### Al Sudoku Solver

Marine AI Practice in English Mid-term Report

# Sometimes, Sudoku is too hard...

### Let's make it a bit easier

Use Image recognition to recognize written digits

Select the sudoku problem contours from an image map the numbers and solve the problem

Create an easy interface to use the program

### PROJECT SETUP / BACKGROUND

### What tools are we using (so far)

### Python

- Tensorflow (Keras)
- Cv2
- Numpy
- MatPlotLib

### **Information**

### Kaggle

- Deep Learning Course
- Computer Vision Course

### Research

- No Routing Needed Between Capsules (Adam Byerly, Tatiana Kalganova, Ian Kear)
- Sudo OpenCV Blog
- Tensorflow Documentation

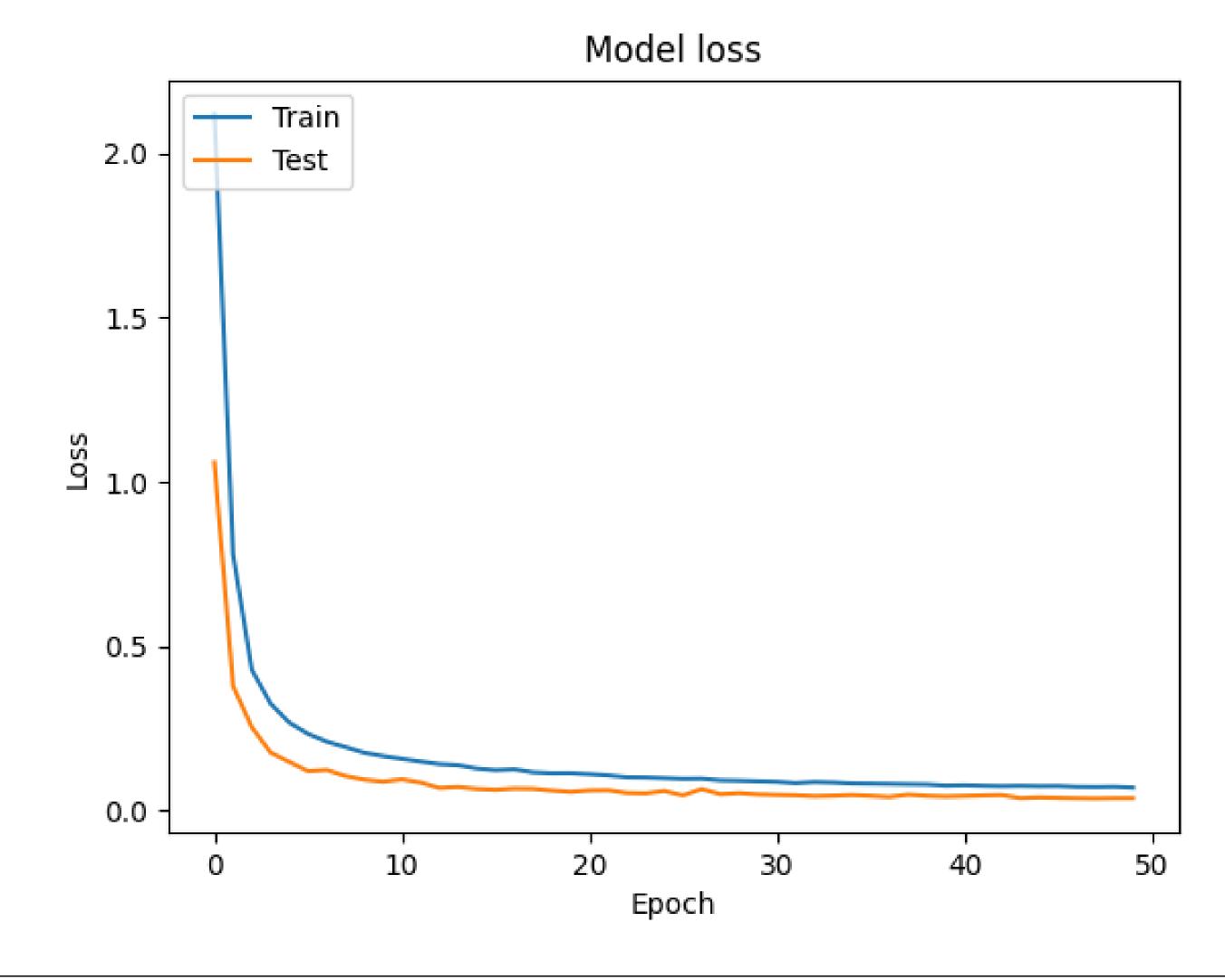
### MNIST dataset

### Our current model

Current Accuracy: 98.53%

Current Loss: 4.61%

```
# Create the model with Input layer instead of input_shape
self.model = tf.keras.Sequential([]
    # Input layer specifying the input shape
    layers.Input(shape=[28, 28, 1]), # Input layer
    # data augmentation
    layers.RandomRotation(fill_mode='constant', fill_value=0.0, factor=0.3),
    # First Convolutional Block (3 layers of Conv2D with 64 filters)
    layers.Conv2D(filters=64, kernel_size=3, activation='relu', padding='same'),
    layers.MaxPool2D(),
    layers.Conv2D(filters=64, kernel_size=3, activation='relu',padding='same'),
    layers.MaxPool2D(),
    layers.Conv2D(filters=64, kernel_size=3, activation='relu',padding='same'),
    layers.MaxPool2D(),
    # Second Convolutional Block
    layers.Dense(activation='relu', units=64),
    layers.Dropout(0.3),
    layers.Conv2D(filters=64, kernel_size=3, activation='relu',padding='same'),
  Hermansen, 1 hour ago • feat: enhance prediction output in sudokuNet cl...
    # Head
    layers.Flatten(),
    layers.Dense(units=10, activation="softmax"),
```



### Key Problems and Current Solutions

### **Data preprocessing**

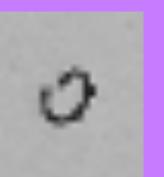
Finding a good way to process our own images of handwritten numbers

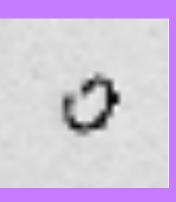
### **Problematic Numbers**

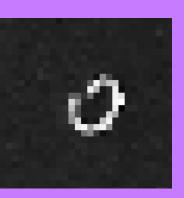
Since you'll say more in a bit, you can keep these short.

### Finding good ways to process our handwritten data









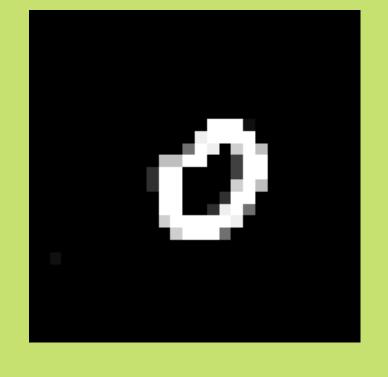


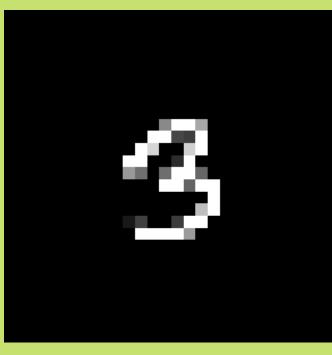
1/1		
