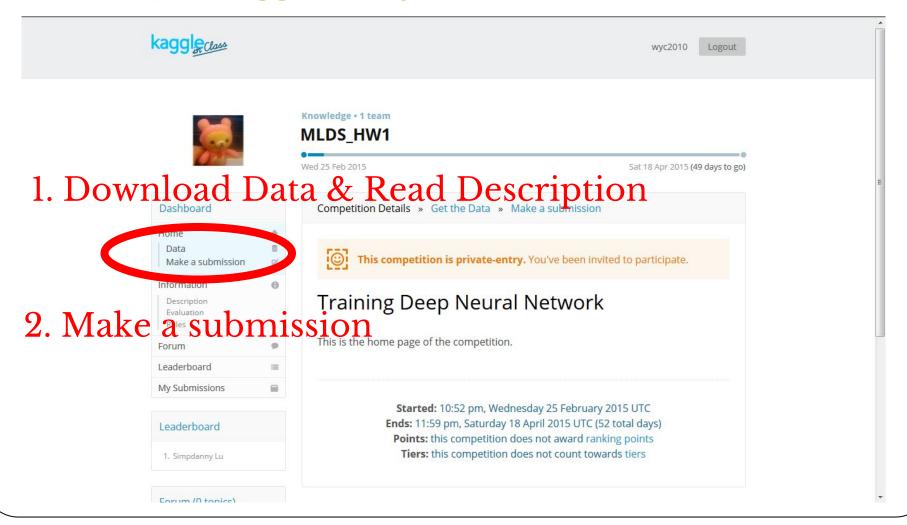
Registration

kaggle **Customer Solutions** Competitions Community • wyc2010 Logout wyc2010 Hi wyc2010! We'd like to welcome you to Kaggle. View / Edit Profile Since you're new, here's just a few ways to get started: On the Forums Table comparing supervised learning algorithms Viewing winner's code like in Topcoder Methodological ML dispute with a Explore the competitions Meet the community Udacity teacher: Could you help Read more about me understanding if I'm right? how Kaggle works Visit the forums for each Download some active What was your favorite 2014 competition? Check out How it Works. competition data files and competition to discuss Are there other Kaggle contests Ask a question on the a sample entry. Or practice methods & results. with no labels of data? Kaggle Forum. on a completed sentiment analysis dataset competition. SUCCESS in the Blog **Active Competitions** Defending Champions Winners' ... Reviewing the American Epilep... March Macline Learning Mania 2015 14 days Kaggle InClass: Stanford's "G... 206 teams Predict the 2015 ICAA Basketball Tournament CIFAR-10 Competition Winners:... \$15,000

Convolutional Nets and CIFAR-...

2nd Place: The Hunt for Probi

https://kaggle.com/join/104_1_mlds_hw1



1 team

1 player

entry

Team Mergers

Team mergers are not allowed in this competition.

Team Limits

There is no maximum team size.

Submission Limits

You may submit a maximum of 2 entries per day. You may select up to 2 final submissions for judging.

Competition Timeline

Start Date: 2/25/2015 10:52:45 PM UTC

Merger Deadline: None

First Submission Deadline: **None** End Date: **4/18/2015 11:59:00 PM UTC**

Rules Acceptance

I understand and accept

I do not accept

By clicking on the "I understand and accept" button below, you are indicating that you agree to be bound to the above rules.



wvc2010

Logout

Thank you for accepting the rules.





SMS Verification

To make a submission, you must verify your Kaggle account via your mobile phone. You will only need to do this once. Do not use a public number or share your

8869XX-XXX-XXX

Phone number (inc. country code)





Remember to add country code(i.e. 886)





wyc2010

Logout



SMS Verification

You should have received your verification

123456Enter your code here



Didn't receive a code?

Try again with the same or a different number.





wyc2010

Logout

Your account has been successfully verified. Treat it nice, it's your only one!



