## Lab6-The\_Perceptron

February 6, 2025

## 1 Imports

```
[5]: import numpy as np
   from sklearn.datasets import make_blobs
   import pandas as pd
   from pandas import DataFrame
   %matplotlib inline
   import matplotlib.pyplot as plt
   from mlxtend.plotting import plot_decision_regions
```

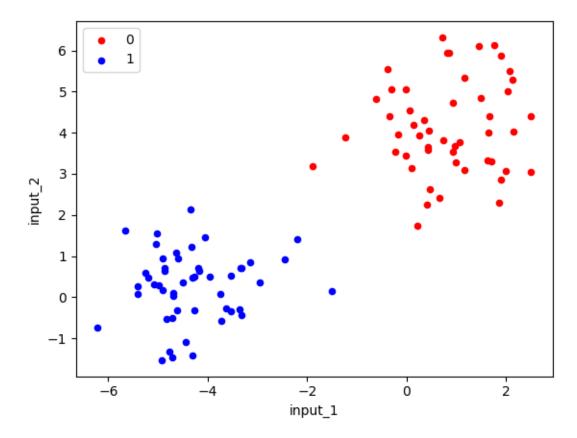
## 2 Dataset

```
[6]: def prep_data():
         X, labels = make_blobs(n_samples=100, centers=2, n_features=2,__

¬cluster_std=1, random_state=3)
         # labels = np.where(labels == 0, -1, 1)
         \#df = DataFrame(dict(constant=np.repeat(1, len(X)), input_1=X[:,0], len(X))
      →input_2=X[:,1], label=labels))
         df = DataFrame(dict(input_1=X[:,0], input_2=X[:,1], label=labels))
         print(df.head())
         colors = {1:'blue', 0:'red'}
         fig, ax = plt.subplots()
         grouped = df.groupby('label')
         for key, group in grouped:
             group.plot(ax=ax, kind='scatter', x='input_1', y='input_2', label=key, u
      ⇔color=colors[key])
         plt.show()
         #y = np.where(labels == 0, -1, 1)
         y = labels
         # sepal length and petal length # from sklearn.datasets.samples_generator_
      ⇔import make_blobs
         X = df.iloc[0:100, [0,1]].values
         return X, y
     X, y = prep_data()
```

```
input_1 input_2 label 0 -3.755938 0.067538 1
```

```
1 -2.948434 0.365538 1
2 -3.367091 -0.303440 1
3 1.890244 2.869420 0
4 -4.770407 -1.330004 1
```



## 3 The Perceptron Implementation

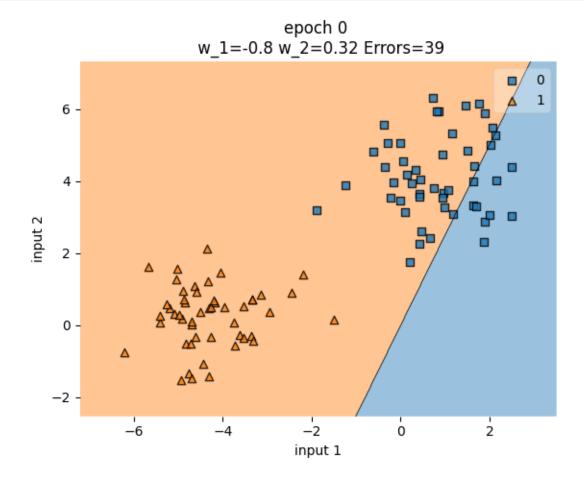
```
[7]: class Perceptron(object):
    # """Perceptron class

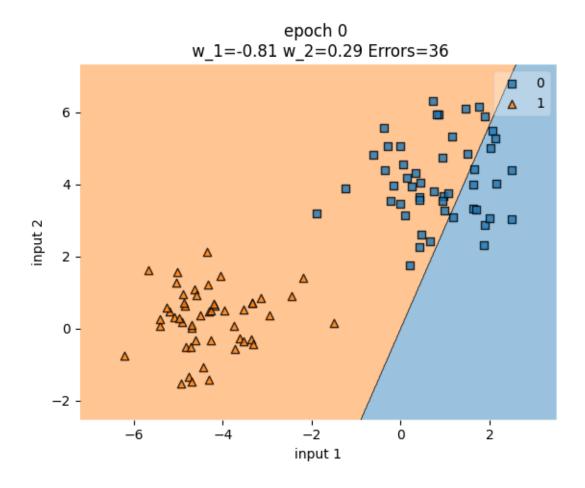
# Usage:
    # ppn = Perceptron(epochs=10, lr=0.01)
# ppn.train(X, y)
# ppn.plot_delta()

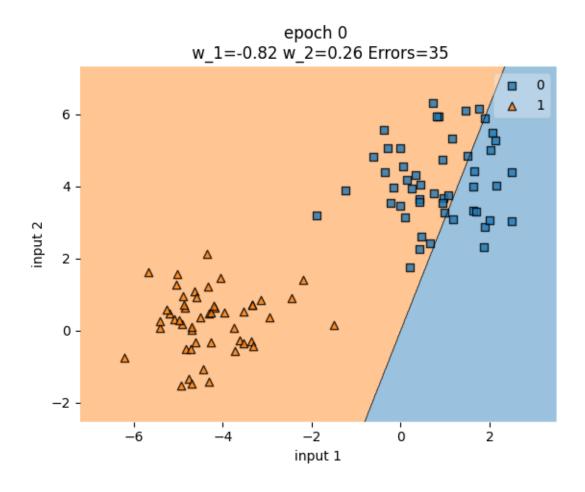
# Author: Uzair Ahmad
# """
    def __init__(self, lr=0.001, epochs=50):
        self.lr = lr
        self.epochs = epochs
```

