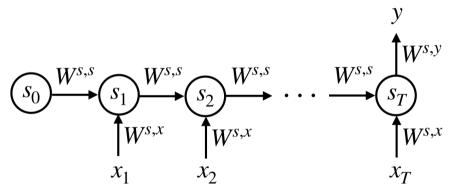
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Problem 6



Recurrent neural networks (RNN) can be used as classification models for time series data. Here we have a simple RNN as shown in the figure above, where

$$s_t = f_1\left(W^{s,s}s_{t-1} + W^{s,x}x_t
ight), \quad t = 1,2,\ldots,T$$

and

$$y=f_{2}\left(W^{s,y}s_{T}+W_{0}
ight)$$

We assume all offsets are 0 except W_0 for the final output layer and we decide the two activation functions to be:

$$f_{1}\left(z
ight) =\operatorname{RELU}\left(z
ight) =\max\left(0,z
ight)$$

and

$$f_{2}\left(z
ight)=\mathrm{sign}\left(z
ight)=\left\{egin{array}{ll} 1, & if & z\geq0\ 0, & if & z<0 \end{array}
ight.$$

Note that the $\operatorname{RELU}(z)$ can be applied elementwise if z is a vector.

Suppose we want to apply this model to classify sentences into different categories (e.g. positive/negative sentiment), we need to encode each word in a sentence into a vector as the input x_t to the model. One way to do this is to represent the tth word as a column vector of length |V|, where V is the set of the entire vocabulary. The ith element of x_t is 1 if the word is the ith word in the vocabulary and all other elements are zero.

6. (1)

2.0/2 points (graded)

We first explore a simple scenario where our vocabulary contains only 2 words, $V=\{A,B\}$. Let $s_t\in\mathbb{R}^2$ and we set the initial state s_0 and the weights before the last layer as follows:

$$s_0 = egin{bmatrix} 0 \ 0 \end{bmatrix}, \quad W^{s,s} = egin{bmatrix} -1 & 0 \ 0 & 1 \end{bmatrix}, \quad W^{s,x} = egin{bmatrix} 1 & 0 \ 0 & 1 \end{bmatrix}$$

Now given 3 training sentences: AA, ABB, BAA

Encode each of them into a sequence of vectors. As an example, the sentence AA is encoded as $x^{(1)}=(x_1^{(1)},x_2^{(1)})$, where $x_1^{(1)}=x_2^{(1)}=[1,0]^T$.

(To enter the sequence above, type [[1,0],[1,0]].)

Now encode the other 2 sentences into $x^{(2)}$ and $x^{(3)}$.

Solution:

A is encoded as $\left[1,0
ight]^T$ and B is encoded as $\left[0,1
ight]^T$, so we have

$$x^{(2)} \; = (x_1^{(2)}, x_2^{(2)}, x_3^{(2)}) = ([1, 0]^T, [0, 1]^T, [0, 1]^T)$$

$$x^{(3)} \; = (x_1^{(3)}, x_2^{(3)}, x_3^{(3)}) = ([0,1]^T, [1,0]^T, [1,0]^T)$$

You have used 1 of 5 attempts

1 Answers are displayed within the problem

6. (2)

3.0/3 points (graded)

Now compute the final hidden state $s_T^{(1)}$, $s_T^{(2)}$, $s_T^{(3)}$ for each of the three senteces AA, ABB, BAA in this RNN.

(Enter [0,0] for $S_T = \left[0,0
ight]^T$.)

$$s_T^{(1)} = \boxed{ \left[ext{0,0}
ight]}$$
 $ightharpoons 200$ Answer: $\left[ext{0,0}
ight]$

$$s_T^{(2)} = \boxed{ ext{[0,2]} }$$

Solution:

Using $s_t = RELU\left(W^{s,s}s_{t-1} + W^{s,x}x_t
ight)$, we can compute

$$s_T^{\left(1
ight)} \ = \left[0,0
ight]^T$$

$$s_T^{(2)} \; = [0,2]^T$$

$$s_{T}^{(3)} \ = \left[0,1
ight]^{T}$$

You have used 1 of 5 attempts

1 Answers are displayed within the problem

6. (3)

0/1 point (graded)

Fixing s_0 , $W^{s,s}$, $W^{s,x}$ and by only learning the linear classifier in the final layer, can this RNN separate the 3 examples regardless of how they were labeled?





Solution:

No. As the final layer is a linear classifier and $s_T^{(1)}$, $s_T^{(2)}$, $s_T^{(2)}$ are collinear points, they are not linear seperable in general. A concrete example is when $y^{(1)}=y^{(2)}=1$ and $y^{(3)}=-1$.

You have used 2 of 3 attempts

1 Answers are displayed within the problem

6. (4)

1/1 point (graded)

A simpler model to classify sentences is to represent the entire sentence into a vector z and apply a linear model on z, i.e.

$$y=sign\left(W^{z,y}z
ight)$$

The vector z has length |V| and the ith element of z is the count of how many times the ith word appears in the sentence. For example, the sentence ABA with $V=\{A,B\}$ will be encoded as $z=[2,1]^T$. If we want the RNN we described earlier to match the output of this linear model given any input sentences, Which of the following is a possible setting of the weights and initial state s_0 of the RNN? Check all that apply.

