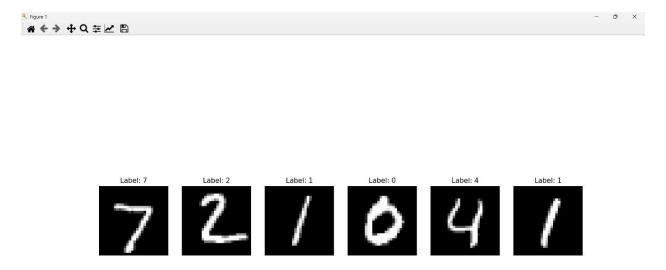
# PATTERN RECOGNITION & COMPUTER VISION Project 5: Recognition using Deep Networks BASIL REJI & KEVIN SANI

#### **Description:**

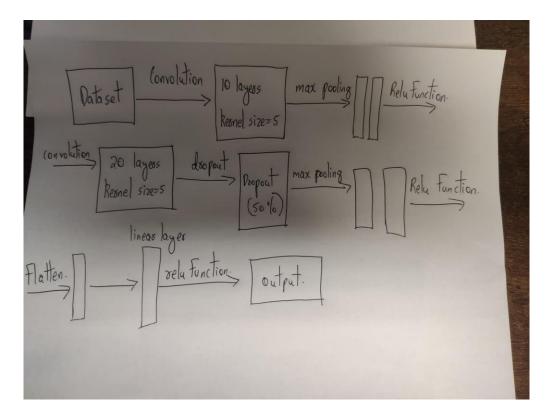
The project entails designing, implementing, training, and experimenting on a neural network to be fit for image recognition. First, the MNIST digits dataset was used. This is the chosen dataset, as it is well known for balancing simplicity and the level of challenge while designing and training a convolution neural network (CNN). The architecture of the model was initially designed; it kept in mind convolutional layers, pooling, and dropout for effective recognition of handwritten digits. To generalize this exploration further, that very network was adapted to recognize Greek letters. This was done by modifying the final layer to classify three different classes and retraining with a custom dataset. Afterwards, we varied the architecture of the network systematically along certain dimensions: the number of convolutional filters, the sizes of the kernels, and the rate of the dropouts. This research, in that respect, sought to run a set of experiments with a variety of configurations to optimize the network's performance and efficiency of the MNIST Fashion dataset. Very useful in grasping the behavior of the network in all setups, it clearly enabled us to understand how some architectural decisions affect both learning effectiveness and computational demands.

### Task 1:Build and train a network to recognize digits

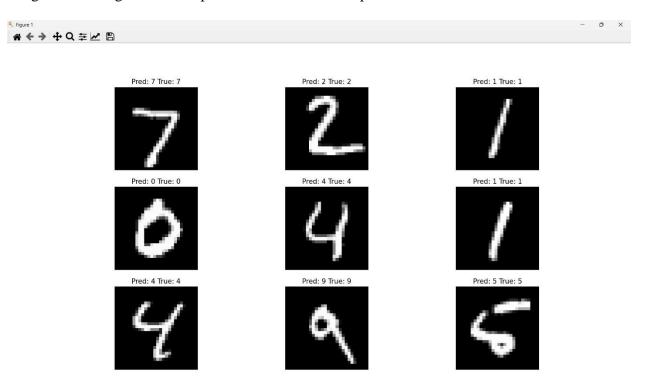
First six example digits from the test set

