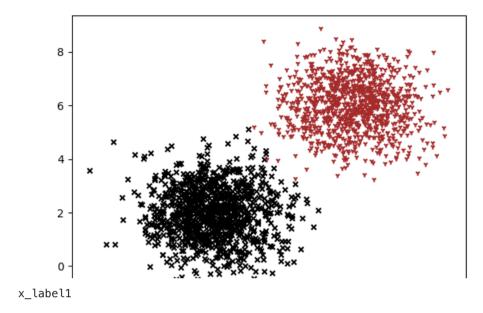
### ▼ Exercise 1

```
import tensorflow as tf
import matplotlib.pyplot as plt
import numpy as np
import random
import keras
from keras.models import Sequential
from keras.layers import Dense
Two gangs
Dataset:
[0]*10+[1]*10
     [0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1]
x_label1 = np.random.normal(3, 1, 1000)
y_label1 = np.random.normal(2, 1, 1000)
x_{label2} = np.random.normal(7, 1, 1000)
y_label2 = np.random.normal(6, 1, 1000)
xs = np.append(x label1, x label2)
ys = np.append(y label1, y label2)
labels = np.asarray([0.]*len(x_label1)+[1.]*len(x_label2))
labels
    array([0., 0., 0., ..., 1., 1., 1.])
plt.scatter(x_label1, y_label1, c='black', marker='x', s=20)
plt.scatter(x label2, y label2, c='brown', marker='1', s=20)
plt.show()
```



```
1.4043319/,
            Z./000////, 5.WOOWWOY,
                                                  4.30984000,
                                                               4.UU013Z3 ,
            3.30948566.
                        2.65892936.
                                     4.54708041.
                                                  3.3536456 .
                                                               2.64132637.
            1.91270728.
                        2.28289622.
                                     1.98510396.
                                                  4.50041241.
                                                               3.02702409.
                                     3.16128403,
                                                  3.62381545.
                                                               3.31061833,
            3.43281255.
                        3.09853993,
            3.10332631.
                        4.65475797,
                                     2.82132597, 5.55659173,
                                                               0.1992336 .
            3.3286668 ,
                        2.30486388,
                                     2.1906411 , 1.23671327,
                                                               2.61964142.
                                                  1.67535553,
            2.61999692,
                        2.45492048.
                                     4.9837424 ,
                                                               2.63047348.
                                     2.68707446.
            3.73767474.
                        4.0986719 .
                                                  3.62350634.
                                                               1.45070148.
            4.52935102,
                        4.09046056.
                                     3.45308144,
                                                  2.75901958,
                                                               2.39938623,
            2.70931953,
                        3.34270314,
                                     2.26173305,
                                                  3.81247003,
                                                               2.74178278.
            3.36031086,
                        4.60967789.
                                     2.86390102,
                                                  4.19866739.
                                                               2.27490456.
            4.43047177.
                        2.03006323.
                                     3.10469332,
                                                  2.79190053,
                                                               3.19264955.
            2.08360951,
                        4.65623311,
                                     1.27400255, 3.2272532,
                                                               2.64194927.
                                     3.73287552,
                                                  3.62950425,
            1.89099932,
                        1.38163391.
                                                               3.7473768
            3.53349219, 1.66943243,
                                     4.18337928, 2.25480363,
                                                               2.37689968.
            1.63581301,
                        3.90752455,
                                     3.93871394,
                                                  2.41022709,
                                                               3.36137868,
            4.74228693.
                        0.48701304,
                                     2.73775202,
                                                  3.55490039,
                                                               2.36108197.
            4.97994702, 4.54542903,
                                     3.72602175, 2.78391812,
                                                               3.77055318,
            3.3955416 , 2.67375828,
                                     2.25612879, 1.56181916,
                                                               3.76082979,
            4.3261713 , 3.36876662,
                                     2.69526173, 1.90672271,
                                                               3.18184957,
            2.3843636 , 2.99288444,
                                     2.65162821, 2.49780499,
                                                               3.79412158,
            3.17941147. 2.98437068.
                                     2.39272109. 5.202545
                                                               3.44023152.
            3.63917235, 3.00954373, 1.9515941,
                                                  2.43108033, 2.09541243,
            1.88771844,
                        2.45698092, 2.57063964,
                                                  3.91299906.
                                                               2.97756476.
            2.98083935. 4.69267141. 2.39218397. 1.70003808.
                                                               3.94808565.
            2.67958261, 3.47190478, 2.48356196, 3.16802027, 1.78518477,
            2.80122073, 4.47161755, 2.72641721, 3.35765132,
                                                               3.65103761.
            2.20791223, 2.32981877, 3.45861151, 1.45496228,
                                                               5.3351428
            4.7084647 , 1.78428851, 3.59957314, 2.56614516, 3.19995689])
def loss fn grad(y, y model):
return tf.reduce mean(-y*tf.math.log(y model)-(1-y)*tf.math.log(1-y model))
def split dataset(data points, label, subset size):
   arr = np.arange(len(data points))
   l=len(data points)
   s=int(subset size*l)
   np.random.shuffle(arr)
   data points val =data_points[arr[0:s]]
   label val = label[arr[0:s]]
   data_points_train = data_points[arr[:int(l*(1-subset_size))]]
   label train = label[arr[:int(l*(1-subset size))]]
   return data_points_train, label_train, data_points_val, label_val
```

```
def subset_dataset(x_dataset, y_dataset, label, subset_size):
    arr = np.arange(len(x_dataset))
    np.random.shuffle(arr)
    x_train = x_dataset[arr[0:subset_size]]
    y_train = y_dataset[arr[0:subset_size]]
    label_train = label[arr[0:subset_size]]
    return x_train, y_train, label_train

def subset_dataset_concatenated(data, label, subset_size):
    arr = np.arange(len(data))
    np.random.shuffle(arr)
    data_train = data[arr[0:subset_size]]
    label_train = label[arr[0:subset_size]]
    return data_train, label_train

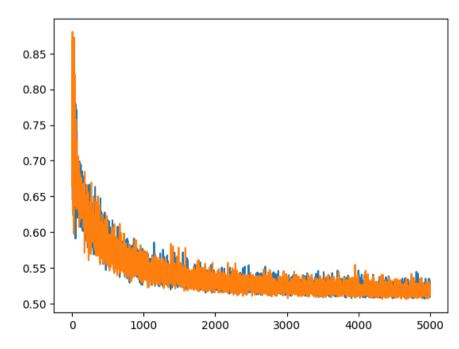
labels.shape
    (2000,)
```

```
27.11.2023. 22:15
   Loss = []
   Val loss = []
   epochs = 5000
   learning rate = 0.1
   batch size = 20
   w = tf.Variable(np.random.random((2, 2)))
   b = tf.Variable(np.random.random((2)))
   data = np.column stack((xs,ys))
   data train, label train, data val, label val = split dataset (data, labels, 0.2)
   for _ in range(epochs):
     data batch, labels batch = subset dataset concatenated(data train, label train, batch size)
     data val batch, labels val batch = subset dataset concatenated(data val, label val, batch size)
     with tf.GradientTape() as tape:
       pred l=tf.nn.softmax(tf.matmul(data batch, w) + b)
       pred l val=tf.nn.softmax(tf.matmul(data val batch, w) + b)
       labels batch one hot = tf.one hot(labels batch, depth=2)
       labels val batch one hot = tf.one hot(labels val batch, depth=2)
       loss = tf.keras.losses.BinaryCrossentropy(from_logits=True)(labels_batch_one_hot, pred_l)
       Loss.append(loss.numpy())
       val_loss = tf.keras.losses.BinaryCrossentropy(from_logits=True)(labels_val_batch_one_hot, pred_l_val)
       Val loss.append(val loss.numpy())
       print("loss",loss,"val loss",val loss)
     dloss dw,dloss db = tape.gradient(loss, [w, b])
     w.assign sub(learning rate*dloss dw )
     b.assign_sub(learning_rate*dloss_db )
```

