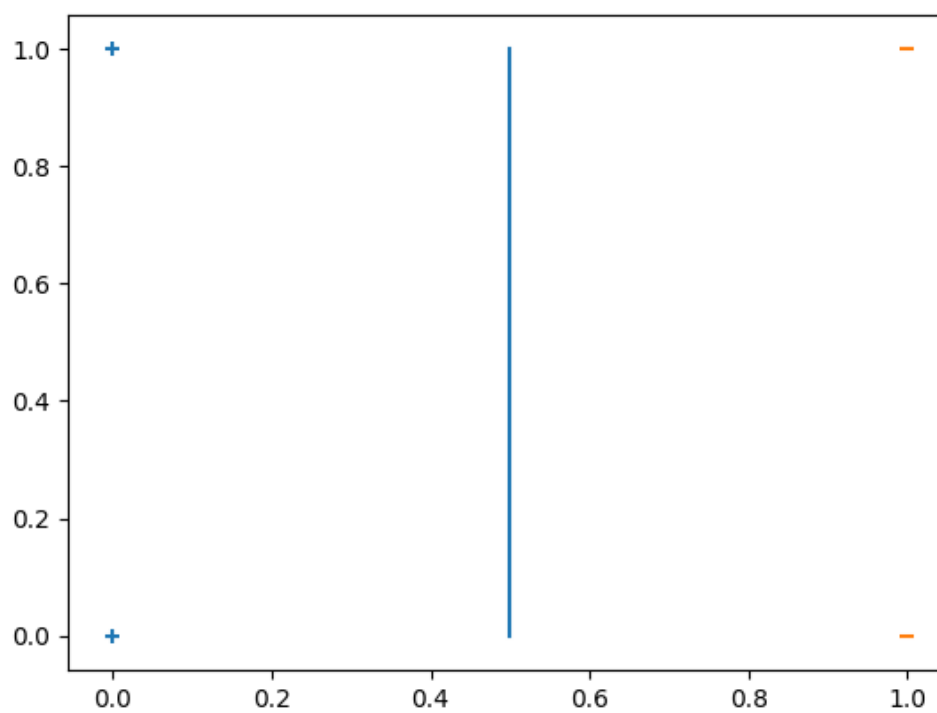


# Homework 1

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## 1 Perceptron Algorithm and Convergence Analysis