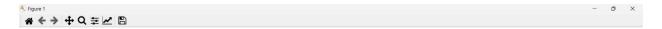


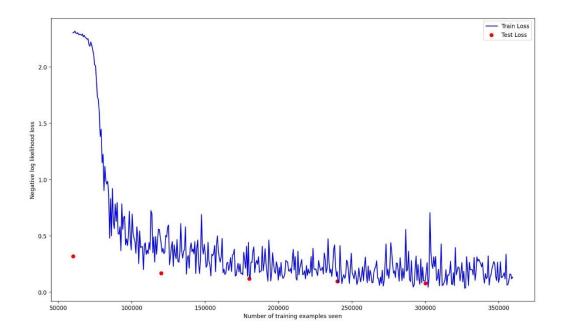
# Network Model



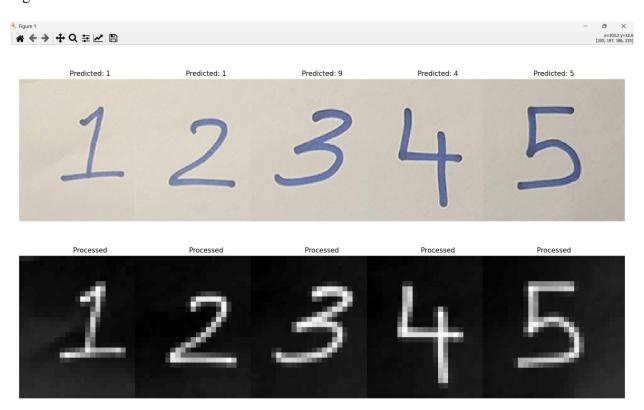
9 digits as a 3x3 grid with the prediction for each example above it





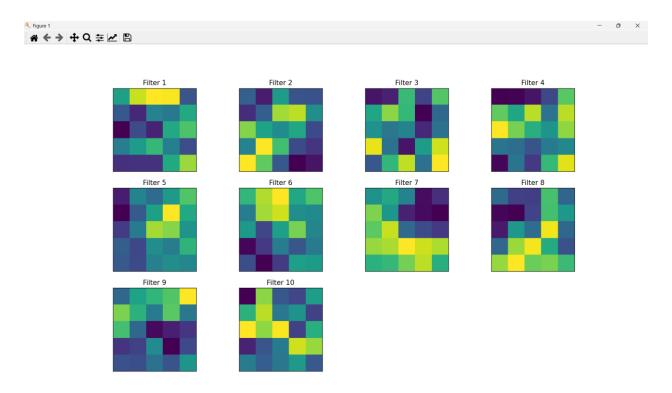


# digits and their classified result for written dataset

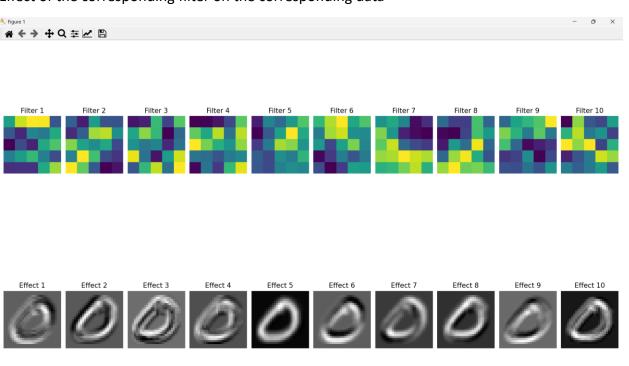


# Task 2: Examine your network

Visualization of the filters



Effect of the corresponding filter on the corresponding data



# Task 3: Transfer Learning on Greek Letters

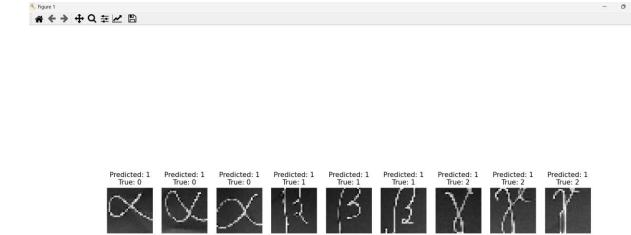
Training loss over the new network and the Greek Letter Dataset



Number of training examples seen

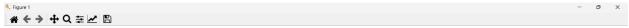
# Specifications of the new Network Model

