CSC 4080: Artificial Intelligence in

Medical Imaging and Health

Homework Set 1

Due date: Feb. 20, 2022

Short Answer questions:

1. Explain the image acquisition process for digital CT Imaging in math. (10%)

For a 2D attenuation function $\mu(x, y)$, describing the X-ray attenuation per unit volume in some slice z in the human body, the Radon transform is given by all line integrals through this function:

$$R(\rho,\varphi)[\mu(x,y)] = \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} \mu(x,y) \delta(\rho - x\cos\varphi - y\sin\varphi) dxdy$$

where δ is the Dirac-delta function and s is the line.

When light spreads in the homogeneous (均匀) medium, its intensity attenuation model is: $I = I_0 e^{-\int_L \mu(x) dx}$, where u is the attenuation coefficient and x is the length. The Radon Transform is $ln(\frac{I}{I_0}) = -\int_L \mu(x) dx = R_L \mu$.

Two variables (ρ and ϕ) in radon transform, assume that ϕ is fixed, only ρ changes. We apply Fourier transformation to $R_L\mu$:

$$\begin{split} F_{\rho}R_{\mu}(\rho,\varphi) &= \int_{-\infty}^{\infty} e^{-2\pi i r \rho} R_{\mu}(\rho,\varphi) d\rho \\ &= \int_{-\infty}^{\infty} e^{-2\pi i r \rho} \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} \mu(x_1,x_2) \delta(\rho - x_1 cos\varphi - x_2 sin\varphi) dx_1 dx_2 d\rho \\ &= \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} \mu(x_1,x_2) \int_{-\infty}^{\infty} e^{-2\pi i r \rho} \, \delta(\rho - x_1 cos\varphi - x_2 sin\varphi) d\rho dx_1 dx_2 \\ &= \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} \mu(x_1,x_2) e^{-2\pi i r(x_1 cos\varphi + x_2 sin\varphi)} = F\mu(\xi_1,\xi_2) \end{split}$$

Take the inverse of Fourier transform, μ can be calculated. Then the information of the thickness of organs is found. The result is:

$$\mu(\xi_1, \xi_2) = \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} F_{\rho} R_{\mu}(\rho, \varphi) e^{-2\pi i r (x_1 \cos \varphi + x_2 \sin \varphi)} dx_1 dx_2$$

$$= \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} \ln(\frac{I(\rho, \varphi)}{I_0(\rho, \varphi)}) e^{-2\pi i r (x_1 \cos \varphi + x_2 \sin \varphi)} e^{-2\pi i r \rho} d\rho dx_1 dx_2)$$

2. What is the purpose of filtering with a Hamming window filter when reconstructing CT images? (10%)

Combined with High pass filters, to reduce the amplification of high-frequencies. It is also a low pass filter, which presents a high degree of smoothing.

- 3. What is the advantage and disadvantages of X-ray, CT, ultrasound, and MRI? (10%)
 - Advantages of X-ray: It provides a lucid picture of humans' bones, and the thickness and the shape of organs. It's cheap.
 - Shortcomings for X-ray: It cannot view the 3D appearance of the human body. The radiation damages humans' health. Short at distinguishing organs with similar density and thickness and adjacent positions.
 - Advantages of CT: Form images of non-living objects. Create an image of one or more slices through the body.
 - Shortcomings of CT: Time consuming, expensive and a great amount of radiation.
 - Advantages of ultrasound: fast and timely, no pain and danger, non-invasive. It has been popularized in clinical applications and it is an important part of medical imaging. Because the equipment is not as expensive as CT or MRI equipment, US diagnosis can obtain arbitrary cross-sectional images of organs, and can also observe the activity of moving organs. It's cheap and convenient.
 - Shortcomings of ultrasound: the contrast resolution and spatial resolution of the image are not as high as that of CT or MRI.
 - Advantages of MRI: Examination of nervous system, head, cervical spine (颈椎), thoracic spine, lumbar spine, limbs and other parts has unique advantages. The biggest advantage is that it can not only display diseased tissues, but also reflect the physiological and biochemical information of living tissue functions and metabolic processes.

Shortcomings of MRI: the contrast resolution and spatial resolution of the image are not as high as that of CT or MRI.

