

Applied Deep Learning

Chapter 6: Recurrent NNs

Ali Bereyhi

`ali.bereyhi@utoronto.ca`

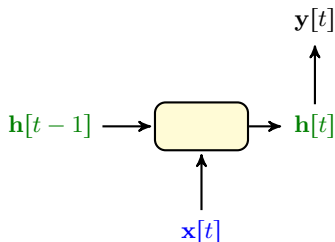
Department of Electrical and Computer Engineering
University of Toronto

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Bidirectional RNNs

We have up to now considered *unidirectional* RNNs

we *start from beginning* of the sequence and *move in one direction*



But, can't we *learn from future* input as well?

Bidirectional RNNs

Future entries can have information about **past**: say our RNN wants to fill the empty field

... the color that many people assume is the color of **sun** ...

Obviously, **future** input in the sequence is helping in this example!

- + But, how can we **get information** from future?
- Well, we have **the whole sequence**: we could move **once from left to right** and **once from right to left**

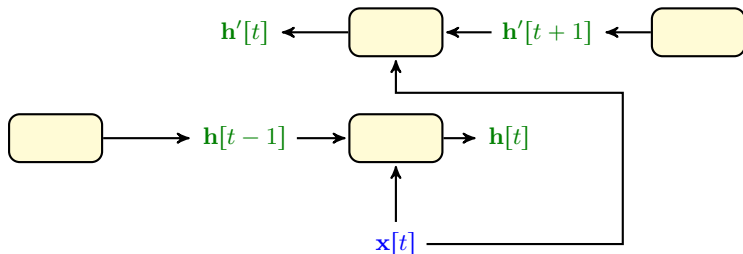
Bidirectional RNNs

Bidirectional RNNs

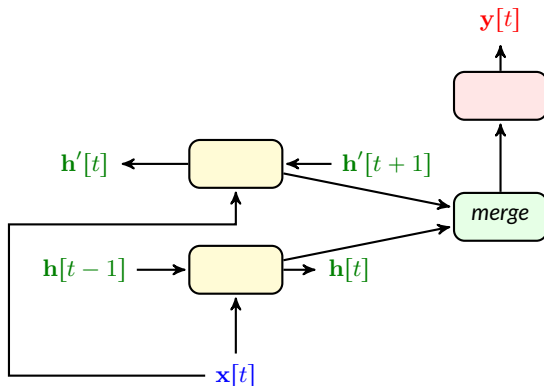
A *bidirectional* RNN (BRNN) consists of two RNNs

- one that starts with an *initial hidden state* at $t = 0$ and computes $\mathbf{h}[t]$ from $\mathbf{h}[t - 1]$ and $\mathbf{x}[t]$
- another that starts with an *initial hidden state* at $t = T + 1$ and computes $\mathbf{h}'[t]$ from $\mathbf{h}'[t + 1]$ and $\mathbf{x}[t]$

Output at time t is determined from *merged* version of the two hidden states

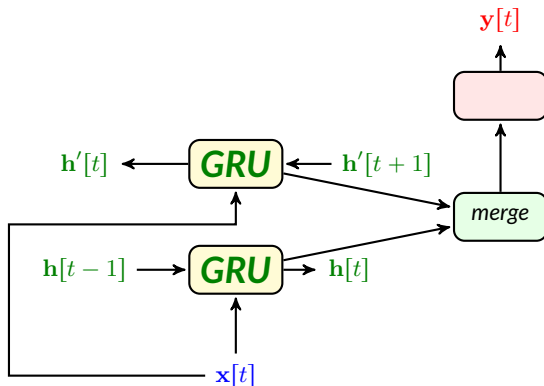


Bidirectional RNNs



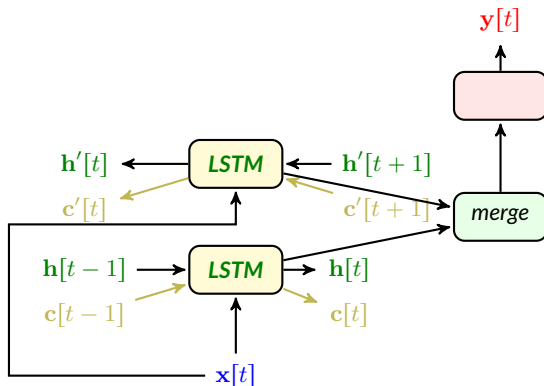
- + What exactly is this *merge* block?
- It gets the *two states* and returns a vector that matches *output layer*

Bidirectional GRU



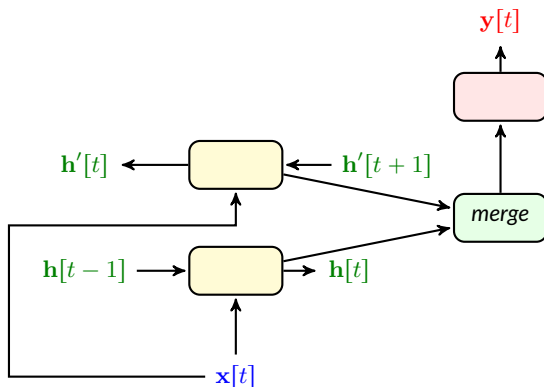
- + Should we use *any RNN* here?
- Sure! We may use *GRU*

Bidirectional LSTM



- + Should we use **any RNN** here?
- Sure! We may use **LSTM**

RNNs in PyTorch



- + Any suggestion for *merging* the *hidden states*?
- Sure! Let's see some code

RNNs in PyTorch

In PyTorch, we can access a *basic RNN* in `torch.nn` module

```
torch.nn.RNN()
```

We can make it *deep* by simply choosing `num_layers` *more than one* and *bidirectional* by setting `bidirectional` to `True`. Same with *GRU* and *LSTM*

```
torch.nn.LSTM()
```

```
torch.nn.GRU()
```

In *bidirectional* case, we get access to *both states*. To *merge* them, we could

- add the two states
- average them
- concatenate them, i.e., $\mathbf{h}_c[t] = (\mathbf{h}[t], \mathbf{h}'[t])$

or do *any other operation* that we find useful