Chapter ML:IX (continued)

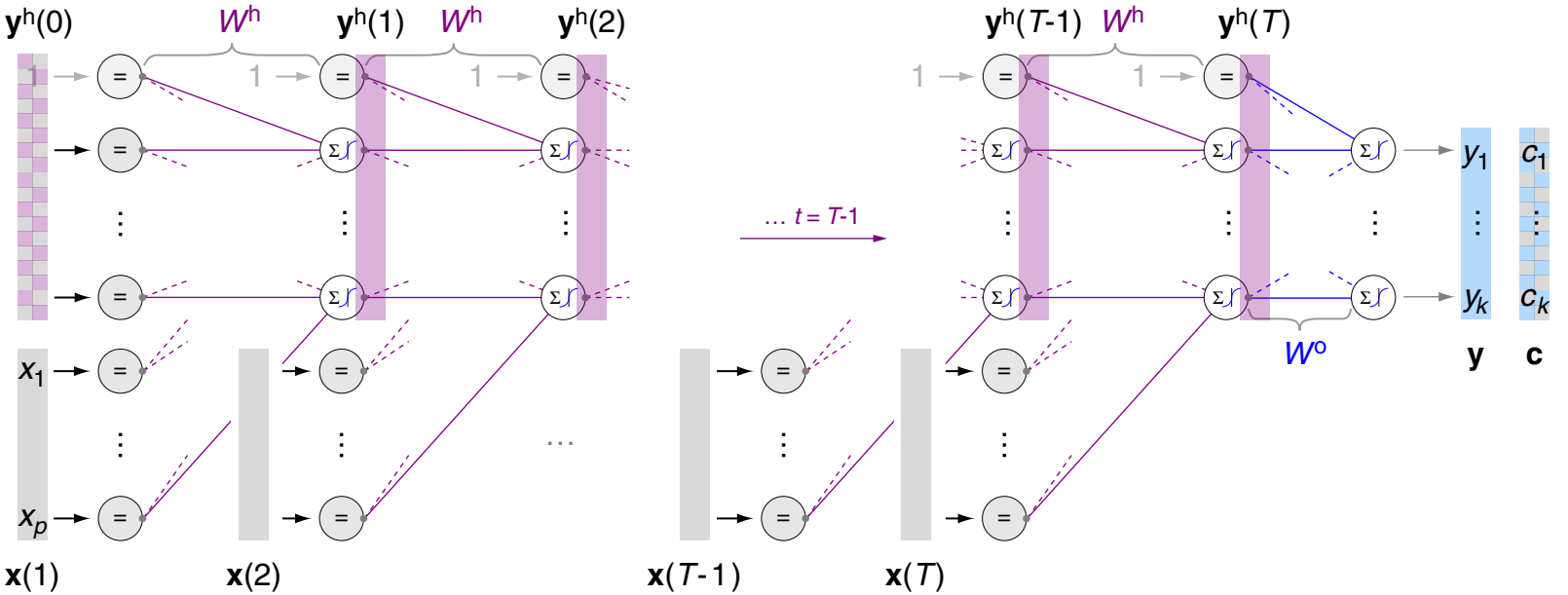
IX. Deep Learning

- 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

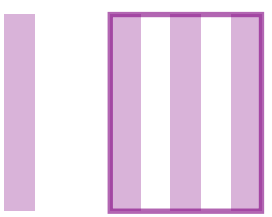
Notation II (computational graph)

