MLDS HW1-3

TAs ntu.mldsta@gmail.com

Outline

- Timeline
- Task Descriptions
- Q&A

Timeline

Three Parts in HW1

- (1-1) Deep vs Shallow:
 - Simulate a funtion.
 - Train on actual task using shallow and deep models.
- (1-2) Optimization
 - Visualize the optimization process.
 - Observe gradient norm during training.
 - What happens when gradient is almost zero?
- (1-3) Generalization
 - Can network fit random labels?
 - Number of parameters v.s. Generalization
 - Flatness v.s. Generalization

Schedule

- 3/9:
 - Release HW1-1
- 3/16:
 - Release HW1-2
- 3/23:
 - Deadline to team-up by yourselves
 - Release HW1-3
- 3/30:
 - Deadline to team-up by TAs
- 4/6:
 - All HW1 due (including HW1-1, HW1-2 and HW1-3)

Task Descriptions

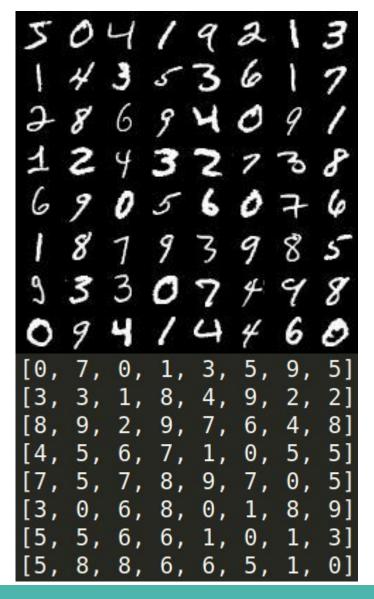
HW1-3: Generalization

- Three subtask
 - Can network fit random labels?
 - Number of parameters v.s. Generalization
 - o Flatness v.s. Generalization
- Train on MNIST or CIFAR-10...

Can network fit random labels? 1/2

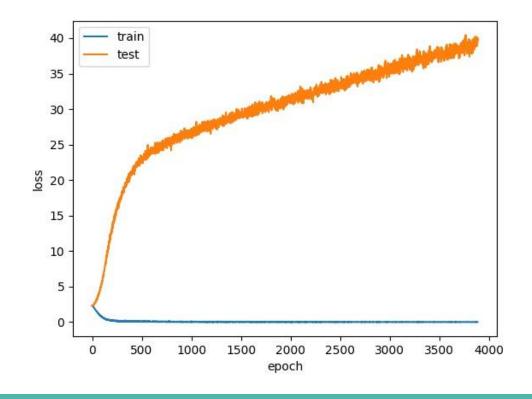
- Requirement
 - Train on MNIST or CIFAR-10
 - Randomly shuffle the label before training.
 - Try to fit the network with these random labels.

Can network fit random labels? 2/2



MNIST

3 hidden layers with 256 nodes.

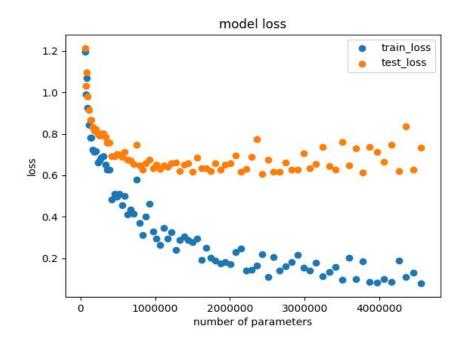


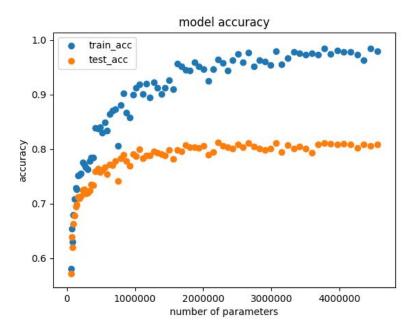
Number of parameters v.s. Generalization 1/2

- Requirement
 - Train on MNIST or CIFAR-10
 - At least 10 similar-structured models with different amount of parameters
 - Record both training and testing, loss and accuracy

Number of parameters v.s. Generalization 2/2

CIFAR-10



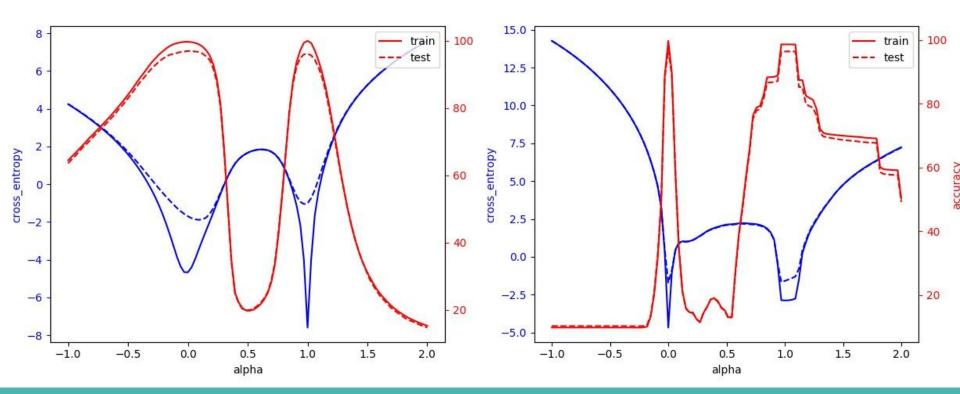


Flatness v.s. Generalization - part1 1/2

- Visualize the line between two trained models
- Requirement:
 - Train two models (m1 and m2) with different training approach. (e.g. batch size 64 and 1024)
 - Record the loss and accuracy of the model which is linear interpolation between m1 and m2.
 - \circ $\theta_{\alpha} = (1 \alpha)\theta_1 + \alpha\theta_2$, where α is the interpolation ratio, θ is the parameter of the model.

Flatness v.s. Generalization - part1 2/2

- MNIST (The cross_entropy is log scale)
 - o batch size 64 vs. batch size 1024
 - learning rate 1e-3 vs. 1e-2



Flatness v.s. Generalization - part2 1/

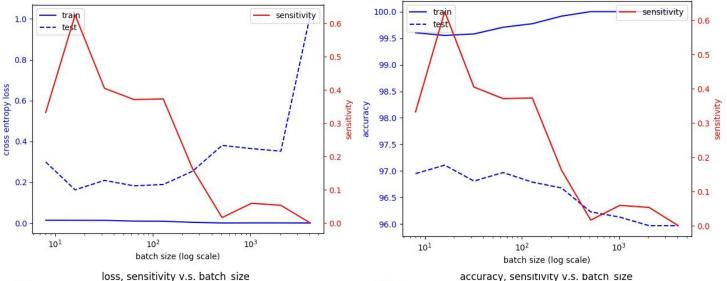
- Requirement:
 - Train at least 5 models with different training approach.
 - Record the loss and accuracy of all models.
 - Record the sensitivity of all models.

Flatness v.s. Generalization - part2 2/

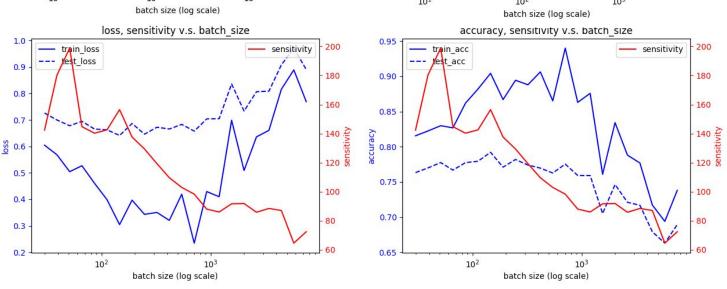
- What is sensitivity:
 - Reference: https://arxiv.org/pdf/1802.08760.pdf
 - Original definition:
 - Frobenius norm of Jacobian matrix of model output (class probability) to input
 - Computationally expensive for us
 - Our definition:
 - Frobenius norm of gradients of loss to input

Flatness v.s. Generalization - part2 3/

MNIST:



CIFAR10:



HW1-3 Report Questions (10%)

- Can network fit random labels?
 - Describe your settings of the experiments. (e.g. which task, learning rate, optimizer) (1%)
 - Plot the figure of the relationship between training and testing, loss and epochs. (1%)
- Number of parameters v.s. Generalization
 - Describe your settings of the experiments. (e.g. which task, the 10 or more structures you choose) (1%)
 - Plot the figures of both training and testing, loss and accuracy to the number of parameters. (1%)
 - Comment your result. (1%)
- Flatness v.s. Generalization
 - o Part 1:
 - Describe the settings of the experiments (e.g. which task, what training approaches) (0.5%)
 - Plot the figures of both training and testing, loss and accuracy to the number of interpolation ratio. (1%)
 - Comment your result. (1%)
 - o Part 2:
 - Describe the settings of the experiments (e.g. which task, what training approaches) (0.5%)
 - Plot the figures of both training and testing, loss and accuracy, sensitivity to your chosen variable. (1%)
 - Comment your result. (1%)
 - o Bonus: Use other metrics or methods to evaluate a model's ability to generalize and concretely describe it and comment your results.
- Check TA's examples to know what to plot!

