

# IN3063/INM702 - Programming and Mathematics for Artificial Intelligence

10.B - RNNs

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# A lot of data has a temporal dimension

- As humans, all our experiences have temporality
  - Moving objects
  - Ego-motion
  - Causality



# Humans learn naïve physics

- When very young, we don't have a notion of naïve physics

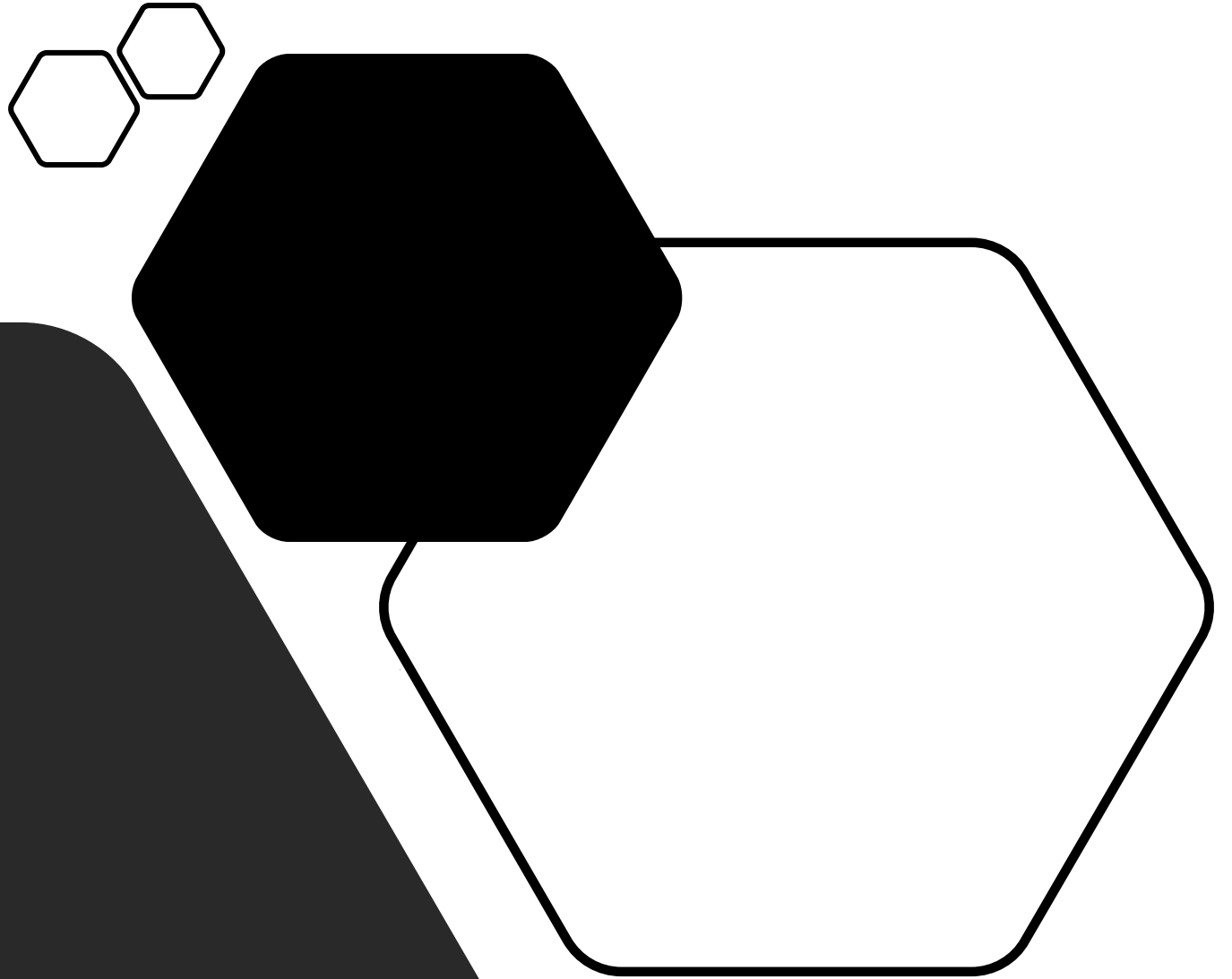


# Sound, and sequences of words

- Temporal order of sounds matters a lot
- Sequence of words
- Sequence of letters