Calculation problems:

1. Get the derivatives for the sigmoid function $\sigma(x) = \frac{1}{1 + \exp(-x)}$ and softmax function $z_k = \operatorname{softmax}(x_k) = \frac{\exp(x_k)}{\sum_{i=1}^K \exp(x_i)}$ (you should write down the explicit matrix form different from the slides). Talk about the similarities of these two functions and their applications, respectively. (10%)

$$\sigma'(x) = \begin{bmatrix} \frac{d(\frac{1}{1 + \exp(-x_1)})}{dx_1} & \frac{d(\frac{1}{1 + \exp(-x_1)})}{dx_2} & \dots & \frac{d(\frac{1}{1 + \exp(-x_1)})}{dx_n} \\ \frac{d(\frac{1}{1 + \exp(-x_2)})}{dx_1} & \frac{d(\frac{1}{1 + \exp(-x_2)})}{dx_2} & \dots & \frac{d(\frac{1}{1 + \exp(-x_2)})}{dx_n} \\ \frac{d(\frac{1}{1 + \exp(-x_n)})}{dx_1} & \frac{d(\frac{1}{1 + \exp(-x_n)})}{dx_2} & \dots & \frac{d(\frac{1}{1 + \exp(-x_n)})}{dx_n} \end{bmatrix} = \\ \begin{bmatrix} \frac{\exp(-x_1)}{(1 + \exp(-x_1))^2} & 0 & \dots & 0 \\ 0 & \frac{\exp(-x_2)}{(1 + \exp(-x_2))^2} & \dots & 0 \\ \dots & \dots & \dots & \dots \\ 0 & 0 & 0 & \frac{\exp(-x_n)}{(1 + \exp(-x_n))^2} \end{bmatrix}$$

$$z_{k}'(x) = \begin{bmatrix} \frac{d(\frac{\exp(x_{1})}{\sum_{i=1}^{K} \exp(x_{i})})}{dx_{1}} & \frac{d(\frac{\exp(x_{1})}{\sum_{i=1}^{K} \exp(x_{i})})}{dx_{2}} & \dots & \frac{d(\frac{\exp(x_{1})}{\sum_{i=1}^{K} \exp(x_{i})})}{dx_{K}} \\ \frac{d(\frac{\exp(x_{2})}{\sum_{i=1}^{K} \exp(x_{i})})}{dx_{1}} & \frac{d(\frac{\exp(x_{2})}{\sum_{i=1}^{K} \exp(x_{i})})}{dx_{2}} & \dots & \frac{d(\frac{\exp(x_{2})}{\sum_{i=1}^{K} \exp(x_{i})})}{dx_{K}} \\ \dots & \dots & \dots \\ \frac{d(\frac{\exp(x_{K})}{\sum_{i=1}^{K} \exp(x_{i})})}{dx_{1}} & \frac{d(\frac{\exp(x_{K})}{\sum_{i=1}^{K} \exp(x_{i})})}{dx_{2}} & \dots & \frac{d(\frac{\exp(x_{K})}{\sum_{i=1}^{K} \exp(x_{i})})}{dx_{K}} \end{bmatrix} = 0$$

$$\begin{bmatrix} \frac{\sum_{i=1,i\neq 1}^{K} \exp{(x_i + x_1)}}{\left(\sum_{i=1}^{K} \exp{(x_i)}\right)^2} & \frac{-\exp{(x_2 + x_1)}}{\left(\sum_{i=1}^{K} \exp{(x_i)}\right)^2} & \dots & \frac{-\exp{(x_K + x_1)}}{\left(\sum_{i=1}^{K} \exp{(x_i)}\right)^2} \\ \frac{-\exp{(x_1 + x_2)}}{\left(\sum_{i=1}^{K} \exp{(x_i)}\right)^2} & \frac{\sum_{i=1,i\neq 2}^{K} \exp{(x_i + x_2)}}{\left(\sum_{i=1}^{K} \exp{(x_i)}\right)^2} & \dots & \frac{-\exp{(x_K + x_2)}}{\left(\sum_{i=1}^{K} \exp{(x_i)}\right)^2} \\ \dots & \dots & \dots & \dots \\ \frac{-\exp{(x_1 + x_K)}}{\left(\sum_{i=1}^{K} \exp{(x_i)}\right)^2} & \frac{-\exp{(x_2 + x_K)}}{\left(\sum_{i=1}^{K} \exp{(x_i)}\right)^2} & \frac{\sum_{i=1,i\neq K}^{K} \exp{(x_i + x_K)}}{\left(\sum_{i=1}^{K} \exp{(x_i)}\right)^2} \end{bmatrix}$$

Similarities:

Both of functions have a range (0, 1). The outputs can represent probabilities in the output layer of CNN and NN in classification problems. However, sigmoid functions can only be used in the binary classification. Softmax functions is an extension of sigmoid functions at some extend.