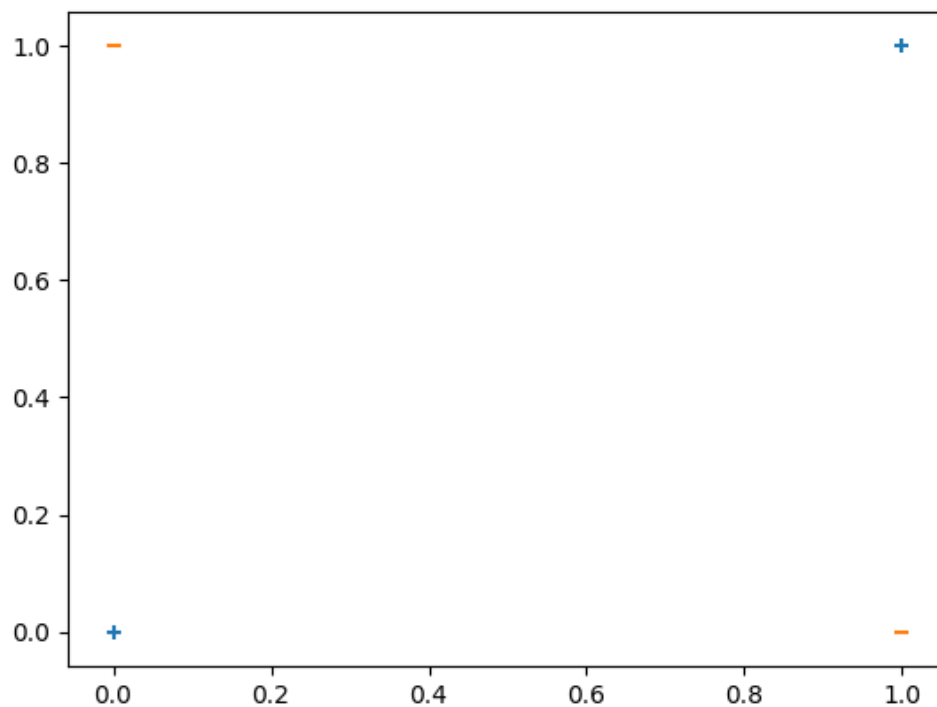
- (a)  $y = f(x_1, x_2) = x_1$



- (b) A truth table for this function could be:

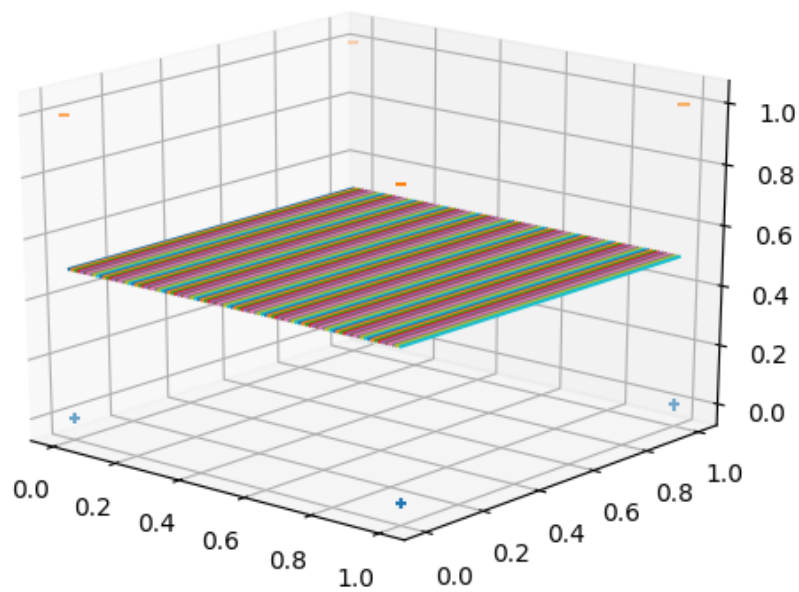
$x_1$	$x_2$	$y$
0	0	1
0	1	0
1	0	0
1	1	1

We can plot those points in a figure.



The perceptron algorithm gives us a weight vector which corresponds to its orthogonal hyperplane to divide points into 2 parts. In this figure since we can't find a hyperplane to divide the 2 kinds, this function can't be separated by single perceptron.

(c)  $y = f(x_1, x_2, x_3) = x_3$



2.  $\beta$  is a weight vector perpendicular to the hyperplane the signed Euclidean Distance of the point  $x$  to the hyperplane is given by:

$$d = \frac{(x - x_0) \cdot \beta}{\|\beta\|_2}$$

where

$$x_0 \in \{x | f(x) = 0\}$$

$$\therefore f(x_0) = \beta_0 + \beta^T x_0$$

$$\therefore \beta^T x_0 = -\beta_0$$

Therefore

$$\begin{aligned}
d &= \frac{y(x \cdot \beta - x_0 \cdot \beta)}{\|\beta\|_2} \\
&= \frac{y(\beta^T x - \beta^T x_0)}{\|\beta\|_2} \\
&= \frac{y(\beta^T x + \beta_0)}{\|\beta\|_2} \\
&= \frac{yf(x)}{\|\beta\|_2}
\end{aligned} \tag{1}$$