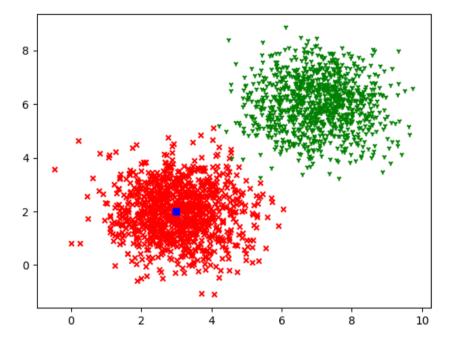
```
LUSS LI.IENSUI(V.3122W4313421W/, SNdpe=(), ULYpe=1LUdLO4) Val LUSS LI.IENSUI(V.310646122161W332, SNdpe=(), ULYpe=1LUdLO4)
loss tf.Tensor(0.5256652933447883, shape=(), dtype=float64) val loss tf.Tensor(0.5166535053015568, shape=(), dtype=float64)
loss tf.Tensor(0.5178850405770185, shape=(), dtype=float64) val loss tf.Tensor(0.5203052381603911, shape=(), dtype=float64)
loss tf.Tensor(0.517922249328594, shape=(), dtype=float64) val loss tf.Tensor(0.5154060740405815, shape=(), dtype=float64)
loss tf.Tensor(0.5152846908257349, shape=(), dtype=float64) val loss tf.Tensor(0.5111547236876955, shape=(), dtype=float64)
loss tf.Tensor(0.5179300538121787, shape=(), dtype=float64) val loss tf.Tensor(0.5142287949998849, shape=(), dtype=float64)
loss tf.Tensor(0.512815642658637, shape=(), dtype=float64) val loss tf.Tensor(0.5183142711873845, shape=(), dtype=float64)
loss tf.Tensor(0.5147071733618589, shape=(), dtvpe=float64) val loss tf.Tensor(0.5195010178465658, shape=(), dtvpe=float64)
loss tf.Tensor(0.517624940687482, shape=(), dtype=float64) val loss tf.Tensor(0.5092411047700509, shape=(), dtype=float64)
loss tf.Tensor(0.5148768149278122, shape=(), dtype=float64) val loss tf.Tensor(0.5097451641217741, shape=(), dtype=float64)
loss tf.Tensor(0.5123402296159247, shape=(), dtype=float64) val loss tf.Tensor(0.5242343871866244, shape=(), dtype=float64)
loss tf.Tensor(0.5158228739727162, shape=(), dtype=float64) val loss tf.Tensor(0.5176552019419643, shape=(), dtype=float64)
loss tf.Tensor(0.5181808717704712, shape=(), dtype=float64) val loss tf.Tensor(0.5302942459486595, shape=(), dtype=float64)
loss tf.Tensor(0.5210640558914632, shape=(), dtype=float64) val loss tf.Tensor(0.5091564510948471, shape=(), dtype=float64)
loss tf.Tensor(0.5196437340793592, shape=(), dtype=float64) val loss tf.Tensor(0.5108230031725508, shape=(), dtype=float64)
loss tf.Tensor(0.5101729655144, shape=(), dtype=float64) val_loss tf.Tensor(0.5100541946290931, shape=(), dtype=float64)
loss tf.Tensor(0.5073725707414354, shape=(), dtvpe=float64) val loss tf.Tensor(0.5256094437149509, shape=(), dtvpe=float64)
loss tf.Tensor(0.5140487890774331, shape=(), dtype=float64) val loss tf.Tensor(0.5185798935276333, shape=(), dtype=float64)
loss tf.Tensor(0.5147100753424215, shape=(), dtype=float64) val_loss tf.Tensor(0.5126580177094191, shape=(), dtype=float64)
loss tf.Tensor(0.512593865203139, shape=(), dtype=float64) val loss tf.Tensor(0.523060146363972, shape=(), dtype=float64)
loss tf.Tensor(0.5165768522340816, shape=(), dtype=float64) val loss tf.Tensor(0.5215936957953248, shape=(), dtype=float64)
loss tf.Tensor(0.5145151858330945, shape=(), dtype=float64) val loss tf.Tensor(0.5132791642210766, shape=(), dtype=float64)
loss tf.Tensor(0.511823165370401, shape=(), dtype=float64) val loss tf.Tensor(0.5165108258351669, shape=(), dtype=float64)
loss tf.Tensor(0.5340399098517743, shape=(), dtype=float64) val loss tf.Tensor(0.5075444861725433, shape=(), dtype=float64)
loss tf.Tensor(0.5169254816341875, shape=(), dtvpe=float64) val loss tf.Tensor(0.5119881931845672, shape=(), dtvpe=float64)
loss tf.Tensor(0.5210333395143512, shape=(), dtype=float64) val loss tf.Tensor(0.5140113103480203, shape=(), dtype=float64)
loss tf.Tensor(0.515202840404932, shape=(), dtvpe=float64) val loss tf.Tensor(0.512332780709589, shape=(), dtvpe=float64)
loss tf.Tensor(0.5150698649983365, shape=(), dtype=float64) val loss tf.Tensor(0.5120866121899313, shape=(), dtype=float64)
loss tf.Tensor(0.5304406833370937, shape=(), dtype=float64) val loss tf.Tensor(0.5156165411408342, shape=(), dtype=float64)
loss tf.Tensor(0.5159516568757614, shape=(), dtype=float64) val loss tf.Tensor(0.5313480072067179, shape=(), dtype=float64)
loss tf.Tensor(0.527278300045394, shape=(), dtype=float64) val loss tf.Tensor(0.5242581866241394, shape=(), dtype=float64)
loss tf.Tensor(0.5300226215717415, shape=(), dtype=float64) val loss tf.Tensor(0.5203665836874464, shape=(), dtype=float64)
loss tf.Tensor(0.5162360599561746, shape=(), dtype=float64) val loss tf.Tensor(0.5132870609061891, shape=(), dtype=float64)
loss tf.Tensor(0.5348840565998073, shape=(), dtype=float64) val loss tf.Tensor(0.5116777943532278, shape=(), dtype=float64)
loss tf.Tensor(0.5213577937715219, shape=(), dtype=float64) val loss tf.Tensor(0.5092627719822975, shape=(), dtype=float64)
loss tf.Tensor(0.5102712250713457, shape=(), dtvpe=float64) val loss tf.Tensor(0.5122154219339361, shape=(), dtvpe=float64)
loss tf.Tensor(0.5192766449738351, shape=(), dtype=float64) val loss tf.Tensor(0.5203141652220289, shape=(), dtype=float64)
loss tf.Tensor(0.5133585054523104, shape=(), dtype=float64) val_loss tf.Tensor(0.5112405492728064, shape=(), dtype=float64)
loss tf.Tensor(0.5161412473586546, shape=(), dtype=float64) val loss tf.Tensor(0.5224799881012637, shape=(), dtype=float64)
loss tf.Tensor(0.5114006490589237, shape=(), dtype=float64) val loss tf.Tensor(0.5136758197089615, shape=(), dtype=float64)
loss tf.Tensor(0.508903866092442, shape=(), dtype=float64) val loss tf.Tensor(0.5127944070751355, shape=(), dtype=float64)
loss tf.Tensor(0.5321286128639814, shape=(), dtype=float64) val loss tf.Tensor(0.5152197623047291, shape=(), dtype=float64)
loss tf.Tensor(0.5126511748973185, shape=(), dtype=float64) val loss tf.Tensor(0.5119886462276876, shape=(), dtype=float64)
loss tf.Tensor(0.5115058768604321, shape=(), dtype=float64) val loss tf.Tensor(0.5270552301924957, shape=(), dtype=float64)
loss tf.Tensor(0.5148654887241977, shape=(), dtype=float64) val loss tf.Tensor(0.5129883036155126, shape=(), dtype=float64)
loss tf.Tensor(0.5125567617774643, shape=(), dtype=float64) val loss tf.Tensor(0.514186096690384, shape=(), dtype=float64)
```

print(data\_train.size,label\_train.size,data\_val.size,label\_val.size)

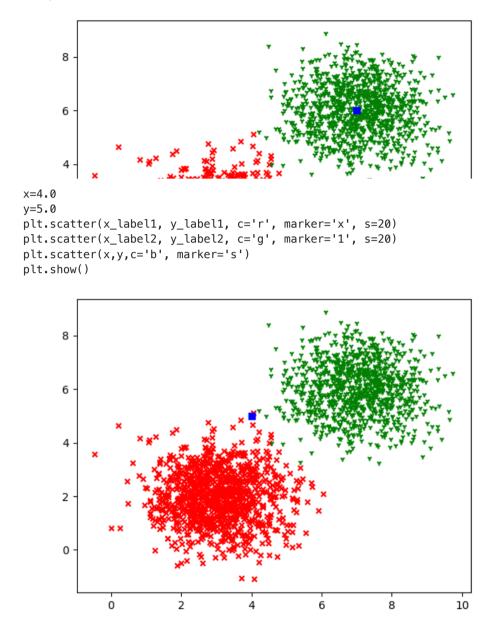
3200 1600 800 400



```
y=2.0
plt.scatter(x_label1, y_label1, c='r', marker='x', s=20)
plt.scatter(x_label2, y_label2, c='g', marker='1', s=20)
plt.scatter(x,y,c='b', marker='s')
plt.show()
```



```
x=7.0
y=6.0
plt.scatter(x_label1, y_label1, c='r', marker='x', s=20)
plt.scatter(x_label2, y_label2, c='g', marker='1', s=20)
plt.scatter(x,y,c='b', marker='s')
plt.show()
```



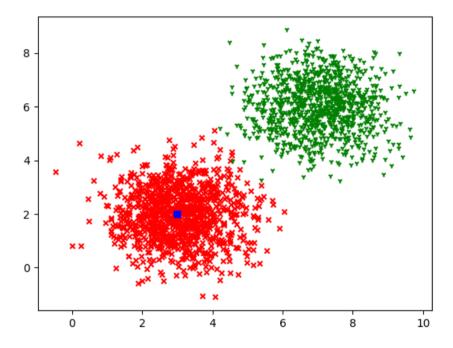
## → Hiperparametria

### → Learning rate 0.01

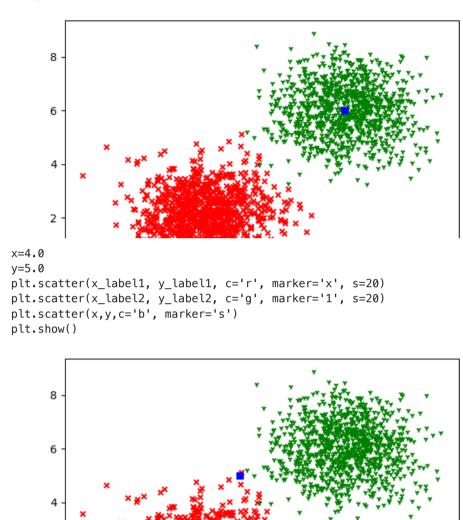
```
Loss = []
Val loss = []
epochs = 5000
learning rate = 0.01
batch\_size = 20
w = tf.Variable(np.random.random((2, 2)))
b = tf.Variable(np.random.random((2)))
data = np.column stack((xs,ys))
data train, label train, data val, label val = split dataset(data, labels, 0.2)
for in range(epochs):
 data_batch,labels_batch = subset_dataset_concatenated(data_train,label_train,batch_size)
 data_val_batch, labels_val_batch = subset_dataset_concatenated(data_val, label_val, batch_size)
 with tf.GradientTape() as tape:
    pred l=tf.nn.softmax(tf.matmul(data batch, w) + b)
    pred_l_val=tf.nn.softmax(tf.matmul(data_val_batch, w) + b)
    labels batch one hot = tf.one hot(labels batch, depth=2)
    labels val batch one hot = tf.one hot(labels val batch, depth=2)
    loss = tf.keras.losses.BinaryCrossentropy(from logits=True)(labels batch one hot, pred l)
    Loss.append(loss.numpy())
    val_loss = tf.keras.losses.BinaryCrossentropy(from_logits=True)(labels_val_batch_one_hot, pred_l_val)
    Val loss.append(val loss.numpy())
    print("loss", loss, "val_loss", val_loss)
 dloss_dw,dloss_db = tape.gradient(loss, [w, b])
 w.assign_sub(learning_rate*dloss_dw )
 b.assign sub(learning rate*dloss db )
```

