

Recognizing Famous Places on Android

Seminar

**Practical Applications of Multimedia Retrieval
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Tim Oesterreich, Romain Granger

Supervisors:

Haojin Yang, Christian Bartz
Prof. Dr. Christoph Meinel

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Contents

1	Introduction	3
2	CNNDroid	4
2.1	Setup and Integration into Android Project	4
2.2	Structure Overview of necessary CNNDroid Files	5
2.3	Convert Trained Models into CNNDroid-compatible Format	6
2.4	CPU vs GPU performance	7
3	Google Streetview Crawler	8
3.1	Setup of viewing parameters	8
3.2	State of Automation (i.e. taking one image per viewing angle)	8
3.3	Current Limitations	8
3.4	Possible Improvements	8
3.5	Google Places API/shutterstock/Flickr	8
4	PlaceRecognizer Application	9
4.1	CNNDroid Integration/Image Classifier	9
4.2	Real-Time Frame Capture	9
4.3	GPS Logger	9
4.4	Wikipedia Parser	9
4.5	How to setup project in Android Studio	9
5	Outlook	10
	References	12

1 Introduction

As deep learning is becoming an increasingly big influence in everyday applications, more and more focus is put into increasing its distribution to different platforms. During the winter semester 2016/2017 a project seminar emerged, that laid focus on developing a mobile phone application for Google's Android operating system that is capable of recognizing famous sights in big cities from images taken with a smartphone camera.

Modern smartphones can in some cases outperform mid-range notebooks from a couple of years ago and, most interestingly, often have a dedicated GPU¹. GPUs are usually used for deep learning because of their high parallelization capabilities.

Part of this seminar was the evaluation of a fairly recent deep learning framework called CNNDroid [1], which uses GPU acceleration for classification and promises substantial performance improvements compared to CPU classification.

¹Graphics Processing Unit

2 CNNDroid

CNNDroid is an open source deep learning library for Android. It is able to execute convolutional neural networks, supporting most CNN layers used by existing desktop/server deep learning frameworks, namely Caffe, Theano and Torch. Supported layers, as of March 2017 are:

- Convolutional Layer
- Pooling Layer
- Local Response Normalization Layer
- Fully-Connected Layer
- Rectified Linear Unit Layer (ReLU)
- Softmax Layer
- Accuracy and Top-K Layer

Due to the library being open source, it is possible to add additional layers, such as batch normalization or sigmoid.

The library also supports a variety of customizations, like maximum memory usage, GPU or CPU acceleration and automatic performance tuning.

2.1 Setup and Integration into Android Project

CNNDroid is a source code library. That means that integration into an existing Android Project is fairly straight forward and doesn't require any third-party dependencies.

The only prerequisites are:

- A functional Android development environment (e.g. Android Studio²)
- Android phone or emulator running at least Android SDK version 21.0 (Lollipop)

To integrate CNNDroid into the project, the project has to be cloned from its GitHub³ page. Inside the 'CNNDroid Source Package' folder, there are three folders. 'java', 'rs' and 'libs'.

The 'java' and 'rs' folders need to be copied into the your 'app/src/main/' directory, merging the 'java' folders.

The 'libs' folder has to be copied and merged into your 'app/' directory.

For effective usage of CNNDroid, it needs read and write access to storage on the smart-phone. Android policies require an application to request these permissions before the

²<https://developer.android.com/studio/index.html>

³<https://github.com/ENCP/CNNDroid>

app starts. These permission requests are specified inside the ‘AndroidManifest.xml’ file. Inside this file, the following lines have to be added before the ‘<application>’ section:

```
<uses-permission android:name="android.permission.READ_EXTERNAL_STORAGE"/>
<uses-permission android:name="android.permission.WRITE_EXTERNAL_STORAGE"/>
```

To cope with the high computational requirements, CNNDroid potentially needs a large amount of memory. Therefore, the heap has to be increased, which is done by adding

```
android::largeHeap="true"
```

inside the ‘<application>’ section.

Now, everything should be set up correctly and programming can commence.

After importing the package `network.CNNDroid`, the `CNNDroid` object is accessible, which can call the function `CNNDroid::classify`. This is the keyword to start execution of the trained model. The specification of the model is explained in the following section.

2.2 Structure Overview of necessary CNNDroid Files

Of course, CNNDroid needs to know how to classify an input. For that it either uses converted binary layer files, created by a desktop deep learning framework or internal conversion functions (e.g. for pooling or ReLU).

The layer blobs are generally used for more complicated layers, such as convolutional or fully connected layers, which have a high dimensional and variable amount of variables. These files have to be converted into the MessagePack⁴ format and put onto the smart-phones storage.

The order in which the layers are executed, additional configurations and layers which don’t need a blob file are defined inside the definition file. This file is used as a settings file for the `CNNDroid` classifier.

On creation of the `CNNDroid` object, the absolute path to this file has to be specified.

Configurations, other than the layer definitions are:

- the absolute path to the layer blob files
(`root_directory: "/absolute/path/to/layer/files"`)
- the maximum amount of RAM the classifier should use on the smartphone
(`allocated_ram: Amount_in_MB`)
- whether or not automatic performance tuning should be used
(`auto_tuning: "on|off"`)
- whether or not classification should be GPU accelerated (if possible)
(`execution_mode: "sequential|parallel"`)

⁴<http://msgpack.org/>

Layers are defined as shown in figure 1.

```
root_directory: "/sdcard/Download/berlin_sights_data/"
allocated_ram: 100
auto_tuning: "off"
execution_mode: "parallel"

layer {
  type: "Convolution"
  name: "conv1"
  parameters_file: "model_param_conv1.msg"
  pad: 2
  group: 1
  stride: 1
}
```

Figure 1: Excerpt of a CNNDroid definition file

Not essential for the execution of CNNDroid, but very helpful for processing the final results, is a labels file. This file should specify human-readable names for the extractable classes. The order inside this file should be mapped to the output order of the last layer (probably a fully-connected layer), i.e. the first value of the output array should correspond to the first class name inside the labels file.

