CS 5600/6600: Intelligent Systems

Project 1: Part 2: Classification of Images and Audio Samples with Convolutional Networks

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Learning Objectives

- 1. Convolutional Networks
- 2. Image Classification
- 3. Audio Classification

Introduction

This project has two objectives. The first objective of this project is for you to design, train, evaluate, and persist one ConvNet for the BEE1 dataset and one ConvNet for the BEE2 dataset. The second objective is for you to design, train, evaluate, and persist one ConvNet for the BUZZ1 dataset and one ConvNet for the BUZZ2 dataset.

Datasets

For the sake of completeness, I kept the descriptions of BEE1, BEE2, BUZZ1, and BUZZ2 datasets. These descriptions are the same as in the first part of Project 1 that I posted under Assignments in Canvas a week ago.

BEE1

BEE1 contains 54,382 32 x 32 images labeled with two categories BEE (if an image contains at least one bee) or NO-BEE (if it contains no bees or only a small part of a bee). All images were obtained from BeePi monitors deployed on beehives with Italian honeybees in Logan and North Logan, UT in 2017. This dataset is available at https://usu.app.box.com/s/0a5yurmn7ija15cp236awa1re65mbcnr.

The dataset contains six folders — bee_test, bee_train, bee_valid, no_bee_test, no_bee_train, and no_bee_valid. The folders that start with bee_ contain BEE images and the folders that start with no_bee_ contain NO-BEE images. You can use the images in the folders that end with _train and _test for training and testing your ConvNets and the images in the folders that end with _valid for validating their performance.

BEE2

BEE2 is a dataset of 112,879 labeled 90 x 90 images obtained from the videos captured by four BeePi monitors on four Langstroth beehives with Carniolan honeybee colonies in Logan, UT in 2018. This dataset consists of two subsets: BEE2_1S and BEE2_2S. BEE2_1S contains images from the camera placed on top of the first super of a beehive. BEE2_2S contains images from the camera placed on top of the second super of a beehive.

BEE2_1S is available at https://usu.app.box.com/s/p7y8v95ot9u9jvjbbayci61no3lzbgx3. It consists of three folders: training, testing, and validation. You can use the images in the first two folders for training and testing your ConvNets and the images in the third one – for validating them.

BEE2_2S is available at https://usu.app.box.com/s/3ccizd5b1qzcqcs4t0ivawmrxbgva7ym. It consists of three folders: training, testing, and validation. You can use the images in the first two folders for training and testing your ConvNets and the images in the third one – for validating them.

BUZZ1

BUZZ1 contains 10,260 audio samples of manually labeled audio samples (bee buzzing, cricket chirping, and ambient noise). BUZZ1 consists of audio samples captured from beehives with Italian honeybees in Logan and North Logan, UT in 2017. BUZZ1 is available at https://usu.app.box.com/v/BeePiAudioData/file/295296630735.

BUZZ1 contains the folders bee, cricket, noise, and out_of_sample_data_for_validation. You can use wav files in the first three folders for training and testing your ConvNets and in the last folder – for validating them.

BUZZ2

BUZZ2 contains 12,914 audio samples of manually labeled audio samples (bee buzzing, cricket chirping, and ambient noise). BUZZ2 consits of audio samples captured from beehives with Italian and Carniolan honeybee colonies in Logan and North Logan, UT in 2018. BUZZ2 is available at https://usu.app.box.com/v/BeePiAudioData.

BUZZ2 contains the folders test, train, out_of_sample_data_for_validation, You can use wav files in the first two folders for training and testing your ConvNets and in the last folder – for validating them.

