

ENGLISH CHARACTER IMAGE GENERATOR IMAGE PREDICTOR

ALAN WONG



CONTENTS

- PROBLEM DEFINITION
- INTRODUCTION TO DEEP LEARNING
- SOLUTION
 - IMAGE PREDICTOR
 - IMAGE GENERATOR
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PROBLEM DEFINITION

**CREATE A PROGRAMME THAT FULFILS THE
FOLLOWING CRITERIA**



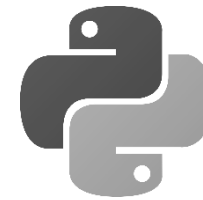
**ENGLISH
CHARACTER
GENERATOR**



**ENGLISH
CHARACTER
PREDICTOR**



**DEEP NEURAL
NETWORK**



PYTHON 3

2

INTRODUCTION TO DEEP LEARNING

MACHINE LEARNING TASK TYPES



SUPERVISED LEARNING

INPUT AND
OUTPUT DATA

MAKE PREDICTION



UNSUPERVISED LEARNING

INPUT DATA ONLY

FIND SIMILARITY
AND DIFFERENCES
IN DATA



REINFORCEMENT LEARNING

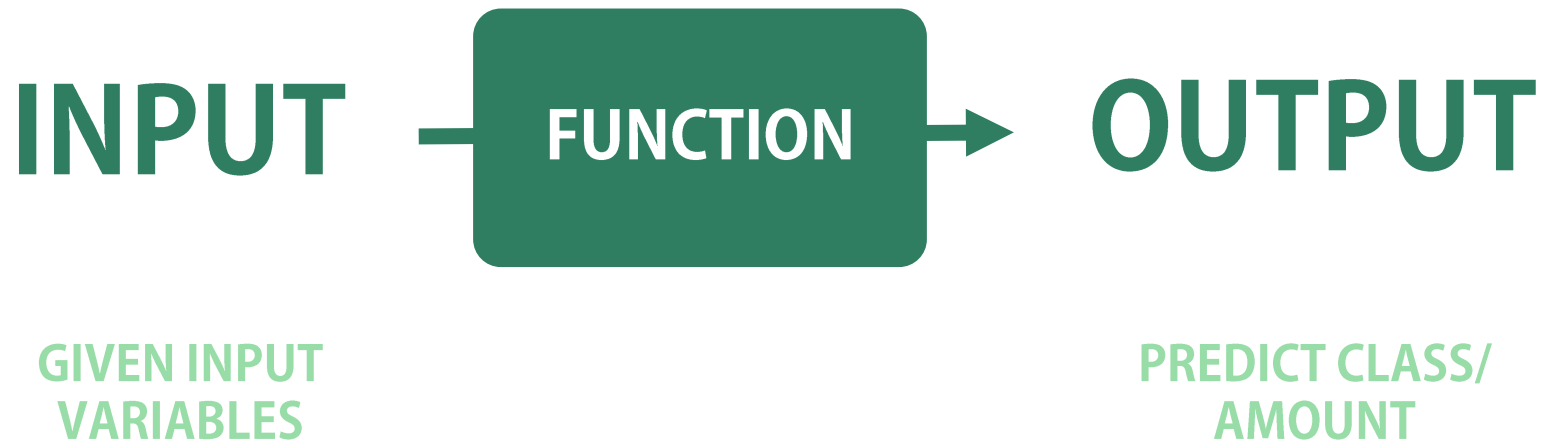
INPUT DATA ONLY

FIND ACTION TO
MAXIMIZE
REWARD

2

INTRODUCTION TO DEEP LEARNING

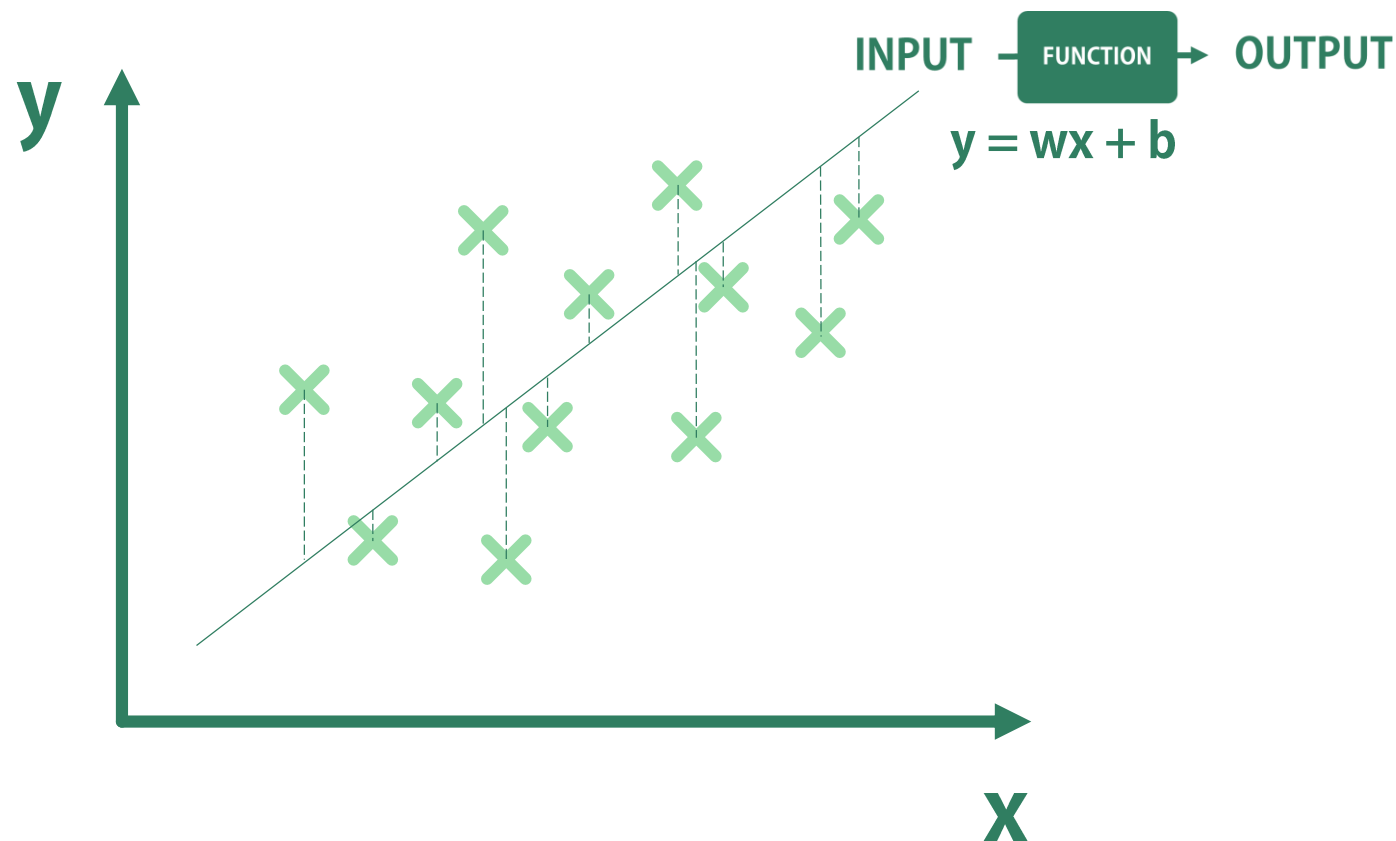
MACHINE LEARNING PREDICTION



2

INTRODUCTION TO DEEP LEARNING

LINEAR REGRESSION BACK TO BASICS



OPTIMIZE (FIND w AND b)
MINIMIZING MEAN SQUARED ERROR

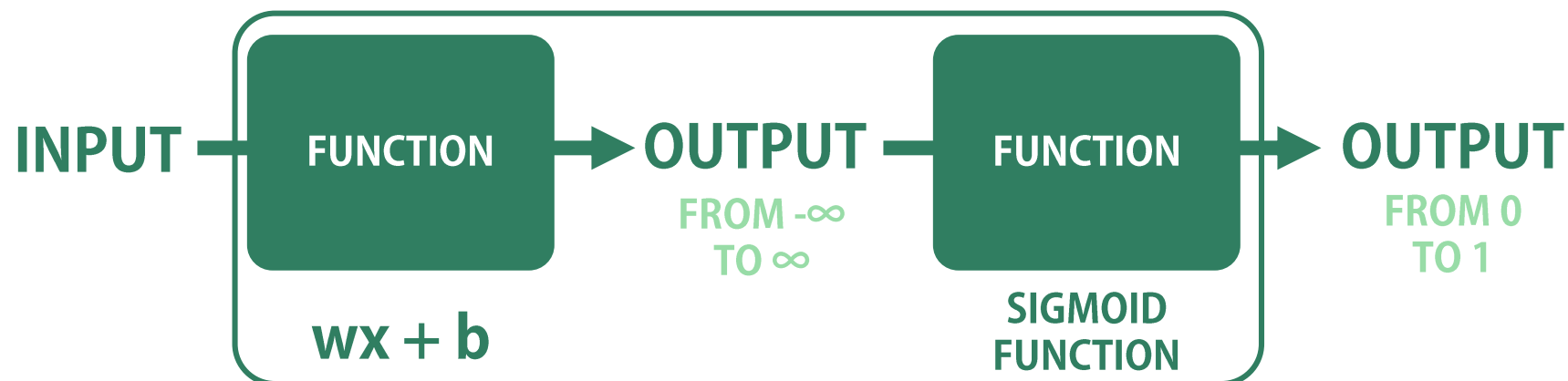
2

INTRODUCTION TO DEEP LEARNING

LOGISTIC REGRESSION

USE LINEAR REGRESSION TO PREDICT CLASS

OUTPUT FROM LINEAR REGRESSION IS FROM $-\infty$ TO ∞



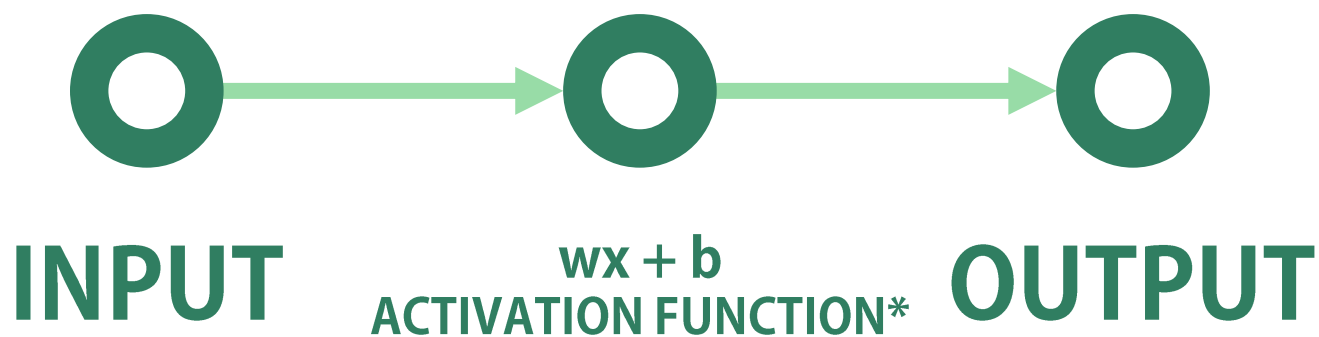
OPTIMIZE (FIND w AND b)
MINIMIZING CROSS ENTROPY

2

INTRODUCTION TO DEEP LEARNING

NEURAL NETWORKS

THE MOST BASIC NEURAL NETWORK LINEAR/LOGISTIC REGRESSION



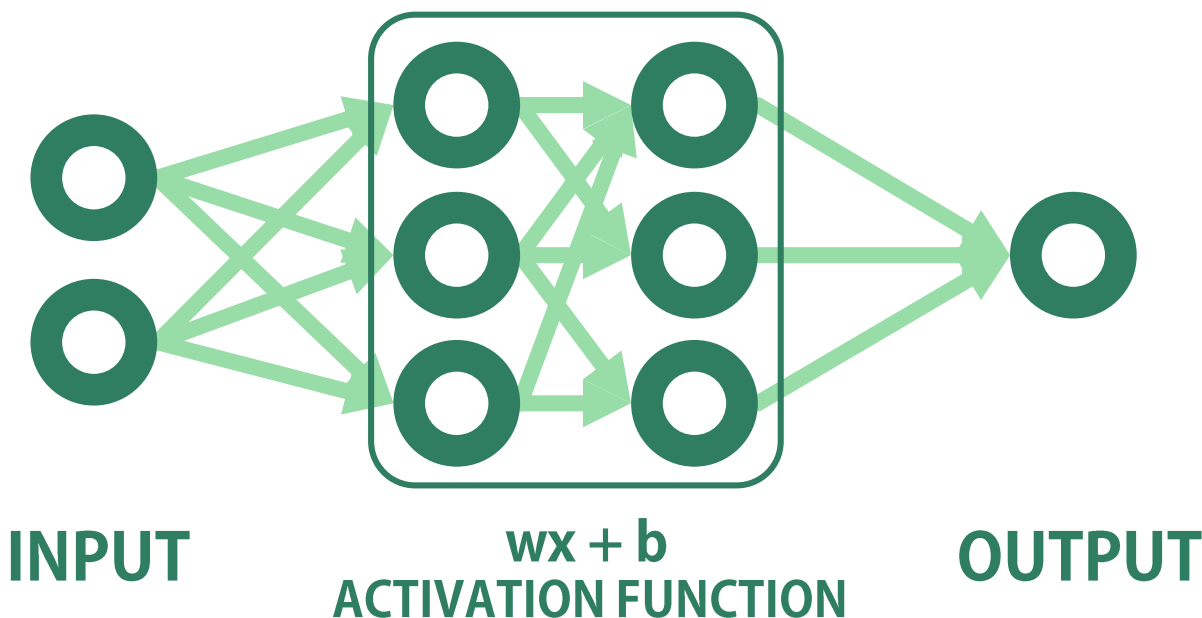
* ACTIVATION FUNCTION IS USED TO INTRODUCE NON-LINEARITY
SIGMOID FUNCTION, RELU, LEAKY RELU

2

INTRODUCTION TO DEEP LEARNING

NEURAL NETWORKS

COMPLEX NEURAL NETWORKS CAN BE BUILT WITH THE SAME IDEA



OPTIMIZE (FIND w AND b AT EACH NODE AND LAYER)
MINIMIZING MEAN SQUARED ERROR/CROSS ENTROPY

SOLUTION OVERVIEW

GENERATIVE ADVERSARIAL NETWORKS AND CONVOLUTIONAL NEURAL NETWORKS ARE USED IN THE CURRENT TASK



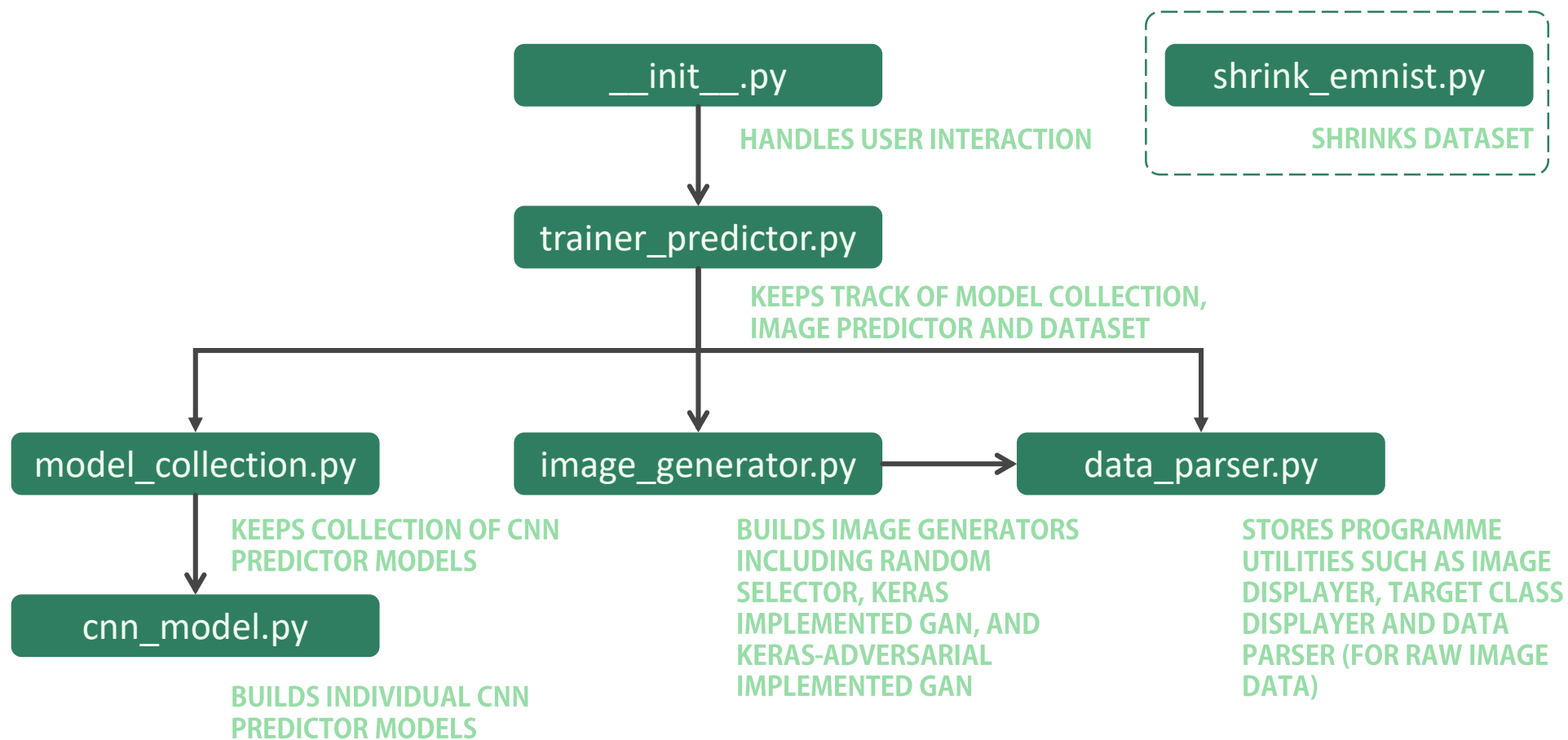
- GENERATIVE ADVERSARIAL NETWORK (GAN)
 - KERAS IMPLEMENTATION
 - KERAS-ADVERSARIAL IMPLEMENTATION
- RANDOM SELECTION FROM SAMPLE