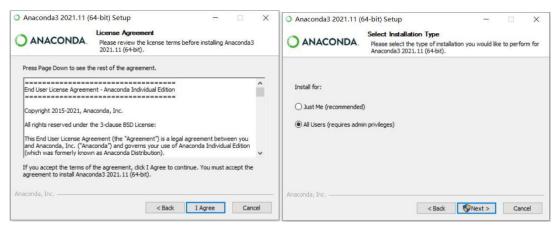
2. Write down the cross-entropy loss function, and its derivatives. Discuss how it works for classification tasks. Is there any disadvantage of using naïve cross-entropy for classification? (10%)

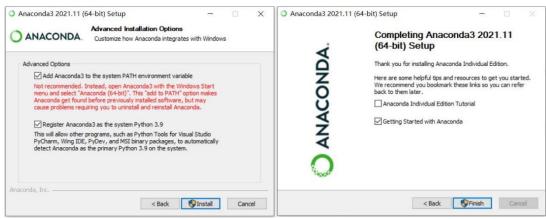
Assume \mathbf{x} is the result vector of the last layer, and the probability generate by softmax function. Then \mathbf{p} =softmax(\mathbf{x}), Assume the target one is the i th vector, then the indicator function \mathbf{y} with $y_i = 1$, $y_j = 0$ if $j \neq i$. $L = -\sum_i y_i \log(p_i)$, $\frac{dL}{dx_i} = -\sum_i y_i \frac{dlog(p_i)}{dp_i} \frac{dp_i}{dx_i} = \sum_i y_i \frac{1}{p_i} \frac{dp_i}{dx_i}$. In the task, the softmax function transform the result into probability. Cross entropy function provides a judgement for the efficiency of the model. And the cross entropy function is differentiable, which means it can update the model by gradient descending. Yes, the disadvantage is that if p_i is very small, then the gradient will be very large(gradient explode) since it's in the denominator. And softmax function is similar to an exponential function. The repeating calculation of exponential and log functions.

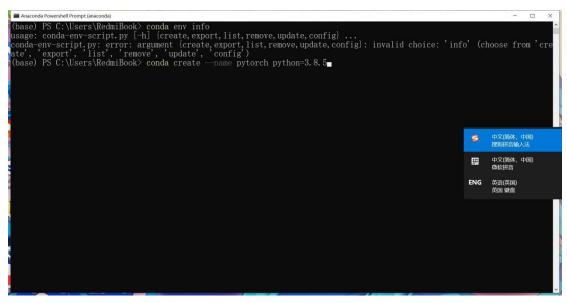
Programming problems:

