Project: Test Accuracy of Using PyTorch and Our Matlab/Octave Implementation
Goal

- Making test accuracy of using PyTorch and our code be the same or similar
In our timing comparison you may have noticed that even by using the simplest stochastic gradient steps, the two packages (PyTorch and ours) give different test accuracy.

Many things can cause the difference.

For example,

- data pre-processing
- initial solution
- learning rate
- mini-batch selection
- and others
Now we see why checking the reproducibility of a paper is so difficult

For this project, let’s modify the PyTorch settings to be the same as those for Matlab/Octave (see details below)

The goal is to have that the two packages give same (or very similar) test accuracy

Sometimes this is easy but sometimes it’s tremendously difficult

Of course both run the same number of epochs

Present your findings and thoughts
Let’s check the default settings of our own implementation
Network Architecture

- See the config file of the previous project
- It should be the same as the one for the first project, where PyTorch was used
- Therefore, this part should be all set though you want to double check
Pre-processing I

- **Min-max normalization.**
  For each pixel of every image $Z^{1,i}$, we have

  $$Z^{1,i}_{a,b,d} \leftarrow \frac{Z^{1,i}_{a,b,d} - \min}{\max - \min},$$

  where max/min is the maximum/minimum value of all pixels in $Z^{1,i}$.

- **Zero-centering.**
  This is commonly applied before training CNN (Krizhevsky et al., 2012; Zeiler and Fergus, 2014).
For every pixel in image $Z^{1,i}$, we have

$$Z^{1,i}_{a,b,d} \leftarrow Z^{1,i}_{a,b,d} - \text{mean}(Z^{1,:}_{a,b,d}),$$

where $\text{mean}(Z^{1,:}_{a,b,d})$ is the per-pixel mean value across all the training images.
We follow He et al. (2015) to set the weight values by multiplying random values from the $\mathcal{N}(0, 1)$ distribution and

$$\sqrt{\frac{2}{n_{in}^m}}, \quad \text{where} \quad n_{in}^m = \begin{cases} d^m \times h^m \times h^m & \text{if} \ m \leq L^c, \\ n_m & \text{otherwise}. \end{cases}$$

How about b?? Neex to mention it in the paper.
Parameters for Stochastic Gradient I

- Mini-batch size: we use 128
- For the algorithm with momentum, we use

\[ \alpha = 0.9 \]

for

\[
\begin{align*}
\mathbf{v} & \leftarrow \alpha \mathbf{v} - \eta \left( \frac{\theta}{C} + \frac{1}{|S|} \nabla_{\theta} \sum_{i:i \in S} \xi(\theta; \mathbf{y}^i, Z^{1,i}) \right) \\
\theta & \leftarrow \theta + \mathbf{v}
\end{align*}
\]
Parameters for Stochastic Gradient II

- We run 1,000 epochs and use the model obtained in the end for prediction
- Initial learning rates?? need a table to specify values for the two sets
Other Parameters

- Regularization parameter: ??
  Did we mention it in the paper?