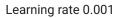
```
LUSS LI.IENSOI(@./29969110064902, Shape=(), Ulype=Float04) Val_LOSS LI.IENSOI(@./00002690260094, Shape=(), Ulype=Float04)
     loss of Tonsor(A 0A26/000700//61 shano-() dougno-float6/) val loss of Tonsor(A 65/02A0A000/2A701 shano-() dougno-float6/)
print(data_train.size, label_train.size, data_val.size, label_val.size)
     3200 1600 800 400
np.max(Loss),np.min(Loss)
     (0.9331220401407656, 0.556582217420378)
np.max(Val_loss),np.min(Val_loss)
     (1.003188993512679, 0.5808985670668569)
print(w.numpy())
print(b.numpy())
     [[1.25592541 0.31700005]
     [1.21573953 0.16564472]]
     [1.05052379 0.08972305]
plt.plot(Loss)
plt.plot(Val_loss)
plt.show()
```

```
x=3.0
y=2.0
plt.scatter(x_label1, y_label1, c='r', marker='x', s=20)
plt.scatter(x_label2, y_label2, c='g', marker='1', s=20)
plt.scatter(x,y,c='b', marker='s')
plt.show()
```



```
x=7.0
y=6.0
plt.scatter(x_label1, y_label1, c='r', marker='x', s=20)
plt.scatter(x_label2, y_label2, c='g', marker='1', s=20)
plt.scatter(x,y,c='b', marker='s')
plt.show()
```





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2

2 ·

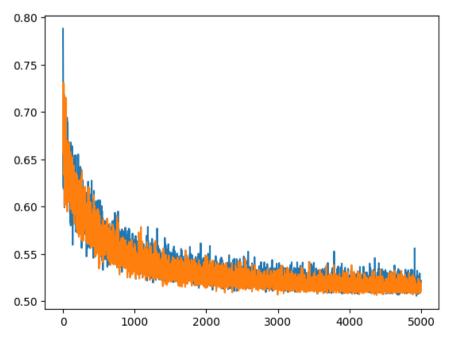
0

10

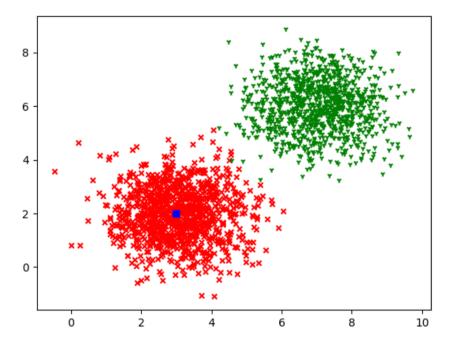
8

```
Loss = []
Val loss = []
epochs = 5000
learning rate = 0.1
batch size = 20
w = tf.Variable(np.random.random((2, 2)))
b = tf.Variable(np.random.random((2)))
data = np.column stack((xs,ys))
data train,label train,data val,label val = split dataset(data,labels,0.2)
for _ in range(epochs):
 data batch, labels batch = subset dataset concatenated(data train, label train, batch size)
 data val batch, labels val batch = subset dataset concatenated(data val, label val, batch size)
 with tf.GradientTape() as tape:
    pred_l=tf.nn.softmax(tf.matmul(data_batch, w) + b)
    pred l val=tf.nn.softmax(tf.matmul(data val batch, w) + b)
    labels batch one hot = tf.one hot(labels batch, depth=2)
    labels val batch one hot = tf.one hot(labels val batch, depth=2)
    loss = tf.keras.losses.BinaryCrossentropy(from logits=True)(labels batch one hot, pred l)
    Loss.append(loss.numpy())
    val_loss = tf.keras.losses.BinaryCrossentropy(from_logits=True)(labels_val_batch_one_hot, pred_l_val)
   Val_loss.append(val_loss.numpy())
    print("loss",loss,"val_loss",val_loss)
 dloss_dw,dloss_db = tape.gradient(loss, [w, b])
 w.assign sub(learning rate*dloss dw )
 b.assign sub(learning rate*dloss db )
```

