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Ex 5

### Task I

Padding size formula:  $\left(\frac{k-1}{2}\right)$  here is zero

Output size formula:  $\frac{\text{InputSize} - \text{kernelSize} + 2 \times \text{PaddingSize}}{\text{Stride}} + 1$

Number of multiply-add operations in CNL:  $\text{OutputSize} \times \text{OutputChannels} \times (\text{kernelSize}^2 \times \text{inputChannels} + 1)$

Number of parameters in CNL:  $\text{OutputChannels} \times (\text{kernelSize}^2 \times \text{inputChannels} + 1)$

Fully connected layers number of multiply-add operations:  $\text{OutputNeurons} \times (\text{inputneurons} + 1)$

Number of parameters in FCL:  $\text{OutputNeurons} \times (\text{inputNeurons} + 1)$

Layer details	input	inputSize	outputSize	#Parameters	#multiply-add
CNL: 5x5x32 - stride 1		47x47x3	43x43x32	2432	104576
CNL: 3x3x32 - S2		43x43x32	21x21x32	9248	194208
CNL: 3x3x32 - S1		21x21x32	19x19x32	9248	175712
CNL: 3x3x32 - S2		19x19x32	9x9x32	9248	83232
CNL: 3x3x32 - S1		9x9x32	7x7x32	9248	64736
Flattening		7x7x32	1568	0	0
FC - 128		1568	128	200832	200832
FC -		128	10	1290	1290
overall - - - - -				241546	824586

## Amir Hossein Eyvazkhani - 1747696

Ex 05

### Task 2

The coverage is checked in Collab. It reaches more than 92 percent accuracy and less than 0.22 loss in the first 20 epochs. In my local machine, I did not have any compute power for showing the output. Sorry about that...

```
In [ ]: import tensorflow as tf
        from tensorflow import keras
        from keras import layers
        import numpy as np
        import matplotlib.pyplot as plt

        class MyBatchNormalization(keras.layers.Layer):
            def __init__(
                self,
                epsilon=1e-5
            ):
                super().__init__()
                self.epsilon = epsilon

            def build(self, input_shape):
                param_shape = (input_shape[-1],)
                self.gamma = self.add_weight(
                    name="gamma",
                    shape=param_shape,
                    dtype='float32',
                    trainable=True,
                )
                self.beta = self.add_weight(
                    name="beta",
                    shape=param_shape,
                    dtype='float32',
                    trainable=True,
                )
                self.built = True
```

```

def call(self, inputs):
    batch_mean = tf.reduce_mean(inputs, axis=0)
    batch_variance = tf.math.reduce_std(inputs, axis=0)
    normalized_inputs = (inputs - batch_mean) / tf.sqrt(batch_variance + self.epsilon)
    outputs = self.gamma * normalized_inputs + self.beta
    return outputs

```

*# for the last part of the questuion*

```

class MyBatchNormalization_joint_version(keras.layers.Layer):

```

```

    def __init__(
        self,
        epsilon=1e-5
    ):
        super().__init__()
        self.epsilon = epsilon

```

```

    def build(self, input_shape):
        self.gamma = self.add_weight(
            name="gamma",
            shape=(1,),
            dtype='float32',
            trainable=True,
        )
        self.beta = self.add_weight(
            name="beta",
            shape=(1,),
            dtype='float32',
            trainable=True,
        )
        self.built = True

```

```

    def call(self, inputs):
        batch_mean = tf.reduce_mean(inputs)
        batch_variance = tf.math.reduce_std(inputs)
        normalized_inputs = (inputs - batch_mean) / tf.sqrt(batch_variance + self.epsilon)
        outputs = self.gamma * normalized_inputs + self.beta
        return outputs

```

```

def define_model():

```

```

inputs = keras.Input(shape=(28,28,1))

K = 20 # number of convolution layers per block
L = 3 # number of blocks
x = inputs
for i in range(0,L):
    for j in range(0,K):
        x = layers.Conv2D(32, 3, activation="relu",padding="same")(x)
        x = MyBatchNormalization()(x)
        # for the second part we should use this version of BatchNormalization
        # x = MyBatchNormalization_joint_version()(x)
    x = layers.MaxPooling2D(3)(x)
x = layers.GlobalMaxPooling2D()(x)
outputs = layers.Dense(10,activation='softmax')(x)
model = keras.Model(inputs,outputs)
# model.summary()
return model

```

```

In [ ]: # Load and preprocess training data (Fashion-MNIST)
(train_images, train_labels), (test_images, test_labels) = tf.keras.datasets.fashion_mnist.load_data()
train_images = train_images / 255.0
test_images = test_images / 255.0
train_labels = tf.keras.utils.to_categorical(train_labels)
test_labels = tf.keras.utils.to_categorical(test_labels)

# Define and train model
model = define_model()
model.compile(loss=keras.losses.CategoricalCrossentropy(),optimizer=keras.optimizers.Adam(),metrics=["accuracy"])
model.fit(train_images,train_labels, batch_size=64, epochs=100)

```

## Task 3

```

In [ ]: import tensorflow as tf
from tensorflow import keras
from keras import layers
import numpy as np
import matplotlib.pyplot as plt

def define_model():

```

```

# I define the implementation for the model manually here
# in other words, it is not generalized with K, M, and L
# and that is because I could not figure it out when we needed spatial sampling
# also it was mentioned in the question to consider the graph for the implementation
# the name of the layer has the following format: [LAYER_TYPE]_[L]_[K]
# and each M = 2 layer a residual connection is added
inputs = keras.Input(shape=(28,28,1))
conv2d_0_0 = layers.Conv2D(32, 3, activation="relu", padding="same")(inputs)
conv2d_0_1 = layers.Conv2D(32, 3, activation="relu", padding="same")(conv2d_0_0)
matchChannel0 = layers.Conv2D(32, 1, padding="same")(inputs)
add0 = layers.add([conv2d_0_1, matchChannel0])
conv2d_0_2 = layers.Conv2D(32, 3, activation="relu", padding="same")(add0)
conv2d_0_3 = layers.Conv2D(32, 3, activation="relu", padding="same")(conv2d_0_2)
add1 = layers.add([conv2d_0_3, add0])
maxPooling2D_1 = layers.MaxPooling2D(pool_size=(3,3), padding='valid', strides=(1,1))(add1)
conv2d_1_0 = layers.Conv2D(32, 3, activation="relu", padding="same")(maxPooling2D_1)
conv2d_1_1 = layers.Conv2D(32, 3, activation="relu", padding="same")(conv2d_1_0)
averagePooling2D_1 = layers.MaxPooling2D(pool_size=(3,3), padding='valid', strides=(1,1))(add1)
add2 = layers.add([conv2d_1_1, averagePooling2D_1])
conv2d_1_2 = layers.Conv2D(32, 3, activation="relu", padding="same")(add2)
conv2d_1_3 = layers.Conv2D(32, 3, activation="relu", padding="same")(conv2d_1_2)
add3 = layers.add([conv2d_1_3, add2])
maxPooling2D_2 = layers.MaxPooling2D(pool_size=(3,3), padding='valid', strides=(1,1))(add3)
conv2d_2_0 = layers.Conv2D(32, 3, activation="relu", padding="same")(maxPooling2D_2)
conv2d_2_1 = layers.Conv2D(32, 3, activation="relu", padding="same")(conv2d_2_0)
averagePooling2D_2 = layers.MaxPooling2D(pool_size=(3,3), padding='valid', strides=(1,1))(add3)
add4 = layers.add([conv2d_2_1, averagePooling2D_2])
conv2d_2_2 = layers.Conv2D(32, 3, activation="relu", padding="same")(add4)
conv2d_2_3 = layers.Conv2D(32, 3, activation="relu", padding="same")(conv2d_2_2)
add5 = layers.add([conv2d_2_3, add4])
falttening = layers.GlobalMaxPooling2D()(add5)
outputs = layers.Dense(10,activation='softmax')(falttening)
model = keras.Model(inputs,outputs)
# model.summary()
return model

```

```

# Load and preprocess training data (Fashion-MNIST)
(train_images, train_labels), (test_images, test_labels) = tf.keras.datasets.fashion_mnist.load_data()
train_images = train_images / 255.0
test_images = test_images / 255.0
train_labels = tf.keras.utils.to_categorical(train_labels)
test_labels = tf.keras.utils.to_categorical(test_labels)

```

```
# Define and train model
model = define_model()
model.compile(loss=keras.losses.CategoricalCrossentropy(),optimizer=keras.optimizers.Adam(),metrics=["accuracy"])
model.fit(train_images,train_labels, batch_size=64, epochs=5) # just tested with 5 epochs because of the limited comp
```

Epoch 1/5

WARNING:tensorflow:From h:\Uni\WiSe 2024\ML LAB\ml\_lab\_venv\Lib\site-packages\keras\src\utils\tf\_utils.py:492: The name tf.ragged.RaggedTensorValue is deprecated. Please use tf.compat.v1.ragged.RaggedTensorValue instead.

WARNING:tensorflow:From h:\Uni\WiSe 2024\ML LAB\ml\_lab\_venv\Lib\site-packages\keras\src\engine\base\_layer\_utils.py:384: The name tf.executing\_eagerly\_outside\_functions is deprecated. Please use tf.compat.v1.executing\_eagerly\_outside\_functions instead.

938/938 [=====] - 182s 187ms/step - loss: 0.6447 - accuracy: 0.7724

Epoch 2/5

938/938 [=====] - 172s 183ms/step - loss: 0.3844 - accuracy: 0.8622

Epoch 3/5

938/938 [=====] - 180s 192ms/step - loss: 0.3252 - accuracy: 0.8834

Epoch 4/5

938/938 [=====] - 175s 187ms/step - loss: 0.2957 - accuracy: 0.8949

Epoch 5/5

938/938 [=====] - 173s 185ms/step - loss: 0.2743 - accuracy: 0.9021

Out[ ]: <keras.src.callbacks.History at 0x204c6bb65d0>

It reaches suitable accuracy and loss without batchnorm.

In [ ]: