

Machine Learning HW5 Report

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1. (1%) 請說明你實作之 RNN 模型架構及使用的 word embedding 方法，回報模型的正確率並繪出訓練曲線*

RNN 模型架構:

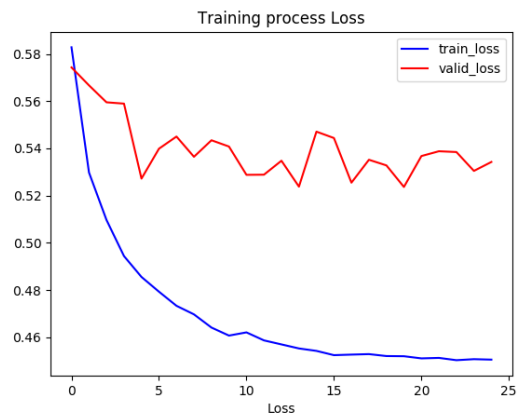
一層 LSTM (參數: hidden_size=vector_space, n_layers=6, bidirectional=True, dropout=0.5)

一層 Linear 降到 2 維後過 LogSoftmax

word embedding 方法:

使用 gensim.models.Word2Vec (參數:size=25, window=5, min_count=1)

Public score: 0.78604



2. (1%) 請實作 BOW+DNN 模型，敘述你的模型架構，回報模型的正確率並繪出訓練曲線*。

BOW 總共有 5205 個 words，使用 tfidf

nn.Linear(5205,2048),

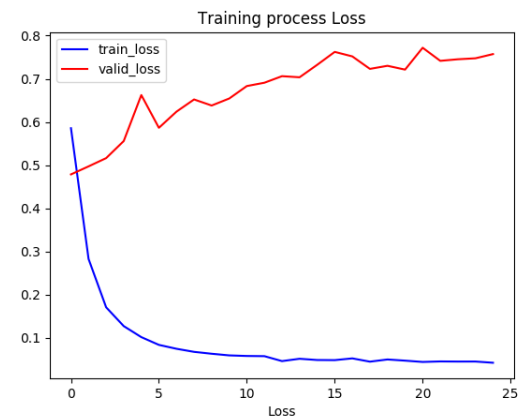
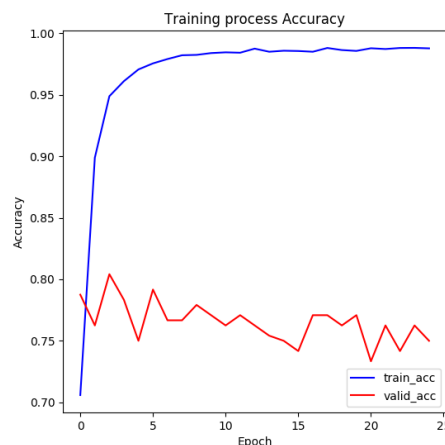
nn.Linear(2048,1024),

nn.Linear(1024,512),

nn.Linear(512,2),

nn.LogSoftmax(dim=1)

Public score: 0.77209



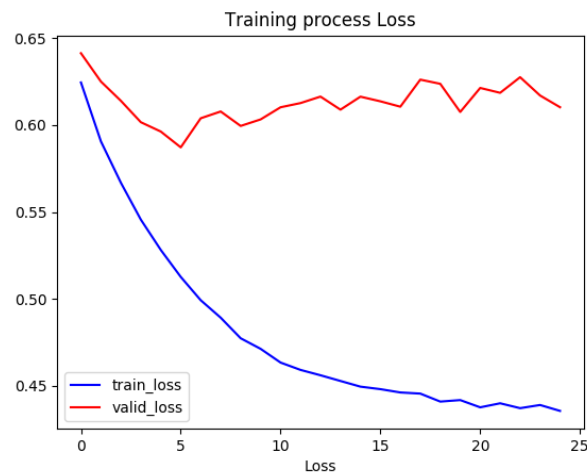
3. (1%) 請敘述你如何 improve performance (preprocess, embedding, 架構等) , 並解釋為何這些做法可以使模型進步。

LSTM 的 n_layers 太大會 train 不起來, 最後選擇為 6

embedding 的維度不是越大越好, 太高會讓 validation score 下降, 最後選擇為 25, 原因是維度太大無法確實表示單字之間的關係, 一堆沒意義的數字反而會導致 overfit

4. (1%) 請比較不做斷詞 (e.g., 用空白分開) 與有做斷詞, 兩種方法實作出來的效果差異, 並解釋為何有此差別。

不做斷詞會讓 word 的數量增加, 且有些會是沒有意義的, 那些無意義的單字會導致訓練學到的東西只是單純湊出來的, 對 testing 沒有幫助



5. (1%) 請比較 RNN 與 BOW 兩種不同 model 對於 "Today is hot, but I am happy."與"I am happy, but today is hot." 這兩句話的分數 (model output) , 並討論造成差異的原因。

| | "Today is hot, but I am happy." | "I am happy, but today is hot." |
|-----|---------------------------------|---------------------------------|
| RNN | (-0.062, -2.810) | (-0.066, -2.750) |
| BOW | (-0.009, -4.687) | (-0.009, -4.687) |

