Machine Learning HW3

tags: Machine Learning

Files Description

• train.py \ test.py \ models/ \ datasets/

• main.py: 執行全部作業的執行檔

• parameters.csv、 parameters_test.csv: 訓練參數檔

• part_1_result.csv: 第一部分訓練及測試結果

• part_1_result/: 第一部分訓練曲線及模型參數檔

• part_2_6_result.csv: 第二部分最佳表現的訓練及測試結果

• part_2_6_result/: 第二部分最佳表現的訓練曲線及模型參數檔

• hw3-report.pdf: 作業報告書

Execution description

執行訓練

• 訓練曲線及模型參數檔存於 train_result

\$ python train.py

執行測試

\$ python test.py

執行完整作業

• 包含兩部分的訓練及測試

\$ python main.py

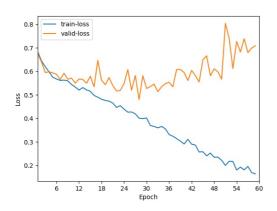
Experimental results

Part 1

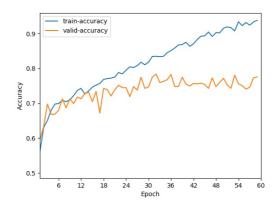
Part 1訓練結果

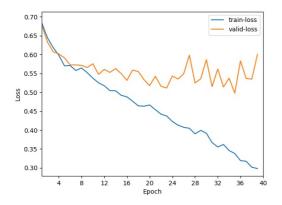
■ batch size	<u>Aa</u> epochs	≡ Ir.	train loss	valid loss	train acc.	valid acc.	test acc. (%)
32	<u>40</u>	0.01	0.2954	0.5902	0.8762	0.768	75.9
32	<u>50</u>	0.01	0.2131	0.6244	0.9157	0.773	75.1
32	<u>20</u>	0.01	0.4714	0.5242	0.7720	0.737	74.1
32	<u>60</u>	0.01	0.1554	0.7433	0.9397	0.776	73.9
16	<u>20</u>	0.01	0.4838	0.5354	0.7698	0.735	71.5
8	<u>20</u>	0.01	0.5317	0.5518	0.7392	0.7120	69.1
32	<u>20</u>	0.001	0.5410	0.5773	0.7183	0.699	66.4
32	<u>20</u>	0.1	0.6970	0.6940	0.4908	0.4990	57.7

訓練曲線

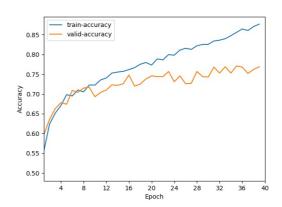


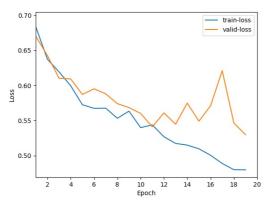




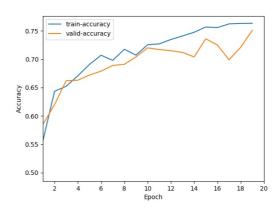


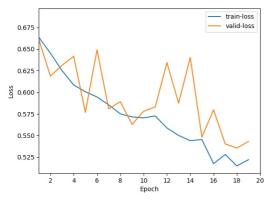
epochs = 40, batch size = 32, lr = 0.01



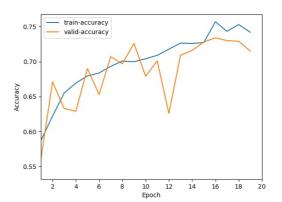


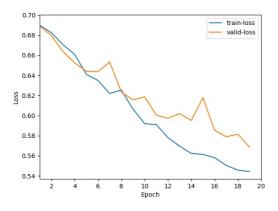
epochs = 20, batch size = 32, lr = 0.01



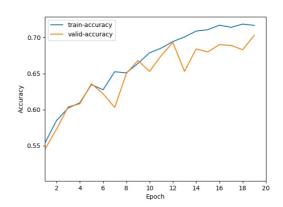


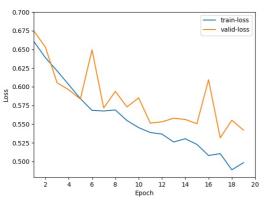
epochs = 20, batch size = 8, Ir = 0.01



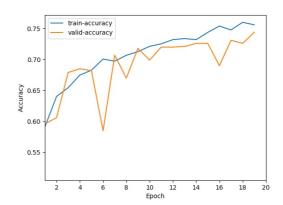


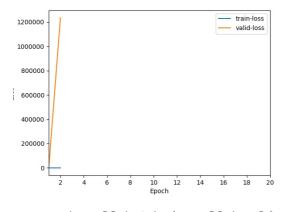
epochs = 20, batch size = 32, lr = 0.001