3. By perceptron convergence theorem we get that the perceptron algorithm makes at most  $\frac{1}{\delta^2}$  mistakes where  $\frac{1}{\delta^2}$  is the lower bound for  $y_i(w^{sep^T} x_i)$  and  $w^{sep}$  is the “good separator”:

$$T \leq \frac{1}{\delta^2}$$

where

$$y_i(w^{sep} x_i) \geq \delta$$

Since

$$y_i w^{sep} x_i = 1$$

we have

$$T \leq 1$$

Since  $w^0 = 0$ ,  $\|w^{sep}\| = 1$ , we have:

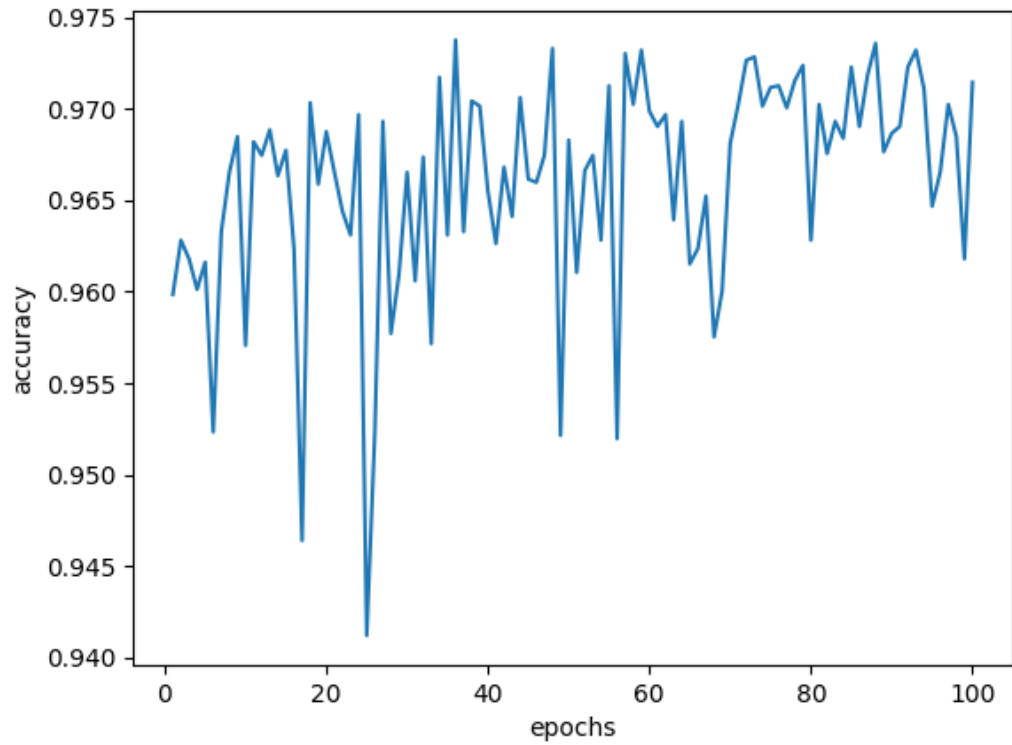
$$\|w^{(0)} - w^{sep}\|_2^2 = \|w^{sep}\|_2^2 = 1$$

