







# Handwriting recognition in Bengali




Mohammed Hannan  
Capstone Project



# Bengali Characters



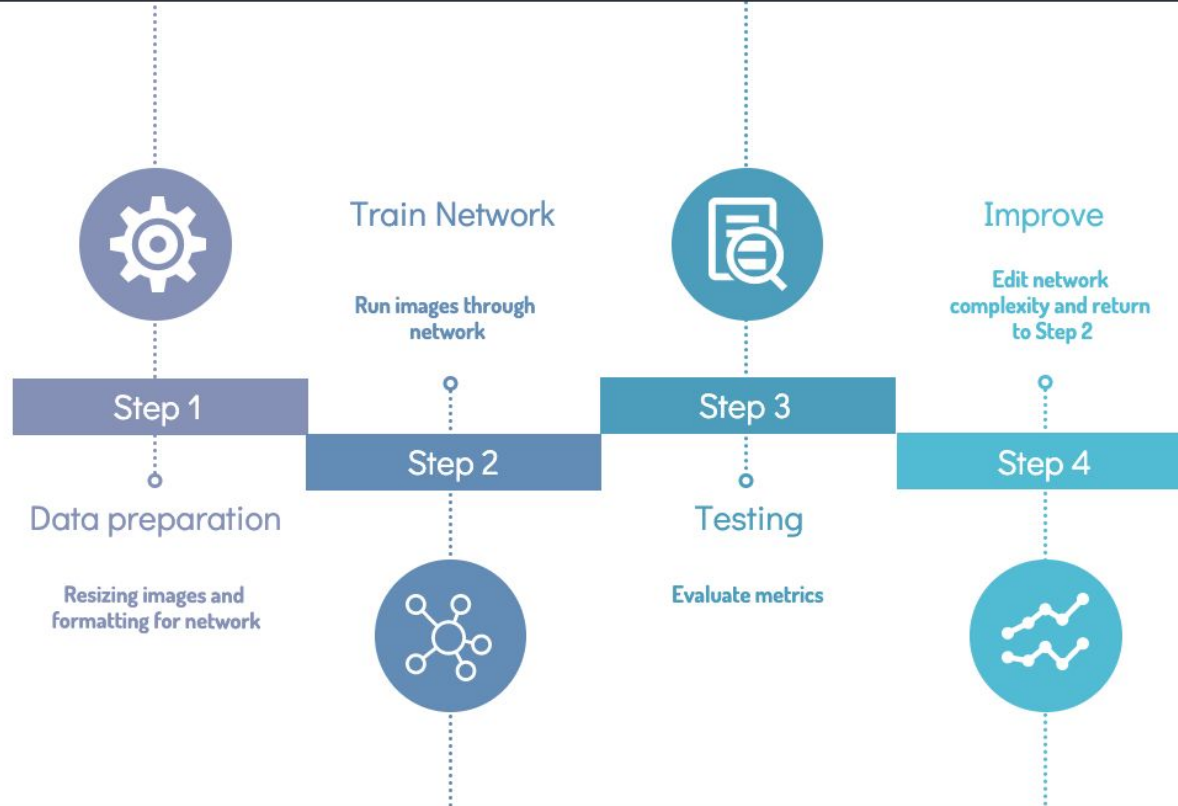
ই	কৃ	বি	রী	মৌ	কৈ	পু	টু	1
								2
								3



Accents can exist below the root, on the left or the right, or on both the left and right

There are over 10,000 combinations of root, vowel and consonant accents. The aim is to use approximately 1000 different combinations to be able to classify the other 9000+ combinations.