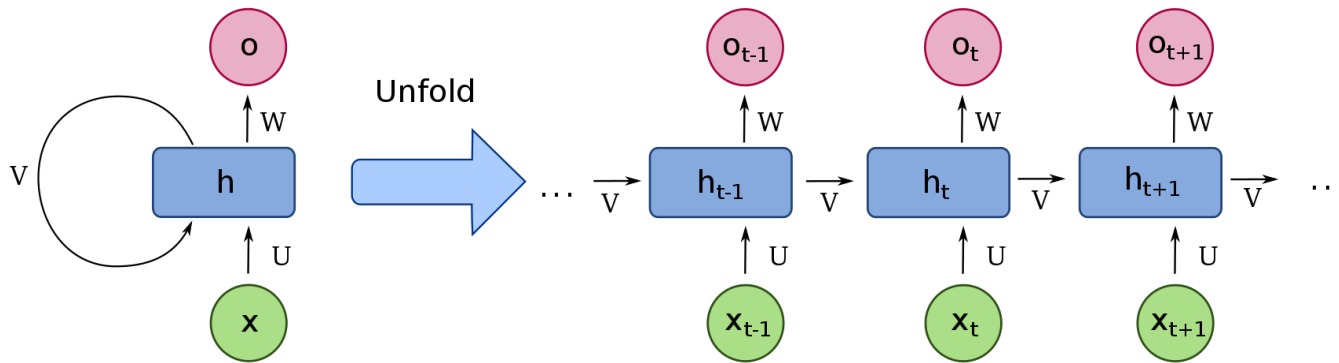


# Learning sequences



# Recurrent Neural Networks

- Transition from one time-step to the next

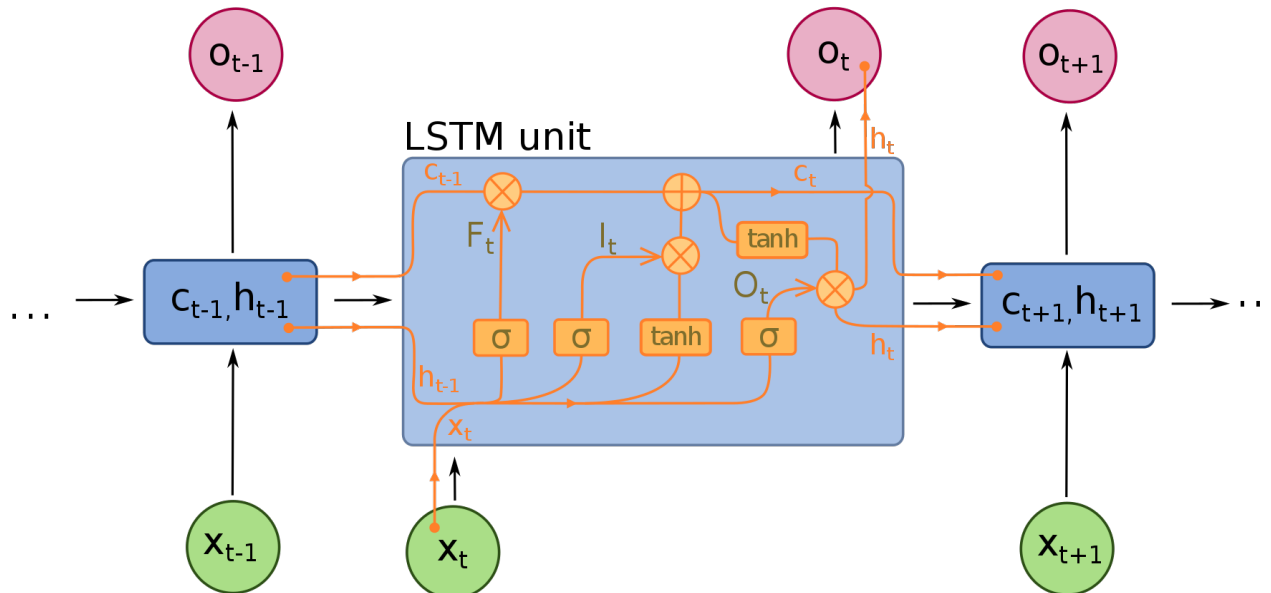


$$h_t = \sigma_h(W_h x_t + U_h h_{t-1} + b_h)$$
$$y_t = \sigma_y(W_y h_t + b_y)$$