Therefore,

$$T \leq \|w^{(0)} - w^{sep}\|_2^2$$

## 2 Programming Assignment

1. Please refer code in "perceptron.py"

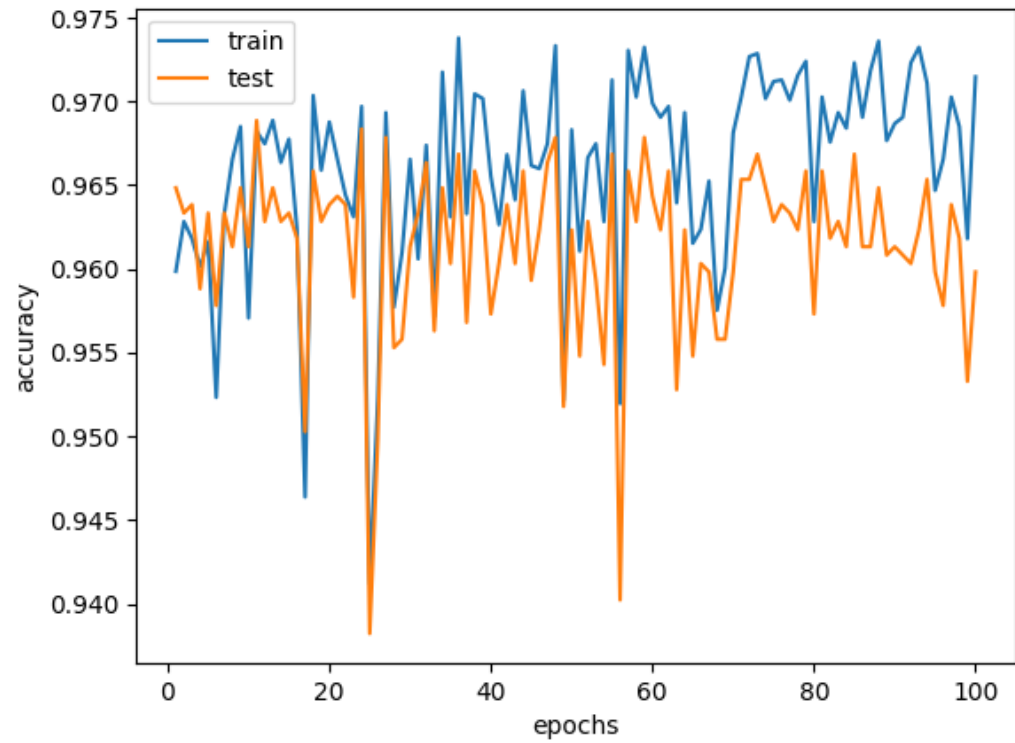


(a)

Since first epoch the accuracy is already pretty good.

Then it shakes drastically at first (within a small bound, in general the accuracy is always above 0.94), then becomes more steady, with the moving average increasing slowly.

By increasing/decreasing number of epochs, the curve becomes denser/sparser – just revealing more/less information in the figure, the points' location won't change.

