#### Classification with Convolutional Neural Networks

Classify the data using a Convolutional Neural Network with the following architecture:

• input features x • convolutional layer with 32, 3x3 filters, stride 1, padding 1x1 • max pooling layer 2,2 • convolutional layer with 32, 3x3 filters, stride 1, padding 1x1 • flatten the output • output layer: one fully connected layer with five output values • a SoftMax layer to transform the outputs into a multi-class probability distribution for classification • activation functions: internal layers all use ReLU activation • optimizer: Stochastic Gradient Descent

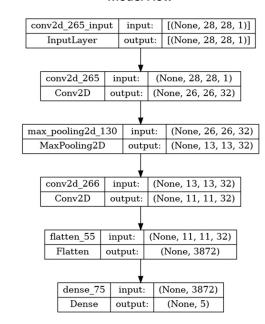
#### **Model Summary**

Model: "sequential 55"

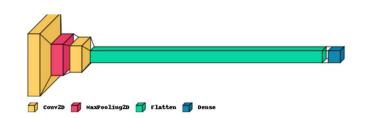
Layer (type)	Output Shape	Param #
conv2d_265 (Conv2D)	(None, 26, 26, 32)	320
max_pooling2d_130 (MaxPooli ng2D)	(None, 13, 13, 32)	0
conv2d_266 (Conv2D)	(None, 11, 11, 32)	9248
flatten_55 (Flatten)	(None, 3872)	0
dense_75 (Dense)	(None, 5)	19365

Non-trainable params: 0

#### **Model Flow**

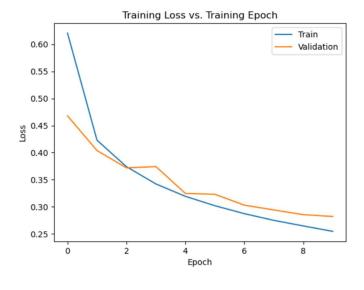


#### **Model Architecture**



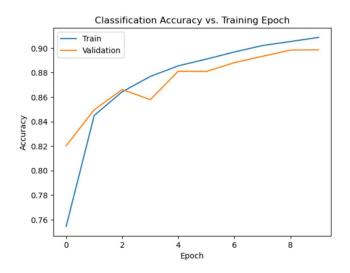
#### Runtime Information (Epochs, Loss, Accuracy)

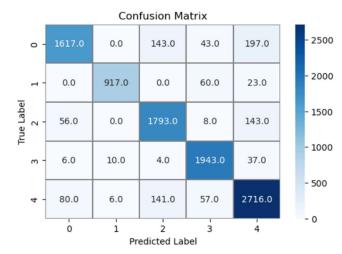
#### **Evaluation of Model and Performance Metrics**



#### **Classification Report**

	precision	recall	f1-score	support
0	0.92	0.81	0.86	2000
1	0.98	0.92	0.95	1000
2	0.86	0.90	0.88	2000
3	0.92	0.97	0.95	2000
4	0.87	0.91	0.89	3000
accuracy			0.90	10000
macro avg	0.91	0.90	0.90	10000
weighted avg	0.90	0.90	0.90	10000





Using Default Network with the given layers and hyperparameters, we observed **90.93** % **Train Accuracy and 90.39** % **Test Accuracy** with **10 epochs** and with the **default Batch Size**. Observations:

- From the Training Loss vs Training Epoch Plot, Loss value for Train went down for every epoch. The similar pattern we observed for Validation Set.
- From the Classification Accuracy vs Training Epoch Plot, Accuracy value for Train went up for every epoch. The similar pattern we observed for Validation Set.
- we don't observe overfitting within these 10 epochs for both the plots.
- From the classification report,
  - o **Precision** for the five classes ranges from 86% to 98%, meaning that 86% to 98%, of Predicted positive classes were truly positive.
  - o Recall for the five classes ranges from 81% to 97%, meaning that 81% to 97% of Actual positive classes were truly positive.
  - F1 Score for the five classes is above 86% on an average, meaning that it is correctly identifying the positive cases and avoiding false positives.

