

Mnist

September 22, 2024

```
[35]: import tensorflow
      from tensorflow import keras
      from tensorflow.keras import Sequential
      from tensorflow.keras.layers import ␣
      ↪Dense, Flatten, Conv2D, MaxPooling2D, Dropout, BatchNormalization
```

```
[3]: (X_train, y_train), (X_test, y_test) = keras.datasets.mnist.load_data()
```

Downloading data from <https://storage.googleapis.com/tensorflow/tf-keras-datasets/mnist.npz>
11490434/11490434 0s
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```
[4]: X_train.shape
```

```
[4]: (60000, 28, 28)
```

```
[5]: X_test.shape
```

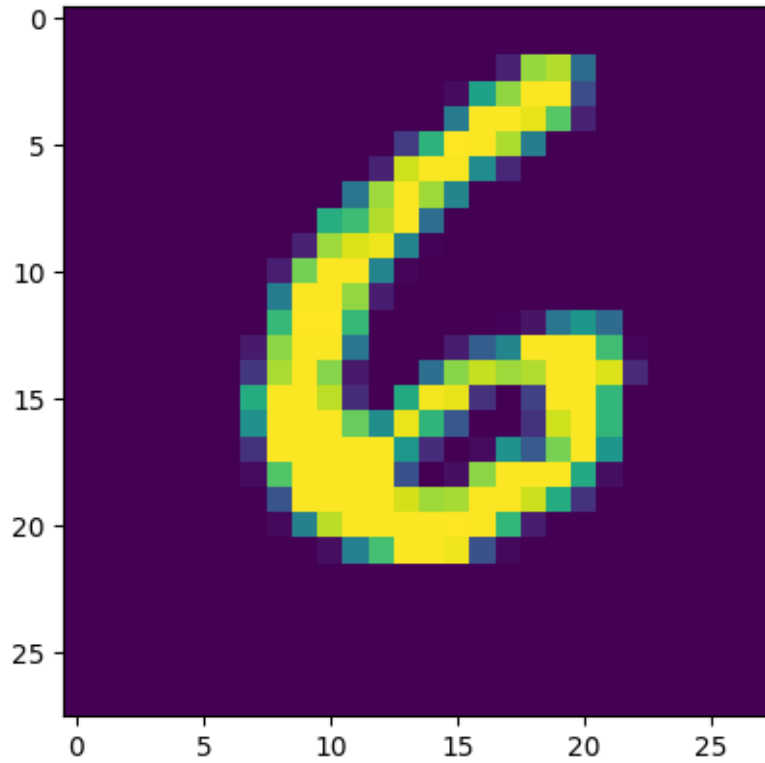
```
[5]: (10000, 28, 28)
```

```
[6]: y_test.shape
```

```
[6]: (10000,)
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```
[7]: import matplotlib.pyplot as plt
      plt.imshow(X_train[90])
```

```
[7]: <matplotlib.image.AxesImage at 0x7b20bbb09960>
```



```
[8]: X_train= X_train.reshape(-1, 28, 28, 1).astype('float32') / 255.0
      X_test = X_test.reshape(-1, 28, 28, 1).astype('float32') / 255.0
```

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[9]: X_train[0]
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```
[10]: X_train, X_val = X_train[:-10000], X_train[-10000:]
      y_train, y_val = y_train[:-10000], y_train[-10000:]
```

```
[11]: X_train.shape
```

```
[11]: (50000, 28, 28, 1)
```

```
[12]: y_train
```

```
[12]: array([5, 0, 4, ..., 8, 4, 8], dtype=uint8)
```

```
[13]: # Build a simple CNN model with regularization
model = Sequential()

# Add layers to the model
model.add(Conv2D(32, (3, 3), activation='relu', input_shape=(28, 28, 1)))
model.add(MaxPooling2D((2, 2)))
model.add(Conv2D(64, (3, 3), activation='relu'))
model.add(MaxPooling2D((2, 2)))
model.add(Flatten())
model.add(Dense(64, activation='relu'))
model.add(Dense(10, activation='softmax'))
```

```
/usr/local/lib/python3.10/dist-
packages/keras/src/layers/convolutional/base_conv.py:107: UserWarning: Do not
pass an `input_shape`/`input_dim` argument to a layer. When using Sequential
models, prefer using an `Input(shape)` object as the first layer in the model
instead.
```

```
super().__init__(activity_regularizer=activity_regularizer, **kwargs)
```

```
[14]: model.summary()
```

