



# DeCaptcha

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# Our Problem

The goal is for this model is to be able to read text from a "Completely Automated Public Turing test to tell Computers and Humans Apart (CAPTCHA)." The dataset I will be using can be found here: [Link](#)

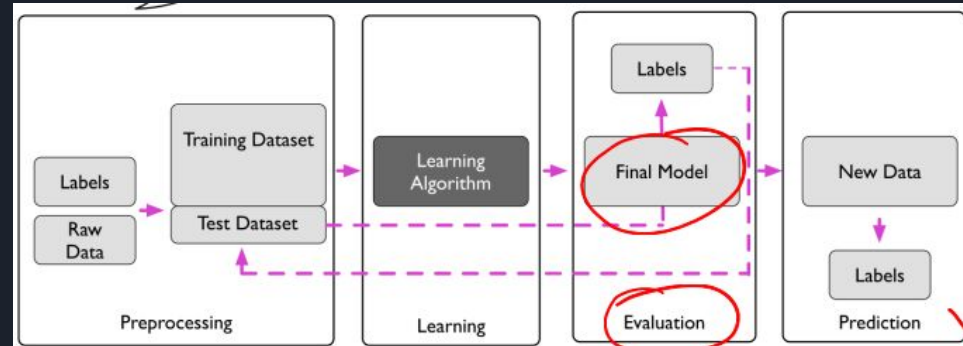
Captchas are a pain in the butt to solve (We get it wrong 29% of the time, [Stanford](#)), so why don't we go and make an AI to automatically solve these problems with just a copy & paste?



# Our Process

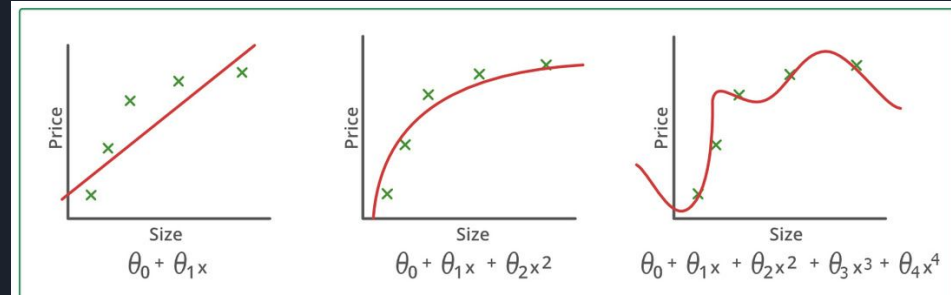
We have split the task up into 3 parts:

- Data processing
- Training & Validation
- Prediction



# Some lingo:

Overfitting:



Bootstrapping : It is any test or metric that uses random sampling with replacement, and falls under the broader class of resampling methods ( $1/e \approx .368$ )

Epoch: When every training sample is processed, you do epoch multiple times to improve accuracy.

# Data Preprocessing (V1)

111XM.jpg (2.46 kB)



Each image is 150x40 pixels. The images consist of texts with many different colors, with the text being listed in the title.

Our data currently is consisted of 113,000 images, and we will be converting this into a csv for ease of use.

We are planning to quantize each image's pixel in a format similar to this:

| Image | Pixel 1 | Pixel 2 | Pixel 3 | Pixel 4 | Pixel 5 | ... | Pixel n |
|-------|---------|---------|---------|---------|---------|-----|---------|
| 12345 | (r,g,b) | (r,g,b) | (r,g,b) | (r,g,b) | (r,g,b) | ... | (r,g,b) |
| 23456 | (r,g,b) | (r,g,b) | (r,g,b) | (r,g,b) | (r,g,b) | ... | (r,g,b) |
| 34567 | (r,g,b) | (r,g,b) | (r,g,b) | (r,g,b) | (r,g,b) | ... | (r,g,b) |
|       | (r,g,b) | (r,g,b) | (r,g,b) | (r,g,b) | (r,g,b) | ... | (r,g,b) |



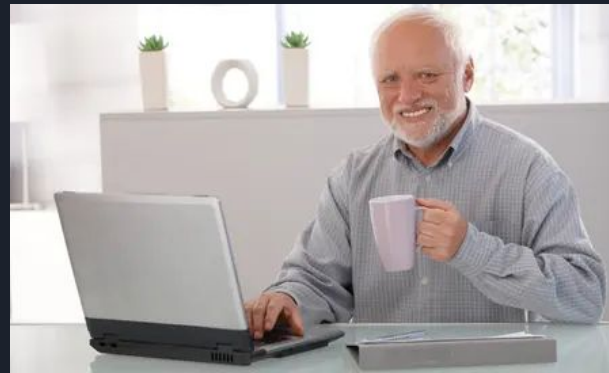
# A little roadblock

```
2 from PIL import Image
3 def imgConvert(imageName):
4
5
6 img = Image.open("imageName") #Placeholder
7 rgb_img = img.convert('RGB')
8 for i <= 149,i=0,i++
9     for n <= 39, n = 0, n++
10         r,g,b = rgb_img.getpixel((i,n))
11         print(r,g,b)
```

We decided that instead of bootstrapping we would manually split the images 36% of the data into the validation folder.

We use the OS library's `listdir` and OpenCV `imread` methods to fix this issue and to further speed up our preprocessing speeds.

Additionally, we were planning on training our model on the captchas themselves but we realized that we'd need to train the model individually



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We

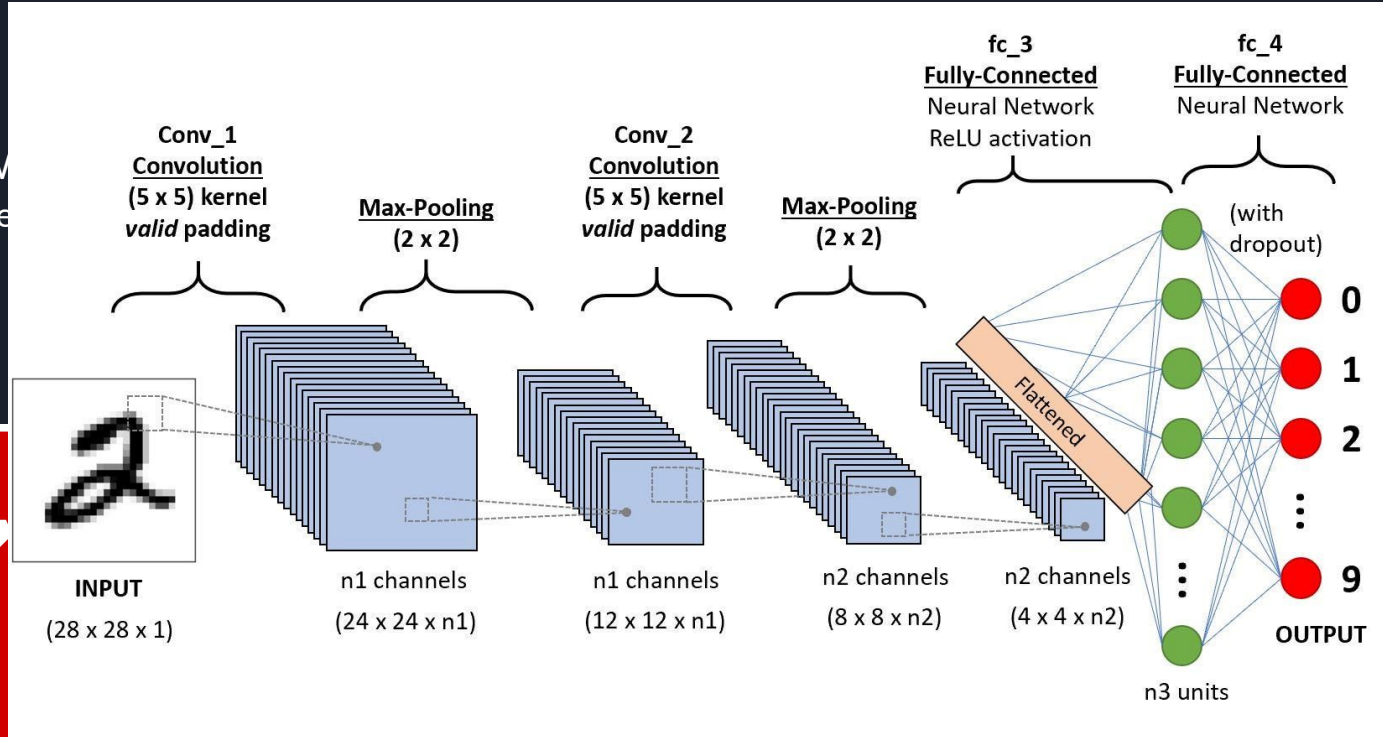
to

We

```
1 import numpy as np
2 import cv2 as cv
3
4 def pp(directory): # Input is the directory of images you want to convert into x
5
6     num_images = len(os.listdir(directory))
7
8     X = np.zeros((num_images, 40, 150, 3)) # creates an array with "num_images" *
9     y = np.zeros((5, num_images, lencharacters)) # gets an array to be able to as
10
11     for i, pic in enumerate(os.listdir(directory)): # gives each image with a num
12         img = cv.imread(os.path.join(directory, pic)) # reads the data of the ima
13         name = pic[:-4] # Takes the name without the .jpg
14
15         if len(name) < 6: # makes sure machine is not tripping
16             img = img / 255.0 # changes the rgb values so that it is in between
17             img = np.reshape(img, (40, 150, 3)) # changes image into a large arra
18
19             target = np.zeros((5, lencharacters)) # creates an array 5 by 62 with
20
21             for j, k in enumerate(name): # labels each letter
22                 index = characters.find(k) # gets the index of the letter we want
23                 target[j, index] = 1 # puts a 1 in the spot k was in allowing us
24
25             X[i] = img # stores the image for later
26             y[:,i] = target # stores all the data for the labels in that image
27
28     return X, y
```

the model

# Training & Validation







# Results of Training

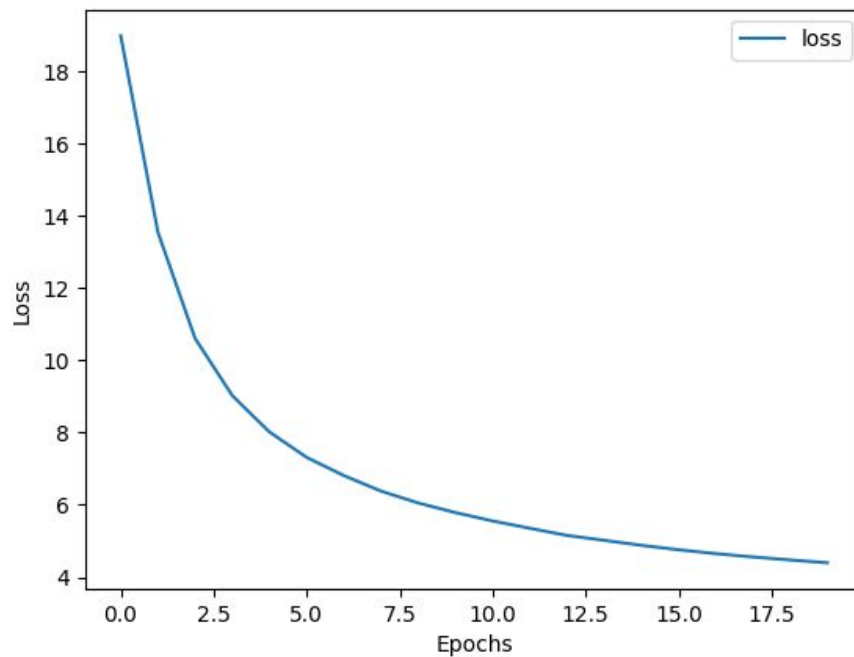
On our first model we started with doing 32 Batches, 20 Epochs, and our optimizer was ADAM. With this model we had a 79.53% success rate (That's a B baby!)

On our third model we used 128 Batches, 20 Epochs, and our optimizer was RMSProp, and we changed the validation to 20%. With this model we had roughly a 94% success rate.

On our fourth model we used 128 Batches, 20 Epochs, and our optimizer was RMSProp, we changed the validation to 20%, and this time we used weight decay at .2% to prevent overfitting. With this model we had achieved a 97.2% success rate.

# Some Graphs:

Model 3





Our Final Product!





## Sources:

<https://www.kaggle.com/code/yassineghouzam/introduction-to-cnn-keras-0-997-top-6>

<https://medium.com/@manvi./captcha-recognition-using-convolutional-neural-network-d191ef91330e>

[burszstein\\_2010\\_captcha.pdf \(stanford.edu\)](#)



This notebook is open with private outputs. Outputs will not be saved. You can disable this in [Notebook settings](#).

## DeCapthaMoreRAM.ipynb

Last saved at 9:46AM

```
1 layers.Input(shape = size)# initialize keras tensor and set the size of the input
2 layers.Conv2D(32,(5,5), padding = 'same', activation = 'relu')(img) # 1st convoluti
3 layers.MaxPooling2D(pool_size = (2,2),padding='same')(conv1) # 1st pooling layer
4 layers.Conv2D(64,(5,5), padding = 'same', activation = 'relu')(mp1) # 2nd convoluti
5 layers.MaxPooling2D(pool_size = (2,2),padding='same')(conv2) # 2nd pooling layer
6 dp = layers.Dropout(0.25)(mp2)
7 conv3 = layers.Conv2D(64,(3,3), padding = 'same', activation = 'relu')(dp) # 3rd convoluti
8 mp3 = layers.MaxPooling2D(pool_size = (2,2),padding='same')(conv3)
9 flat = layers.Flatten()(mp3) # flattens the layers into a vector or sm
10
11 outs = []
12 for _ in range(5): # for each layer of the captcha
13     dens1 = layers.Dense(256, activation='relu')(flat) # normal Neural
14     drop = layers.Dropout(0.5)(dens1) # drops half of the nodes
15     res = layers.Dense(10, activation = 'sigmoid')(drop) #
16
17     outs.append(res) # adds the results of the layers to the end of the
18
19 model = Model(img, outs) # creates the model
20 model.compile(optimizer = optimizer , loss = "categorical_crossentropy")
21 return model
```

```
[ ] 1 X_train, y_train = pp(train)
```

```
[ ] 1 #initializes model
2 model = model.compile(optimizer = optimizer, loss = "categorical_crossentropy")
3 model.summary()
```

```
[ ] 1 #Apply model
2 hist = model.fit(X_train, y_train[0], y_train[1], y_train[2], y_train[3], y_train[4], batch_size=32, epochs=10, validation_data=(X_test, y_test))
```

```
[ ] 1 model.save('model15.h5')
```

```
1 #graph of loss vs epochs
```

Select all Images with bridges



Are you still there?

Colab is for [interactive](#)



VERIFY