Astral Algorithmists: Assignment 3.2

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# Environment Setup & Mini Project

For this assignment we have to create an environment for deep learning, in order to discover the use of this new AI solution. Therefore, we first had to choose the framework to use for this solution. Hence, we chose to use PyTorch in order to discover it. Indeed, previously we used Tensorflow. On the other hand, the use of PyTorch’s dynamic graphs could be a valuable asset for our AI solution. In fact, we will be freer when it comes to creating and customizing the architecture of our future deep learning solution. So, we had to install PyTorch before we could start creating our working environment. Here is a snippet of the installation of PyTorch:

|  |
| --- |
| Une image contenant texte, capture d’écran, Police  Le contenu généré par l’IA peut être incorrect. |
| PyTorch installation |

As for the creation of the virtual environment, this was done quite easily as you can see the commands below:

|  |
| --- |
|  |
| Creation of the environment |
|  |
| Activation of the environment |

In addition, as most of the members of the group have already used Github before, the creation of a collaborative directory has been simplified. This directory will allow in the future the management of the different versions of our AI. Here is the link of this latter: <https://github.com/Vorpalin/AI_Assignment>

However, the GPU Check recognition caused us the most problem. Indeed, as first we did not understand why CUDA was not available and we thought that was due to a bad installation of PyTorch on our part. After several hours of hard research, we finally understood that this was not a problem itself and that it was normal since we did not have a GPU adapted to CUDA.

Finally, thanks to the code that was provided for this assignment, we were able to test the creation and the results of our model, which works as expected. Here is a snippet of the code performing these tests:

|  |
| --- |
| Une image contenant texte, Appareils électroniques, capture d’écran, logiciel  Le contenu généré par l’IA peut être incorrect. |
| The python script |
| Une image contenant texte, capture d’écran, Police  Le contenu généré par l’IA peut être incorrect. |
| Its result |

To summarize, this homework familiarizes us with the structure and creation of a deep learning model as well as the PyTorch Framework and project management. These various skills will be useful to us both for our current term project but also for all our future projects.

# Practical Training Pipeline & Mini Project

The goal of this part is to set up a training pipeline for CNN, with data loading and preprocessing, data augmentation and a training loop.

## 2.1) Practical Training Pipeline

### 2.1.1) Data Pipeline

The first action to take when you want to create an AI solution is to load data. This time, instead of using the panda library as usual, we used functions specific to PyTorch, allowing us on the one hand to use transformers to increase the size of our data by adding new images derived from those already present and to perform preprocessing, but also to load datasets already present in the library. Indeed, this week we preferred to use a dataset (the MNIST for digit recognition) whose quality and balance we are sure of, which was not necessarily the case with our previous image datasets. Here is a snippet of code performing the preprocessing, the data loading and augmentation:

|  |
| --- |
| Une image contenant texte, capture d’écran, Police  Le contenu généré par l’IA peut être incorrect. |
| Transformers |
| Une image contenant texte, capture d’écran, Police  Le contenu généré par l’IA peut être incorrect. |
| Data loading and augmentation |

### 2.1.2) Training loop and validation

Next, to create our AI solution, we need to define its architecture, which we did by creating the CNN class, inheriting from Module. Our class therefore consists of three alternations between convolution and maxpooling layers with two dropouts to prevent overfitting ending up on two linear layers. This description is found in the forwarding function of the class. Here is a snippet of the class CNN:

Une image contenant texte, capture d’écran, Police

Le contenu généré par l’IA peut être incorrect.