Model: "sequential"

Layer (type)	Output Shape	
Param #		
conv2d (Conv2D)	(None, 26, 26, 32)	
↪ 320		
max_pooling2d (MaxPooling2D)	(None, 13, 13, 32)	
↪ 0		
conv2d_1 (Conv2D)	(None, 11, 11, 64)	
↪ 18,496		
max_pooling2d_1 (MaxPooling2D)	(None, 5, 5, 64)	
↪ 0		
flatten (Flatten)	(None, 1600)	
↪ 0		

```

dense (Dense)                                     (None, 64)
↳102,464

dense_1 (Dense)                                    (None, 10)
↳650

```

Total params: 121,930 (476.29 KB)

Trainable params: 121,930 (476.29 KB)

Non-trainable params: 0 (0.00 B)

```

[15]: optimizer = tensorflow.keras.optimizers.Adam(learning_rate=0.001)

[16]: model.compile(loss = "sparse_categorical_crossentropy",optimizer = optimizer,
↳metrics=['accuracy'])

[17]: history = model.fit(X_train, y_train, epochs=10, validation_data=(X_val, y_val))

```

```

Epoch 1/10
1563/1563          64s 40ms/step -
accuracy: 0.8957 - loss: 0.3627 - val_accuracy: 0.9790 - val_loss: 0.0723
Epoch 2/10
1563/1563          50s 32ms/step -
accuracy: 0.9832 - loss: 0.0552 - val_accuracy: 0.9859 - val_loss: 0.0495
Epoch 3/10
1563/1563          48s 31ms/step -
accuracy: 0.9891 - loss: 0.0360 - val_accuracy: 0.9889 - val_loss: 0.0381
Epoch 4/10
1563/1563          83s 31ms/step -
accuracy: 0.9916 - loss: 0.0247 - val_accuracy: 0.9896 - val_loss: 0.0384
Epoch 5/10
1563/1563          81s 31ms/step -
accuracy: 0.9936 - loss: 0.0190 - val_accuracy: 0.9894 - val_loss: 0.0389
Epoch 6/10
1563/1563          48s 31ms/step -
accuracy: 0.9956 - loss: 0.0135 - val_accuracy: 0.9880 - val_loss: 0.0468
Epoch 7/10
1563/1563          83s 31ms/step -
accuracy: 0.9960 - loss: 0.0108 - val_accuracy: 0.9885 - val_loss: 0.0464
Epoch 8/10
1563/1563          47s 30ms/step -
accuracy: 0.9974 - loss: 0.0081 - val_accuracy: 0.9894 - val_loss: 0.0459
Epoch 9/10
1563/1563          87s 33ms/step -

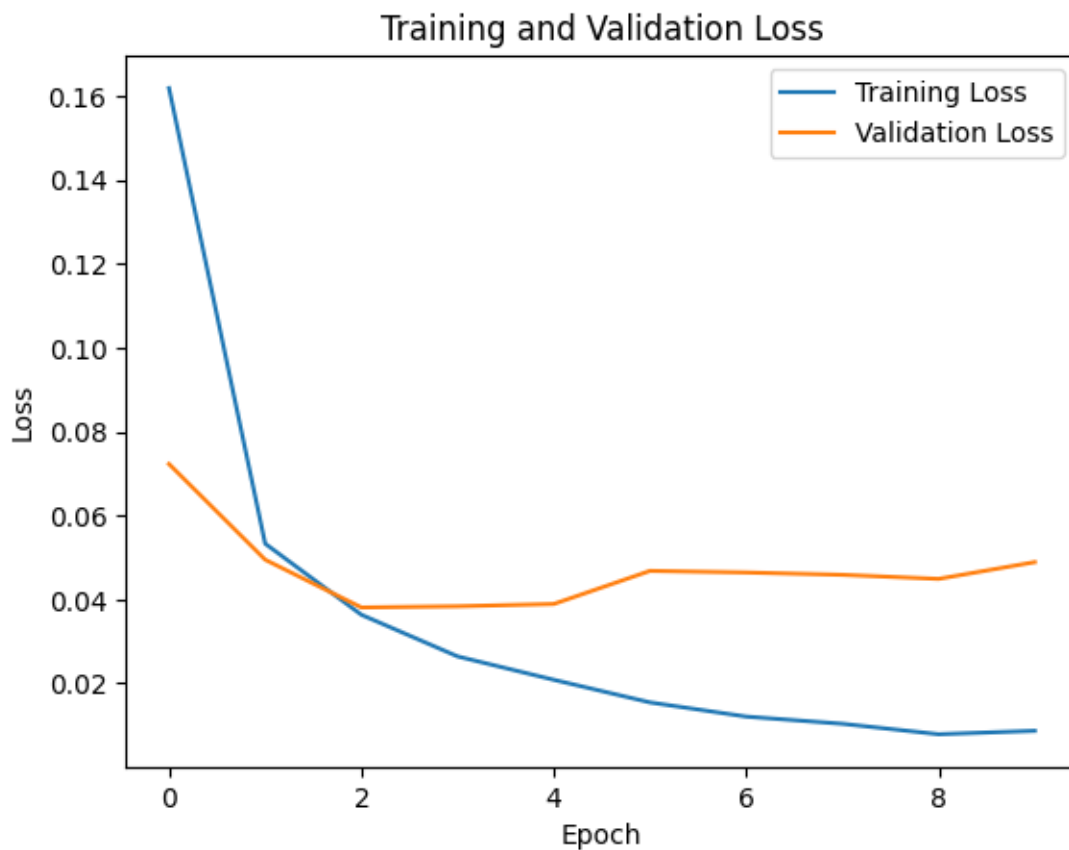
```

