深度學習模型的比較研究 – 以 MNIST 為例

A Github Paper

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Abstract

不同的神經網路模型，經訓練之後的正確率可能差異很大。本文針對手寫數字辨識的 MNIST 資料庫進行測試，以便觀察模型的表現，並分析其背後的原因。

***K*eywords** 神經網路 · 深度學習 · MNIST

1 簡介

近幾年深度學習技術讓人工智慧領域有了很大的進展，也吸引到了學術界與產業界共同投入研究，相繼開發出更好，但也相對更複雜的模型。

為何有些模型表現好，有些模型表現差，各個網路層的效用是甚麼，為何需要加入某些層，若拿掉的話會有甚麼不良反應嗎？這就是本研究所想要探討的問題！

2 背景

手寫數字辨識的 MNIST 是影像辨識領域中最常被拿來測試的資料集，而 CNN 卷積神經網路架構的 LeNet 則是 Yann Le Cun 1989 年在研究手寫辨識問題時，提出來的辨識模型，實驗發現 LeNet 在手寫辨識上有相當高的正確率。

不過，其他的模型，像是使用多層感知器，也可以達到 90% 以上的正確率，

3 方法

簡易的『爬山演算法』如下圖所示 1 所示。

```
Algorithm Hill-Climbing(pi)
 p = pi // 設定粒子 p 為起始粒子 pi
 while not isEnd()
 pn = p.neighbor(step) //選擇粒子 p 的鄰居 pn
 if pn.fitness()>=p.fitness() //如果更好，就接受
 p = pn;
End Algorithm
```

3.1 高度函數如何設計

Measure Measure Measure Measure Measure Measure Measure Measure Measure Measure Measure Measure Measure Measure

*Use footnote for providing further information about author (webpage, alternative address)—*not* for acknowledging funding agencies.

Table 1: 不同模型的 MNIST 正確率

模型	正確率	說明
fc1	74%	
fc1s	92%	
fc2	10%	損失負無限大
fc2s	92%	
fc2net	95%	
LeNet	97%	

$$\xi_{ij}(t) = P(x_t = i, x_{t+1} = j|y, v, w; \theta) = \frac{\alpha_i(t) a_{ij}^{w_t} \beta_j(t+1) b_j^{v_{t+1}}(y_{t+1})}{\sum_{i=1}^{N} \sum_{j=1}^{N} \alpha_i(t) a_{ij}^{w_t} \beta_j(t+1) b_j^{v_{t+1}}(y_{t+1})} \tag{1}$$

3.1.1 如何選取好的鄰居？

任何的參數變動，都可以創造出新的鄰居模型，因此、鄰居的選擇性是無限多的，我們面臨的問題是，該如何從無限多的鄰居當中，選擇一個有可能更好的適當鄰居呢？

在此、我們用了一些啟發式法則如下

1. 加一個新層
2. 將一層換成另一層
3. 調整某層的參數

Paragraph Paragraph Paragraph Paragraph Paragraph Paragraph Paragraph Paragraph Paragraph Paragraph Paragraph Paragraph Paragraph Paragraph Paragraph

4 Experiments

[1, 2] and see [3].

The documentation for `natbib` may be found at

`http://mirrors.ctan.org/macros/latex/contrib/natbib/natnotes.pdf`

Of note is the command `\citet`, which produces citations appropriate for use in inline text. For example,

```
\citet{hasselmo} investigated\dots
```

produces

Hasselmo, et al. (1995) investigated…

`https://www.ctan.org/pkg/booktabs`

4.1 Figures

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[2] Sample of the first footnote.

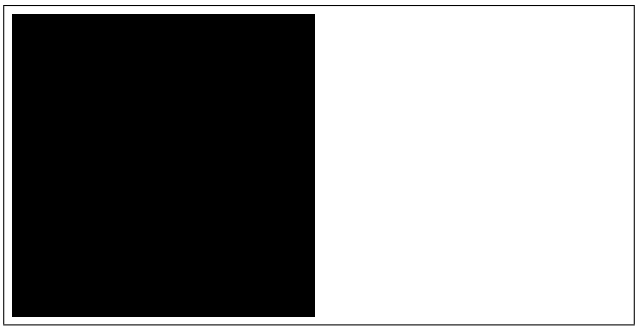

Figure 1: Sample figure caption.

Table 2: Sample table title

Part		
Name	Description	Size (μm)
Dendrite	Input terminal	$\sim$100
Axon	Output terminal	$\sim$10
Soma	Cell body	up to 10^6

4.2 Tables

Etiam euismod. Fusce facilisis lacinia dui. Suspendisse potenti. In mi erat, cursus id, nonummy sed, ullamcorper eget, sapien. Praesent pretium, magna in eleifend egestas, pede pede pretium lorem, quis consectetuer tortor sapien facilisis magna. Mauris quis magna varius nulla scelerisque imperdiet. Aliquam non quam. Aliquam porttitor quam a lacus. Praesent vel arcu ut tortor cursus volutpat. In vitae pede quis diam bibendum placerat. Fusce elementum convallis neque. Sed dolor orci, scelerisque ac, dapibus nec, ultricies ut, mi. Duis nec dui quis leo sagittis commodo. See awesome Table 2.

4.3 Lists

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- consectetur adipiscing elit.
- Aliquam dignissim blandit est, in dictum tortor gravida eget. In ac rutrum magna.

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