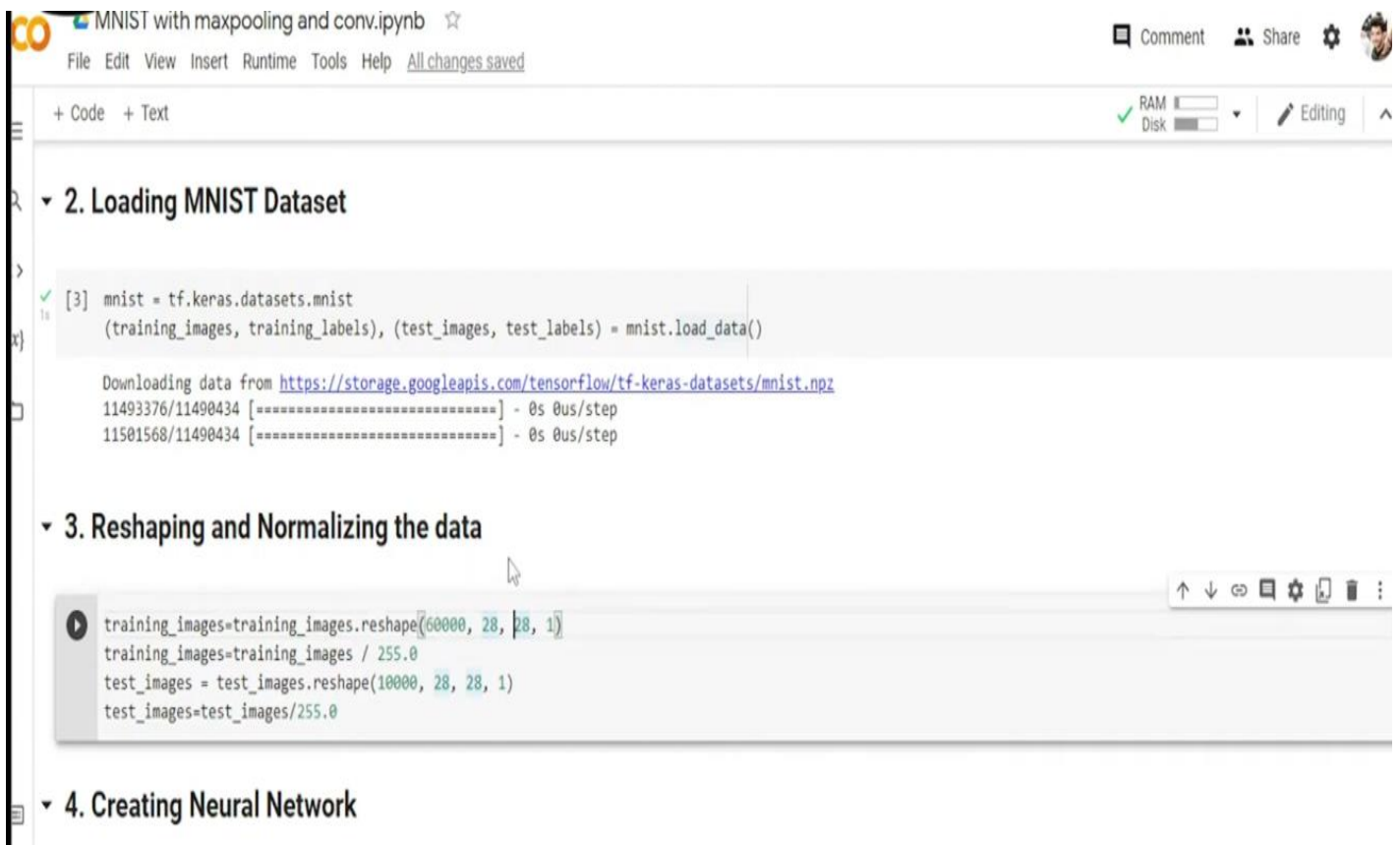
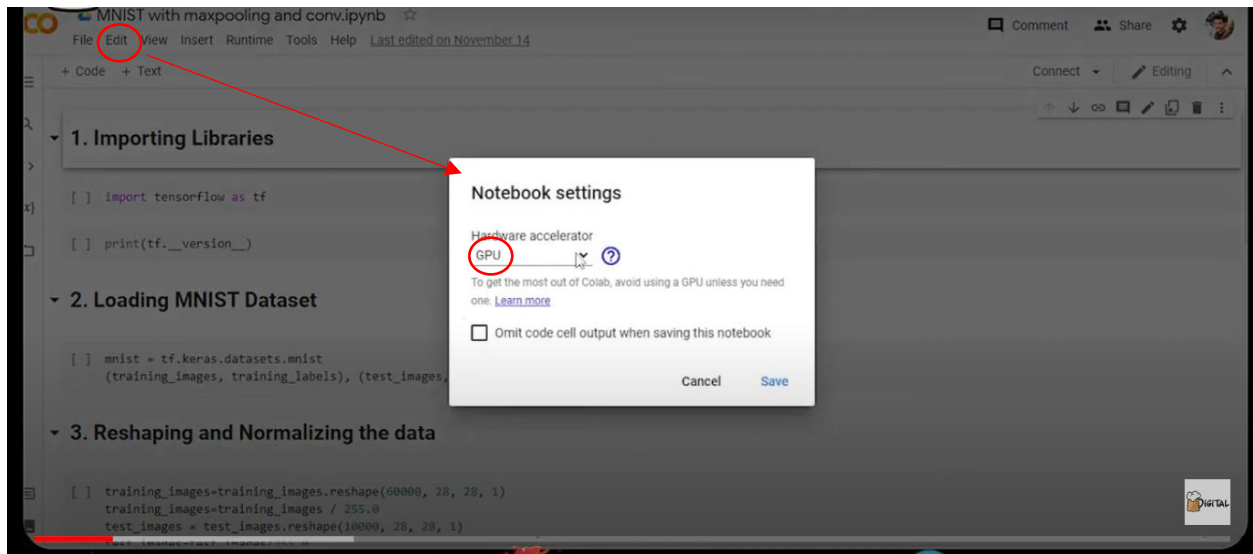