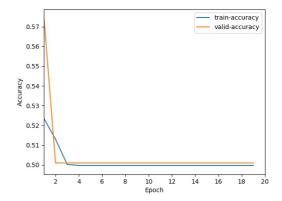


epochs = 20, batch size = 16, lr = 0.01





epochs = 20, batch size = 32, Ir = 0.1



Part 2

Try 1

• 多一層卷積層和池化層

模型架構

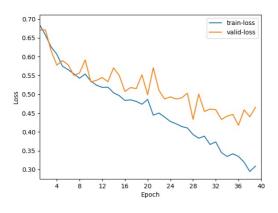
Layer (type)	Output Shape	Param #				
Conv2d-1	 [-1, 64, 224, 224]	1,792				
ReLU-2	[-1, 64, 224, 224]	0				
MaxPool2d-3	[-1, 64, 112, 112]	0				
Conv2d-4	[-1, 64, 112, 112]	36,928				
ReLU-5	[-1, 64, 112, 112]	0				
MaxPool2d-6	[-1, 64, 56, 56]	0				
Conv2d-7	[-1, 64, 56, 56]	36,928				
ReLU-8	[-1, 64, 56, 56]	0				
MaxPool2d-9	[-1, 64, 28, 28]	0				
Conv2d-10	[-1, 64, 28, 28]	36,928				
ReLU-11	[-1, 64, 28, 28]	0				
MaxPool2d-12	[-1, 64, 14, 14]	0				
Linear-13	[-1, 512]	6,423,040				
ReLU-14	[-1, 512]	0				
Linear-15	[-1, 512]	262,656				
ReLU-16	[-1, 512]	0				
Linear-17	[-1, 2]	1,026				
Tatal manager 6, 700, 300						
Total params: 6,799,298 Trainable params: 6,799,298 Non-trainable params: 0						
Input size (MB): 0.57 Forward/backward pass size (MB): 73.23 Params size (MB): 25.94						
Estimated Total Size (MB): 9	99.74					

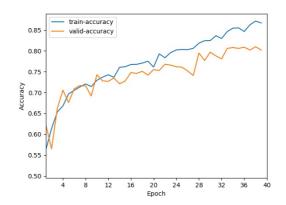
Try 1訓練結果

≡ batch size	<u>Aa</u> epochs	E Ir.		valid loss		valid acc.	test acc. (%)
32	<u>40</u>	0.01	0.3093	0.4657	0.8670	0.802	81.9
32	<u>50</u>	0.01	0.2258	0.4387	0.9060	0.829	81.4
32	<u>60</u>	0.01	0.1658	0.4447	0.9285	0.836	80.5

訓練曲線

只放效果最好的





epochs = 40, batch size = 32, lr = 0.01

Try 2

• 奠基在 Try 1 上,再多一層卷積層(不加池化層)

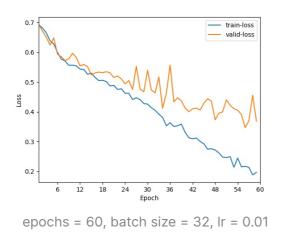
Layer (type)	Output Shape	Param #			
Conv2d-1 [-	1, 64, 224, 224]	1,792			
ReLU-2	1, 64, 224, 224	0			
MaxPool2d-3 [-	1, 64, 112, 112	0			
Conv2d-4 [-	1, 64, 112, 112]	36,928			
ReLU-5 [-	1, 64, 112, 112]	0			
MaxPool2d-6	[-1, 64, 56, 56]	0			
Conv2d-7	[-1, 64, 56, 56]	36,928			
ReLU-8	[-1, 64, 56, 56]	0			
MaxPool2d-9	[-1, 64, 28, 28]	0			
Conv2d-10	[-1, 64, 28, 28]	36,928			
ReLU-11	[-1, 64, 28, 28]	0			
Conv2d-12	[-1, 64, 28, 28]	36,928			
ReLU-13	[-1, 64, 28, 28]	0			
MaxPoo12d-14	-1, 64, 14, 14	Ø			
Linear-15	[-1, 512]	6,423,040			
ReLU-16	[-1, 512]	0			
Linear-17	[-1, 512]	262,656			
ReLU-18	[-1, 512]	0			
Linear-19	[-1, 2]	1,026			
Total params: 6,836,226 Trainable params: 6,836,226 Non-trainable params: 0 Input size (MB): 0.57 Forward/backward pass size (MB): 73.99 Params size (MB): 26.08 Estimated Total Size (MB): 100.65					

