SPEECH LAB EVAL

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Paper Summary:

The paper "Speech Commands: A Dataset for Fine-Grained Speech Recognition" presents a dataset with 65,000 one-second recordings of 12 spoken commands, such as "yes," "no," "up," and "down." It aims to advance research in speech recognition by offering a standardized benchmark for fine-grained audio classification, focusing on real-world performance in command recognition.

Google Collab files link:

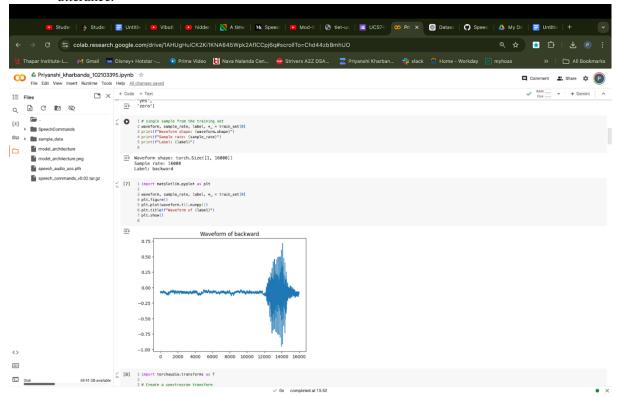
https://colab.research.google.com/drive/1dcbddNXmg2VLQ_4t2HFQZc86bKP_z6cC

https://colab.research.google.com/drive/1k2v-4TAlplImoT-EbDsSzEr4eQBCU640#scrollTo=SdDDteRj07ft

 $\underline{https://colab.research.google.com/drive/1auuc3uhHo1DOKkCUVxBDmp3nud7o6dVU?usp=sharing}$

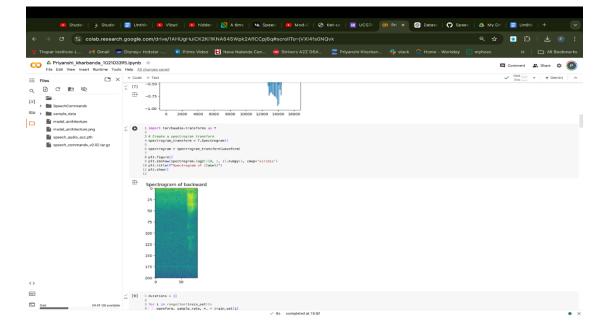
Analysis of data

1. A data point in the SPEECHCOMMANDS dataset is a tuple made of a waveform (the audio signal), the sample rate, the utterance (label), the ID of the speaker, the number of the utterance.



Number of training samples: 84843 Number of testing samples: 11005

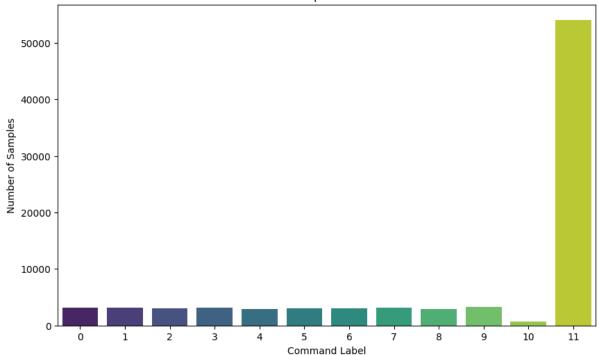
2. After applying fourier transformation, here is what a spectogram of label backward looks like

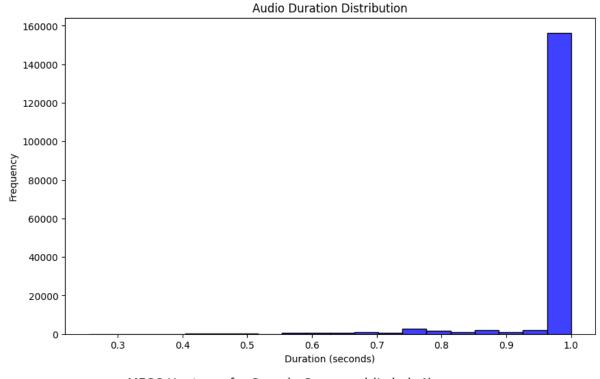


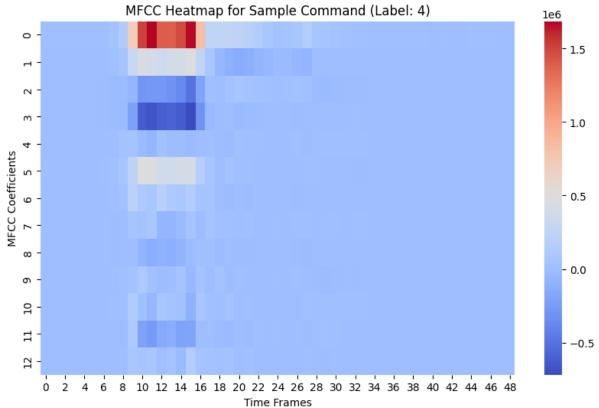
3. Analysis done of duration of samples in our speechcommand dataset.

Some snippets of analysis:









Steps considered while setting up data loading and batching for training and testing a model using the SPEECHCOMMANDS dataset with PyTorch.

1. Padding Sequences (pad_sequence function)

- **Purpose**: Ensures all tensors in a batch are the same length by padding with zeros.
- Details:
 - Transposes each tensor in the batch.
 - o Uses torch.nn.utils.rnn.pad_sequence to pad tensors to the same length.
 - Permutes the dimensions of the padded tensor to match the expected input shape for your model (usually (batch_size, channels, sequence_length)).

2. Collate Function (collate_fn function)

- **Purpose**: Defines how to collate (combine) individual data samples into a batch.
- Details:
 - o Extracts waveforms and labels from each sample in the batch.
 - o Applies pad_sequence to ensure waveforms in the batch have the same length.
 - o Converts labels to indices using label_to_index and stacks them into a tensor.

3. DataLoader Configuration

- Purpose: Creates DataLoader instances for the training and test datasets.
- Details:
 - train_loader: Loads training data with batching, shuffling, and custom collate function.
 - batch size=256
 - shuffle=True: Shuffles data at each epoch.
 - num_workers and pin_memory are set based on whether a GPU (cuda) or CPU is used.
 - o **test_loader**: Loads test data with batching and custom collate function.
 - batch_size=256
 - shuffle=False: No shuffling for test data.
 - drop_last=False: Keeps the last batch even if it's smaller than the batch size.

Model Used (M5 Class)

Purpose: Defines a CNN model for classification.

Components:

- Convolutional Layers:
 - o conv1: 1D convolution with kernel size 80, stride 16.
 - o conv2: 1D convolution with kernel size 3.
 - o conv3: 1D convolution with kernel size 3, output channels doubled.
 - o conv4: 1D convolution with kernel size 3, output channels doubled.
- Batch Normalization:
 - o bn1, bn2, bn3, bn4: Normalize outputs of respective convolutional layers.
- Pooling Layers:
 - o pool1, pool2, pool3, pool4: Max pooling with kernel size 4.
- Fully Connected Layer:
 - o fc1: Linear layer that maps from the output channels to the number of classes.

• Forward Pass:

- Applies convolutions, batch normalization, ReLU activation, pooling, and average pooling.
- Reshapes and passes through a fully connected layer followed by log_softmax for classification.