1. Install Anaconda (a very good Python version that can help you easily implement data science applications).

Based on Anaconda, install some commonly used packages for Python, including Numpy, Pandas, Matplotlib, Seaborn, PyTorch (CPU version is OK if you don't have GPU on your device). Record your installation process in detail, screenshots for every step are needed. (10%)







```
Anaconda Powershell Prompt (anaconda)
                                                                                                                                                                                                                                                                                                                                                   П
            package
                                                                                                                                  build
                                                                                                      haa95532 4
py38haa95532_2
h2bbff1b 0
py38haa95532_0
h5fd99c_1
py38haa95532_0
h2bbff1b_0
pyhd3eb1b0_0
py38haa95532_2
           ca-certificates-2021.10.26
certifi-2021.10.8
openss1-1.1.1m
pip-21.2.2
python-3.8.5
setuptools-58.0.4
sqlite-3.37.2
wheel-0.37.1
                                                                                                                                                                         116 KB
152 KB
4.8 MB
1.9 MB
15.7 MB
779 KB
799 KB
33 KB
15 KB
                                                                                                                                                                        24.2 MB
 The following NEW packages will be INSTALLED:

        ca-certificates certifi
        pkgs/main/win-64::ca-certificates-2021. 10. 26-haa95532_4

        certifi
        pkgs/main/win-64::certifi-2021. 10. 8-py38haa95532_2

        ppenssl
        pkgs/main/win-64::openssl-1. 1. lm-h2bbfflb_0

        pip
        pkgs/main/win-64::ptp-21. 2. 2-py38haa95532_0

        python
        pkgs/main/win-64::pthon-3. 8. 5-h5fd99cc_1

        setuptools
        pkgs/main/win-64::setuptools-58. 0. 4-py38haa95532_0

        sqlite
        pkgs/main/win-64::setuptools-58. 0. 4-py38haa95532_0

        pkgs/main/win-64::scllite-3. 37. 2-h2bbfflb_0

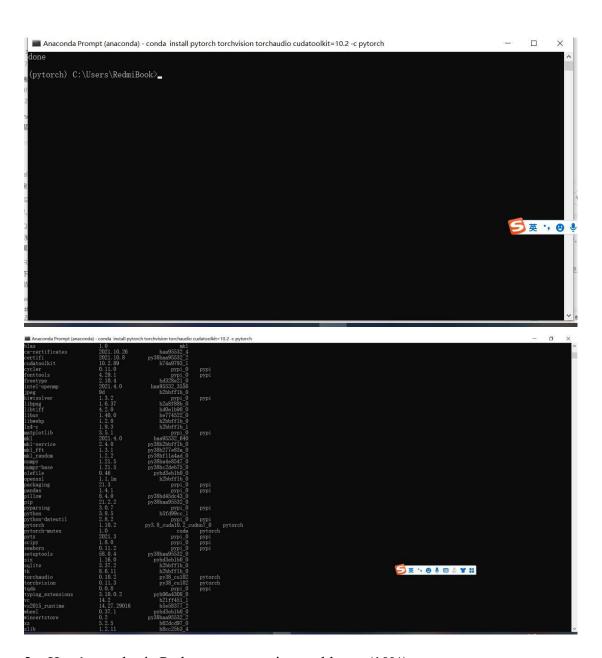
        vs2015_runtime
        pkgs/main/win-64::vc-14. 2-h2lff451_1

        pkgs/main/win-64::vs2015_runtime-14. 27. 29016-h5e58377_2

        pkgs/main/win-64::vs015_runtime-14. 27. 29016-h5e58375_2

        pkgs/main/win-64::vs015_runtime-14. 27. 29016-h5e58375_2

Proceed ([y]/n)?
Anaconda Powershell Prompt (anaconda)
                                                                                                                                                                                                                                                                                                                                                  □ ×
       To activate this environment, use
                  $ conda activate pytorch
                 $ conda deactivate
  (base) PS C:\Users\RedmiBook>
Anaconda Powershell Prompt (anaconda)
      To activate this environment, use
   (base) PS C:\Users\RedmiBook> conda activate pytorch
(pytorch) PS C:\Users\RedmiBook> conda list
‡ packages in environment at D:\anaconda\envs\pytorch:
  三英,●∮⊞ ▮
```



- 2. Here's two basic Python programming problems: (10%)
 - 1) Write a program to sum all the numbers in a list;

2) Write a program to find all prime numbers in a list.

```
def isprime(M):
    for i in range(2,int(np.sqrt(M))+1):
        if M%i==0:
            return True
    return False
    list= [i for i in range(100)]

ans=[]
for i in list:
    if isprime(i)==False:
        ans.append(i)
ans
```

- 3. Here's two basic Numpy programming problems: (10%)
 - 1) Given an array, calculate its softmax result;

2) use the softmax result to calculate the cross-entropy loss (you are free to construct the input x, and ground truth label y by yourself).

```
x=np.array([1,2,3])
y=softmax(x)
-np.log(y[0])

2.40760596444438
```

