

ECE695DL: Homework 3

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one_neuron_classifier_sgd_plus.py

```
import random
import numpy as np
import matplotlib.pyplot as plt
import operator

seed = 0
random.seed(seed)
np.random.seed(seed)

from ComputationalGraphPrimer import *

class CGPPlus(ComputationalGraphPrimer):
    def __init__(self, *args, **kwargs):
        super().__init__(*args, **kwargs)
        self.old_step=[]
        self.mu = 0

    def set_mu(self, mu):
        self.mu = mu

    def backprop_and_update_params_one_neuron_model(self, y_error,
        vals_for_input_vars, deriv_sigmoid):
        input_vars = self.independent_vars
        vals_for_input_vars_dict = dict(zip(input_vars,
            list(vals_for_input_vars)))
        for i, param in enumerate(self.vals_for_learnable_params):
            step = self.mu*self.old_step[i] + self.learning_rate * y_error *
            vals_for_input_vars_dict[input_vars[i]] * deriv_sigmoid
            self.vals_for_learnable_params[param] += step
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        self.old_step[i] = step
    self.bias_step =
    self.mu*self.bias_step+self.learning_rate*y_error*deriv_sigmoid
    self.bias += self.bias_step

def reset_step(self):
    for i in range(len(self.learnable_params)):
        if len(self.old_step) < len(self.vals_for_learnable_params):
            self.old_step.append(0)
        else:
            self.old_step[i] = 0

def run_training_loop_one_neuron_model(self):
    training_data = self.training_data
    self.vals_for_learnable_params = {param: random.uniform(0,1)
    for param in self.learnable_params}
    self.bias = random.uniform(0,1)
    self.bias_step=0

class DataLoader:
    def __init__(self, training_data, batch_size):
        self.training_data = training_data
        self.batch_size = batch_size
        self.class_0_samples = [(item, 0) for item in
        self.training_data[0]]
        self.class_1_samples = [(item, 1) for item in
        self.training_data[1]]
    def __len__(self):
        return len(self.training_data[0]) + len(self.training_data[1])
    def _getitem(self):
        cointoss = random.choice([0,1])
        if cointoss == 0:
            return random.choice(self.class_0_samples)
        else:
            return random.choice(self.class_1_samples)
    def getbatch(self):
        batch_data, batch_labels = [], []
        maxval = 0.0
        for _ in range(self.batch_size):
            item = self._getitem()
            if np.max(item[0]) > maxval:
                maxval = np.max(item[0])
            batch_data.append(item[0])
            batch_labels.append(item[1])
        batch_data = [item/maxval for item in batch_data]
        batch = [batch_data, batch_labels]

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        return batch

data_loader = DataLoader(training_data, batch_size=self.batch_size)
loss_running_record = []
avg_loss_over_literations = 0.0
self.reset_step()
for i in range(self.training_iterations):
    data = data_loader.getbatch()
    data_tuples = data[0]
    class_labels = data[1]
    y_preds, deriv_sigmoids =
self.forward_prop_one_neuron_model(data_tuples)
    loss = sum([(abs(class_labels[i] - y_preds[i]))**2 for i in
range(len(class_labels))])
    loss_avg = loss / float(len(class_labels))
    avg_loss_over_literations += loss_avg
    if i%(self.display_loss_how_often) == 0:
        avg_loss_over_literations /= self.display_loss_how_often
        loss_running_record.append(avg_loss_over_literations)
        print("[iter=%d]  loss = %.4f" %
        (i+1, avg_loss_over_literations))
        avg_loss_over_literations = 0.0
    y_errors = list(map(operator.sub, class_labels, y_preds))
    y_error_avg = sum(y_errors) / float(len(class_labels))
    deriv_sigmoid_avg = sum(deriv_sigmoids) / float(len(class_labels))
    data_tuple_avg = [sum(x) for x in zip(*data_tuples)]
    data_tuple_avg = list(map(operator.truediv, data_tuple_avg,
        [float(len(class_labels))] *
        len(class_labels) ))
    self.backprop_and_update_params_one_neuron_model(y_error_avg,
    data_tuple_avg, deriv_sigmoid_avg)

return loss_running_record

```

```

cgpplus = CGPPlus(
    one_neuron_model = True,
    expressions = ['xw=ab*xa+bc*xb+cd*xc+ac*xd'],
    output_vars = ['xw'],
    dataset_size = 5000,
    learning_rate = 1e-3,
    training_iterations = 40000,
    batch_size = 16,
    display_loss_how_often = 100,

```

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        debug = True,
    )

cgp = CGPPlus(
    one_neuron_model = True,
    expressions = ['xw=ab*xa+bc*xb+cd*xc+ac*xd'],
    output_vars = ['xw'],
    dataset_size = 5000,
    learning_rate = 1e-3,
    training_iterations = 40000,
    batch_size = 16,
    display_loss_how_often = 100,
    debug = True,
)

cgp.parse_expressions()
cgp.gen_training_data()
sgd_loss = cgp.run_training_loop_one_neuron_model()

cgpplus.parse_expressions()
cgpplus.gen_training_data()
cgpplus.set_mu(0.99)
sgdplus_loss = cgpplus.run_training_loop_one_neuron_model()

plt.figure()
plt.plot(sgd_loss, label = 'SGD Training Loss')
plt.plot(sgdplus_loss, label = 'SGD+ Training Loss')
plt.legend()
plt.show()

```

multi_neuron_classifier_sgd_plus.py

```

import random
import numpy as np
import operator
import matplotlib.pyplot as plt

```

```

seed = 0
random.seed(seed)
np.random.seed(seed)

from ComputationalGraphPrimer import *

class CGPPlus(ComputationalGraphPrimer):
    def __init__(self, *args, **kwargs):
        super().__init__(*args, **kwargs)
        self.old_step = []
        self.mu=0

    def set_mu(self, mu):
        self.mu = mu

    def backprop_and_update_params_multi_neuron_model(self, y_error, class_labels):
        pred_err_backproped_at_layers = {i: [] for i in
            range(1, self.num_layers - 1)}
        pred_err_backproped_at_layers[self.num_layers - 1] = [y_error]
        for back_layer_index in reversed(range(1, self.num_layers)):
            input_vals = self.forw_prop_vals_at_layers[back_layer_index - 1]
            input_vals_avg = [sum(x) for x in zip(*input_vals)]
            input_vals_avg = list(map(operator.truediv, input_vals_avg,
                [float(len(class_labels))] * len(class_labels)))
            deriv_sigmoid = self.gradient_vals_for_layers[back_layer_index]
            deriv_sigmoid_avg = [sum(x) for x in zip(*deriv_sigmoid)]
            deriv_sigmoid_avg = list(map(operator.truediv, deriv_sigmoid_avg,
                [float(len(class_labels))] *
                    len(class_labels)))
            vars_in_layer = self.layer_vars[back_layer_index]
            ## a list like ['xo']
            vars_in_next_layer_back = self.layer_vars[back_layer_index - 1]
            ## a list like ['xw', 'xz']

            layer_params = self.layer_params[back_layer_index]
            ## note that layer_params are stored in a dict like
            ##      {1: [['ap', 'aq', 'ar', 'as'], ['bp', 'bq', 'br', 'bs']], 2:
            ##      [['cp', 'cq']]}
            ## "layer_params[idx]" is a list of lists for the link weights
            ## in layer whose output nodes are in layer "idx"
            transposed_layer_params = list(zip(*layer_params))
            ## creating a transpose of the link matrix

            backproped_error = [None] * len(vars_in_next_layer_back)
            for k, varr in enumerate(vars_in_next_layer_back):

