

Applied Deep Learning

Chapter 6: Recurrent NNs

Ali Bereyhi

ali.bereyhi@utoronto.ca

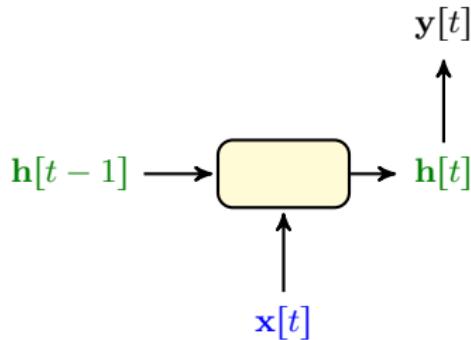
Department of Electrical and Computer Engineering
University of Toronto

Fall 2025

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

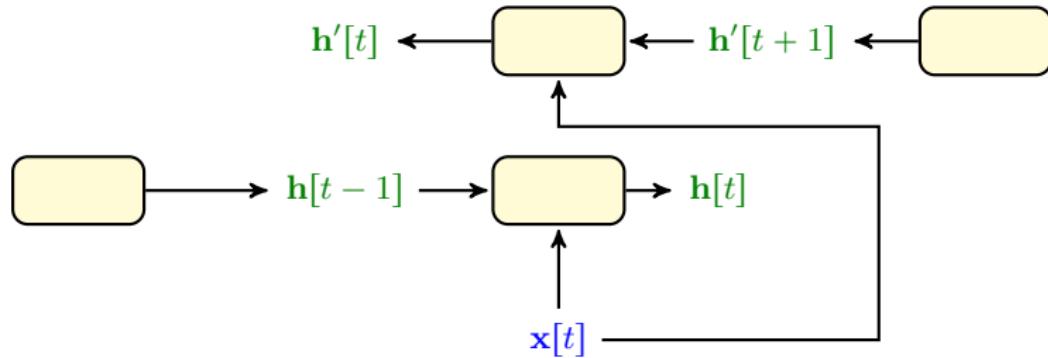
Bidirectional RNNs

Bidirectional RNNs

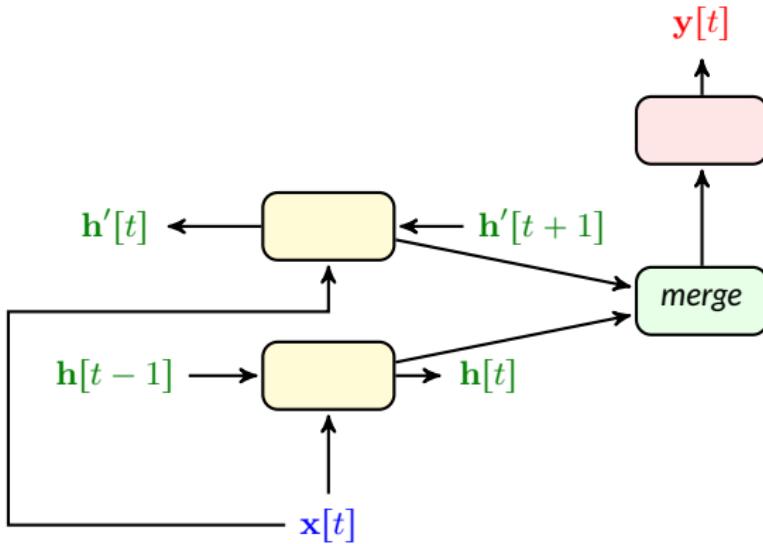
A **bidirectional** RNN (BRNN) consists of two RNNs