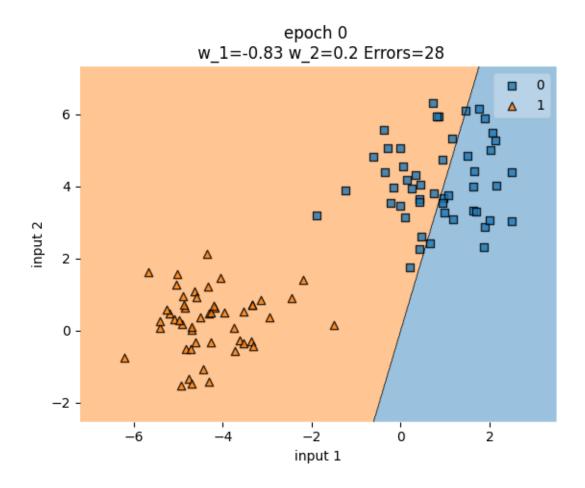
```
def predict(self, X):
                  preds = np.dot(X, self.w_)
                  return np.where(preds >= 0, 1, 0)
       def train(self, X, y):
                  # Initialize weights
                  self.w_ = np.random.randn(2)
                  self.errors = []
                  for epoch in range(self.epochs):
                              # Calculate classification errors
                              self.errors.append(np.sum(np.abs(y - self.predict(X))))
                              self.plot_dr(X, y, title='epoch {0} \n w_1={1} w_2={2} Errors={3}'.
oformat(epoch, np.round(self.w_[0],2), np.round(self.w_[1],2),
                 int(np.sum(np.abs(y - self.predict(X)))))
                              for xi, yi in zip(X, y):
                                         output = self.predict(xi)
                                         # calculate update
                                         update = self.lr * (yi - output) * xi
                                         # Update w
                                         self.w_ += update
                                         if np.all(update != 0):
                                               self.plot_dr(X, y, title='epoch \{0\} \setminus w_1=\{1\} \setminus w_2=\{2\}
General Structure = Struc
                 int(np.sum(np.abs(y - self.predict(X)))))
                  return self
       def plot_dr(self, X, y, title):
                  plot_decision_regions(X, y, clf=self)
                  plt.title(title)
                  plt.xlabel('input 1')
                  plt.ylabel('input 2')
                  plt.show()
       def plot_delta(self):
             plt.plot(range(1, len(self.errors)+1), self.errors, marker='o')
            plt.title('Errors / Epoch')
            plt.xlabel('Epochs')
            plt.ylabel('Error Count')
            plt.show()
```

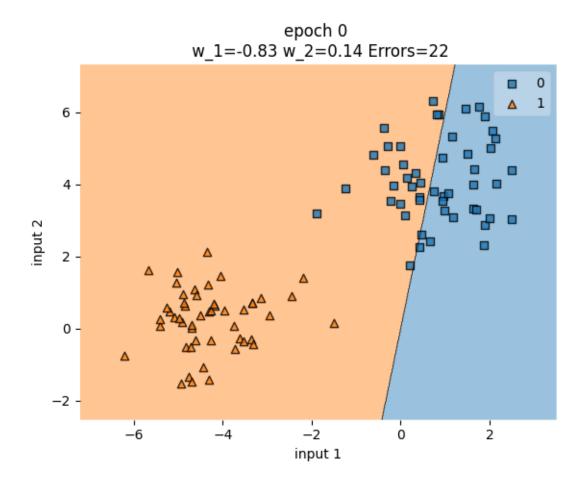
```
[8]: np.random.seed(1000)
    ppn = Perceptron(epochs=10, lr=0.01)
    ppn.train(X, y)
    ppn.plot_delta()
```

