

# ACSE-4 Mini Project #4



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# Project Aim

To create and implement a neural network to maximise the prediction accuracy of Kuzushiji characters by training on the KMNIST Dataset



# Training Approach / Sustainability

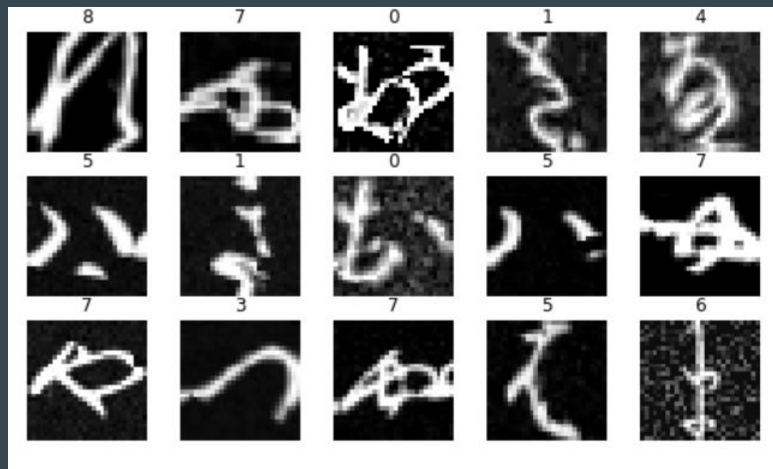
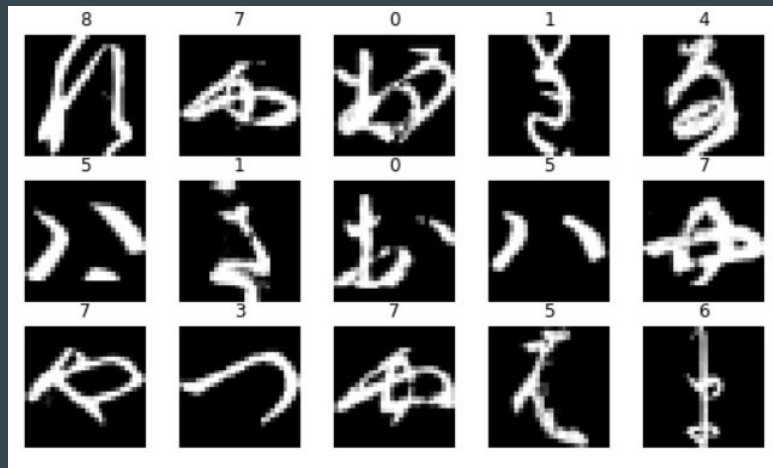
```
# Create Supervised Learning Object
learning = SupervisedLearning(X_aug, y_aug.long(), alexnet, optimiser, loss_function, batch_size, test_batch_size,
                             device=device,
                             transform=False,
                             seed=42, n_epochs=n_epochs,
                             val_ratio=0.1, n_splits=n_folds+1,
                             early_stop = True,
                             patience = patience,
                             tol = early_stop_tol)
```

## ‘SupervisedLearning’ wrapper class

- Sustainability and Coherence
- Parameter Tuning
- Input models, optimisers and loss functions
- Early stop
- In-built Data Augmentation
- Sharing models on drive
- Github integration
- Working on different Colab Notebooks
- Saving function to save model and parameters
- Easy to perform Transfer Learning to continue training our own model

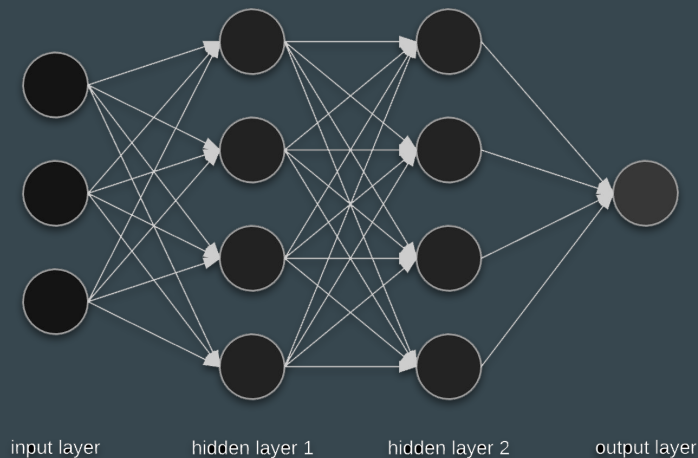
# Data Pre-processing and Augmentation

- Exploratory Data Analysis (EDA)
- Normalisation based on training set
- PyTorch transforms (class in-built)
  - Random Rotation
  - Random Crop
- Albumentation transforms (external)
  - Blur
  - Noise
  - Distortions



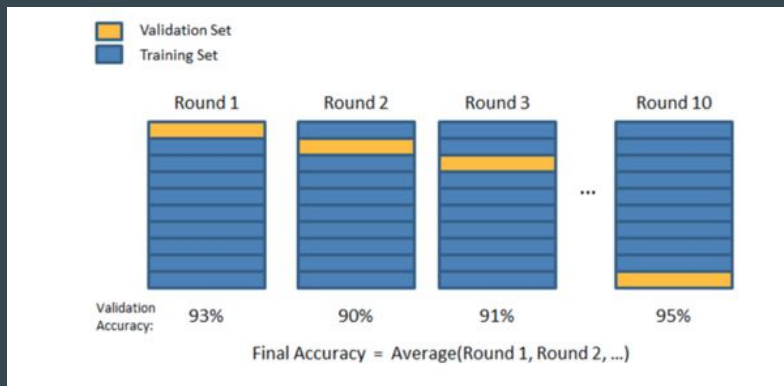
# Neural Network Architectures

- Modifications made to AlexNet and LeNet5 to improve validation accuracy
  - AlexNetMod
  - DeepAlexNet
  - SDeepAlexNet
  - mLeNet5
- Increase in network depth
- Regularisation with dropout and batch normalisation
- Ensembles



# Validation

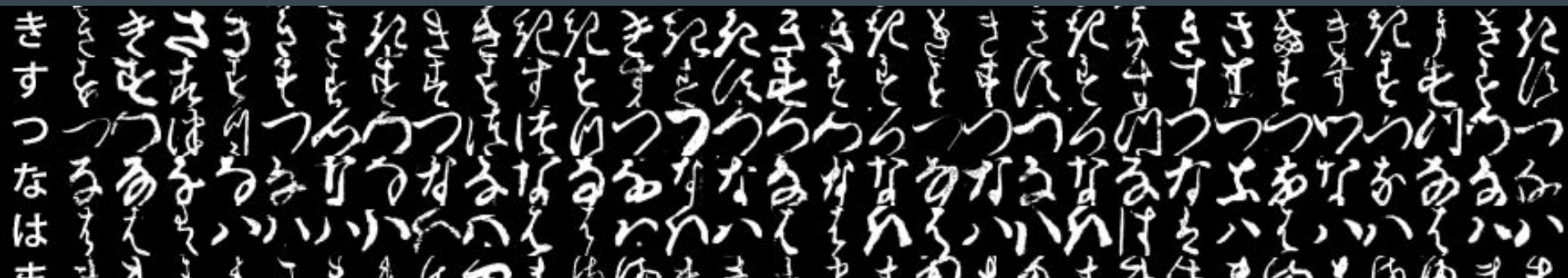
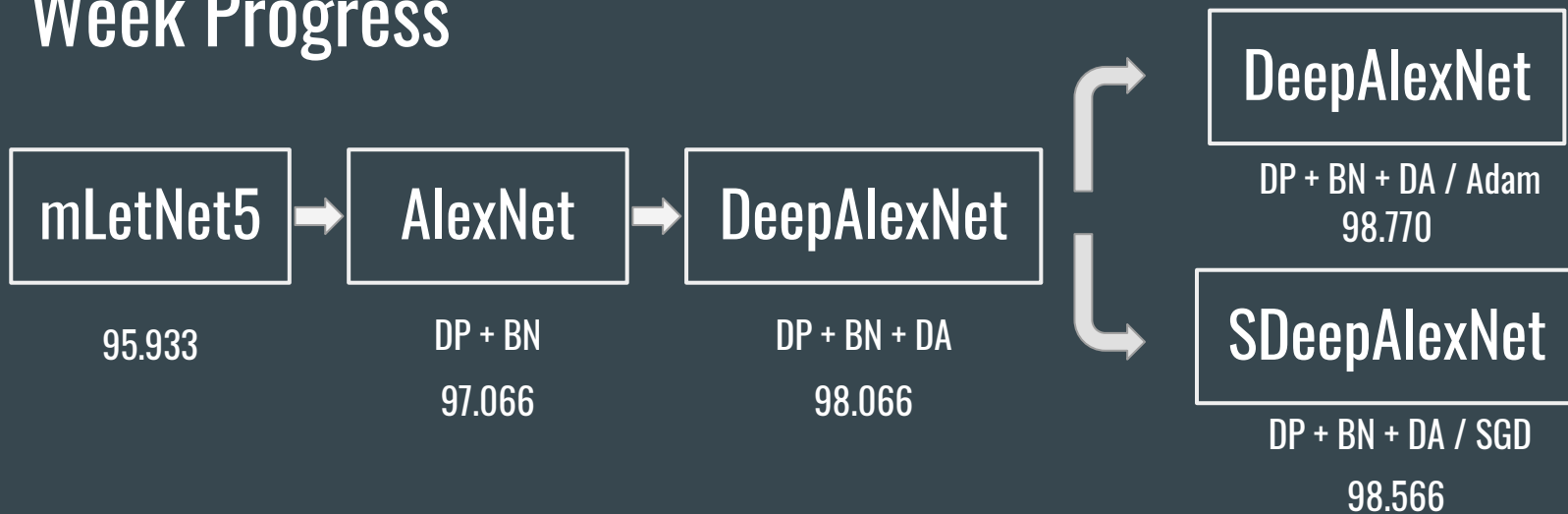
- Large dataset: balanced data split
- Hold-out validation (quicker!)
- K-Fold validation for better estimation of generalised error (much slower!)