CNN class

As for AI training, we implemented it outside of the class. To create this training loop, we must first choose the number of epochs. Then for each batch of images (64 images) we apply the forwarding function, we compute the loss, and we update the Artificial Neuron Network with the gradient descent as well as the parameter with our optimizer. We also added an early stopping so that our training stops earlier if our program detects the start of overfitting, so our model does not improve for 5 epochs in a row. Here is a snippet of the code performing this action:

|  |
| --- |
| Une image contenant texte, capture d’écran, Police, logiciel  Le contenu généré par l’IA peut être incorrect. |
| The training loop |
|  |
| Its result |
| Une image contenant texte, capture d’écran, ligne, diagramme  Le contenu généré par l’IA peut être incorrect. |
| Loss over training |

Finally, all that remains is to validate our model by testing it on our test data and see the results of this process. Here is snippet of the code performing this action:

|  |
| --- |
| Une image contenant texte, capture d’écran, logiciel  Le contenu généré par l’IA peut être incorrect. |
| The testing loop |
| Une image contenant texte, Police, capture d’écran, typographie  Le contenu généré par l’IA peut être incorrect. |
| Its result |

As we can see with the metrics obtained by the graph on the previous page and the accuracy by number just above, our model works very well without being victim of overfitting.

Finally, as an improvement we used the wandb library to preserve the performance of our model, with a backup of our best AI in terms of accuracy. Here is snippet of the code performing this action:

|  |
| --- |
|  |
| Initilisation of the wandb library |
| Une image contenant texte, capture d’écran, Police  Le contenu généré par l’IA peut être incorrect. |
| Data backup |

## 2.2) Individual Reflection

### 2.2.1) AMIOT Tom

During this assignment, one of the challenges we faced was the newly use of the PyTorch framework, we previously worked on TensorFlow and us adapting to PyTorch’s dynamic required some time. Another hard time we had was, like said before, during the training sessions, we were faced with many compatibility problems between layers, data stacking together, etc.

Also, the tuning of CNN’s hyperparameters was another small difficulty we encounter. To prevent overfitting, we needed to find the right balance between the layers and the dropout. Throught many experiments, we learned how small adjustments could impact the model’s accuracy and risk of overfitting.

With these difficulties, we gain a lot more experience in building and optimizing AI models, future creation in AI will be more effective.

### 2.2.2) AMRADE Madin

### 2.2.3) MIALON Alexis

The main challenge that Astral Algorithmists encountered during this assignment was the model training part and the model architecture.

Indeed, during the various training sessions, the group often realized that there were compatibility problems between layers or that data had a tendency to group together in blocks, so all the images corresponding to 0 were together and the other numbers the same. So, we had to revisit our architecture several times and we made sure that our dataset shuffled each training epoch. We also had trouble finding a good architecture, but we overcame this by using architecture already used previously on other projects. Finally, the last complication that the group encountered was the training time, which was long, wasting a lot of time in the event of an implementation error. To overcome this, we redirected our calculations to our CPU and made sure that our training processes the images in mini batches, further increasing the efficiency of the training.

Solving these problems has allowed the group to gain experience in the field of AI creation, which will facilitate the creation of future models.

### 2.2.4) MOSSAND Thomas

During this assignment, the **change of environment** from **TensorFlow to PyTorch** and the use of the **wandb library** were two of the challenging moments in our progress.

Unlike TensorFlow, where many operations are automated, PyTorch requires a **more detailed approach.** This represents **a difficulty** at first sight but ultimately enables **greater precision and control** in the training process. Also, the implementation of wandb, which enables deep learning experiments to be monitored and training metrics to be recorded and visualized, was **not a difficulty** in itself. On the other hand, understanding **the structure of the logs**, in particular how to correctly **record and load saved models**, was more **complicated**.

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### 2.2.5) MULOT Antoine

A key challenge in this assignment was transitioning from TensorFlow to PyTorch for model development. While PyTorch’s dynamic graphs offered more flexibility, it required adjustments in model implementation and debugging.

Additionally, setting up the training pipeline introduced compatibility issues between layers, and the dataset needed proper shuffling to avoid grouping patterns during training. Long training times also complicated the process, leading to shifting towards mini-batch processing for improved efficiency.

These challenges provided valuable experience in deep learning model optimization and framework adaptation.