

# CREDIT CARD DIGIT SEGMENTATION

Presented By

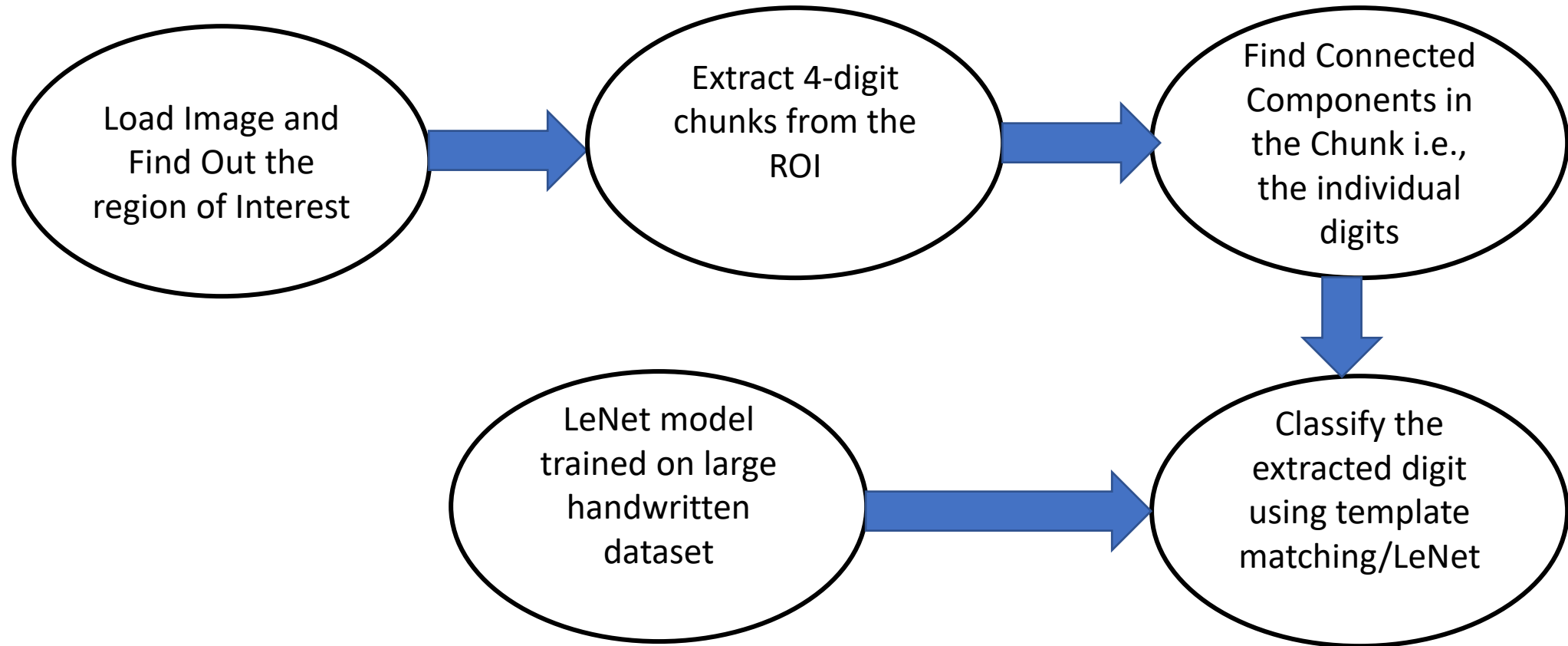
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# Introduction

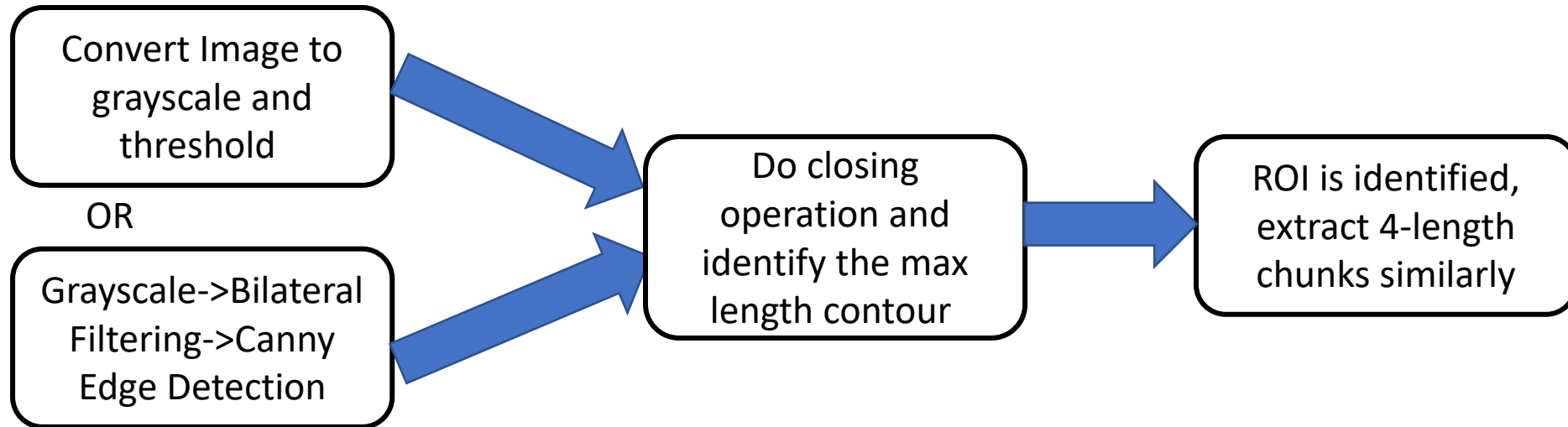
- The aim of our project is to segment and recognize each of 16 digits in credit/debit cards.
- From these detected digits, the card's company i.e.- Visa, Mastercard, or American Express and many other useful information can also be determined.
- In this project, we have created an image processing pipeline to locate the 16 digits in the card.
- Various algorithms like thresholding, morphological operations, and contour detection have been used for the purpose.
- Finally, the digits have been identified using a trained CNN classifier.

# Plan of Action



# Preprocessing

Identifying the 4-length chunks in the image:



# Finding Out the Region of Interest(ROI)

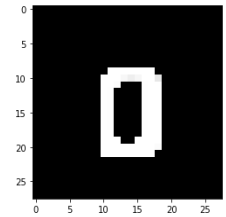
- We tried to first identify the 16 length ROI automatically by drawing a bounding box around the largest length contour as follows:



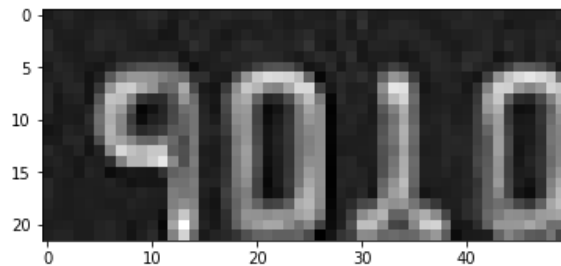
- But we encountered many problems in general such as spurious long contours, contours wrongly connected due to closing, etc.
- Rather than handling each case separately we decided to resize the image to a fix size  $\sim(179 \times 280)$  and assume the approximate location of the ROI (16-chunk)
- After the ROI is found, we tried to find the 4 length chunks as bilateral filtering  $\rightarrow$  canny  $\rightarrow$  contours

# Extracting each digit and classification

- So we extracted the ROI template:
- We centred each of the digits identified by connected components from the 4-length chunks, to provide some background info.
- We then tried template matching and LeNet to classify the digits accordingly.



A 4-digit chunk highlighting poor quality after extraction:

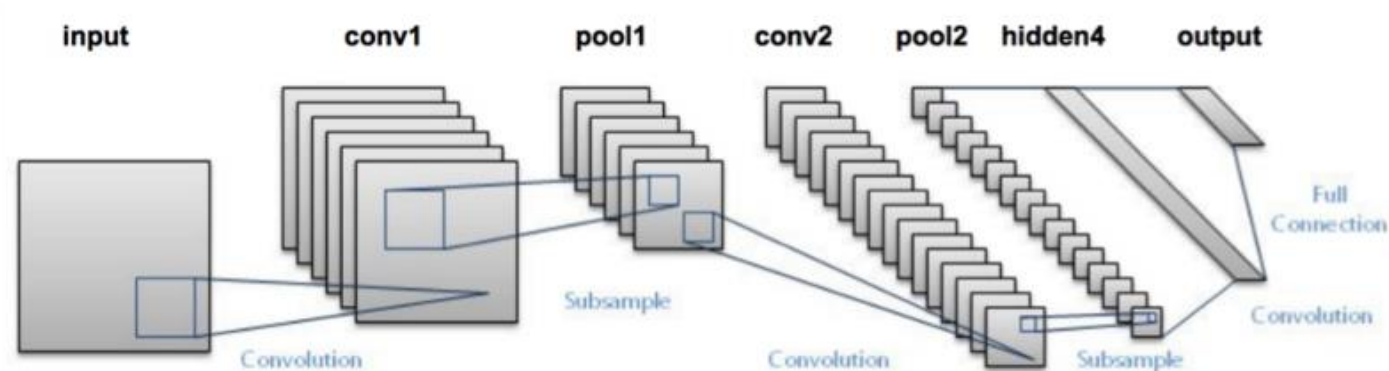


# Classification using Template matching

- First, we used standard OCR-digit images templates to match with our extracted digits. But as the extracted digit had a lot of non-uniformity (because it is drawn from a small part of the large credit card image) we were not getting satisfactory results.
  - We then extracted templates themselves from the credit cards but faced the following problems:
    - Each digit had different sizes across various images
    - Each extracted digit varied across images due to the closing/dilation operations which were used to coalesce broken parts of same digit
- So we moved to adopting LeNet based classification of the digits.

# LeNet-5 Architecture

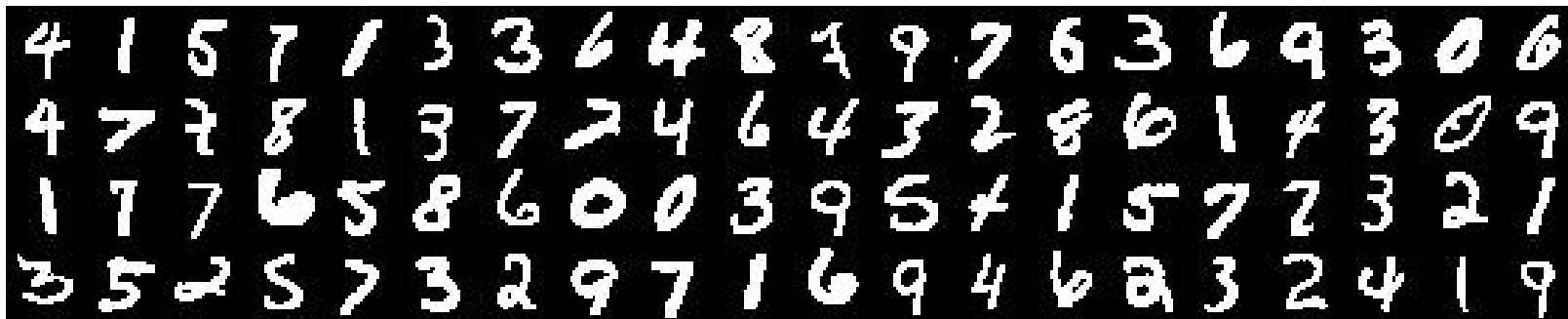
- LeNet-5 is a simple convolutional neural network (CNN) architecture.
- It consists of two sets of convolutional, activation, and pooling layers, followed by two fully-connected layers, and finally a soft max classifier.
- Size of the input is 28x28 and convolution kernels used are of size 5x5.
- Activation function used is 'tanh' activation function.
- Average pooling with kernel of size 2x2 is used.





# MNIST dataset and Training

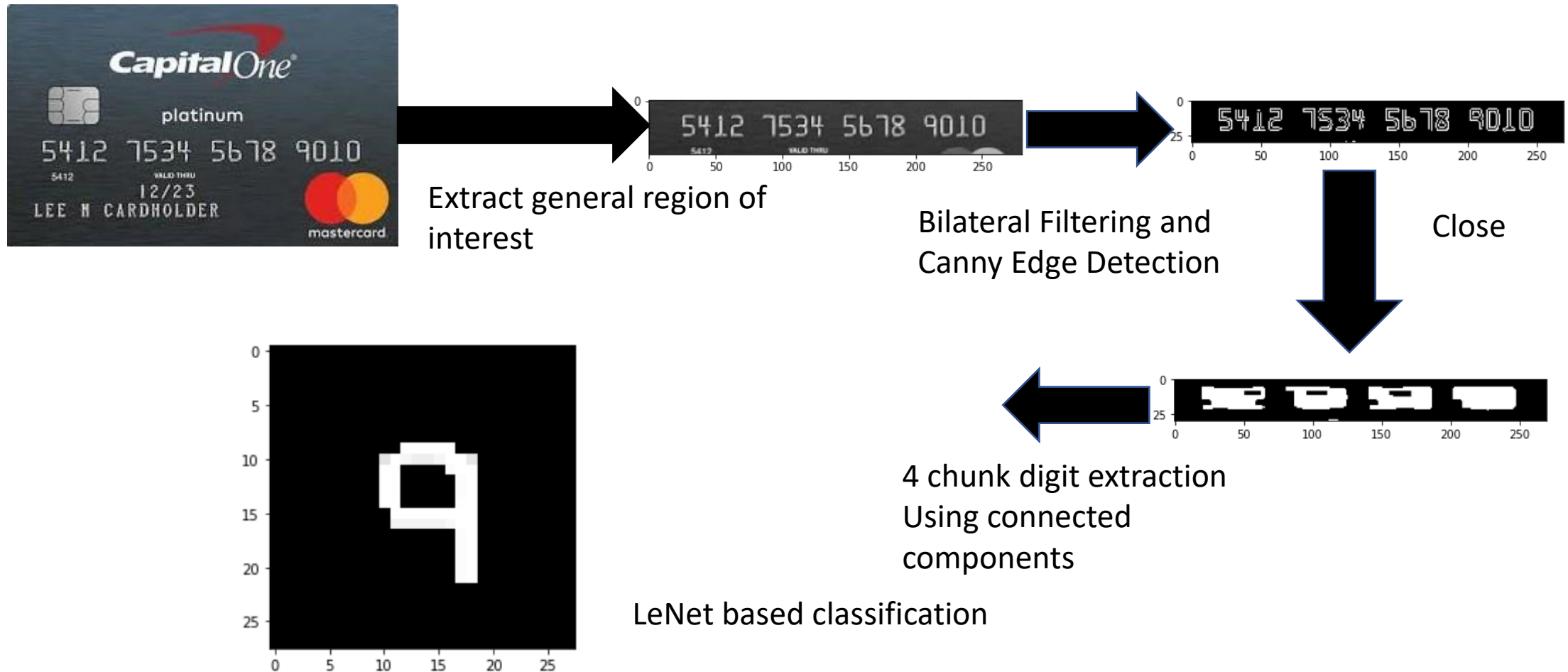
- MNIST dataset is a large dataset containing images of handwritten digits 0-9.
- There are a total of 70,000 images in the dataset. 60,000 images for training and 10,000 images for validation purpose.
- Each sample is a 28x28 grayscale image of one of 10 digits.
- This dataset has been used to train the LeNet model which will classify each of the 10 digits.



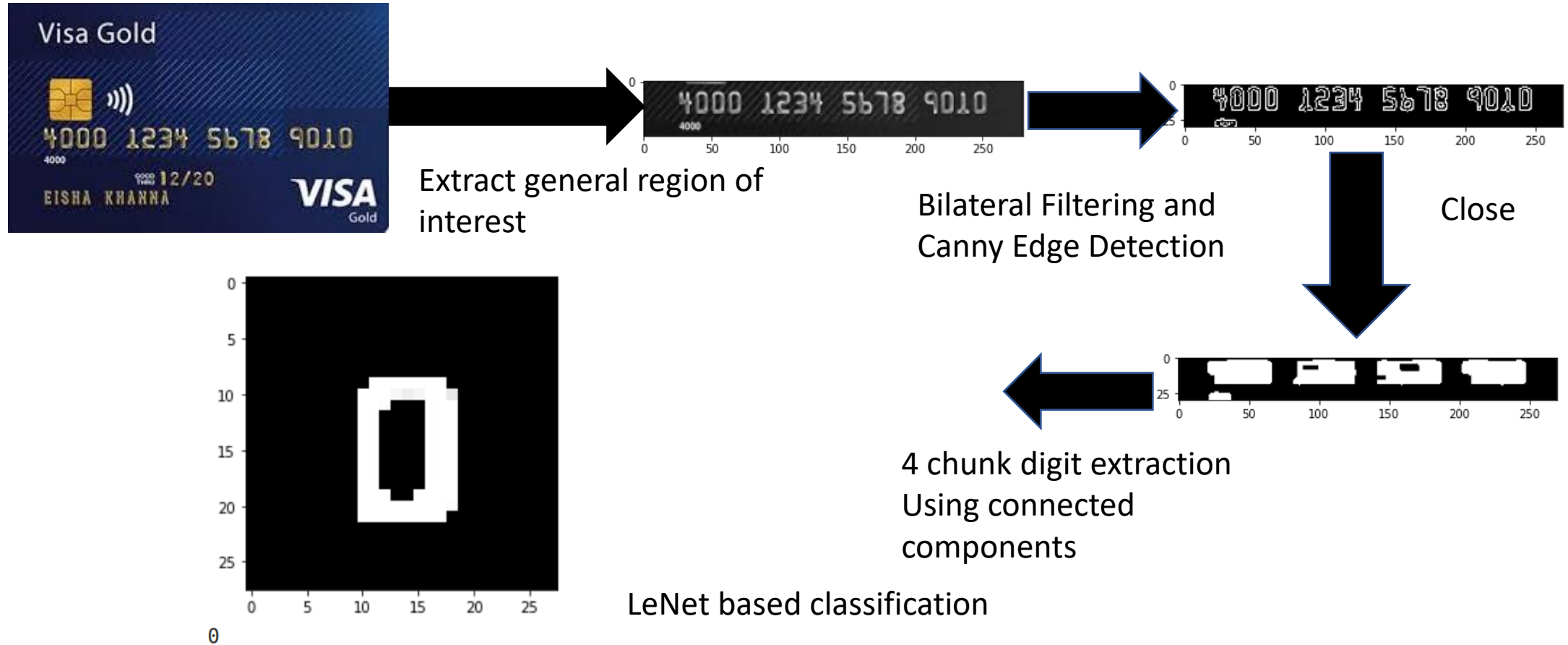
# Classification using CNN

- We obtained better (but not 100% accurate results) after training with the LeNet model.
- The challenges we faced are:
  - OCR font is a bit different from handwritten digits in general
  - To do transfer learning/fine-tuning we should have large number of samples of 28x28 size (standard LeNet size)
  - Also after resizing images we found sort of aliasing effect (some blurred boundaries on the periphery of the digits ,etc.)

# Result



# Result



# Conclusion and Comments

- The methods work best when background is smooth and has no illumination problems.
- The LeNet-5 based classification is not accurate which was unexpected as it was trained on a large set of handwritten images and not OCR-A fonts.

THANK YOU