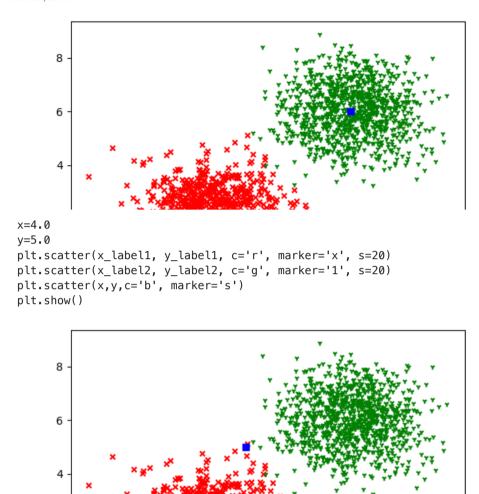
```
LUSS LI.IENSUI(W.DZ/DWG/DD9411300, SNdpe=(), ULYpe=1LUdLO4) VdL LUSS LI.IENSUI(W.D1//Z9D/930002, SNdpe=(), ULYpe=1LUdLO4)
loss tf.Tensor(0.5189430605543036, shape=(), dtype=float64) val loss tf.Tensor(0.5155052180067216, shape=(), dtype=float64)
loss tf.Tensor(0.513511756142648, shape=(), dtype=float64) val loss tf.Tensor(0.5127919148822238, shape=(), dtype=float64)
loss tf.Tensor(0.5101764309597876, shape=(), dtype=float64) val loss tf.Tensor(0.5189958658493179, shape=(), dtype=float64)
loss tf.Tensor(0.5196424005534986, shape=(), dtype=float64) val loss tf.Tensor(0.5118618637521382, shape=(), dtype=float64)
loss tf.Tensor(0.5188501951362374, shape=(), dtype=float64) val loss tf.Tensor(0.5112260413259253, shape=(), dtype=float64)
loss tf.Tensor(0.5209071051856041, shape=(), dtype=float64) val loss tf.Tensor(0.5133233718033687, shape=(), dtype=float64)
loss tf.Tensor(0.5078124238422042, shape=(), dtype=float64) val loss tf.Tensor(0.513196475979453, shape=(), dtype=float64)
loss tf.Tensor(0.5197721386633857, shape=(), dtype=float64) val_loss tf.Tensor(0.5177385955155718, shape=(), dtype=float64)
loss tf.Tensor(0.5146268990527558, shape=(), dtype=float64) val loss tf.Tensor(0.5103813188940525, shape=(), dtype=float64)
loss tf.Tensor(0.5085737981766647, shape=(), dtype=float64) val loss tf.Tensor(0.5116532543527562, shape=(), dtype=float64)
loss tf.Tensor(0.5183905189085314, shape=(), dtype=float64) val loss tf.Tensor(0.524131139021734, shape=(), dtype=float64)
loss tf.Tensor(0.5195840583549654, shape=(), dtype=float64) val loss tf.Tensor(0.5099307892047982, shape=(), dtype=float64)
loss tf.Tensor(0.5174512131164086, shape=(), dtype=float64) val loss tf.Tensor(0.5132162402928945, shape=(), dtype=float64)
loss tf.Tensor(0.5191790627562589, shape=(), dtype=float64) val loss tf.Tensor(0.5125322726205608, shape=(), dtype=float64)
loss tf.Tensor(0.5253541363078215, shape=(), dtype=float64) val loss tf.Tensor(0.5103687762148537, shape=(), dtype=float64)
loss tf.Tensor(0.5137894618497632, shape=(), dtvpe=float64) val loss tf.Tensor(0.5126329741052267, shape=(), dtvpe=float64)
loss tf.Tensor(0.5109347437264484, shape=(), dtype=float64) val loss tf.Tensor(0.5111489410830776, shape=(), dtype=float64)
loss tf.Tensor(0.511982794179852, shape=(), dtype=float64) val loss tf.Tensor(0.5198977272035724, shape=(), dtype=float64)
loss tf.Tensor(0.5177734536616434, shape=(), dtype=float64) val loss tf.Tensor(0.5106665128852244, shape=(), dtype=float64)
loss tf.Tensor(0.5116969042983642, shape=(), dtype=float64) val loss tf.Tensor(0.5165137157732602, shape=(), dtype=float64)
loss tf.Tensor(0.5092038370077286, shape=(), dtype=float64) val loss tf.Tensor(0.5214779758534336, shape=(), dtype=float64)
loss tf.Tensor(0.5103259937873792, shape=(), dtype=float64) val_loss tf.Tensor(0.5137889764512019, shape=(), dtype=float64)
loss tf. Tensor (0.521585317213763, shape=(), dtype=float64) val loss tf. Tensor (0.5139719993535377, shape=(), dtype=float64)
loss tf.Tensor(0.5139860644674012, shape=(), dtype=float64) val loss tf.Tensor(0.5091751884853066, shape=(), dtype=float64)
loss tf.Tensor(0.5194482426706601, shape=(), dtype=float64) val loss tf.Tensor(0.5142152177471087, shape=(), dtype=float64)