# MODEL 1 - Variation of filters with more convolution layers

#### Model - 1 Summary

Model: "sequential\_56"

Layer (type)	Output Shape	Param #
conv2d_267 (Conv2D)	(None, 26, 26, 64)	640
conv2d_268 (Conv2D)	(None, 24, 24, 64)	36928
max_pooling2d_131 (MaxPooli ng2D)	(None, 12, 12, 64)	0
conv2d_269 (Conv2D)	(None, 10, 10, 128)	73856
conv2d_270 (Conv2D)	(None, 8, 8, 128)	147584
max_pooling2d_132 (MaxPooli ng2D)	(None, 4, 4, 128)	0
conv2d_271 (Conv2D)	(None, 2, 2, 256)	295168
flatten_56 (Flatten)	(None, 1024)	0
dense_76 (Dense)	(None, 5)	5125

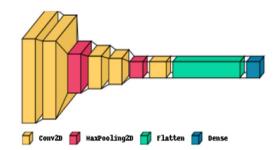
-----

Total params: 559,301 Trainable params: 559,301 Non-trainable params: 0

#### **Runtime Information (Epochs, Loss, Accuracy)**

Epoch 1/25 235/235 [====================================
Epoch 2/25 235/235 [====================================
Epoch 3/25 235/235 [====================================
Epoch 4/25 235/235 [====================================
Epoch 5/25 235/235 [====================================
Epoch 6/25 235/235 [====================================
Epoch 7/25 235/235 [====================================
Epoch 8/25 235/235 [====================================
Epoch 9/25 235/235 [====================================
Epoch 10/25 235/235 [====================================
Epoch 11/25 235/235 [====================================
Epoch 12/25 235/235 [====================================
Epoch 13/25 235/235 [====================================
Epoch 14/25 235/235 [====================================
Epoch 15/25 235/235 [====================================
Epoch 16/25 235/235 [====================================
Epoch 17/25 235/235 [===============] - 3s 14ms/step - loss: 0.3138 - accuracy: 0.8862 - val_loss: 0.3656 - val_accuracy: 0.8617
Epoch 18/25 235/235 [====================================
Epoch 19/25 235/235 [====================================
Epoch 20/25 235/235 [====================================
Epoch 21/25 235/235 [====================================
Epoch 22/25 235/235 [====================================
Epoch 23/25 235/235 [====================================
Epoch 24/25 235/235 [====================================
Epoch 25/25 235/235 [====================================

#### **Model Architecture**



## MODEL 2 - With activation function - LeakyReLU & addition of Dropout

#### Model - 2 Summary

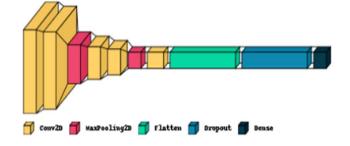
Model: "sequential\_57"

Layer (type)	Output Shape	Param #
conv2d_272 (Conv2D)	(None, 26, 26, 64)	640
conv2d_273 (Conv2D)	(None, 24, 24, 64)	36928
<pre>max_pooling2d_133 (MaxPooli ng2D)</pre>	(None, 12, 12, 64)	0
conv2d_274 (Conv2D)	(None, 10, 10, 128)	73856
conv2d_275 (Conv2D)	(None, 8, 8, 128)	147584
<pre>max_pooling2d_134 (MaxPooli ng2D)</pre>	(None, 4, 4, 128)	0
conv2d_276 (Conv2D)	(None, 2, 2, 256)	295168
flatten_57 (Flatten)	(None, 1024)	0
dropout_45 (Dropout)	(None, 1024)	0
dense_77 (Dense)	(None, 5)	5125

Total params: 559,301 Trainable params: 559,301 Non-trainable params: 0

# Runtime Information (Epochs, Loss, Accuracy)

Epoch 1/25 235/235 [====================================	
Epoch 2/25 235/235 [====================================	
Epoch 3/25 235/235 [====================================	
Epoch 4/25 235/235 [====================================	
Epoch 5/25 235/235 [====================================	
Epoch 6/25 235/235 [====================================	
Epoch 7/25 235/235 [=============] - 4s 15ms/step - loss: 0.4611 - accuracy: 0.8172 - val_loss: 0.4890 - val_accuracy: 0.8117	
Epoch 8/25 235/235 [===========-=	
Epoch 9/25 235/235 [===============] - 4s 16ms/step - loss: 0.4198 - accuracy: 0.8360 - val_loss: 0.4335 - val_accuracy: 0.8301	
Epoch 10/25 235/235 [====================================	
Epoch 11/25 235/235 [====================================	
Epoch 12/25 235/235 [====================================	
Epoch 13/25 235/235 [====================================	
Epoch 14/25 235/235 [====================================	
Epoch 15/25 235/235 [====================================	
Epoch 16/25 235/235 [====================================	
Epoch 17/25 235/235 [====================================	
Epoch 18/25 235/235 [====================================	
Epoch 19/25 235/235 [====================================	
Epoch 20/25 235/235 [====================================	
Epoch 21/25 235/235 [====================================	
Epoch 22/25 235/235 [====================================	
Epoch 23/25 235/235 [====================================	
Epoch 24/25 235/235 [====================================	
Epoch 25/25 235/235 [====================================	