文字順序不影響 BOW 產生的 input, 這 2 個輸入對於 BOW 是一樣的

1. LSTM Cell

$$x_1 = (0, 1, 0, 3), C_1 = 0, g(z_1) = 3, f(z_1^i) = f(q_0), f(z_1^o) = f(-10) \\ C_1' = f(q_0) \times 3 + 0 \times f(z_1^f) = 3, y_1 = f(-10) \times 3 = 0_{\neq}$$

$$x_2 = (1, 0, 1, -2), C_2 = 3, g(z_2) = -2, f(z_2^i) = f(q_0), f(z_2^o) = f(q_0) \\ C_2' = f(q_0) \times -2 + 3 \times f(z_2^f) = -2 + 3 = 1, y_2 = f(q_0) \times 1 = 1_{\neq}$$

$$x_3 = (1, 1, 1, 4), C_3 = 1, g(z_3) = 4, f(z_3^i) = f(1q_0), f(z_3^o) = f(q_0) \\ C_3' = f(1q_0) \times 4 + 1 \times f(z_3^f) = 4 + 0 = 4, y_3 = f(q_0) \times 4 = 4_{\neq}$$

$$x_4 = (0, 1, 1, 0), C_4 = 4, g(z_4) = 0, f(z_4^i) = f(q_0), f(z_4^o) = f(q_0) \\ C_4' = f(q_0) \times 0 + 4 \times f(z_4^f) = 0 + 4 = 4, y_4 = f(q_0) \times 4 = 4_{\neq}$$

$$x_5 = (0, 1, 0, 2), C_5 = 4, g(z_5) = 2, f(z_5^i) = f(q_0), f(z_5^o) = f(-10) \\ C_5' = f(q_0) \times 2 + 4 \times f(z_5^f) = 2 + 4 = 6, y_5 = f(-10) \times 6 = 0_{\neq}$$

$$x_6 = (0, 0, 1, -4), C_6 = 6, g(z_6) = -4, f(z_6^i) = f(-10), f(z_6^o) = f(q_0) \\ C_6' = f(-10) \times -4 + 6 \times f(z_6^f) = 0 + 6 = 6, y_6 = f(q_0) \times 6 = 6_{\neq}$$

$$x_7 = (1, 1, 1, 1), C_7 = 6, g(z_7) = 1, f(z_7^i) = f(1q_0), f(z_7^o) = f(q_0) \\ C_7' = f(1q_0) \times 1 + 6 \times f(z_7^f) = 1 + 0 = 1, y_7 = f(q_0) \times 1 = 1_{\neq}$$

$$x_8 = (1, 0, 1, 2), C_8 = 1, g(z_8) = 2, f(z_8^i) = f(q_0), f(z_8^o) = f(q_0) \\ C_8' = f(q_0) \times 2 + 1 \times f(z_8^f) = 2 + 1 = 3, y_8 = f(q_0) \times 3 = 3_{\neq}$$

2. Word Embedding

$$\frac{\partial L}{\partial W_{ij}^T} = \sum_{k=1}^V \sum_{c=1}^C \frac{\partial L}{\partial u_{c,k}} \cdot \frac{\partial u_{c,k}}{\partial W_{ij}^T}$$

$$\frac{\partial L}{\partial u_{c,k}} = -\delta_{kk_c^*} + y_{c,k} \quad \cdot \quad \delta_{ij} = \begin{cases} 0 & \text{if } i \neq j \\ 1 & \text{if } i = j \end{cases} \quad \cdot \quad k_c^* \text{ 是 } W_c \text{ one-hot encoding 为 1 的位置}$$

$$\frac{\partial u_{c,k}}{\partial W_{ij}^T} = h_i = \sum_{k=1}^V W_{ik} x_k$$

$$\frac{\partial L}{\partial W_{ij}^T} = \sum_{c=1}^C (-\delta_{kk_c^*} + y_{c,k}) \cdot \sum_{k=1}^V W_{ki} x_k \quad \#$$

$$\frac{\partial L}{\partial W_{ij}^T} = \sum_{k=1}^V \sum_{c=1}^C \frac{\partial L}{\partial u_{c,k}} \cdot \frac{\partial u_{c,k}}{\partial W_{ij}^T}$$

$$u_{c,k} = \sum_{m=1}^N \sum_{l=1}^V W'_{mk} W_{lm} x_l$$

$$\frac{\partial u_{c,k}}{\partial W_{ij}^T} = W'_{jk} x_i$$

$$\frac{\partial L}{\partial W_{ij}^T} = \sum_{k=1}^V \sum_{c=1}^C (-\delta_{kk_c^*} + y_{c,k}) W'_{jk} x_i \quad \#$$