loss tf.Tensor(0.5277311523814807, shape=(), dtype=float64) val loss tf.Tensor(0.516572866145303, shape=(), dtype=float64)
loss tf.Tensor(0.5275317581187624, shape=(), dtype=float64) val loss tf.Tensor(0.5163296244159018, shape=(), dtype=float64)
loss tf.Tensor(0.5165166439724884, shape=(), dtype=float64) val loss tf.Tensor(0.5216401444847845, shape=(), dtype=float64)
loss tf.Tensor(0.5222676220368603, shape=(), dtype=float64) val loss tf.Tensor(0.5165763816909597, shape=(), dtype=float64)
loss tf.Tensor(0.529621248932491, shape=(), dtype=float64) val loss tf.Tensor(0.5159660732874464, shape=(), dtype=float64)
loss tf.Tensor(0.523467165584597, shape=(), dtype=float64) val loss tf.Tensor(0.5112844699205696, shape=(), dtype=float64)
loss tf.Tensor(0.5167164293050123, shape=(), dtype=float64) val loss tf.Tensor(0.5162130818230658, shape=(), dtype=float64)
loss tf.Tensor(0.5146041560776438, shape=(), dtype=float64) val loss tf.Tensor(0.512288435507049, shape=(), dtype=float64)
loss tf.Tensor(0.5153606554209712, shape=(), dtype=float64) val loss tf.Tensor(0.514768528786061, shape=(), dtype=float64)
loss tf.Tensor(0.5127451097266329, shape=(), dtvpe=float64) val loss tf.Tensor(0.5111088140193374, shape=(), dtvpe=float64)
loss tf.Tensor(0.5177285492384475, shape=(), dtype=float64) val loss tf.Tensor(0.5085369618061211, shape=(), dtype=float64)
loss tf.Tensor(0.5157969115330815, shape=(), dtype=float64) val_loss tf.Tensor(0.5170849031067417, shape=(), dtype=float64)
loss tf.Tensor(0.5200538685976491, shape=(), dtype=float64) val loss tf.Tensor(0.5110168127815878, shape=(), dtype=float64)
loss tf.Tensor(0.5155814958632463, shape=(), dtype=float64) val loss tf.Tensor(0.5118421676207223, shape=(), dtype=float64)
loss tf.Tensor(0.5137902806278498, shape=(), dtype=float64) val loss tf.Tensor(0.5096147667692611, shape=(), dtype=float64)
loss tf.Tensor(0.5202994927257037, shape=(), dtype=float64) val loss tf.Tensor(0.510025462480552, shape=(), dtype=float64)
loss tf. Tensor(0.516016810108774, shape=(), dtype=float64) val loss tf. Tensor(0.5161654938882435, shape=(), dtype=float64)
loss tf.Tensor(0.5169116659842842, shape=(), dtype=float64) val loss tf.Tensor(0.5173557271158268, shape=(), dtype=float64)
loss tf.Tensor(0.510662568305275, shape=(), dtype=float64) val loss tf.Tensor(0.5111941127459524, shape=(), dtype=float64)
loss tf. Tensor(0.513524858127526. shape=(). dtvpe=float64) val loss tf. Tensor(0.5129694240868237. shape=(). dtvpe=float64)
loss tf.Tensor(0.5217809560686683, shape=(), dtype=float64) val loss tf.Tensor(0.515749584229652, shape=(), dtype=float64)
loss tf.Tensor(0.5146141541539442, shape=(), dtype=float64) val loss tf.Tensor(0.5193874062781482, shape=(), dtype=float64)
loss tf.Tensor(0.521246963600418, shape=(), dtype=float64) val loss tf.Tensor(0.5110375766913942, shape=(), dtype=float64)
loss tf.Tensor(0.5123742983787526, shape=(), dtype=float64) val loss tf.Tensor(0.5133331853144293, shape=(), dtype=float64)
```



```
x=3.0
y=2.0
plt.scatter(x_label1, y_label1, c='r', marker='x', s=20)
plt.scatter(x_label2, y_label2, c='g', marker='1', s=20)
plt.scatter(x,y,c='b', marker='s')
plt.show()
```



```
x=7.0
y=6.0
plt.scatter(x_label1, y_label1, c='r', marker='x', s=20)
plt.scatter(x_label2, y_label2, c='g', marker='1', s=20)
plt.scatter(x,y,c='b', marker='s')
plt.show()
```