accuracy: 0.9972 - loss: 0.0077 - val_accuracy: 0.9888 - val_loss: 0.0449
Epoch 10/10
1563/1563 47s 30ms/step -
accuracy: 0.9976 - loss: 0.0071 - val_accuracy: 0.9894 - val_loss: 0.0489

```
[18]: #history = model.fit(X_train, y_train, epochs=10, validation_data=(X_val, y_val), batch_size=32)

# Plotting the training and validation loss
import matplotlib.pyplot as plt

plt.plot(history.history['loss'], label='Training Loss')
plt.plot(history.history['val_loss'], label='Validation Loss')
plt.title('Training and Validation Loss')
plt.xlabel('Epoch')
plt.ylabel('Loss')
plt.legend()
plt.show()
```



```
[19]: test_loss, test_accuracy = model.evaluate(X_test, y_test)
      print(f"Test Accuracy: {test_accuracy*100:.2f}%")
```

```
313/313          3s 9ms/step -
accuracy: 0.9895 - loss: 0.0402
Test Accuracy: 99.18%
```

Batch Normalization and its usage:

Code after applying drop out layer

```
[21]: X_train
```

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[21]: array([[[[0.],
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```

[26]: model_with_drop_out = Sequential()

      # Input and first convolutional layer
      model_with_drop_out.add(Conv2D(32, (3, 3), activation='relu', input_shape=(28, 28, 1)))
      model_with_drop_out.add(MaxPooling2D((2, 2)))

      model_with_drop_out.add(Dropout(0.25))

```

```

# Second convolutional layer
model_with_drop_out.add(Conv2D(64, (3, 3), activation='relu'))
model_with_drop_out.add(MaxPooling2D((2, 2)))

model_with_drop_out.add(Dropout(0.25))

model_with_drop_out.add(Flatten())

model_with_drop_out.add(Dense(64, activation='relu'))
model_with_drop_out.add(Dropout(0.5))

model_with_drop_out.add(Dense(10, activation='softmax'))

```

```

/usr/local/lib/python3.10/dist-
packages/keras/src/layers/convolutional/base_conv.py:107: UserWarning: Do not
pass an `input_shape`/`input_dim` argument to a layer. When using Sequential
models, prefer using an `Input(shape)` object as the first layer in the model
instead.

```

```

super().__init__(activity_regularizer=activity_regularizer, **kwargs)

```

```

[27]: optimizer = tensorflow.keras.optimizers.Adam(learning_rate=0.001)

```

```

[28]: model_with_drop_out.compile(loss = "sparse_categorical_crossentropy",optimizer_
↪ optimizer, metrics=['accuracy'])

```

```

[29]: history_with_dropout = model_with_drop_out.fit(X_train, y_train, epochs=10,
↪ validation_data=(X_val, y_val))

```

