

Lab 2 experiment report

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1. Introduction

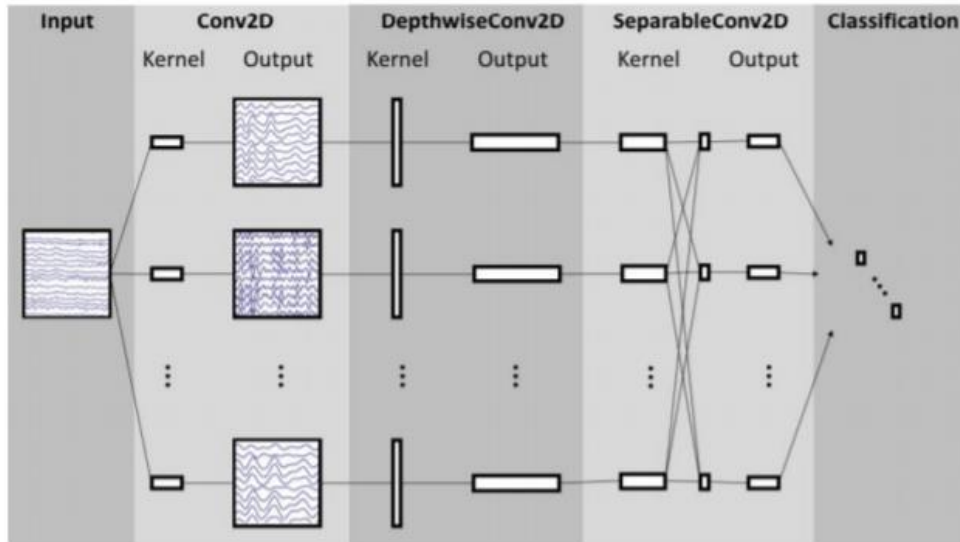
In this homework, we need to train two kinds of deep convolution networks, EEGNet and DeepConvNet, the dataset is BCI Competition III. It is a dataset with 2 classes (left hand, right hand). The dimension of input is $[B, 1, 2, 750]$, the output dimension is $[B, 2]$, and the label dimension is $[B]$.

2. Experiment Set Up

A. The detail of model

◆ EEGNet

EEGNet is composed of Conv2D, DepthwiseConv2D, Separable2D and Classification.



More detail of the model

C = number of channels, T = number of time points, F_1 = number of temporal filters, D = depth multiplier (number of spatial filters), F_2 = number of pointwise filters, and N = number of classes.

Block	Layer	# filters	size	# params	Output	Activation	Options
1	Input				(C, T)		
	Reshape				$(1, C, T)$		
	Conv2D	F_1	$(1, 64)$	$64 * F_1$	(F_1, C, T)	Linear	mode = same
	BatchNorm			$2 * F_1$	(F_1, C, T)		
	DepthwiseConv2D	$D * F_1$	$(C, 1)$	$C * D * F_1$	$(D * F_1, 1, T)$	Linear	mode = valid, depth = D , max norm = 1
	BatchNorm			$2 * D * F_1$	$(D * F_1, 1, T)$		
	Activation				$(D * F_1, 1, T)$	ELU	
	AveragePool2D		$(1, 4)$		$(D * F_1, 1, T // 4)$		
	Dropout*				$(D * F_1, 1, T // 4)$		$p = 0.25$ or $p = 0.5$
2	SeparableConv2D	F_2	$(1, 16)$	$16 * D * F_1 + F_2 * (D * F_1)$	$(F_2, 1, T // 4)$	Linear	mode = same
	BatchNorm			$2 * F_2$	$(F_2, 1, T // 4)$		
	Activation				$(F_2, 1, T // 4)$	ELU	
	AveragePool2D		$(1, 8)$		$(F_2, 1, T // 32)$		
	Dropout*				$(F_2, 1, T // 32)$		$p = 0.25$ or $p = 0.5$
Classifier	Flatten				$(F_2 * (T // 32))$		
	Dense	$N * (F_2 * T // 32)$			N	Softmax	max norm = 0.25

