


CS50's Introduction to Artificial Intelligence with Python

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Traffic

Write an AI to identify which traffic sign appears in a photograph.

```
$ python traffic.py gtsrb
Epoch 1/10
500/500 [=====] - 5s 9ms/step - loss: 3.7139 - accuracy: 0
Epoch 2/10
500/500 [=====] - 6s 11ms/step - loss: 2.0086 - accuracy:
Epoch 3/10
500/500 [=====] - 6s 12ms/step - loss: 1.3055 - accuracy:
Epoch 4/10
500/500 [=====] - 5s 11ms/step - loss: 0.9181 - accuracy:
Epoch 5/10
500/500 [=====] - 7s 13ms/step - loss: 0.6560 - accuracy:
Epoch 6/10
500/500 [=====] - 9s 18ms/step - loss: 0.5078 - accuracy:
Epoch 7/10
```

```
500/500 [=====] - 9s 18ms/step - loss: 0.4216 - accuracy:
Epoch 8/10
500/500 [=====] - 10s 20ms/step - loss: 0.3526 - accuracy:
Epoch 9/10
500/500 [=====] - 10s 21ms/step - loss: 0.3016 - accuracy:
Epoch 10/10
500/500 [=====] - 10s 20ms/step - loss: 0.2497 - accuracy:
333/333 - 5s - loss: 0.1616 - accuracy: 0.9535
```

When to Do It

By Sunday, January 1, 2023, 4:59 AM GMT ⌚ (<https://time.cs50.io/2022-12-31T23:59:00-05:00>).

How to Get Help

1. Ask questions via [Ed \(https://cs50.edx.org/ed\)](https://cs50.edx.org/ed)!
2. Ask questions via any of CS50's [communities](#)!

Background

As research continues in the development of self-driving cars, one of the key challenges is [computer vision \(https://en.wikipedia.org/wiki/Computer_vision\)](https://en.wikipedia.org/wiki/Computer_vision), allowing these cars to develop an understanding of their environment from digital images. In particular, this involves the ability to recognize and distinguish road signs – stop signs, speed limit signs, yield signs, and more.

In this project, you'll use [TensorFlow \(https://www.tensorflow.org/\)](https://www.tensorflow.org/) to build a neural network to classify road signs based on an image of those signs. To do so, you'll need a labeled dataset: a collection of images that have already been categorized by the road sign represented in them.

Several such data sets exist, but for this project, we'll use the [German Traffic Sign Recognition Benchmark \(http://benchmark.ini.rub.de/?section=gtsrb&subsection=news\)](http://benchmark.ini.rub.de/?section=gtsrb&subsection=news) (GTSRB) dataset, which contains thousands of images of 43 different kinds of road signs.

Getting Started

- Download the distribution code from <https://cdn.cs50.net/ai/2020/x/projects/5/traffic.zip> (<https://cdn.cs50.net/ai/2020/x/projects/5/traffic.zip>) and unzip it.
- Download the [data set \(https://cdn.cs50.net/ai/2020/x/projects/5/gtsrb.zip\)](https://cdn.cs50.net/ai/2020/x/projects/5/gtsrb.zip) for this project and unzip it. Move the resulting `gtsrb` directory inside of your `traffic` directory.

- Inside of the `traffic` directory, run `pip3 install -r requirements.txt` to install this project's dependencies: `opencv-python` for image processing, `scikit-learn` for ML-related functions, and `tensorflow` for neural networks.

Understanding

First, take a look at the data set by opening the `gtsrb` directory. You'll notice 43 subdirectories in this dataset, numbered `0` through `42`. Each numbered subdirectory represents a different category (a different type of road sign). Within each traffic sign's directory is a collection of images of that type of traffic sign.

Next, take a look at `traffic.py`. In the `main` function, we accept as command-line arguments a directory containing the data and (optionally) a filename to which to save the trained model. The data and corresponding labels are then loaded from the data directory (via the `load_data` function) and split into training and testing sets. After that, the `get_model` function is called to obtain a compiled neural network that is then fitted on the training data. The model is then evaluated on the testing data. Finally, if a model filename was provided, the trained model is saved to disk.

The `load_data` and `get_model` functions are left to you to implement.

Specification

An automated tool assists the staff in enforcing the constraints in the below specification. Your submission will fail if any of these are not handled properly, if you import modules other than those explicitly allowed, or if you modify functions other than as permitted.

Complete the implementation of `load_data` and `get_model` in `traffic.py`.

- The `load_data` function should accept as an argument `data_dir`, representing the path to a directory where the data is stored, and return image arrays and labels for each image in the data set.
 - You may assume that `data_dir` will contain one directory named after each category, numbered `0` through `NUM_CATEGORIES - 1`. Inside each category directory will be some number of image files.
 - Use the OpenCV-Python module (`cv2`) to read each image as a `numpy.ndarray` (a `numpy` multidimensional array). To pass these images into a neural network, the images will need to be the same size, so be sure to resize each image to have width `IMG_WIDTH` and height `IMG_HEIGHT`.
 - The function should return a tuple `(images, labels)`. `images` should be a list of all of the images in the data set, where each image is represented as a

`numpy.ndarray` of the appropriate size. `labels` should be a list of integers, representing the category number for each of the corresponding images in the `images` list.