# Hyperparameter Tuning on the AlexNet

- Optimiser: Adam vs Stochastic Gradient Descent
- Learning Rate:  $1e-2$ ,  $1e-4$
- Adam unstable on larger learning rate
- Weight Decay: 0 ,  $1e-3$ ,  $1e-4$ ,  $1e-5$

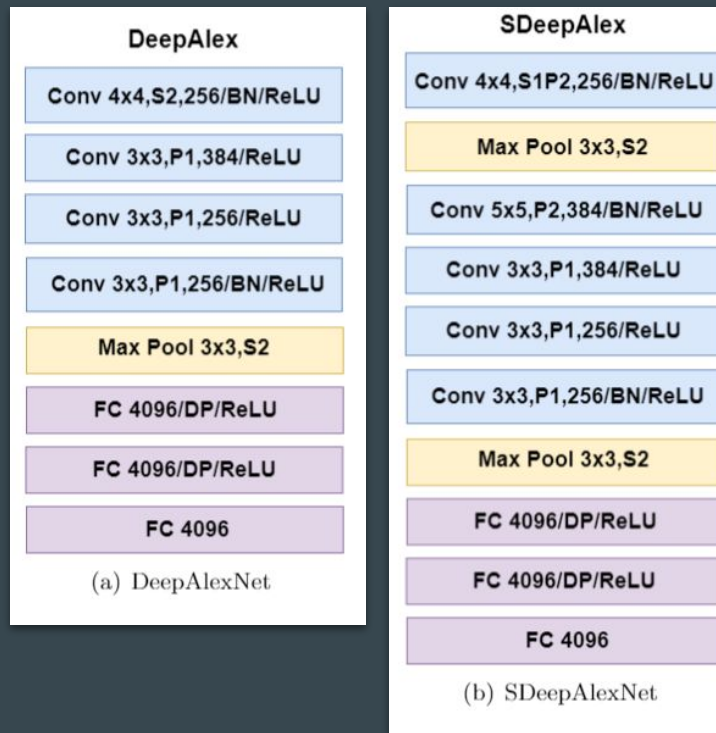
# Week Progress





# Highest Scoring Models

- After validation, trained on full dataset
- Transfer Learning on previously ran models -- time saving!!
- DeepAlex and SDeepAlex with best results



DeepAlex	SGD 50 Epochs Transfer Learning:, last layer training with external 'albumentation' transforms (20 epochs)	Batch Normalisation Dropout Weight Decay 1e-4	97.90%
DeepAlex	Adam 50 Epochs In-built data augmentation		98.77%
SDeepAlex	SGD 50 Epochs In-built data augmentation Weight Decay 1e-4		98.57%

# Limitations and Further Work

- Time and Computational Power!
- Explore different activation and loss functions
- Tune other hyperparameters
- Full-run with 'albummentation' DA instead of transfer learning
- Explore other well-established models and ensemble them

