Lab 5 Exercise - A little Linear Regression

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This is the exercise that you need to work through on your own after completing the fifth lab session. You’ll need to write up your results/answers/findings and submit this to ECS handin as a PDF document along with the other lab exercises near the end of the module (1 pdf document per lab).

We expect that you will use no more than one side of A4 to cover your responses to this exercise. This exercise is worth 5% of your overall module grade.

1 An initial attempt

In the lab exercise you built and trained a couple of CNNs for image classification. Your now going to try something different and build some CNNs for image-to-vector regression task - in particular you’re going to implement a network that takes an image of a scatter plot, and predicts the parameters of the line of best fit.

The following code is used to generate the datasets for this task (get it in a useable form here: https://gist.github.com/jonhare/73a59dcc5416729548a086a983e81f07):

```python
import torch
from torchvision import transforms
from torch.utils.data import Dataset

class MyDataset(Dataset):
    def __init__(self, size=5000, dim=40, random_offset=0):
        super(MyDataset, self).__init__()
        self.size = size
        self.dim = dim
        self.random_offset = random_offset

    def __getitem__(self, index):
        if index >= len(self):
            raise IndexError('{} out of range'.format(self.__class__.__name__))

        rng_state = torch.get_rng_state()
        torch.manual_seed(index + self.random_offset)

        while True:
            img = torch.zeros(self.dim, self.dim)
            dx = torch.randint(-10, 10, (1,), dtype=torch.float)
            dy = torch.randint(-10, 10, (1,), dtype=torch.float)
            c = torch.randint(-20, 20, (1,), dtype=torch.float)

            params = torch.cat((dy/dx, c))
            xy = torch.randint(0, img.shape[1], (20, 2), dtype=torch.float)
```

1
xy[:,1] = xy[:,0] * params[0] + params[1]

xy.round_()
xy = xy[ xy[:,1] > 0 ]
xy = xy[ xy[:,1] < self.dim ]
xy = xy[ xy[:,0] < self.dim ]

for i in range(xy.shape[0]):
x, y = xy[i][0], self.dim - xy[i][1]
img[int(y), int(x)]=1
if img.sum() > 2:
    break

torch.set_rng_state(rng_state)
return img.unsqueeze(0), params

def __len__(self):
    return self.size

train_data = MyDataset()
val_data = MyDataset(size=500, random_offset=33333)
test_data = MyDataset(size=500, random_offset=99999)

1.1 A simple CNN baseline (2 marks)
Implement the following CNN model, and train it using Adam (default parameters) for 100 epochs (use
a GPU and be prepared to wait 6 or 7 minutes!). Use shuffled batches of 128 items. State the loss
function you're using. Comment on the performance of the model.
Convolution2D, channels=48, size=3x3, stride=1, padding=1
ReLU
Linear, 128 outputs
ReLU
Linear, 2 outputs

2 A second attempt
Clearly the CNN implemented in Section 1 has many parameters in its final hidden layers. One com-
mon way of reducing this is to use Global Max Pooling to flatten the feature maps into a vector (called

2.1 A simple CNN with global pooling (1 mark)
Implement the following CNN model, and train it using Adam (default parameters) for 100 epochs. Use
shuffled batches of 128 items. Comment on the model performance.
Convolution2D, channels=48, size=3x3, stride=1, padding=1
ReLU
Convolution2D, channels=48, size=3x3, stride=1, padding=1
ReLU
Global Max Pool
Linear, 128 outputs
ReLU
Linear, 2 outputs
3 Something that actually works?

The two models so far likely have a few issues. We’re now going to try and fix this.

### 3.1 Let’s regress (2 marks)

Modify the model from Section 2 as follows:

1. Modify the number of input channels to the first convolutional layer to be 3 instead of 1

2. In the forward pass, before the first convolution, modify the input, $x$, using the following code:

   ```python
   idxx = torch.repeat_interleave(
       torch.arange(-20,20, dtype=torch.float).unsqueeze(0) / 40.0,
       repeats=40, dim=0).to(x.device)
   idxy = idxx.clone().t()
   idx = torch.stack([idxx, idxy]).unsqueeze(0)
   idx = torch.repeat_interleave(idx, repeats=x.shape[0], dim=0)
   x = torch.cat([x, idx], dim=1)
   ```

Train the modified model using Adam (default parameters) for 100 epochs. Use shuffled batches of 128 items. **Comment** on the model performance. **Describe** the rationale for the modification that was made.