# **FontGen**

# Generating fonts using a neural net

Matt Gaikema

```
In[1]:= ClearAll["Global`*"]
SetDirectory[NotebookDirectory[]];
```

I spend a lot of time trying to identify and download fonts, so a program that could do it for me would be wonderful. There are many academic papers on the subject, but none of them have published their code, so I decided to implement it myself and learn more about deep learning along the way.

#### Data

The data is a whopping 13+ gigabytes and was scraped from the internet by someone else. It can be downloaded here.

```
Import["fonts.hdf5", {"Dimensions"}]  \langle \big| / \text{fonts} \rightarrow \{56443, 62, 64, 64\} \big| \rangle
```

I debugged the neural net using a much smaller dataset, consisting of my personal collection of fonts. The smaller dataset was created by running a script written by the same guy who got the original data.



### **Network**

Each letter is 64x64 pixels.

```
(* We use 4 letters as input... *)
4 * 64 ^ 2
(* ... and get 62 letters as output. *)
62 * 64 ^ 2
16384
```

253 952

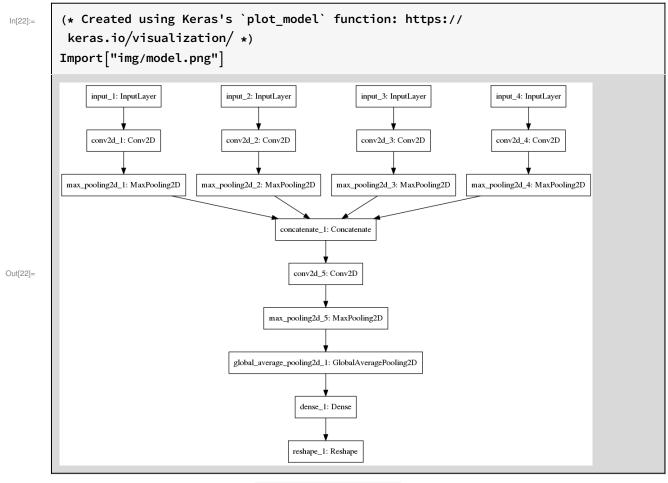
There are  $4 * 64^2 = 16$ , 384 input neurons, and  $62 * 64^2 = 253$ , 953 output neurons. The letters chosen for the inputs are "B", "A", "S", and "Q".

```
(* http://bit.ly/2oeHyZX *)
Range[0, 25];
letters = AssociationThread[ToUpperCase[#] & /@ Alphabet[], %];
Lookup[letters, {"B", "A", "S", "Q"}]
Image[f[[1]][[#+1]] & /@%
\{1, 0, 18, 16\}
```



#### **Architecture**

Here is the architecture that I used for the model. Many of the parameters were determined through tedious trial-and-error.



After training is complete, running model.save weights() saves the model and its weights to an HDF5 file, which can be loaded to recreate the model (see the documentation).

```
Import["model.hdf5", {"Dimensions"}]
  In[21]:=
                                                    \langle \ | \ / conv2d_1/conv2d_1/bias:0 \rightarrow \{8\}, \ / conv2d_1/conv2d_1/kernel:0 \rightarrow \{4, 4, 1, 8\}, \ / conv2d_1/conv2d_1/kernel:0 \rightarrow \{4, 4, 1, 8\}, \ / conv2d_1/conv2d_1/kernel:0 \rightarrow \{4, 4, 1, 8\}, \ / conv2d_1/kernel:0 \rightarrow \{4, 4, 4, 1, 8\}, \ / conv2d_1/kernel:0 \rightarrow \{4, 4, 4, 1, 8\}, \ / conv2d_1/kernel:0 \rightarrow \{4, 4, 4, 1, 8\}, \ / conv2d_1/kernel:0 \rightarrow \{4, 4, 4, 1, 8\}, \ / conv2d_1/kernel:0 \rightarrow \{4, 4, 4, 1, 8\}, \ / conv2d_1/kernel:0 \rightarrow \{4, 4, 4, 1, 8\}, \ / conv2d_1/kernel:0 \rightarrow \{4, 4, 4, 1, 8\}, \ / conv2d_1/kernel:0 \rightarrow \{4, 4, 4, 1, 8\}, \ / conv2d_1/kernel:0 \rightarrow \{4, 4, 4, 1, 8\}, \ / conv2d_1/kernel:0 \rightarrow \{4, 4, 4, 1, 8\}, \ / conv2d_1/kernel:0 \rightarrow \{4, 4, 4, 1, 8\}, \ / conv2d_1/kernel:0 \rightarrow \{4, 4, 4, 1, 8\}, \ / conv2d_1/kernel:0 \rightarrow \{4, 4, 4, 1, 8\}, \ / conv2d_1/kernel:0 \rightarrow \{4, 4, 4, 1, 8\}, \ / conv2d_1/kernel:0 \rightarrow \{4, 4, 4, 1, 8\}, \ / conv2d_1/kernel:0 \rightarrow \{4, 4, 4, 1, 8\}, \ / conv2d_1/kernel:0 \rightarrow \{4, 4, 4, 1, 8\}, \ / conv2d_1/kernel:0 \rightarrow \{4, 4, 4, 1, 8\}, \ / conv2d_1/kernel:0 \rightarrow \{4, 4, 4, 1, 8\}, \ / conv2d_1/kernel:0 \rightarrow \{4, 4, 4, 1, 8\}, \ / conv2d_1/kernel:0 \rightarrow \{4, 4, 4, 1, 8\}, \ / conv2d_1/kernel:0 \rightarrow \{4, 4, 4, 1, 8\}, \ / conv2d_1/kernel:0 \rightarrow \{4, 4, 4, 1, 8\}, \ / conv2d_1/kernel:0 \rightarrow \{4, 4, 4, 1, 8\}, \ / conv2d_1/kernel:0 \rightarrow \{4, 4, 4, 1, 8\}, \ / conv2d_1/kernel:0 \rightarrow \{4, 4, 4, 1, 8\}, \ / conv2d_1/kernel:0 \rightarrow \{4, 4, 4, 1, 8\}, \ / conv2d_1/kernel:0 \rightarrow \{4, 4, 4
Out[21]=
                                                           /conv2d_2/conv2d_2/bias:0 \rightarrow \{8\}, /conv2d_2/conv2d_2/kernel:0 \rightarrow \{4, 4, 1, 8\},
                                                           /conv2d_3/conv2d_3/bias:0 \rightarrow \{8\}, /conv2d_3/conv2d_3/kernel:0 \rightarrow \{4, 4, 1, 8\},
                                                           /conv2d_4/conv2d_4/bias:0 \rightarrow \{8\}, /conv2d_4/conv2d_4/kernel:0 \rightarrow \{4, 4, 1, 8\},
                                                           /conv2d_5/conv2d_5/bias:0 \rightarrow \{10\}, /conv2d_5/conv2d_5/kernel:0 \rightarrow \{4, 4, 32, 10\},
                                                           /dense\_1/dense\_1/bias:0 \rightarrow \{253\,952\}\,,\,\,/dense\_1/dense\_1/kernel:0 \rightarrow \{10\,,\,253\,952\}\,\big|\,\rangle
```

## **Evaluation**

```
(* Transforms a list of image arrays to a list of images. *)
displayFont[f_]:=Map[Map[Image,#]&,f];
```

I tested the model on the font Ubuntu Mono, which is available for download here.

Out[20]=

{{		Say,				
				\$ \\ \\ \\ \\ \\ \\ \\ \\ \\ \\ \\ \\ \\		
						, ,
	,					
Assa'						
,		,	, ,	**************************************	, , , ,	(*) 

# References

```
(* There is no better way to insert references on Mac/Linux. see http://
 bit.ly/2lo3Pkc *)
Import["sources.bib"]
@inproceedings{paper,
    title = {Learning Typographic Style},
    author = {Shumeet Baluja},
    year = \{2016\},
    URL = {http://arxiv.org/abs/1603.04000},
    booktitle = {arXiv}
@misc{deepfont,
    url =
  {https://erikbern.com/2016/01/21/analyzing-50k-fonts-using-deep-neural-networks
    title = {Analyzing 50k fonts using deep neural networks},
    author = {Erik Bernhardsson},
    year = \{2016\},
    month = JAN
@misc{fonsi,
    author = {Fonsi Bonilla},
    howpublished = {Personal Correspondance},
    year = \{2018\},
    month = DEC
```