# Convolutional Neural Network

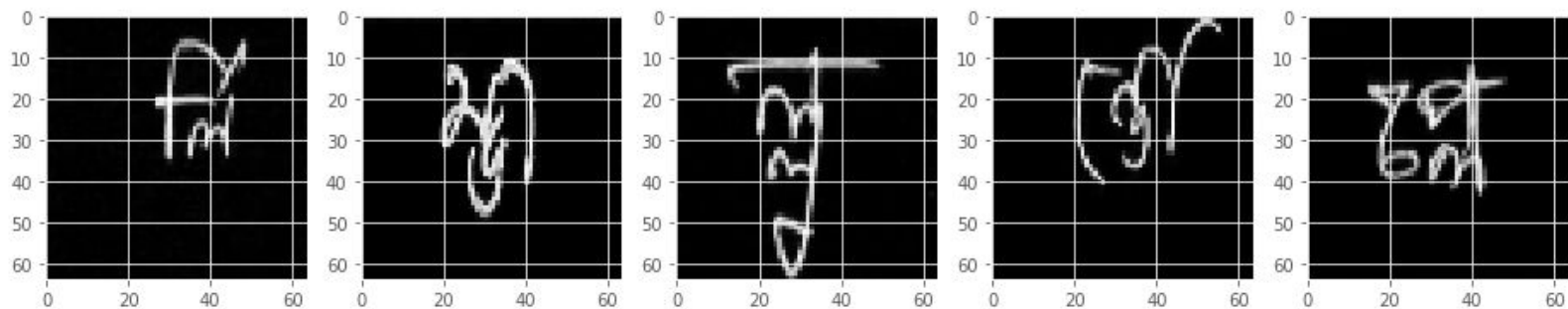




# Model performance

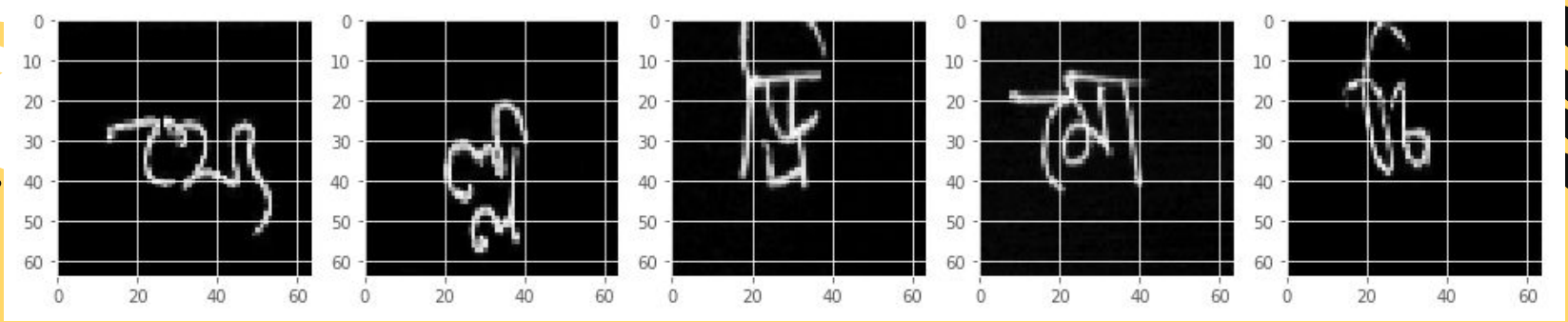
How well are the characters identified?

# First Model



	predicted						true					
	root	r_comp	vowel	v_comp	consonant	c_comp	root	r_comp	vowel	v_comp	consonant	c_comp
image_0	72	ଦ	2	ି	2	ର୍	124	ଲ	2	ି	2	ର୍
image_1	119	ଛ	1	ା	0	0	119	ଛ	1	ା	0	0
image_2	132	ଜ୍ଞ	4	ୁ	0	0	132	ଜ୍ଞ	4	ୁ	0	0
image_3	71	ଥ	9	ୋ	2	ର୍	86	ଛ	9	ୋ	0	0
image_4	153	ଞ୍ଜ	7	େ	0	0	101	ଜ୍ଞ	7	େ	0	0

# Final Model



	predicted						true					
	root	r_comp	vowel	v_comp	consonant	c_comp	root	r_comp	vowel	v_comp	consonant	c_comp
image_0	71	ଥ	7	େ	4	ା	71	ଥ	7	େ	4	ା
image_1	132	୩	0	0	0	0	132	୩	0	0	0	0
image_2	56	ଝ	2	ି	5	ୱ	56	ଝ	2	ି	5	ୱ
image_3	115	ମ	9	ୋ	0	0	115	ମ	9	ୋ	0	0
image_4	38	ଟ	2	ି	0	0	38	ଟ	2	ି	0	0



92.15%

Root prediction accuracy

98.00%

Vowel prediction accuracy


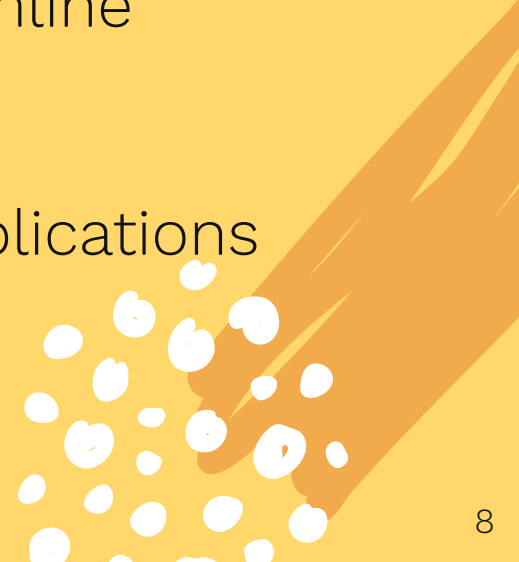
97.84%

Consonant prediction accuracy





# Why?

- Handwriting recognition systems can help digitally input information to streamline operations (bank cheques, postal addresses)
  - Can also be used in translation applications
- 
- 





# Thanks!

Any questions?



# Appendix

## Network Structure

```
inputs = Input(shape = (64, 64, 1))

model = Conv2D(filters=32, kernel_size=(3, 3), padding='SAME', activation='relu', input_shape=(64, 64, 1))(inputs)
model = Conv2D(filters=32, kernel_size=(3,3), padding='SAME', activation='relu')(model)
model = BatchNormalization(momentum=0.15)(model)
model = MaxPool2D(pool_size=(2, 2))(model)
model = Conv2D(filters=32, kernel_size=(5,5), padding='SAME', activation='relu')(model)
model = BatchNormalization(momentum=0.15)(model)
model = Dropout(rate=0.2)(model)

model = Conv2D(filters=64, kernel_size=(3, 3), padding='SAME', activation='relu')(model)
model = BatchNormalization(momentum=0.15)(model)
model = MaxPool2D(pool_size=(2, 2))(model)
model = Conv2D(filters=64, kernel_size=(5,5), padding='SAME', activation='relu')(model)
model = BatchNormalization(momentum=0.15)(model)
model = Dropout(rate=0.2)(model)

model = Conv2D(filters=128, kernel_size=(3, 3), padding='SAME', activation='relu')(model)
model = BatchNormalization(momentum=0.15)(model)
model = Conv2D(filters=128, kernel_size=(5,5), padding='SAME', activation='relu')(model)
model = BatchNormalization(momentum=0.15)(model)
model = MaxPool2D(pool_size=(2, 2))(model)
model = Dropout(rate=0.2)(model)

model = Flatten()(model)
model = Dense(1024, activation = "relu")(model)
model = Dropout(rate=0.3)(model)
dense = Dense(512, activation = 'relu')(model)

root = Dense(168, activation = 'softmax')(dense)
vowel = Dense(11, activation = 'softmax')(dense)
consonant = Dense(7, activation = 'softmax')(dense)

model = Model(inputs=inputs, outputs=[root, vowel, consonant])
```

# Network improvement