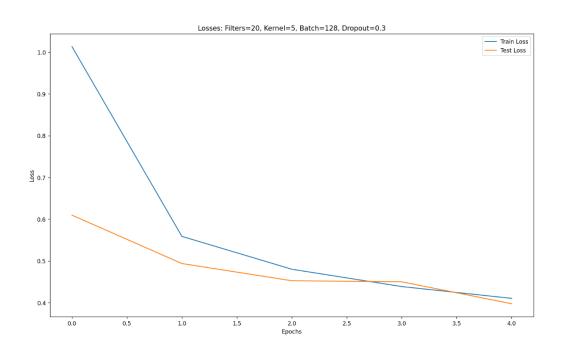


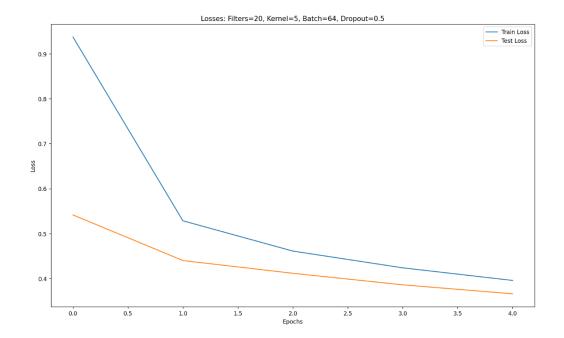


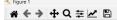
# Task 4: Design your own experiment

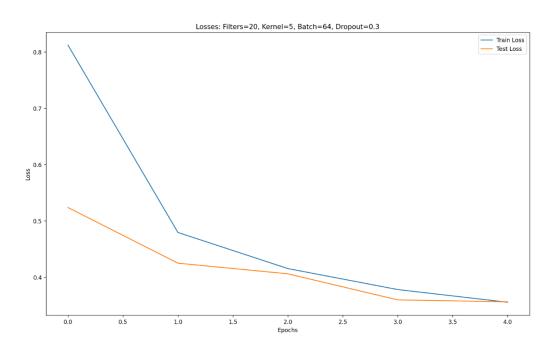
<u>Collection of images of training and testing error with different parameters run on the custom network the corresponding parameters used are displayed as titles for the images</u>

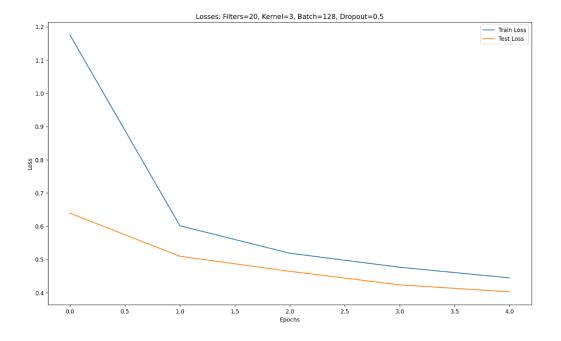


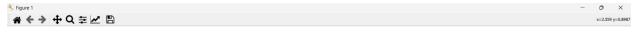


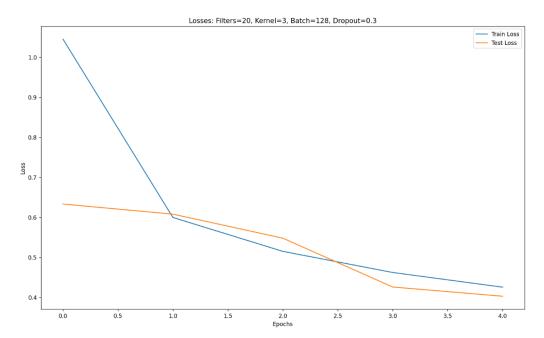


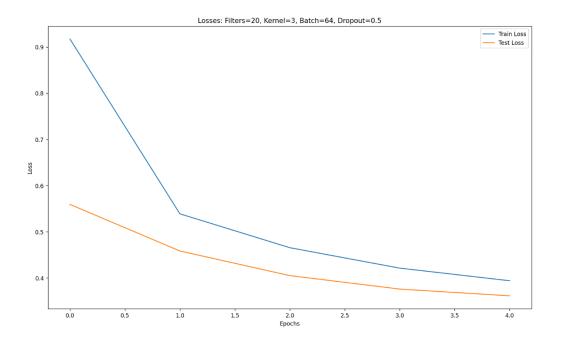


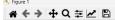


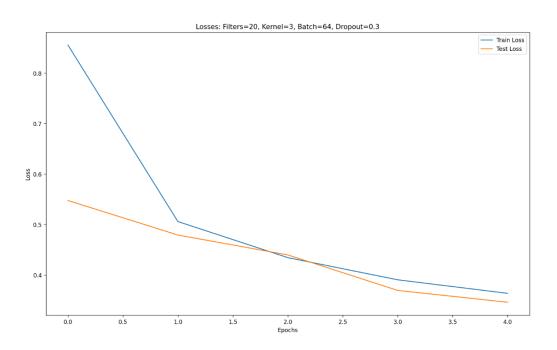


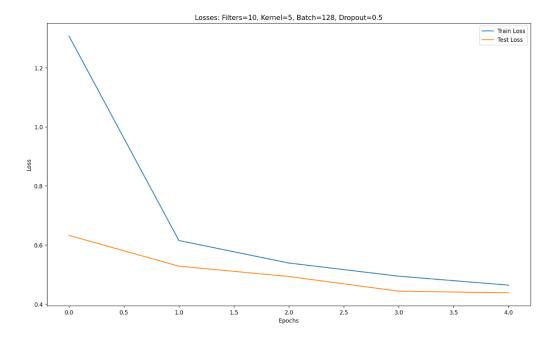


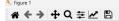




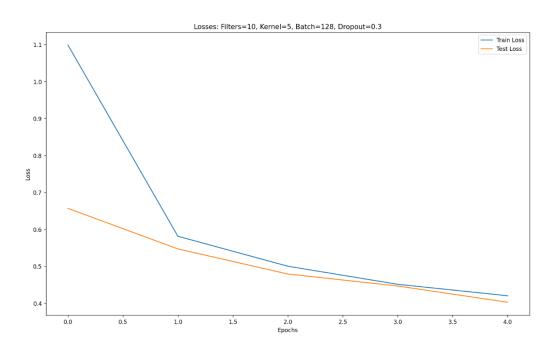


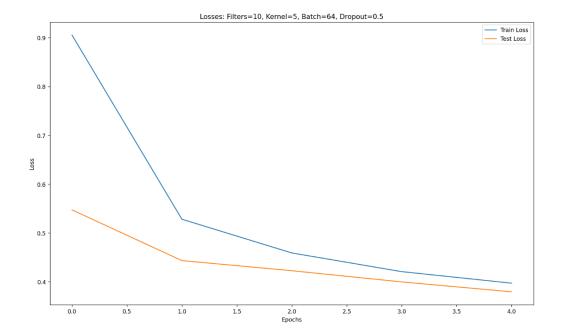


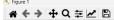


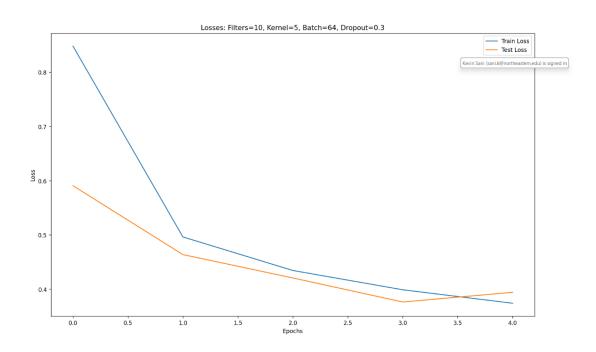


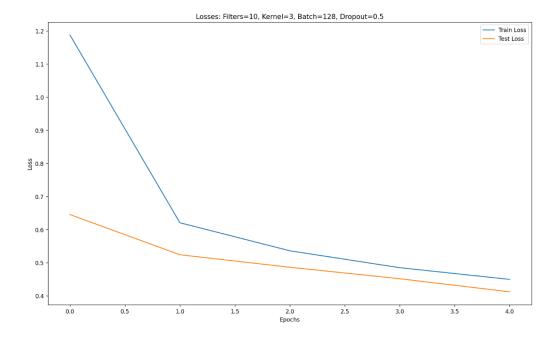




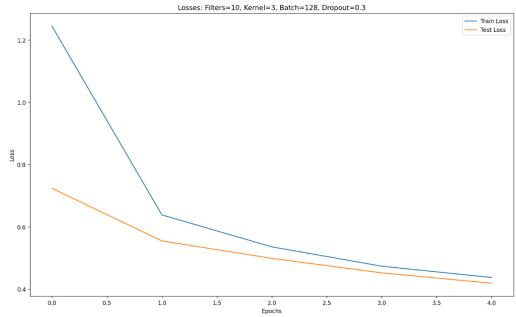


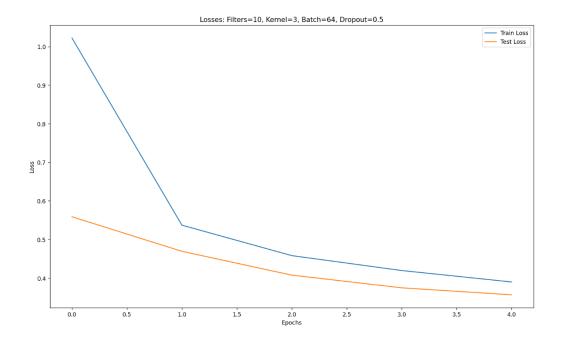


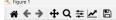


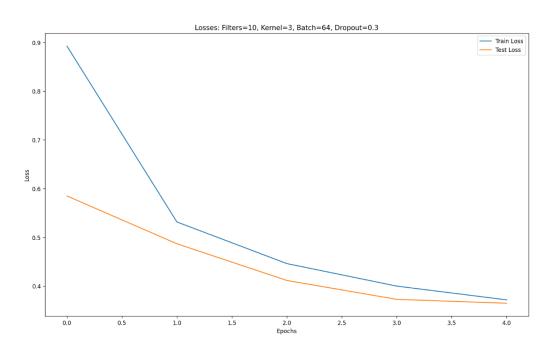


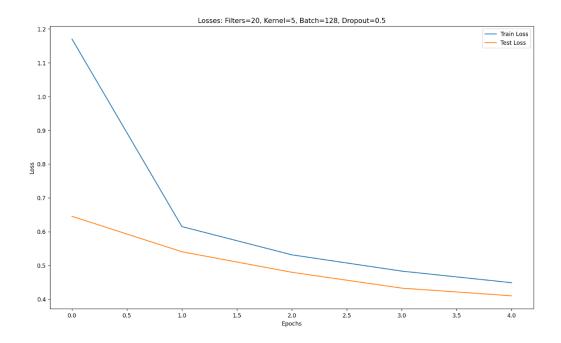




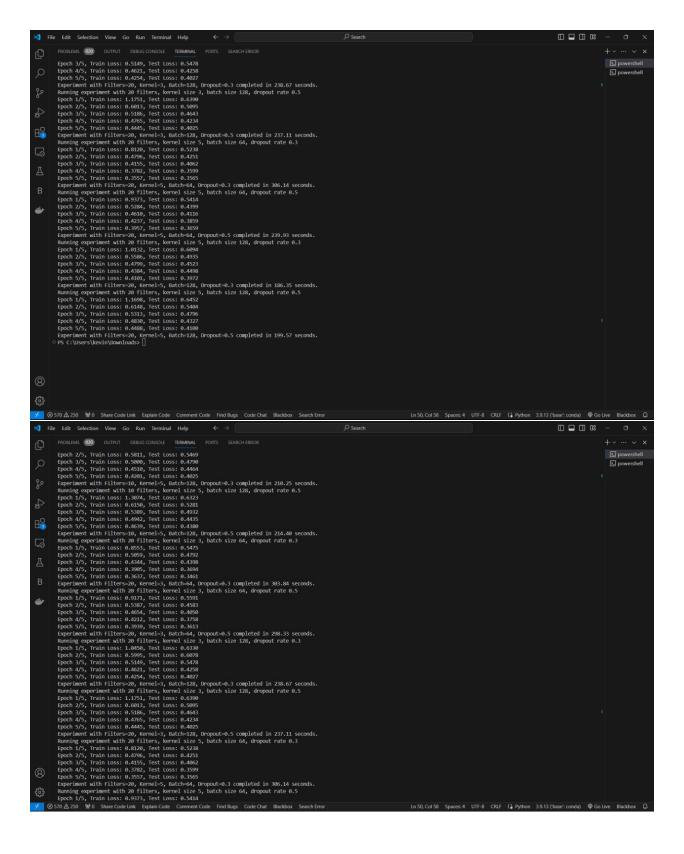








The corresponding time taken for the network to complete its execution the order format is the same as the order of the images



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#### **Reflection:**

The project gave me an opportunity to delve into the specifics of how to build and optimize convolutional neural networks (CNNs) for image recognition exercises with the MNIST and Fashion MNIST datasets. I then took the digit recognition model from the foundational model I had built earlier and adapted it for a completely different, more complex task: recognizing handwritten Greek letters. It has been journeying through important aspects of designing a neural network, e.g., how to choose and adjust layers, filters, and other hyperparameters so that, in turn, they can increase model accuracy and efficiency. Experimentation over several such configurations of networks shed light on the fine balance between model complexity and computational demand, hence elaborating on the significance of thoughtful architecture design. I have automated lots of systematic experiments with the network that evaluates many possible architectures and have learned much more about architectural choices and their effect on performance than if I had to do all of that by hand. This project further polished my knowledge of the principles and practices of computer vision and how important the pytorch library is important in both python and computer vision.

#### **Acknowledgements & References:**

- Computer Vision: Algorithms and Applications, 2nd ed.
- Pytorch Documentation
- Object Detection Made Easy: Nicolai Nielson, YouTube Channel