2.3 Convert Trained Models into CNNDroid-compatible Format

As mentioned before, CNNDroid uses a format called MessagePack for the layer definitions. MessagePack is a binary serialization format for exchanging data, comparable to JSON⁵, but other than JSON it is not human-readable. The omission of this constraint makes it possible to reduce size and increase parsing speed of the data stored inside. This is especially important due to the limited storage that most smartphones possess, especially compared to big servers, where deep learning is usually executed.

Figure 2 illustrates how MessagePack saves storage space by using types (JSON encodes everything as strings, so for example, ‘true’ uses four Bytes, while it only uses one Byte using MessagePack) and dropping control symbols, like the curly braces {}.

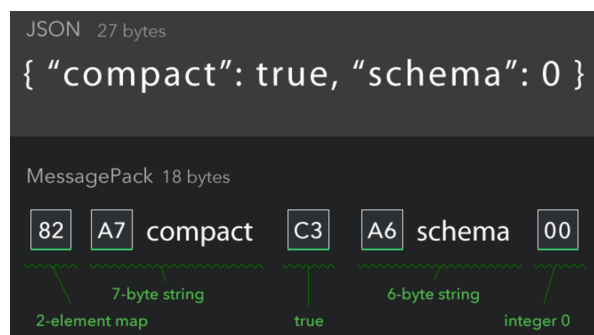


Figure 2: Comparison JSON and MsgPack

⁵JavaScript Object Notation

In order to convert models from the aforementioned desktop/server deep learning frameworks, CNNDroid provides conversion scripts. The script for Caffe is written in Python and requires the packages `numpy` and `msgpack` to be installed.

Three variables need to be defined. The absolute path to the trained model, the deployment prototxt file, which specifies the parameters for the layers, and the saving path. The output are a couple of files in MessagePack format. The definition and labels file has to be created manually, still. For the definition file, most of the content of the prototxt deployment file can be used. Only some parts of the syntax has to be adapted to a CNNDroid parsable format.

2.4 CPU vs GPU performance

Comparison of computation time of CPU (sequential) and GPU (parallel) mode on in-memory images using CIFAR10 (in-memory to minimize error using camera, CIFAR10 could be exchanged with CaffeNet if too fast)

3 Google Streetview Crawler

3.1 Setup of viewing parameters

Explain csv file

3.2 State of Automation (i.e. taking one image per viewing angle)

3.3 Current Limitations

Full automation not possible as of current street view API state; wrong latitude/longitude; Streetview Image API returns different images than JavaScript API (which is being used on google maps website)

3.4 Possible Improvements

Use Classifier to identify things in photosphere

3.5 Google Places API/shutterstock/Flickr

4 PlaceRecognizer Application

Overview: What does it do. Whom is it for. How does it achieve its task?

4.1 CNNDroid Integration/Image Classifier

Explain ImageClassifier Class; Including Variables that need adaption when changing Layers or DataSets

How to put msgpack on phone

4.2 Real-Time Frame Capture

How does the camera talk to the Image Classifier?

4.3 GPS Logger

How do we get GPS values and how can we integrate them?

4.4 Wikipedia Parser

How do we get the text for a classified image from Wikipedia?

4.5 How to setup project in Android Studio

5 Outlook

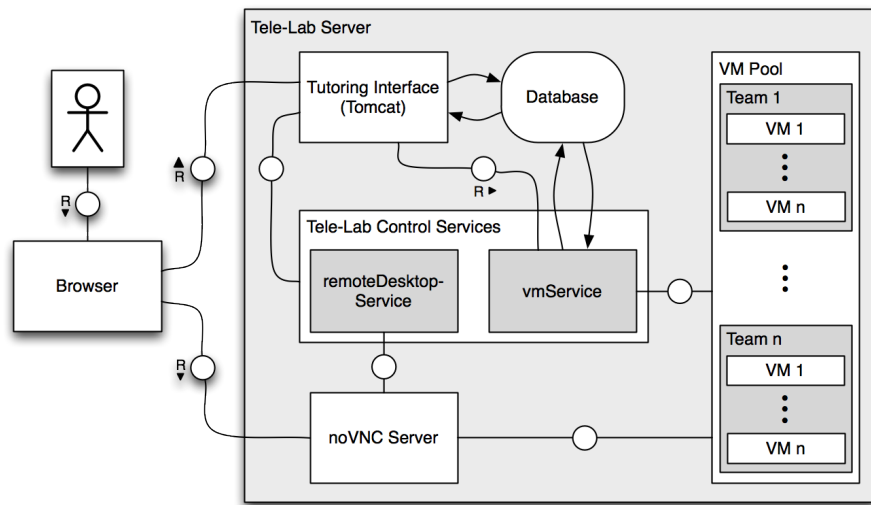


Figure 3: Eine Abbildung, Quelle: [1]

References

- [1] Seyyed Salar Latifi Oskoueï, Hossein Golestani, Matin Hashemi, and Soheil Ghiasi. Cnndroid: Gpu-accelerated execution of trained deep convolutional neural networks on android. In *Proceedings of the 2016 ACM on Multimedia Conference*, MM '16, pages 1201–1205, 2016.