Kravspecifikation för del 2

- 1. W Börja använda basala MLOps best-practices:
 - A. Ordna så att du har ett bra sätt att övervaka träningen, samt att spara (och hålla ordning på!) versioner av modellerna du tränar upp. (Tips: Foldrar, smart valda filnamn, etc).
 - B. H Kom ihåg att det troligen inte är den sista epokens checkpoint (sparad version) som är den bästa, p.g.a. overfitting om du kör många epochs. Du behöver själv se till att spara checkpoints lite då och då under träningen.
 - C. Versionshantering av parametrar, kod, och checkpoints (modellens parametrar) blir snabbt väldigt viktigt när man börjar jobba seriöst med machine learning. Om man inte håller reda på vilken körning som gjorts med vilka parametrar, så blir det svårt att jämföra prestanda, vidareutveckla modellen senare, etc. Ingen vill ha en modell levererad där programmeraren inte minns hur den gjordes.
 - D. **Fördjupning/överkurs:** Vilka verktyg finns som kan hjälpa till att hålla ordning på versioner, körningar, resultat, hyper parameters, etc?
- 2. Fördjupning/överkurs: Addera performance metrics, så att du kan se hur lång tid respektive tränings-körning tar. (En körning = Alla epochs av träning för en given modell med specificerad uppsättning hyper parameters).
- 3. Applicera lämpliga data augmentation methods för att artificiellt variera, och till och med "förstora" datamängden lite. Använd exempelvis skalning, rotation, färgvariation, brus, (spegelvändning?), etc. Här är lite repetition om data augmentation:
 - A. Se avsnittet "Uppgift 1 Perceptron för OCR" längre ner i denna uppgift (del 3).
 - B. Bra 10-minuters video om ämnet: Pytorch-Data-Augmentation-using-Torchvision
 - C. Observera att det inte är säkert att data augmentation leder till bättre resultat i alla lägen, eftersom det beror på detaljerna i vad man bygger, och vilken data man har.
- 4. Det är nu dags att testa CNN istället för FFN. Byt ut de första lagren i modellen till convolutional layers för att ta upp translationsinvarians i bilden. Sök själv upp vad man brukar ha direkt efter varje convolutional layer för att hålla ordning på antalet dimensioner till nästa lager.
 - (Svårt ord "translations-invarians": translation=förflyttning, invarians="samma oavsett" ⇒ translationsinvarians = "spelar ingen roll var i bilden")
- 5. Experimentera därefter med att addera ett till (eller flera) convolutional layer(s). Tanken är att testa **olika nät-arkitekturer**. Använd Google/Chat för initial gissning för modell-arkitektur, men sedan är det trial-and-error som gäller. Att fundera på:

A. Thur påverkas resultatet av olika modell-arkitektur? Var noga med att inte jämföra modeller med stor skillnad i antal parameters, om aspekter som tex lagertyp ska utvärderas.

- B. Silvilken typ av features detekteras typiskt av de senare convolutional-lagren jämfört med det första convolutional-lagret?
 - a. Hint: Videon från lektion om convolutional layers: Convolutional-Neural-Networks-Explained. (Självfallet kan du också plotta weight-matrices från din egen modell också om du vill!)
- 6. Applicera lämpliga **regularization methods** för att minimera risken för overfitting och göra cost-function-landskapet fördelaktigt för back-prop.
 - A. Exempel på sådana metoder är: drop-out, weight-decay, noise injection, batch normalization, etc.
 - B. Mer info om regularization: understanding-regularization-with-pytorch
- 7. **Fördjupning:** Skapa en lista av kombinationer av hyper parameters, och låt datorn träna din modell på nytt för samtliga hyperparameter settings i din lista. Detta kallas för **hyper-parameter tuning**, d.v.s. att via trial-and-error hitta inställningar som funkar.
 - Enklast möjliga hyper-param-tuning: Gör en lista med t.ex. 10 rader (en per körning), där varje rad specificerar alla hyper-parameters som kan vara svårt att gissa optimalt värde på:
 - Antal neuroner f\u00f6r respektive lager
 - Typ av activation function
 - Vilka lager som ska vara convolutional, storlek på convolutional kernel, antal convolutional kernels per lager, etc.
 - Learning rate / hyper-params till optimizer ADAM
 - Drop-out level
 - Settings för olika typer av data augmentation (rotation, skalning, noise, etc)
 - Det finns även en mängd automatiserade typer av hyper-param-tuning. Se exempelvis: https://en.wikipedia.org/wiki/Hyperparameter_optimization

```
import os
import time
import logging
import random
import torch
import torch.nn as nn
import torch.optim as optim
from datetime import datetime
from torch.utils.data import DataLoader, random_split
from torchvision import datasets, transforms

# Define the perceptron neural-network model
class Perceptron(nn.Module):
    # Define the constructor
    def __init__(self, dropout_rate=0.25):
```

```
super().__init__()
        self.dropout rate = dropout rate
        # Convolutional layers
        self.conv layers = nn.Sequential(
            nn.Conv2d(1, 32, kernel size=3, padding=1), # 32x32 with 1 paddi
            nn.BatchNorm2d(32), # Add batch normalization after convolution
            nn.ReLU(), # Apply ReLU activation function
            nn.MaxPool2d(kernel size=2), # 14x14 output dimensions
            nn.Dropout2d(dropout rate) # Add dropout to convolutional featur
        )
        # Fully connected layers
        self.fc layers = nn.Sequential(
            nn.Linear(32 * 14 * 14, 128), # 128 neurons in the hidden layer
            nn.BatchNorm1d(128), # Add batch normalization after linear layε
            nn.ReLU(), # Apply ReLU activation function
            nn.Dropout(dropout rate), # Add dropout to hidden layer
            nn.Linear(128, 10) # 10 neurons in the output layer
        )
   # Define the forward pass
   def forward(self, x):
       # Pass through the convolutional layers
       logits = self.conv layers(x)
       # Flatten the output
       logits = logits.view(logits.size(0), -1)
        # Pass through the fully connected layers
       logits = self.fc layers(logits)
        return logits
def train model(model, training dataset, testing dataset, hyperparameters, m
   # Extract hyperparameters
   num epochs = hyperparameters['num epochs']
   learning rate = hyperparameters['learning rate']
   batch size = hyperparameters['batch size']
   l1 lambda = hyperparameters['l1 lambda']
   12 lambda = hyperparameters['l2 lambda']