- CONVOLUTIONAL NEURAL NETWORKS (CNN)
 - TESTED 11 HYPERPARAMETER SETTINGS

3
SOLUTION

SOLUTION OVERVIEW

THE SOLUTION CONSISTS MAINLY OF 6 PYTHON FILES, WITH AN ADDITIONAL FILE TO SHRINK THE DATASET



3 SOLUTION

MODEL SELECTION PROCESS

DATASET

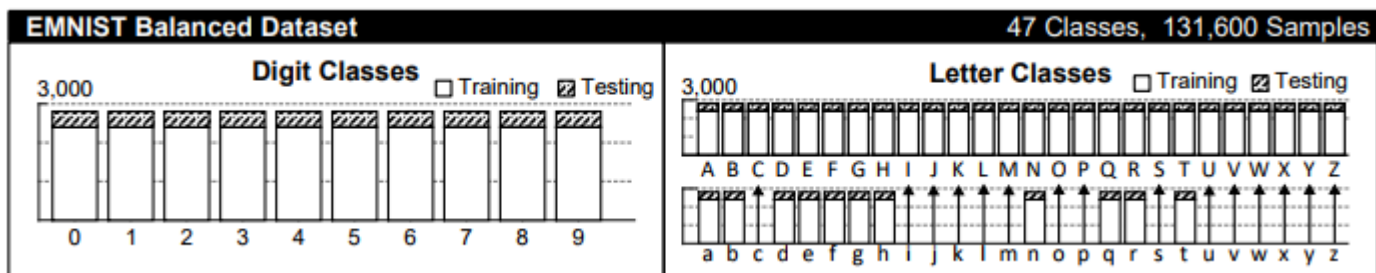
THE EMNIST BALANCED DATASET IS USED FOR THE CURRENT TASK

47 CLASSES - 10 DIGITS AND 37 CHARACTER CLASSES

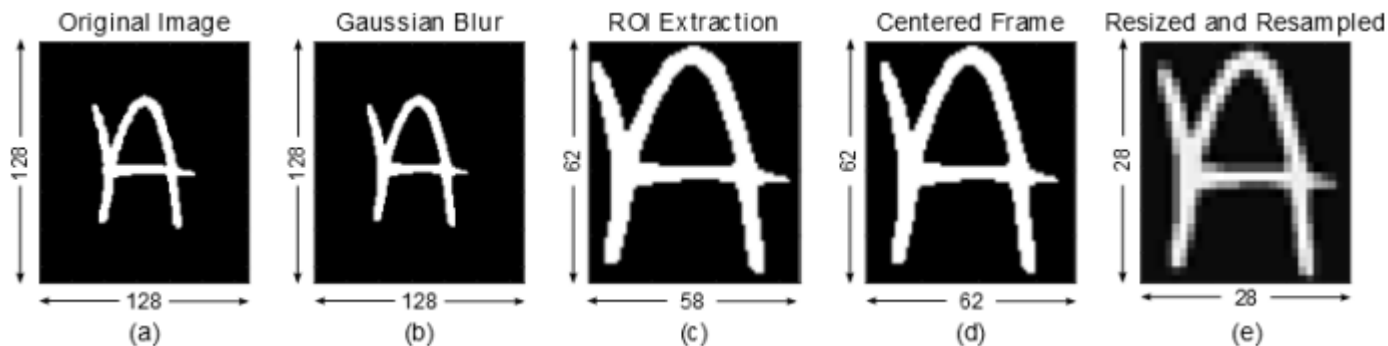
15 LOWER CASE CHARACTER CLASSES MERGED WITH UPPER CASE

2,800 SAMPLES IN EACH CLASS

PRE-CLASSIFIED TRAINING AND TESTING SETS



EACH IMAGE IS 28 x 28 SIZE



3

SOLUTION

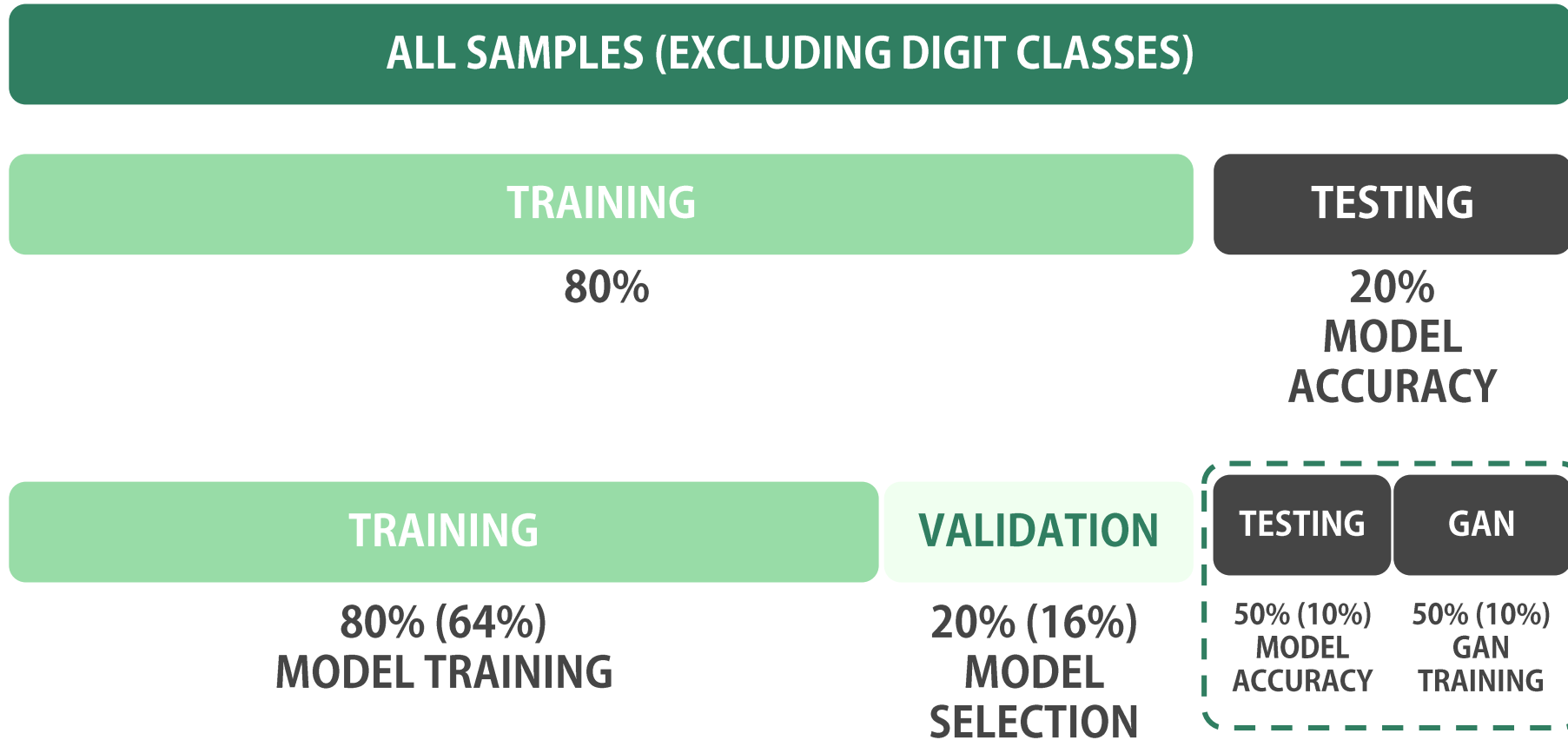
MODEL SELECTION PROCESS

DATA SPLITTING

DATA IS SPLIT INTO TRAINING, VALIDATION AND TESTING SETS

3

SOLUTION

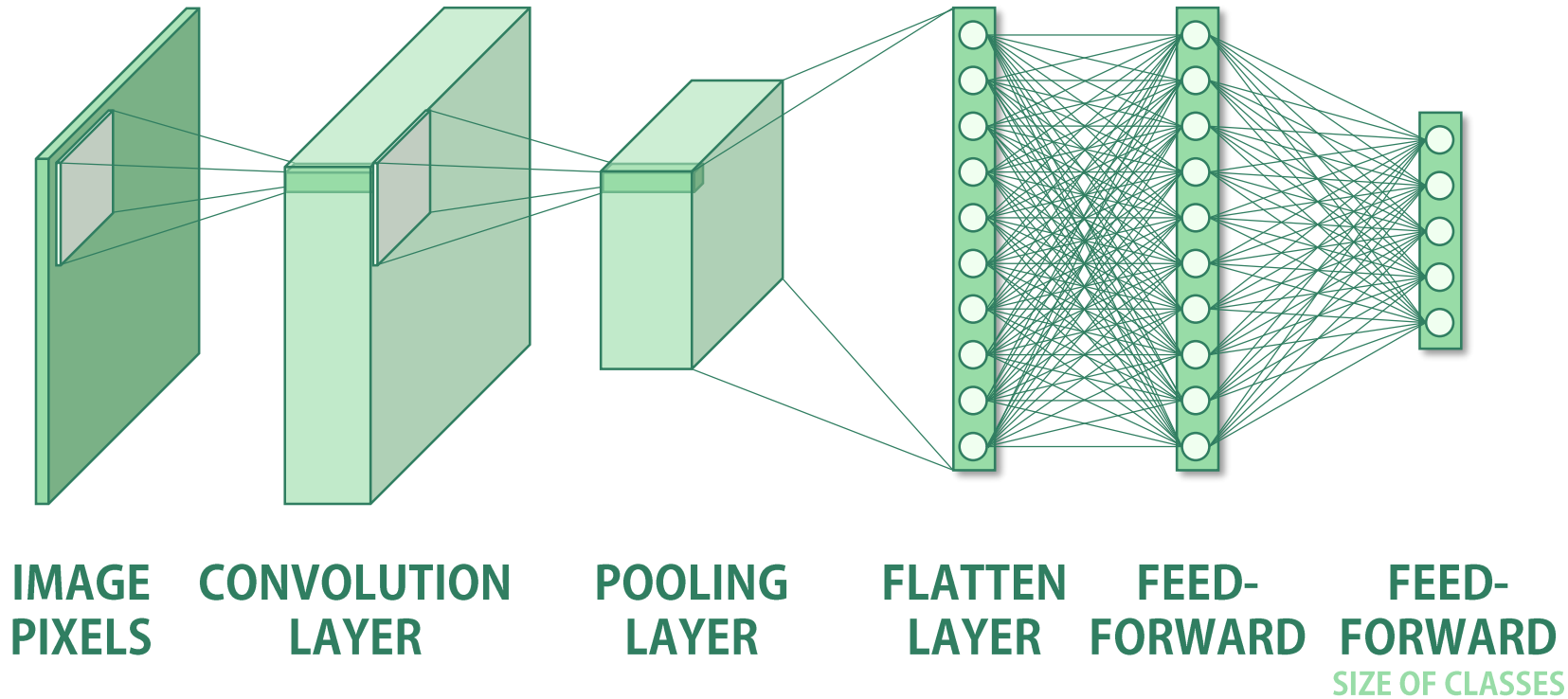


CONVOLUTIONAL NEURAL NETWORKS (CNN)

