

# Project 1 Report: CIFAR-10 Classification with Custom ResNet

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## Abstract

This project addresses the CIFAR-10 image classification challenge using a custom ResNet architecture under the 5-million-parameter constraint. We propose a [brief description of approach], achieving [X

## 1 Introduction

The CIFAR-10 dataset is a standard benchmark for image classification, consisting of 60K images spread across 10 classes. We aim to design a custom ResNet architecture with fewer than 5 million parameters, trained from scratch to achieve high accuracy on this dataset.

## 2 Methodology

### 2.1 Data Preprocessing and Augmentation

We use standard data normalization, random cropping, and horizontal flipping to increase effective training data size.

### 2.2 Model Architecture

Our proposed network modifies the standard ResNet block to reduce parameter count while maintaining representational power. The final layer configuration is shown in Table 1.

### 2.3 Training Configuration

We train using stochastic gradient descent (SGD) with momentum, a batch size of 128, and an initial learning rate of 0.1 with a cosine annealing scheduler.

Table 1: Key Architecture Details and Parameter Counts

Layer	Output Size	# Params
Conv1	$32 \times 32$	XXX
ResBlock1	$16 \times 16$	XXX
ResBlock2	$8 \times 8$	XXX
...	...	...
FC	10	XXX

### 3 Experiments and Results

#### 3.1 Ablation Studies (Optional)

### 4 Discussion

Our experiments show that data augmentation and a well-tuned learning rate schedule are critical to achieving high accuracy within the parameter limit. We found that deeper networks risk exceeding 5 million parameters, while shallower networks underfit.

### 5 Conclusion

We demonstrated a custom ResNet for CIFAR-10 that respects the 5-million-parameter cap while maintaining competitive accuracy. Future work could explore novel regularization strategies or advanced optimization techniques.

### 6 References