

CSCI964 Computational Intelligence: Lab #1

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Task 1

```
#include <iostream>
#include <eigen3/Eigen/Core>

using namespace std;
using namespace Eigen;

double SigmoidFunc(double x)
{
    return 1.0 / (1.0 + exp(-x));
}

void Sigmoid(VectorXf &src, VectorXf &dst)
{
    float *src_data = src.data();
    float *dst_data = dst.data();
    for (int i = 0; i < src.size(); ++i)
    {
        dst_data[i] = SigmoidFunc(src_data[i]);
    }
}

double forwardProp(int numberOfLayers, MatrixXd input, MatrixXd *weights, MatrixXd *thetas)
{
    MatrixXd output = input;
    // output.transposeInPlace();
    for (int i = 0; i < numberOfLayers; i++)
    {
        output = (output * weights[i] - thetas[i]).unaryExpr(&SigmoidFunc);
    }
    return output(0);
}

double gradDst(int batch_size, int layernumber, int dataset_size, double LR, double maxerror,
               int maxt, MatrixXd *x, MatrixXd *weights, MatrixXd *thetas, double *y)
{
    int times = 0, ctr = 0, idx = 0;
    double error = 100, sumerror = 0;
    MatrixXd deltaW(3, 1);
    double deltaTheta = 0;
    deltaW << 0, 0, 0;
    while (error > maxerror && times < maxt)
    {
        idx = (idx + 1) % dataset_size;
        double t = forwardProp(layernumber, x[idx], weights, thetas);
        deltaW += LR * (t - y[idx]) * t * (1 - t) * x[idx].transpose();
        deltaTheta += LR * (t - y[idx]) * t * (1 - t);
        ctr = (ctr + 1) % batch_size;
        sumerror += 0.5 * (t - y[idx]) * (t - y[idx]);
        if (ctr == 0)
        {
            weights[layernumber - 1] -= deltaW;
        }
    }
}
```

```
        thetas[layernumber - 1](0) += deltaTheta;
        deltaW << 0, 0, 0;
        deltaTheta = 0;
    }
    times++;
    if (idx == 0)
    {
        error = sumerror / dataset_size;
        sumerror = 0;
    }

    cout << "epochs: " << times << endl;
    cout << "error: " << error << endl;
    cout << "w" << weights[0] << endl;
    cout << "theta: " << thetas[0] << endl;
}
return error;
}

int main()
{
    MatrixXd x0(1, 3), x1(1, 3), x2(1, 3), x3(1, 3);
    MatrixXd weight(3, 1), bias(1, 1);
    x0 << 0, 0, 1;
    x1 << 0, 1, 1;
    x2 << 1, 0, 1;
    x3 << 1, 1, 1;
    weight << 6, 0, 0;
    bias << 2;
    double y0 = 0, y1 = 0, y2 = 1, y3 = 1;
    MatrixXd x[] = {x0, x1, x2, x3};
    double y[] = {y0, y1, y2, y3};
    MatrixXd biases[] = {bias};
    MatrixXd weights[] = {weight};
    //gradDst(1, 1, 4, 0.01, 0.001, 1000000, x, weights, biases, y); // online learning, set
    // batch to 1
    gradDst(3, 1, 4, 0.01, 0.001, 1000000, x, weights, biases, y); // batch method , set batch to
    1
}
```

```
epochs: 117690
error: 0.00100001
w  8.48812
   -0.116891
   -2.0642
theta: 2.0642

epochs: 117691
error: 0.00100001
w  8.48814
   -0.116871
   -2.06418
theta: 2.06418

epochs: 117692
error: 0.000999998
w  8.48814
   -0.116871
   -2.06421
theta: 2.06421
```

Figure 1: Online Learning Method

```
epochs: 119996
error: 0.00100001
w  8.51003
   -0.116895
   -2.06968
theta: 2.06968

epochs: 119997
error: 0.00100001
w  8.51003
   -0.116895
   -2.06968
theta: 2.06968

epochs: 119998
error: 0.000999997
w  8.51003
   -0.116895
   -2.06968
theta: 2.06968
```

Figure 2: Batch Method (batchsize = 100)

```
epochs: 152373
error: 0.00100001
w  8.76612
   -0.117385
   -2.13364
theta: 2.13364

epochs: 152374
error: 0.001
w  8.76612
   -0.117385
   -2.13364
theta: 2.13364

epochs: 152375
error: 0.000999997
w  8.76612
   -0.117385
   -2.13364
theta: 2.13364
```

Figure 3: Batch Method (batchsize = 200)