    dropout rate = hyperparameters['dropout rate']
   early stopping patience = hyperparameters['early stopping patience']
   criterion = nn.CrossEntropyLoss()
   optimizer = optim.Adam(model.parameters(), lr=learning rate, weight deca
   # Split training data into train and validation subsets
   training_subset_size = int(0.8 * len(training dataset))
   validation subset size = len(training dataset) - training subset size
   training subset, validation subset = random split(training dataset, [tra
   # Create DataLoaders
   train loader = DataLoader(training subset, batch size=batch size, shuffl
   validation loader = DataLoader(validation subset, batch size=batch size,
   testing_loader = DataLoader(testing_dataset, batch size=batch size, shuf
```

```
# Create model directory
model dir = os.path.join('models', f'run {run id}', f'model {model id}')
os.makedirs(model dir, exist ok=True)
# Create checkpoints directory
checkpoints dir = os.path.join(model dir, 'checkpoints')
os.makedirs(checkpoints dir, exist ok=True)
# Set up logging
logging.basicConfig(level=logging.DEBUG, format='%(asctime)s - %(levelname)
logger = logging.getLogger()
log file = os.path.join(model dir, f'model {model id}.log')
fhandler = logging.FileHandler(filename=log file, mode='a')
formatter = logging.Formatter('%(asctime)s - %(name)s - %(levelname)s -
fhandler.setFormatter(formatter)
logger.addHandler(fhandler)
# Log hyperparameters
logger.info("=" * 100)
logger.info(f"Model ID: {model id}")
logger.info(f"Run ID: {run id}")
logger.info(f"Training configuration:")
logger.info(f" Learning rate: {learning rate}")
logger.info(f" Batch size: {batch size}")
logger.info(f" L1 regularization: {l1 lambda}")
logger.info(f" L2 regularization: {l2 lambda}")
logger.info(f" Dropout rate: {dropout rate}")
logger.info(f" Epochs: {num epochs}")
logger.info(f" Early stopping patience: {early stopping patience}")
logger.info(f" Optimizer: Adam")
logger.info(f" Loss function: CrossEntropyLoss")
# Training and validation loop
training start time = time.time()
best val loss = float('inf')
best model path = None
early stopping counter = 0
epochs without improvement = 0
for epoch in range(num epochs):
    # Cache the start time of the epoch
    epoch start time = time.time()
    # Training phase
    model.train()
    running train loss = 0.0
    for x, y in train loader:
        # Move data to device
        x, y = x.to(device), y.to(device)
        # Forward pass
        outputs = model(x)
        # Calculate base loss
        base_loss = criterion(outputs, y)
```

```
# L1 regularization (L2 is handled by weight decay)
    l1 reg = torch.tensor(0., device=device)
    for param in model.parameters():
        l1 reg += torch.norm(param, 1)
    # Total loss with L1 regularization
    loss = base loss + l1 lambda * l1 reg
    # Backward pass and optimization
    optimizer.zero grad()
    loss.backward()
    optimizer.step()
    # Update running loss
    running train loss += loss.item()
# Calculate average loss
avg train loss = running train loss / len(train loader)
# Print training loss
logger.info("="*100)
logger.info(f"Epoch [{epoch+1}/{num epochs}]")
logger.info(f"Training Loss: {avg train loss:.4f}")
# Validation phase
model.eval()
running val loss = 0.0
correct = 0
total = 0
with torch.no grad():
    for x, y in validation loader:
        # Move data to device
        x, y = x.to(device), y.to(device)
        # Forward pass
        outputs = model(x)
        loss = criterion(outputs, y)
        running val loss += loss.item()
        # Calculate accuracy
        _, predicted = torch.max(outputs, 1)
        total += y.size(0)
        correct += (predicted == y).sum().item()
# Calculate average loss and accuracy
avg val loss = running val loss / len(validation loader)
val accuracy = 100 * correct / total
# Print validation loss and accuracy
logger.info(f"Validation Loss: {avg val loss:.4f}")
logger.info(f"Validation Accuracy: {val accuracy:.2f}%")
# Save the checkpoint
checkpoint filename = f'epoch {epoch+1}.pth'
checkpoint path = os.path.join(checkpoints dir, checkpoint filename)
```

```
torch.save(model.state dict(), checkpoint path)
    # Update the best model if the current model has a lower validation
    if avg val loss < best val loss:</pre>
        best model path = checkpoint path # Cache the path to the best n
        best val loss = avg val loss
        epochs without improvement = 0
        logger.info(f"New best validation loss: {best val loss:.4f}")
    else:
        epochs without improvement += 1
        logger.info(f"No improvement for {epochs without improvement} er
    # Early stopping check
    if epochs without improvement >= early stopping patience:
        logger.info(f"Early stopping triggered after {epoch+1} epochs")
        break
    # Log the duration of the epoch
    epoch end time = time.time()
    epoch duration = epoch end time - epoch start time
    logger.info(f"Epoch {epoch+1} duration: {epoch duration:.2f} seconds
training end time = time.time()
training duration = training end time - training start time
logger.info(f"Training duration: {training duration:.2f} seconds")
if epochs without improvement >= early stopping patience:
    logger.info(f"Training stopped early at epoch {epoch+1}/{num epochs}
else:
    logger.info(f"Completed all {num epochs} epochs")
# Get the best model for testing
logger.info(f"Loading best model from: {best model path}")
model.load state dict(torch.load(best model path))
# Testing loop
testing start time = time.time()
model.eval()
running test loss = 0.0
correct = 0
total = 0
with torch.no grad():
    for x, y in testing loader:
        # Move data to device
        x, y = x.to(device), y.to(device)
        # Forward pass
        outputs = model(x)
        loss = criterion(outputs, y)
        running test loss += loss.item()
        # Calculate accuracy
        , predicted = torch.max(outputs, 1)
        total += y.size(0)
        correct += (predicted == y).sum().item()
# Calculate average loss and accuracy
```

```
avg test loss = running test loss / len(testing loader)
   test accuracy = 100 * correct / total
   # Log the results
   testing end time = time.time()
   testing duration = testing end time - testing start time
   total run time = testing end time - training start time
   logger.info("="*100)
