



Model Optimization and Tuning Phase Template

Date	15 March 2024
Team ID	739868
Project Title	Real Time Communication System Powered By AI For Specially Abled
Maximum Marks	10 Marks

Model Optimization and Tuning Phase

The Model Optimization and Tuning Phase involves refining neural network models for peak performance. It includes optimized model code, fine-tuning hyperparameters, comparing performance metrics, and justifying the final model selection for enhanced predictive accuracy and efficiency.

Hyperparameter Tuning Documentation (8 Marks):

Model	Tuned Hyperparameters
CNN	 Batch Size: The number of samples processed before updating the model's weights. Larger batch sizes accelerate training but require more memory, while smaller batches are more memory-efficient. Example: batch_size=32 Epochs: The number of times the entire training dataset is passed through the model. More epochs improve learning but may lead to overfitting. Example: epochs=10 Learning Rate: Determines the step size during optimization. A smaller rate slows down the training process, while a larger rate might cause the model to miss optimal solutions. Example: learning_rate=0.001 Optimizer: Determines how the model's weights are updated based on the loss function. Common optimizers include Adam (adaptive learning rate) and SGD (stochastic gradient descent). Example: optimizer=Adam() Loss Function: Measures how well the model's predictions match the true values. Categorical cross-entropy is widely used for multi-class classification.





```
Initialize the model

model = Sequential()

Add the convolution layer

model.add(Convolution2D(32,(3,3),input_shape=(64,64,1),activation = 'relu'))

Add the pooling layer

model.add(MaxPooling2D(pool_size=(2,2)))

Add the flatten layer

model.add(Flatten())

Adding the dense layers

model.add(Dense(units=512,activation='relu'))
 model.add(Dense(units=9,activation='softmax'))
```

Final Model Selection Justification (2 Marks):

Final Model	Reasoning
	The model processes images, extracts features using convolutional
	layers, and makes predictions through fully connected layers. It utilizes
	ReLU activation for hidden layers and Softmax for multi-class
	classification. The model is trained with categorical cross-entropy
	loss and the Adam optimizer . It's designed to recognize specific
	patterns or gestures in images, enabling real-time communication for
CNN	specially-abled individuals.