model quantization xillinx

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1 Model quantization

If you plan on using this implementation, please cite our work (https://www.sciencedirect.com/science/article/pii/S0141933119302844):

@article{Nalepa2020MaM, author = {Jakub Nalepa and Marek Antoniak and Michal Myller and Pablo {Ribalta Lorenzo} and Michal Marcinkiewicz}, title = {Towards resource-frugal deep convolutional neural networks for hyperspectral image segmentation}, journal = {Microprocessors and Microsystems}, volume = {73}, pages = {102994}, year = {2020}, issn = {0141-9331}, doi = {https://doi.org/10.1016/j.micpro.2020.102994}, url = {https://www.sciencedirect.com/science/article/pii/S0141933119302844}}

To perform model quantization, we use the Xillinx DNNDK tool (https://www.xilinx.com/support/documentation/sw_manuals/ai_inference/v1_6/ug1327-dnndk-user-guide.pdf).

```
[1]: import os
import sys
sys.path.append(os.path.dirname(os.getcwd()))
```

```
[4]: import os import subprocess

import tensorflow as tf

from scripts.quantization import evaluate_graph, freeze_model from scripts import prepare_data, artifacts_reporter, train_model, □ → evaluate_model from ml_intuition.data.utils import plot_training_curve, show_statistics
```

2 Prepare the data

To fit into the the pipeline, the data has to be preprocessed. It is achieved by the prepare_data.main function. It accepts a path to a .npy file with the original cube as well as the corresponding ground truth. In this example, we randomly extract 250 samples from each class (balanced scenario), use 10% of them as validation set, and extract only spectral information of a pixel. The returned object is a dictionary with three keys: train, test and val. Each of them contains an additional dictionary with data and labels keys, holding corresponding numpy.ndarray objects with the data. For more details about the parameters, refer to the documentation of prepare_data.main function (located in scripts/prepare_data).

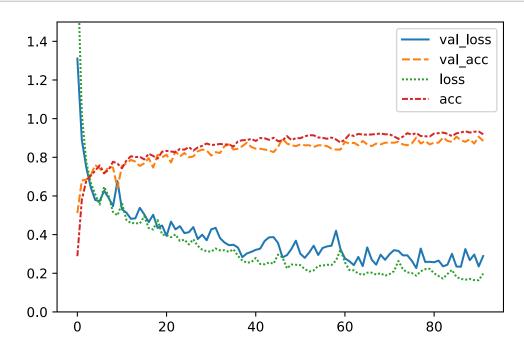
3 Train the model

The function train_model.train executed the training procedure. Trained model will be stored under experiment_dest_path folder path.

```
[7]: train_model.train(model_name='model_2d',
                          kernel_size=5,
                          n_kernels=200,
                          n layers=1,
                          dest_path=experiment_dest_path,
                          data=data_path,
                          sample_size=103,
                          n_classes=9,
                          lr=0.001,
                          batch_size=128,
                          epochs=200,
                          verbose=0,
                          shuffle=True,
                          patience=15,
                          noise=[],
                          noise_sets=[])
```

Layer (type)	Output Shape	Param #
conv2d (Conv2D)	(None, 99, 1, 200)	1200
conv2d_1 (Conv2D)	(None, 32, 1, 200)	200200
conv2d_2 (Conv2D)	(None, 14, 1, 200)	200200
conv2d_3 (Conv2D)	(None, 5, 1, 200)	200200
flatten (Flatten)	(None, 1000)	0
dense (Dense)	(None, 200)	200200
dense_1 (Dense)	(None, 128)	25728
dense_2 (Dense)	(None, 9)	1161

Total params: 828,889 Trainable params: 828,889 Non-trainable params: 0



4 Evaluate full precision model

Evaluate performance of the model in full precision to later compare to the quantized one.

```
[]: evaluate_model.evaluate(
          model_path=os.path.join(experiment_dest_path, 'model_2d'),
          data=data_path,
          dest_path=experiment_dest_path,
          n_classes=9,
          batch size=1024,
          noise=[],
          noise sets=[])
      tf.keras.backend.clear_session()
[10]: show_statistics(os.path.join(experiment_dest_path, "inference_metrics.csv"))
[10]:
        accuracy_score balanced_accuracy_score
                                                 cohen_kappa_score
                                                                      Class_0 \
               0.842077
                                        0.903612
                                                           0.795369
                                                                     0.843755
                    Class_2
          Class 1
                              Class_3 Class_4
                                                 Class_5
                                                           Class_6
                                                                     Class_7
      0 0.783575
                  0.831801
                            0.964463 0.99726 0.903327
                                                         0.953704 0.856061
          Class 8
                  inference_time
      0 0.998565
                         7.275701
```

5 Freeze model

Freeze the tensorflow model into the .pb format.

6 Quantize the model

Perform the quantization by running the quantize.sh bash script with appropriate parameters. It executes the decent_q command from the Xillinx DNNDK library. The output is the quantize_eval_model.pb file and a deploy_model.pb file, which should be used for compilation for a specific DPU.

```
[12]:
```

7 Evaluate the quantized model (graph)

Evaluate the performance of the quantized model to check whether there was any loss in performance. Results for the graph are stored in inference_graph_metrics.csv.

```
[]: graph_path = os.path.join(experiment_dest_path, 'quantize_eval_model.pb')
      evaluate_graph.main(graph_path=graph_path,
                          node_names_path=node_names_file,
                          dataset path=data path,
                          batch_size=1024)
      tf.keras.backend.clear session()
[14]: show_statistics(os.path.join(experiment_dest_path, "inference_graph_metrics.

csv"))
        accuracy_score balanced_accuracy_score cohen_kappa_score
[14]:
                                                                      Class_0 \
              0.874402
                                        0.888489
                                                           0.832185 0.908322
                  Class_2
                            Class_3
                                      {\tt Class\_4}
                                                Class_5
                                                           Class_6
                                                                    Class_7 \
        Class_1
      0 0.89119 0.868037 0.933547 0.998174 0.799121 0.887037 0.712413
         Class_8 inference_time
      0 0.998565
                       21.373184
 []:
```