<https://www.youtube.com/watch?v=WX4pHOb9P-w>



MNIST with maxpooling and conv.ipynb ☆

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```
[4] test_images = test_images.reshape(10000, 28, 28, 1)
test_images=test_images/255.0
```

4. Creating Neural Network

```
model = tf.keras.models.Sequential([
    tf.keras.layers.Conv2D(32, (3,3), activation='relu', input_shape=(28, 28, 1)),
    tf.keras.layers.MaxPooling2D(2, 2),
    tf.keras.layers.Flatten(),
    tf.keras.layers.Dense(128, activation='relu'),
    tf.keras.layers.Dense(10, activation='softmax')
])
```

Activation function = relu

2D = two dimension

32=Filter

3x3 = matrix

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

```
[ ] model.compile(optimizer='adam', loss='sparse_categorical_crossentropy', metrics=['accuracy'])
```

5. Training and evaluating the data

MNIST with maxpooling and conv.ipynb ☆

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```
tf.keras.layers.Dense(128, activation='relu'),
[5] tf.keras.layers.Dense(10, activation='softmax')
])
```

```
[6] model.summary() #Check summary
```

```
Model: "sequential"
```

Layer (type)	Output Shape	Param #
conv2d (Conv2D)	(None, 26, 26, 32)	320
max_pooling2d (MaxPooling2D)	(None, 13, 13, 32)	0
flatten (Flatten)	(None, 5408)	0
dense (Dense)	(None, 128)	692352
dense_1 (Dense)	(None, 10)	1290

```

Total params: 693,962
Trainable params: 693,962
Non-trainable params: 0

```

DIGITAL

MNIST with maxpooling and conv.ipynb

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```

model = tf.keras.models.Sequential([
    tf.keras.layers.Conv2D(32, (3,3), activation='relu', input_shape=(28, 28, 1)),
    tf.keras.layers.MaxPooling2D(2, 2),
    tf.keras.layers.Flatten(),
    tf.keras.layers.Dense(128, activation='relu'),
    tf.keras.layers.Dense(10, activation='softmax')
])

```

[6] model.summary()

Model: "sequential"

Layer (type)	Output Shape	Param #
conv2d (Conv2D)	(None, 26, 26, 32)	320
max_pooling2d (MaxPooling2D)	(None, 13, 13, 32)	0
flatten (Flatten)	(None, 5408)	0
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dense_1 (Dense)	(None, 10)	1290

Total params: 693,962
Trainable params: 693,962
Non-trainable params: 0

```

[ ] model.compile(optimizer='adam', loss='sparse_categorical_crossentropy', metrics=['accuracy'])

```

5. Training and evaluating the data

```

[ ] model.fit(training_images, training_labels, epochs=10)

```

The diagram illustrates a 3x3 convolution operation on a handwritten digit '0'. The kernel is applied to the input image, resulting in a 6x6 grid of values. The values in the grid are: 170, 238, 85, 255, 221, 0; 68, 136, 17, 170, 119, 68; 221, 0, 238, 136, 0, 255; 119, 255, 85, 170, 136, 238; 238, 17, 221, 68, 119, 255; 85, 170, 119, 221, 17, 136.

MNIST with maxpooling and conv.ipynb

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```

conv2d (Conv2D) (None, 26, 26, 32) 320
max_pooling2d (MaxPooling2D) (None, 13, 13, 32) 0
flatten (Flatten) (None, 5408) 0
dense (Dense) (None, 128) 692352
dense_1 (Dense) (None, 10) 1290

```

Total params: 693,962
Trainable params: 693,962
Non-trainable params: 0

```

model.compile(optimizer='adam', loss='sparse_categorical_crossentropy', metrics=['accuracy'])

```

Can use with the help of "call back", it is best to use call back

Compiling of created model

5. Training and evaluating the data

```

model.fit(training_images, training_labels, epochs=10)

```

```

Epoch 1/10
1875/1875 [-----] - 39s 5ms/step - loss: 0.1427 - accuracy: 0.9583
Epoch 2/10
1875/1875 [-----] - 9s 5ms/step - loss: 0.0499 - accuracy: 0.9845
Epoch 3/10
1875/1875 [-----] - 9s 5ms/step - loss: 0.0312 - accuracy: 0.9904
Epoch 4/10
1875/1875 [-----] - 9s 5ms/step - loss: 0.0207 - accuracy: 0.9932
Epoch 5/10
1875/1875 [-----] - 9s 5ms/step - loss: 0.0145 - accuracy: 0.9955
Epoch 6/10
1875/1875 [-----] - 9s 5ms/step - loss: 0.0101 - accuracy: 0.9967
Epoch 7/10
1875/1875 [-----] - 9s 5ms/step - loss: 0.0075 - accuracy: 0.9976
Epoch 8/10
1875/1875 [-----] - 9s 5ms/step - loss: 0.0070 - accuracy: 0.9976
Epoch 9/10
1875/1875 [-----] - 9s 5ms/step - loss: 0.0043 - accuracy: 0.9986
Epoch 10/10
1875/1875 [-----] - 9s 5ms/step - loss: 0.0039 - accuracy: 0.9988

```

```

test_loss, test_acc = model.evaluate(test_images, test_labels)
print(test_acc)

```

```
[8] Epoch 3/10
1875/1875 [=====] - 9s 5ms/step - loss: 0.0312 - accuracy: 0.9984
Epoch 4/10
1875/1875 [=====] - 9s 5ms/step - loss: 0.0207 - accuracy: 0.9932
Epoch 5/10
1875/1875 [=====] - 9s 5ms/step - loss: 0.0145 - accuracy: 0.9955
Epoch 6/10
1875/1875 [=====] - 9s 5ms/step - loss: 0.0101 - accuracy: 0.9967
Epoch 7/10
1875/1875 [=====] - 9s 5ms/step - loss: 0.0075 - accuracy: 0.9976
Epoch 8/10
1875/1875 [=====] - 9s 5ms/step - loss: 0.0070 - accuracy: 0.9976
Epoch 9/10
1875/1875 [=====] - 9s 5ms/step - loss: 0.0043 - accuracy: 0.9986
Epoch 10/10
1875/1875 [=====] - 9s 5ms/step - loss: 0.0039 - accuracy: 0.9988
<keras.callbacks.History at 0x7f39730fed50>

[9] test_loss, test_acc = model.evaluate(test_images, test_labels)
print(test_acc)

313/313 [=====] - 1s 4ms/step - loss: 0.0672 - accuracy: 0.9844
0.9843999743461609
```