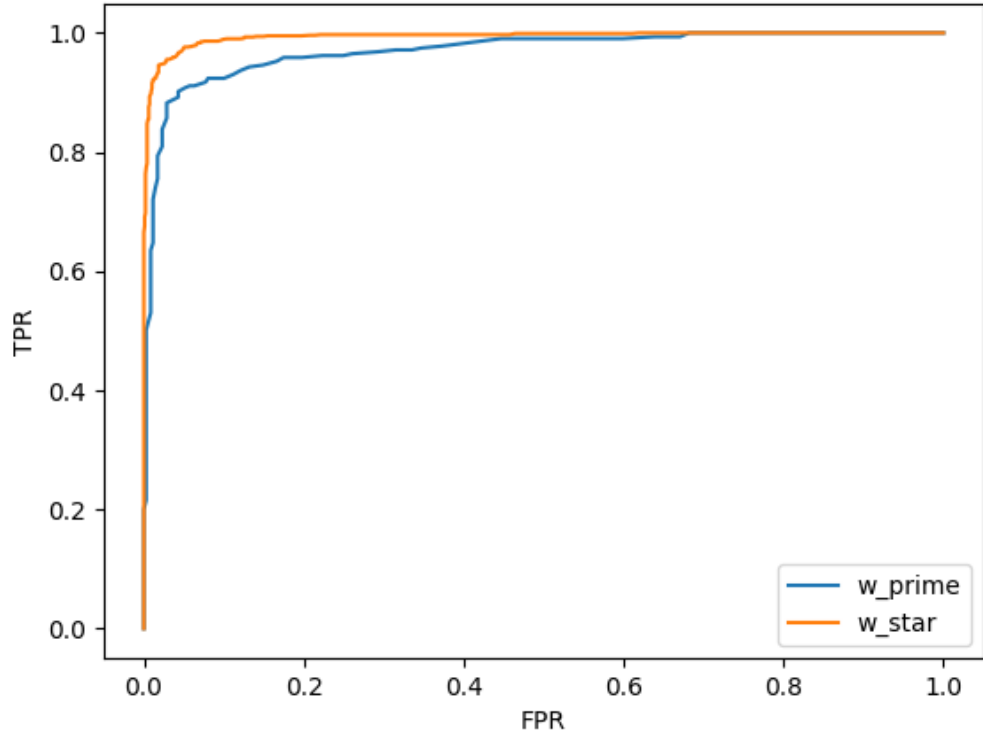


- (b) The curve for test set roughly follows the same pattern of training set. With the epoch number going up, it becomes lower than the training set.

(c) accuracy: 0.9939728779507785

confusion matrix:

$$\begin{bmatrix} 968 & 39 \\ 41 & 943 \end{bmatrix}$$

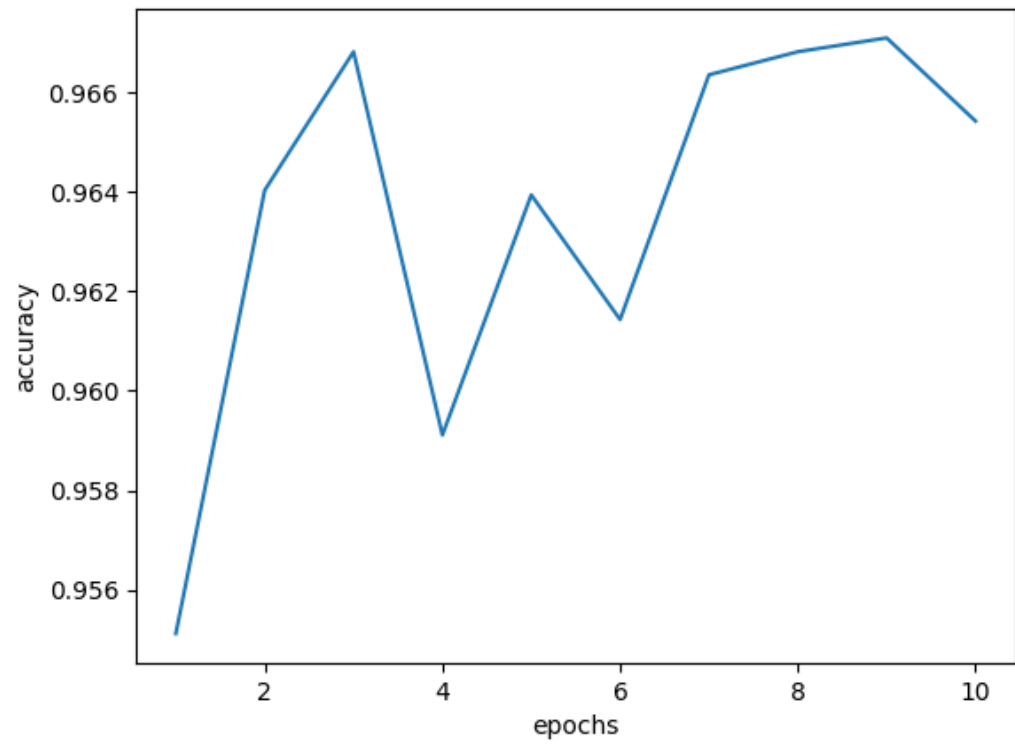


(d)

(e)  $AUC(w') = 0.9877518533678995$

$AUC(w^*) = 0.9999666948582919$  The greater the AUC value, the bigger area that is under the ROC curve, the 'upper' that the ROC curve goes.

2. (a) Evolution for training set after 10 epochs:



For test set:  
accuracy: 0.9653440482169764  
confusion matrix:

$$\begin{bmatrix} 980 & 40 \\ 29 & 942 \end{bmatrix}$$

(b)