   logger.info(f"Testing duration: {testing duration:.2f} seconds")
   logger.info(f"Total duration: {total run time:.2f} seconds")
   logger.info("="*100)
   logger.info(f"Best Model: {best model path}")
   logger.info(f"Best Validation Loss: {best val loss:.4f}")
   logger.info(f"Test Loss: {avg test loss:.4f}")
   logger.info(f"Test Accuracy: {test accuracy:.2f}%")
   logger.info("="*100)
   return test accuracy
if name == " main ":
   # Create a unique ID for the run
   run id = datetime.now().strftime("%Y%m%d %H%M%S")
   # Create directory structure
   os.makedirs('models', exist ok=True)
   run dir = os.path.join('models', f'run {run id}')
   os.makedirs(run dir, exist ok=True)
   # Set up logging
   logging.basicConfig(level=logging.DEBUG, format='%(asctime)s - %(levelne
   logger = logging.getLogger()
   log file = os.path.join(run dir, f"run {run id}.log")
   os.makedirs(os.path.dirname(log file), exist ok=True)
   fhandler = logging.FileHandler(filename=log file, mode='a')
   formatter = logging.Formatter('%(asctime)s - %(name)s - %(levelname)s -
   fhandler.setFormatter(formatter)
   logger.addHandler(fhandler)
   # Log start of experiment
   logger.info("="*100)
   logger.info(f"Starting new experiment with run ID: {run id}")
   logger.info("="*100)
   # Select device to run on
   device = torch.accelerator.current accelerator().type if torch.accelerat
   logger.info(f"Using device: {device}")
   # Transformations for training data with data augmentation
   training transform = transforms.Compose([
       transforms.RandomRotation(10), # Rotate by up to 10 degrees
       transforms.RandomAffine(0, scale=(0.8, 1.2)), # Random scaling
       transforms.ToTensor(), # Convert to tensor
       transforms.Normalize((0.1307,), (0.3081,)), # Normalize with MNIST
   ])
```

```
# Load the MNIST dataset
training dataset = datasets.MNIST(root='./data', train=True, download=Tr
testing dataset = datasets.MNIST(root='./data', train=False, download=Tr
logger.info(f"Loaded datasets - Training: {len(training dataset)} sample
# Hyperparameters options
num variants = 5
num epochs = 10
batch size learning rate options = [(32, 0.0001), (64, 0.0001), (128, 0.0001)]
ll lambda options = [0.0001, 0.00001, 0.000001]
12 \text{ lambda options} = [0.001, 0.0001, 0.00001]
dropout rate options = [0.1, 0.25, 0.5]
early stopping patience = 5
hyperparameters list = []
for in range(num variants):
    # Randomly select hyperparameters
    batch_size, learning_rate = random.choice(batch_size_learning_rate_d
    l1 lambda = random.choice(l1 lambda options)
    12 lambda = random.choice(l2 lambda options)
    dropout rate = random.choice(dropout rate options)
    # Construct the hyperparameters dictionary
    hyperparameters = {
        'num epochs': num epochs,
        'learning rate': learning rate,
        'batch size': batch size,
        '<mark>l1_lambda</mark>': l1 lambda,
        'l2 lambda': l2 lambda,
        'dropout rate': dropout rate,
        'early stopping patience': early stopping patience
    }
    # Ensure hyperparameters are unique
    if hyperparameters not in hyperparameters list:
        hyperparameters list.append(hyperparameters)
# Track best model and accuracy
best model accuracy = 0
best model id = None
# Track all model results
model results = []
# Train a model for each set of hyperparameters
for i, hyperparameters in enumerate(hyperparameters list):
    # Create a unique ID for this model
    model id = f"variant {i+1}"
    # Log start of training for this model
    logger.info("="*100)
    logger.info(f"Starting training for model {model id}")
    logger.info(f"Hyperparameters:")
    logger.info(f"Learning rate: {hyperparameters['learning rate']}")
    logger.info(f"Batch size: {hyperparameters['batch size']}")
    logger.info(f"L1 lambda: {hyperparameters['l1 lambda']}")
```

```
logger.info(f"L2 lambda: {hyperparameters['l2 lambda']}")
    logger.info(f"Dropout rate: {hyperparameters['dropout rate']}")
    logger.info(f"Early stopping patience: {hyperparameters['early stopp
    # Initialize a fresh model for each set of hyperparameters with the
    model = Perceptron(dropout rate=hyperparameters['dropout rate']).to(
    # Train the model and get test accuracy
    accuracy = train model(model, training dataset, testing dataset, hyp
    # Log the results for this model
    logger.info(f"Completed training for model {model id}")
    logger.info(f"Test accuracy: {accuracy:.2f}%")
    # Store model results
    model results.append({
        'model id': model id,
        'accuracy': accuracy,
        'hyperparameters': hyperparameters
    })
    # Track the best model
    if accuracy > best model accuracy:
        best model accuracy = accuracy
        best model id = model id
        logger.info(f"New best model: {model_id} with accuracy {accuracy
# Log summary of all models
logger.info("="*100)
logger.info("Training completed for all models")
logger.info("Summary of model performances:")
for result in sorted(model results, key=lambda x: x['accuracy'], reverse
    logger.info(f"Model {result['model id']}: Accuracy {result['accuracy
    logger.info(f" Learning rate: {result['hyperparameters']['learning
    logger.info(f" Batch size: {result['hyperparameters']['batch size']
    logger.info(f" L1 lambda: {result['hyperparameters']['l1 lambda']}"
    logger.info(f" L2 lambda: {result['hyperparameters']['l2 lambda']}"
    logger.info(f" Dropout rate: {result['hyperparameters']['dropout_ra
logger.info(f" Early stopping patience: {result['hyperparameters'][
    logger.info("-" * 50)
# Log best model details
logger.info("="*100)
logger.info(f"Best model: {best model id} with accuracy: {best model acc
best model hp = next(result['hyperparameters'] for result in model resul
logger.info(f"Best model hyperparameters:")
logger.info(f" Learning rate: {best model hp['learning rate']}")
                Batch size: {best model hp['batch size']}")
logger.info(f"
logger.info(f" L1 lambda: {best model hp['l1 lambda']}")
logger.info(f" L2 lambda: {best model hp['l2 lambda']}")
                Dropout rate: {best model hp['dropout rate']}")
logger.info(f"
logger.info(f"
                Early stopping patience: {best model hp['early stopping
logger.info("="*100)
```

```
______
2025-04-27 21:09:47,944 - INFO - Starting new experiment with run ID: 202504
27 210947
