

Chapter ML:IX (continued)

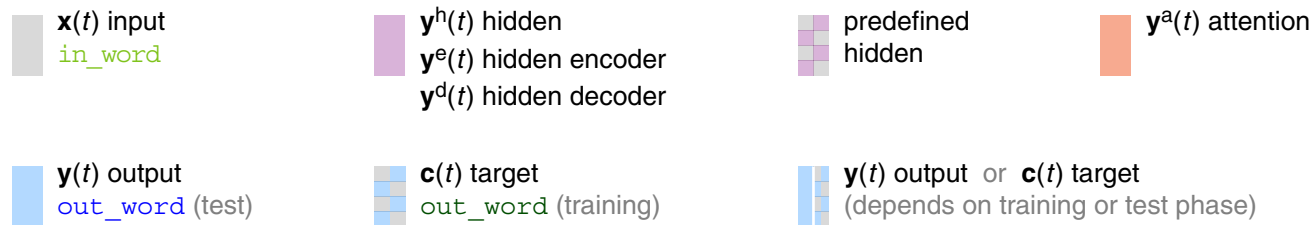
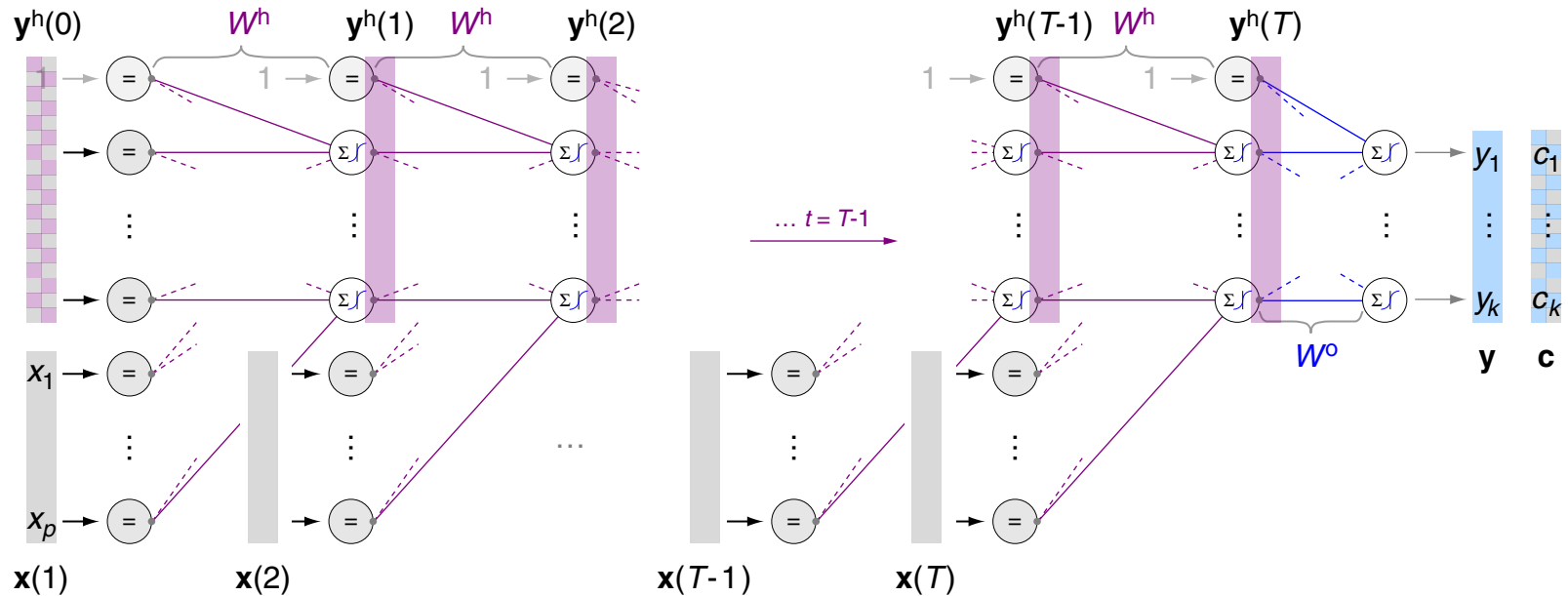
IX. Deep Learning