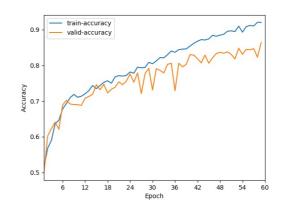
Try 2 訓練結果

■ batch size	<u>Aa</u> epochs	≡ Ir.	■ train loss	valid loss	train acc.	valid acc.	test acc. (%)
32	<u>60</u>	0.01	0.1967	0.3685	0.9207	0.864	85.0
32	<u>55</u>	0.01	0.2170	0.4047	0.9125	0.831	84.5
32	<u>50</u>	0.01	0.2770	0.4347	0.8822	0.820	84.3
32	<u>40</u>	0.01	0.3421	0.4168	0.8500	0.814	82.0
32	<u>55</u>	0.001	0.4933	0.5371	0.7612	0.727	72.1

訓練曲線

只放效果最好的





Try 3

• 奠基在 Try 2 上,再多好幾層卷積層(不加池化層)

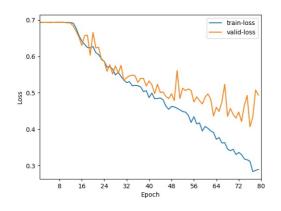
Layer (type)	Output Shape	Param #					
Conv2d-1	[-1, 64, 224, 224]	1,792					
ReLU-2	[-1, 64, 224, 224]	0					
MaxPool2d-3	[-1, 64, 112, 112]	0					
Conv2d-4	[-1, 64, 112, 112]	36,928					
ReLU-5	[-1, 64, 112, 112]	0					
Conv2d-6	[-1, 64, 112, 112]	36,928					
ReLU-7	[-1, 64, 112, 112]	0					
MaxPool2d-8	[-1, 64, 56, 56]	0					
Conv2d-9	[-1, 64, 56, 56]	36,928					
ReLU-10	[-1, 64, 56, 56]	0					
Conv2d-11	[-1, 64, 56, 56]	36,928					
ReLU-12	[-1, 64, 56, 56]	0					
MaxPool2d-13	[-1, 64, 28, 28]	0					
Conv2d-14	[-1, 64, 28, 28]	36,928					
ReLU-15	[-1, 64, 28, 28]	0					
Conv2d-16	[-1, 64, 28, 28]	36,928					
ReLU-17	[-1, 64, 28, 28]	0					
MaxPool2d-18	[-1, 64, 14, 14]	0					
Linear-19	[-1, 512]	6,423,040					
ReLU-20	[-1, 512]	0					
Linear-21	[-1, 512]	262,656					
ReLU-22	[-1, 512]	0					
Linear-23	[-1, 2]	1,026					
Total params: 6,910,082 Trainable params: 6,910,082 Non-trainable params: 0							
Input size (MB): 0.57 Forward/backward pass size (MB): 89.31 Params size (MB): 26.36 Estimated Total Size (MB): 116.24							

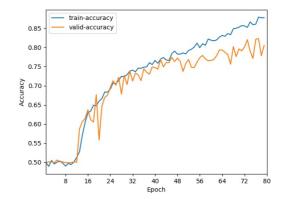
Try 3 訓練結果

■ batch size	<u>Aa</u> epochs	≡ Ir.	■ train loss	valid loss		valid acc.	test acc. (%)
32	<u>80</u>	0.01	0.2896	0.4936	0.8775	0.805	81.1
32	<u>70</u>	0.01	0.3420	0.4388	0.8508	0.796	77.9
32	<u>60</u>	0.01	0.4336	0.4764	0.8013	0.764	75.1
32	<u>55</u>	0.01	0.4427	0.4991	0.7963	0.760	73.9
64	<u>60</u>	0.01	0.4800	0.5363	0.7752	0.729	73.8

