

NNC_Final_Project

December 15, 2019

Code

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[4]: import numpy as np

class Model:
    def __init__(self, n0, n1, n2):
        self.n_0 = n0
        self.n_1 = n1
        self.n_2 = n2

        self.reinitialize_weights = True

        # weights and biases initialized as placeholders.
        # reinitialized later
        self.weights_1 = np.zeros((self.n_0, self.n_1), dtype=int)
        self.weights_2 = np.zeros((self.n_1, self.n_2), dtype=int)
        self.biases_1 = np.zeros((1, self.n_1), dtype=int)
        self.biases_2 = np.zeros((1, self.n_2), dtype=int)

        self.activations_1 = np.zeros((1, self.n_1), dtype=int)
        self.activations_2 = np.zeros((1, self.n_2), dtype=int)
        self.sensitivities_1 = np.zeros((1, self.n_1), dtype=int)
        self.sensitivities_2 = np.zeros((1, self.n_2), dtype=int)

        self.hyper_params = HyperParams()

        self.model_info = Info()

class HyperParams:
    def __init__(self):
        self.alpha_list = [0.1, 0.2, 0.3]
        self.zeta_list = [0.5, 1, 1.5]
        self.x0_list = [0.5, 1, 1.5]
        self.max_epochs = 700 # empirically chosen
        self.tolerance = 0.05
        self.learning_rate = self.alpha_list[1]
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        self.zeta = self.zeta_list[1]
        self.x0 = self.x0_list[1]
        self.cost_fn = 0 # {0: quadratic, 1: cross-entropy}

class Info:
    def __init__(self, total_epochs=0, last_epoch_error=0.0, convergence=False):
        self.total_epochs_req = total_epochs
        self.last_epoch_error = last_epoch_error
        self.converged = convergence

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[3]: import pandas as pd

sheets = {}

def update_sheet(writer, sheet_name, sheet_obj):
    df_obj = {
        'Model architecture': sheet_obj['model_arch_list'],
        'Model weights': sheet_obj['model_weight_list'],
        'Model biases': sheet_obj['model_bias_list'],
        '# Training epochs': sheet_obj['total_epochs_req_list'],
        'Learning Rate': sheet_obj['learning_rate_list'],
        'Zeta': sheet_obj['zeta_list'],
        'X0': sheet_obj['x0_list'],
        'Cost Function': sheet_obj['cost_fn_list'],
        'Last epoch error': sheet_obj['last_epoch_error_list'],
        'Did converge?': sheet_obj['converged_list'],
    }
    df = pd.DataFrame(df_obj)
    df.to_excel(writer, sheet_name=sheet_name, index=False)

def save_data(sheet_name, model):
    try:
        sheet_obj = sheets[sheet_name]
    except KeyError:
        sheet_obj = {}
    if not sheet_obj:
        sheet_obj = {'model_arch_list': [], 'model_weight_list': [],
→ 'model_bias_list': [],
                        'total_epochs_req_list': [], 'learning_rate_list': [],
→ 'zeta_list': [],
                        'x0_list': [], 'cost_fn_list': [], 'last_epoch_error_list':
→ [],
                        'converged_list': []}

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        sheet_obj['model_arch_list'].append("[ " + str(model.n_0) + ", " + str(model.
→n_1) +
                                ", " + str(model.n_2) + "]" )
        sheet_obj['model_weight_list'].append("Weights1 = " + str(model.weights_1) +
                                "\nWeights2 = " + str(model.weights_2))
        sheet_obj['model_bias_list'].append("Biases1 = " + str(model.biases_1) +
                                "\nBiases2 = " + str(model.biases_2))
        sheet_obj['total_epochs_req_list'].append(model.model_info.total_epochs_req)
        sheet_obj['learning_rate_list'].append(model.hyper_params.learning_rate)
        sheet_obj['zeta_list'].append(model.hyper_params.zeta)
        sheet_obj['x0_list'].append(model.hyper_params.x0)
        sheet_obj['cost_fn_list'].append("Quadratic" if model.hyper_params.cost_fn_
→== 0
                                else "Cross-Entropy")
        sheet_obj['last_epoch_error_list'].append(model.model_info.last_epoch_error)
        sheet_obj['converged_list'].append("Yes" if model.model_info.converged else
→"No")
        sheets[sheet_name] = sheet_obj

def export_data():
    print("Starting export")
    writer = pd.ExcelWriter('Results.xlsx', engine='xlsxwriter')
    if not writer:
        print("Error while opening writer. Exiting.")
        return
    for sheet_name in sheets:
        sheet_obj = sheets[sheet_name]
        if not sheet_obj:
            print("Skipping sheet - %s " % sheet_name)
            continue
        print("Updating sheet - %s " % sheet_name)
        update_sheet(writer, sheet_name, sheet_obj)
    writer.save()
    print("Data exported")

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[1]: import getopt
import sys
import numpy as np

from model import Model, HyperParams
from save import save_data, export_data

export_to_excel = False

def transfer_ftn(n_l, x0):

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a_l = np.tanh(n_l / (2 * x0))
return a_l

# we only save a_l NOT n_l if using bipolar sigmoid transfer function
def derivative_transfer_ftn(a_l, x0):
    derivative = ((1 + a_l) * (1 - a_l)) / (2 * x0)
    return derivative

def init_weights_biases(model):
    if model.reinitialize_weights:
        model.weights_1 = np.random.uniform(-1 * model.hyper_params.zeta,
                                              model.hyper_params.zeta, model.
→weights_1.shape)

        model.weights_2 = np.random.uniform(-1 * model.hyper_params.zeta,
                                              model.hyper_params.zeta, model.
→weights_2.shape)

        model.biases_1 = np.random.uniform(-1 * model.hyper_params.zeta,
                                             model.hyper_params.zeta, model.
→biases_1.shape)

        model.biases_2 = np.random.uniform(-1 * model.hyper_params.zeta,
                                             model.hyper_params.zeta, model.
→biases_2.shape)

    return [model.weights_1, model.weights_2], [model.biases_1, model.biases_2]

def train_nn(x_train, y_train, model):
    Q = len(x_train)
    weight_list, bias_list = init_weights_biases(model)
    weight_list_len = len(weight_list)
    for epoch in range(model.hyper_params.max_epochs):
        epoch_error = 0
        for iteration in range(Q):
            x = x_train[iteration]
            y = y_train[iteration]
            x = np.array(x).reshape((1, len(x)))
            y = np.array(y).reshape((1, len(y)))

            # Calculate activations for all layers
            # don't need to save n_l if we are using bipolar sigmoid transfer_
→function
            a_l_list = [x]

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for i in range(len(weight_list)):
    n_l = np.matmul(a_l_list[-1], weight_list[i]) + bias_list[i]
    a_l = transfer_ftn(n_l, model.hyper_params.x0)
    a_l_list.append(a_l)

    # calculating the error for this example
    y_hat = a_l_list[-1] # activation of the last layer
    example_error = np.matmul(y_hat - y, (y_hat - y).T)
    example_error = np.asscalar(example_error)
    epoch_error = epoch_error + example_error

    # Calculate sensitivities for last layer. Performs element-wise
    →multiplication.
    # quadratic cost function
    if model.hyper_params.cost_fn == 0:
        s_L = np.multiply((y_hat - y), derivative_transfer_ftn(y_hat,
    →model.hyper_params.x0))
    # cross entropy cost function
    elif model.hyper_params.cost_fn == 1:
        s_L = y_hat - y

    # Calculate sensitivities for other layers
    sensitivities_list = [s_L]

    for l in range(weight_list_len - 1, 0, -1):
        s_l = np.multiply(np.matmul(sensitivities_list[0],
    →weight_list[l].T), \
                                derivative_transfer_ftn(a_l_list[l], model.
    →hyper_params.x0))
        sensitivities_list.insert(0, s_l)

    # Update weights and biases
    for l in range(weight_list_len):
        weight_list[l] = weight_list[l] - \
            (model.hyper_params.learning_rate *
            np.matmul(a_l_list[l].T,
    →sensitivities_list[l]))

        bias_list[l] = bias_list[l] - \
            (model.hyper_params.learning_rate *
    →sensitivities_list[l])

    # epoch error is not normalized (not divided by number of examples)
    if epoch_error < model.hyper_params.tolerance:
        break

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num_training_epochs = epoch + 1
if num_training_epochs < model.hyper_params.max_epochs:
    convergence = True
else:
    convergence = False

    update_model_info(model, weight_list, bias_list, num_training_epochs,
→epoch_error, convergence)

    return model

def update_model_info(model, weight_list, bias_list, num_training_epochs,
→epoch_error, convergence):
    model.weights_1 = weight_list[0]
    model.weights_2 = weight_list[1]
    model.biases_1 = bias_list[0]
    model.biases_2 = bias_list[1]
    model.model_info.total_epochs_req = num_training_epochs
    model.model_info.last_epoch_error = epoch_error
    model.model_info.converged = convergence

def extract_model_info(model, sheet_name, verbose=True):
    if verbose:
        □
→print("-----")
        print(f"Learning rate = {model.hyper_params.learning_rate} | "
              f"Zeta = {model.hyper_params.zeta} | "
              f"x0 = {model.hyper_params.x0}")
        print(f"Convergence = {model.model_info.converged} | "
              f"Training Epochs = {model.model_info.total_epochs_req} | "
              f"Squared Error = {model.model_info.last_epoch_error}")
        □
→print("-----")

    if export_to_excel:
        save_data(sheet_name, model)

def part_2a(x_train, y_train, model, sheet_name):
    # TODO
    # Look for patterns when do we get non-convergent results
    # Try all 3X3X3=27 hyper parameter combinations of alpha, zeta and x0
    num_convergence = 0
    for alpha in model.hyper_params.alpha_list:
        for zeta in model.hyper_params.zeta_list:

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        for x0 in model.hyper_params.x0_list:
            model.hyper_params.learning_rate = alpha
            model.hyper_params.zeta = zeta
            model.hyper_params.x0 = x0
            model = train_nn(x_train, y_train, model)
            if model.model_info.converged:
                num_convergence += 1
            extract_model_info(model, sheet_name, verbose=True)

    print("-----")
    print(f"Number of convergent hyper parameter combinations = {num_convergence} (out of 27)")

def part_2b(x_train, y_train, cost_fn, sheet_name):
    N1_list = [1, 2, 4, 6, 8, 10]
    convergence_list = []
    for i in range(len(N1_list)):
        model = Model(2, N1_list[i], 1)
        model.hyper_params.cost_fn = cost_fn
        num_convergence = 0
        for iters in range(100):
            model = train_nn(x_train, y_train, model)
            if model.model_info.converged:
                num_convergence += 1
            extract_model_info(model, sheet_name, verbose=False)
        convergence_list.append(num_convergence)
        print(f"Convergence for N1 = {N1_list[i]} -> {num_convergence}")

    print(f"Convergence results for N1 = [1,2,4,6,8,10] (out of 100): {convergence_list}")

    # Results mostly converge for N1=4 and above. For N1=2, almost 70% of the
    times,
    # it converges. For N1=1, it doesn't converge at all.
    # This is probably because the XOR problem is not linearly separable and we
    need a higher
    # number of neurons in the hidden layer to approximate the function (see
    universality theorem).

def xor_weight_validation(x_train, y_train, model, sheet_name):
    model.hyper_params.max_epochs = 1

    # Setting initial weights and biases for xor weight validation
    model.weights_1 = np.array([[0.197, 0.3191, -0.1448, 0.3594],

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                                [0.3099, 0.1904, -0.0347, -0.4861]]).
→reshape(model.weights_1.shape)
    model.weights_2 = np.array([0.4919, -0.2913, -0.3979, 0.3581]).reshape(model.
→weights_2.shape)
    model.biases_1 = np.array([-0.3378, 0.2771, 0.2859, -0.3329]).reshape(model.
→biases_1.shape)
    model.biases_2 = np.array([-0.1401]).reshape(model.biases_2.shape)
    model.reinitialize_weights = False

    model = train_nn(x_train, y_train, model)

    print("W1=", model.weights_1, sep="\n")
    print("b1=", model.biases_1, sep="\n")
    print("W2=", model.weights_2, sep="\n")
    print("b2=", model.biases_2, sep="\n")

    extract_model_info(model, sheet_name, verbose=False)

