Evaluation

Evaluate a fully trained neural network using the model accuracy as the evaluation metric.

Chapter Goals:

• Evaluate model performance on a test set

A. Evaluating using accuracy

After training a model, it is a good idea to evaluate its performance. We do this by using a *test set* (i.e. data points not used in model training) and observe the model's prediction accuracy on the test set.

The code for this chapter makes use of the accuracy metric defined in Chapter 4. The accuracy represents the classification accuracy of an already trained model, i.e. the proportion of correct predictions the model makes on a test set.

B. Different amounts of data

Since we used None when defining our placeholder shapes, it allows us to run training or evaluation on any number of data points, which is super helpful since we normally want to evaluate on many more data points than the training batch size.

It is good practice to split up a dataset into a three sets:

- Training set (~80% of dataset): Used for model training and optimization
- *Validation set* (~10% *of dataset*): Used to evaluate the model in between training runs, e.g. when tweaking model parameters like batch size
- *Test set* (~10% *of dataset*): Used to evaluate the final model, usually through some accuracy metric

Time to Code!

The coding exercise for this chapter uses the accuracy metric from Chapter 4 (which is initialized in the backend). We also provide the test set's data and

labels (test_data and test_labels) as NumPy arrays initialized in the backend, as well as the inputs and labels placeholders.

We've taken the liberty of loading a pretrained model in the backend using a tf.Session object called sess. You'll be evaluating the accuracy of the pretrained model on the test data and labels.

Set feed_dict equal to a python dictionary with key-value pairs inputs: test_data and labels: test_labels.

Set eval_acc equal to the output of sess.run, with first argument accuracy and a keyword argument feed_dict=feed_dict.