CNN IS THE MOST COMMON CHOICE FOR IMAGE RECOGNITION

3A

SOLUTION
IMAGE
PREDICTOR



MODEL SELECTION PROCESS

MODELS TESTED

MODEL 7 GAVE THE BEST RESULTS ON VALIDATION SET
TEST SET ACCURACY IS 83.58%

| MODEL | CONVOLUTION LAYER 1 (FILTERS) | DROP OUT | CONVOLUTION LAYER 2 (FILTERS) | DROP OUT | FEED- FORWARD (NEURONS) | DROP OUT | FEED- FORWARD (NEURONS) | DROP OUT | RESULTS (VALID) |
|-------|-------------------------------------|-------------|-------------------------------------|-------------|-------------------------------|-------------|-------------------------------|-------------|--------------------|
| 0 | 32 | 25% | | | 64 | 25% | | | 51.17% |
| 1 | 64 | 25% | | | 64 | 25% | | | 50.78% |
| 2 | 64 | 25% | 32 | 25% | 64 | 25% | | | 39.42% |
| 3 | 64 | 25% | 64 | 25% | 64 | 25% | | | 41.52% |
| 4 | 32 | 25% | 32 | 25% | 64 | 25% | | | 40.50% |
| 5 | 32 | 75% | 32 | 50% | 64 | 25% | | | 53.79% |
| 6 | 32 | 75% | 32 | 75% | 64 | 50% | | | 77.15% |
| 7 | 32 | 75% | 32 | 75% | 128 | 50% | 64 | 50% | 83.62% |
| 8 | 64 | 75% | 32 | 75% | 128 | 50% | 64 | 50% | 76.51% |
| 9 | 64 | 75% | 64 | 75% | 128 | 50% | 64 | 50% | 71.24% |
| 10 | 64 | 75% | 64 | 75% | 256 | 50% | 128 | 50% | 57.59% |

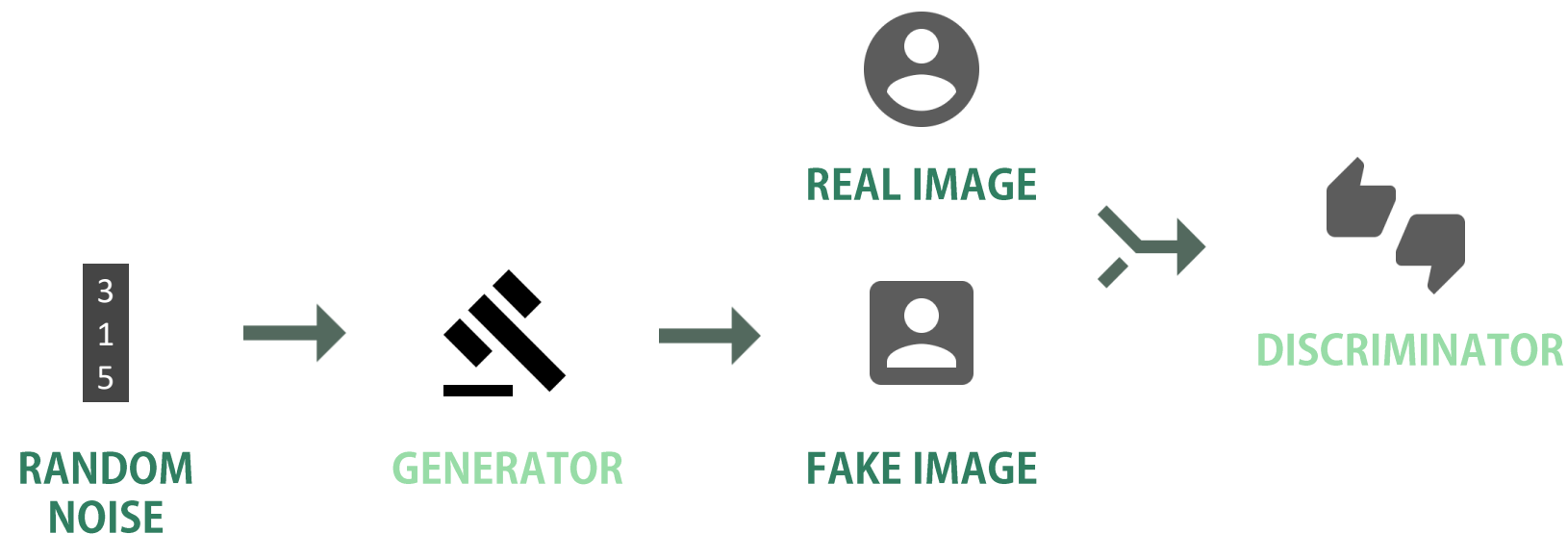


3A

SOLUTION
IMAGE
PREDICTOR

GENERATIVE ADVERSARIAL NETWORKS (GAN)

TWO NEURAL NETWORKS COMPETING WITH EACH OTHER
AT CONVERGE, DISCRIMINATOR UNABLE TO DISTINGUISH FAKE FROM REAL
GENERATOR CREATES REALISTIC IMAGES



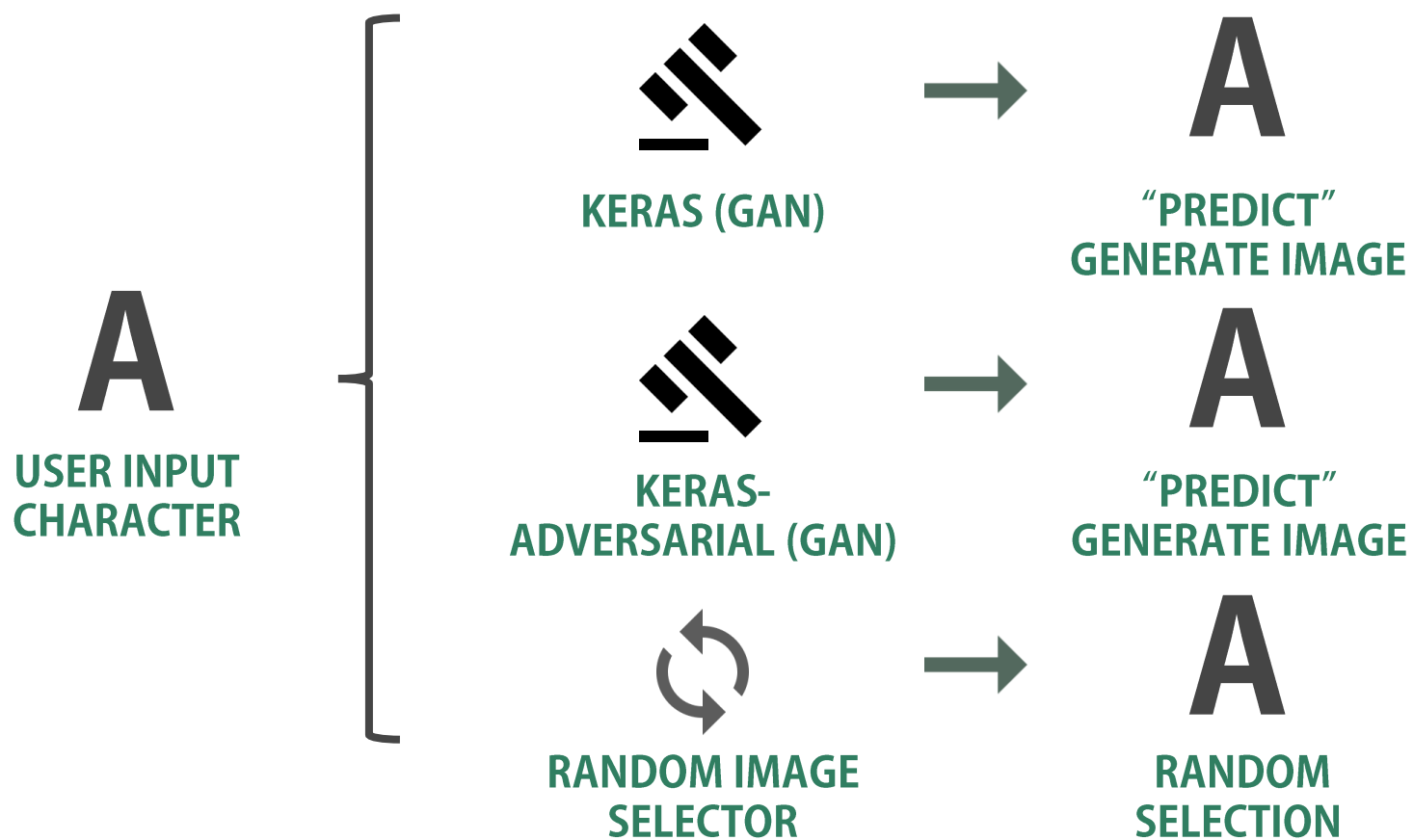
CROSS ENTROPY
DISCRIMINATOR RECOGNIZES AS REAL

CROSS ENTROPY
DISCRIMINATOR DISTINGUISHES
REAL AND FAKE IMAGES

3B
SOLUTION
IMAGE
GENERATOR

IMAGE GENERATOR PROCESS

THE IMAGE GENERATOR ALLOWS USERS TO SELECT FROM ONE OF THREE IMPLEMENTATIONS

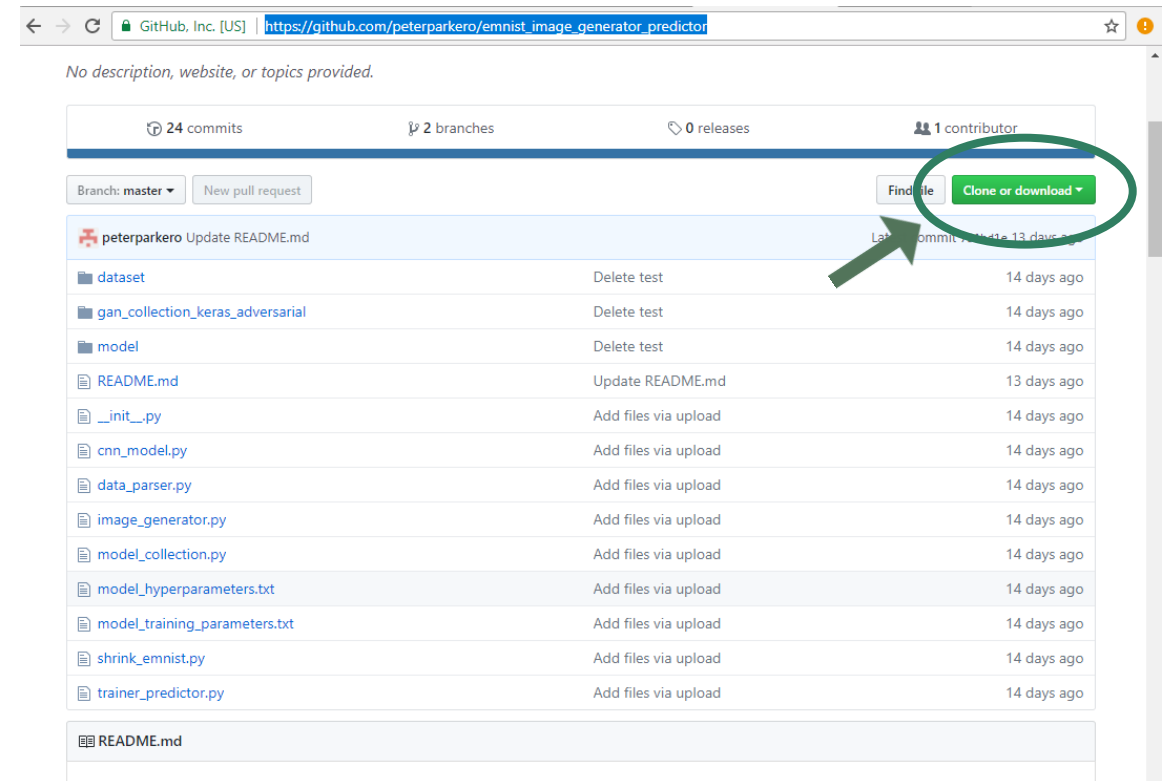
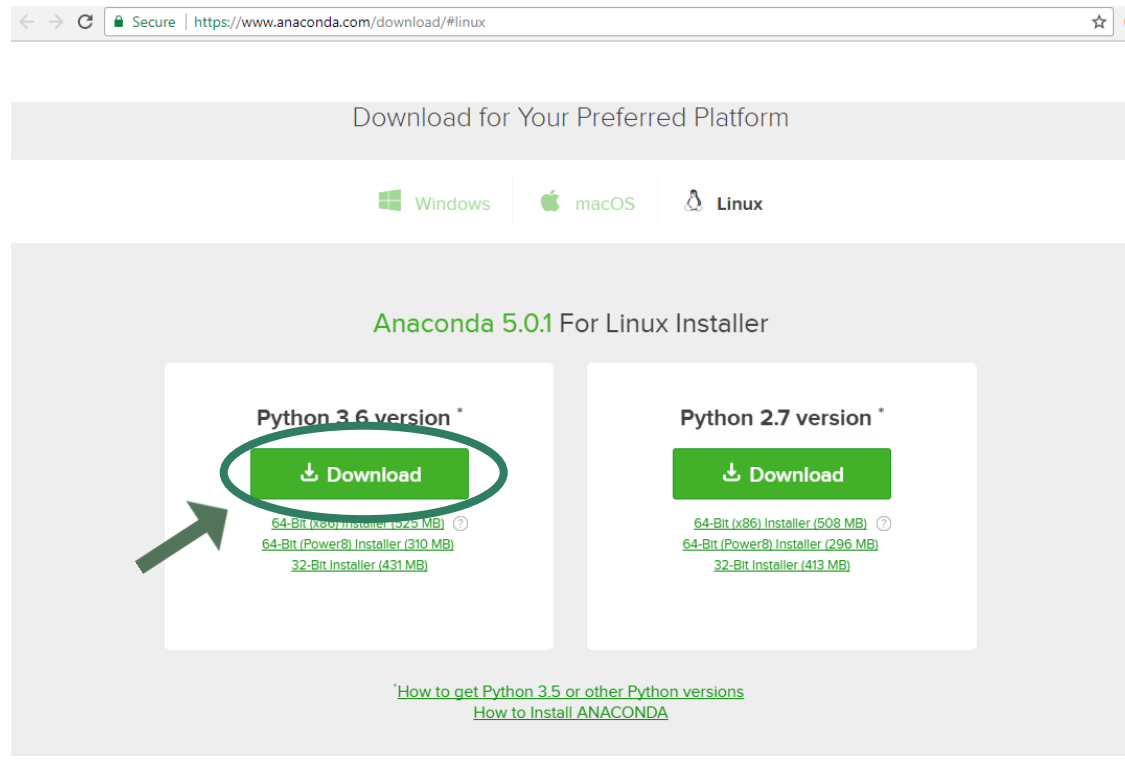


3B

SOLUTION
IMAGE
GENERATOR

4 DEMONSTRATION

STEP 1 - INSTALL PYTHON (ANACONDA) AND REQUIRED PACKAGES DOWNLOAD AND EXTRACT FILES FROM GITHUB

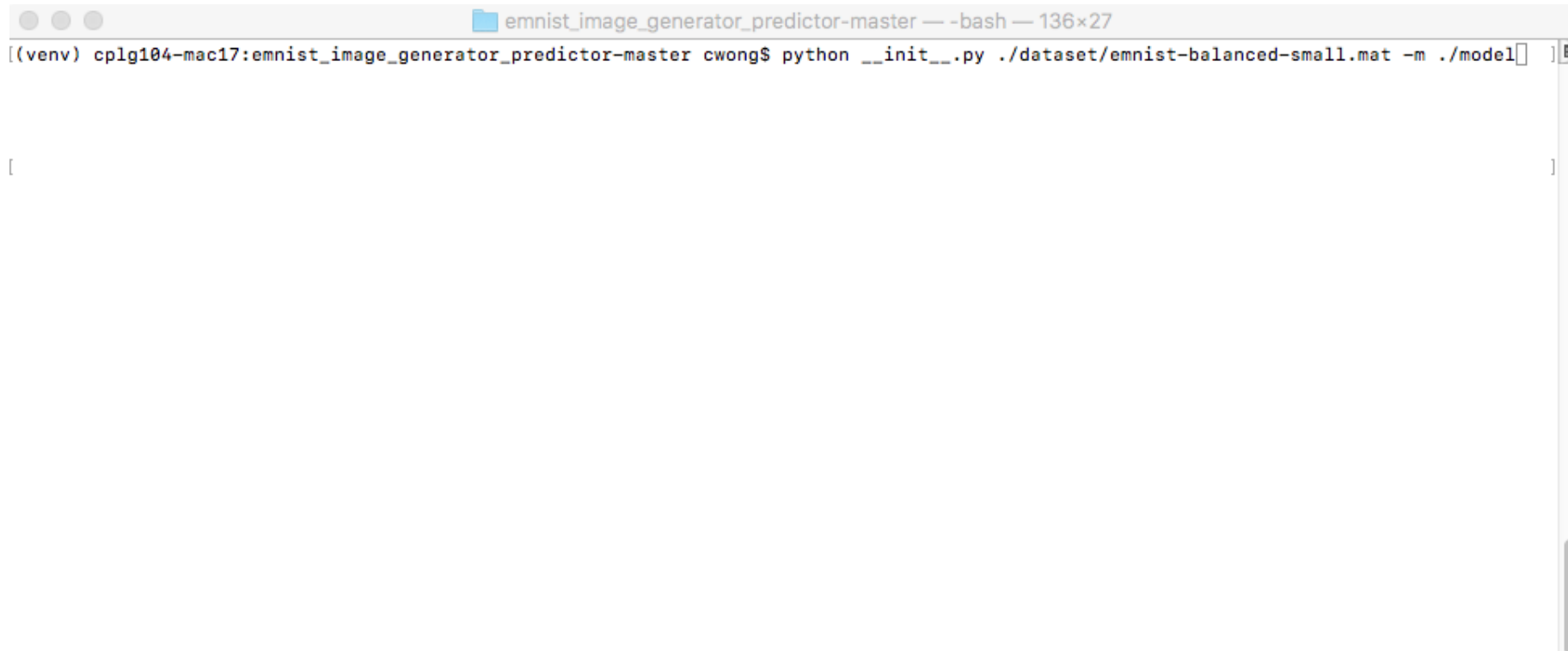


4

DEMONSTRATION

**STEP 2 - OPEN TERMINAL/COMMAND PROMPT
GO TO FILE DIRECTORY AND EXECUTE**

`python __init__.py ./dataset/emnist-balanced-small.mat -m ./model`



```
emnist_image_generator_predictor-master — bash — 136x27
((venv) cplg104-mac17:emnist_image_generator_predictor-master cwong$ python __init__.py ./dataset/emnist-balanced-small.mat -m ./model ]
[ ]
```

4

DEMONSTRATION

STEP 3 - REACH MAIN MENU UPON COMPLETION OF LOADING REQUIRED PACKAGES AND MODELS WARNINGS FROM TENSORFLOW CAN BE IGNORED

```
emnist_image_generator_predictor-master — Python __init__.py ./dataset/emnist-balanced-small.mat -m ./model — 136x27
[(venv) cplg104-mac17:emnist_image_generator_predictor-master cwong$ python __init__.py ./dataset/emnist-balanced-small.mat -m ./model ]
Using TensorFlow backend.
/Library/Frameworks/Python.framework/Versions/3.6/lib/python3.6/importlib/_bootstrap.py:205: RuntimeWarning: compiletime version 3.5 of
module 'tensorflow.python.framework.fast_tensor_util' does not match runtime version 3.6
    return f(*args, **kwargs)
[2017-11-17 20:19:43.124519: I tensorflow/core/platform/cpu_feature_guard.cc:137] Your CPU supports instructions that this TensorFlow bin
ary was not compiled to use: SSE4.1 SSE4.2 AVX

Model collection file loaded.

Would you like to Train a Model, Generate and Predict a Image, Save Model Collection, Load Model Collection, Display Results, or Exit
[Train/Predict/Save/Load/Results/Exit]
█
```

4

DEMONSTRATION

TRAIN - TYPE “train” TO TRAIN NEW MODELS BASED ON SPECIFIED HYPERPARAMETER SETTINGS ENTER HYPERPARAMETER AND TRAINING PARAMETER FILE PATHS

```
Would you like to Train a Model, Generate and Predict a Image, Save Model Collection, Load Model Collection, Display Results, or Exit  
[Train/Predict/Save/Load/Results/Exit]  
Train
```

```
Please input the hyperparameter_layer_filepath.  
./model_hyperparameters.txt
```

```
Please input the training_parameters_filepath.  
./model_training_parameters.txt
```

```
Please input whether you like to retrain the models if the same hyperparameter settings are given. [Y/N]  
Y
```

| Layer (type) | Output Shape | Param # |
|---|--------------------|---------|
| conv2d_1 (Conv2D) | (None, 28, 28, 32) | 320 |
| max_pooling2d_1 (MaxPooling2D) | (None, 14, 14, 32) | 0 |
| dropout_1 (Dropout) | (None, 14, 14, 32) | 0 |
| batch_normalization_1 (Batch Normalization) | (None, 14, 14, 32) | 128 |
| flatten_1 (Flatten) | (None, 6272) | 0 |
| dense_1 (Dense) | (None, 64) | 401472 |
| dropout_2 (Dropout) | (None, 64) | 0 |
| dense_2 (Dense) | (None, 47) | 3055 |

Total params: 404,975
Trainable params: 404,911
Non-trainable params: 64

```
Epoch 1/10  
2560/2560 [=====] - 6s 2ms/step - loss: 3.4509 - acc: 0.1465  
Epoch 2/10  
2560/2560 [=====] - 3s 1ms/step - loss: 2.4600 - acc: 0.3512  
Epoch 3/10  
2560/2560 [=====] - 3s 1ms/step - loss: 2.4600 - acc: 0.3512
```

PREDICT - TYPE "predict" TO GENERATE AND PREDICT A CHARACTER IMAGE
ENTER "basic" FOR RANDOM IMAGE SELECTOR

What character would you like to generate image on and make a prediction?
 Note that not all characters are available for prediction.
 Unavailable characters are: c, i, j, k, l, m, o, p, s, u, v, w, x, y and z.
 a

Preview of selected image:

This character image is predicted to be:
a

PREDICT - TYPE "predict" TO GENERATE AND PREDICT A CHARACTER IMAGE
ENTER "keras" FOR GAN IMPLEMENTED USING KERAS

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PREDICT - TYPE "predict" TO GENERATE AND PREDICT A CHARACTER IMAGE
ENTER "keras_adversarial" FOR GAN IMPLEMENTED USING KERAS-ADVERSARIAL

What character would you like to generate image on and make a prediction?
Note that not all characters are available for prediction.
Unavailable characters are: c, i, j, k, l, m, o, p, s, u, v, w, x, y and z.
a
Preview of selected image:

```
The performance of this model:
Validation Score - 83.62%
Test Set Score - 83.58%

This character image is predicted to be:
a
```


4

DEMONSTRATION

SAVE - TYPE “save” TO SAVE THE CURRENT MODEL COLLECTION TO A FOLDER

```
Would you like to Train a Model, Generate and Predict a Image, Save Model Collection, Load Model Collection, Display Results, or Exit
[Train/Predict/Save/Load/Results/Exit]
Save

Please input the output folder path.
./model

Files saved to ./model.
```

4

DEMONSTRATION

LOAD - TYPE “load” TO LOAD A PREVIOUSLY SAVED MODEL COLLECTION FROM A FOLDER
LOADING THE PROGRAM WITHOUT “-m ./model” AND LOADING THE MODELS AFTER
REACHING THE MENU PAGE IS EQUIVALENT TO RUNNING
“python __init__.py ./dataset/emnist-balanced-small.mat - m ./model”

```
Would you like to Train a Model, Generate and Predict a Image, Save Model Collection, Load Model Collection, Display Results, or Exit  
[Train/Predict/Save/Load/Results/Exit]  
Load
```

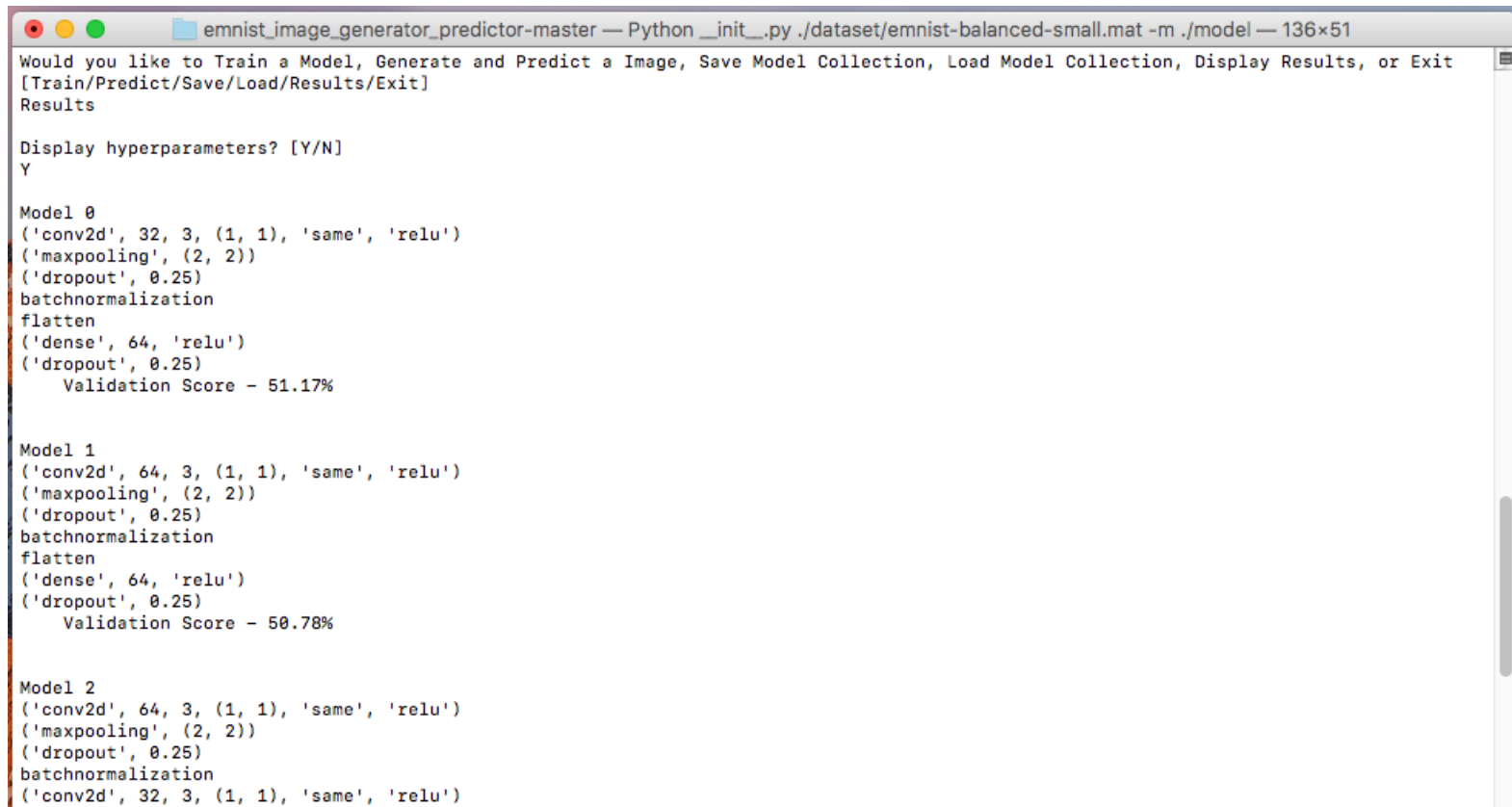
```
Please input the folder path to load the model collection from.  
./model
```

```
Model collection file loaded.
```

4

DEMONSTRATION

RESULTS - TYPE “results” TO DISPLAY THE VALIDATION SCORES OF ALL TRAINED MODELS FROM THE MODEL COLLECTION



```
emnist_image_generator_predictor-master — Python __init__.py ./dataset/emnist-balanced-small.mat -m ./model — 136x51
Would you like to Train a Model, Generate and Predict a Image, Save Model Collection, Load Model Collection, Display Results, or Exit
[Train/Predict/Save/Load/Results/Exit]
Results

Display hyperparameters? [Y/N]
Y

Model 0
('conv2d', 32, 3, (1, 1), 'same', 'relu')
('maxpooling', (2, 2))
('dropout', 0.25)
batchnormalization
flatten
('dense', 64, 'relu')
('dropout', 0.25)
    Validation Score - 51.17%

Model 1
('conv2d', 64, 3, (1, 1), 'same', 'relu')
('maxpooling', (2, 2))
('dropout', 0.25)
batchnormalization
flatten
('dense', 64, 'relu')
('dropout', 0.25)
    Validation Score - 50.78%

Model 2
('conv2d', 64, 3, (1, 1), 'same', 'relu')
('maxpooling', (2, 2))
('dropout', 0.25)
batchnormalization
('conv2d', 32, 3, (1, 1), 'same', 'relu')
```

4

DEMONSTRATION

EXIT - TYPE “exit” TO EXIT THE PROGRAM

```
Would you like to Train a Model, Generate and Predict a Image, Save Model Collection, Load Model Collection, Display Results, or Exit  
[Train/Predict/Save/Load/Results/Exit]  
Exit  
(venv) cplg104-mac17:emnist_image_generator_predictor-master cwong$
```

