COEN 166L: Assignment #6

- 1. Python Practice
- 2. Vacuum World
- 3. Pac-man I
- 4. Pac-man II
- 5. Face Recognition
- 6. Neural Network

No Demo;

Report & code due by Wed,12/11

Overview

Problem 1- Recognition

Build a CNN network for recognition with the fashion MNIST dataset

- 1. **Epochs = 5.**
- 2. Error function is **cross-entropy**
- Give the recognition accuracy rate and show the confusion matrix, both for the test set.

Problem 2 - Image Compression

Implement an image compression system using a neural network.

- 1. Epochs=10, Batch size=64.
- 2. The loss/error function is the **mean-squared-error (mse)**
- 3. Train with different P values, P = 10, 50, 200
- 4. Calculate PSNR
- 5. Display the images compressed and without compressed

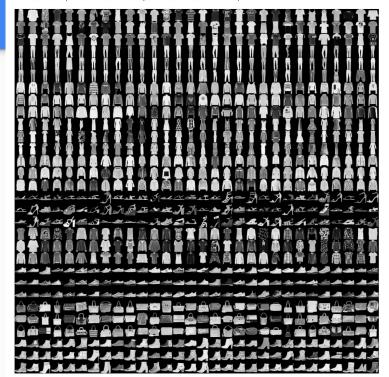
Fashion-MNIST

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Fashion-MNIST is a dataset of Zalando's article images—consisting of a training set of 60,000 examples and a test set of 10,000 examples. Each example is a 28x28 grayscale image, associated with a label from 10 classes. We intend Fashion-MNIST to serve as a direct drop-in replacement for the original MNIST dataset for benchmarking machine learning algorithms. It shares the same image size and structure of training and testing splits.

Here's an example how the data looks (each class takes three-rows):



https://github.com/zalandoresearch/fashion-mnist

Pre-work

Packages needed:

- Tensorflow (Keras)
- Numpy
- Sklearn
- matplotlib

Implementation Problem 1: Overall Structure

```
t tensorflow as tf
  om tensorflow.keras import datasets, layers, models
 rom sklearn.metrics import confusion_matrix
fashion_mnist = keras.datasets.fashion_mnist
(train_images, train_labels), (test_images, test_labels) = fashion_mnist.load_data()
train_images = train_images.reshape((60000, 28, 28, 1))
test_images = test_images.reshape((10000, 28, 28, 1))
# Normalize pixel values to be between 0 and 1
train images, test images = train images / 255.0, test images / 255.0
####-----decide the model layers-----
model = models.Sequential()
model.add(layers.Conv2D(32, (3, 3), activation='relu', input_shape=(28, 28, 1)) # the 1st 2d-convolutional layer
# model.add()
# model.add()
# model.add()
# ... to be completed by yourself
#####---- compile the model -----
model.compile()
### ---- evaluate the model -----
model.evaluate()
#####---- train the model -----
model.fit()
#####---- predict with the model -----
model.predict()
confusion matrix()
```

Adopt the following convolutional neural network structure:

- 1. Input layer
- 2. 2d-convolutional layer: filter size 3x3, depth=32, ReLU activation function
- 3. 2x2 max pooling layer
- 4. 2d-convolutional layer: filter size 3x3, depth=64, ReLU activation function
- 5. 2x2 max pooling layer
- 6. 2d-convolutional layer: filter size 3x3, depth=64, ReLU activation function
- 7. Fully-connected layer: 64 units, ReLU activation function
- 8. (output) Fully-connected layer: 10 units, softmax activation function

(Refer to Lab assignment 6.pdf)

How to construct a neural network (CNN)?

mnist_cnn.py

```
(Camino -> lecture ->camino->lecture session-> lecture slides folder)
model = model.Sequential()
model.add()
model.add()
model.add()
model.summary()
model.compile(optimizer = ' ', loss=' ', metrics =' ')
model.fit( train_images, train_labels, epochs = 5 ,batch_size= 64)
loss, acc = model.evaluate(test_images, test_labels)
v_test_hat = model.predict (test_images)
```

Refer to 'mnist_cnn.py' or mnist_nn.py' about how to

construct a CNN or neural network.

```
import tensorflow as tf
from tensorflow.keras import datasets, layers, models
 import numpy as np
(train images, train labels), (test images, test labels) = datasets.mnist.load data()
train images = train images.reshape((60000, 28, 28, 1))
test_images = test_images.reshape((10000, 28, 28, 1))
# Normalize pixel values to be between 0 and 1
train_images, test_images = train_images / 255.0, test_images / 255.0
model = models.Sequential()
# 32: number of filters
model.add(layers.Conv2D(32, (3, 3), activation='relu', input_shape=(28, 28, 1)))
model.add(layers.MaxPooling2D((2, 2)))
model.add(layers.Conv2D(64, (3, 3), activation='relu'))
model.add(layers.MaxPooling2D((2, 2)))
model.add(layers.Conv2D(64, (3, 3), activation='relu'))
model.add(layers.Flatten())
model.add(layers.Dense(64, activation='relu'))
model.add(layers.Dense(64, activation='relu'))
model.add(layers.Dense(10, activation='softmax'))
model.summary()
model.compile(optimizer='adam',
              loss='sparse categorical crossentropy',
              metrics=['accuracy'])
model.fit(train_images, train_labels, epochs=5,batch_size=64)
test_loss, test_acc = model.evaluate(test_images, test_labels)
print(test acc)
y_test_hat_mat = model.predict(test_images)
y test hat = np.argmax(y test hat mat, axis=1)
 from sklearn.metrics import confusion matrix
cm=confusion_matrix(test_labels, y_test_hat, labels=range(10))
```

You may need to (probably not in this lab):

1.

If download using fashion_mnist.load_data() is too slow, try to download manually from the website, but you need to write some code to load the downloaded .gz data file

Get the Data

Many ML libraries already include Fashion-MNIST data/API, give it a try!

You can use direct links to download the dataset. The data is stored in the same format as the original MNIST data.

Name	Content	Examples	Size	Link	MD5 Checksum
train-images- idx3-ubyte.gz	training set images	60,000	26 MBytes	Download	8d4fb7e6c68d591d4c3dfef9ec88bf0d
train-labels- idx1-ubyte.gz	training set labels	60,000	29 KBytes	Download	25c81989df183df01b3e8a0aad5dffbe
t10k-images- idx3-ubyte.gz	test set images	10,000	4.3 MBytes	Download	bef4ecab320f06d8554ea6380940ec79
t10k-labels- idx1-ubyte.gz	test set labels	10,000	5.1 KBytes	Download	bb300cfdad3c16e7a12a480ee83cd310

Alternatively, you can clone this GitHub repository; the dataset appears under data/fashion . This repo also contains some scripts for benchmark and visualization.

2.

Save the model : model.save

Load the model for the next time usage: model.load

Can also serialize to other format:

- JSON HD5
- YAML HD5

https://machinelearningmastery.com/save-load-keras-deep-learning-models/

https://github.com/zalandoresearch/fashion-mnist

Implementation Problem 2: Overall Structure

The network includes <u>compressed</u> + <u>decompressed</u>, so the workflow is:

Original image - > network(first compress, then decompress) -> output decompressed image

Overall structure is similar to Problem 1

But some differences:

- 1. model.compile()
- 2. model.fit() # input and output are both train images
- 3. model.predict() # in problem 2: predicted output is decompressed images
 (in problem 1: prediction output is probability of each class)
- 4. need to calculate PSNR between test_images and test_decompressed_images