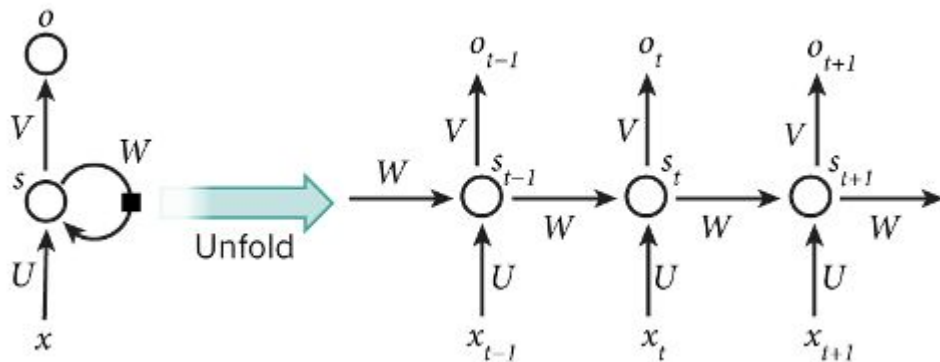


# **Natural Language Processing (CS5803)**

Lecture 6  
(Recurrent Neural Networks)

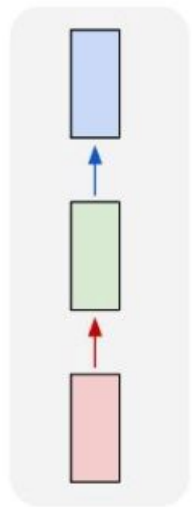
# RNNs

- A neural network architecture for modeling recurrence
- Input (previous state  $s_{t-1}$ , current observation  $x_t$ )
- Produces (current state  $s_t$ , current output  $o_t$ )
- Can be unrolled through time
- Useful for sequential data

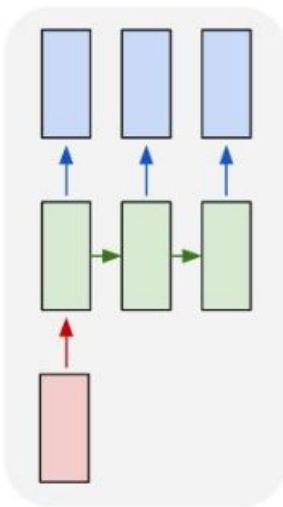


# Different Flavors of Sequence Modeling

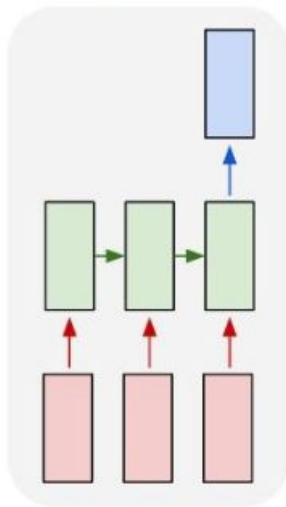
one to one



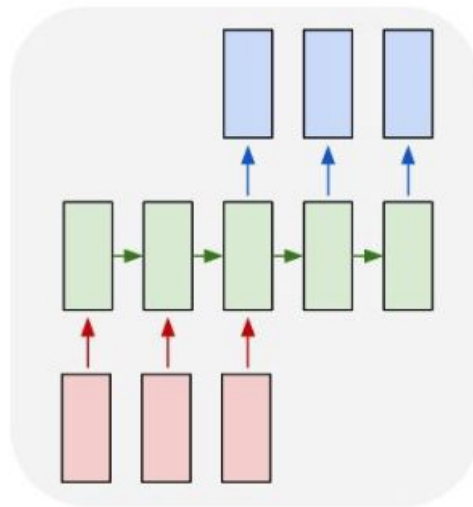
one to many



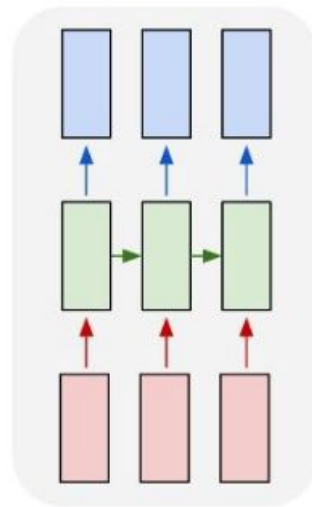
many to one



many to many

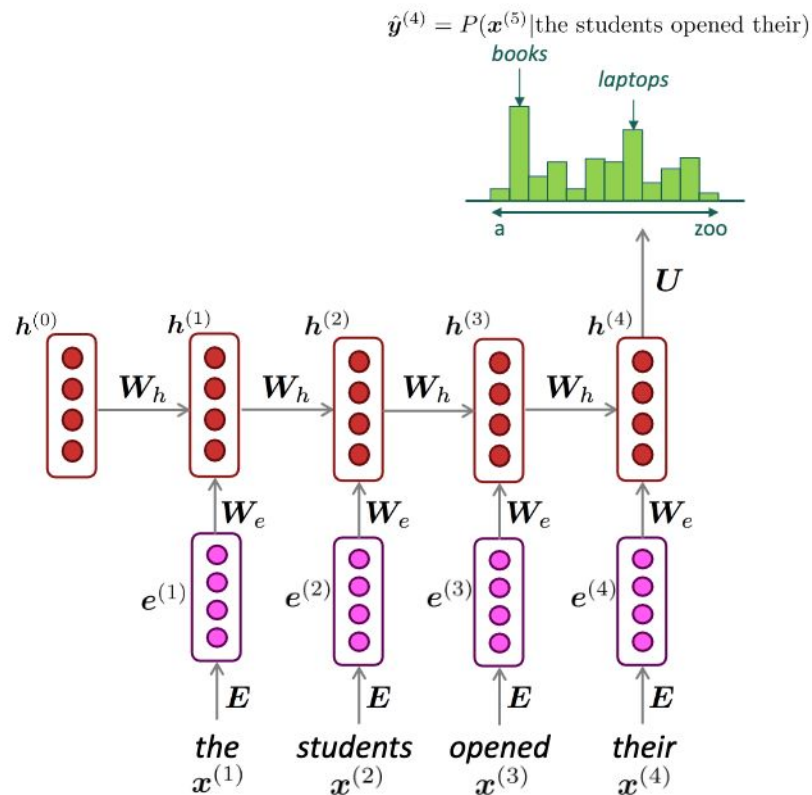


many to many



# RNN for language modeling

- Input: tokens in sequence
- Output: next token

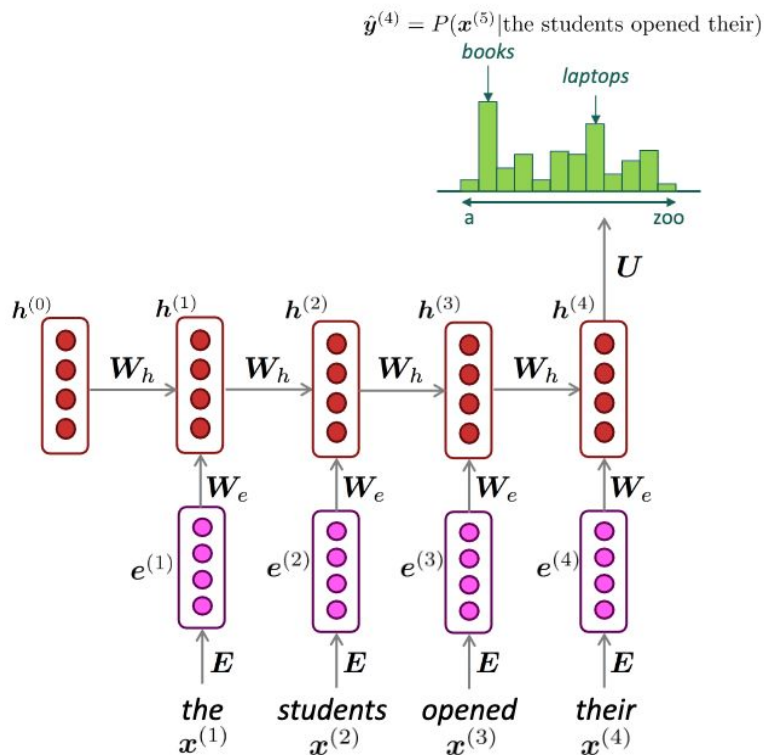


# How to Train?

- Get a large collection of documents
- Train for predicting next words for each sequence from the corpus
- Predict the next word, given the words seen so far
  - Do it for every word
- Loss function?

$$J^{(t)}(\theta) = CE(\mathbf{y}^{(t)}, \hat{\mathbf{y}}^{(t)}) = - \sum_{w \in V} \mathbf{y}_w^{(t)} \log \hat{\mathbf{y}}_w^{(t)} = - \log \hat{\mathbf{y}}_{x_{t+1}}^{(t)}$$

- This objective is called Language Modeling Objective



# RNN and LSTM

Will use [this reference](#).