# → Number of epchos - 100

2

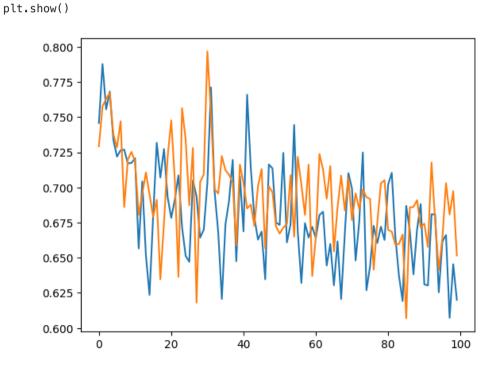
0

8

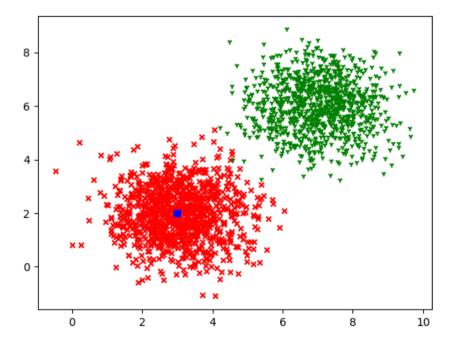
10

```
Loss = []
Val loss = []
epochs = 100
learning rate = 0.1
batch\_size = 20
w = tf.Variable(np.random.random((2, 2)))
b = tf.Variable(np.random.random((2)))
data = np.column stack((xs,ys))
data train,label train,data val,label val = split dataset(data,labels,0.2)
for in range(epochs):
  data batch, labels batch = subset dataset concatenated(data train, label train, batch size)
  data val batch, labels val batch = subset dataset concatenated(data val, label val, batch size)
  with tf.GradientTape() as tape:
    pred l=tf.nn.softmax(tf.matmul(data batch, w) + b)
    pred_l_val=tf.nn.softmax(tf.matmul(data_val_batch, w) + b)
    labels batch one hot = tf.one hot(labels batch, depth=2)
    labels val batch one hot = tf.one hot(labels val batch, depth=2)
    loss = tf.keras.losses.BinaryCrossentropy(from logits=True)(labels batch one hot, pred l)
    Loss.append(loss.numpy())
    val loss = tf.keras.losses.BinaryCrossentropy(from logits=True)(labels val batch one hot, pred l val)
    Val_loss.append(val_loss.numpy())
    print("loss",loss,"val_loss",val_loss)
  dloss_dw,dloss_db = tape.gradient(loss, [w, b])
  w.assign_sub(learning_rate*dloss_dw )
  b.assign sub(learning rate*dloss db )
```