```

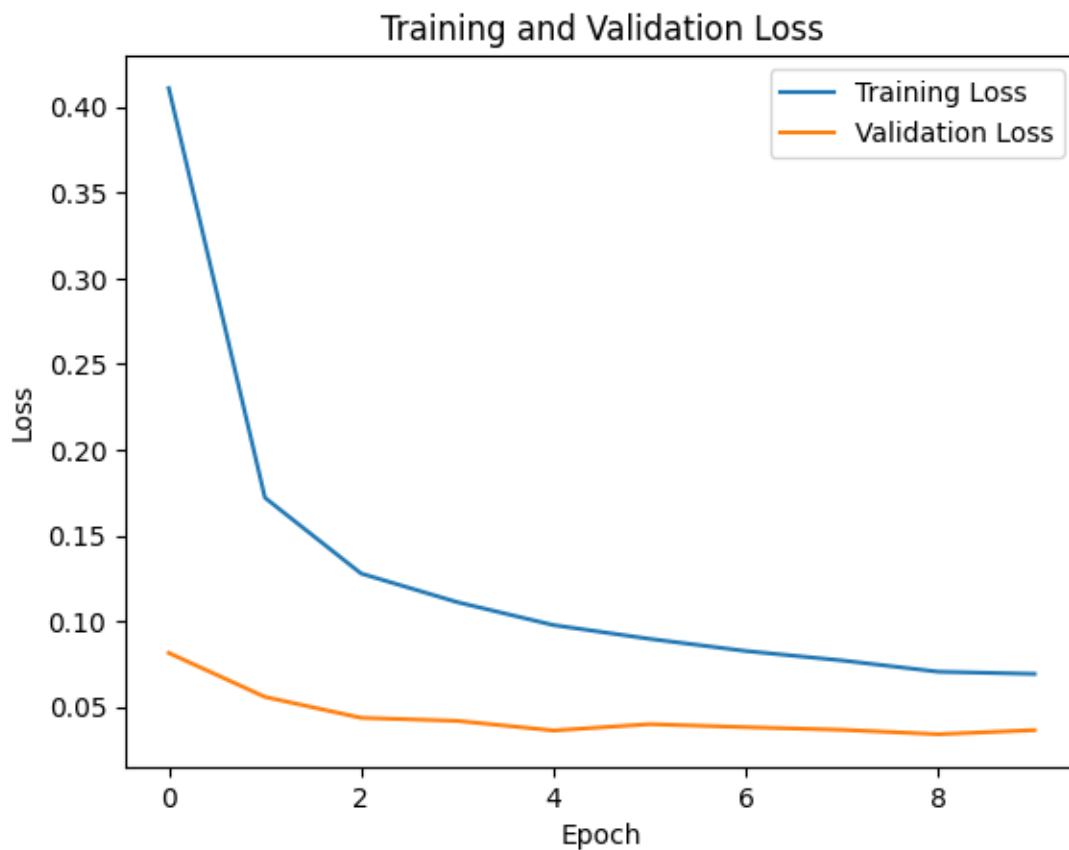
Epoch 1/10
1563/1563          64s 40ms/step -
accuracy: 0.7611 - loss: 0.7232 - val_accuracy: 0.9759 - val_loss: 0.0815
Epoch 2/10
1563/1563          55s 35ms/step -
accuracy: 0.9425 - loss: 0.1841 - val_accuracy: 0.9832 - val_loss: 0.0559
Epoch 3/10
1563/1563          82s 35ms/step -
accuracy: 0.9616 - loss: 0.1337 - val_accuracy: 0.9876 - val_loss: 0.0437
Epoch 4/10
1563/1563          82s 35ms/step -
accuracy: 0.9654 - loss: 0.1131 - val_accuracy: 0.9879 - val_loss: 0.0419
Epoch 5/10
1563/1563          82s 35ms/step -
accuracy: 0.9689 - loss: 0.0972 - val_accuracy: 0.9897 - val_loss: 0.0363
Epoch 6/10
1563/1563          54s 35ms/step -
accuracy: 0.9735 - loss: 0.0851 - val_accuracy: 0.9897 - val_loss: 0.0400

```

Epoch 7/10
1563/1563 54s 35ms/step -
accuracy: 0.9747 - loss: 0.0823 - val_accuracy: 0.9902 - val_loss: 0.0382
Epoch 8/10
1563/1563 54s 35ms/step -
accuracy: 0.9767 - loss: 0.0777 - val_accuracy: 0.9905 - val_loss: 0.0367
Epoch 9/10
1563/1563 82s 35ms/step -
accuracy: 0.9792 - loss: 0.0692 - val_accuracy: 0.9914 - val_loss: 0.0341
Epoch 10/10
1563/1563 54s 35ms/step -
accuracy: 0.9799 - loss: 0.0671 - val_accuracy: 0.9902 - val_loss: 0.0365