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        for j, var2 in enumerate(vars_in_layer):
            backproped_error[k] =
                sum([self.vals_for_learnable_params[transposed_layer_params[k]
                    [i]] *
                    pred_err_backproped_at_layers
                    [back_layer_index][i]
                    for i in range(len(vars_in_layer))])
            # deriv_sigmoid_avg[i]
        for i in range(len(vars_in_layer))]
        pred_err_backproped_at_layers[back_layer_index - 1] = backproped_error
        input_vars_to_layer = self.layer_vars[back_layer_index - 1]
        for j, var in enumerate(vars_in_layer):
            layer_params = self.layer_params[back_layer_index][j]
            for i, param in enumerate(layer_params):
                gradient_of_loss_for_param = input_vals_avg[i] *
                pred_err_backproped_at_layers[back_layer_index][j]
                step = self.mu*self.old_step[i]+self.learning_rate *
                gradient_of_loss_for_param * deriv_sigmoid_avg[j]
                self.vals_for_learnable_params[param] += step
                self.old_step[i] = step
        self.bias_step = self.mu*self.bias_step+self.learning_rate * sum(
            pred_err_backproped_at_layers[back_layer_index]) *
            sum(deriv_sigmoid_avg) / len(deriv_sigmoid_avg)
        self.bias[back_layer_index - 1] += self.bias_step

def run_training_loop_multi_neuron_model(self):
    training_data = self.training_data
    self.vals_for_learnable_params = {param: random.uniform(0,1) for param in
    self.learnable_params}
    self.bias = [random.uniform(0,1) for _ in range(self.num_layers-1)]
    self.bias_step=0

class DataLoader:
    def __init__(self, training_data, batch_size):
        self.training_data = training_data
        self.batch_size = batch_size
        self.class_0_samples = [(item, 0) for item in
            self.training_data[0]]
        self.class_1_samples = [(item, 1) for item in
            self.training_data[1]]
    def __len__(self):
        return len(self.training_data[0]) + len(self.training_data[1])
    def _getitem(self):
        cointoss = random.choice([0,1])
        if cointoss == 0:

```

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        return random.choice(self.class_0_samples)
    else:
        return random.choice(self.class_1_samples)
def getbatch(self):
    batch_data, batch_labels = [], []
    maxval = 0.0
    for _ in range(self.batch_size):
        item = self._getitem()
        if np.max(item[0]) > maxval:
            maxval = np.max(item[0])
        batch_data.append(item[0])
        batch_labels.append(item[1])
    batch_data = [item/maxval for item in batch_data]
    batch = [batch_data, batch_labels]
    return batch

data_loader = DataLoader(self.training_data, batch_size=self.batch_size)
loss_running_record = []
i = 0
avg_loss_over_literations = 0.0
self.reset_step()
for i in range(self.training_iterations):
    data = data_loader.getbatch()
    data_tuples = data[0]
    class_labels = data[1]
    self.forward_prop_multi_neuron_model(data_tuples)
    predicted_labels_for_batch =
    self.forw_prop_vals_at_layers[self.num_layers-1]
    y_preds = [item for sublist in predicted_labels_for_batch
    for item in sublist]
    loss = sum([(abs(class_labels[i] - y_preds[i]))**2
    for i in range(len(class_labels))])
    loss_avg = loss / float(len(class_labels))
    avg_loss_over_literations += loss_avg
    if i%(self.display_loss_how_often) == 0:
        avg_loss_over_literations /= self.display_loss_how_often
        loss_running_record.append(avg_loss_over_literations)
        print("[iter=%d] loss = %.4f" % (i+1, avg_loss_over_literations))
        avg_loss_over_literations = 0.0
    y_errors = list(map(operator.sub, class_labels, y_preds))
    y_error_avg = sum(y_errors) / float(len(class_labels))
    self.backprop_and_update_params_multi_neuron_model
    (y_error_avg, class_labels)
return loss_running_record

```

```

def reset_step(self):
    for i in range(len(self.learnable_params)):
        if len(self.old_step) < len(self.vals_for_learnable_params):
            self.old_step.append(0)
        else:
            self.old_step[i] = 0

cgpplus = CGPPlus(
    num_layers = 3,
    layers_config = [4,2,1],
    # num of nodes in each layer
    expressions = ['xw=ap*xp+aq*xq+ar*xr+as*xs',
                  'xz=bp*xp+bq*xq+br*xr+bs*xs',
                  'xo=cp*xw+cq*xz'],
    output_vars = ['xo'],
    dataset_size = 5000,
    learning_rate = 1e-3,
    training_iterations = 40000,
    batch_size = 8,
    display_loss_how_often = 100,
    debug = True,
)

cgp = CGPPlus(
    num_layers = 3,
    layers_config = [4,2,1],
    # num of nodes in each layer
    expressions = ['xw=ap*xp+aq*xq+ar*xr+as*xs',
                  'xz=bp*xp+bq*xq+br*xr+bs*xs',
                  'xo=cp*xw+cq*xz'],
    output_vars = ['xo'],
    dataset_size = 5000,
    learning_rate = 1e-3,
    training_iterations = 40000,
    batch_size = 8,
    display_loss_how_often = 100,
    debug = True,
)

cgp.parse_multi_layer_expressions()
cgp.gen_training_data()
sgd_loss = cgp.run_training_loop_multi_neuron_model()

cgpplus.parse_multi_layer_expressions()

```



```
cgpplus.gen_training_data()
cgpplus.set_mu(0.99)
sgdplus_loss=cgpplus.run_training_loop_multi_neuron_model()

plt.figure()
plt.plot(sgd_loss, label = 'SGD Training Loss')
plt.plot(sgdplus_loss, label = 'SGD+ Training Loss')
plt.legend()
plt.show()
```

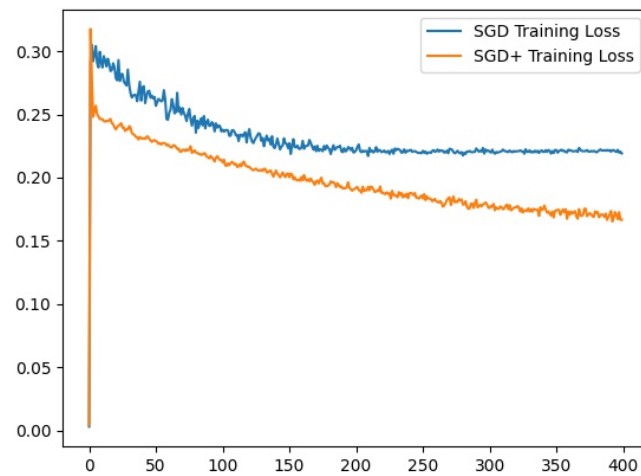


Figure 1: one_neuron_loss.jpg

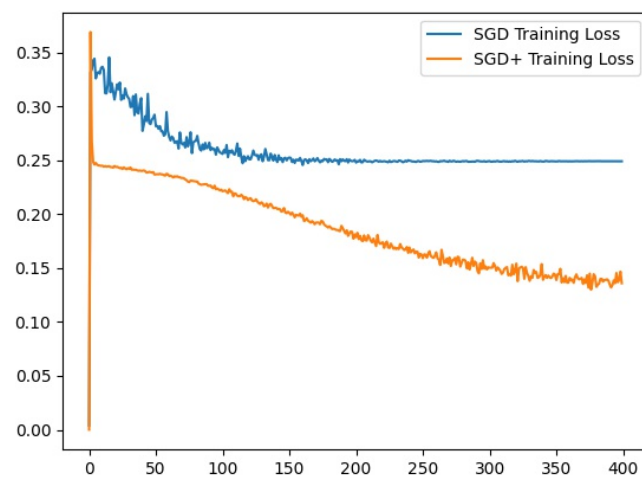


Figure 2: multi_neuron_loss.jpg