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1/1	Model loaded successfull	y!
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1/1	5	
Predicting 6daniel.png  1/1		
Predicting 6daniel.png 1/1	1/1 ————	0s
1/1	5	
Predicting 9shen.png  1/1	Predicting 6daniel.png	
Predicting 9shen.png 1/1	1/1 ———	0s
1/1	5	
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	5	

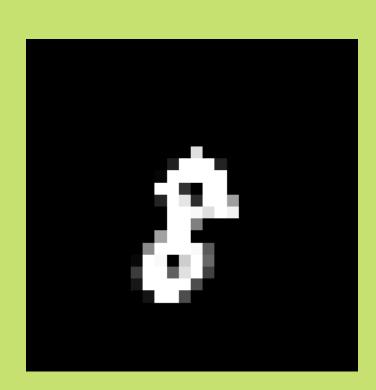
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1/1		1mads.png
8		
Pred <b>1/1</b>		6daniel.pn
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1/1		2mads.png
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8 Brod	licting	4mads.png
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8 Pred	licting	5daniel.png
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8		

	8shen.png		
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2: 5 (4.98			
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2: 7 (13.6			
3: 8 (4.4% Predicting	0shen.png		
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2: 3 (25.6 3: 5 (13.1			
Predicting			
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2: 3 (4.93 3: 5 (1.0%			
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	ຈາ 7shen.png		
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2: 3 (0.25 3: 9 (0.17			
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	5daniel.png		
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1: 5 (97.3			
2: 8 (2.27 3: 9 (0.31			
3. 3 (0.31	-0,-		

## Some numbers are hard for our model to predict







# Live Demo & Showcase

### TIMELINE

### 4 weeks

**Week 1**: Fine-tune our model to be able to identify problematic numbers

Week 2: Be able to scan and identify a sudoku board

Week 3: Pinpoint the position of numbers & solve the sudoku problem

Week 4: Create a mobile demo and web server to serve results

### Goals

### Solve a Sudoku Problem

**Goal 1**: We want to be able to take an image of a Sudoku board and solve it

Goal 2: We want to have a high accuracy of identifying handwritten numbers (90%?)

### Optimistic Goals

### Mobile Interface

If previous goal is achieved we want to expand into a simple mobile app to take a picture, upload image to our web server and produce the solution for the user

# "I love sudoku solver, it helped me bring joy to my family and friends"

Future Sudoku Solver User, North Korea