Here $c^{|V|}$ stands for a vector of length |V| in which every element is c (e.g. $[1,1,1]^T$ if c=1 and |V|=3), $I_{|V|}$ stands for the identity matrix of size |V|.

$$lacksquare s_0=1^{|V|}$$
 , $W^{s,s}=I_{|V|}$, $W^{s,x}=I_{|V|}$, $W^{s,y}=-W^{z,y}$, $W_0=\sum_i W_i^{z,y}$

$$lacksquare s_0=1^{|V|}$$
 , $W^{s,s}=I_{|V|}$, $W^{s,x}=I_{|V|}$, $W^{s,y}=W^{z,y}$, $W_0=-\sum_i W_i^{z,y}$ 🗸

$$lacksquare s_0=0^{|V|}$$
 , $W^{s,s}=I_{|V|}$, $W^{s,x}=-I_{|V|}$, $W^{s,y}=-W^{z,y}$, $W_0=0$

$$extbf{ extit{Y}} \; s_0 = 0^{|V|}$$
 , $W^{s,s} = I_{|V|}$, $W^{s,x} = I_{|V|}$, $W^{s,y} = W^{z,y}$, $W_0 = 0$ 🗸



Solution:

By setting $W^{s,s}=I_{|V|}$ and $W^{s,x}=I_{|V|}$, we have $s_t=RELU$ $(s_{t-1}+x_t)$. If we initialize $s_0=0^{|V|}$, then s_T will be the same as z, thus $W^{s,y}=W^{z,y}$ and $W_0=0$. If we initialize $s_0=1^{|V|}$, then $s_T=z+1^{|V|}$. To make $W^{z,y}z=W^{s,y}$ $(z+1^{|V|})+W_0$, we have $W^{s,y}=W^{z,y}$ and $W_0=-W^{s,y}\cdot 1^{|V|}=-\sum_i W_i^{z,y}$

Submit

You have used 1 of 3 attempts

1 Answers are displayed within the problem

6. (5)

1/1 point (graded)

Now suppose we want to use indicators instead of counts for the vector z. That is the ith element of z will be 1 if the ith word appears anywhere in the sentence. Which of the following is a possible setting of the weights and initial state s_0 of the RNN? Check all that apply.

Note (Sept 8): In the choices below, W_0 is a scalar, and the summation $\sum_i W_i^{z,y}$ over i is summing over the elements of the vector $W^{z,y}$.

$$lacksquare s_0=1^{|V|}$$
 , $W^{s,s}=I_{|V|}$, $W^{s,x}=-I_{|V|}$, $W^{s,y}=W^{z,y}$, $W_0=-\sum_i W_i^{z,y}$

$$otin s_0 = 1^{|V|}$$
 , $W^{s,s} = I_{|V|}$, $W^{s,x} = -I_{|V|}$, $W^{s,y} = -W^{z,y}$, $W_0 = \sum_i W_i^{z,y}$ 🗸

$$lacksquare s_0=0^{|V|}$$
 , $W^{s,s}=I_{|V|}$, $W^{s,x}=-I_{|V|}$, $W^{s,y}=-W^{z,y}$, $W_0=0$

$$lacksquare s_0=0^{|V|}$$
 , $W^{s,s}=I_{|V|}$, $W^{s,x}=I_{|V|}$, $W^{s,y}=W^{z,y}$, $W_0=0$



Solution:

As the RELU activation function will map all non-positive input to 0, so whenever a word appears, we can minus the corresponding state by 1. If we set the initial state to be 1, representing a word does not appear, then as long as a word appears, no matter how many times, the state will be 0.

With this idea, we choose $s_0=1^{|V|}$, $W^{s,s}=I_{|V|}$ and $W^{s,x}=-I_{|V|}$. By doing so, we have $s_T=1^{|V|}-z$.

To make $W^{z,y}z=W^{s,y}\left(1^{|V|}-z
ight)+W_0$,

We have $W^{s,y} = -W^{z,y}$ and $W_0 = \sum_i W_i^{z,y}$.

You have used 1 of 3 attempts

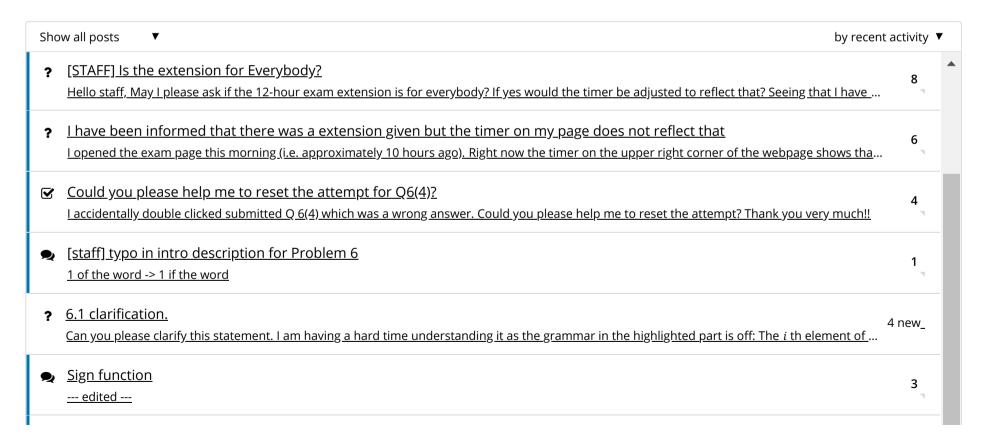
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