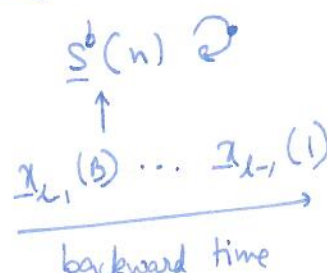
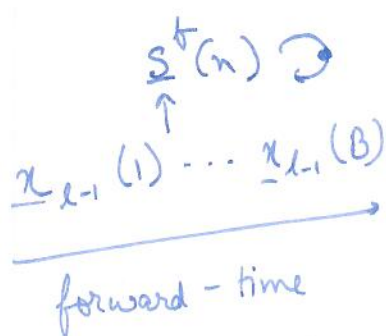


8.2 Bidirectional Recurrent Neural Network

Causal and non-causal systems

RNN is a causal system as it is dependent on past and present.

$$8-12 \quad \underline{x}_L(n) = \begin{bmatrix} \underline{s}_L^f(n) \\ \underline{s}_L^b(B-n) \end{bmatrix} \quad 1 \leq n \leq B$$



~~Q How is the above eqn possible due to~~

8-13

Training: backpropagation of derivatives through the unfolded bidirectional graph.

8.3 Long Short Term Memory (LSTM)

RNN is more difficult to train than CNN due to the exploding/vanishing gradient

DNN in Ch 5 and 7: large $L = \#$ layers

→ long chain rule over layers

RNN

: large $B = \#$ time recursions

→ long chain rule over time in addition

Difficult choice of B :

$B \downarrow$: noisy gradient, short memory

$B \uparrow$: vanishing/exploding gradient

Solution : LSTM - special RNN

8-14

forget gate - clear the memory for next time
 input gate - to overwrite the input or not
 output gate - to control if the output is to be obtained or not

~~8-15~~

→ calculated by machine

8-15

why sigmoid $\sigma \in (0, 1) \rightarrow$ so $0 \rightarrow$ gate closed
 $1 \rightarrow$ gate opened

8-16

There are in total 7 operations in each LSTM cell/layer.

Effects of gating

$0 < \text{gate signal} < 1$
 ≈ 0
 ≈ 1

input/output gate	forget gate
close gate	clear memory
open gate	keep memory

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8-18

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8-20

⋮

8-23

9 Autoencoders and generative models

9.1 Autoencoder

- 9-1
- 9-2
- 9-3
- 9-4

Autoencoder (AE):

an unsupervised learned NN to learn an efficient representation/code of the input.