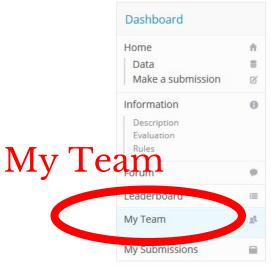


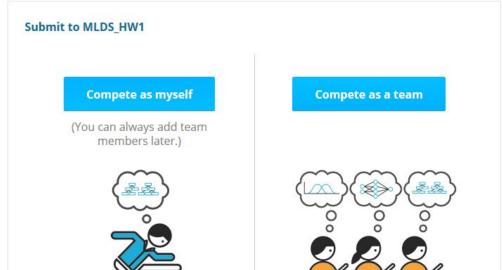
Knowledge • 1 team

MLDS_HW1

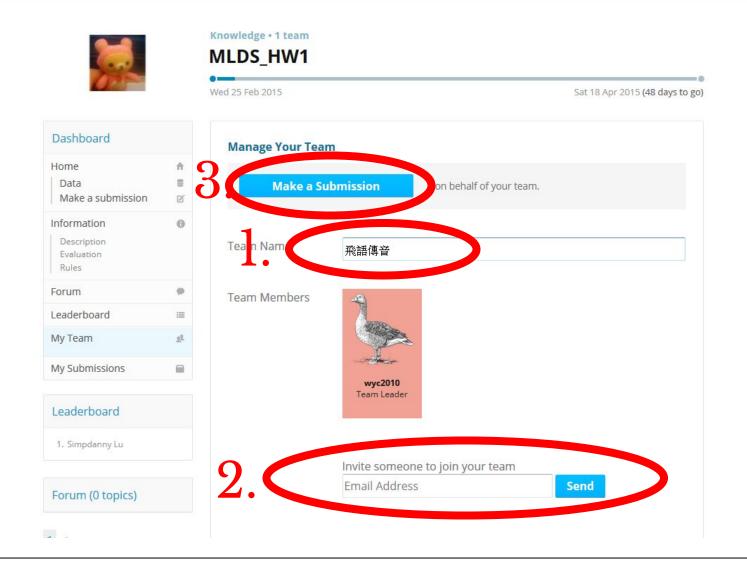
Wed 25 Feb 2015

Sat 18 Apr 2015 (48 days to go)

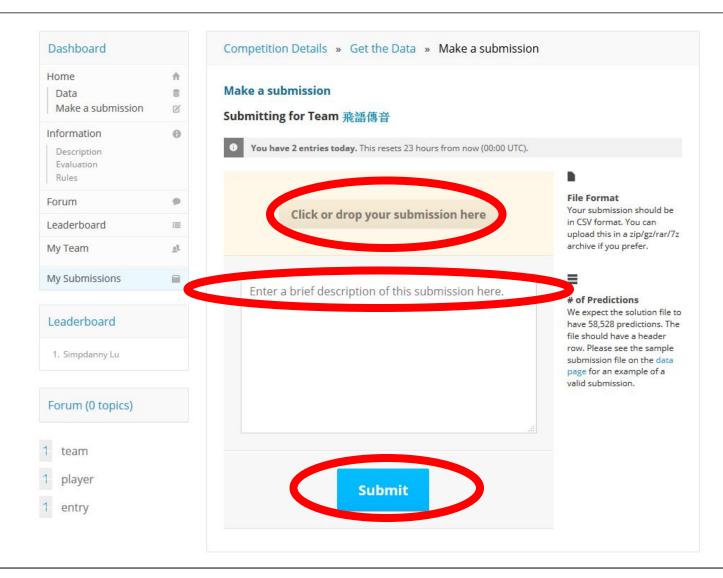




Manage Your Team



Submitting for Team



Final Rules

- Please team-up with a group name in Kaggle in 1 week.
- Any submission by individuals would be neglected and team mergers forbidden.
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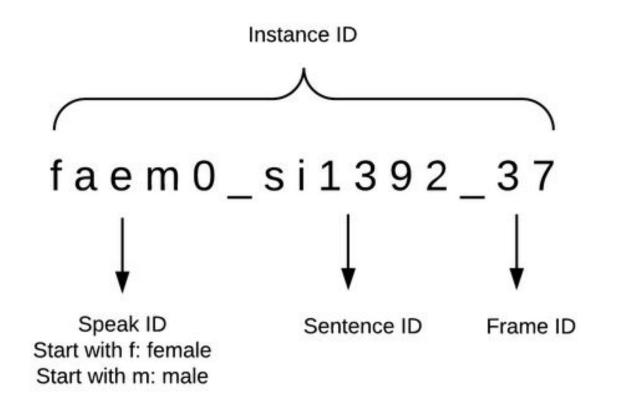
Dataset and Format

Dataset

- TIMIT(Texas Instrument and Massachusetts Institute of Technology)
- Well-transcribed speech of American English speakers of different sexes and dialects.
- Designed for the development and evaluation of ASR systems.

Dataset

- Each instance consists of 3 parts:
 - speaker faem0, sentence sil392, the 37th frame



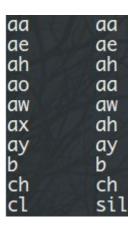
- WAV file: Speak-Sentence ID + .wav
 - Check by your ear(s)
- ARK file: Instance ID+ features
 - 1 faem0_si1392_1 3.48541 3.60388 4.100700 3.520525 1.077076 6.677959 7.571936 7.836775 7.328356 7.501573 8.236745 7.386778 8.365101 9.175194 10.44421 9.683798 10.80442 11.04268 10.06622 10.55132 10.28832 11.0653 6 11.58939 0.2704574 0.05309612 -0.1586174 0.0615164 0.2801027 -0.226 3938 -0.412236 -0.5605489 -0.3172206 -2.384186e-07 -1.192093e-07 0.186 0594 -0.1076797 -0.2 96915 -0.4659706 -0.2504535 -0.4384818 -0.254075 0.005485415 -2.384186 07 0.1432481 -0.166646 -0.1194608 -0.125215 -0.09690015 -0.2077889 -0.07074384 0.01680622 -0.06037173 -0.1455567 -0.1195837 -0.0516952 -0.04357988 -0.07634205 0.03012398 -0.02076791 -0.0 7573968 -0.155601 -0.06261331 3.1235719 -0.1034293 8.340697e-08 -0.03 100524 -0.0533277 -0.0441806 -0.06928831 2 faem0_si1392_2 4.677213 4.400319 4.795176 3.526529 4.637282 6.426411 7.397112 6.217342 0.035976 7.336027 6.236745 8.184175 6.209169 8.50935 9.612121 9.683798 10.4058 10.89597 10.55984 10.9507 11.21045 10.79408 11.63718 -0.1763853 -0.19311 -0.8089488 -0.307582 0.196072 -0.3270134 -0.4397185 -0.5730056 -0.3994629 -0.1466 -0.2009004 0.1417596 -0.071786 4 -0.307953 -0.657351 -0.2283547 -0.478344 -0.3447649 0.02193951 -0.0599072 0.03480816 -0.01525 -0.1672461 -0.28692 -0.1340673 -0.110322 -0.02306862 -0.07882925 -0.01635064 -0.07832487 0.05356358 0.06696881 -0.05751204 -0.09442315 0.02923772 -0.001794666 0.01581386 -0.016641 91 -0.008102864 -0.01395154 0.02923772 -0.001794666 0.01581386 -0.016641 91 -0.008102864 -0.01395154 0.02923772 -0.001794666 0.01581386 -0.016641 91 -0.008102864 -0.01395154 0.05501553 -0.03839263 -0.04873157 -0.101 0142 -0.0005813509 -0.0769349

- LAB file: Instance ID + comma + label
- state label and 48 phones
- you have to map them to 39 phonemes yourselves.

```
mueb0 si1\11 2
maeb0 si
maeb0 si
maeb0 si
maeb0 si
maeb0 si
```

- MAP file: 2 mapping
 - State ID 48 phones 39 phones
 - 48 phones 39 phones
 - Delimiter: tab '\t'

0	sil	sil
1	aa	aa
2	ae	ae
3	ah	ah
4	ao	aa
5	aw	aw
6	ax	ah
7	ay b	ay b
8	b	b
0 1 2 3 4 5 6 7 8 9 1 8	ch	ch
10	cl	sil



- **CSV file**: Your prediction when submission.
- Must mapped to 39-phonemes
- With header row: "Id, Prediction"
- Instance ID + comma + 39-phonemes

```
Id,Prediction
fadg0_si1279_1,sil
fadg0_si1279_2,sil
fadg0_si1279_3,sil
fadg0_si1279_4,sil
fadg0_si1279_5,sil
fadg0_si1279_6,sil
fadg0_si1279_7,sil
fadg0_si1279_8,sil
fadg0_si1279_10,sil
```

SUBMISSION REQUIREMENT

Submission(Kaggle)

- Your predict.csv.
- Twice a day.
- Public score: validation set of TIMIT
- Private score: test set of TIMIT

# ∆1d	Team Name	Score ②	Entries	Last Submission UTC (Best - Last Submission)
1 -	Simpdanny Lu	0.60880	2	Mon, 02 Mar 2015 15:12:21 (-4.3d)
2	Simple Baseline	0.55088		
53.	wyc2010	Waiting to pounce	-	

Submission(Ceiba)

- Your source code and documentation
 - Usage and environment Setting
 - Package dependency
 - NO NN-based toolkit is allowed.
 - Only matrix manipulation acceleration package can be utilized.
- Report
- Upload 1 copy per group.

Report

- Group Information 10%
 - Group Name, Student Name, Student ID
 - Member Contribution
- What have you done? 40%
 - Data structure and algorithm design
 - Data preprocessing
 - Implementation tips and obstacles
 - Bug(s) and how to solve it(them).

Report

- Experiment Setting and Results 50%
 - How you design your experiments
 - Compare different models/techniques
- No more than 4 pages with font 12, A4 size.

Grading Policy

- Accuracy 60%
 - Simple Baseline in Kaggle (released day 7)
 - Once achieve the baseline, you can get the full credit.
 - 1 % less absolute accuracy = 1% absolute credit loss
- Report 40%
- Implementation 20%.
- Bonus
 - First Place 15%
 - First Runner-up 10%
 - Second Runner-up 5%

Grading Policy

- Once you achieve simple baseline, you get 60% full credit **no matter how well your model performs**.
- Please do not get stuck in tuning learning rates or other basic parameters. However, they matter indeed.
- Your team members could be in charge of different methods to improve the results.

Grading Policy

- 40% score is based on how many methods or extensions you have tried, with the experimental **results** in comparison.
- Momentum, dropout, CNN, RBM, learning rate policies, optimization algorithms or...
- Even if your score is not improved by some specific method, the trial is worthy and you will get credits thereby.

Penalty

- Latency
 - Half-life = 24 hrs
 - The excess time will round up to hours.
 - The time is determined by ceiba upload.
- Usage on NN-based toolkit
 - You shall get 0 credit in both implementation and accuracy parts.
 - Please ask TAs whether you could use the package or not.

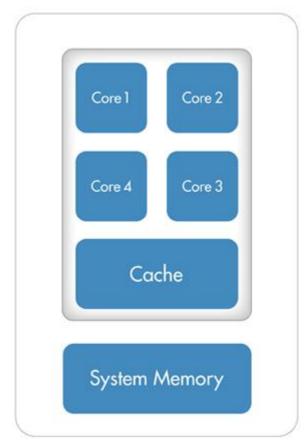
GPU Acceleration

Scalability Issue

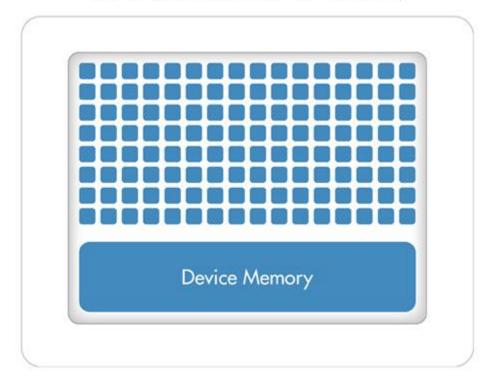
- GPU Acceleration
 - C/C++: CUDA library
 - Python: Theano
 - Matlab: GPU toolbox
- Typically, acceleration lies about 5-20 times.
 - 20 80 hrs -> 4 hrs
 - You got 3 weeks to have your HW1 done.

GPU Acceleration

CPU (Multiple Cores)



GPU (Hundreds of Cores)



PRO: Highly parallel with many micro floating point processors.

CON: Memory transportation bottleneck.

GPU Acceleration - C/C++

- CUDA
- A good toolkit
- https://github.
 com/botonchou/libcumatrix
- Or write your toolkit! (Hard)
- https://developer.nvidia.com/cublas

GPU Acceleration - Python

- Theano is recommended
- Follow the instructions of Theano configuration
- http://deeplearning.
 net/software/theano/library/config.html

GPU Acceleration - Matlab

- gpuArray
- Usage

```
Elapsed time is 1.542150 seconds.
Elapsed time is 0.004115 seconds.
```

• Reference

```
% Init data
A = rand( 2000, 4000 );
gpu_A = gpuArray(A);
B = rand( 4000, 5000 );
gpu_B = gpuArray(B);
% Perform A*B
tic;
C = A * B;
toc;
tic;
gpu_C = gpu_A* gpu_B;
toc;
```

About Toolkits

- Theano is recommended for you to better comprehend GPU programming with easy usage.
- However, theano.nnet is forbidden
- Any package that wraps DNN together is not allowed.
- Ask if you're unsure about the usage of some specific package.

Recommendation

- Start as early as possible.
 - In TA's <u>poor</u> machine(Nvidia GTX 660), 1 epoch costs about 200 seconds.
 - 10 epochs are sufficient to achieve baseline if correctly implemented.
 - Bug is everywhere.

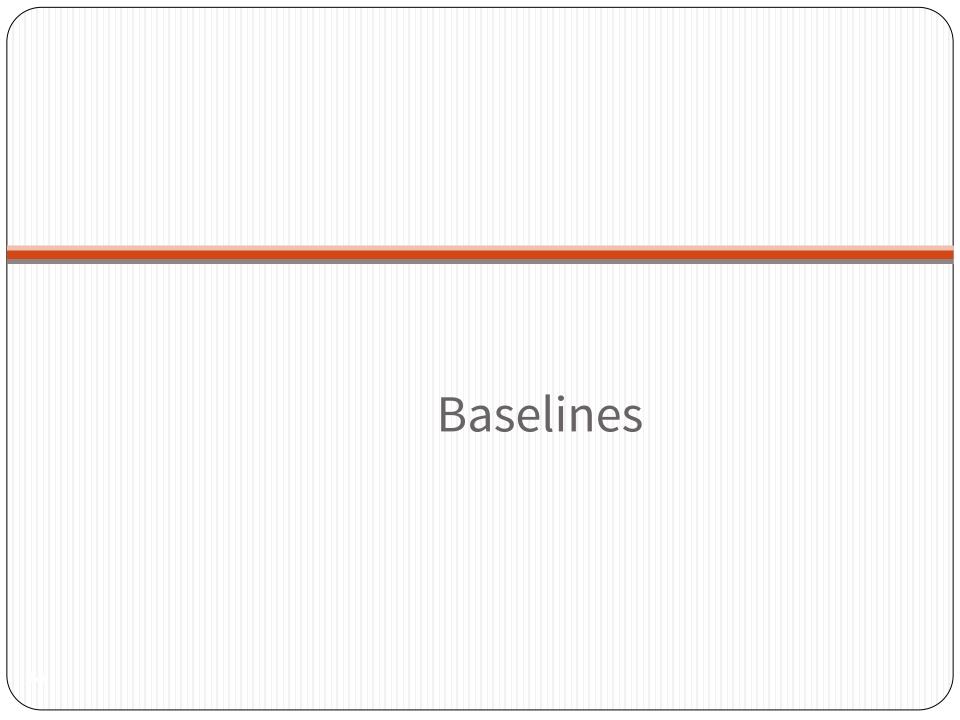


- Start-ups!
 - Group discussion and task assignments
 - Derive all formulas. Everyone.
 - Get familiar with Kaggle.
 - Which language?

- Understand what happened....
 - Look over training data and testing data
 - Many decision
 - Data preprocessing.
 - Basic structure and core methods implementation.
 - Test on tiny set(make it yourselves!).

- Everything settled down
 - Start training and report testing results.
 - Debugging...
 - Can you overfit the training data?
 - Precision overflow? (exponential)
 - Matrix multiplication?
 - Error backpropagation?

- Basic learning techniques
 - Parameter selection
 - batch size, learning rate, or...
 - Frame shuffle?
 - Splice features in context window?
 - Regularization(weight decay)?
 - Learning rate decay?



Simple Baseline - 0.5310

- Once achieve this score, you will get full credit in the grading part of accuracy.
- However, only pure DNN is considered. Your mix learning techniques would help you better win the competition in Kaggle, but your DNN alone should be powerful enough.
- You must present your experiment results with pure DNN in the report.

Simple Baseline - 0.5310

Input	69	fbank features
Hidden	128	only one hidden layer
Output	48	-
Batchsize	128	-
Initial Learning Rate	0.0001	-
Learning Rate Decay	0.9999	decay with each batch updating
Initialization Variance	0.1	Model initial weights gaussian variance
Momentum	0.9	-
Training Time	about 5.5 mins	-

Real Baseline - 0.6179

- Due to the structure of the speech,
 splicing contextual frames (say previous 4 and subsequent 4) would usually helps
- Increase the model complexity with usage of **state level** pronunciation unit.
- A real 2-3 days training time required.

Real Baseline - 0.6179

Input	621	previous 4 frames + subsequent 4 frames + center 1 frame = 9 frames fbank features
Hidden	1024-1024-1024-1024	4 layers
Output	1943	state label
Batchsize	256	-
Learning Rate	0.001	-
Dropout	0.1	-
Initialization	random [-0.1. 0.1]	uniform
Momentum	0.5	-
Training Time	2-3 days	600-700 seconds per epoch