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Deep Learning Project
on
Street View House Number Recognition

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Problem Statement

- Recognition of street-level house numbers is a critical task for urban mapping, navigation systems, and infrastructure assessment.
- In this project, we aim to develop a deep learning model that can accurately recognize multi-digit house numbers from real-world images using the Street View House Numbers (SVHN) dataset.

Goal

- The goal is to automate visual number recognition in complex environments.

Dataset Description

- The **SVHN dataset** is a real-world image dataset for developing models that recognize digits from street view imagery.
- Images are in **.png** format with variable resolution and real-world distortions.
- Each digit in the image is labeled and has a bounding box stored in a **.mat** file.

| File | Description | No of Digits |
|------------------|-------------------------|--------------|
| train.zip | Labeled training Digits | 73,257 |
| test.zip | Labeled test Digits | 26,032 |

Dataset Description(Cont.)



Fig 1 : Sample Images

Labels

- **10 classes** for digits: '1' → 1, '9' → 9, '0' → 10 (not 0).
- Each image may contain **multiple digits**.

Annotation Format

- **digitStruct.mat**: Contains bounding box info per digit:
 - **name**: Image filename
 - **bbox**: Array of structs → height, width, label, top, left

Preprocessing

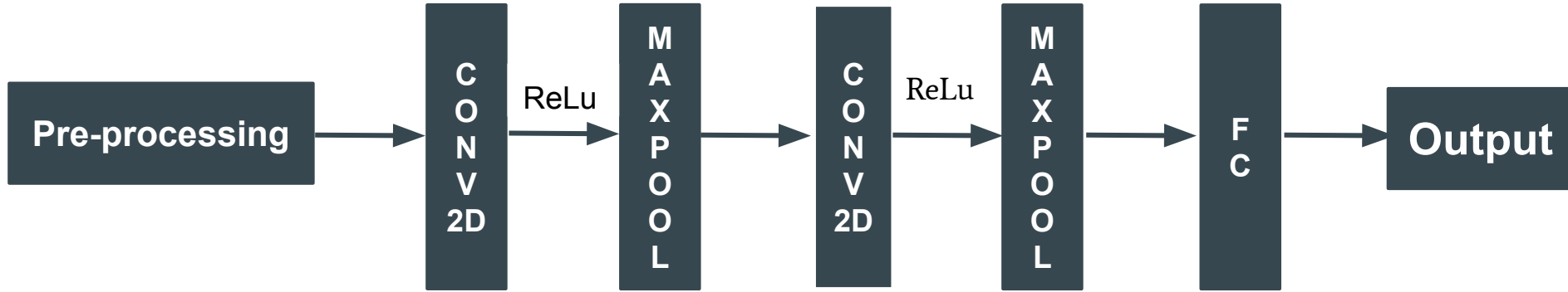
1. Extracted Data from .tar.gz archives (train/test).
2. Parsed digitStruct.mat for filenames, bounding boxes, and digit labels.
3. Converted labels ($10 \rightarrow 0$) and created structured CSV files.
4. Filled missing digits with -1 for consistency.

Visualizations

- Digit length distribution.
- Digit frequency.

Output: Clean CSVs ready for model training.

CNN Architectures

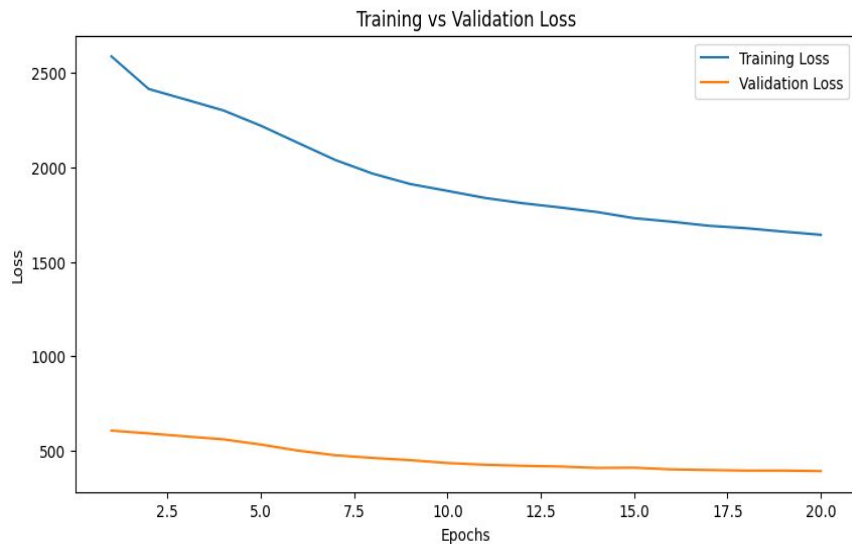


Architecture 1 - Custom Architecture



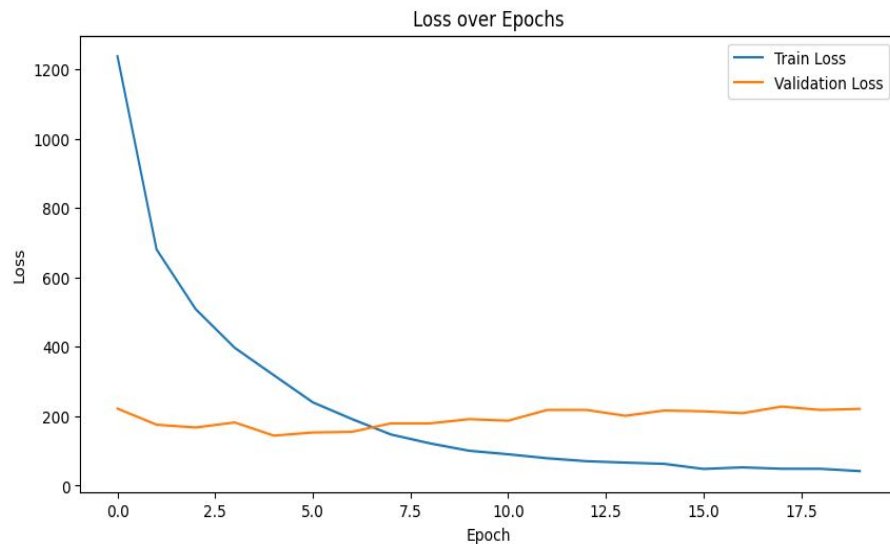
Architecture 2 - Modified Pretrained Architecture

Results



Custom Architecture's
Epoch vs Loss Curve

Test Accuracy : 72.36%



ResNet Architecture's
Epoch vs Loss Curve

Test Accuracy : 86.33%