- Limitations: Vanishing gradient

# Long Short-Term Memory

- Input gate: regulates the input
- Forget gate: regulates what is kept in memory
- Output gate: regulates what is useful for current output



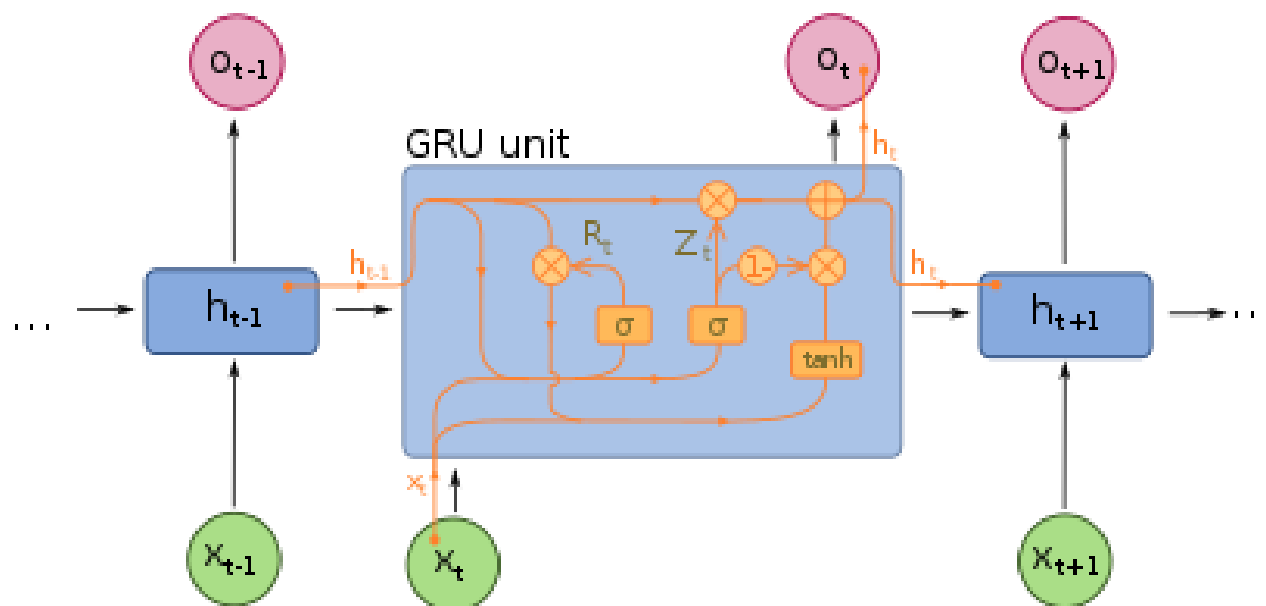
$$\begin{aligned}f_t &= \sigma_g(W_f x_t + U_f h_{t-1} + b_f) \\i_t &= \sigma_g(W_i x_t + U_i h_{t-1} + b_i) \\o_t &= \sigma_g(W_o x_t + U_o h_{t-1} + b_o)\end{aligned}$$



$$\begin{aligned}\tilde{c}_t &= \sigma_c(W_c x_t + U_c h_{t-1} + b_c) \\c_t &= f_t \circ c_{t-1} + i_t \circ \tilde{c}_t \\h_t &= o_t \circ \sigma_h(c_t)\end{aligned}$$

# GRU

- Simplified LSTM
- Only two gates: update and reset gates



$$\hat{h}_t = \phi_h(W_h x_t + U_h (r_t \odot h_{t-1}) + b_h)$$
$$h_t = (1 - z_t) \odot h_{t-1} + z_t \odot \hat{h}_t$$



# RNNs in PyTorch

- Already Implemented
- But, take care about the order:
  - Sequence length, Batch size, Input size

`nn.RNN`

Applies a multi-layer Elman RNN with `tanh` or `ReLU` non-linearity to an input sequence.