[notation: color, graph, language]



x(t) input

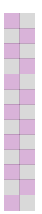
`in_word`



y^{h}(t) hidden (vector, matrix)

y^{e}(t) hidden encoder

y^{d}(t) hidden decoder



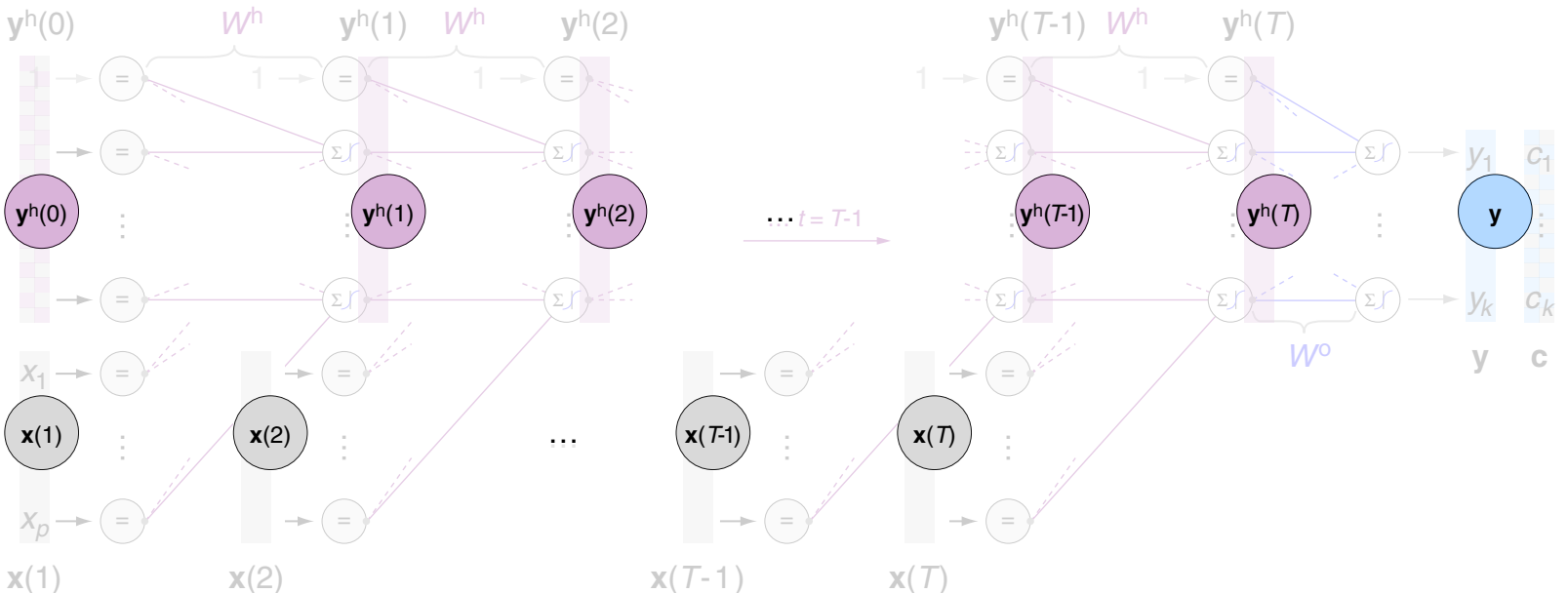
predefined hidden

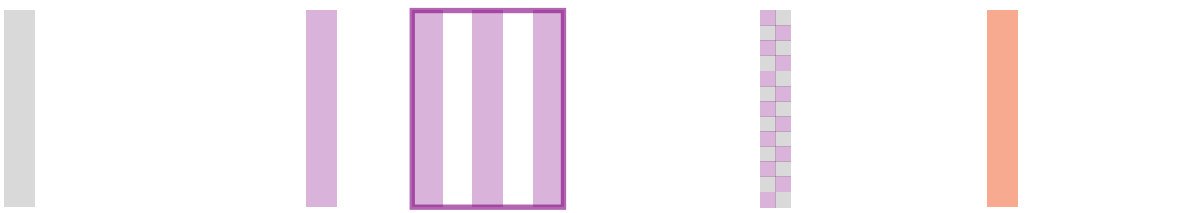
y^{a}(t) attention

Long-Term Dependencies

Notation II (computational graph)

[notation: color, graph, language]





x(t) input
`in_word`

y^{h}(t) hidden (vector, matrix)
y^{e}(t) hidden encoder
y^{d}(t) hidden decoder

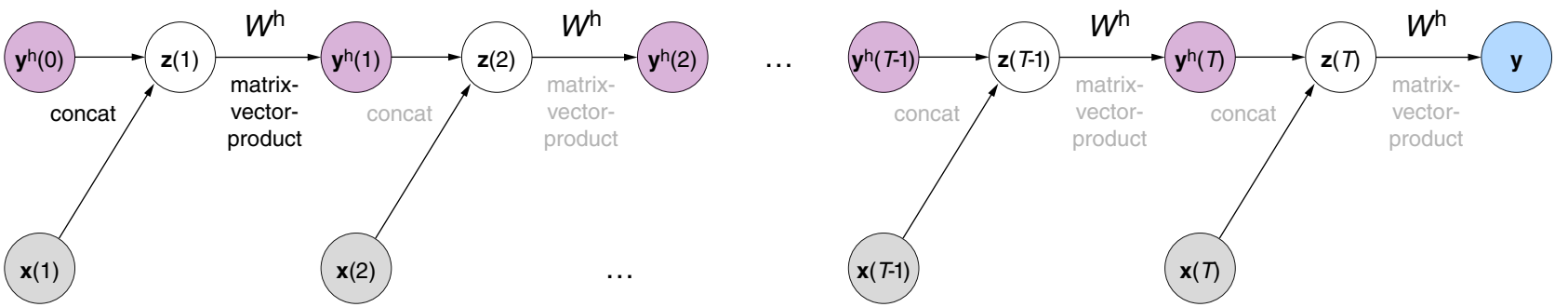
predefined hidden

y^{a}(t) attention

Long-Term Dependencies

Notation II (computational graph)

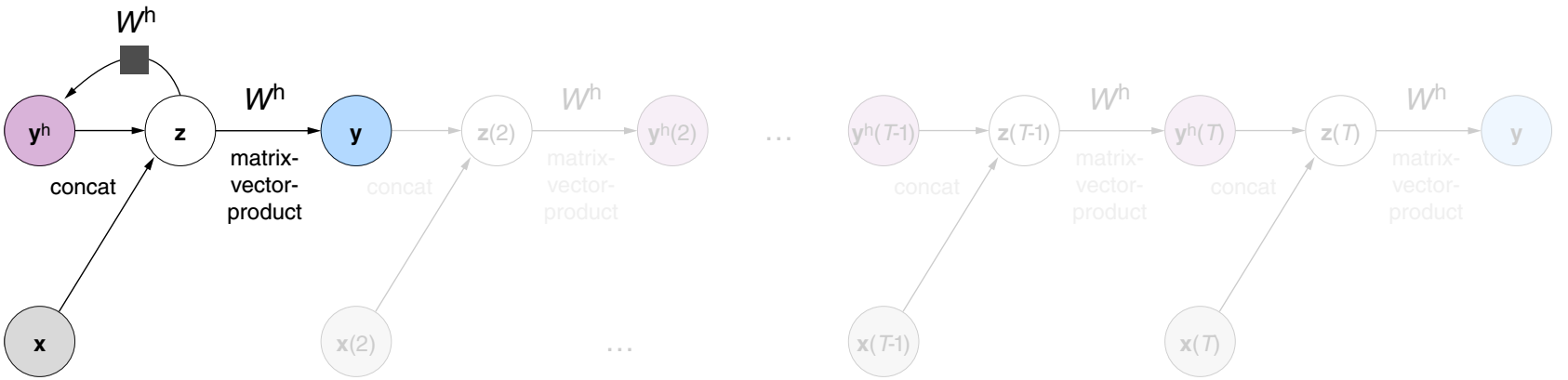
[notation: color, graph, language]



Long-Term Dependencies

Notation II (computational graph)

[notation: color, graph, language]



Remarks (computational graph) :

- The computational graph notation shown 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*]