訓練曲線

只放效果最好的





epochs = 80, batch size = 32, lr = 0.01

Try 4

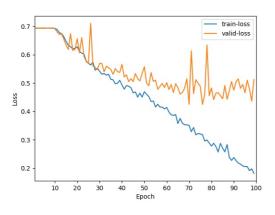
• 奠基在 Try 3 上,減少一層卷積層

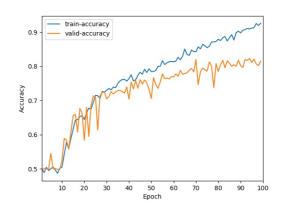
Layer (type)	Output Shape	Param #			
 Conv2d-1	[-1, 64, 224, 224]	1,792			
ReLU-2	[-1, 64, 224, 224]	0			
MaxPool2d-3	[-1, 64, 112, 112]	0			
Conv2d-4	[-1, 64, 112, 112]	36,928			
ReLU-5	[-1, 64, 112, 112]	0			
MaxPool2d-6	[-1, 64, 56, 56]	0			
Conv2d-7	[-1, 64, 56, 56]	36,928			
ReLU-8	[-1, 64, 56, 56]	0			
Conv2d-9	[-1, 64, 56, 56]	36,928			
ReLU-10	[-1, 64, 56, 56]	0			
MaxPool2d-11	[-1, 64, 28, 28]	96 020			
Conv2d-12 ReLU-13	[-1, 64, 28, 28]	36 , 928 0			
KeLU-13 Conv2d-14	[-1, 64, 28, 28] [-1, 64, 28, 28]	ە 36 , 928			
ReLU-15	[-1, 64, 28, 28]	30,926 0			
MaxPool2d-16	[-1, 64, 14, 14]	0			
Linear-17	[-1, 512]	6,423,040			
Rel U-18	[-1, 512]	0,423,640			
Linear-19	[-1, 512]	262,656			
ReLU-20	[-1, 512]	0			
Linear-21	[-1, 2]	1,026			
Fotal params: 6,873,154 Frainable params: 6,873,154 Fon-trainable params: 0	ı				
Input size (MB): 0.57 Forward/backward pass size (MB): 77.06 Params size (MB): 26.22 Estimated Total Size (MB): 103.85					

Try 4 訓練結果

≡ batch size	<u>Aa</u> epochs	≡ Ir.	■ train loss	valid loss		valid acc.	test acc. (%)
32	<u>100</u>	0.01	0.1814	0.5125	0.9260	0.815	78.3

訓練曲線





epochs = 100, batch size = 32, lr = 0.01

<u>Try 5</u>

• 奠基在 Try 2 上,更改卷積層的 channel 參數,及全連接層的輸出入參數

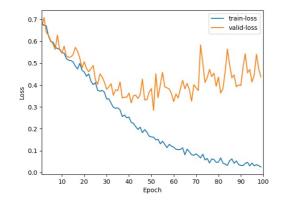
Layer (type)	Output Shape	 Param #				
Conv2d-1 ReLU-2 MaxPool2d-3 Conv2d-4 ReLU-5 MaxPool2d-6 Conv2d-7 ReLU-8 MaxPool2d-9 Conv2d-10 ReLU-11 Conv2d-12 ReLU-13 MaxPool2d-14 Linear-15 ReLU-16	[-1, 64, 224, 224] [-1, 64, 112, 112] [-1, 64, 112, 112] [-1, 128, 112, 112] [-1, 128, 112, 112] [-1, 128, 56, 56] [-1, 256, 56, 56] [-1, 256, 56, 56] [-1, 256, 28, 28] [-1, 512, 28, 28] [-1, 512, 28, 28] [-1, 512, 28, 28] [-1, 512, 28, 28] [-1, 512, 14, 14] [-1, 1024]	1,792 0 0 73,856 0 0 295,168 0 0 1,180,160 0 2,359,808 0 0				
Linear-17 ReLU-18 Linear-19	[-1, 512] [-1, 512] [-1, 2]	524,800 0 1,026				
Total params: 107,198,082 Trainable params: 107,198,082 Non-trainable params: 0						
Input size (MB): 0.57 Forward/backward pass size (MB): 109.51 Params size (MB): 408.93 Estimated Total Size (MB): 519.01						

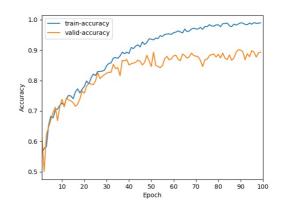
Try 5 訓練結果

■ batch size		≡ Ir.	■ train loss	valid loss	train acc.	valid acc.	test acc. (%)
32	<u>100</u>	0.01	0.0252	0.4365	0.9895	0.893	87.3

訓練曲線

Overfitting !!!





