Al Assignment #5 Report

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4. Testing Network:

For this part the corresponding script was used.

```
- code_sajjad pythond_neural_net_texter.py_staple

Training on OR_date
train: 1869|
18080/18080 [08:82-08:80, 4836.881t/s]
weights: (slA(11.17), w2A(11.17), w4(6.93))
Trainod weights:
Weight ***\text{1.17290}
Weight **\text{1.17290}
Texting on Retest data
Inst((0.1, 0.1, 0)) returnes: 0.88791316473484] **> 0 [correct]
text((0.1, 0.1, 0)) returnes: 0.88791316473484] **> 1 [correct]
text((0.9, 0.1, 1)) returnes: 0.9857875715894018 **> 1 [correct]
text((0.9, 0.1, 1)) returnes: 0.9857875715894018 **> 1 [correct]
Accuracy: 1.080400

Training on AMD data
train: 1095|
10000/10000 [00:02-00:00, 4886.961t/s]
weights: (slA(30.50), wA(10.50), wA(14.37)]
Trained weights:
Weight **\text{1.4}(1.8.50), wA(10.50), wA(14.37)]
Trained weights:
Weight **\text{1.4}(1.8.50) vA(10.50), wA(14.37)]
Trained weights:
Weight **\text{1.4}(1.8.50) vA(10.50), wA(14.37)]
Texting on AMD text-data
text((0.1, 0.1, 0)) returnes: 0.0245240502173127 **> 0 [correct]
text((0.1, 0.9, 0)) returnes: 0.0245240502173127 **> 0 [correct]
text((0.1, 0.9, 0)) returnes: 0.0245240502173127 **> 0 [correct]
text((0.9, 0.9, 1)) returnes: 0.0345240502173127 **> 0 [correct]
```

5. Finite difference:

For this part member function finite_difference(self) was implemented for class Network using which correctness of derivatives implementations was tested by simple test

```
ed python3 neural net tester.py simple
 fraining on OR data
train:: 100%|
                                                                                                                                                                                                                                             10000/10000 [00:02<00:00, 3805.441t/s]
weights: [vtA(11.17), v2A(11.17), vA(6.93)]
passed finite difference test with acc = 1e 68
 rained weights
Weight 'wlA': 11.172780
Weight 'w2A': 11.173843
Weight 'w4: 6.933418
 lesting on OR test-data
test((6.1, 8.1, 8)) returnes: 8.889823648728838386 -> 8 [correct]
test((8.1, 8.9, 1)) returnes: 8.9837913164734841 -> 1 [correct]
test((8.9, 8.1, 1)) returnes: 8.9857875715894888 -> 1 [correct]
test((8.9, 8.9, 1)) returned: 8.999988187788629 => 1 [correct]
Accuracy: 1.888889
                                                                                                                                                                                                                                            | 10000/10000 [00:03<00:00, 3318.701t/s]
 eights: [VIA(10.58), V2A(10.58), VA(14.37)]
 basses finite difference test with acc = le-88
 Trained weights:
Weight 'w1A': 10,499844
Weight 'w2A': 16,499561
Weight 'wA': 14,356979
Weight 'WA : 14.350979
Testing on AMD test-data
Testi(0.1, 0.1, 0.1) returned: 4.784254617957318e-86 => 8 [correct]
Testi(0.1, 0.9, 0.1) returned: 0.020484490369173127 => 8 [correct]
Testi(0.1, 0.1, 0.1) returned: 0.0204890380372059 => 0 [correct]
Testi(0.1, 0.1, 0.1) returned: 0.0204890380372059 => 1 [correct]
```

6. Two Layer Neural Network:

Using API provided by neural_net.py the requested network was implemented.

OR:

```
Training on OR data
train; 100%|
weights: [viA(-4.17), v2A(-4.15), vA(-2.08), w18(-5.25), w28(-5.25), w8(-2.04), wAC(-5.94), w8C(-0.10)]
Trained weights:
Weight 'w1A': -4.106633
Weight 'w2A': 4.149468
Weight 'w3A': -5.281165
Weight 'w3B': -5.234993
Weight 'w3B': -5.234993
Weight 'w3B': -5.234998
Weight 'w6': -8.209305
Weight 'w6': -8.209305
Weight 'w6': -8.155650
Testing on OR test-data
testi(0.1, 0.1, 0)) returned: 0.0030250790206227775 ⇒ 0 [correct]
testi(0.1, 0.9, 1)) returned: 0.9919040619425431 ⇒ 1 [correct]
testi(0.9, 0.1, 1)) returned: 0.9919040619425431 ⇒ 1 [correct]
testi(0.9, 0.9, 1)) returned: 0.9978001861300183 ⇒ 1 [correct]
Accuracy: 1.000000
```

AND:

```
Training on AMD data
                                                                                                                                                                                                            | 10000/10000 [00:09<00:00, 1092.361t/s]
weights: [w1Al-3.57], w2A(-3.61), wA(-4.72), w1B(4.82), w2B(3.98), wB(5.31), wAC(-7.77), wBC(8.86), wC(0.96)]
Trained weights
Weight 'w1A': -3.567641
Weight 'w2A': -3.600579
Weight
           'wA': -4.719848
Weight 'w10': 4.821667
Weight
           'w281: 3.981724
Weight 'w8': 5.386268
Weight
           'wAC': -7.765878
Weight 'v8C': 8.655540
Weight 'vC': 8.962835
Testing on AMD test-data test[(8.1, 8.1, 8.1) returned: 0.89828451042688453718 => 0 [correct]
\begin{array}{lll} test((\theta,1,\;\theta,9,\;\theta)) & returned: \;\theta,\theta\theta6783813285854222 \implies \theta \; [correct] \\ test((\theta,9,\;\theta,1,\;\theta)) & returned: \;\theta,\theta96789528452711841 \implies \theta \; [correct] \\ test((\theta,9,\;\theta,9,\;1)) & returned: \;\theta,995518965449981 \implies 1 \; [correct] \\ \end{array}
Accuracy: 1.668668
```

EQUAL:

```
Training on EQUAL data
Train:: 185%]
                                                                                                                                                                                                                   | 10800/10800 [08:05-08:08, 1728.7811/s]
 weights: [MIÅ(-6.68), MZA(-6.85), MA(-2.78), WIB(-5.81), WZB(-5.86), WB(-7.58), MAC(18.63), MBC(-10.51), WC(-5.81)]
Trained weights:
Weight 'WLA': -6.682529
Weight 'W2A': -6.845288
Weight 'WA': -2.782586
Weight WAB: -5.612516
Meight
             WZB":
                         5.656722
Meight
             MB :
                      -7.501407
Meight
             'wat': 18.631280
Weight
             "MUC": -10.505650
 Metght 'wC': -5,000055
Testing on FALM test-data
test((8.1, 8.1, 1)) returned: 8.9572879757678457 -> 1 [correct] test((8.1, 8.0, 9)) returned: 8.811226265758113587 -> 8 [correct] test((8.9, 8.1, 9)) returned: 8.811258471362683439 -> 8 [correct] test((8.9, 8.9, 1)) returned: 8.9681879694444555 -> 1 [correct] Accuracy: 1.898896
```

HORIZONTAL BAND:

VERTICAL BAND:

```
Training on vertical-banes data

train:: [804]

weights: [VIA(-4.63], w2A(8.67), wA(-11.29), v1B(-6.66), w2B(8.19], wB(-3.28), wAC(12.17), vBC(-11.88), wC(5.92)]

Trained weights:
Weight: w1A:: 4.631307

Weight: w1A:: -8.73538

Weight: w1A:: -11.20081

Weight: w2:: -1.1.20081

Weight: w2:: -3.106700

Weight: w2:: -3.
```

DIAGONAL BAND:

INVERSE DIAGONAL BAND:

```
Training on inverse diagonal band data

train: 1691

| 18008/18008 [88:24<80:80, 412.3811/s] |

Training weight: | kHA[-4.52], w2A[4.81], wA(5.95], w18[4.82], w2B(-4.35), wB(5.96], wAK[10.80], wK(5.101] |

Training weight: | wA(5.32) |

Weight 'wA(6.32) | wA(6.95) |

Weight 'wA(7.32) | wA(7.95) |

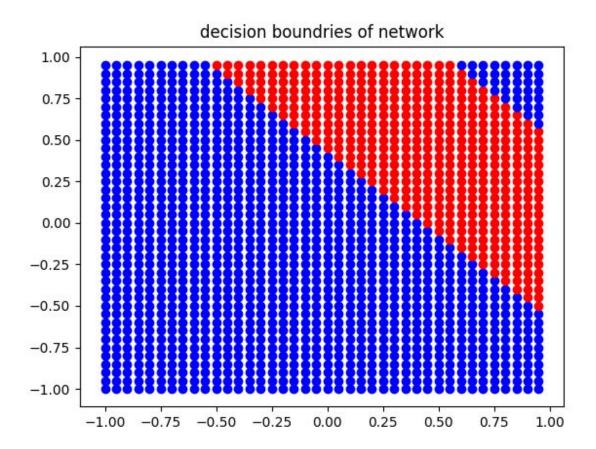
Weight 'wA(7.32) |
```

NOT EQUAL:

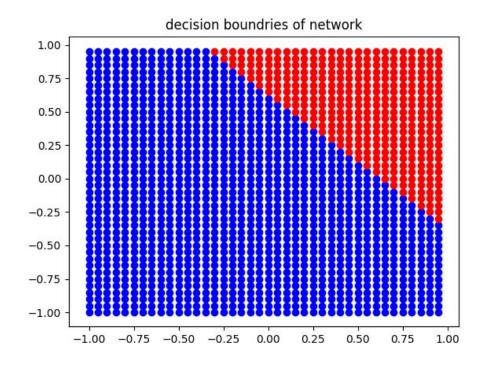
```
Training on NOT FQUAL data
train:: 1004|
weights: [vtA(-6.58), vvA(6.50), vA(3.50), vtB(-5.97), vVB(6.23), vB(-2.07), vAC(10.27), vBC(-10.01), vC(-4.79)|
Trained weights:
Weight 'wtA': -6.587623
Weight 'wtA': -6.587623
Weight 'wtA': 3.588471
Weight 'wtA': 3.588471
Weight 'wtA': 3.588471
Weight 'wtA': 2.973460
Weight 'wtA': 2.973460
Weight 'wtC: 19.996440
Weight 'wtC: 19.996440
Weight 'wtC: 4.788817
Testing on HOT EQUAL test data
test([0.1, 0.1, 0]) returned: 0.9018214787819529 => 0 [correct]
test([0.9, 0.1, 1]) returned: 0.9058289458612442 => 1 [correct]
test([0.9, 0.1, 1]) returned: 0.9058289458612442 => 0 [correct]
test([0.9, 0.1, 0]) returned: 0.90828494981703212 => 0 [correct]
test([0.9, 0.9, 0]) returned: 0.910428404981703212 => 0 [correct]
test([0.9, 0.9, 0]) returned: 0.910428404981703212 => 0 [correct]
Accuracy: 1.000000
```

7. Plot Decision Boundary:

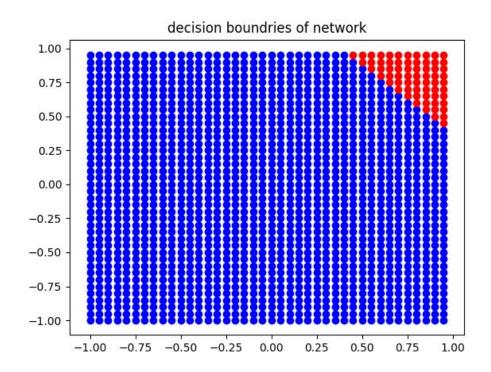
To plot the corresponding decision boundary plt.scatter was used. NOT EQ boundaries:



OR boundaries:

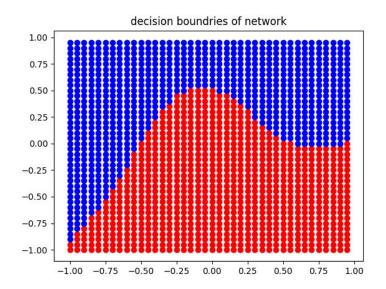


AND boundaries:



8. Overfitting and Regularization:

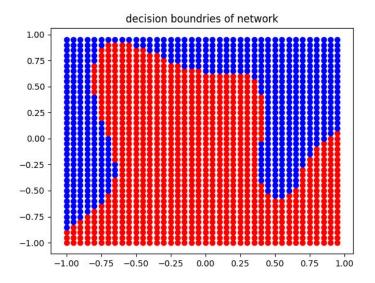
In short, Regularization in machine learning is the process of regularizing the parameters that constrain, regularizes, or shrinks the coefficient estimates towards zero. In other words, this technique discourages learning a more complex or flexible model, avoiding the risk of Overfitting. Iteration = 100



Iteration = 500

code_sajjad python3 neural net tester.py two_moons

Training on two-moons data
train:: 188%|
decision surface:: 1888it [08:08, 178.091t/s]
| 500/500 [02:11<00:08, 3.741t/



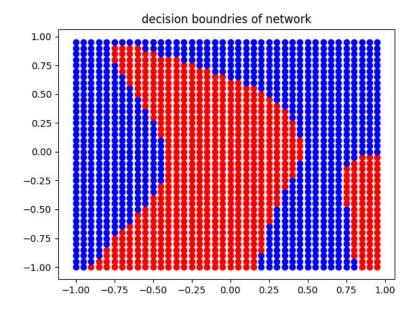
Iteration 1000:

Training on two-moons data

train: 1865|

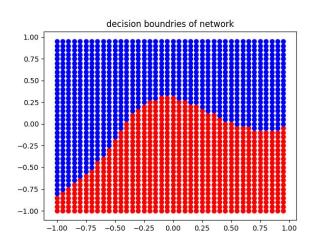
decision surface: 1888:t [88:88, 182.38it/s]

Accuracy: 8.866888



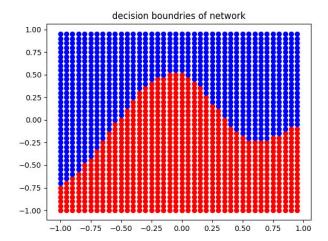
Iteration = 100 with regularization:





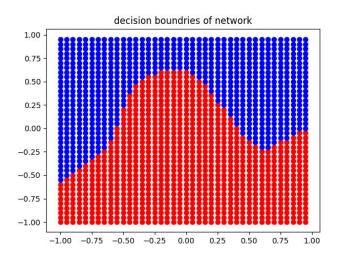
Iteration = 500 with regularization:





Iteration = 1000 with regularization:

```
| code_sajjad python3 <u>neural net tester.py</u> two moons
| raining on two-moons data
| 1000| | 1000/1000 | 04:57-00:00, 3.36it/s|
| constance: 1000it | 00:10, 157:11it/s|
| constance: 0.900000
```



In the first part as the number of iterations increased, model was overfitted on training data. It happens mainly for the extreme flexibility of neural networks. One of the signs of overfitting is complex decision boundaries which occurred in this part.

When regularization term is added, it makes parameters to constantly drift to 0 thus prevents overfitting by putting additional yet desirable constraints. As a result decisions boundaries would appear more smooth.

The problem of overfitting is obviously solved!