Assignment 1 Pokemon Type Classification by Images

[21/22 WS] Introduction to Deep Learning
Hyemin Ahn

Human-centered Assistive Robotics

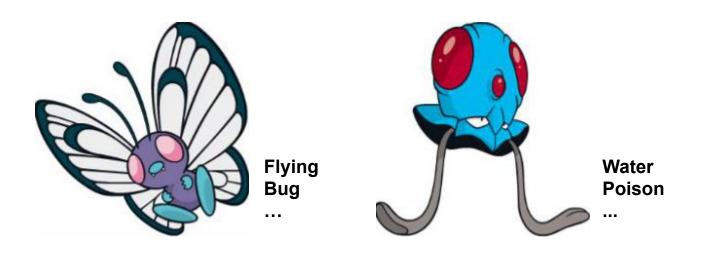
Technische Universität München





Assignment Description

To Do: Pokemon Type Classification based on the Images









Assignment Description

Dataset for Train/Test: Will be distributed to Moodle

10 Classes as below

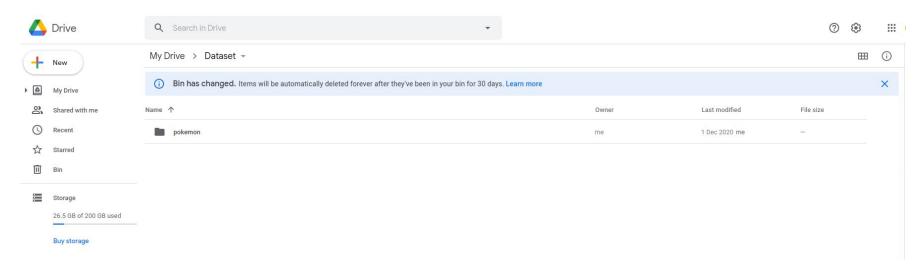






Things you must check before start (in Colab)

1. Put the dataset in your Google Drive!







Things you must check before start (in Colab)

- 2. Load your Google Drive to your Colab session
- 3. Define the path of your Dataset to your notebook code

```
Let's define some path, and our PokeMon dataset

• Put the "pokemon" folder to somewhere of your Google Drive, and define the train/test path to "train_path" and "test_path"

• To 'model_dir', put the drive's directory path that you want to save your model

train_path = './drive/MyDrive/Dataset/pokemon/train'
test_path = './drive/MyDrive/Dataset/pokemon/test'
model_dir = './drive/MyDrive/Codes/models' #./drive/MyDrive/Path/To/Save/Your/Model
classes = ['bug', 'electric', 'fighting', 'fire', 'flying', 'grass', 'ground', 'phychic', 'poison', 'water']
```





Things you must check before start (in Colab)

4. Give the "train_path" or "test_path" as an input argument when creating an instance of PokemonDataset.

```
class PokemonDataset(Dataset):
    def __init__(self, data_path, classes):
        self.data_path = data_path
        self.classes = classes

# organize path information for __len__ and __getitem__
        self.img_path_label = list()
        for c in self.classes:
            img_list = os.listdir(os.path.join(self.data_path, c))
            for fp in img_list:
                full_fp = os.path.join(self.data_path, c, fp)
                self.img_path_label.append((full_fp, c, self.classes.index(c)))
```

```
batch_size = 64

train_dataset = PokemonDataset(train_path, classes)
train_dataloader = DataLoader(train_dataset, batch_size=batch_size, shuffle=True)
test_dataset = PokemonDataset[[test_path, classes]]
test_dataloader = DataLoader(test_dataset, batch_size=batch_size, shuffle=False)
num_classes = len(classes)
```





- 1. Define your "Model" class!
 - Please name your class as "Model", to make grading easier!!

```
class Model (nn.Module):
    def __init__(self, feat_dim = 2048, output_dim =num_classes):
        super(Model, self).__init__()

        self.feat_dim = feat_dim
        self.output_dim = output_dim

        self.backbone = torchvision.models.resnet50(pretrained=True)

        self.backbone.fc = nn.Linear(feat_dim, output_dim)

def forward(self, img):
        out = self.backbone(img)
        return out
```





- 2. Conduct your Training Phase
- 3. Save your model in "Checkpoint" (file_name.pth)

```
# save the model and optimizer's information before the evaulation
checkpoint = {
    'model' : Model(),
    'model_state_dict': model.state_dict(),
    'optimizer_state_dict': optimizer.state_dict(),
}

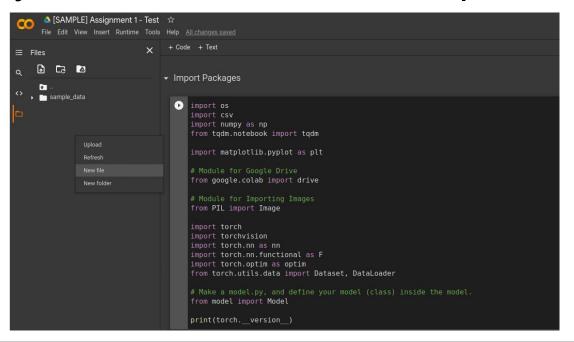
# Save the checkpoint - you can try to save the "best" model with the validation accuracy/loss
torch.save(checkpoint, tmp_path)
if (epoch+1) % save_stride == 0:
    torch.save(checkpoint, os.path.join(model_dir, 'pokemon_{{}}.pth'.format(epoch+1)))
torch.save(checkpoint, os.path.join(model_dir, 'pokemon_recent.pth'))
```

model_dir = './drive/MyDrive/Codes/models'





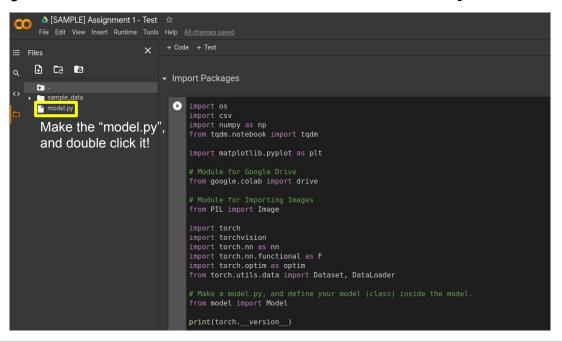
4. Check if your saved model works well on sample test code.







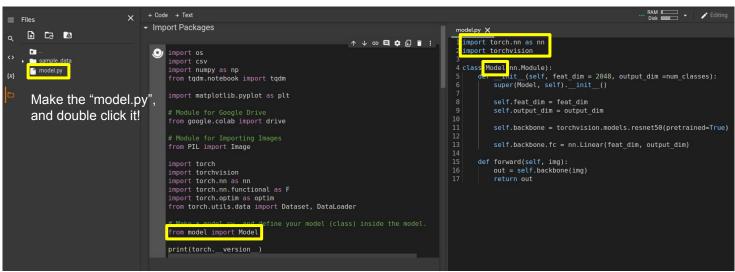
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4. Check if your saved model works well on sample test code.



- (1) Add your "Model" class's definition to "model.py"
- (2) In there, also import dependent packages. (i.e., nn, torchvision...)
- (3) In .ipynb, check if "from model import Model" works.





- 5. Zip your codes (train.ipynb, model.py) + saved "only one best" checkpoint (.pth)
- 6. Submit <firstname_lastname>.zip file to hyemin.ahn@tum.de

```
# save the model and optimizer's information before the evaulation
checkpoint = {
    'model' : Model(),
    'model_state_dict': model.state_dict(),
    'optimizer_state_dict': optimizer.state_dict(),
}

# Save the checkpoint - you can try to save the "best" model with the validation accuracy/loss
torch.save(checkpoint, tmp_path)
if (epoch+l) % save_stride == 0:
    torch.save(checkpoint, os.path.join(model_dir, 'pokemon_r50_{{}}.pth'.format(epoch+l)))
torch.save(checkpoint, os.path.join(model_dir, 'pokemon_r50_recent.pth'))
```

model dir = './drive/MyDrive/Codes/models'





Evaluation

- 1. TOP1 accuracy based on the test dataset will be used for grading.
 - a. Tutor has her own test dataset. Distributed test dataset is only part of it.
- 2. This is not a team project!
 - a. Every student need to submit one's own code and trained model.
- 3. Deadline: 17th of December.
- 4. Only THE FIRST SINGLE SUBMISSION IS ACCEPTED!
 - a. Do not submit your .zip file multiple times.





Some Hints

- Try to make your own "validation" dataset by separating (100% training data) into (80% training data) + (20% validation data).
 - When separating data, try to make (pokemon1) is not included in both (80% training data) and (20% validation data).
 - (Various validation) -> (check test result) -> (various validation)....
 - Try k-fold cross validation. Submit the most robust model.
- Try to add data augmentation tricks.
- Try to fix/freeze the initial layers of ResNet or its variants.
 - Or you can suggest new CNN-based model.





Some Tips for Colab Usage

https://stackoverflow.com/questions/61254168/prevent-a-google-colab-process-from-being-disconnected





Criteria for Grading

Accuracy (%)	0	60	65	70	75	80	85	90	95	100
Score	0	20	30	40	50	60	70	80	90	100

Score will be linearly distributed based on this criteria (i.e., accuracy 72.5% →45)

Experiment Results from Tutor's side (based on Google Colab, best among 10 epoch)

Resnet 18 : (Accu. of provided test data): 75.0% (Accu. of tutor's own test data): 75.7%
Resnet 34 : (Accu. of provided test data): 68.3% (Accu. of tutor's own test data): 71.8%
Resnet 50 : (Accu. of provided test data): 78.3% (Accu. of tutor's own test data): 81.2%

**Running time can be different from account to account.