Flatness v.s. Generalization - bonus example 1/8

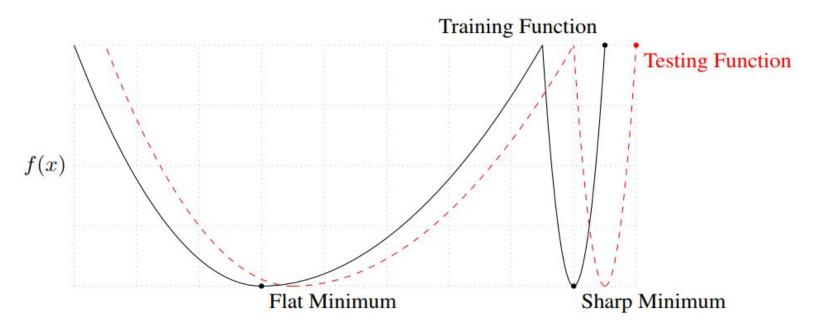
- Reference: https://arxiv.org/pdf/1609.04836.pdf
- Requirement
 - Train on MNIST or CIFAR-10
 - Use at least ten different approaches to train the same model
 - Calculate the sharpness of trained models
- Tips
 - Train on MNIST
 - Train with different batch size

Flatness v.s. Generalization - bonus example 2/8

- There is a generalization gap when using large-batch (LB) methods (instead of small-batch (SB) methods) for training deep learning models.
- The reasons (maybe more than these):
 - LB methods lack the explorative properties of SB methods and tend to zoom-in on the minimizer closest to the initial point.
 - SB and LB methods converge to qualitatively different minimizers with differing generalization properties (i.e. SB converges to flat minimizer, LB converges to sharp minimizer)
- We will focus on the second reason.

Flatness v.s. Generalization - bonus example 3/8

Visually, it means that :



How to measure sharpness (or flatness)?

Flatness v.s. Generalization - bonus example 4/8

- Many methods can measure sharpness, but we will only utilize one in this assignment.
- Notations:
 - \circ θ : vector of all parameters
 - \circ $L(\theta)$: loss of the model given the paramters
 - \circ $B_2(\epsilon, heta)$: a Euclidean ball centers at heta with radius ϵ
- Sharpness:

$$rac{\max_{ heta' \in B_2(\epsilon, heta)}(L(heta') - L(heta))}{1 + L(heta)}$$

Flatness v.s. Generalization - bonus example 5/8

- How to calculate this : $\frac{\max_{\theta' \in B_2(\epsilon,\theta)}(L(\theta')-L(\theta))}{1+L(\theta)}$
- Original paper : Use L-BFGS-B to maximize $L(\theta')$
- Around a critical point : $L(\theta') = L(\theta) + \frac{1}{2}(\theta' \theta)^T (\nabla^2 L)(\theta)(\theta' \theta) + o(\|\theta' \theta\|_2^2)$
- Then:

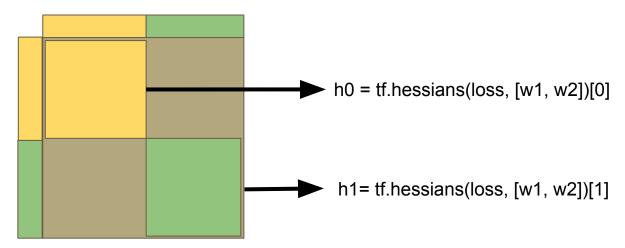
$$\frac{\max_{\theta' \in B_2(\epsilon,\theta)} (L(\theta') - L(\theta))}{1 + L(\theta)} \longrightarrow \frac{\|(\nabla^2 L)(\theta)\|_2 \epsilon^2}{2(1 + L(\theta))}$$

Since 2-norm of a matrix is defined as :

$$||A||_2 = \max\{||Ax||_2 : x \in R^n \text{with} ||x||_2 = 1\} = \sqrt{\lambda_{\max}(A^T A)}$$

Flatness v.s. Generalization - bonus example 6/8

- How to calculate Hessian matrices efficiently:
 - Use GPU: tf.hessians
 - Calculate only 500 out of 60000 examples in MNIST
- But tf.hessians only return block-diagonal part:
 - vector of all paramters: w1 w2
 - Hessian matrix :