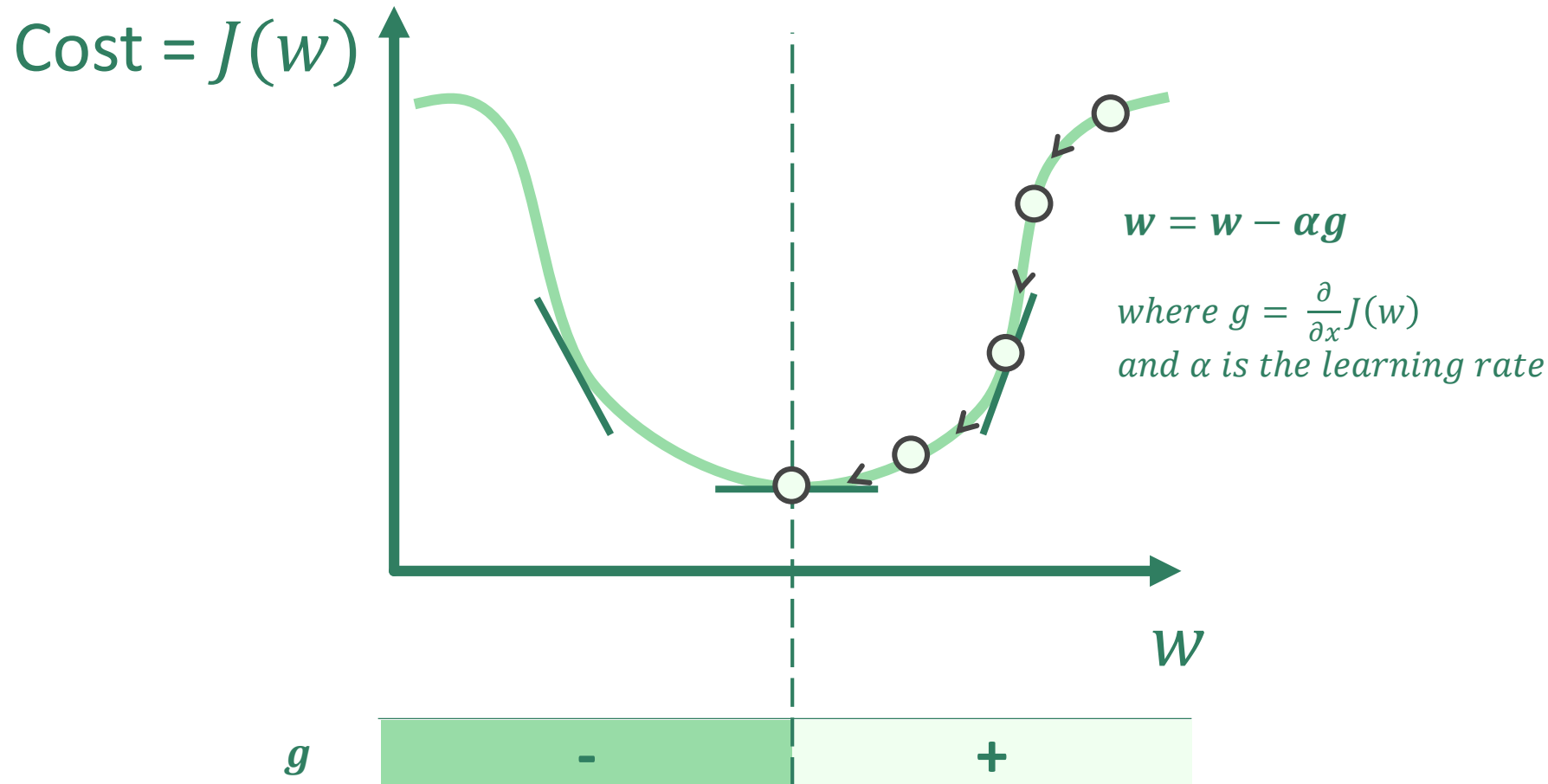
APPENDIX

- GRADIENT DESCENT
- ADAM
- BATCH NORMALIZATION
- BACKPROPAGATION
- DROPOUT
- FORMULA

5

APPENDIX

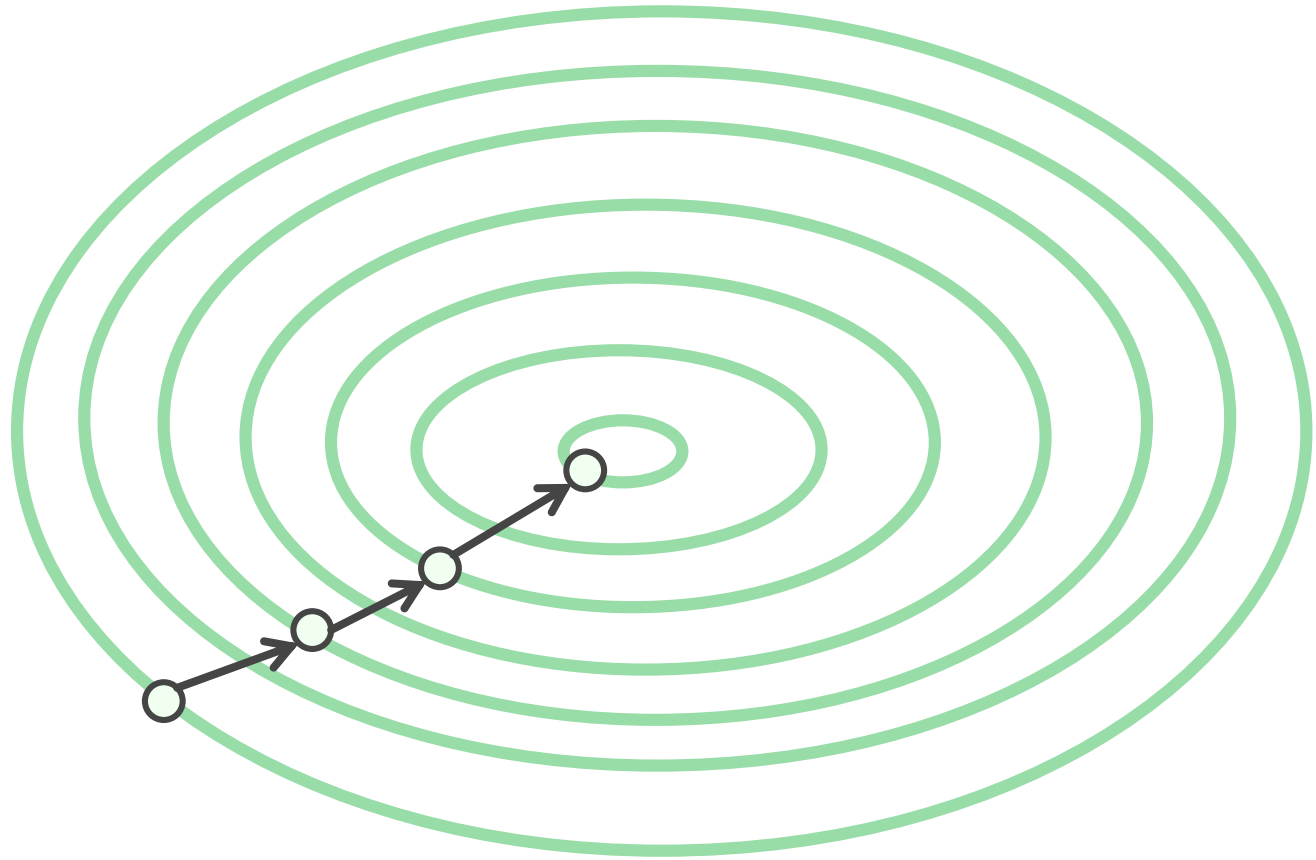
GRADIENT DESCENT



GRADIENT DESCENT

5

APPENDIX



5

APPENDIX

OTHER OPTIMIZATION TECHNIQUES

MOMENTUM

INCLUDES A “MOMENTUM” FACTOR IN EACH UPDATE
LINEAR COMBINATION OF GRADIENT OF PREVIOUS UPDATE

$$\begin{aligned}\Delta w &= -\alpha g + \mu \Delta w \\ w &= w + \Delta w\end{aligned}$$

ADAPTIVE GRADIENT (ADAGRAD)

INCREASES LEARNING RATE FOR MORE SPARSE PARAMETERS
WORKS WELL WITH SPARSE GRADIENTS

$$w_t = w_{t-1} - \alpha \frac{g_{t-1}}{\sqrt{\sum_{i=1}^t g_i^2}}$$

ROOT MEAN SQUARE PROPAGATION (RMSPROP)

EXPONENTIAL MOVING AVERAGE CONTROLLED BY γ
WORKS WELL IN NON-STATIONARY SETTINGS (E.G. TIME SERIES)

$$\begin{aligned}R_t &= \gamma R_{t-1} + (1 - \gamma) g_{t-1}^2 \\ w_t &= w_{t-1} - \alpha \frac{g_{t-1}}{\sqrt{R_t}}\end{aligned}$$

5

APPENDIX

OTHER OPTIMIZATION TECHNIQUES

ADAPTIVE MOMENT ESTIMATION (ADAM)

COMBINE THE ADVANTAGES OF ADAGRAD AND RMSPROP

$$M_0 = 0, R_0 = 0$$

For $t = 1$ to T :

$$M_t = \beta_1 M_{t-1} + (1 - \beta_1) g_{t-1}$$

$$R_t = \beta_2 R_{t-1} + (1 - \beta_2) (g_{t-1})^2$$

$$\widehat{M}_t = \frac{M_t}{1 - \beta_1^t}$$

$$\widehat{R}_t = \frac{R_t}{1 - \beta_2^t}$$

$$W_t = W_{t-1} - \alpha \frac{\widehat{M}_t}{\sqrt{\widehat{R}_t + \epsilon}}$$

INITIALIZE

1ST MOMENT ESTIMATE

2ND MOMENT ESTIMATE

1ST MOMENT BIAS CORRECTION

2ND MOMENT BIAS CORRECTION

UPDATE PROPORTIONAL TO

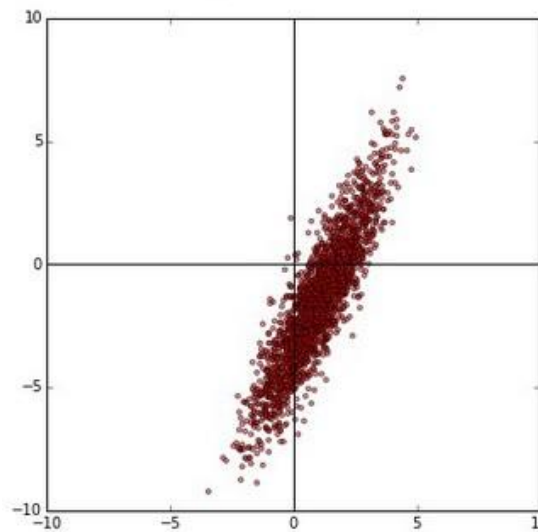
$$\frac{\text{average gradient}}{\sqrt{\text{average squared gradient}}}$$

BATCH NORMALIZATION

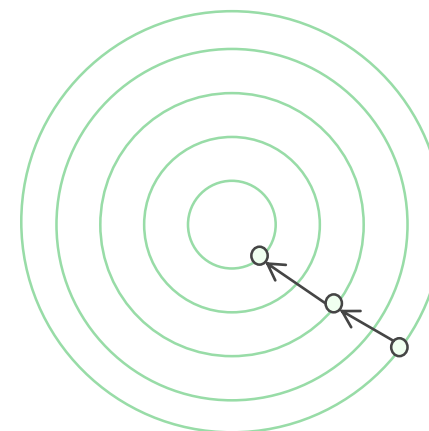
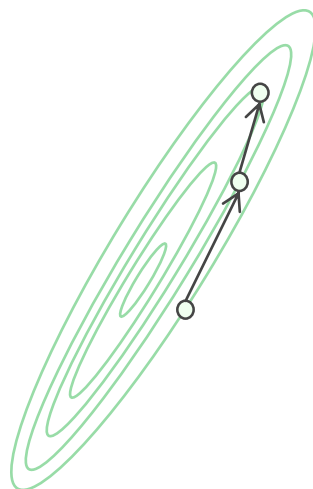
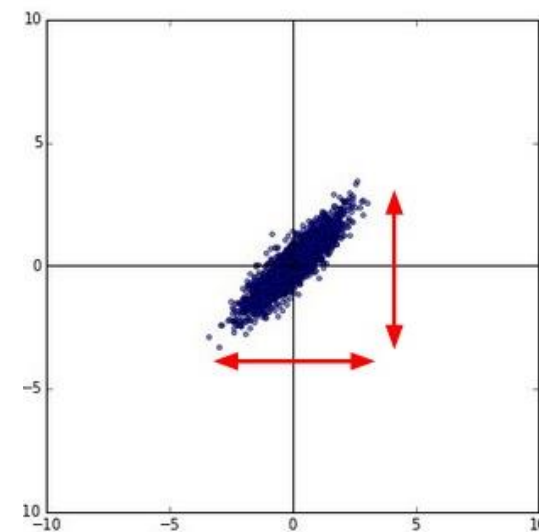
5

APPENDIX

original data



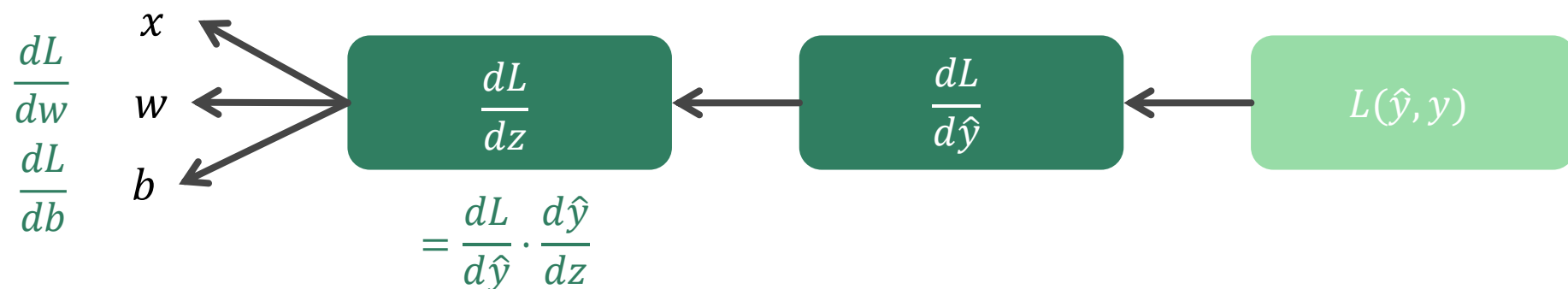
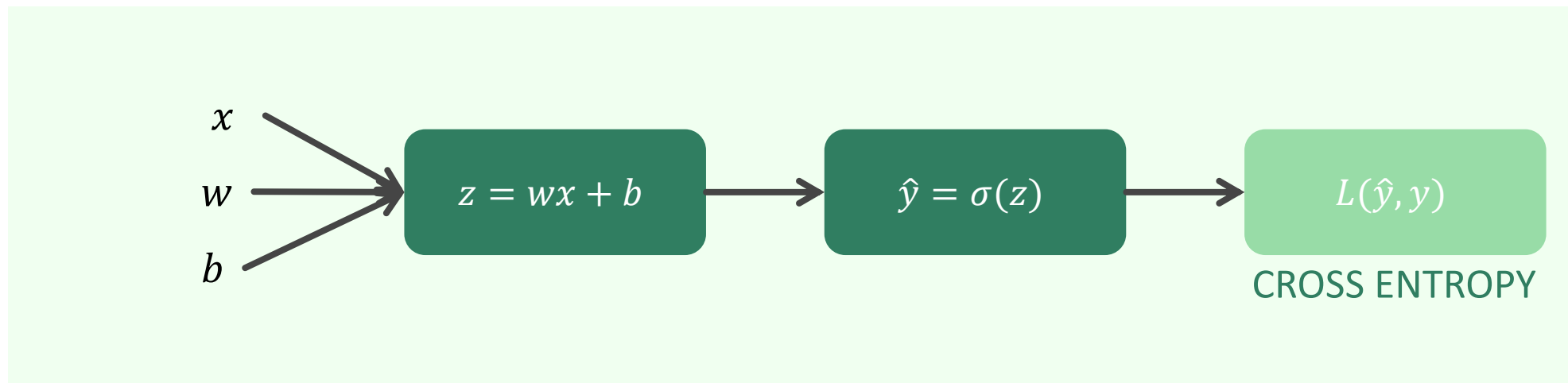
normalized data



5

APPENDIX

BACKPROPAGATION



APPLY CHAIN RULE

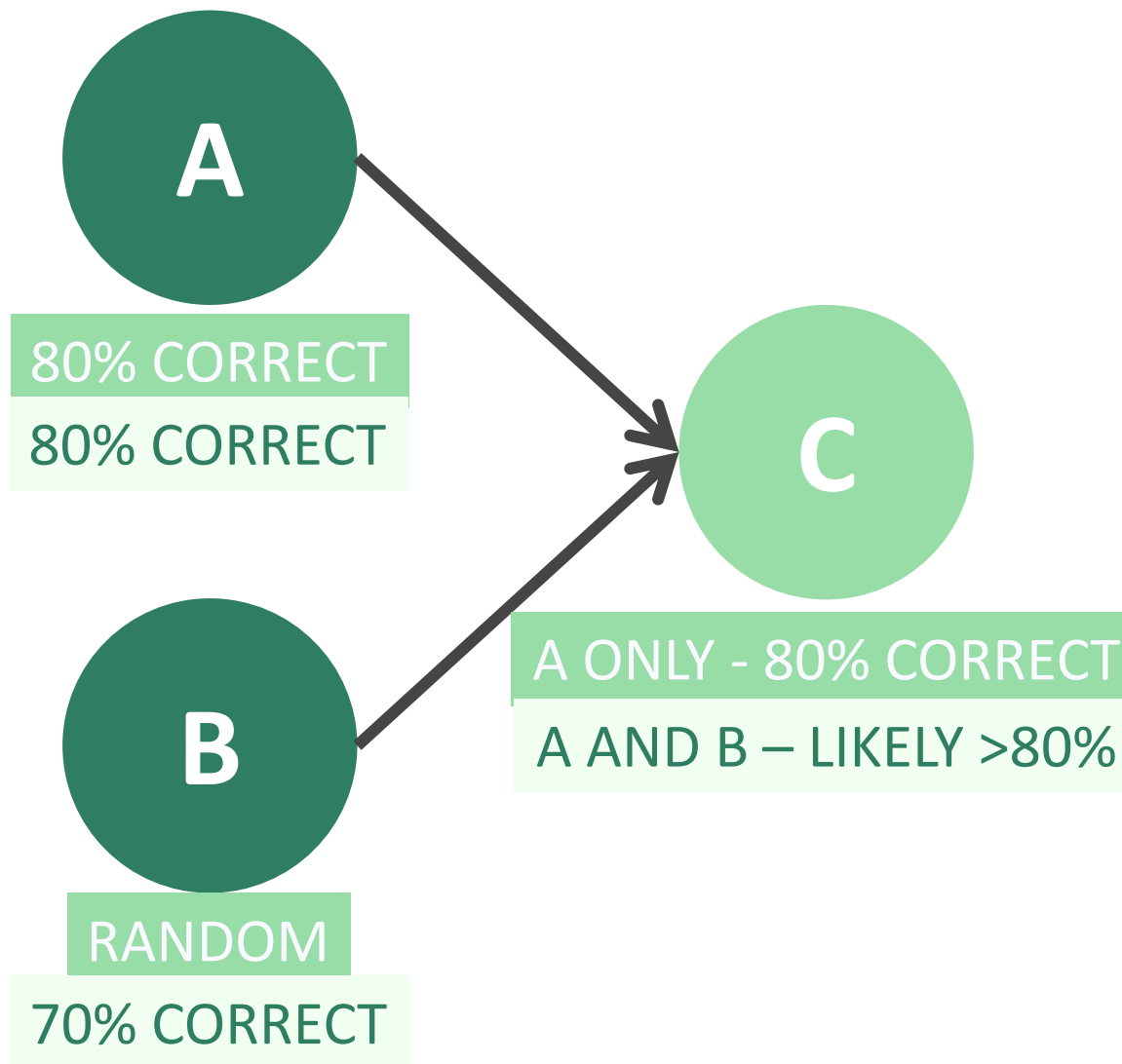
APPLY CHAIN RULE

* CROSS ENTROPY: $L(\hat{y}, y) = -y \log(\hat{y}) - (1 - y) \log(1 - \hat{y})$

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APPENDIX

DROPOUT



5

APPENDIX

FORMULA

LINEAR REGRESSION

$$y = \sum_{i=1}^n w_i x_i + b$$

MEAN SQUARED ERROR

$$MSE = \frac{1}{n} \sum_{i=1}^n (\widehat{Y}_i - Y_i)^2$$

LOGISTIC REGRESSION

$$y = \frac{1}{1 + e^{-(\sum_{i=1}^n w_i x_i + b)}}$$

CROSS ENTROPY

$$CE = \sum_{i=1}^n y^{(i)} \log(\widehat{y}^{(i)}) + (1 - y^{(i)}) \log(1 - \widehat{y}^{(i)})$$