______
2025-04-27 21:09:47,945 - INFO - Using device: cuda
2025-04-27 21:09:47,944 - INFO - Starting new experiment with run ID: 202504
27 210947
_____
2025-04-27 21:09:47,945 - INFO - Using device: cuda
2025-04-27 21:09:48,009 - INFO - Loaded datasets - Training: 60000 samples,
Testing: 10000 samples
______
2025-04-27 21:09:48,010 - INFO - Starting training for model variant 1
2025-04-27 21:09:48,010 - INFO - Hyperparameters:
2025-04-27 21:09:48,011 - INFO - Learning rate: 0.0001
2025-04-27 21:09:48,011 - INFO - Batch size: 128
2025-04-27 21:09:48,011 - INFO - L1 lambda: 1e-06
2025-04-27 21:09:48,012 - INFO - L2 lambda: 1e-05
2025-04-27 21:09:48,012 - INFO - Dropout rate: 0.5
2025-04-27 21:09:48,012 - INFO - Early stopping patience: 5
______
2025-04-27 21:09:48,021 - INFO - Model ID: variant 1
2025-04-27 21:09:48,021 - INFO - Run ID: 20250427_210947
2025-04-27 21:09:48,022 - INFO - Training configuration:
2025-04-27 21:09:48,022 - INFO - Learning rate: 0.0001
2025-04-27 21:09:48,023 - INFO - Batch size: 128
2025-04-27 21:09:48,023 - INFO - L1 regularization: 1e-06
2025-04-27 21:09:48,023 - INFO - L2 regularization: 1e-05
2025-04-27 21:09:48,024 - INFO - Dropout rate: 0.5
2025-04-27 21:09:48,024 - INFO - Epochs: 10
2025-04-27 21:09:48,025 - INFO - Early stopping patience: 5
2025-04-27 21:09:48,025 - INFO - Optimizer: Adam
2025-04-27 21:09:48,025 - INFO - Loss function: CrossEntropyLoss
______
2025-04-27 21:09:58,803 - INFO - Epoch [1/10]
2025-04-27 21:09:58,803 - INFO - Training Loss: 0.9644
2025-04-27 21:10:01,291 - INFO - Validation Loss: 0.5133
2025-04-27 21:10:01,292 - INFO - Validation Accuracy: 90.58%
2025-04-27 21:10:01,297 - INFO - New best validation loss: 0.5133
2025-04-27 21:10:01,297 - INFO - Epoch 1 duration: 13.27 seconds
______
2025-04-27 21:10:12,017 - INFO - Epoch [2/10]
2025-04-27 21:10:12,017 - INFO - Training Loss: 0.5255
2025-04-27 21:10:14,517 - INFO - Validation Loss: 0.3081
2025-04-27 21:10:14,518 - INFO - Validation Accuracy: 93.79%
2025-04-27 21:10:14,524 - INFO - New best validation loss: 0.3081
2025-04-27 21:10:14,524 - INFO - Epoch 2 duration: 13.23 seconds
______
```

```
2025-04-27 21:10:25,222 - INFO - Epoch [3/10]
2025-04-27 21:10:25,223 - INFO - Training Loss: 0.3868
2025-04-27 21:10:27,683 - INFO - Validation Loss: 0.2268
2025-04-27 21:10:27,683 - INFO - Validation Accuracy: 95.03%
2025-04-27 21:10:27,688 - INFO - New best validation loss: 0.2268
2025-04-27 21:10:27,689 - INFO - Epoch 3 duration: 13.16 seconds
2025-04-27 21:10:38,553 - INFO - ========================
______
2025-04-27 21:10:38,554 - INFO - Epoch [4/10]
2025-04-27 21:10:38,554 - INFO - Training Loss: 0.3232
2025-04-27 21:10:41,125 - INFO - Validation Loss: 0.1859
2025-04-27 21:10:41,126 - INFO - Validation Accuracy: 95.59%
2025-04-27 21:10:41,131 - INFO - New best validation loss: 0.1859
2025-04-27 21:10:41,131 - INFO - Epoch 4 duration: 13.44 seconds
_____
2025-04-27 21:10:52,298 - INFO - Epoch [5/10]
2025-04-27 21:10:52,298 - INFO - Training Loss: 0.2828
2025-04-27 21:10:54,808 - INFO - Validation Loss: 0.1516
2025-04-27 21:10:54,809 - INFO - Validation Accuracy: 96.21%
2025-04-27 21:10:54,814 - INFO - New best validation loss: 0.1516
2025-04-27 21:10:54,814 - INFO - Epoch 5 duration: 13.68 seconds
______
2025-04-27 21:11:05,659 - INFO - Epoch [6/10]
2025-04-27 21:11:05,660 - INFO - Training Loss: 0.2563
2025-04-27 21:11:08,181 - INFO - Validation Loss: 0.1393
2025-04-27 21:11:08,181 - INFO - Validation Accuracy: 96.53%
2025-04-27 21:11:08,186 - INFO - New best validation loss: 0.1393
2025-04-27 21:11:08,187 - INFO - Epoch 6 duration: 13.37 seconds
______
2025-04-27 21:11:18,981 - INFO - Epoch [7/10]
2025-04-27 21:11:18,982 - INFO - Training Loss: 0.2350
2025-04-27 21:11:21,499 - INFO - Validation Loss: 0.1234
2025-04-27 21:11:21,500 - INFO - Validation Accuracy: 96.66%
2025-04-27 21:11:21,505 - INFO - New best validation loss: 0.1234
2025-04-27 21:11:21,505 - INFO - Epoch 7 duration: 13.32 seconds
_____
2025-04-27 21:11:32,599 - INFO - Epoch [8/10]
2025-04-27 21:11:32,600 - INFO - Training Loss: 0.2281
2025-04-27 21:11:35,128 - INFO - Validation Loss: 0.1130
2025-04-27 21:11:35,129 - INFO - Validation Accuracy: 96.91%
2025-04-27 21:11:35,136 - INFO - New best validation loss: 0.1130
2025-04-27 21:11:35,137 - INFO - Epoch 8 duration: 13.63 seconds
______
2025-04-27 21:11:45,923 - INFO - Epoch [9/10]
2025-04-27 21:11:45,924 - INFO - Training Loss: 0.2183
2025-04-27 21:11:48,411 - INFO - Validation Loss: 0.1101
2025-04-27 21:11:48,412 - INFO - Validation Accuracy: 97.00%
2025-04-27 21:11:48,417 - INFO - New best validation loss: 0.1101
2025-04-27 21:11:48,418 - INFO - Epoch 9 duration: 13.28 seconds
```

```
2025-04-27 21:11:59,187 - INFO - Epoch [10/10]
2025-04-27 21:11:59,187 - INFO - Training Loss: 0.2123
2025-04-27 21:12:01,765 - INFO - Validation Loss: 0.1047
2025-04-27 21:12:01,765 - INFO - Validation Accuracy: 97.02%
2025-04-27 21:12:01,770 - INFO - New best validation loss: 0.1047
2025-04-27 21:12:01,771 - INFO - Epoch 10 duration: 13.35 seconds
2025-04-27 21:12:01,771 - INFO - Training duration: 133.75 seconds
2025-04-27 21:12:01,772 - INFO - Completed all 10 epochs
2025-04-27 21:12:01,772 - INFO - Loading best model from: models/run 2025042
7 210947/model variant 1/checkpoints/epoch 10.pth
_____
2025-04-27 21:12:02,489 - INFO - Testing duration: 0.71 seconds
2025-04-27 21:12:02,489 - INFO - Total duration: 134.46 seconds
_____
2025-04-27 21:12:02,490 - INFO - Best Model: models/run 20250427 210947/mode
l variant 1/checkpoints/epoch 10.pth
2025-04-27 21:12:02,491 - INFO - Best Validation Loss: 0.1047
2025-04-27 21:12:02,491 - INFO - Test Loss: 0.7590
2025-04-27 21:12:02,492 - INFO - Test Accuracy: 72.61%
______
2025-04-27 21:12:02,493 - INFO - Completed training for model variant 1
2025-04-27 21:12:02,493 - INFO - Test accuracy: 72.61%
2025-04-27 21:12:02,494 - INFO - New best model: variant 1 with accuracy 72.
61%
_____
2025-04-27 21:12:02,495 - INFO - Starting training for model variant 2
2025-04-27 21:12:02,495 - INFO - Hyperparameters:
2025-04-27 21:12:02,496 - INFO - Learning rate: 0.0001
2025-04-27 21:12:02,496 - INFO - Batch size: 64
2025-04-27 21:12:02,496 - INFO - L1 lambda: 1e-06
2025-04-27 21:12:02,497 - INFO - L2 lambda: 1e-05
2025-04-27 21:12:02,497 - INFO - Dropout rate: 0.25