- ❑ Elements of Deep Learning
- ❑ Convolutional Neural Networks
- ❑ Autoencoder Networks
- ❑ Recurrent Neural Networks
- ❑ Long-Term Dependencies
- ❑ RNNs for Machine Translation
- ❑ Attention Mechanism
- ❑ Self Attention and Transformers
- ❑ Transformer Language Models

Long-Term Dependencies

Notation II (computational graph)

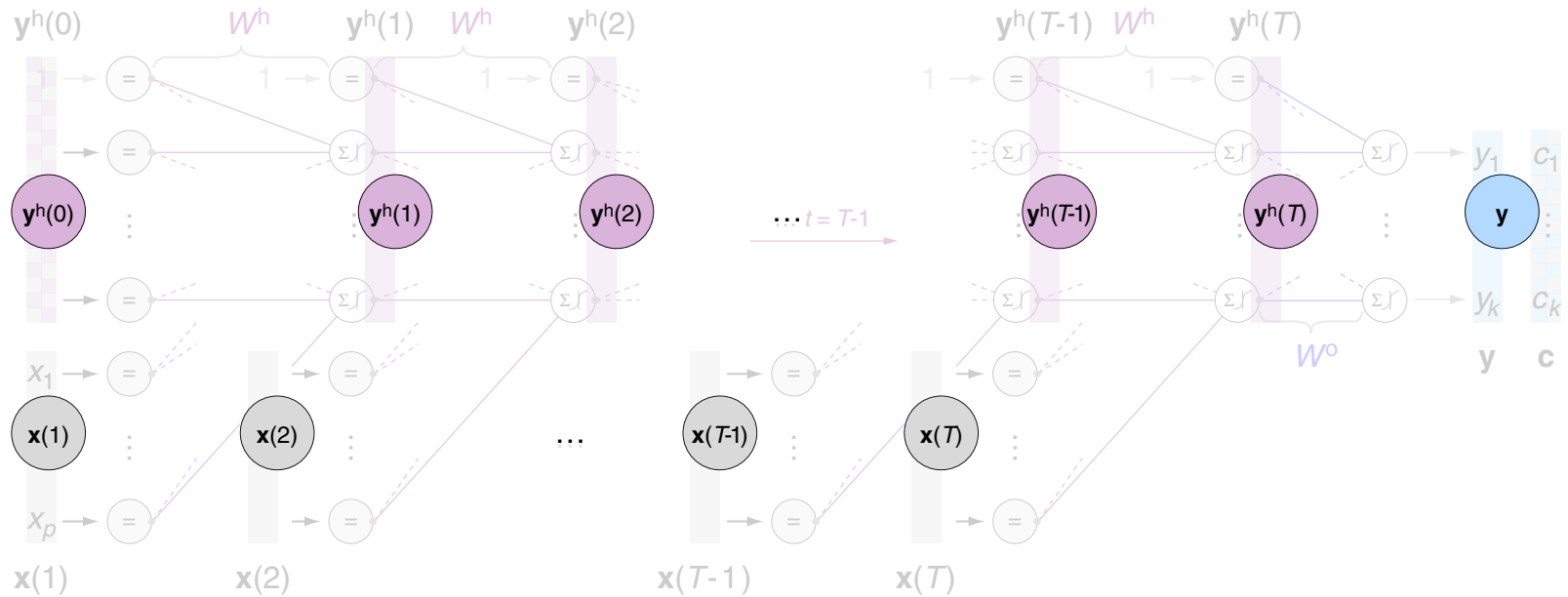
[notation: color, graph, language]



Long-Term Dependencies

Notation II (computational graph)

[notation: color, graph, language]



$x(t)$ input
in_word

$y^h(t)$ hidden
 $y^e(t)$ hidden encoder
 $y^d(t)$ hidden decoder

predefined
hidden

$y^a(t)$ attention

$y(t)$ output
out_word (test)

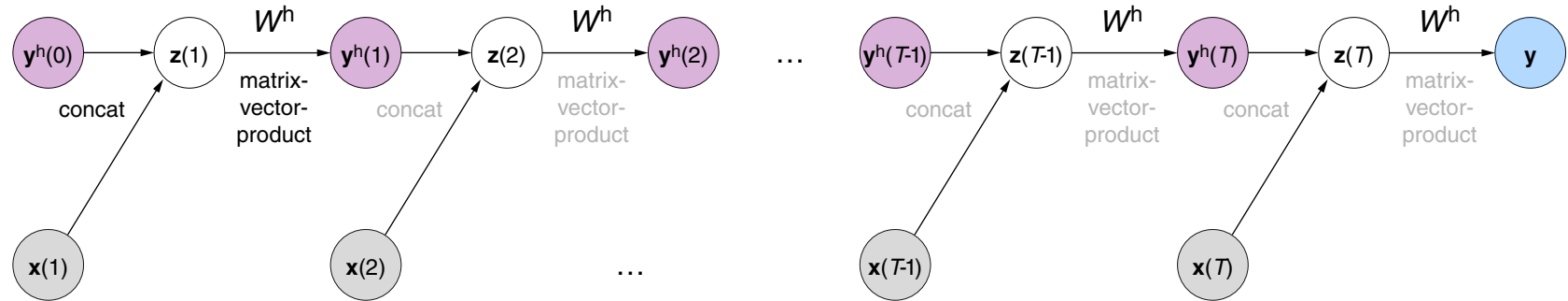
$c(t)$ target
out_word (training)

$y(t)$ output or $c(t)$ target
(depends on training or test phase)

Long-Term Dependencies

Notation II (computational graph)

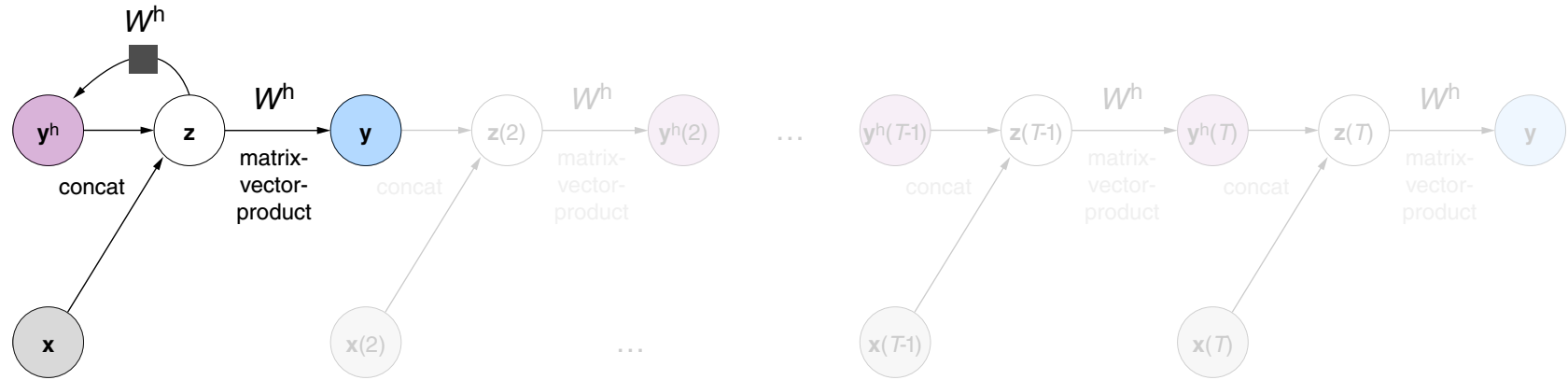
[notation: color, graph, language]



Long-Term Dependencies

Notation II (computational graph)

[notation: color, graph, language]



Remarks (computational graph) :

- The computational graph notation used here follows Goodfellow/Bengio/Courville 2016 :
 1. Each node in the graph indicates a variable. A variable may be a scalar, vector, matrix, tensor, or be of another type.
 2. An operation is a function of one or more variables. An operation returns a single output variable, which does not lose generality because the output variable can have multiple entries, such as a vector.

If a variable b is computed by applying an operation to a variable a , a directed edge is drawn from a to b .

Long-Term Dependencies

Vanishing Gradient Problem

[*TODO*]

Long-Term Dependencies

RNN with Long Short-Term Memory (LSTM)

[*TODO*]

Remarks:

- ❑ LSTM is a recurrent neural network architecture that is very efficient at remembering long term dependencies and that is less vulnerable to the vanishing gradient problem.

Long-Term Dependencies

RNN with Gated Recurrent Units (GRU)

[*TODO*]