◆ DeepConvNet

The structure of DeepConvNet, in this lab, C=2, T=750, N = 2

Layer	# filters	size	# params	Activation	Options
Input		(C, T)			
Reshape		(1, C, T)			
Conv2D	25	(1, 5)	150	Linear	mode = valid, max norm = 2
Conv2D	25	(C, 1)	$25 * 25 * C + 25$	Linear	mode = valid, max norm = 2
BatchNorm			$2 * 25$		epsilon = 1e-05, momentum = 0.1
Activation				ELU	
MaxPool2D		(1, 2)			
Dropout					p = 0.5
Conv2D	50	(1, 5)	$25 * 50 * C + 50$	Linear	mode = valid, max norm = 2
BatchNorm			$2 * 50$		epsilon = 1e-05, momentum = 0.1
Activation				ELU	
MaxPool2D		(1, 2)			
Dropout					p = 0.5
Conv2D	100	(1, 5)	$50 * 100 * C + 100$	Linear	mode = valid, max norm = 2
BatchNorm			$2 * 100$		epsilon = 1e-05, momentum = 0.1
Activation				ELU	
MaxPool2D		(1, 2)			
Dropout					p = 0.5
Conv2D	200	(1, 5)	$100 * 200 * C + 200$	Linear	mode = valid, max norm = 2
BatchNorm			$2 * 200$		epsilon = 1e-05, momentum = 0.1
Activation				ELU	
MaxPool2D		(1, 2)			
Dropout					p = 0.5
Flatten					
Dense	N			softmax	max norm = 0.5

B. Activation function

◆ ReLU

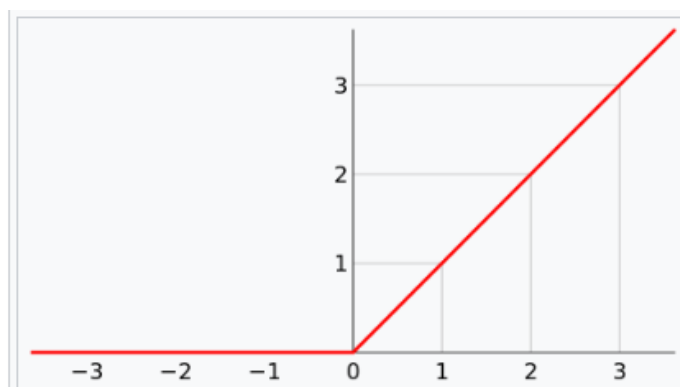
When x is less than 0, f(x) will be 0, and if x is bigger than 0, f(x) will be x.

$$f(x) = \max(0, x)$$

For the neural network that use the activation function, the neuron output will be:

$$\max(0, \mathbf{w}^T \mathbf{x} + b)$$

x is the input of neuron



◆ Leaky ReLU

When $x \leq 0$, there is a small gradient for x .

$$f(x) = \begin{cases} x & \text{if } x > 0, \\ 0.01x & \text{otherwise.} \end{cases}$$

◆ ELU

The *exponential linear unit* (ELU) with $0 < \alpha$ is

$$f(x) = \begin{cases} x & \text{if } x > 0 \\ \alpha (\exp(x) - 1) & \text{if } x \leq 0 \end{cases}, \quad f'(x) = \begin{cases} 1 & \text{if } x > 0 \\ f(x) + \alpha & \text{if } x \leq 0 \end{cases}.$$

α value for the ELU formulation. Default: 1.0

3. Experiment results

Accuracy screenshot of two models:

Highest accuracy: **87.31** (EEGNet with activation function ReLU)

EEGNet

```
Train Accuracy: 98.52 %
Epoch: 294 | train loss: 0.1114
Test Accuracy: 82.59 %

Train Accuracy: 99.07 %
Epoch: 295 | train loss: 0.0294
Test Accuracy: 82.31 %

Train Accuracy: 98.52 %
Epoch: 296 | train loss: 0.0513
Test Accuracy: 80.93 %

Train Accuracy: 98.70 %
Epoch: 297 | train loss: 0.0067
Test Accuracy: 82.22 %

Train Accuracy: 99.17 %
Epoch: 298 | train loss: 0.0225
Test Accuracy: 82.22 %

Train Accuracy: 99.07 %
Epoch: 299 | train loss: 0.0089
Test Accuracy: 83.52 %

Highest Test Accuracy (ELU): 82.59 %
Highest Test Accuracy (ReLU): 87.31 %
Highest Test Accuracy (LReLU): 84.07 %
```

DeepConvNet

```
Train Accuracy: 92.87 %
Epoch: 294 | train loss: 0.0784
Test Accuracy: 73.33 %

Train Accuracy: 90.46 %
Epoch: 295 | train loss: 0.6395
Test Accuracy: 73.43 %

Train Accuracy: 92.31 %
Epoch: 296 | train loss: 0.1575
Test Accuracy: 73.06 %

Train Accuracy: 90.19 %
Epoch: 297 | train loss: 0.2564
Test Accuracy: 71.94 %

Train Accuracy: 91.94 %
Epoch: 298 | train loss: 0.1954
Test Accuracy: 71.02 %

Train Accuracy: 91.11 %
Epoch: 299 | train loss: 0.0819
Test Accuracy: 73.98 %

Highest Test Accuracy (ELU): 78.52 %
Highest Test Accuracy (ReLU): 76.30 %
Highest Test Accuracy (LReLU): 77.31 %
```

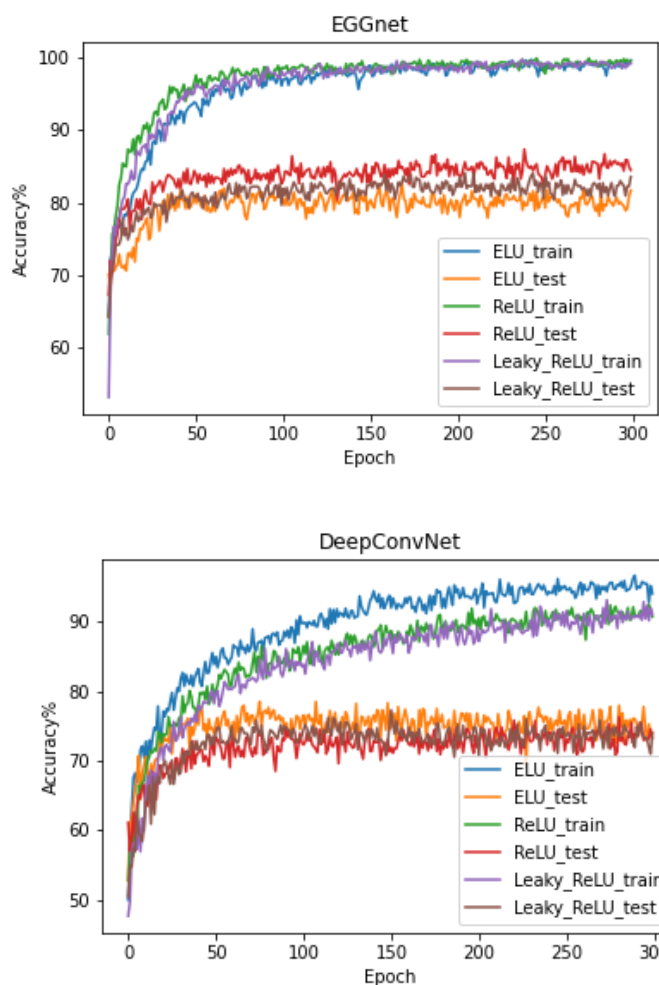
Hyper parameter of EEGnet:

```
EPOCH = 300
LR = 0.005
BATCH_SIZE = 128
```

Hyper parameters of DeepConvNet:

```
EPOCH = 300
LR = 0.01
BATCH_SIZE = 64
```

Comparison figures of models:



4. Discussion

In this Lab, I learned how to build two different kinds of deep convolution networks, I tried to use different parameters to train the models, and find out that when setting batch size:128, learning rate:0.005 and use the activation function ReLU on the EEGNet Model can get the best result(87.31). When building the model, I also found out that between the last linear layer and the last convolutional layer, it's important to use a function view(-1, in_features of linear layer) to make the output of convolutional layer fit the input of linear layer.

5. Reference:

EEGNet

<https://arxiv.org/pdf/1611.08024.pdf>

ELU

<https://arxiv.org/pdf/1511.07289.pdf>

ReLU

[https://en.wikipedia.org/wiki/Rectifier_\(neural_networks\)](https://en.wikipedia.org/wiki/Rectifier_(neural_networks))

Leaky ReLU

<https://zh.wikipedia.org/wiki/%E7%BA%BF%E6%80%A7%E6%95%B4%E6%B5%81%E5%87%BD%E6%95%B0>