- 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

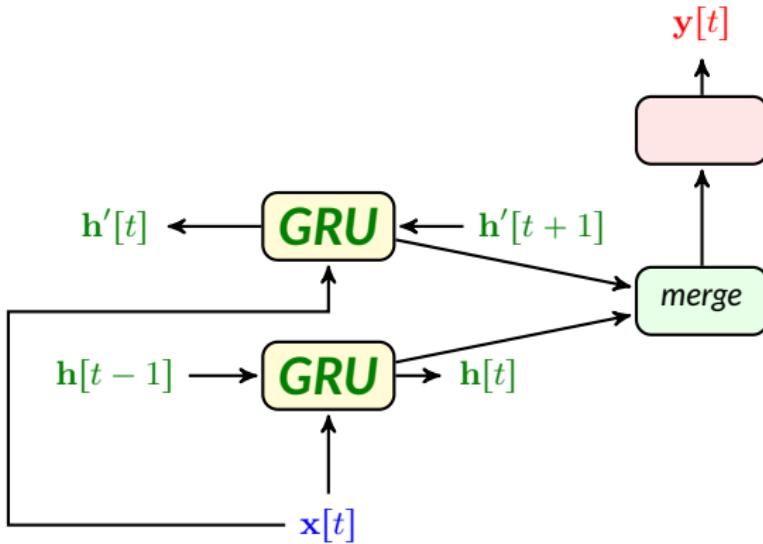


Bidirectional RNNs



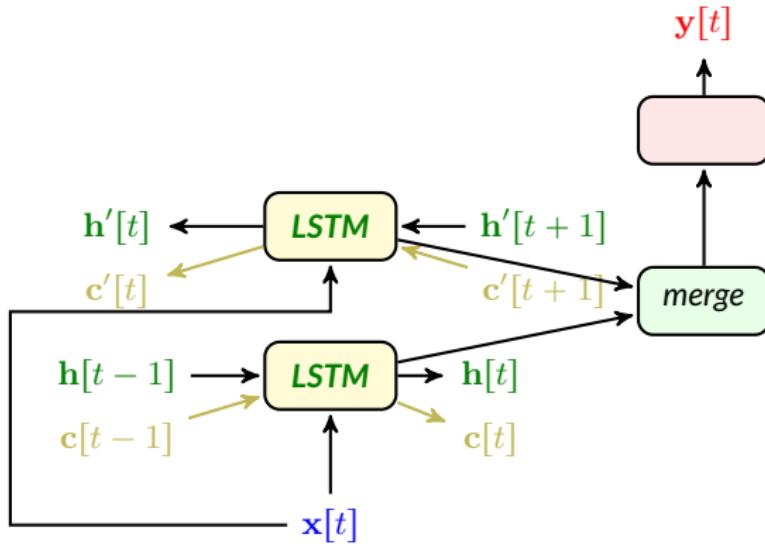
- + What exactly is this *merge* block?
 - It gets the **two states** and *returns a vector that matches output layer*

Bidirectional GRU



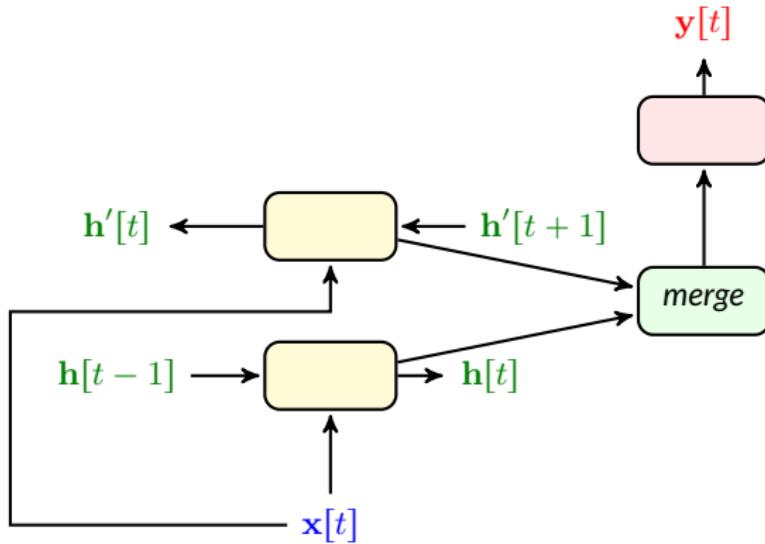
- + Should we use *any RNN* here?
- Sure! We may use *GRU*

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