4. Medical data exploration. You are going to work with a real medical image dataset, including exploring data labels, visualizing and observing the data, along with processing the data. Please fill in the blanks in "data_exploration.ipynb" and successfully run all the cells. (20%)

```
# Plot a histogram of the distribution of the pixels
### Your code here ###
raw_image=raw_image.reshape(1,1048576)
sns.distplot(raw_image,bins=len(np.unique(raw_image)),kde=False,color='b')
import matplotlib.patches as mp
info = mp.Patch(label='pixel Mean: %.4f & Standard Deviation %.4f '%(raw_image.std(),raw_image.mean()))
### End your code ###
     ### End your code ###
plt.legend(handles=[info],loc='upper center')
plt.tile('Distribution of Pixel Intensities in the Image')
plt.xlabel('Pixel Intensity')
plt.ylabel[['# Pixels in Image')]
D:\anaconda\envs\pytorch\lib\site-packages\seaborn\distributions.py:2619: FutureWarning: `distplot` is a deprecated function and will be removed in a
future version. Please adapt your code to use either 'displot' (a figure-level function with similar flexibility) or 'histplot' (an axes-level
function for histograms).
   warnings.warn(msg, FutureWarning)
 Text(0, 0.5, '# Pixels in Image')
                        Distribution of Pixel Intensities in the Image
                    Pixel Mean: 0.2294 & Standard Deviation 0.5800
     25000
      20000
      15000
     10000
       5000
                                         0.4 0.6
Pixel Intensity
                             0.2
```

```
# Create transformation for data standardization
    norm_mean = [0.6, 0.6, 0.6]
    norm_std = [0.225, 0.225, 0.225]
    data_transform = transforms.Compose([
        transforms.Resize((320, 320)),
        transforms.ToTensor(),
        ### Your code here, use API like transforms.XXX() ###
        transforms.Normalize(norm_mean, norm_std)
    dataset = XRayDataset('nih/images-small', data transform)
   data (module) transforms ns.Compose([
      transforms.Resize((320, 320)),
      transforms.RandomCrop(300),
      transforms.ToTensor()
   dataset = XRayDataset('nih/images-small', data_transform)
   sns.set_style("white")
   generated_image = dataset.__getitem__(0)
   inverted_image = transform_invert(copy.deepcopy(generated_image), data_transform)
   plt.imshow(inverted_image, cmap='gray')
   plt.colorbar()
   plt.title('Chest X Ray Image after Random Cropping')
Text(0.5, 1.0, 'Chest X Ray Image after Random Cropping')
   Chest X Ray Image after Random Cropping
 50
                                   - 200
 100
                                   150
 150
                                    100
 200
 250
```

0

100

200

```
data transform = transforms.Compose([
       transforms.Resize((320, 320)),
       transforms.RandomHorizontalFlip(p=1),
       transforms.ToTensor()
   dataset = XRayDataset('nih/images-small', data_transform)
   sns.set_style("white")
   generated_image = dataset.__getitem__(0)
   inverted_image = transform_invert(copy.deepcopy(generated_image), data_transform)
   plt.imshow(inverted_image, cmap='gray')
   plt.colorbar()
   plt.title('Chest X Ray Image after Flipping')
Text(0.5, 1.0, 'Chest X Ray Image after Flipping')
       Chest X Ray Image after Flipping
  50
                                      200
 100
                                       - 150
 150
 200
                                      100
 250
 300
    0
             100
                      200
                               300
```

```
data_transform = transforms.Compose([
       transforms.Resize((320, 320)),
       transforms.RandomRotation(20),
       transforms.ToTensor()
   dataset = XRayDataset('nih/images-small', data_transform)
   sns.set style("white")
   generated_image = dataset.__getitem__(0)
   inverted_image = transform_invert(copy.deepcopy(generated_image), data_transform)
   plt.imshow(inverted_image, cmap='gray')
   plt.colorbar()
   plt.title('Chest X Ray Image after Rotating')
✓ 0.2s
Text(0.5, 1.0, 'Chest X Ray Image after Rotating')
       Chest X Ray Image after Rotating
   0
  50
                                       200
 100
                                        150
 150
                                       - 100
 200
 250
 300
             100
                      200
                                300
```

```
data_transform = transforms.Compose([
        transforms.Resize((320, 320)),
        transforms.ColorJitter(brightness=0.1, contrast=0, saturation=1, hue=0.1),
        transforms.ToTensor()
   dataset = XRayDataset('nih/images-small', data_transform)
   sns.set_style("white")
   generated_image = dataset.__getitem__(0)
inverted_image = transform_invert(copy.deepcopy(generated_image), data_transform)
   plt.imshow(inverted_image, cmap='gray')
   plt.colorbar()
   plt.title('Chest X Ray Image after Color Jittering')
Text(0.5, 1.0, 'Chest X Ray Image after Color Jittering')
      Chest X Ray Image after Color Jittering
  50
                                          - 200
 100
                                          150
 150
 200
                                           100
 250
 300
```

100

200

300