`nn.LSTM`

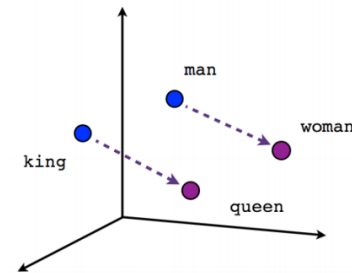
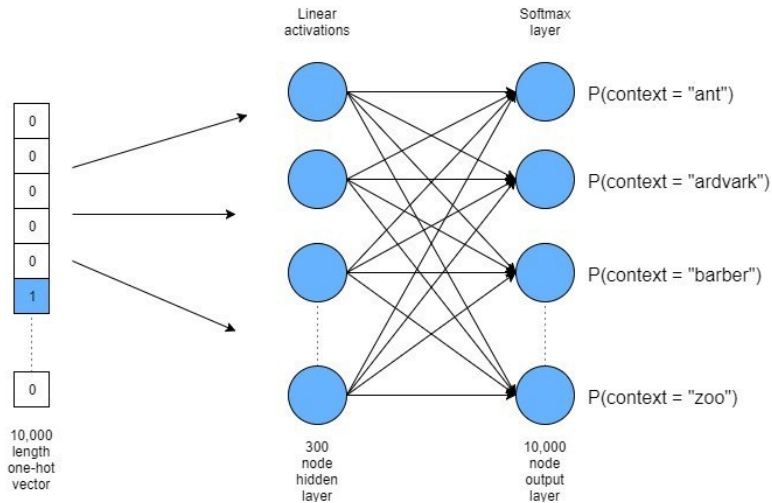
Applies a multi-layer long short-term memory (LSTM) RNN to an input sequence.

`nn.GRU`

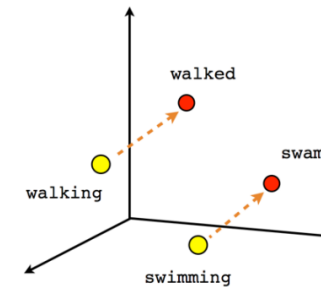
Applies a multi-layer gated recurrent unit (GRU) RNN to an input sequence.

# Word embeddings

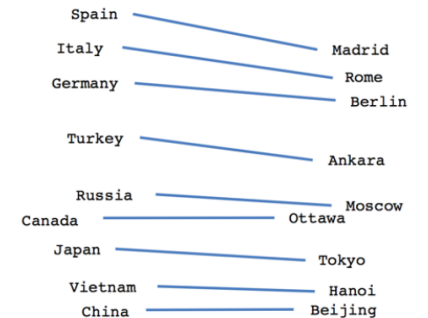
- Convert a word into a vector
- Use sequences of text to learn an embedding



Male-Female

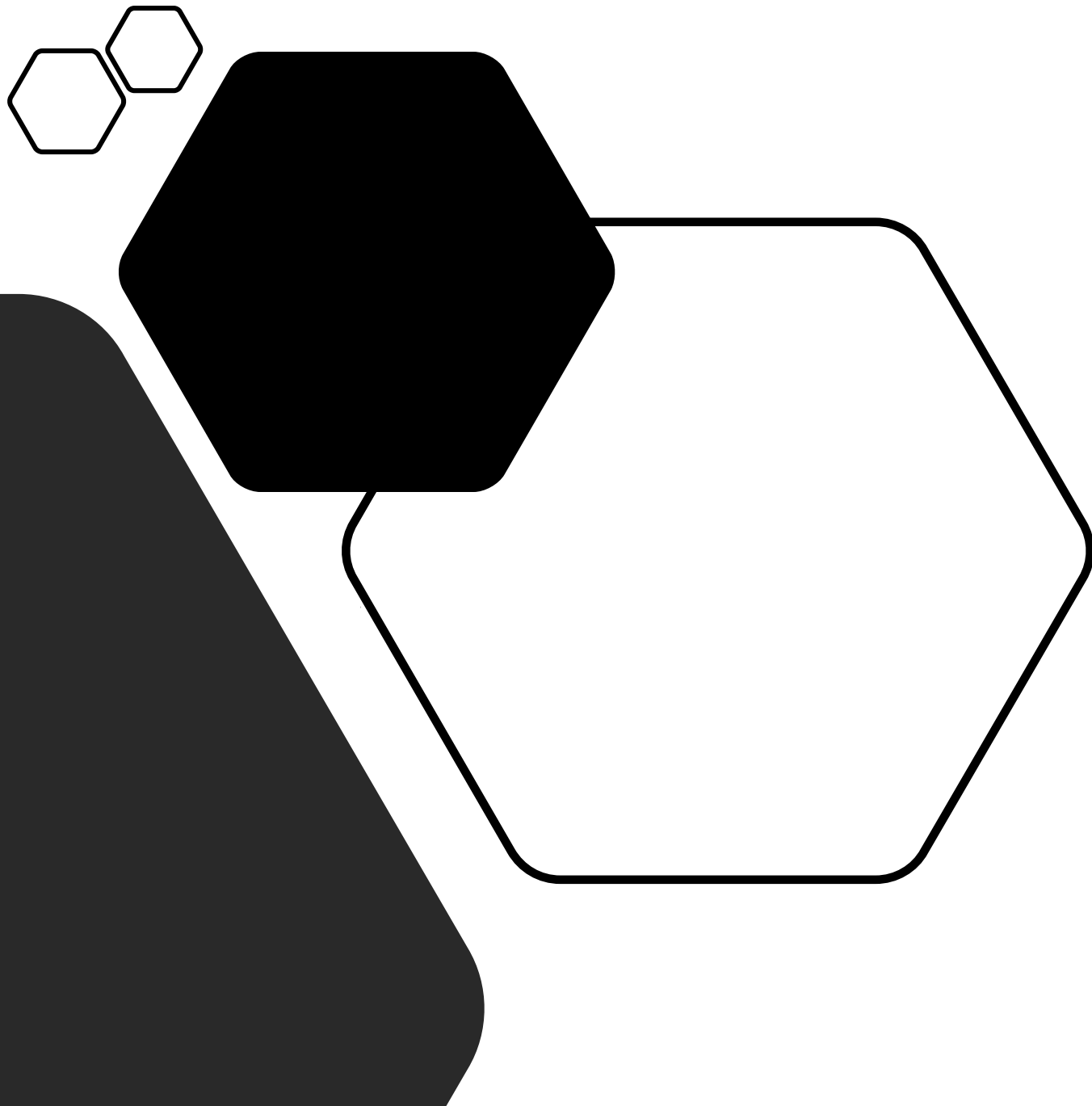


Verb tense



Country-Capital

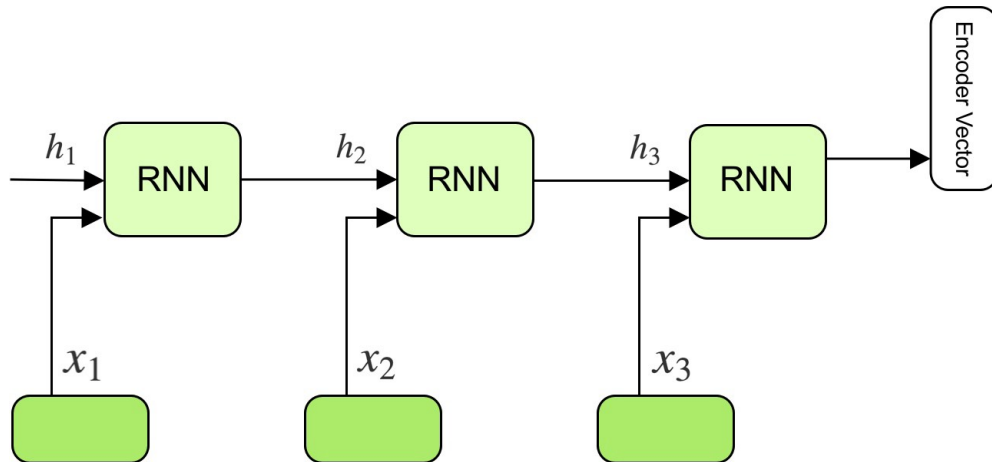
# Sequence Learning problems



# Sequence encoding

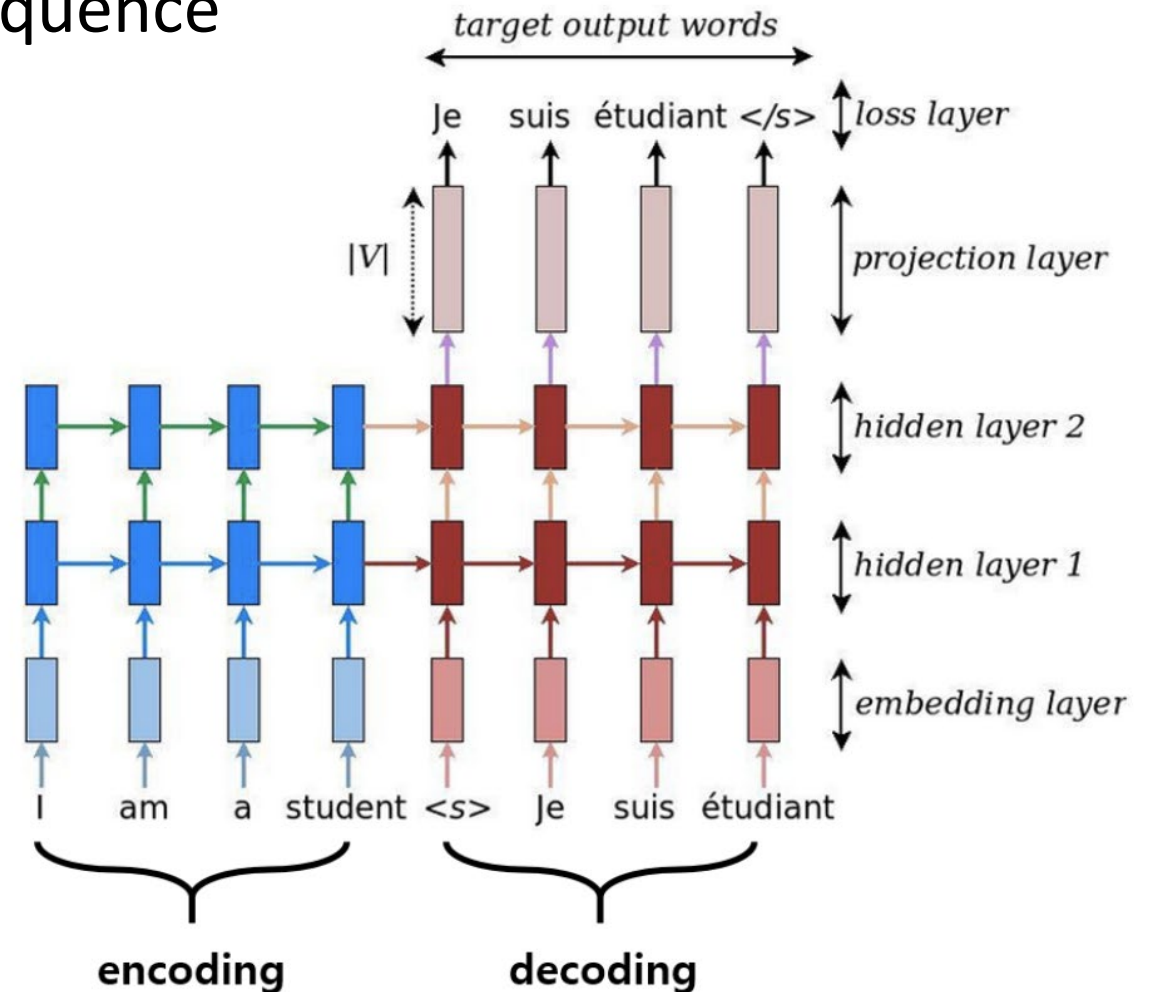
- Encode a sequence into a vector of representation
- Example: sentiment analysis

## Encoder



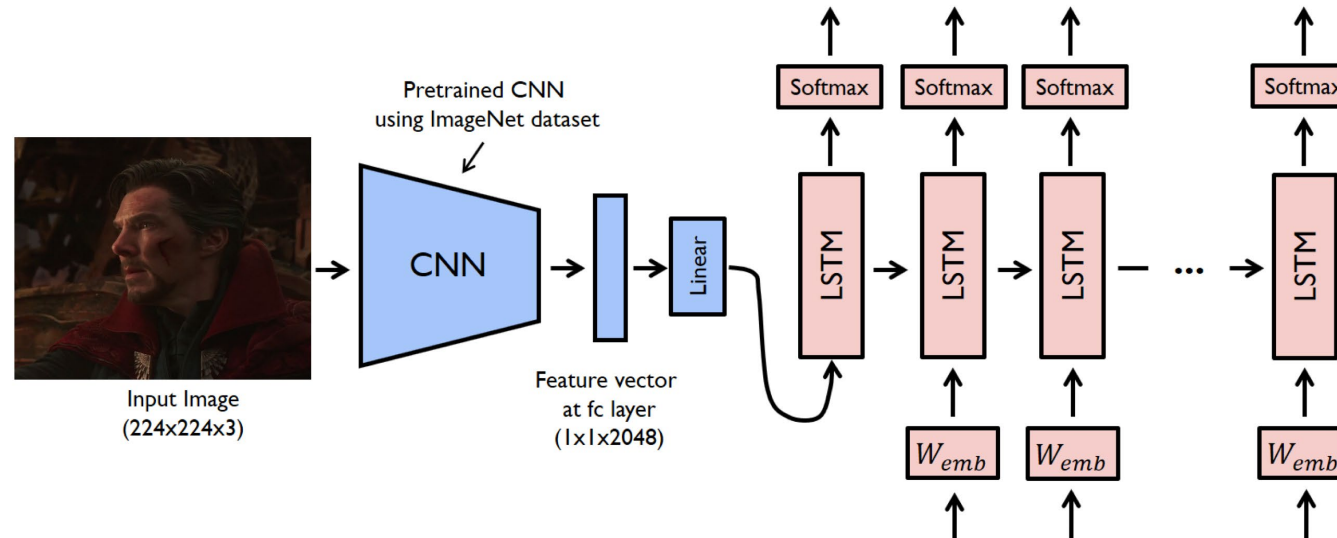
# Encoder-decoder architectures

- Encode a sequence, decode a sequence
- Example: translation



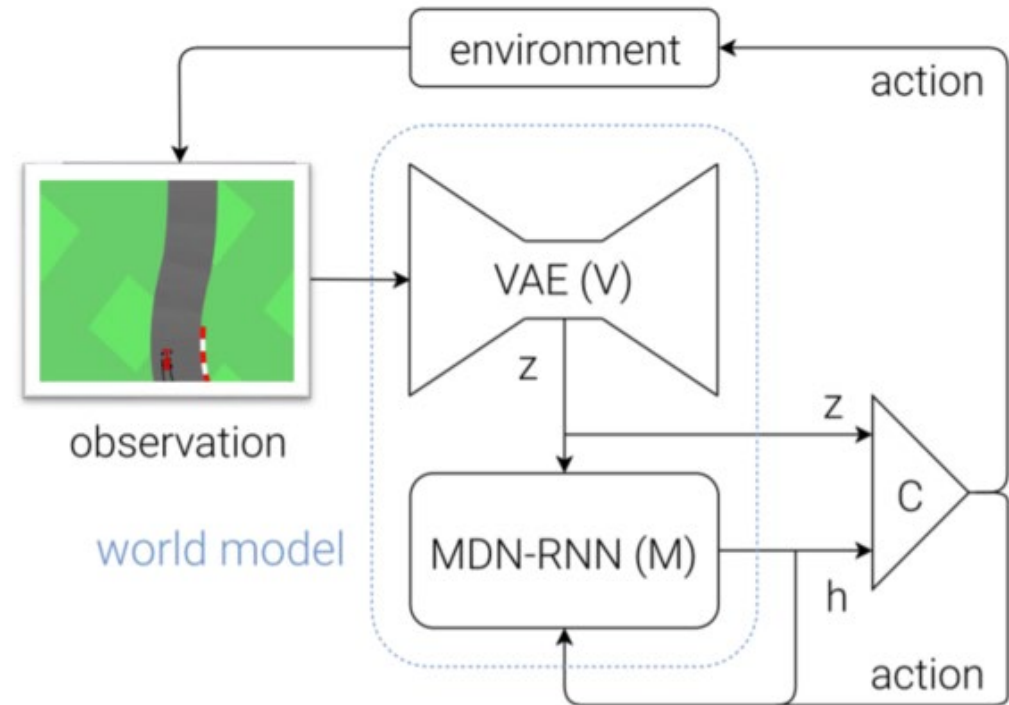
# Sequence generation

- Example: Image captioning

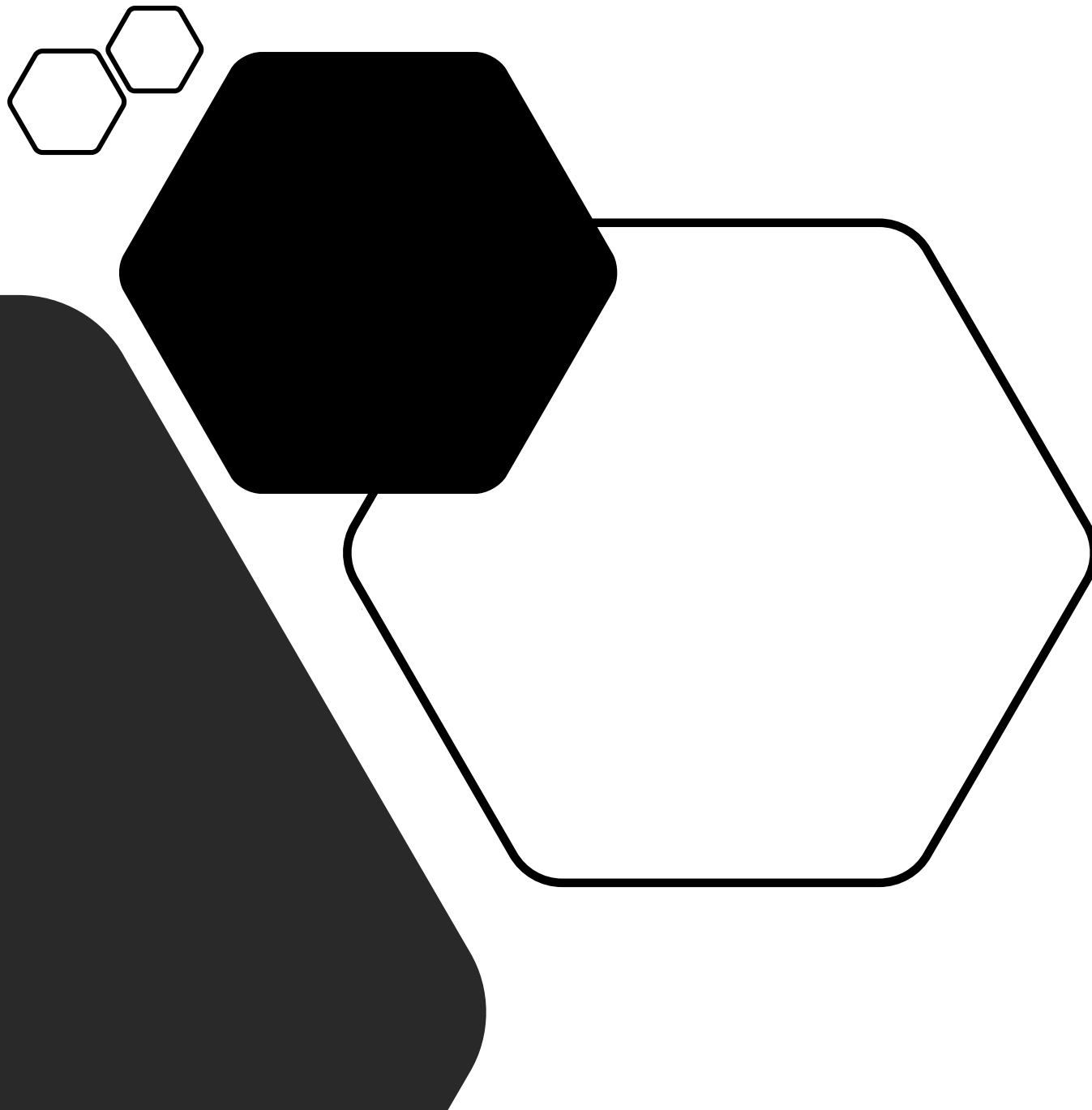


# State encoding

- An agent moves around and tries to solve tasks
- Sequence of past observations is much more informative than still image
- We can learn the state of an agent



# Lab 10 – 2





# Lab 10 – part 2

- IMDB sentiment classification
- 25000 training points / 25000 testing points
- Sentences in English associated with good/bad review
- (I will not provide solutions for this one)