```
[32]: import matplotlib.pyplot as plt

plt.plot(history_with_dropout.history['loss'], label='Training Loss')
plt.plot(history_with_dropout.history['val_loss'], label='Validation Loss')
plt.title('Training and Validation Loss with drop out layer')
plt.xlabel('Epoch')
plt.ylabel('Loss')
plt.legend()
plt.show()
```



```
[33]: test_loss, test_accuracy = model_with_drop_out.evaluate(X_test, y_test)
      print(f"Test Accuracy: {test_accuracy*100:.2f}%")
```

```
313/313          3s 10ms/step -
accuracy: 0.9899 - loss: 0.0323
Test Accuracy: 99.17%
```

```
[37]: model_with_BN = Sequential()

      # Add layers to the model
      model_with_BN.add(Conv2D(32, (3, 3), activation='relu', input_shape=(28, 28, 1)))
      model_with_BN.add(BatchNormalization())
      model_with_BN.add(MaxPooling2D((2, 2)))
      model_with_BN.add(Conv2D(64, (3, 3), activation='relu'))
      model_with_BN.add(BatchNormalization())
      model_with_BN.add(MaxPooling2D((2, 2)))
      model_with_BN.add(Flatten())
      model_with_BN.add(Dense(64, activation='relu'))
      model_with_BN.add(BatchNormalization())
      model_with_BN.add(Dense(10, activation='softmax'))
```

```
[38]: optimizer = tensorflow.keras.optimizers.Adam(learning_rate=0.001)
```

```
[39]: model_with_BN.compile(loss = "sparse_categorical_crossentropy", optimizer = optimizer, metrics=['accuracy'])
```

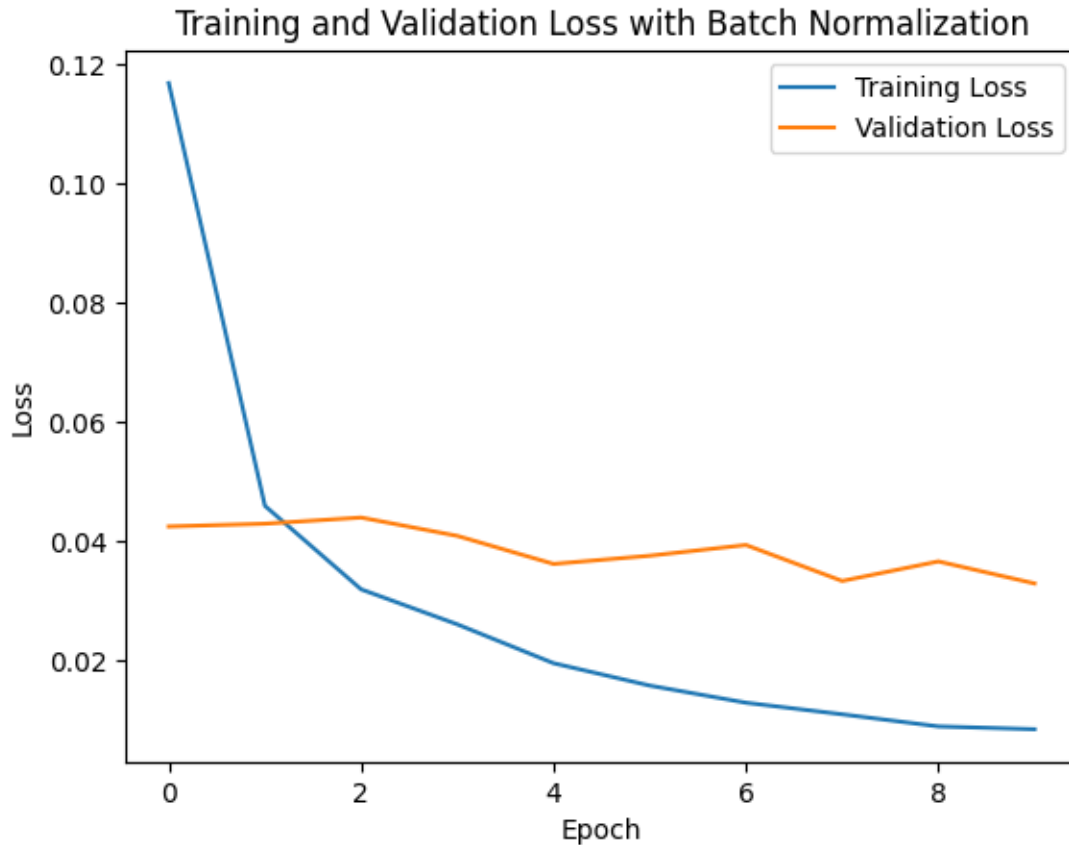
```
[40]: history_with_BN = model_with_BN.fit(X_train, y_train, epochs=10, validation_data=(X_val, y_val))
```

```
Epoch 1/10
1563/1563          91s 56ms/step -
accuracy: 0.9294 - loss: 0.2394 - val_accuracy: 0.9877 - val_loss: 0.0425
Epoch 2/10
1563/1563          70s 45ms/step -
accuracy: 0.9879 - loss: 0.0431 - val_accuracy: 0.9873 - val_loss: 0.0430
Epoch 3/10
1563/1563          81s 44ms/step -
accuracy: 0.9911 - loss: 0.0291 - val_accuracy: 0.9876 - val_loss: 0.0440
Epoch 4/10
1563/1563          82s 44ms/step -
accuracy: 0.9924 - loss: 0.0226 - val_accuracy: 0.9889 - val_loss: 0.0410
Epoch 5/10
1563/1563          83s 45ms/step -
accuracy: 0.9943 - loss: 0.0189 - val_accuracy: 0.9892 - val_loss: 0.0363
Epoch 6/10
```



```
1563/1563          82s 44ms/step -  
accuracy: 0.9952 - loss: 0.0147 - val_accuracy: 0.9894 - val_loss: 0.0376  
Epoch 7/10  
1563/1563          82s 44ms/step -  
accuracy: 0.9964 - loss: 0.0114 - val_accuracy: 0.9902 - val_loss: 0.0394  
Epoch 8/10  
1563/1563          72s 46ms/step -  
accuracy: 0.9972 - loss: 0.0095 - val_accuracy: 0.9916 - val_loss: 0.0334  
Epoch 9/10  
1563/1563          79s 44ms/step -  
accuracy: 0.9975 - loss: 0.0079 - val_accuracy: 0.9902 - val_loss: 0.0367  
Epoch 10/10  
1563/1563          82s 44ms/step -  
accuracy: 0.9980 - loss: 0.0064 - val_accuracy: 0.9908 - val_loss: 0.0330
```

```
[41]: import matplotlib.pyplot as plt  
  
plt.plot(history_with_BN.history['loss'], label='Training Loss')  
plt.plot(history_with_BN.history['val_loss'], label='Validation Loss')  
plt.title('Training and Validation Loss with Batch Normalization')  
plt.xlabel('Epoch')  
plt.ylabel('Loss')  
plt.legend()  
plt.show()
```



```
[42]: test_loss, test_accuracy = model_with_BN.evaluate(X_test, y_test)
      print(f"Test Accuracy: {test_accuracy*100:.2f}%")
```

```
313/313          4s 12ms/step -
accuracy: 0.9893 - loss: 0.0361
Test Accuracy: 99.10%
```

1 Definition

Batch Normalization and its usage:

Batch normalization is a technique which makes deep learning training more faster and stable. It consists of normalizing activation vectors from hidden layers using the mean and variance of the current batch. This is applied right before or after the nonlinear function

Drop out layer and its usage:

Dropout layer is a technique used to reduce overfitting of the model by dropping of some neurons randomly from hidden layer to avoid overfitting

2 Inferences:

I have trained the model without dropout layer, later with dropout layer and batch normalization layer.

observations:

- 1) Adding drop out layer doesnot make any significant change in the output and test accuracy is almost same i.e 99.17. Additionallly training loss is initially higher in model with drop out compared with model without drop out layer

overfitting is not a concern, as test accuracy is same in both cases

- 2) Model trained with Batch Normalization significantly improves the model's performance by normalizing the inputs at each layer, which helps in stabilizing the training process. This is evident from the consistently lower training and validation losses compared to models without it. Such normalization reduces drastic shifts in input distribution—a common problem in deep networks—making training faster and more reliable. The close tracking of training and validation losses indicates that the model generalizes well and doesn't overfit, making Batch Normalization a highly effective strategy for enhancing neural network performance.

Model requires batch processing layer and doesnt require drop out based on the observations

[]:

[42]: