Introduction to Deep Learning

TP R 3 - Convolutional Neural Networks 5th year Statistics and Data Science

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all data of this TP is available in this link:

https://drive.google.com/drive/folders/1ls07T82766Fg1tpsi_bB9zmvfdLSRSbf

1. The concept of convolution

using magick

- Q1. read the museum picture provided using magick tools
- Q2. create a horizontal filter and flip it to become a vertical filter
- Q3. apply these two filters to the image
- **Q4.** try other filters:

```
matrix1 <- matrix(c(0,1,1,0,1,1,0,1,1), nrow = 3, ncol = 3)
edge = matrix(c(0,0,0,0,0,0,1,1,1), 3, 3)
sharpen = matrix(c(0,-1,0, -1,6,-1, 0,-1,0),3, 3)
embos = matrix(c(-2,0,0, 0,1,0, 0,0,2),3, 3)
edge2 = matrix(c(-1,-1,-1, -1,8,-1, -1,-1,-1),3, 3)
edge3 = matrix(c(0,1,0, 1,-4,1, 0,1,0), 3, 3)</pre>
```

using a predefined function for images with one channel

Q5. complete this function to create a convolution operation function

```
# Define a 2D convolution function
convolve2d <- function(image, kernel) {
    # Get dimensions of image and kernel

# Output image placeholder
output <- matrix(0, nrow = image_height, ncol = image_width)

# Calculate padding</pre>
```

```
# Apply convolution

}

return(output)
}
```

Q6. read the image using this code

```
image <- readImage(system.file("images", "sample-color.png", package = "EBImage"))</pre>
```

- **Q7.** display the image
- Q8. define the same filters as part 1 plus a gaussian blur filter
- **Q9.** apply all these filters to the provided image
- Q10. (advanced) try to modify this function to deal with images with 3 channel

2. MNIST with convolution

Use a convolutional model (of your choice) to estimate the numbers of MNIST dataset.

3. CNN for time series:

toy example

- Q1. Load the libraries
- Q2. Simulate a simple sine wave as time series data (use a seed of 123 to insure the reproduction of the data)
- ${f Q3.}$ define a function to prepare the time series for a CNN model. The function need to have a parameter : look_back
- Q4. use a look back to 10 and prepare the data
- Q5. use a 80% rows of the data as training set
- Q6. define a CNN model and compile it
- **Q7.** fit the model by using 20% as a validation
- **Q8.** predict the model on testing set
- **Q9.** plot the predicted vs actual data

real data

- $\mathbf{Q10.}$ read the data set "2014-12.csv" choose the variable \mathbf{INDPRO} . Do the necessary transformations and answer all the previous questions
- Q11. try other time series

4. Data Augmentaion in images

The goal of this exercise is to predict if a picture contains a cat or a dog using the data set provided "cats_and_dogs_filtered.zip"

- Q1. How many images are there in the train and test (validation) folder
- **Q2.** plot some random images of cats and dogs (you can use the EBImage package)
- Q3. define a CNN model compile it and summarize it
- Q4. normalize the data by multiplying by a rescaling factor
- Q5. flow the training images using a 20 images batches
- Q6. do the same for the testing set images
- **Q7.** fit the model using the information from Q4-Q6.
- **Q8.** evaluate the model. Is the model performant?

To increase the performance of the model, we will use data augmentation technique

- Q9. define data augmentation by transforming images using rotation, shifting, shearing, zooming,...
- Q10. take a sample image and apply these transformations to it. Plot the results
- Q11. define a CNN model and fit it on the augmented data
- Q12. is there any improvement?
- Q13. use a CNN model including dropout
- Q14. comment on the use of data augmentation

5. Cifar dataset

- Q1. load the dataset CIFAR (if you don't have it you can use Fashion MNIST)
- **Q2.** what are the dimension of images in the dataset
- Q3. plot some random images of the dataset
- Q4. get the training and testing data into right formatting format
- Q5. define an appropriate CNN model
- **Q6.** compile the model using predefined learning rate = 0.0001 and a weight decay 1e-6
- Q7. fit the model on data and use a validation set
- **Q8.** predict the data on the testing set
- **Q9.** refit the model using an appropriate format of data augmentation
- Q10. compare the results.

6. Advanced ideas: Convolution ships

Read the data set containing ships images (defined in the JSON) or read it directly.

link to the dataset;

https://drive.google.com/drive/folders/1ls07T82766Fg1tpsi bB9zmvfdLSRSbf

Create a CNN model to predict if the image contain a ship or not.

7. Advanced ideas: Exploring convolutional layers

- Q1. read the MNIST data set and create the training and testing images
- **Q2.** reshape the training and testing images :
 - convert it to a 4-dimensional array (1 in the 4th array)
 - convert the grey scale to 0-1 scale
- Q3. visualize some random images from the test set. (for example image 35)
- Q4. create a CNN with 2 convolutional layers and a two layers MLP
- Q5. compile the model
- Q6. using the compiled model get all the outputs the layers of the convolutional part of the model
- Q7. create a model for this convolutional part only (definition input and output)
- **Q8.** predict this model on the image from question 3
- **Q9.** get the result of each layer separately.
- Q10. (many intermediate steps) plot all the channels for each layer.
- $\mathbf{Q11.}$ repeat question 10 with the image of the museum from exercise 1