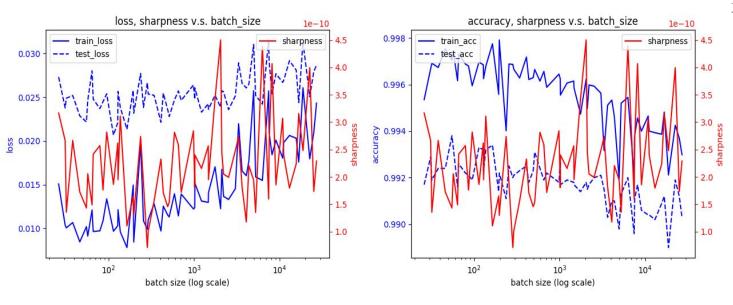
Flatness v.s. Generalization - bonus example 7/8

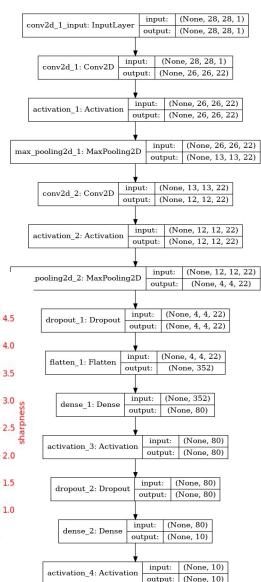
- If we assume off-block-diagonal elements is negligable:
 - Square of block-diagonal matrix is also block-diagonal.
 - Eigenvalues of a block-diagonal matrix is the list of all eigenvalues of each block submatrix.
 - Since we only want the largest eigenvalue, we can conclude that the 2-norm of a block-diagonal matrix is the 2-norm of block submatrix that contains the largest eigenvalue itself.
 - 2-norm of matrix A in tensorflow : tf.norm(A, 2)
 - 2-norm of matrix A in numpy : np.linalg.norm(A, 2)

Flatness v.s. Generalization - bonus example 8/8

MNIST:

- 20000~30000 parameters (in order to calculate hessian matrices while maintaining enough model capacity)
- Calculate hessian matrices as mentioned in previous slide
- o epsilon: 1e-4





Flatness v.s. Generalization - more possible bonus

- Reference: https://arxiv.org/pdf/1703.04933.pdf
- This paper shows that several metrics (including sharpness) do not indicates ability of generalization for any RELU-based deep models.
- Reparametrize:

$$\circ$$
 relu $(x \cdot (lpha heta_1)) \cdot heta_2 = ext{relu}(x \cdot heta_1) \cdot (lpha heta_2)$ if $lpha > 0$

Allow Packages

- python 3.6
- TensorFlow r1.6
- PyTorch 0.3 / torchvision
- Keras 2.0.7 (TensorFlow backend only)
- MXNet 1.1.0
- CNTK 2.4
- matplotlib
- scikit-learn 0.19.1
- Python Standard Library
- Pandas
- If you want to use other packages, please ask TAs for permission first!

Submission

- Deadline: 2018/4/6 23:59 (GMT+8)
- Write the questions of HW1-1, HW1-2 and HW1-3 in one report.
- Chinese unless you are not familiar with Chinese
- At most 10 pages for HW1-1, HW1-2 and HW1-3
- Your github must have several files under directory hw1/
 - Readme.*
 - Report.pdf
 - other code
- In your Readme, state clearly how to run your program to generate the results in your report.
- Files for training is required.
- Check github collaborator. If you still not receive the confirmation, please invite TA's account (mldsta) again.

Q&A

ntu.mldsta@gmail.com

組隊

1.組隊結果公布在

https://docs.google.com/spreadsheets/d/1zcbUvkgjbVV4ghG0EbHfEZ2 uvxWdcDmN2ddvsaAkk7o/edit?usp=sharing

請大家確認, 之後助教幫忙組隊的名單也會公佈在這裡。

- 2.自行組隊將於3/23 23:59 UTC +8 (今天)截止, 之後助教將會以盡量每組湊 到三個人為原則幫大家組隊
- 3.如果你們有兩個人已經一隊,希望找到第三個人一組,但是找不到,請不要填表單並於3/23 23:59 UTC +8(今天)前寄信到助教信箱。助教會盡量從尚未組隊的名單中找人跟你們一隊(不保證找得到,且不保證找到的是強者),不會把你們拆散。