My Project

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## README

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Tested on Ubuntu 20 and python>=3.8

This work is done for the CSE600A (Object Oriented Programming and Design) course requirement for group number 41. The problem statement is given in problem.pdf.

#### 1.0.0.1 Overview

- 1. In this work we implement a simple 2 layer feed-forward arrtificial neural network (ANN) from scratch using numpy only. Gradients are calculated are calculated similarly to this work: http://cs231n.←stanford.edu/slides/2021/discussion\_2\_backprop.pdf.
- 2. Doxygen file is at latex/refman.pdf
- 3. For the profiler file refer to the test/test.txt
- 4. ann/example2l (2l = 2 layer) class has the implementation of training and testing of the 2 layer network.
- 5. We achieve an accuracy of 95 percent when trained for 1 epoch.
- 6. Data and output predictions are in data folder (after you erxtract the data.tar.gz).

#### 1.0.0.2 Model specifications

```
i_d = 784 # input dimension
m_d = 100 # no of cells in the mid layer
o_d = 10 # output dimension
epoch = 1 # we train for 1 epoch only
act = 'relu' # activation type
lr = 0.00001 # learning rate
X,Y,X_test,Y_test = train_features, train_label, test_features, test_label
```

# 1.0.0.3 Run these commands to set up the environment. Conda need to be installed. If not than you have to manually install the dependencies.

```
conda env create -f environment.yml
conda activate oopd
python setup.py sdist bdist_wheel
pip install dist/ann-0.0.1-py3-none-any.whl
```

2 README

#### 1.0.0.4 Run these commands to extract the MNIST dataset and to train the example network on it.

tar -xvf data.tar.gz

This will extract the database required to train the network on MNIST dataset.

After the traning the predicions and their true values will be saved in data/test\_pred\_true.csv file. To conver this into a sql database run the below commands.

#### 1.0.0.5 Run these commands to save the prediction to a sql database.

.mode csv
.import data/test\_pred\_true.csv oopd
.save data/oopd\_pred\_true.db

# Namespace Index

## 2.1 Namespace List

Here is a list of all documented namespaces with brief descriptions:

ann.activation	??
ann.data	??
ann.layer	??
ann.loss	??

4 Namespace Index

# **Hierarchical Index**

## 3.1 Class Hierarchy

This inheritance list is sorted roughly, but not completely, alphabetically:

n.loss.Cross_Ent_Loss	??
ann.example2l.Net	??
n.layer.Linear	??
ann.example2l.Net	??
n.data.Mnist	
n.activation.Relu	??
ann.example2l.Net	??
n.activation.Sigmoid	??
ann.example2l.Net	??

6 Hierarchical Index

# **Class Index**

## 4.1 Class List

Here are the classes, structs, unions and interfaces with brief descriptions:

ann.loss.Cross_Ent_Loss	?
ann.layer.Linear	?
ann.data.Mnist	7
ann.example2l.Net	1
ann.activation.Relu	7
ann.activation.Sigmoid	

8 Class Index

# **Namespace Documentation**

### 5.1 ann.activation Namespace Reference

#### **Classes**

- class Relu
- class Sigmoid

#### 5.1.1 Detailed Description

@package docstring
This module contains the different activation functions used in AI.

### 5.2 ann.data Namespace Reference

#### **Classes**

• class Mnist

#### 5.2.1 Detailed Description

 ${\tt Qpackage}$  docstring This module contains the different dataset download functions.

### 5.3 ann.layer Namespace Reference

#### Classes

class Linear

### 5.3.1 Detailed Description

@package docstring
This module contains all the different layers used in neural networks

## 5.4 ann.loss Namespace Reference

#### **Classes**

• class Cross\_Ent\_Loss

### 5.4.1 Detailed Description

@package docstring
This module contains all the los functions

## **Class Documentation**

### 6.1 ann.loss.Cross\_Ent\_Loss Class Reference

Inheritance diagram for ann.loss.Cross\_Ent\_Loss:

```
classann_1_1loss_1_1Cross__Ent__Loss-eps-converted-to.
```

#### **Public Member Functions**

- def forward\_loss (self, logits, true)
- def backward loss (self)
- def softmax (self, x)

#### **Public Attributes**

- pred
- true
- z

#### 6.1.1 Detailed Description

Applies the combined cross entropy and softmax loss function element-wise. For easier backprop calculation.

Args: None

#### 6.1.2 Member Function Documentation

#### 6.1.2.1 backward\_loss()

```
def ann.loss.Cross_Ent_Loss.backward_loss ( self \ ) 
 Implements \ the \ backprop \ calculation.
```

#### 6.1.2.2 forward\_loss()

#### 6.1.2.3 softmax()

The documentation for this class was generated from the following file:

• ann/loss.py

## 6.2 ann.layer.Linear Class Reference

Inheritance diagram for ann.layer.Linear:

```
classann_1_1layer_1_1Linear-eps-converted-to.pdf
```

#### **Public Member Functions**

```
def __init__ (self, i_d, o_d)
def forward_l (self, x)
def backward_l (self, x)
def update_l (self, alpha)
def reset_l (self)
```

#### **Public Attributes**

- · weight
- · bias
- · weight\_grad
- · bias\_grad
- x

#### 6.2.1 Detailed Description

```
This class implements the linear layer.

Args:
    i_d > input dimension
    o_d > output dimension

Attributes:
    weight: the learnable weight of the module.
    bias: the learnable bias of the module.

    weight_grad: gradient for the weight matrix.
    bias_grad: gradient for the bias matrix.

Examples:
    >>> m = ann.Linear(10,20)
    >>> inp = np.random.rand(i_d,1)
    >>> out = m.forward(inp)
    >>> out.shape
[o_d,1]
```

#### 6.2.2 Member Function Documentation

#### 6.2.2.1 backward\_I()

```
def ann.layer.Linear.backward_l ( self, \\ x \ ) Implements the Backprop calculations given the gradeints.
```

#### 6.2.2.2 forward\_I()

#### 6.2.2.3 reset\_I()

```
def ann.layer.Linear.reset_l ( self \ ) To reset the gradients to zero.
```

#### 6.2.2.4 update\_I()

The documentation for this class was generated from the following file:

ann/layer.py

#### 6.3 ann.data.Mnist Class Reference

#### **Public Member Functions**

```
def __init__ (self, path)
def from_sql (self, train_db='oopd_train.db', test_db='oopd_test.db')
def data (self)
def fetch (self, url)
```

• def sql\_database (self, path\_to\_db)

#### **Public Attributes**

path

#### 6.3.1 Detailed Description

```
Download the MNIST dataset and load it in a required format.

Args:
    path: where to save the dataset.

Return:
    X: training input [60000,784]
    Y: training output [60000,1]

X_test = test input [10000,784]
    Y_test = test output [10000,1]
```

#### 6.3.2 Member Function Documentation

#### 6.3.2.1 data()

```
def ann.data.Mnist.data ( self \;) Download the data from source Save it to the disk in self.path/ directory
```

#### 6.3.2.2 fetch()

```
def ann.data.Mnist.fetch ( self, \\ url \; ) Downloads the MNIST data given the url.
```

#### 6.3.2.3 from\_sql()

#### 6.3.2.4 sql\_database()

```
def ann.data.Mnist.sql_database ( self, \\ path\_to\_db \ ) Read the data from a sql database.
```

The documentation for this class was generated from the following file:

· ann/data.py

### 6.4 ann.example2I.Net Class Reference

Inheritance diagram for ann.example2l.Net:

```
classann_1_1example21_1_1Net-eps-converted-to.pdf
```

#### **Public Member Functions**

```
• def __init__ (self, i_d, m_d, o_d, act, Ir, epoch, X, Y, X_test, Y_test)
```

- def forward (self, x)
- def backward (self)
- def update (self)
- def reset (self)
- def loss (self, pred, index)
- def evaluate (self, X=None, Y=None, save=True)
- def train (self, step\_u=1, step=100000, test=False)

#### **Public Attributes**

- · alpha
- · epoch
- · layer1
- · act1
- · layer2
- · cross\_ent\_loss
- Y
- Y\_test
- · X\_test

#### 6.4.1 Detailed Description

```
Inheritance, Encapsulation, Polymorphism.
Create a 2 layer feed forward artifical neural network (ANN).
Args:
    i_d: input dimension.
    m_d: number of cells in 1st layer.
    o_d: output dimension.
    act: activation function to use.
    lr: Learning rate
    epoch: number of epochs to train.
    X: training input [60000,784].
    Y: training output [60000,1].
    X_{\text{test}} = \text{test input [10000,784]}.
    Y_{\text{test}} = \text{test output [10000,1]}.
Returns:
    None
Example:
    i_d, m_d, o_d = 784, 100, 10
    act = 'relu'
    lr, epoch = 0.00001, 10
    X,Y,X_test,Y_test = data.Mnist(path='data').data()
    net = example21.Net(i_d,m_d,o_d,act, lr,epoch, X,Y,X_test,Y_test)
    net.train()
```

#### 6.4.2 Constructor & Destructor Documentation

#### 6.4.2.1 \_\_init\_\_()

Reimplemented from ann.layer.Linear.

#### 6.4.3 Member Function Documentation

#### 6.4.3.1 backward()

```
def ann.example21.Net.backward ( self \ ) Calls backprop fucntion for each layer to calculate the gradients.
```

#### 6.4.3.2 evaluate()

```
def ann.example21.Net.evaluate ( self, \\ X = None, \\ Y = None, \\ save = True )
```

Evaulate the created model given the test data.

#### 6.4.3.3 forward()

```
def ann.example21.Net.forward ( self, \\ x \ )
```

Calls the forward fucntion for each layer.

#### 6.4.3.4 loss()

Calculate the cross entropy loss.

#### 6.4.3.5 reset()

```
def ann.example21.Net.reset ( self \ ) Calls the reset fucntion of each layer to reset the gradients back to 0.
```

#### 6.4.3.6 train()

```
def ann.example21.Net.train ( self, step_u = 1, step = 100000, test = False)
```

Train the created model given the input.

#### 6.4.3.7 update()

```
def ann.example21.Net.update ( self \ ) Calls the update fucntion of each layer to update the gradients.
```

The documentation for this class was generated from the following file:

• ann/example2l.py

#### 6.5 ann.activation.Relu Class Reference

Inheritance diagram for ann.activation.Relu:

```
classann_1_1activation_1_1Relu-eps-converted-to.pdf
```

#### **Public Member Functions**

- def forward\_a (self, x)
- def backward\_a (self, x)

#### **Public Attributes**

• z

#### 6.5.1 Detailed Description

```
Applies the rectified linear unit function element-wise.  \label{eq:applies}  \mbox{Args:}  None
```

#### 6.5.2 Member Function Documentation

#### 6.5.2.1 backward\_a()

```
def ann.activation.Relu.backward_a ( self, \\ x \ )
```

Implements the backprop calculation.

#### 6.5.2.2 forward\_a()

```
def ann.activation.Relu.forward_a ( self, \\ x \ )
```

Implements the Forward calculation given an input.

The documentation for this class was generated from the following file:

· ann/activation.py

### 6.6 ann.activation.Sigmoid Class Reference

Inheritance diagram for ann.activation.Sigmoid:

```
classann_1_1activation_1_1Sigmoid-eps-converted-to.pdf
```

#### **Public Member Functions**

- def forward\_a (self, x)
- def backward\_a (self, x)

#### **Public Attributes**

• z

#### 6.6.1 Detailed Description

```
Applies the sigmoid function element-wise.

Args:
None
```

#### 6.6.2 Member Function Documentation

#### 6.6.2.1 backward\_a()

```
\begin{tabular}{ll} $\operatorname{def ann.activation.Sigmoid.backward\_a} & $\operatorname{self}, \\ & $x$ ) \end{tabular}
```

Implements the backprop calculation.

#### 6.6.2.2 forward\_a()

```
def ann.activation.Sigmoid.forward_a ( self, \\ x \ )
```

Implements the Forward calculation given an input.

The documentation for this class was generated from the following file:

· ann/activation.py