2025-04-27 21:12:02,497 - INFO - Early stopping patience: 5
_____
2025-04-27 21:12:02,506 - INFO - Model ID: variant 2
2025-04-27 21:12:02,506 - INFO - Run ID: 20250427 210947
2025-04-27 21:12:02,507 - INFO - Training configuration:
2025-04-27 21:12:02,507 - INFO - Learning rate: 0.0001
2025-04-27 21:12:02,508 - INFO - Batch size: 64
2025-04-27 21:12:02,508 - INFO - L1 regularization: 1e-06
2025-04-27 21:12:02,508 - INFO - L2 regularization: 1e-05
2025-04-27 21:12:02,509 - INFO -
                          Dropout rate: 0.25
2025-04-27 21:12:02,509 - INFO -
                          Epochs: 10
2025-04-27 21:12:02,510 - INFO -
                          Early stopping patience: 5
2025-04-27 21:12:02,510 - INFO - Optimizer: Adam
2025-04-27 21:12:02,510 - INFO - Loss function: CrossEntropyLoss
______
2025-04-27 21:12:14,093 - INFO - Epoch [1/10]
2025-04-27 21:12:14,093 - INFO - Training Loss: 0.6118
2025-04-27 21:12:16,632 - INFO - Validation Loss: 0.2700
```

```
2025-04-27 21:12:16,633 - INFO - Validation Accuracy: 95.02%
2025-04-27 21:12:16,637 - INFO - New best validation loss: 0.2700
2025-04-27 21:12:16,638 - INFO - Epoch 1 duration: 14.13 seconds
_____
2025-04-27 21:12:28,174 - INFO - Epoch [2/10]
2025-04-27 21:12:28,174 - INFO - Training Loss: 0.2741
2025-04-27 21:12:30,713 - INFO - Validation Loss: 0.1626
2025-04-27 21:12:30,714 - INFO - Validation Accuracy: 96.41%
2025-04-27 21:12:30,719 - INFO - New best validation loss: 0.1626
2025-04-27 21:12:30,720 - INFO - Epoch 2 duration: 14.08 seconds
2025-04-27 21:12:42,410 - INFO - =======================
______
2025-04-27 21:12:42,410 - INFO - Epoch [3/10]
2025-04-27 21:12:42,411 - INFO - Training Loss: 0.2018
2025-04-27 21:12:44,989 - INFO - Validation Loss: 0.1217
2025-04-27 21:12:44,990 - INFO - Validation Accuracy: 96.96%
2025-04-27 21:12:44,994 - INFO - New best validation loss: 0.1217
2025-04-27 21:12:44,995 - INFO - Epoch 3 duration: 14.27 seconds
_____
2025-04-27 21:12:56,704 - INFO - Epoch [4/10]
2025-04-27 21:12:56,705 - INFO - Training Loss: 0.1718
2025-04-27 21:12:59,243 - INFO - Validation Loss: 0.1069
2025-04-27 21:12:59,244 - INFO - Validation Accuracy: 97.31%
2025-04-27 21:12:59,249 - INFO - New best validation loss: 0.1069
2025-04-27 21:12:59,249 - INFO - Epoch 4 duration: 14.25 seconds
_____
2025-04-27 21:13:10,743 - INFO - Epoch [5/10]
2025-04-27 21:13:10,743 - INFO - Training Loss: 0.1530
2025-04-27 21:13:13,264 - INFO - Validation Loss: 0.0955
2025-04-27 21:13:13,264 - INFO - Validation Accuracy: 97.39%
2025-04-27 21:13:13,269 - INFO - New best validation loss: 0.0955
2025-04-27 21:13:13,270 - INFO - Epoch 5 duration: 14.02 seconds
_____
2025-04-27 21:13:24,702 - INFO - Epoch [6/10]
2025-04-27 21:13:24,703 - INFO - Training Loss: 0.1426
2025-04-27 21:13:27,230 - INFO - Validation Loss: 0.0873
2025-04-27 21:13:27,231 - INFO - Validation Accuracy: 97.51%
2025-04-27 21:13:27,236 - INFO - New best validation loss: 0.0873
2025-04-27 21:13:27,236 - INFO - Epoch 6 duration: 13.97 seconds
2025-04-27 21:13:38,784 - INFO - ==================
______
2025-04-27 21:13:38,785 - INFO - Epoch [7/10]
2025-04-27 21:13:38,786 - INFO - Training Loss: 0.1314
2025-04-27 21:13:41,308 - INFO - Validation Loss: 0.0784
2025-04-27 21:13:41,309 - INFO - Validation Accuracy: 97.90%
2025-04-27 21:13:41,313 - INFO - New best validation loss: 0.0784
2025-04-27 21:13:41,314 - INFO - Epoch 7 duration: 14.08 seconds
_____
2025-04-27 21:13:52,809 - INFO - Epoch [8/10]
2025-04-27 21:13:52,810 - INFO - Training Loss: 0.1263
2025-04-27 21:13:55,375 - INFO - Validation Loss: 0.0768
```

```
2025-04-27 21:13:55,375 - INFO - Validation Accuracy: 97.78%
2025-04-27 21:13:55,380 - INFO - New best validation loss: 0.0768
2025-04-27 21:13:55,380 - INFO - Epoch 8 duration: 14.07 seconds
______
2025-04-27 21:14:06,971 - INFO - Epoch [9/10]
2025-04-27 21:14:06,972 - INFO - Training Loss: 0.1191
2025-04-27 21:14:09,554 - INFO - Validation Loss: 0.0690
2025-04-27 21:14:09,555 - INFO - Validation Accuracy: 98.03%
2025-04-27 21:14:09,560 - INFO - New best validation loss: 0.0690
2025-04-27 21:14:09,560 - INFO - Epoch 9 duration: 14.18 seconds
______
2025-04-27 21:14:21,037 - INFO - Epoch [10/10]
2025-04-27 21:14:21,037 - INFO - Training Loss: 0.1159
2025-04-27 21:14:23,567 - INFO - Validation Loss: 0.0719
2025-04-27 21:14:23,568 - INFO - Validation Accuracy: 97.90%
2025-04-27 21:14:23,573 - INFO - No improvement for 1 epochs (patience: 5)
2025-04-27 21:14:23,573 - INFO - Epoch 10 duration: 14.01 seconds
2025-04-27 21:14:23,574 - INFO - Training duration: 141.06 seconds
2025-04-27 21:14:23,574 - INFO - Completed all 10 epochs
2025-04-27 21:14:23,574 - INFO - Loading best model from: models/run 2025042
7 210947/model variant 2/checkpoints/epoch 9.pth
_____
2025-04-27 21:14:24,329 - INFO - Testing duration: 0.75 seconds
2025-04-27 21:14:24,329 - INFO - Total duration: 141.82 seconds
______
2025-04-27 21:14:24,330 - INFO - Best Model: models/run 20250427 210947/mode
l variant 2/checkpoints/epoch 9.pth
2025-04-27 21:14:24,331 - INFO - Best Validation Loss: 0.0690
2025-04-27 21:14:24,331 - INFO - Test Loss: 0.5928
2025-04-27 21:14:24,332 - INFO - Test Accuracy: 81.12%
_____
2025-04-27 21:14:24,333 - INFO - Completed training for model variant 2
2025-04-27 21:14:24,334 - INFO - Test accuracy: 81.12%
2025-04-27 21:14:24,334 - INFO - New best model: variant 2 with accuracy 81.
______
2025-04-27 21:14:24,335 - INFO - Starting training for model variant_3
2025-04-27 21:14:24,335 - INFO - Hyperparameters:
2025-04-27 21:14:24,336 - INFO - Learning rate: 0.0001
2025-04-27 21:14:24,336 - INFO - Batch size: 64
2025-04-27 21:14:24,337 - INFO - L1 lambda: 0.0001
2025-04-27 21:14:24,337 - INFO - L2 lambda: 1e-05
2025-04-27 21:14:24,337 - INFO - Dropout rate: 0.5
2025-04-27 21:14:24,338 - INFO - Early stopping patience: 5
_____
2025-04-27 21:14:24,345 - INFO - Model ID: variant 3
2025-04-27 21:14:24,346 - INFO - Run ID: 20250427_210947
2025-04-27 21:14:24,346 - INFO - Training configuration:
2025-04-27 21:14:24,347 - INFO - Learning rate: 0.0001
```

```
2025-04-27 21:14:24,347 - INFO -
                           Batch size: 64
2025-04-27 21:14:24,347 - INFO -
                           L1 regularization: 0.0001
2025-04-27 21:14:24,348 - INFO -
                           L2 regularization: 1e-05
2025-04-27 21:14:24,348 - INFO -