Data Processing

Before you can start training your ConvNets, you need to convert examples (images or audio) into numpy arrays. For images, you can install OpenCV (www.opencv.org). OpenCV is an open source computer vision library available for Linux, Mac OS, and Windows. If you're on Ubuntu and use Py 2, you can use this blog entry to install OpenCV: https://www.pyimagesearch.com/2015/06/22/install-opencv-3-0-and-python-2-7-on-ubuntu/. If you're on Ubuntu and use Py 3, you can use this blog entry to install OpenCV: https://www.pyimagesearch.com/2015/07/20/install-opencv-3-0-and-python-3-4-on-ubuntu/. I successfully used both entries to install OpenCV for Py 2 and Py 3. Once you have installed OpenCV, here is how you can check from Python if your installation is successful.

```
>>> import cv2
>>> cv2.__version__
'3.0.0'
```

Here is how you can read an image into a numpy array.

```
>>> img = cv2.imread('/home/vladimir/AI/project_01/images/yes_bee.png')
>>> img.shape
(32, 32, 3)
>>> type(img)
<type 'numpy.ndarray'>
```

Converting wav samples into 1D numpy arrays can be done with scipy (https://www.scipy.org/install.html). Once you have scipy installed, here is how you can read a wav file into a numpy array.

```
>>> from scipy.io import wavfile
>>> samplerate, audio = wavfile.read('/home/vladimir/teaching/AI/project_01/bee25.wav')
>>> samplerate
44100
>>> type(audio)
<type 'numpy.ndarray'>
>>> audio
array([30, 36, 37, ..., 31, 50, 38], dtype=int16)
```

If you need to scale your audio file to have floats between 0 and 1, you can do it as follows.

To define ConvNets, You'll need to install tflearn and all the required packages. You can follow the installation instructions at http://tflearn.org/installation/. We'll go through a few examples in class.

ConvNet Training and Testing

Design and train two ConvNets: ImageConvNet and AudioConvNet and pickle your trained networks in ImageConvNet.pck and AudioConvNet.pck, respectively. Write a Python driver file convnet_tester.py with the following functions.

1. fit_image_convnet(convnet, image_path) - this function reads an image from a file specified by image_path, does the necessary pre-processing of the image to convert it to the appropriate input to the convnet, feeds the input to convnet, and returns a two element binary array where the first element is set to 1 if the image in image_path belongs to the yes_bee class and the second element is set to 1 if the image in image_path belongs to the no_bee class. Suppose that the file yes_bee.png contains an image with a bee and the file no_bee.png contains an image without a bee and your ImageConvNet is persisted in ImageConvNet.pck. Then here is a sample test run.

```
def load(file_name):
    with open(file_name, 'rb') as fp:
        nn = pickle.load(fp)
    return nn
```

```
>>> convnet = load('ImageConvNet.pck')
>>> fit_image_convnet(convnet, 'yes_bee.png')
array([1, 0])
>>> fit_image_convnet(convnet, 'no_bee.png')
array([0, 1])
```

2. fit_audio_convnet(convnet, audio_path) - this function reads an audio wav file from a file specified by audio_path, does the necessary pre-processing of the audio to convert it to the appropriate input to the convnet, feeds the input to convnet, and returns a three element binary array where the first element is set to 1 if the wav file in audio_path belongs to the bee_buzzing class, the second element is set to 1 if the audio sample in audio_path belongs to the cricket_chirping class, and the third element is set to 1 if the audio sample in audio_path belongs to the ambient_noise class. Suppose that the file audio_1.wav contains bee buzzing, the file audio_2.wav contains cricket chirping, and the file audio_3.wav contains ambient noise, and your AudioConvNet is persisted in AudioConvNet.pck. Here is a sample run.

```
def load(file_name):
    with open(file_name, 'rb') as fp:
        nn = pickle.load(fp)
    return nn

>>> convnet = load('AudioConvNet.pck')
>>> fit_audio_convnet(convnet, 'audio_1.wav')
array([1, 0, 0])
>>> fit_audio_convnet(convnet, 'audio_2.wav')
array([0, 1, 0])
>>> fit_audio_convent(convnet, 'audio_3.wav')
array([0, 0, 1])
```

What to Submit

- 1. training_convnets.py this is the file with your training source code and where your ConvNets are defined; your comments should have clearly state the architecture of each network and the parameters with which it was trained;
- 2. ImageConvNet.pck pickled file with your trained ImageConvNet;
- 3. AudioConvNet.pck pickled file with your trained ImageConvNet.
- 4. convnet_report.pdf this is the file with your report where you should write up your training and testing results; your report should have at least 2 pages.

Happy Hacking!