- Your function should be platform-independent: that is to say, it should work regardless of operating system. Note that on macOS, the `/` character is used to separate path components, while the `\` character is used on Windows. Use `os.sep` (<https://docs.python.org/3/library/os.html>) and `os.path.join` (<https://docs.python.org/3/library/os.path.html#os.path.join>) as needed instead of using your platform's specific separator character.
- The `get_model` function should return a compiled neural network model.
 - You may assume that the input to the neural network will be of the shape `(IMG_WIDTH, IMG_HEIGHT, 3)` (that is, an array representing an image of width `IMG_WIDTH`, height `IMG_HEIGHT`, and `3` values for each pixel for red, green, and blue).
 - The output layer of the neural network should have `NUM_CATEGORIES` units, one for each of the traffic sign categories.
 - The number of layers and the types of layers you include in between are up to you. You may wish to experiment with:
 - different numbers of convolutional and pooling layers
 - different numbers and sizes of filters for convolutional layers
 - different pool sizes for pooling layers
 - different numbers and sizes of hidden layers
 - dropout
- In a separate file called *README.md*, document (in at least a paragraph or two) your experimentation process. What did you try? What worked well? What didn't work well? What did you notice?

Ultimately, much of this project is about exploring documentation and investigating different options in `cv2` and `tensorflow` and seeing what results you get when you try them!

You should not modify anything else in `traffic.py` other than the functions the specification calls for you to implement, though you may write additional functions and/or import other Python standard library modules. You may also import `numpy` or `pandas`, if familiar with them, but you should not use any other third-party Python modules. You may modify the global variables defined at the top of the file to test your program with other values.

Hints

- Check out the official [Tensorflow Keras overview](https://www.tensorflow.org/guide/keras/overview) (<https://www.tensorflow.org/guide/keras/overview>) for some guidelines for the syntax of building neural network layers. You may find the lecture source code useful as well.

- The **OpenCV-Python** (https://docs.opencv.org/4.5.2/d2/d96/tutorial_py_table_of_contents_imgproc.html) documentation may prove helpful for reading images as arrays and then resizing them.
- Once you've resized an image `img`, you can verify its dimensions by printing the value of `img.shape`. If you've resized the image correctly, its shape should be `(30, 30, 3)` (assuming `IMG_WIDTH` and `IMG_HEIGHT` are both `30`).
- If you'd like to practice with a smaller data set, you can download a **modified dataset** (<https://cdn.cs50.net/ai/2020/x/projects/5/gtsrb-small.zip>) that contains only 3 different types of road signs instead of 43.

How to Submit

You may not have your code in your `ai50/projects/2020/x/traffic` branch nested within any further subdirectories (such as a subdirectory called `traffic` or `project5`). That is to say, if the staff attempts to access <https://github.com/me50/USERNAME/blob/ai50/projects/2020/x/traffic/traffic.py>, where `USERNAME` is your GitHub username, that is exactly where your file should live. If your file is not at that location when the staff attempts to grade, your submission will fail.

1. Visit **this link** (<https://submit.cs50.io/invites/8f7fa48876984cda98a73ba53bcf01fd>), log in with your GitHub account, and click **Authorize cs50**. Then, check the box indicating that you'd like to grant course staff access to your submissions, and click **Join course**.
2. **Install Git** (<https://git-scm.com/downloads>) and, optionally, **install submit50** (<https://cs50.readthedocs.io/submit50/>).
3. If you've installed `submit50`, execute

```
submit50 ai50/projects/2020/x/traffic
```

Otherwise, using Git, push your work to <https://github.com/me50/USERNAME.git>, where `USERNAME` is your GitHub username, on a branch called `ai50/projects/2020/x/traffic`.

Note the additional requirement in this project specification, below!

4. **Record a screencast not to exceed 5 minutes in length** (<https://www.howtogeek.com/205742/how-to-record-your-windows-mac-linux-android-or-ios-screen/>) in which you demonstrate your project's functionality. Be certain that every element of the specification, above, is demonstrated in your video. There's no need to show your code, just your application in action; we'll review your code on GitHub. **Upload that video to YouTube** (<https://www.youtube.com/upload>) (as unlisted or public, but not private) or somewhere else.

Some further guidelines about your video:

- Please do not use the tiniest font in your screencast. What looks good on your monitor is not going to look as good in a YouTube video frame.
- Please do not show multiple runs where you change parameters. Show only the version that ultimately reflects your final code.

5. Submit [this form \(https://forms.cs50.io/02d7a445-f3f2-4a55-93b4-598df07652bd\)](https://forms.cs50.io/02d7a445-f3f2-4a55-93b4-598df07652bd).

You can then go to [https://cs50.me/cs50ai \(https://cs50.me/cs50ai\)](https://cs50.me/cs50ai) to view your current progress!

Acknowledgements

Data provided by J. Stallkamp, M. Schlipsing, J. Salmen, and C. Igel. The German Traffic Sign Recognition Benchmark: A multi-class classification competition. In Proceedings of the IEEE International Joint Conference on Neural Networks, pages 1453–1460. 2011 (<http://benchmark.ini.rub.de/index.php?section=gtsrb&subsection=dataset#Acknowledgements>)