    # np.savez('xor_weight_validation.npz', model=model)
    # results can be loaded from the xor_weight_validation.npz file by
→uncommenting the following
    # data = np.load('Part1_results.npz')
    # model = data['model']

def main():
    # # uncomment this if data needs to be stored in excel
    # global export_to_excel
    # export_to_excel = True

    x_train = [[1, 1], [1, -1], [-1, 1], [-1, -1]]
    y_train = [[-1], [1], [1], [-1]]

    model = Model(2, 4, 1)
    xor_weight_validation(x_train, y_train, model, sheet_name="XOR weights_
→validation")

    # using quadratic cost ftn
    model = Model(2, 4, 1)
    model.hyper_params.cost_fn = 0
    part_2a(x_train, y_train, model, sheet_name="A-Z-X0 variations (Quad)")
    part_2b(x_train, y_train, cost_fn=0, sheet_name="N1 variations (Quad)")

    # using cross entropy cost ftn
    model = Model(2, 4, 1)
    model.hyper_params.cost_fn = 1
    part_2a(x_train, y_train, model, sheet_name="A-Z-X0 variations (CrsEnt)")

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part_2b(x_train, y_train, cost_fn=1, sheet_name="N1 variations (CrsEnt)")

model = Model(2, 4, 1)
model.hyper_params.learning_rate = 0.2
model.hyper_params.zeta = 1.0
model.hyper_params.x0 = 1.0
model.hyper_params.cost_fn = 1
model.hyper_params.max_epochs = 1
model = train_nn(x_train, y_train, model)
extract_model_info(model, sheet_name="Final verification")

# should be set to true above
if export_to_excel:
    export_data()

if __name__ == "__main__":
    main()

```

```

/usr/local/lib/python3.6/dist-packages/ipykernel_launcher.py:62:
DeprecationWarning: np.asscalar(a) is deprecated since NumPy v1.16, use a.item()
instead

```

```

W1=
[[ 0.19347555  0.31675476 -0.144748    0.36374521]
 [ 0.30686735  0.18845288 -0.03301594 -0.48859019]]

```

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b1=
[[-0.32243413  0.26504268  0.27330512 -0.32503622]]

```

```

W2=
[[ 0.47534885]
 [-0.27642811]
 [-0.38395025]
 [ 0.34801327]]

```

```

b2=
[[-0.08027444]]

```

```

-----
Learning rate = 0.1 | Zeta = 0.5 | x0 = 0.5
Convergence = True | Training Epochs = 144 | Squared Error =
0.049873341492876956
-----

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```

-----
Learning rate = 0.1 | Zeta = 0.5 | x0 = 1
Convergence = False | Training Epochs = 700 | Squared Error =
0.055927975170455696
-----

```

```

-----
Learning rate = 0.1 | Zeta = 0.5 | x0 = 1.5
Convergence = False | Training Epochs = 700 | Squared Error = 4.0477393613429244
-----

```

Learning rate = 0.1 | Zeta = 1 | x0 = 0.5
Convergence = True | Training Epochs = 97 | Squared Error = 0.049637115432790604

Learning rate = 0.1 | Zeta = 1 | x0 = 1
Convergence = True | Training Epochs = 640 | Squared Error = 0.04986490264371217

Learning rate = 0.1 | Zeta = 1 | x0 = 1.5
Convergence = False | Training Epochs = 700 | Squared Error = 0.3720739550348952

Learning rate = 0.1 | Zeta = 1.5 | x0 = 0.5
Convergence = False | Training Epochs = 700 | Squared Error = 4.000705200611622

Learning rate = 0.1 | Zeta = 1.5 | x0 = 1
Convergence = True | Training Epochs = 458 | Squared Error =
0.049987887286957175

Learning rate = 0.1 | Zeta = 1.5 | x0 = 1.5
Convergence = False | Training Epochs = 700 | Squared Error =
0.14315651838094595

Learning rate = 0.2 | Zeta = 0.5 | x0 = 0.5
Convergence = False | Training Epochs = 700 | Squared Error = 4.862384115264075

Learning rate = 0.2 | Zeta = 0.5 | x0 = 1
Convergence = False | Training Epochs = 700 | Squared Error = 4.204958347834154

Learning rate = 0.2 | Zeta = 0.5 | x0 = 1.5
Convergence = False | Training Epochs = 700 | Squared Error = 4.094065865517321

Learning rate = 0.2 | Zeta = 1 | x0 = 0.5
Convergence = True | Training Epochs = 52 | Squared Error = 0.048779595735270326

Learning rate = 0.2 | Zeta = 1 | x0 = 1
Convergence = True | Training Epochs = 408 | Squared Error = 0.04991055755562201

Learning rate = 0.2 | Zeta = 1 | x0 = 1.5

Convergence = True | Training Epochs = 631 | Squared Error = 0.04988909483521971

Learning rate = 0.2 | Zeta = 1.5 | x0 = 0.5

Convergence = True | Training Epochs = 69 | Squared Error = 0.04907998529277527

Learning rate = 0.2 | Zeta = 1.5 | x0 = 1

Convergence = True | Training Epochs = 224 | Squared Error = 0.04987026445288031

Learning rate = 0.2 | Zeta = 1.5 | x0 = 1.5

Convergence = True | Training Epochs = 553 | Squared Error = 0.04999293222262656

Learning rate = 0.3 | Zeta = 0.5 | x0 = 0.5

Convergence = True | Training Epochs = 54 | Squared Error = 0.049039109630381855

Learning rate = 0.3 | Zeta = 0.5 | x0 = 1

Convergence = False | Training Epochs = 700 | Squared Error = 4.310545405999978

Learning rate = 0.3 | Zeta = 0.5 | x0 = 1.5

Convergence = False | Training Epochs = 700 | Squared Error = 4.138021047560265

Learning rate = 0.3 | Zeta = 1 | x0 = 0.5

Convergence = True | Training Epochs = 31 | Squared Error = 0.04995314267186744

Learning rate = 0.3 | Zeta = 1 | x0 = 1

Convergence = True | Training Epochs = 185 | Squared Error =
0.049735165791450764

Learning rate = 0.3 | Zeta = 1 | x0 = 1.5

Convergence = True | Training Epochs = 605 | Squared Error = 0.04984200046100855

Learning rate = 0.3 | Zeta = 1.5 | x0 = 0.5

Convergence = True | Training Epochs = 27 | Squared Error = 0.04787247678823797

Learning rate = 0.3 | Zeta = 1.5 | x0 = 1

Convergence = True | Training Epochs = 137 | Squared Error =
0.049822583171420486

Learning rate = 0.3 | Zeta = 1.5 | x0 = 1.5
Convergence = True | Training Epochs = 391 | Squared Error = 0.04992550078242809

Number of convergent hyper parameter combinations = 17 (out of 27)
Convergence for N1 = 1 -> 0
Convergence for N1 = 2 -> 82
Convergence for N1 = 4 -> 100
Convergence for N1 = 6 -> 100
Convergence for N1 = 8 -> 100
Convergence for N1 = 10 -> 98
Convergence results for N1 = [1,2,4,6,8,10] (out of 100): [0, 82, 100, 100, 100, 98]

Learning rate = 0.1 | Zeta = 0.5 | x0 = 0.5
Convergence = True | Training Epochs = 130 | Squared Error = 0.04818403472549024

Learning rate = 0.1 | Zeta = 0.5 | x0 = 1
Convergence = True | Training Epochs = 273 | Squared Error = 0.04920549727660134

Learning rate = 0.1 | Zeta = 0.5 | x0 = 1.5
Convergence = True | Training Epochs = 537 | Squared Error = 0.049217856264777524

Learning rate = 0.1 | Zeta = 1 | x0 = 0.5
Convergence = True | Training Epochs = 59 | Squared Error = 0.04812216331925205

Learning rate = 0.1 | Zeta = 1 | x0 = 1
Convergence = True | Training Epochs = 189 | Squared Error = 0.048827557727704765

Learning rate = 0.1 | Zeta = 1 | x0 = 1.5
Convergence = True | Training Epochs = 481 | Squared Error = 0.04918584143788117

Learning rate = 0.1 | Zeta = 1.5 | x0 = 0.5
Convergence = True | Training Epochs = 32 | Squared Error = 0.04730144103192096

Learning rate = 0.1 | Zeta = 1.5 | x0 = 1
Convergence = True | Training Epochs = 93 | Squared Error = 0.04935925793673039

Learning rate = 0.1 | Zeta = 1.5 | x0 = 1.5
Convergence = True | Training Epochs = 169 | Squared Error = 0.04935280121677305

Learning rate = 0.2 | Zeta = 0.5 | x0 = 0.5
Convergence = True | Training Epochs = 48 | Squared Error = 0.04729646661090205

Learning rate = 0.2 | Zeta = 0.5 | x0 = 1
Convergence = False | Training Epochs = 700 | Squared Error = 4.4181942951797675

Learning rate = 0.2 | Zeta = 0.5 | x0 = 1.5
Convergence = False | Training Epochs = 700 | Squared Error = 4.275074610028772

Learning rate = 0.2 | Zeta = 1 | x0 = 0.5
Convergence = True | Training Epochs = 30 | Squared Error = 0.04450862905880065

Learning rate = 0.2 | Zeta = 1 | x0 = 1
Convergence = True | Training Epochs = 61 | Squared Error = 0.04999468286965289

Learning rate = 0.2 | Zeta = 1 | x0 = 1.5
Convergence = True | Training Epochs = 143 | Squared Error =
0.049281296908125924

Learning rate = 0.2 | Zeta = 1.5 | x0 = 0.5
Convergence = True | Training Epochs = 15 | Squared Error = 0.04682004586369068

Learning rate = 0.2 | Zeta = 1.5 | x0 = 1
Convergence = True | Training Epochs = 56 | Squared Error = 0.04877838809480512

Learning rate = 0.2 | Zeta = 1.5 | x0 = 1.5
Convergence = True | Training Epochs = 174 | Squared Error =
0.048719322156361364

Learning rate = 0.3 | Zeta = 0.5 | x0 = 0.5
Convergence = False | Training Epochs = 700 | Squared Error = 5.302362039146319

Learning rate = 0.3 | Zeta = 0.5 | x0 = 1

```

Convergence = False | Training Epochs = 700 | Squared Error = 4.638025288751011
-----
Learning rate = 0.3 | Zeta = 0.5 | x0 = 1.5
Convergence = False | Training Epochs = 700 | Squared Error = 4.41819429503458
-----
Learning rate = 0.3 | Zeta = 1 | x0 = 0.5
Convergence = True | Training Epochs = 15 | Squared Error = 0.04426839330093334
-----
Learning rate = 0.3 | Zeta = 1 | x0 = 1
Convergence = True | Training Epochs = 26 | Squared Error = 0.043646433081242504
-----
Learning rate = 0.3 | Zeta = 1 | x0 = 1.5
Convergence = True | Training Epochs = 118 | Squared Error = 0.04929856228031971
-----
Learning rate = 0.3 | Zeta = 1.5 | x0 = 0.5
Convergence = True | Training Epochs = 11 | Squared Error = 0.049505345251627336
-----
Learning rate = 0.3 | Zeta = 1.5 | x0 = 1
Convergence = True | Training Epochs = 31 | Squared Error = 0.049029234653004204
-----
Learning rate = 0.3 | Zeta = 1.5 | x0 = 1.5
Convergence = True | Training Epochs = 82 | Squared Error = 0.04785224419009246
-----
Number of convergent hyper parameter combinations = 22 (out of 27)
Convergence for N1 = 1 -> 0
Convergence for N1 = 2 -> 59
Convergence for N1 = 4 -> 90
Convergence for N1 = 6 -> 96
Convergence for N1 = 8 -> 97
Convergence for N1 = 10 -> 100
Convergence results for N1 = [1,2,4,6,8,10] (out of 100): [0, 59, 90, 96, 97, 100]
-----
Learning rate = 0.2 | Zeta = 1.0 | x0 = 1.0
Convergence = False | Training Epochs = 1 | Squared Error = 5.3801314005663095
-----

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[]: