

INTERNSHIP PROGRESS

Modeling curriculum learning

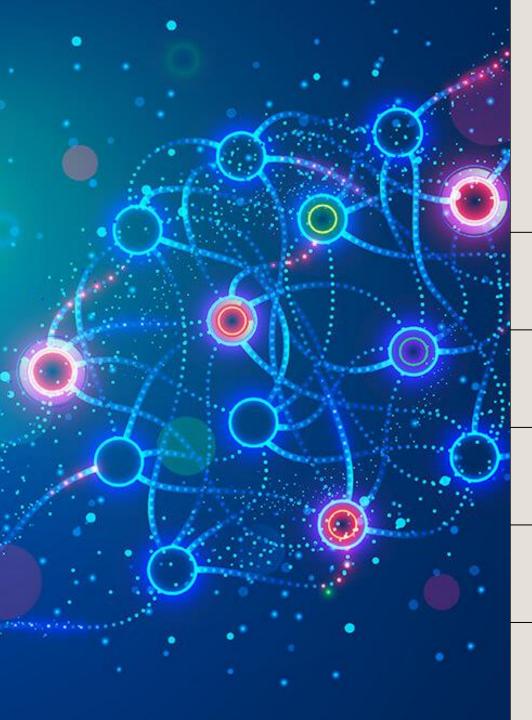


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WEEKLY OBJECTIVES

	FEBRUARY	MARCH	APRIL	MAY
W1	/	Finish 3 models for Level 1 Level 2: accuracy RL learning	Unify the Level 1 models Account for catastrophic interference	Finishing touches Written report
W2	/	Level 2: accuracy RL Level 2: learning progress RL	Level 3: integrate both accuracy and learning progress	Written report
W3	/	Create hypotheses and model ideas for Level 2 & 3 Prepare the presentation	Level 3: explore more options (chaining effect? Between-task learning?)	(exams)
W4	Level 1: implement tasks and neural network	LAB PRESENTATION Integrate all suggestions	Level 3: adjust and compare model performances	(exams)

Primary goals

Past week

Upcoming week

LAST WEEK'S OBJECTIVES



IMPLEMENT A SIMPLE NEURAL NETWORK Using the MNIST dataset

LEARN TO TRACK MODEL PARAMETERS

Such as accuracy, loss, performance

2

CREATE INTRO
FOR THE WEBSITE

 $\overline{4}$

CREATE THE FIRST TASK

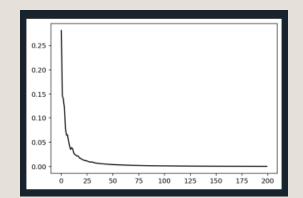
Easy Make questions

 $\overline{3}$

SUMMARISE FINDINGS

Write a short summary of findings, questions and plan for next week.

(5)



#train & test the model
opt = tf.keras.optimizers.Adam(learning_rate = learning_rate)
model.compile(optimizer = opt, loss=tf.keras.losses.MeanSquaredError())
history = model.fit(train_x, train_y, batch_size = 1, epochs = epochs)
model.summary()

test_data = model.predict(test_x)

#report data
print(model.get_weights())

train data: error curve
fig, ax = plt.subplots(1)
ax.plot(history.history["loss"], color = "black")

#test data
print("predictions on the test data:")
print(test_data)

model.save("D:/ULB/MA2/STAGE2/code/internship curriculum model")

What the code looks like

> '''def f(x):

epochs = 100

model.build()

#save the model

learning_rate = 0.5

#define the data set ##AND function

test x = np.copy(train x)

#machine learning model

train_y = np.array([0, 0, 0, 1])
train_y = train_y.reshape(4, 1)

n_input, n_output = train_x.shape[1], 1

model = tf.keras.models.Sequential([
 tf.keras.Input(shape=(n_input,)),

train_x = np.array([[0, 0], [0, 1], [1, 0], [1, 1]])

tf.keras.layers.Dense(n output, activation="sigmoid")

QUESTIONS

QUESTION 1	 Why are we using mean squared error and not cross entropy? → assumes linearity in a binary setting, sensitivity to outliers and greater loss penalty → However, sigmoid neuron which accounts for non-linearity? 	
QUESTION 2	<pre>opt = tf.keras.optimizers.Adam(learning_rate = learning_rate) model.compile(optimizer = opt, loss=tf.keras.losses.MeanSquaredError()) history = model.fit(train_x, train_y, batch_size = 1, epochs = epochs) model.summary()</pre> Should I add a stop requirement? (Early stopping)	
NOTES	XOR is not working yet.	

NEXT WEEK'S OBJECTIVES



IMPLEMENT ALL 3 TASKS

Easy - Hard - Impossible Give them the same structure

LEARN TO TRACK MODEL PARAMETERS

Such as accuracy, loss, performance

2

START LEVEL 2: REINFORCEMENT LEARNING

Research teacher level networks, and implement them using accuracy

UPLOAD CODE TO GITHUB

Improve understanding of the current model, search for potential improvements.

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SUMMARISE FINDINGS

Write a short summary of findings, questions and plan for next week.

(5)