                           Dropout rate: 0.5
2025-04-27 21:14:24,349 - INFO -
                           Epochs: 10
2025-04-27 21:14:24,349 - INFO -
                           Early stopping patience: 5
2025-04-27 21:14:24,350 - INFO -
                           Optimizer: Adam
2025-04-27 21:14:24,350 - INFO - Loss function: CrossEntropyLoss
______
2025-04-27 21:14:35,904 - INFO - Epoch [1/10]
2025-04-27 21:14:35,905 - INFO - Training Loss: 1.2636
2025-04-27 21:14:38,448 - INFO - Validation Loss: 0.3768
2025-04-27 21:14:38,449 - INFO - Validation Accuracy: 92.55%
2025-04-27 21:14:38,453 - INFO - New best validation loss: 0.3768
2025-04-27 21:14:38,454 - INFO - Epoch 1 duration: 14.10 seconds
_____
2025-04-27 21:14:50,192 - INFO - Epoch [2/10]
2025-04-27 21:14:50,192 - INFO - Training Loss: 0.7511
2025-04-27 21:14:52,723 - INFO - Validation Loss: 0.2383
2025-04-27 21:14:52,724 - INFO - Validation Accuracy: 94.62%
2025-04-27 21:14:52,729 - INFO - New best validation loss: 0.2383
2025-04-27 21:14:52,729 - INFO - Epoch 2 duration: 14.27 seconds
______
2025-04-27 21:15:04,190 - INFO - Epoch [3/10]
2025-04-27 21:15:04,191 - INFO - Training Loss: 0.6004
2025-04-27 21:15:06,729 - INFO - Validation Loss: 0.1803
2025-04-27 21:15:06,730 - INFO - Validation Accuracy: 95.60%
2025-04-27 21:15:06,734 - INFO - New best validation loss: 0.1803
2025-04-27 21:15:06,735 - INFO - Epoch 3 duration: 14.01 seconds
_____
2025-04-27 21:15:18,262 - INFO - Epoch [4/10]
2025-04-27 21:15:18,263 - INFO - Training Loss: 0.5409
2025-04-27 21:15:20,832 - INFO - Validation Loss: 0.1600
2025-04-27 21:15:20,833 - INFO - Validation Accuracy: 95.82%
2025-04-27 21:15:20,838 - INFO - New best validation loss: 0.1600
2025-04-27 21:15:20,838 - INFO - Epoch 4 duration: 14.10 seconds
_____
2025-04-27 21:15:32,334 - INFO - Epoch [5/10]
2025-04-27 21:15:32,334 - INFO - Training Loss: 0.5071
2025-04-27 21:15:34,885 - INFO - Validation Loss: 0.1445
2025-04-27 21:15:34,886 - INFO - Validation Accuracy: 96.00%
2025-04-27 21:15:34,890 - INFO - New best validation loss: 0.1445
2025-04-27 21:15:34,891 - INFO - Epoch 5 duration: 14.05 seconds
_____
2025-04-27 21:15:46,508 - INFO - Epoch [6/10]
2025-04-27 21:15:46,508 - INFO - Training Loss: 0.4833
2025-04-27 21:15:49,095 - INFO - Validation Loss: 0.1327
2025-04-27 21:15:49,096 - INFO - Validation Accuracy: 96.26%
2025-04-27 21:15:49,101 - INFO - New best validation loss: 0.1327
2025-04-27 21:15:49,101 - INFO - Epoch 6 duration: 14.21 seconds
```

```
______
2025-04-27 21:16:00,555 - INFO - Epoch [7/10]
2025-04-27 21:16:00,555 - INFO - Training Loss: 0.4685
2025-04-27 21:16:03,103 - INFO - Validation Loss: 0.1229
2025-04-27 21:16:03,103 - INFO - Validation Accuracy: 96.67%
2025-04-27 21:16:03,108 - INFO - New best validation loss: 0.1229
2025-04-27 21:16:03,109 - INFO - Epoch 7 duration: 14.01 seconds
_____
2025-04-27 21:16:14,654 - INFO - Epoch [8/10]
2025-04-27 21:16:14,655 - INFO - Training Loss: 0.4614
2025-04-27 21:16:17,200 - INFO - Validation Loss: 0.1232
2025-04-27 21:16:17,200 - INFO - Validation Accuracy: 96.53%
2025-04-27 21:16:17,205 - INFO - No improvement for 1 epochs (patience: 5)
2025-04-27 21:16:17,206 - INFO - Epoch 8 duration: 14.10 seconds
_____
2025-04-27 21:16:28,723 - INFO - Epoch [9/10]
2025-04-27 21:16:28,724 - INFO - Training Loss: 0.4503
2025-04-27 21:16:31,257 - INFO - Validation Loss: 0.1147
2025-04-27 21:16:31,257 - INFO - Validation Accuracy: 96.73%
2025-04-27 21:16:31,262 - INFO - New best validation loss: 0.1147
2025-04-27 21:16:31,263 - INFO - Epoch 9 duration: 14.06 seconds
______
2025-04-27 21:16:42,802 - INFO - Epoch [10/10]
2025-04-27 21:16:42,803 - INFO - Training Loss: 0.4405
2025-04-27 21:16:45,350 - INFO - Validation Loss: 0.1101
2025-04-27 21:16:45,350 - INFO - Validation Accuracy: 96.83%
2025-04-27 21:16:45,355 - INFO - New best validation loss: 0.1101
2025-04-27 21:16:45,356 - INFO - Epoch 10 duration: 14.09 seconds
2025-04-27 21:16:45,356 - INFO - Training duration: 141.01 seconds
2025-04-27 21:16:45,356 - INFO - Completed all 10 epochs
2025-04-27 21:16:45,357 - INFO - Loading best model from: models/run 2025042
7 210947/model variant 3/checkpoints/epoch 10.pth
_____
2025-04-27 21:16:46,110 - INFO - Testing duration: 0.75 seconds
2025-04-27 21:16:46,111 - INFO - Total duration: 141.76 seconds
_____
2025-04-27 21:16:46,112 - INFO - Best Model: models/run_20250427_210947/mode
l variant 3/checkpoints/epoch 10.pth
2025-04-27 21:16:46,112 - INFO - Best Validation Loss: 0.1101
2025-04-27 21:16:46,112 - INFO - Test Loss: 0.4928
2025-04-27 21:16:46,113 - INFO - Test Accuracy: 96.58%
_____
2025-04-27 21:16:46,114 - INFO - Completed training for model variant 3
2025-04-27 21:16:46,114 - INFO - Test accuracy: 96.58%
2025-04-27 21:16:46,114 - INFO - New best model: variant 3 with accuracy 96.
______
2025-04-27 21:16:46,115 - INFO - Starting training for model variant 4
```

```
2025-04-27 21:16:46,115 - INFO - Hyperparameters:
2025-04-27 21:16:46,116 - INFO - Learning rate: 0.0001
2025-04-27 21:16:46,116 - INFO - Batch size: 32
2025-04-27 21:16:46,116 - INFO - L1 lambda: 1e-06
2025-04-27 21:16:46,117 - INFO - L2 lambda: 0.0001
2025-04-27 21:16:46,117 - INFO - Dropout rate: 0.5
2025-04-27 21:16:46,117 - INFO - Early stopping patience: 5
_____
2025-04-27 21:16:46,125 - INFO - Model ID: variant 4
2025-04-27 21:16:46,125 - INFO - Run ID: 20250427 210947
2025-04-27 21:16:46,125 - INFO - Training configuration:
2025-04-27 21:16:46,126 - INFO - Learning rate: 0.0001
2025-04-27 21:16:46,127 - INFO -
                          Batch size: 32
2025-04-27 21:16:46,127 - INFO -
                          L1 regularization: 1e-06
2025-04-27 21:16:46,127 - INFO - L2 regularization: 0.0001
2025-04-27 21:16:46,128 - INFO - Dropout rate: 0.5
2025-04-27 21:16:46,128 - INFO -
                          Epochs: 10
2025-04-27 21:16:46,129 - INFO -
                          Early stopping patience: 5
2025-04-27 21:16:46,129 - INFO - Optimizer: Adam
2025-04-27 21:16:46,129 - INFO - Loss function: CrossEntropyLoss
______
2025-04-27 21:16:59,405 - INFO - Epoch [1/10]
2025-04-27 21:16:59,406 - INFO - Training Loss: 0.7898
2025-04-27 21:17:02,092 - INFO - Validation Loss: 0.3410
2025-04-27 21:17:02,093 - INFO - Validation Accuracy: 92.63%
2025-04-27 21:17:02,098 - INFO - New best validation loss: 0.3410
2025-04-27 21:17:02,098 - INFO - Epoch 1 duration: 15.97 seconds
_____
2025-04-27 21:17:15,313 - INFO - Epoch [2/10]
2025-04-27 21:17:15,313 - INFO - Training Loss: 0.4159
2025-04-27 21:17:17,950 - INFO - Validation Loss: 0.2074
2025-04-27 21:17:17,951 - INFO - Validation Accuracy: 94.55%
2025-04-27 21:17:17,956 - INFO - New best validation loss: 0.2074
2025-04-27 21:17:17,956 - INFO - Epoch 2 duration: 15.86 seconds
_____
2025-04-27 21:17:31,163 - INFO - Epoch [3/10]
2025-04-27 21:17:31,163 - INFO - Training Loss: 0.3297
2025-04-27 21:17:33,817 - INFO - Validation Loss: 0.1599
2025-04-27 21:17:33,818 - INFO - Validation Accuracy: 95.72%
2025-04-27 21:17:33,823 - INFO - New best validation loss: 0.1599
2025-04-27 21:17:33,823 - INFO - Epoch 3 duration: 15.87 seconds
_____
2025-04-27 21:17:46,925 - INFO - Epoch [4/10]
2025-04-27 21:17:46,926 - INFO - Training Loss: 0.2942
2025-04-27 21:17:49,624 - INFO - Validation Loss: 0.1410
2025-04-27 21:17:49,625 - INFO - Validation Accuracy: 96.16%
2025-04-27 21:17:49,631 - INFO - New best validation loss: 0.1410
2025-04-27 21:17:49,632 - INFO - Epoch 4 duration: 15.81 seconds
_____
2025-04-27 21:18:03,072 - INFO - Epoch [5/10]
```

```
2025-04-27 21:18:03,072 - INFO - Training Loss: 0.2730
2025-04-27 21:18:05,717 - INFO - Validation Loss: 0.1268
2025-04-27 21:18:05,718 - INFO - Validation Accuracy: 96.47%
2025-04-27 21:18:05,724 - INFO - New best validation loss: 0.1268
2025-04-27 21:18:05,724 - INFO - Epoch 5 duration: 16.09 seconds
_____
2025-04-27 21:18:19,047 - INFO - Epoch [6/10]
2025-04-27 21:18:19,047 - INFO - Training Loss: 0.2562
2025-04-27 21:18:21,699 - INFO - Validation Loss: 0.1150
2025-04-27 21:18:21,699 - INFO - Validation Accuracy: 96.78%
2025-04-27 21:18:21,704 - INFO - New best validation loss: 0.1150
2025-04-27 21:18:21,705 - INFO - Epoch 6 duration: 15.98 seconds
_____
2025-04-27 21:18:34,870 - INFO - Epoch [7/10]
2025-04-27 21:18:34,871 - INFO - Training Loss: 0.2418
2025-04-27 21:18:37,510 - INFO - Validation Loss: 0.1126
2025-04-27 21:18:37,511 - INFO - Validation Accuracy: 96.71%
2025-04-27 21:18:37,516 - INFO - New best validation loss: 0.1126
2025-04-27 21:18:37,516 - INFO - Epoch 7 duration: 15.81 seconds
______
2025-04-27 21:18:50,729 - INFO - Epoch [8/10]
2025-04-27 21:18:50,730 - INFO - Training Loss: 0.2330
2025-04-27 21:18:53,378 - INFO - Validation Loss: 0.1097
2025-04-27 21:18:53,379 - INFO - Validation Accuracy: 96.84%
2025-04-27 21:18:53,384 - INFO - New best validation loss: 0.1097
2025-04-27 21:18:53,385 - INFO - Epoch 8 duration: 15.87 seconds
______
2025-04-27 21:19:06,586 - INFO - Epoch [9/10]
2025-04-27 21:19:06,586 - INFO - Training Loss: 0.2306
2025-04-27 21:19:09,242 - INFO - Validation Loss: 0.1018
2025-04-27 21:19:09,242 - INFO - Validation Accuracy: 96.91%
2025-04-27 21:19:09,247 - INFO - New best validation loss: 0.1018
2025-04-27 21:19:09,248 - INFO - Epoch 9 duration: 15.86 seconds
_____
2025-04-27 21:19:22,463 - INFO - Epoch [10/10]
2025-04-27 21:19:22,464 - INFO - Training Loss: 0.2199
2025-04-27 21:19:25,131 - INFO - Validation Loss: 0.0949
2025-04-27 21:19:25,132 - INFO - Validation Accuracy: 97.24%
2025-04-27 21:19:25,137 - INFO - New best validation loss: 0.0949
2025-04-27 21:19:25,137 - INFO - Epoch 10 duration: 15.89 seconds
2025-04-27 21:19:25,138 - INFO - Training duration: 159.01 seconds
2025-04-27 21:19:25,138 - INFO - Completed all 10 epochs
2025-04-27 21:19:25,139 - INFO - Loading best model from: models/run 2025042
7 210947/model variant 4/checkpoints/epoch 10.pth
______
2025-04-27 21:19:25,983 - INFO - Testing duration: 0.84 seconds
2025-04-27 21:19:25,984 - INFO - Total duration: 159.85 seconds
2025-04-27 21:19:25,985 - INFO - Best Model: models/run 20250427 210947/mode
```

```
l variant 4/checkpoints/epoch 10.pth
2025-04-27 21:19:25,985 - INFO - Best Validation Loss: 0.0949
2025-04-27 21:19:25,986 - INFO - Test Loss: 0.6466
2025-04-27 21:19:25,986 - INFO - Test Accuracy: 81.94%
_____
2025-04-27 21:19:25,987 - INFO - Completed training for model variant 4
2025-04-27 21:19:25,988 - INFO - Test accuracy: 81.94%
2025-04-27 21:19:25,988 - INFO - ============================
_____
2025-04-27 21:19:25,989 - INFO - Starting training for model variant 5
2025-04-27 21:19:25,989 - INFO - Hyperparameters:
2025-04-27 21:19:25,990 - INFO - Learning rate: 0.0001
2025-04-27 21:19:25,990 - INFO - Batch size: 32
2025-04-27 21:19:25,990 - INFO - L1 lambda: 1e-05
2025-04-27 21:19:25,991 - INFO - L2 lambda: 0.0001
2025-04-27 21:19:25,991 - INFO - Dropout rate: 0.25
2025-04-27 21:19:25,992 - INFO - Early stopping patience: 5
_____
2025-04-27 21:19:25,999 - INFO - Model ID: variant 5
2025-04-27 21:19:25,999 - INFO - Run ID: 20250427_210947
2025-04-27 21:19:26,000 - INFO - Training configuration:
2025-04-27 21:19:26,000 - INFO - Learning rate: 0.0001
2025-04-27 21:19:26,001 - INFO - Batch size: 32
2025-04-27 21:19:26,001 - INFO - L1 regularization: 1e-05
2025-04-27 21:19:26,002 - INFO - L2 regularization: 0.0001
2025-04-27 21:19:26,002 - INFO -
                            Dropout rate: 0.25
2025-04-27 21:19:26,002 - INFO - Epochs: 10
2025-04-27 21:19:26,003 - INFO - Early stopping patience: 5
2025-04-27 21:19:26,004 - INFO - Optimizer: Adam
2025-04-27 21:19:26,004 - INFO - Loss function: CrossEntropyLoss
2025-04-27 21:19:39,321 - INFO - ============================
_____
2025-04-27 21:19:39,322 - INFO - Epoch [1/10]
2025-04-27 21:19:39,322 - INFO - Training Loss: 0.5892
2025-04-27 21:19:41,974 - INFO - Validation Loss: 0.2142
2025-04-27 21:19:41,975 - INFO - Validation Accuracy: 95.28%
2025-04-27 21:19:41,979 - INFO - New best validation loss: 0.2142
2025-04-27 21:19:41,980 - INFO - Epoch 1 duration: 15.98 seconds
_____
2025-04-27 21:19:55,154 - INFO - Epoch [2/10]
2025-04-27 21:19:55,154 - INFO - Training Loss: 0.2850
2025-04-27 21:19:57,797 - INFO - Validation Loss: 0.1352
2025-04-27 21:19:57,797 - INFO - Validation Accuracy: 96.44%
2025-04-27 21:19:57,802 - INFO - New best validation loss: 0.1352
2025-04-27 21:19:57,803 - INFO - Epoch 2 duration: 15.82 seconds
_____
2025-04-27 21:20:11,009 - INFO - Epoch [3/10]
2025-04-27 21:20:11,009 - INFO - Training Loss: 0.2319
2025-04-27 21:20:13,644 - INFO - Validation Loss: 0.1120
2025-04-27 21:20:13,644 - INFO - Validation Accuracy: 96.85%
2025-04-27 21:20:13,649 - INFO - New best validation loss: 0.1120
2025-04-27 21:20:13,650 - INFO - Epoch 3 duration: 15.85 seconds
```

```
_____
2025-04-27 21:20:26,898 - INFO - Epoch [4/10]
2025-04-27 21:20:26,899 - INFO - Training Loss: 0.2070
2025-04-27 21:20:29,563 - INFO - Validation Loss: 0.0950
2025-04-27 21:20:29,564 - INFO - Validation Accuracy: 97.59%
2025-04-27 21:20:29,569 - INFO - New best validation loss: 0.0950
2025-04-27 21:20:29,570 - INFO - Epoch 4 duration: 15.92 seconds
_____
2025-04-27 21:20:42,767 - INFO - Epoch [5/10]
2025-04-27 21:20:42,767 - INFO - Training Loss: 0.1925
2025-04-27 21:20:45,424 - INFO - Validation Loss: 0.0864
2025-04-27 21:20:45,425 - INFO - Validation Accuracy: 97.49%
2025-04-27 21:20:45,429 - INFO - New best validation loss: 0.0864
2025-04-27 21:20:45,430 - INFO - Epoch 5 duration: 15.86 seconds
______
2025-04-27 21:20:58,675 - INFO - Epoch [6/10]
2025-04-27 21:20:58,676 - INFO - Training Loss: 0.1898
2025-04-27 21:21:01,330 - INFO - Validation Loss: 0.0789
2025-04-27 21:21:01,331 - INFO - Validation Accuracy: 97.85%
2025-04-27 21:21:01,335 - INFO - New best validation loss: 0.0789
2025-04-27 21:21:01,336 - INFO - Epoch 6 duration: 15.91 seconds
______
2025-04-27 21:21:14,556 - INFO - Epoch [7/10]
2025-04-27 21:21:14,557 - INFO - Training Loss: 0.1771
2025-04-27 21:21:17,202 - INFO - Validation Loss: 0.0707
2025-04-27 21:21:17,202 - INFO - Validation Accuracy: 98.07%
2025-04-27 21:21:17,207 - INFO - New best validation loss: 0.0707
2025-04-27 21:21:17,208 - INFO - Epoch 7 duration: 15.87 seconds
______
2025-04-27 21:21:30,434 - INFO - Epoch [8/10]
2025-04-27 21:21:30,434 - INFO - Training Loss: 0.1711
2025-04-27 21:21:33,101 - INFO - Validation Loss: 0.0707
2025-04-27 21:21:33,101 - INFO - Validation Accuracy: 97.92%
2025-04-27 21:21:33,106 - INFO - New best validation loss: 0.0707
2025-04-27 21:21:33,107 - INFO - Epoch 8 duration: 15.90 seconds
2025-04-27 21:21:46,335 - INFO - ===================
______
2025-04-27 21:21:46,336 - INFO - Epoch [9/10]
2025-04-27 21:21:46,336 - INFO - Training Loss: 0.1665
2025-04-27 21:21:48,976 - INFO - Validation Loss: 0.0713
2025-04-27 21:21:48,977 - INFO - Validation Accuracy: 97.91%
2025-04-27 21:21:48,982 - INFO - No improvement for 1 epochs (patience: 5)
2025-04-27 21:21:48,982 - INFO - Epoch 9 duration: 15.88 seconds
2025-04-27 21:22:02,207 - INFO - =============================
______
2025-04-27 21:22:02,208 - INFO - Epoch [10/10]
2025-04-27 21:22:02,209 - INFO - Training Loss: 0.1618
2025-04-27 21:22:04,873 - INFO - Validation Loss: 0.0668
2025-04-27 21:22:04,874 - INFO - Validation Accuracy: 98.03%
2025-04-27 21:22:04,879 - INFO - New best validation loss: 0.0668
2025-04-27 21:22:04,879 - INFO - Epoch 10 duration: 15.90 seconds
```

```
2025-04-27 21:22:04,880 - INFO - Training duration: 158.88 seconds
2025-04-27 21:22:04,880 - INFO - Completed all 10 epochs
2025-04-27 21:22:04,881 - INFO - Loading best model from: models/run 2025042
7 210947/model variant 5/checkpoints/epoch 10.pth
_____
2025-04-27 21:22:05,723 - INFO - Testing duration: 0.84 seconds
2025-04-27 21:22:05,723 - INFO - Total duration: 159.72 seconds
2025-04-27 21:22:05,724 - INFO - ============================
_____
2025-04-27 21:22:05,724 - INFO - Best Model: models/run 20250427 210947/mode
l variant 5/checkpoints/epoch 10.pth
2025-04-27 21:22:05,724 - INFO - Best Validation Loss: 0.0668
2025-04-27 21:22:05,725 - INFO - Test Loss: 0.5400
2025-04-27 21:22:05,725 - INFO - Test Accuracy: 91.01%
_____
2025-04-27 21:22:05,726 - INFO - Completed training for model variant_5
2025-04-27 21:22:05,727 - INFO - Test accuracy: 91.01%
_____
2025-04-27 21:22:05,727 - INFO - Training completed for all models
2025-04-27 21:22:05,728 - INFO - Summary of model performances:
2025-04-27 21:22:05,728 - INFO - Model variant_3: Accuracy 96.58%
2025-04-27 21:22:05,729 - INFO - Learning rate: 0.0001
2025-04-27 21:22:05,729 - INFO - Batch size: 64
2025-04-27 21:22:05,730 - INFO - L1 lambda: 0.0001
2025-04-27 21:22:05,730 - INFO - L2 lambda: 1e-05
2025-04-27 21:22:05,730 - INFO - Dropout rate: 0.5
2025-04-27 21:22:05,731 - INFO - Early stopping patience: 5
2025-04-27 21:22:05,731 - INFO - -------
2025-04-27 21:22:05,732 - INFO - Model variant 5: Accuracy 91.01%
2025-04-27 21:22:05,732 - INFO - Learning rate: 0.0001
2025-04-27 21:22:05,733 - INFO - Batch size: 32
2025-04-27 21:22:05,733 - INFO - L1 lambda: 1e-05
2025-04-27 21:22:05,734 - INFO - L2 lambda: 0.0001
2025-04-27 21:22:05,734 - INFO - Dropout rate: 0.25
2025-04-27 21:22:05,734 - INFO - Early stopping patience: 5
2025-04-27 21:22:05,735 - INFO - ------
2025-04-27 21:22:05,735 - INFO - Model variant_4: Accuracy 81.94%
2025-04-27 21:22:05,736 - INFO - Learning rate: 0.0001
2025-04-27 21:22:05,736 - INFO - Batch size: 32
2025-04-27 21:22:05,737 - INFO - L1 lambda: 1e-06
2025-04-27 21:22:05,737 - INFO - L2 lambda: 0.0001
2025-04-27 21:22:05,738 - INFO -
                            Dropout rate: 0.5
2025-04-27 21:22:05,738 - INFO -
                            Early stopping patience: 5
2025-04-27 21:22:05,739 - INFO - -----
2025-04-27 21:22:05,739 - INFO - Model variant 2: Accuracy 81.12%
2025-04-27 21:22:05,739 - INFO - Learning rate: 0.0001
2025-04-27 21:22:05,740 - INFO - Batch size: 64
2025-04-27 21:22:05,740 - INFO - L1 lambda: 1e-06
2025-04-27 21:22:05,740 - INFO - L2 lambda: 1e-05
2025-04-27 21:22:05,741 - INFO -
                            Dropout rate: 0.25
```

```
2025-04-27 21:22:05,741 - INFO - Early stopping patience: 5
2025-04-27 21:22:05,741 - INFO - ------
2025-04-27 21:22:05,742 - INFO - Model variant 1: Accuracy 72.61%
2025-04-27 21:22:05,742 - INFO - Learning rate: 0.0001
2025-04-27 21:22:05,743 - INFO - Batch size: 128
2025-04-27 21:22:05,744 - INFO - L1 lambda: 1e-06
2025-04-27 21:22:05,744 - INFO - L2 lambda: 1e-05
2025-04-27 21:22:05,744 - INFO - Dropout rate: 0.5
2025-04-27 21:22:05,745 - INFO - Early stopping patience: 5
2025-04-27 21:22:05,745 - INFO - -------
_____
2025-04-27 21:22:05,746 - INFO - Best model: variant 3 with accuracy: 96.58%
2025-04-27 21:22:05,746 - INFO - Best model hyperparameters:
2025-04-27 21:22:05,746 - INFO - Learning rate: 0.0001
2025-04-27 21:22:05,747 - INFO - Batch size: 64
2025-04-27 21:22:05,747 - INFO - L1 lambda: 0.0001
2025-04-27 21:22:05,747 - INFO - L2 lambda: 1e-05
2025-04-27 21:22:05,748 - INFO - Dropout rate: 0.5
2025-04-27 21:22:05,748 - INFO - Early stopping patience: 5
```