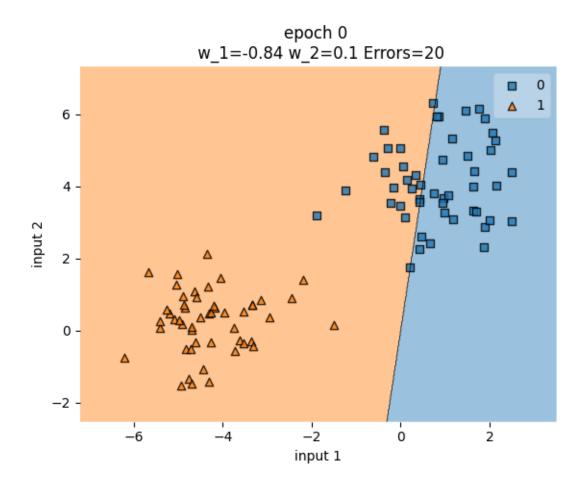


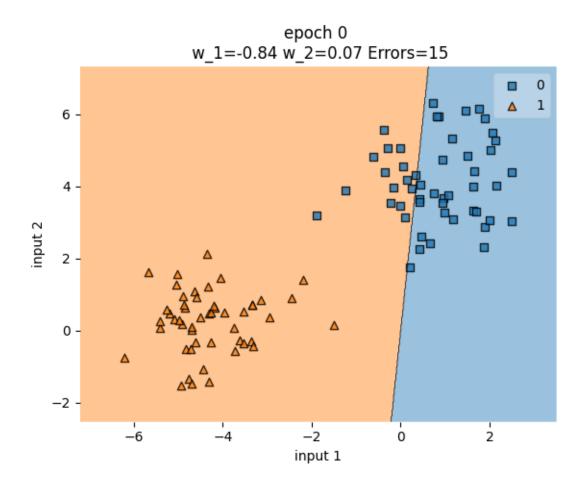


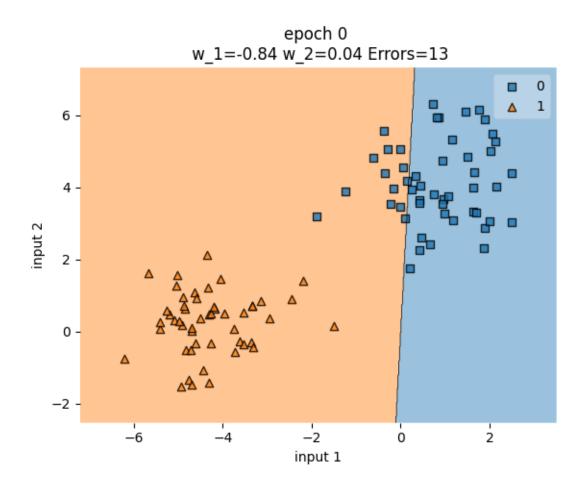


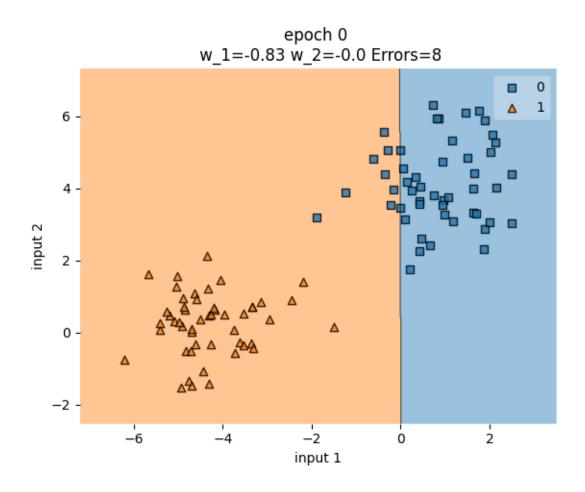


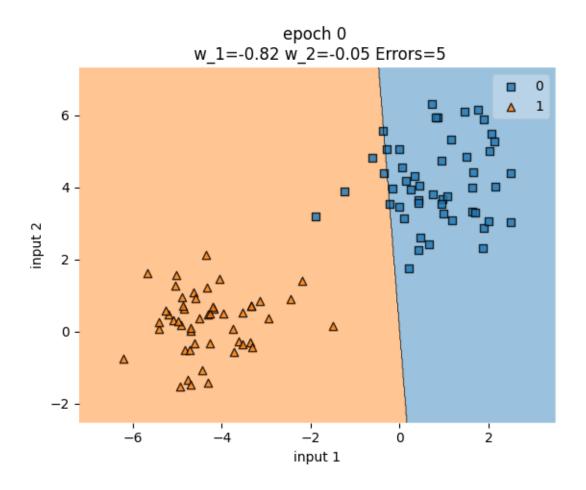


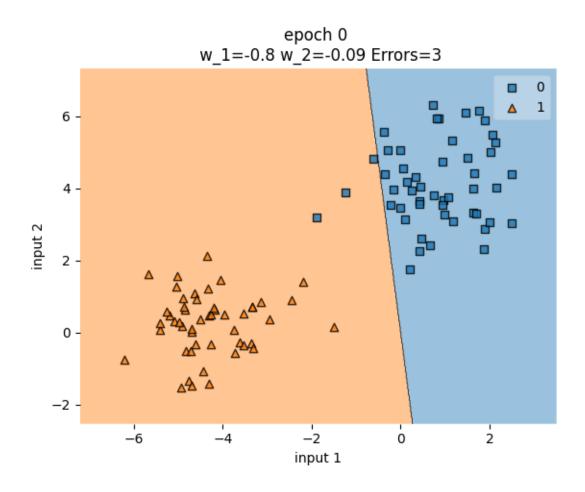


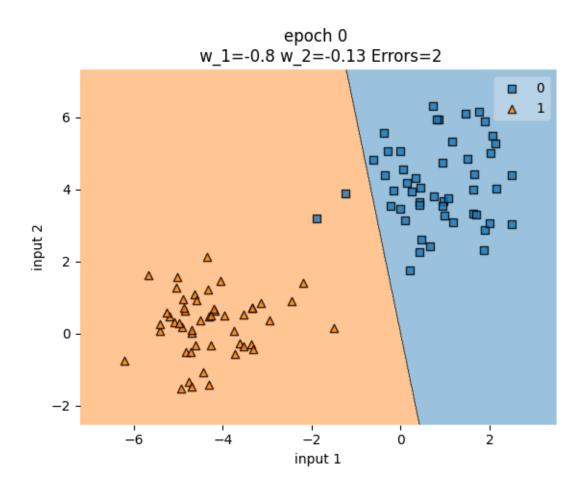


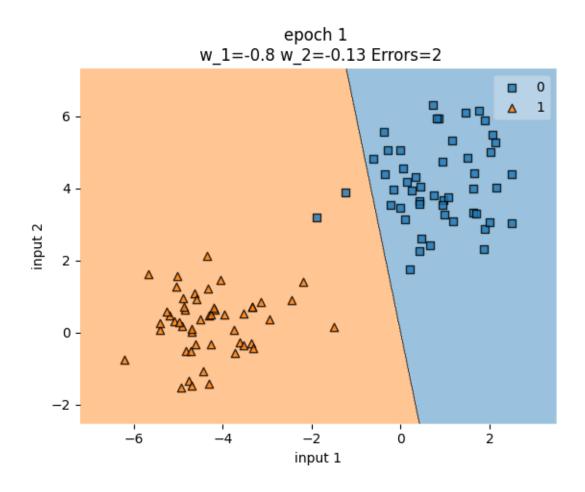


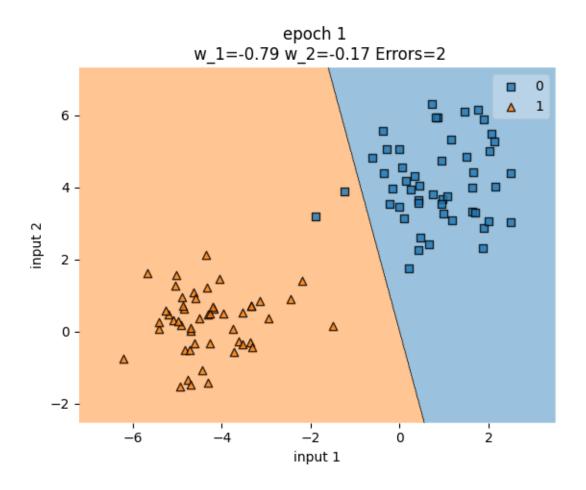


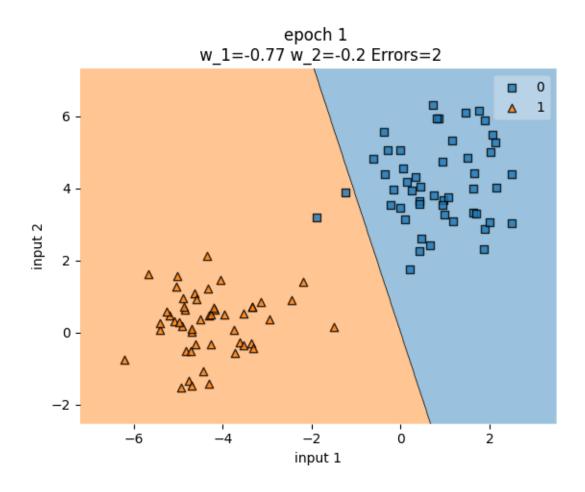


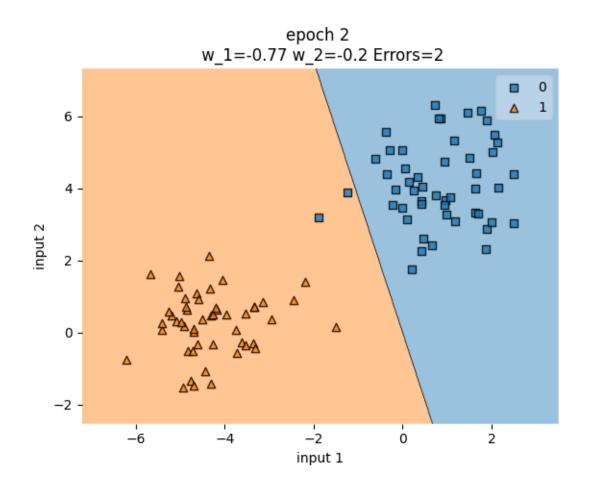


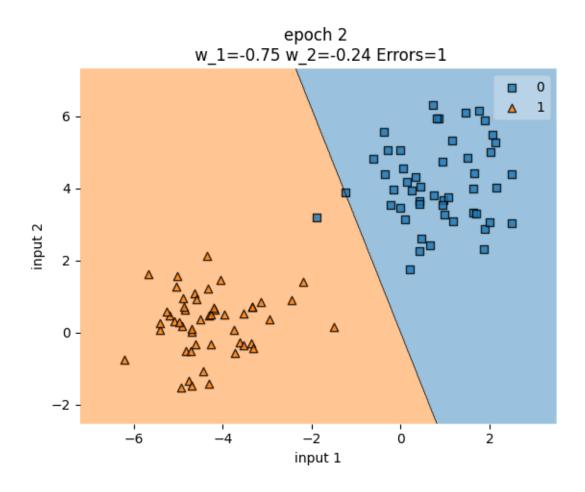


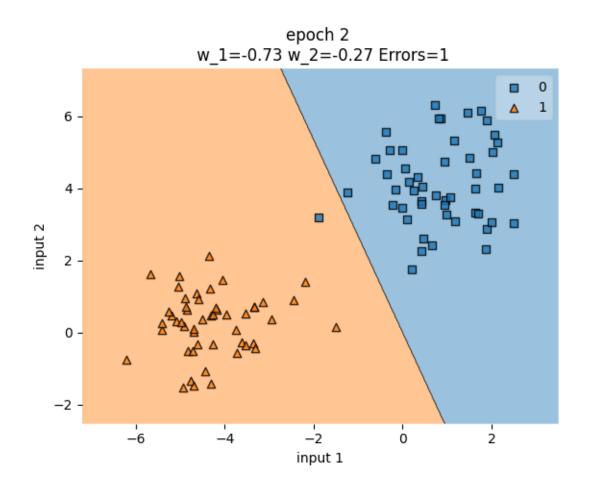


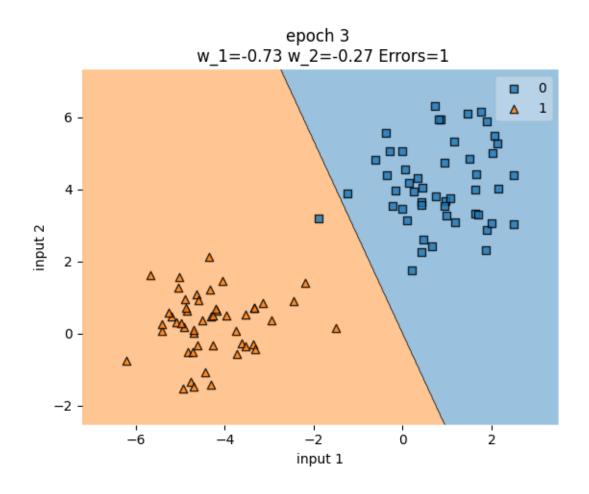


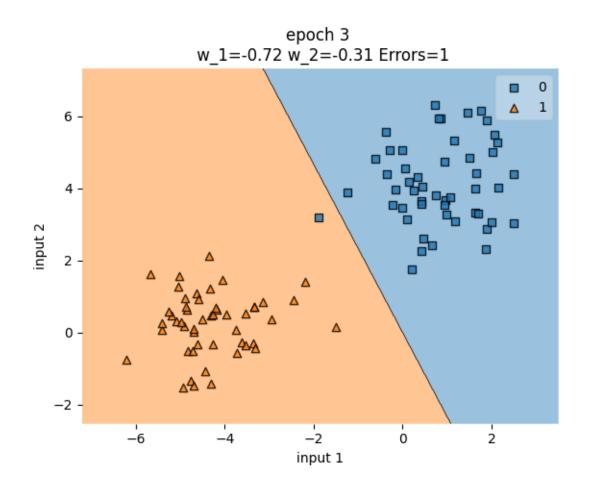


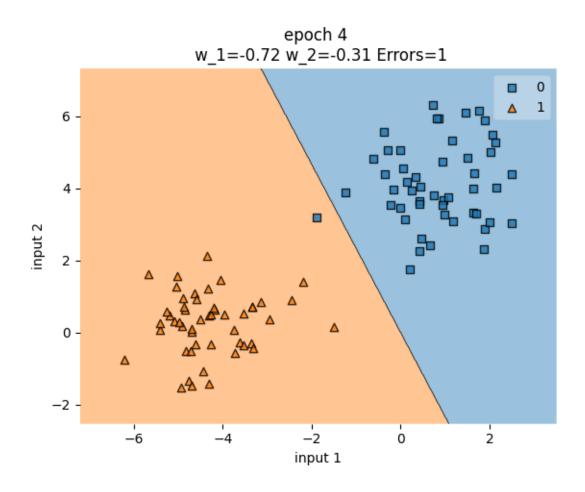


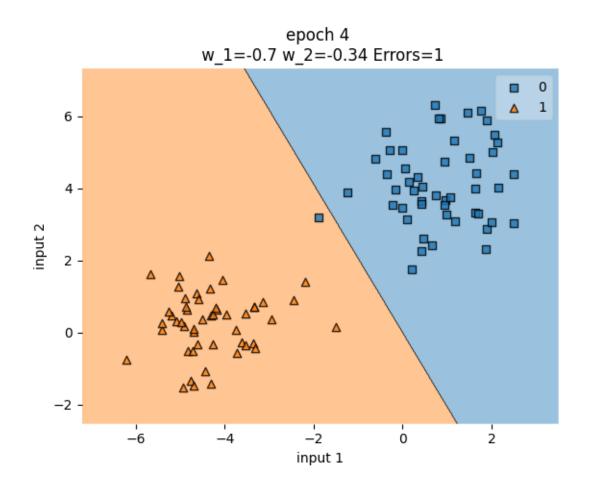


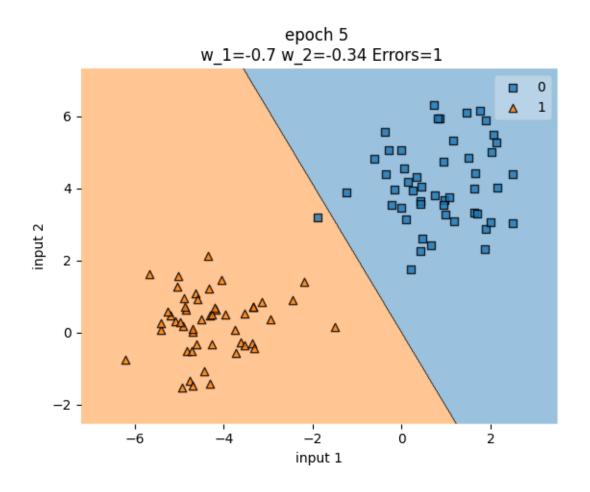


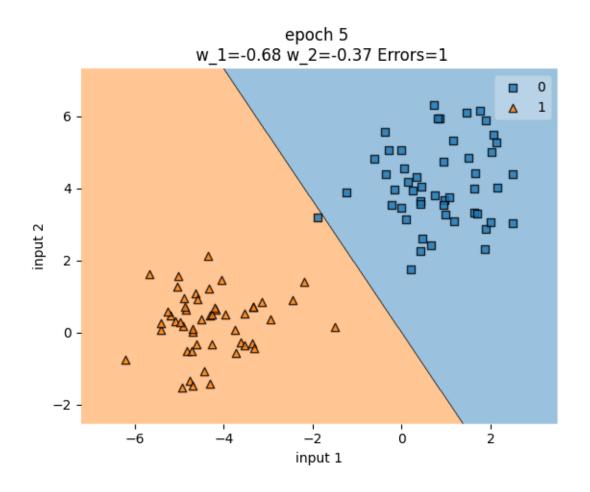


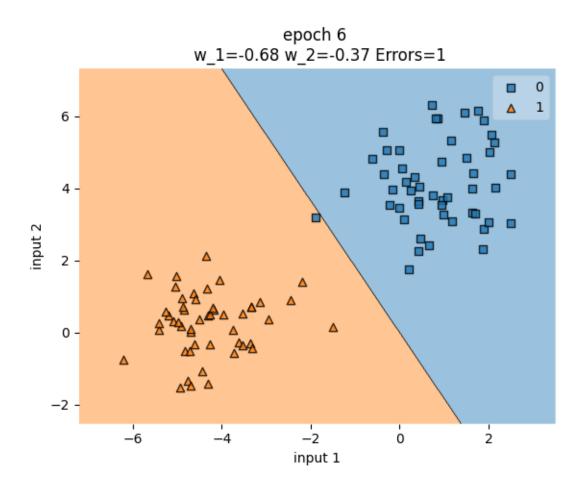


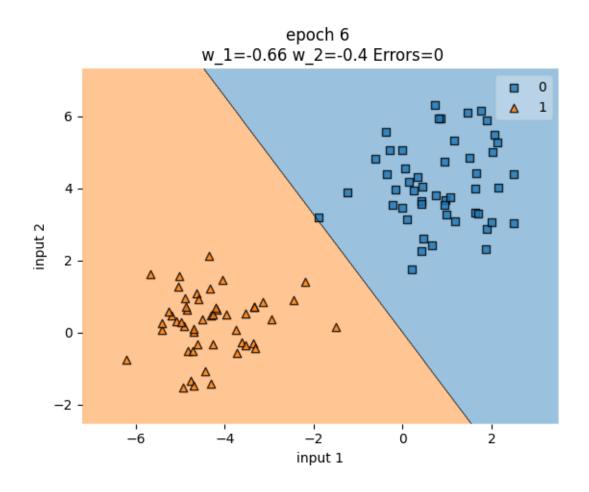


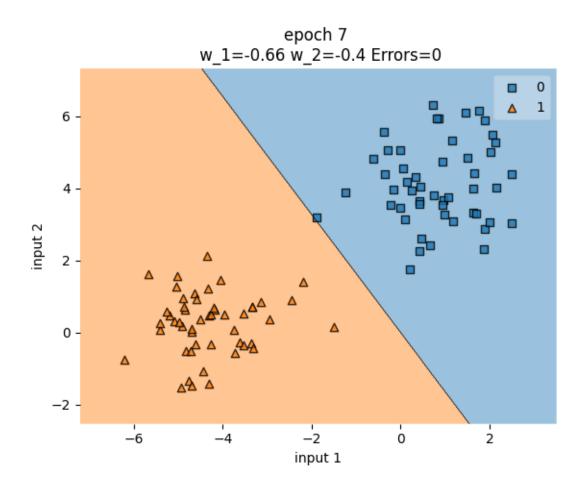


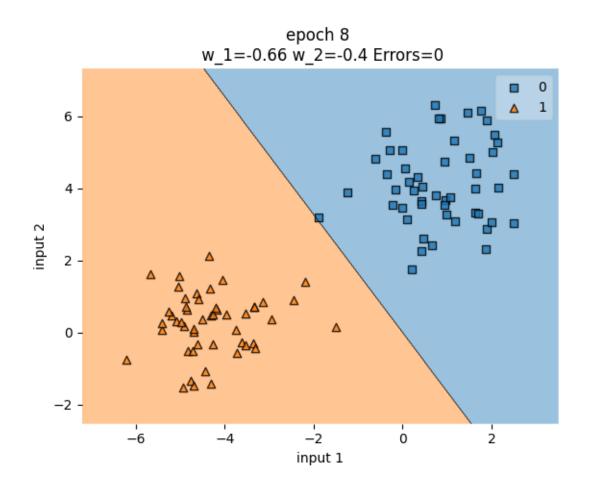


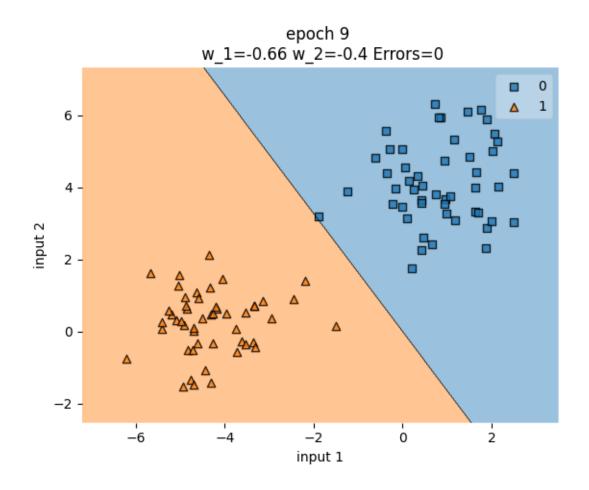


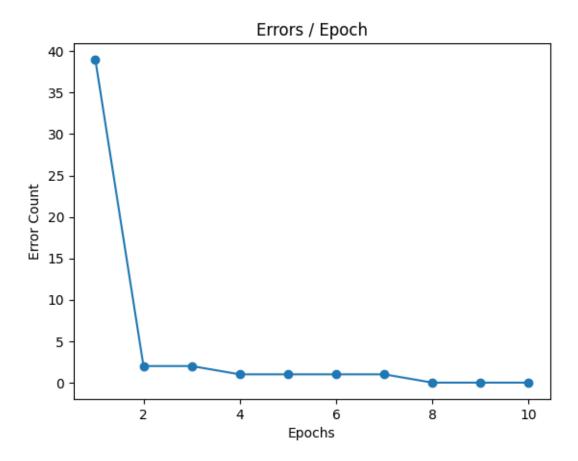












```
[9]: W= np.asarray([-0.68, -0.4])
magnitude_W = np.sqrt(np.dot(W, W))
l = []
for xi, yi in zip(X, y):
    magnitude_xi = np.sqrt(np.dot(xi, xi))
    theta = round(np.arccos(np.dot(W, xi)/(magnitude_W*magnitude_xi)), 2)
    l.append((xi, round(np.degrees(theta)), round(np.cos(theta),2), round(np.dot(W, xi),2), yi))
l.sort(key=lambda r:r[1])
l
# np.arccos(np.dot(W, np.sort(X.T))/np.linalg.norm(W)*np.linalg.norm(X))
[9]: [(array([-4.31126623, -1.40590907]), 13, 0.98, 3.49, 1)
```

```
[9]: [(array([-4.31126623, -1.40590907]), 13, 0.98, 3.49, 1), (array([-4.70716987, -1.45908253]), 13, 0.97, 3.78, 1), (array([-4.93524002, -1.52755815]), 13, 0.97, 3.97, 1), (array([-4.77040684, -1.33000372]), 15, 0.97, 3.78, 1), (array([-4.44314571, -1.08149453]), 17, 0.96, 3.45, 1), (array([-3.73612426, -0.56615694]), 22, 0.93, 2.77, 1), (array([-6.21229878, -0.74422301]), 23, 0.92, 4.52, 1), (array([-3.31907313, -0.43187992]), 23, 0.92, 2.43, 1),
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```

```
[10]: W = np.asarray([-0.68, -0.4])
magnitude_W = np.sqrt(np.dot(W, W))
X_ = np.asarray([[0, 2.0], [-2.0, 0.0]])
1 = []
for xi in X_:
    print(xi)
    magnitude_xi = np.sqrt(np.dot(xi, xi))
    theta = round(np.arccos(np.dot(W, xi)/(magnitude_W*magnitude_xi)), 2)
    yi = 1 if np.dot(W, xi) >= 0 else 0
    l.append((xi, round(np.degrees(theta)), round(np.cos(theta),2), round(np.cot(W, xi),2), yi))
1.sort(key=lambda r:r[1])
1
[0. 2.]
[-2. 0.]
```

[10]: [(array([-2., 0.]), 30, 0.86, 1.36, 1), (array([0., 2.]), 120, -0.5, -0.8, 0)]