## Model – 3 Summary

Model: "sequential\_58"

Layer (type)	Output Shape	Param #
conv2d_277 (Conv2D)	(None, 26, 26, 64)	640
conv2d_278 (Conv2D)	(None, 24, 24, 64)	36928
max_pooling2d_135 (MaxPool ng2D)	i (None, 12, 12, 64)	0
conv2d_279 (Conv2D)	(None, 10, 10, 128)	73856
conv2d_280 (Conv2D)	(None, 8, 8, 128)	147584
max_pooling2d_136 (MaxPool ng2D)	i (None, 4, 4, 128)	0
conv2d_281 (Conv2D)	(None, 2, 2, 256)	295168
flatten_58 (Flatten)	(None, 1024)	0
dropout_46 (Dropout)	(None, 1024)	0
dense_78 (Dense)	(None, 5)	5125

.....

Total params: 559,301 Trainable params: 559,301 Non-trainable params: 0

# Model – 4 Summary

Model: "sequ	ential_59"	
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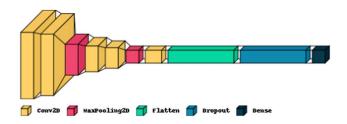
Layer (type)	Output Shape	Param #
conv2d_282 (Conv2D)	(None, 26, 26, 64)	640
conv2d_283 (Conv2D)	(None, 24, 24, 64)	36928
max_pooling2d_137 (MaxPooling2D)	(None, 12, 12, 64)	0
conv2d_284 (Conv2D)	(None, 10, 10, 128)	73856
conv2d_285 (Conv2D)	(None, 8, 8, 128)	147584
max_pooling2d_138 (MaxPooling2D)	(None, 4, 4, 128)	О
conv2d_286 (Conv2D)	(None, 2, 2, 256)	295168
flatten_59 (Flatten)	(None, 1024)	0
dropout_47 (Dropout)	(None, 1024)	0
dense_79 (Dense)	(None, 5)	5125

Total params: 559,301 Trainable params: 559,301 Non-trainable params: 0

## **Runtime Information (Epochs, Loss, Accuracy)**

Epoch 1/25 235/235 [====================================	
Epoch 1/25 235/235 [====================================	
Epoch 3/25 235/235 [====================================	
Epoch 4/25 235/235 [====================================	
Epoch 5/25 235/235 [====================================	
Epoch 6/25 235/235 [====================================	
Epoch 7/25 235/235 [====================================	
Epoch 8/25 235/235 [====================================	
Epoch 9/25 235/235 [====================================	
Epoch 10/25 235/235 [====================================	)
Epoch 11/25 235/235 [====================================	J
Epoch 12/25 235/235 [====================================	)
Epoch 13/25 235/235 [====================================	J
Epoch 14/25 235/235 [====================================	)
Epoch 15/25 235/235 [====================================	)
Epoch 16/25 235/235 [====================================	)
Epoch 17/25 235/235 [====================================	)
Epoch 18/25 235/235 [====================================	0
Epoch 19/25 235/235 [====================================	)
Epoch 20/25 235/235 [====================================	
Epoch 21/25 235/235 [====================================	
Epoch 22/25 235/235 [====================================	
Epoch 23/25 235/235 [====================================	
Epoch 24/25 235/235 [====================================	
Epoch 25/25 235/235 [====================================	
Epoch 23/23 [	,

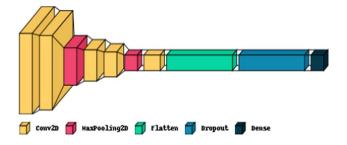
#### **Model Architecture**



## MODEL 4 - With activation function - tanh & addition of Dropout

## **Runtime Information (Epochs, Loss, Accuracy)**

Epoch 1/25 235/235 [====================================
Epoch 2/25 235/235 [====================================
Epoch 3/25 235/235 [====================================
Epoch 4/25 235/235 [====================================
Epoch 5/25 235/235 [====================================
Epoch 6/25 235/235 [====================================
Epoch 7/25 235/235 [====================================
Epoch 8/25 235/235 [====================================
Epoch 9/25 235/235 [====================================
Epoch 10/25 235/235 [====================================
Epoch 11/25 235/235 [====================================
Epoch 12/25 235/235 [====================================
Epoch 13/25 235/235 [====================================
Epoch 14/25 235/235 [====================================
Epoch 15/25 235/235 [====================================
Epoch 16/25 235/235 [====================================
Epoch 17/25 235/235 [====================================
Epoch 18/25 235/235 [====================================
Epoch 19/25 235/235 [====================================
Epoch 20/25 235/235 [====================================
Epoch 21/25 235/235 [====================================
Epoch 22/25 235/235 [====================================
Epoch 23/25 235/235 [====================================
Epoch 24/25 235/235 [====================================
Epoch 25/25 235/235 [====================================



#### Model - 5 Summary

Model: "sequential\_60"

Layer (type)	Output Shape	Param #
conv2d_287 (Conv2D)	(None, 26, 26, 64)	640
conv2d_288 (Conv2D)	(None, 24, 24, 64)	36928
max_pooling2d_139 (MaxPool: ng2D)	(None, 12, 12, 64)	0
conv2d_289 (Conv2D)	(None, 10, 10, 128)	73856
conv2d_290 (Conv2D)	(None, 8, 8, 128)	147584
max_pooling2d_140 (MaxPool: ng2D)	i (None, 4, 4, 128)	0
conv2d_291 (Conv2D)	(None, 2, 2, 256)	295168
flatten_60 (Flatten)	(None, 1024)	0
dropout_48 (Dropout)	(None, 1024)	0
dense_80 (Dense)	(None, 5)	5125

Total params: 559,301

Trainable params: 559,301 Non-trainable params: 0

## Model – 6 Summary

Model: "sequential\_61"

Trainable params: 559,301 Non-trainable params: 0

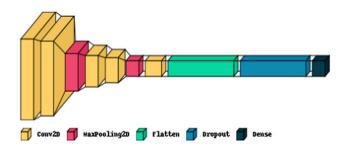
Layer (type)	Output Shape	Param #
conv2d_292 (Conv2D)	(None, 26, 26, 64)	640
conv2d_293 (Conv2D)	(None, 24, 24, 64)	36928
max_pooling2d_141 (MaxPool: ng2D)	(None, 12, 12, 64)	0
conv2d_294 (Conv2D)	(None, 10, 10, 128)	73856
conv2d_295 (Conv2D)	(None, 8, 8, 128)	147584
max_pooling2d_142 (MaxPool: ng2D)	(None, 4, 4, 128)	0
conv2d_296 (Conv2D)	(None, 2, 2, 256)	295168
flatten_61 (Flatten)	(None, 1024)	0
dropout_49 (Dropout)	(None, 1024)	0
dense_81 (Dense)	(None, 5)	5125

## Runtime Information (Epochs, Loss, Accuracy)

MODEL 5 - With activation function - tanh & using Adam optimizer

Epoch 1/25 469/469 [====================================
Epoch 2/25 469/469 [====================================
Epoch 3/25 469/469 [====================================
Epoch 4/25 469/469 [====================================
Epoch 5/25 469/469 [====================================
Epoch 6/25 469/469 [====================================
Epoch 7/25 469/469 [====================================
Epoch 8/25 469/469 [====================================
Epoch 9/25 469/469 [====================================

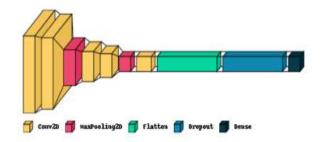
#### **Model Architecture**



# MODEL 6 - With activation function - LeakyReLU & using Adam optimizer

#### Runtime Information (Epochs, Loss, Accuracy)

Epoch 1/25 469/469 [====================================
Epoch 2/25 469/469 [====================================
Epoch 3/25 469/469 [====================================
Epoch 4/25 469/469 [====================================
Epoch 5/25 469/469 [====================================
Epoch 6/25 469/469 [====================================
Epoch 7/25 469/469 [====================================
Epoch 8/25 469/469 [====================================
Epoch 9/25 469/469 [====================================
Epoch 10/25 469/469 [====================================



## Model – 7 Summary

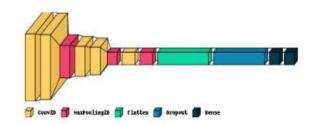
Layer (type)	Output Shape	Param #
conv2d_297 (Conv2D)	(None, 27, 27, 64)	320
conv2d_298 (Conv2D)	(None, 26, 26, 64)	16448
max_pooling2d_143 (MaxPooling2D)	(None, 13, 13, 64)	Θ
conv2d_299 (Conv2D)	(None, 12, 12, 128)	32896
conv2d_300 (Conv2D)	(None, 11, 11, 128)	65664
max_pooling2d_144 (MaxPooling2D)	(None, 5, 5, 128)	0
conv2d_301 (Conv2D)	(None, 4, 4, 256)	131328
max_pooling2d_145 (MaxPooli ng2D)	(None, 2, 2, 256)	0
flatten_62 (Flatten)	(None, 1024)	0
dropout_50 (Dropout)	(None, 1024)	0
dense_82 (Dense)	(None, 5)	5125
dense_83 (Dense)	(None, 5)	30

Trainable params: 251,811 Non-trainable params: 0

## Runtime Information (Epochs, Loss, Accuracy)

Epoch 1/25 469/469 [====================================
Epoch 2/25 469/469 [====================================
Epoch 3/25 469/469 [====================================
Epoch 4/25 469/469 [====================================
Epoch 5/25 469/469 [====================================
Epoch 6/25 469/469 [====================================
Epoch 7/25 469/469 [====================================
Epoch 8/25 469/469 [====================================
Epoch 9/25 469/469 [====================================
Epoch 10/25 469/469 [====================================
Epoch 11/25 469/469 [====================================
Epoch 12/25 469/469 [====================================
Epoch 13/25 469/469 [====================================
Epoch 14/25 469/469 [====================================
Epoch 15/25 469/469 [====================================
Epoch 16/25 469/469 [====================================

## **Model Architecture**



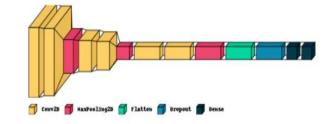
# MODEL 8 - With activation function - LeakyReLU & variation of kernel size & addition of Early Stopping

#### Model – 8 Summary

	None, 27, 27, 128)	
conv2d_382 (Conv2D) ((	None, 27, 27, 128)	
conv2d_383 (Conv2D) (		649
	None, 26, 26, 128)	
max pooling2d 146 (MaxPooli		65664
ng2D)	(None, 13, 13, 128)	6
conv2d_384 (Conv2D) (	None, 12, 12, 256)	131328
conv2d_305 (Conv2D) (	None, 11, 11, 256)	262400
max_pooling2d_147 (MaxPooli ng2D)	(None, 5, 5, 256)	8
conv2d_386 (Conv2D) (	None, 4, 4, 512)	524800
conv2d_387 (Conv2D) (	None, 3, 3, 512)	1849888
max_pooling2d_148 (MaxPooli ng2D)	(None, 1, 1, 512)	8
flatten_63 (Flatten) (	None, 512)	0
dropout_51 (Dropout) (	None, 512)	9
dense_84 (Dense) (	None, 5)	2565
dense_85 (Dense) (	None, 5)	30
	*****	******
otal params: 2,836,515		
rainable params: 2,036,515 kon-trainable params: 0		

# **Runtime Information (Epochs, Loss, Accuracy)**

Epoch 1/25 469/469 [====================================
Epoch 2/25 469/469 [====================================
Epoch 3/25 469/469 [====================================
Epoch 4/25 469/469 [====================================
Epoch 5/25 469/469 [====================================
Epoch 6/25 469/469 [====================================
Epoch 7/25 469/469 [====================================
Epoch 8/25 469/469 [====================================
Epoch 9/25 469/469 [====================================
Epoch 10/25 469/469 [=========] - 11s 24ms/step - loss: 0.1413 - accuracy: 0.9484 - val_loss: 0.2196 - val_accuracy: 0.9268
Epoch 11/25 469/469 [============] - 11s 24ms/step - loss: 0.1269 - accuracy: 0.9538 - val loss: 0.2212 - val accuracy: 0.9296



# Model –9 Summary

#### Model: "sequential 64"

Layer (type)		Shape	Param #
conv2d_308 (Conv2D)			649
conv2d_389 (Conv2D)	(None,	26, 26, 128)	65664
max_pooling2d_149 (MaxPooli ng2D)	(None	, 13, 13, 128)	0
conv2d_318 (Conv2D)	(None,	12, 12, 256)	131328
conv2d_311 (Conv2D)	(None,	11, 11, 256)	262400
max_pooling2d_150 (MaxPooli ng2D)	(None	, 5, 5, 256)	0
conv2d_312 (Conv2D)	(None,	4, 4, 512)	524899
conv2d_313 (Conv2D)	(None,	3, 3, 512)	1049088
max_pooling2d_151 (MaxPooli ng2D)	(None	, 1, 1, 512)	0
flatten_64 (Flatten)	(None,	512)	.0
dropout_52 (Dropout)	(None,	512)	0
dense_86 (Dense)	(None,	5)	2565
dense_87 (Dense)	(None,	5)	30

Trainable params: 2,836,515 Non-trainable params: 8

Total params: 2,036,515

Trainable params: 461,347 Non-trainable params: 0

## Model - 10 Summary

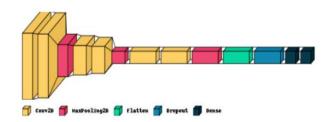
Model: "sequential_65"		
Layer (type)	Output Shape	Param #
	***************	
conv2d_314 (Conv2D)	(None, 27, 27, 128)	649
conv2d_315 (Conv2D)	(None, 26, 26, 128)	65664
conv2d_316 (Conv2D)	(None, 25, 25, 256)	131328
conv2d_317 (Conv2D)	(None, 24, 24, 256)	252400
max_pooling2d_152 (MexPooling2D)	(None, 12, 12, 256)	е
max_pooling2d_153 (MaxPooling2D)	(None, 6, 6, 256)	8
max_pooling2d_154 (MaxPooling2D)	(None, 3, 3, 256)	8
max_pooling2d_155 (MaxPooling2D)	(None, 1, 1, 256)	9
flatten_65 (Flatten)	(None, 256)	е
dropout_53 (Dropout)	(None, 256)	0
dense_88 (Dense)	(None, 5)	1285
dense_89 (Dense)	(None, 5)	30
	**************	
Total params: 461,347		

# MODEL 9 - Multiple activation functions - relu, LeakyReLU & tanh

## Runtime Information (Epochs, Loss, Accuracy)

Epoch 1/25 469/469 [====================================
Epoch 2/25 469/469 [====================================
Epoch 3/25 469/469 [====================================
Epoch 4/25 469/469 [====================================
Epoch 5/25 469/469 [====================================
Epoch 6/25 469/469 [====================================
Epoch 7/25 469/469 [====================================
Epoch 8/25 469/469 [====================================
Epoch 9/25 469/469 [====================================
Epoch 10/25 469/469 [====================================
Epoch 11/25 469/469 [====================================
Epoch 12/25 469/469 [====================================
Epoch 13/25 469/469 [====================================
Epoch 14/25 469/469 [====================================

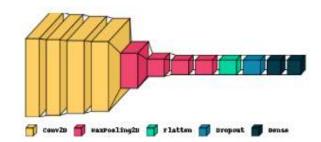
## **Model Architecture**



# MODEL 10 - With activation function - LeakyReLU & varying the order of layers

# Runtime Information (Epochs, Loss, Accuracy)

Epoch 1/25 469/469 [===========] - 22s 42ms/step - loss: 0.7582 - accuracy: 0.6976 - val_loss: 0.5347 - val_accuracy: 0.7872
Epoch 2/25 469/469 [====================================
Epoch 3/25 469/469 [====================================
Epoch 4/25 469/469 [====================================
Epoch 5/25 469/469 [====================================
Epoch 6/25 469/469 [====================================
Epoch 7/25 469/469 [====================================
Epoch 8/25 469/469 [====================================
Epoch 9/25 469/469 [====================================
Epoch 10/25 469/469 [====================================
Epoch 11/25 469/469 [====================================
Epoch 12/25 469/469 [====================================
Epoch 13/25 469/469 [====================================
Epoch 14/25 469/469 [====================================



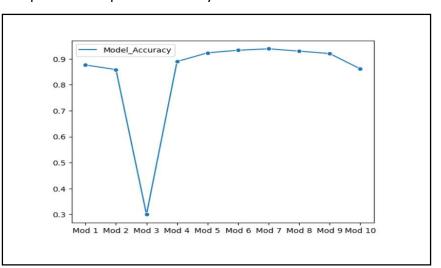
#### **Observations from All 10 models:**

- We tried creating multiple networks by increasing the number of layers and tried varying the sequence of layers. Increasing the number of layers and varying the sequence of layers can potentially improve a neural network's ability to learn complex patterns and features from input data
- We tried different optimizers like Adam and SGD. It adapts the learning rate for each parameter based on estimates of the first and second moments of the gradients, which can help it converge faster and more reliably
- We tried Early Stopping to overcome the overfitting. The idea is to stop the training process before the model starts to overfit the training data. This is done by monitoring the validation loss during training and stopping the training process when the validation loss stops improving.
- We tried Dropout function for regularization. This can help prevent the model from relying too much on any single feature or neuron and can force the model to learn more robust and generalizable representations of the data.
- We tried Varying Activation Functions Like Sigmoid, LeakyReLU, ReLU, tanH. They introduce non-linearity into the model and enable it to learn complex patterns in the data
- We tried using different size of filters in Convolutional layers and Max Pooling Layers. By varying the size of the filters and pooling windows, we can explore different trade-offs between model complexity and performance
- We varied kernel Size for the convolutional layers to ensure that our model is able to capture the relevant patterns in the data and generalize well to new data.

#### **Best Model:**

We evaluated all models with the test data. Here is the table of accuracy comparison and plot of accuracy for all models.

	Mod	lel_Accuracy
Mod	1	0.8762
Mod	2	0.8585
Mod	3	0.3000
Mod	4	0.8895
Mod	5	0.9231
Mod	6	0.9330
Mod	7	0.9386
Mod	8	0.9296
Mod	9	0.9203
Mod	10	0.8616



# When comparing the accuracy of all models we developed, Model 7 gave the best accuracy.

Possibly because We used leaky ReLU as Activation function which is more effective when comparing with other activation functions we used. Besides we used drop out and early stopping to avoid overfitting. Adam optimizer also helped this model to learn optimally.

From the Models we created, we finalized the Model 7 in terms of accuracy. It provides the effective accuracy when comparing with the other models we created.

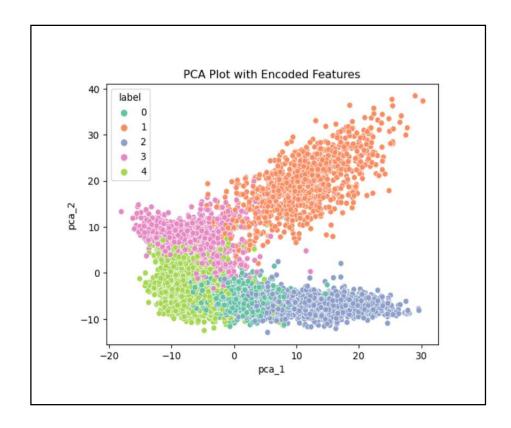
Model 7 has the accuracy of 93.86 %.

We defined an intermediate layer model and we extracted the train encoding sets and Test encoding sets.

## Activity 1: PCA - Visualizing encoding with the first two components

We used test encoding data to extract the PCA components and we plotted for the two components.

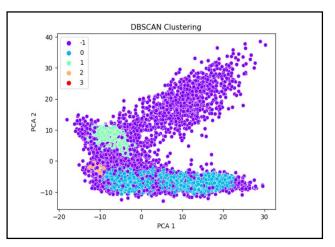
we observed that the different clothing categories are clustered together in certain regions of the plot, indicating that the encoding is effective at capturing the visual similarities between different types of clothing.

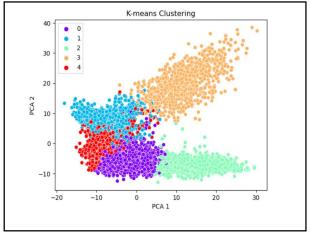


Activity 2: DBSCAN and K means for Clustering and Visualization.

We used test encoding data to perform DBSCAN and K Means and we generated the DB Scan predicted labels and K Means Predicted labels.

From the Graph we observed that for DBSCAN, the labels are filled with outliers (-1) and it clustered very bad but in the case of K Means it clustered effectively and it's really helpful for the prediction of Mystery Labels.

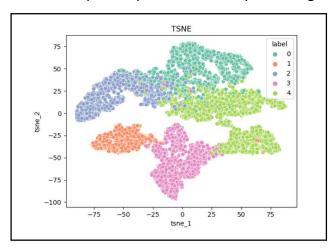


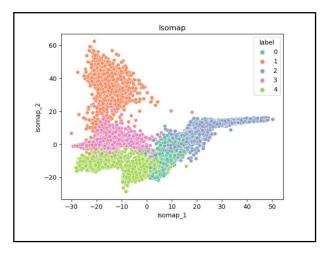


K Means performed well for clustering as it grouped the different classes properly and it has very less outliers but in the case of DBSCAN, it underperformed both in Clustering of classes and outliers.

## Activity 3: t-SNE and ISOMAP for Clustering and Visualization.

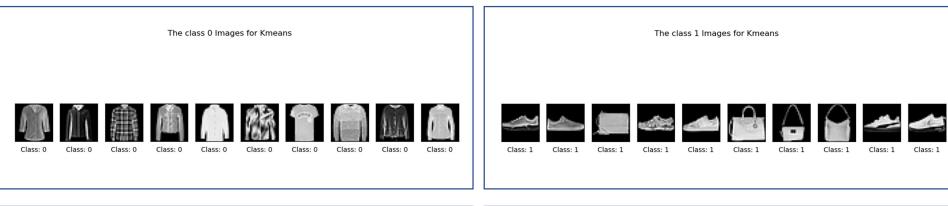
We have tried T-SNE and Isomap (Dimensionality Reduction Techniques) for the visualization to predict the mystery labels. The features extracted from this two techniques helped us effectively to distinguish between the different types of classes (clothing's)

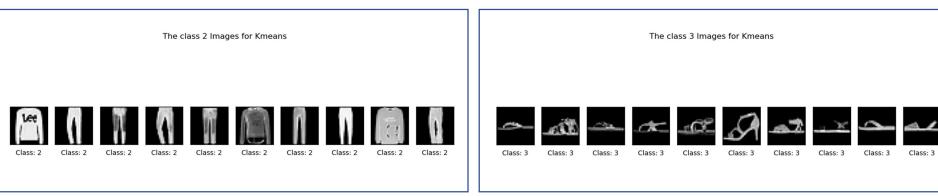


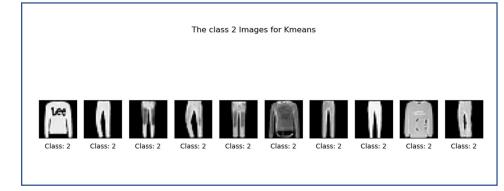


**Activity 4: Prediction of Mystery Labels.** 

# We used K means Labels to identify the Mystery Labels.







From the Images, we predicted the mystery labels for all 5 classes. Class 0 - Shoes and Bags

Class 1 - Casual shirts, T-shirts and Skirts
Class 2 - Lounge Pants and Sweatshirts

Class 3 - Sandals and Flip-flops

Class 4 - Boots and Formal wear

#### **Activity 5: Using this Encoding in a Creative Way.**

We used Encoded data of Model 7 as an Input to the Autoencoders.

Reasons to use Auto Encoders:

- During training, the input to the autoencoder is the noisy data, and the output is the corresponding clean data. The autoencoder learns to encode the noisy input into a latent space representation, and then decode this representation to reconstruct the clean data.
- Autoencoders can learn to extract meaningful features from high-dimensional data. These features can be used for various tasks, such as image recognition, classification, and clustering.

We have used three encoder layers and three decoded layers with activation function as LeakyReLU with Dropout and early stopping for regularization and to avoid overfitting respectively.

We tried with Adam optimizer and loss function as Mean Squared Error with 20 Epochs.

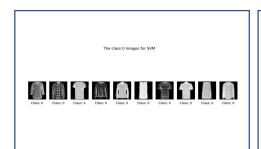
As we applied early stopping, we got Accuracy as 99.45% in the 12<sup>th</sup> Epoch.

With this Auto encoder models, we extracted the train and test datasets.

We used it in SVM Classifier which is considered to be the most efficient image classification model in terms of handling high dimensional spaces and robustness to noise when comparing with other Classification Algorithms like Decision trees and Logistic Regression.

We got the test Accuracy of 94.06 % using SVM.

With the Labels predicted using SVM we found the Mystery Labels.











From the Images, we predicted the mystery labels for all 5 classes.

Class 0 - Casual shirts, T-shirts and Skirts

Class 1 - Sandals and Flip-flops

Class 2 - Lounge Pants and Sweatshirts

Class 3 - Shoes and Bags

Class 4 - Boots and Formal wear