Figure 1: 2-D dataset

For the report, use confusion matrices to demonstrate the accuracy of your MLP, and discuss the methods you used to achieve the highest accuracy. How many nodes do you have on your hidden layer? How did changing the learning rate change your results? What role does increased different number of iterations have on accuracy?

NOTE: The summary of the answers for the questions given above are stated on the last page.

Convention: The above confusion matrix is for the training data and the confusion matrix below is for the Validation data. Since they are in random state, the accuracy and percentage tends to be different from time to time upon repetition, however sometimes increasing the number of iterations and ETA to an acceptable amount (Not less not too much) increases it's chance to be more accurate around 100%. A line starting with the word "Stopped" Indicates the errors where it stopped.

Implementation: From the code given, the earlystopping user defined function is used to give the code a limit the early stopping code uses the mlpfwd and mlptrain in its function s making this the primary code for generating more outputs to the class.. Then after generating those outputs and results, the confmat user defined function is used to check the accuracy of the confusion matrix generated. The input of the confmat is the train, the validation and the whole dataset

3 Hidden layers ,ETA= 0.01, #of iterations : 500

```
Stopped 2.2216926674567397 2.220698494182167 2.2204870340396066

Confusion matrix is:

[[26. 0.]

[14. 44.]]

Percentage Correct: 83.3333333333334

Confusion matrix is:

[[14. 0.]

[ 6. 16.]]

Percentage Correct: 83.3333333333334
```

3 Hidden layers ,ETA= 0.07, #of iterations : 1500

```
Stopped 0.06037995649669422 0.061339603401095434 0.062329914204604375

Confusion matrix is:

[[40. 0.]

[ 0. 44.]]

Percentage Correct: 100.0

Confusion matrix is:

[[20. 0.]

[ 0. 16.]]

Percentage Correct: 100.0
```

3 Hidden layers ,ETA=0.01, #of iterations :600

```
Stopped 2.3622207267415654 2.3632015414056653 2.3641935546529265

Confusion matrix is:

[[26. 0.]

[14. 44.]]

Percentage Correct: 83.33333333333334

Confusion matrix is:

[[14. 0.]

[ 6. 16.]]

Percentage Correct: 83.333333333333
```

3 Hidden layers ,ETA=0.1, #of iterations :600

```
Stopped 2.196985056354456 2.1977534898078472 2.198737143802448

Confusion matrix is:

[[26. 0.]

[14. 44.]]

Percentage Correct: 83.3333333333334

Confusion matrix is:

[[14. 0.]

[ 6. 16.]]

Percentage Correct: 83.33333333333334
```

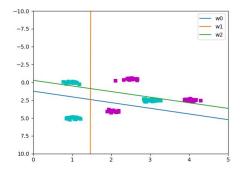
Absurdly increasing numbers(especially ETA) to a weird amount 3 hidden layers, ETA = 5, niterations=900

```
Stopped 8.0 8.0 8.0
Confusion matrix is:
[[40. 44.]
       [ 0. 0.]]
Percentage Correct: 47.61904761904761
Confusion matrix is:
[[20. 16.]
       [ 0. 0.]]
Percentage Correct: 55.55555555556
```

hidden layers= 150, # of iternations =3, ETA 0.1

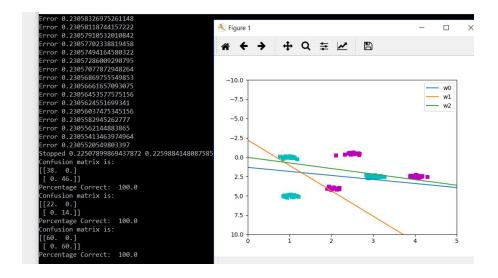
Discussing my results. So far i have 3 hidden layers . Where i normally would assume the first hidden layer would have 3 neurons to draw 3 lines. Well by the code we do not exactly have any option to know how much neurons are there. The learning rate (ETA) largely affects the results as it produces more error if the learning rate is too high or when the learning rate is too small it would not actually train the data set to an error of 0 depending on the number of iterations. The number of iterations say that how much number of epochs will the weights have to adjust itself. So far, depending on the learning rate, the number of iterations can help the weights adjust to an optimal value as long as the ETA is not too large so that the error graph would not oscillate to a result making it erroneous, making the confusion matrix output more errors. So far adjusting the ETA and the number of iterations to a "sweet spot" allows the program to train itself to produce as less error as possible. Where assumptively most of the successful number of iterations is around 1500 and above given you have 3 neurons in this problem as the ETA value is around 0.07 under the logistic output type. These are the results under logistic function. Unfortunately, i cannot provide a fully 100% Correct Percentage as the weight generation in every iteration could vary and is not actually the same compared to the previous runs of codes. So the percentage could potentially vary from time to time. From back propagation the figure below plots the weights in order to visualize how the code performed. WARNING: This is inconsistent because each run produces different percentage correctness

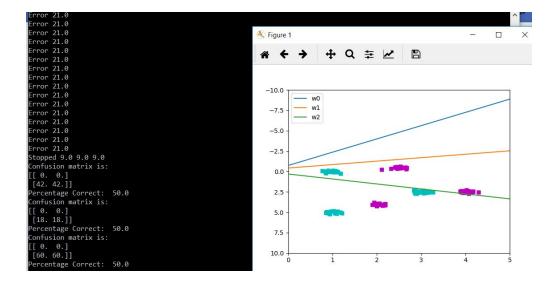
and different errors depending on how the backpropagation decided to perform or how random the testing /validation set was generated.



Taking the result of the most successful iteration and comparing the results in linear and softmax we get these results

LINEAR

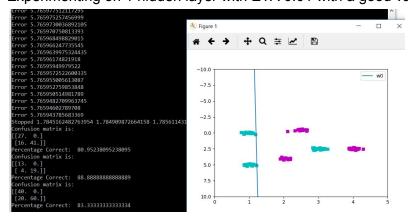




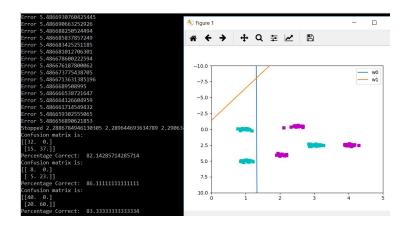
The results of softmax is not very cooperative due to it's equational structure and its update process. Since the update of the errors of softmax very minimal, something tells me softmax is a good function when you are trying to optimize or correct a certain value of weights that are already accurate however one is trying to adjust it to a better optimal value, not weight value starting from 0 or random.

Given that we used a hidden layer of 3 as intuition, normally if we have only 1 or 2, theoretically it will not give 100% accuracy.

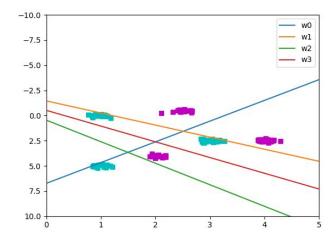
Experimenting on 1 hidden layer with ETA 0.01 with a good value of niterations probably 5000.



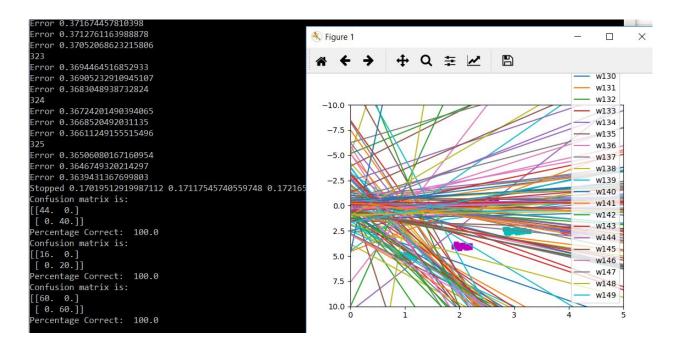
while having 2 neurons, the best performance it can give is similar to this figure



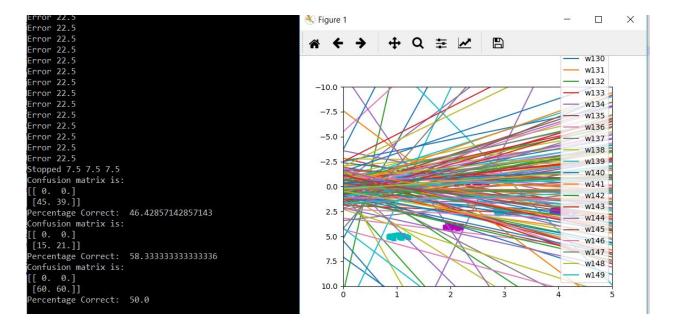
and experimenting with more neurons, the performance of the network can result to a bit of problem and sometimes takes more time to correct itself due to having a new set of weights to affect the system. Despite it did perform, it actually overperform and took more time to train itself compared to the sufficient amount of neurons needed. E.g. 4 neurons



and lastly the experiment results of having 150 neurons and 0.01 ETA done with 3 number of iterations to correct itself but sometimes alot of overlapping information can be found and the neural network configuration is too redundant.



Checking the results even if we try softmax, the results seems to be still inaccurate when using the softmax function



Summary of answers to the questions above:

How many nodes?

Check the figures but for simplicity, the basic i used 3 to atleast make the correct percentage of confusion matrix to be 100%... you need an amount of nodes greater than 3 to achieve 100% accuracy.

How does the learning rate affect the results? The learning rate ETA can affect the correctness of the results in a way that making it too small would need more number of iterations to achieve a good network. While having an ETA that is too big would tend to make your system less stable as if would have a hard time or never even achieve less than 1% error on the system. as your error keeps oscillating or jumping up and down instead of converging to a 0 error network.

What role does the number of iterations do?

It indicates how much and the limit to how many times will the system attempt to train its weights. So far the performance of the system will heavily base on the ETA than the nubmer of iterations so it actually will depend on the combination of both to achieve a good result on neural networks.

Extra: Softmax, Linear or Logistic.

So far from the observed usage of these output types. Linear tends to be straight and botherless compared to logistic or softmax. While logistic has a lesser time to correct itself but tends to be more controllable because of how slow it changes as the error decreases slowly compared to linear. While Softmax, in this application never really changes its percentage correctness and its weights doesn't change at all or maybe changes too small, i'd think the Softmax function is better when you have accurately selected the weights and attempted to use backpropagation just to correct it in very small manners.