Try 6

• 奠基在 Try 5 上,加兩層卷積層及池化層

模型架構

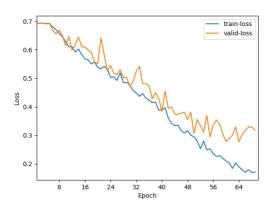
	Layer (type)	Output Shape	Param #			
	Conv2d-1	[-1, 64, 224, 224]	1,792			
	ReLU-2 MaxPool2d-3	[-1, 64, 224, 224] [-1, 64, 112, 112]	0			
	Conv2d-4	[-1, 04, 112, 112]	73,856			
	ReLU-5	[-1, 128, 112, 112]	73,830			
	MaxPool2d-6	[-1, 128, 56, 56]	0			
	Conv2d-7	[-1, 256, 56, 56]	295,168			
	ReLU-8	[-1, 256, 56, 56]	0			
	MaxPool2d-9	[-1, 256, 28, 28]	0			
	Conv2d-10	[-1, 512, 28, 28]	1,180,160			
	ReLU-11	[-1, 512, 28, 28]	0			
	Conv2d-12	[-1, 512, 28, 28]	2,359,808			
	ReLU-13	[-1, 512, 28, 28]	0			
	MaxPool2d-14	[-1, 512, 14, 14]	0			
	Conv2d-15	[-1, 512, 14, 14]	2,359,808			
	ReLU-16	[-1, 512, 14, 14]	0			
	Conv2d-17	[-1, 512, 14, 14]	2,359,808			
	ReLU-18	[-1, 512, 14, 14]	0			
	MaxPool2d-19	[-1, 512, 7, 7]	25 (01 126			
	Linear-20 ReLU-21	[-1, 1024] [-1, 1024]	25,691,136 0			
	Linear-22	[-1, 1024] [-1, 512]	524,800			
	ReLU-23	[-1, 512] [-1, 512]	0			
	linear-24	[-1, 312]	1,026			
======	======================================	[±, 2] 				
Total params: 34,847,362 Trainable params: 34,847,362 Non-trainable params: 0						
Input size (MB): 0.57 Forward/backward pass size (MB): 112.76 Params size (MB): 132.93 Estimated Total Size (MB): 246.27						

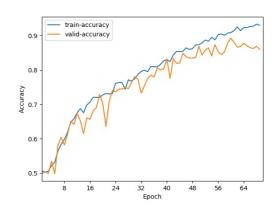
Try 6 訓練結果

■ batch size	<u>Aa</u> epochs	= Ir.		valid loss		valid acc.	test acc. (%)
32	<u>70</u>	0.01	0.1715	0.3179	0.9303	0.86	87.6
32	<u>100</u>	0.01	0.0698	0.2978	0.9730	0.896	87.2

Machine Learning HW3

訓練曲線





epochs = 70, batch size = 32, Ir = 0.01

Conclusion

Part 1

經過實驗,Batch size = 32, epochs = 40-50, learning rate = 0.01 有較好的 訓練學習表現,測試集的準確率為 75.9%。

在訓練次數(epochs)方面,如果訓練到60回合,會有 Overfitting 的可能,所以最佳訓練回合在40-50之間。而learning rate 如果太大,會導致學習過快而準確率過低,甚至 Loss 值有變為 NAN 的可能,而太小則可能要訓練更多回合才有一樣的準確率,所以此模型的 learning rate 訂為 0.01 有更好的表現。Batch Size 方面,過少的量會降低學習效果,但是過多的量也不會有更好的效果,所以訂為 32 是最適合的量。

Part 2

我以 VGG-19 模型為參考來修改模型,經多次增加卷積層、池化層或修改參數的 測試後,發現第六次嘗試的模型設計有較佳的表現,以下為該模型的架構(每一 Conv2d 後都有接 ReLU):

- Conv2d-64
- MaxPool2d
- Conv2d-128

- MaxPool2d
- Conv2d-256
- MaxPool2d
- Conv2d-512
- Conv2d-512
- MaxPool2d
- Conv2d-512
- Conv2d-512
- MaxPool2d
- FC-1024
- FC-512
- FC-2

該模型在 Batch size = 32, epochs = 70, learning rate = 0.01 有較好的訓練學習表現。測試集的準確率為 87.6%。

Discussion

Part 1

在測試時,訓練回合數在40-60都有可能達到最佳的效果,且回合數再多就有 Overfitting 的可能。

Part 2

在以 VGG-19 模型為參考來進行嘗試時,發現越接近輸入的部分不能有較多的卷積層,而以往模型輸出為方向遞增卷積層數。如果要有更好的學習表現,可能要再增加更多層及調高參數量,但訓練時間也因此拉長,所以我最終止步於第六次嘗試,得到87.6%的準確率,雖然不算高,但以自建模型且只利用卷積層及池化層來說已經算高了。