ICT303 – Advanced Machine Learning and Artificial Intelligence

Topic 7: Implementation Details and Discussions

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How to Get in Touch with the Teaching Team

- Internal and External Students
 - Email: H.Laga@murdoch.edu.au.
- Important
 - In any communication, please make sure that you
 - Start the subject of your email with ICT303
 - Include your student ID, name, and the lab slot in which you are enrolled.
 - We will do all our best to answer your queries within 24 hrs.

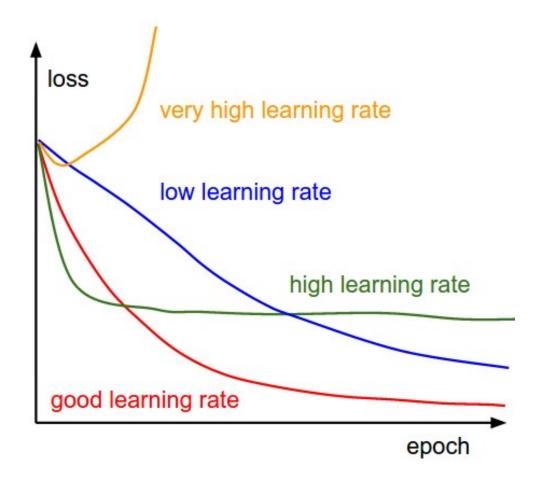
In this Lecture

- Hyper parameters
 - Learning rate
 - Batch size
- Data normalization
- ResNet18 implementation
- Summary

- Learning objectives
 - Understand the effect of some hyper parameters
 - Tune hyperparameters
 - Importance of data normalization
 - Implement ResNet18

Hyperparameters – Learning Rate

Effect of the learning rate on training and loss function



Hyperparameters – Learning Rate

Use a schedule

- Start with a large learning rate (e.g., 0.1)
- Keep decreasing it after a few epochs until epoch == 30 or 50
- Then keep it constant until convergence (at this stage the learning rate is low)

Rate of decrease

- Linear
- Exponential, i.e., every time multiply it by a value that is lower than 1, e.g., 0.98
- Use a decay factor, e.g., Decay = 0.01
 - New LR = Initial LR divided by (1 + Decay X Steps)

Hyperparameters – Learning Rate

- Additional reading
 - Importance of the learning rate
 - https://cs231n.github.io/neural-networks-3/#loss-function
 - Using learning rate scheduler in PyTorch
 - https://machinelearningmastery.com/using-learning-rate-schedule-in-pytorch-training/
- What I didn't cover
 - Momentum update
 - https://cs231n.github.io/neural-networks-3/#loss-function

Hyperparameters – Batch Size

- Affects both the training time and generalization of the model
 - Smaller batch size
 - The network will learn from each individual but it takes long time to converge
 - Larger batch size
 - Fast training
 - Risk of not being able to generalize well it will not capture nuances/variabilities in the data
 - Acceptable sizes
 - 16, 32, 64, 128, 256, 512, 1024
 - Make sure it fits within your GPU memory

Data Normalization

- Normalization is the process of bringing the data into a given range
- Normalizing the input
 - It is important especially of the input are of different types
 - This is usually done by
 - finding the mean and standard deviation
 - Subtract the mean
 - Divide by the standard deviation
 - This transforms the data so that its mean is 0 and its standard deviation is 1.

Data Normalization – Important Note

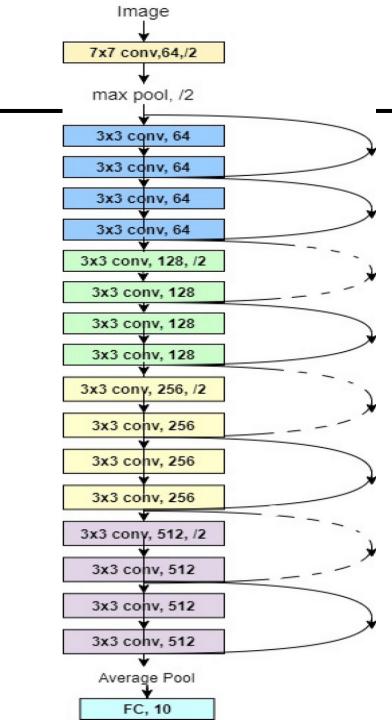
Normalizing the input

- It is important especially of the input are of different types
- This is usually done by
 - finding the mean and standard deviation of the training data
- Normalize the training data
 - Subtract the mean
 - Divide by the standard deviation
- Normalize the validation data
 - Subtract the mean of the training data
 - Divide by the standard deviation of the training data
- Normalize the test data
 - Subtract the mean of the training data
 - Divide by the standard deviation of the training data

Data Normalization – Batch Normalization

- Batch normalization standardizes the inputs to a layer for each mini-batch
 - Important for training very deep neural networks
 - Stabilizing the learning process and dramatically reducing the number of training epochs required to train deep networks
- At each hidden layer
 - Computes the mean and variance of the activation values across the batch
 - Normalizes the activation vector using the computed mean and variance
- PyTorch provides nn.BatchNorm2d or nn.BatchNorm1d
 - Can be applied before or after activation
 - Research shows it is best when applied before

ResNet18 Implementation



Next Week

More about training (practical issues to take into account)

Questions