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NPTEL (<https://swayam.gov.in/explorer?ncCode=NPTEL>) » Deep Learning - IIT Ropar (course)

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## Course outline

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course  
work? ()

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# Week 11: Assignment 11

The due date for submitting this assignment has passed.

Due on 2025-04-09, 23:59 IST.

Assignment submitted on 2025-04-08, 17:56 IST

1) For which of the following problems are RNNs suitable?

1 point

- ☒ Generating a description from a given image.
- ☒ Forecasting the weather for the next  $N$  days based on historical weather data.
- ☒ Converting a speech waveform into text.
- ☐ Identifying all objects in a given image.

Yes, the answer is correct.

Score: 1

Accepted Answers:

*Generating a description from a given image.*

*Forecasting the weather for the next  $N$  days based on historical weather data.*

*Converting a speech waveform into text.*

2) Suppose that we need to develop an RNN model for sentiment classification. The input to the model is a sentence composed of five words and the output is the sentiments (positive or negative). Assume that each word is represented as a vector of length  $100 \times 1$  and the output labels are one-hot encoded. Further, the state vector  $s_t$  is initialized with all zeros of size  $30 \times 1$ . How many parameters (including bias) are there in the network?

4050

No, the answer is incorrect.

Score: 0

Accepted Answers:

*(Type: Numeric) 3992*

1 point

## Week 5 ()

## Week 6 ()

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## Week 11 ()

- Sequence Learning Problems (unit? unit=150&lesson=151)
- Recurrent Neural Networks (unit? unit=150&lesson=152)
- Backpropagation through time (unit? unit=150&lesson=153)
- The problem of Exploding and Vanishing Gradients (unit? unit=150&lesson=154)
- Some Gory Details (unit? unit=150&lesson=155)

3) Select the correct statements about GRUs

1 point

- ☒ GRUs have fewer parameters compared to LSTMs
- ☒ GRUs use a single gate to control both input and forget mechanisms
- ☐ GRUs are less effective than LSTMs in handling long-term dependencies
- ☐ GRUs are a type of feedforward neural network

Yes, the answer is correct.

Score: 1

Accepted Answers:

*GRUs have fewer parameters compared to LSTMs**GRUs use a single gate to control both input and forget mechanisms*

4) What is the main advantage of using GRUs over traditional RNNs?

1 point

- ☐ They are simpler to implement
- ☒ They solve the vanishing gradient problem
- ☐ They require less computational power
- ☐ They can handle non-sequential data

Yes, the answer is correct.

Score: 1

Accepted Answers:

*They solve the vanishing gradient problem*

5) The statement that LSTM and GRU solves both the problem of vanishing and exploding gradients in RNN is

1 point

- ☐ True
- ☒ False

Yes, the answer is correct.

Score: 1

Accepted Answers:

*False*

6) What is the vanishing gradient problem in training RNNs?

1 point

- ☐ The weights of the network converge to zero during training
- ☐ The gradients used for weight updates become too large
- ☐ The network becomes overfit to the training data
- ☒ The gradients used for weight updates become too small

Yes, the answer is correct.

Score: 1

Accepted Answers:

*The gradients used for weight updates become too small*

7) What is the role of the forget gate in an LSTM network?

1 point

- ☐ To determine how much of the current input should be added to the cell state.
- ☒ To determine how much of the previous time step's cell state should be retained.
- ☐ To determine how much of the current cell state should be output.
- ☐ To determine how much of the current input should be output.

Selective Read, Selective Write, Selective Forget - The Whiteboard Analogy (unit? unit=150&lesson=156)

Long Short Term Memory(LSTM) and Gated Recurrent Units(GRUs) (unit? unit=150&lesson=157)

How LSTMs avoid the problem of vanishing gradients (unit? unit=150&lesson=158)

How LSTMs avoid the problem of vanishing gradients (Contd.) (unit? unit=150&lesson=159)

Lecture Material for Week 11 (unit? unit=150&lesson=160)

Week 11 Feedback Form:Deep Learning - IIT Ropar!! (unit? unit=150&lesson=194)

Week 11: Solution (unit? unit=150&lesson=255)

Quiz: Week

Yes, the answer is correct.

Score: 1

Accepted Answers:

*To determine how much of the previous time step's cell state should be retained.*

8) How does LSTM prevent the problem of vanishing gradients?

1 point

- ☐ Different activation functions, such as ReLU, are used instead of sigmoid in LSTM.
- ☐ Gradients are normalized during backpropagation.
- ☐ The learning rate is increased in LSTM.
- ☒ Forget gates regulate the flow of gradients during backpropagation.

Yes, the answer is correct.

Score: 1

Accepted Answers:

*Forget gates regulate the flow of gradients during backpropagation.*

9) We are given an RNN with  $\|W\| = 2.5$ . The activation function used in the RNN is logistic. What can we say about  $\nabla = \left\| \frac{\partial s_{20}}{\partial s_1} \right\|$  1 point

- ☐ Value of  $\nabla$  is very high.
- ☒ Value of  $\nabla$  is close to 0
- ☐ Value of  $\nabla$  is 2.5
- ☐ Insufficient information to say anything.

Yes, the answer is correct.

Score: 1

Accepted Answers:

*Value of  $\nabla$  is close to 0*

10) Select the true statements about BPTT?

1 point

- ☒ The gradients of Loss with respect to parameters are added across time steps
- ☐ The gradients of Loss with respect to parameters are subtracted across time steps
- ☐ The gradient may vanish or explode, in general, if timesteps are too large
- ☒ The gradient may vanish or explode if timesteps are too small

No, the answer is incorrect.

Score: 0

Accepted Answers:

*The gradients of Loss with respect to parameters are added across time steps*

*The gradient may vanish or explode, in general, if timesteps are too large*

**11:  
Assignment  
11  
(assessment?  
name=321)**

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**Week 12 ()**

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**Problem  
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