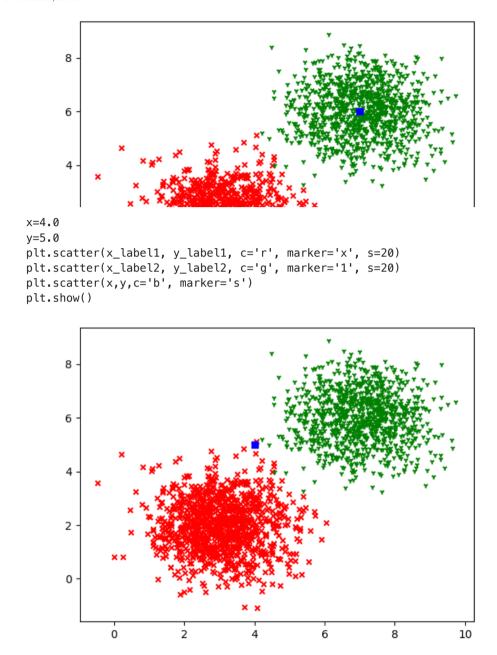
```
LUSS LI.IENSUI(0.07.3393/4910/0034, SNape=(), ULYPE=1LOdLO4) VAL LUSS LI.IENSUI(0.00/1930/1/140401, SNape=(), ULYPE=1LOdLO4)
loss tf.Tensor(0.7246497654675836, shape=(), dtype=float64) val loss tf.Tensor(0.6715672661755365, shape=(), dtype=float64)
loss tf.Tensor(0.6609371367883302, shape=(), dtype=float64) val loss tf.Tensor(0.6739284028778048, shape=(), dtype=float64)
loss tf.Tensor(0.6736872222325825, shape=(), dtype=float64) val loss tf.Tensor(0.7087440940967824, shape=(), dtype=float64)
loss tf.Tensor(0.7445269305375025, shape=(), dtype=float64) val loss tf.Tensor(0.6649892569953518, shape=(), dtype=float64)
loss tf.Tensor(0.6687288281797377, shape=(), dtype=float64) val loss tf.Tensor(0.7217815540626853, shape=(), dtype=float64)
loss tf.Tensor(0.6320415847571409, shape=(), dtype=float64) val loss tf.Tensor(0.7019695016165012, shape=(), dtype=float64)
loss tf.Tensor(0.6746580162563881, shape=(), dtvpe=float64) val loss tf.Tensor(0.6804945477783436, shape=(), dtvpe=float64)
loss tf.Tensor(0.6643928763766408, shape=(), dtype=float64) val loss tf.Tensor(0.7162736208076804, shape=(), dtype=float64)
loss tf.Tensor(0.6719945606367692, shape=(), dtype=float64) val loss tf.Tensor(0.6370117692853579, shape=(), dtype=float64)
loss tf.Tensor(0.6639844689461608, shape=(), dtype=float64) val loss tf.Tensor(0.6647068541250726, shape=(), dtype=float64)
loss tf.Tensor(0.6804345397016153, shape=(), dtype=float64) val loss tf.Tensor(0.7238732801657715, shape=(), dtype=float64)
loss tf.Tensor(0.6827126150558348, shape=(), dtype=float64) val loss tf.Tensor(0.7127205991655815, shape=(), dtype=float64)
loss tf.Tensor(0.644492795199431, shape=(), dtype=float64) val loss tf.Tensor(0.6920971734112922, shape=(), dtype=float64)
loss tf.Tensor(0.6598345707786173, shape=(), dtype=float64) val loss tf.Tensor(0.7152430086286412, shape=(), dtype=float64)
loss tf.Tensor(0.6304325345283306, shape=(), dtype=float64) val loss tf.Tensor(0.6545398400404258, shape=(), dtype=float64)
loss tf.Tensor(0.6616881065420283, shape=(), dtype=float64) val loss tf.Tensor(0.6858405164489018, shape=(), dtype=float64)
loss tf. Tensor(0.620671449366859, shape=(), dtype=float64) val loss tf. Tensor(0.7084741619312193, shape=(), dtype=float64)
loss tf.Tensor(0.6664937455243646, shape=(), dtype=float64) val loss tf.Tensor(0.6837972817732435, shape=(), dtype=float64)
loss tf.Tensor(0.7101757866375109, shape=(), dtype=float64) val loss tf.Tensor(0.7066420417021015, shape=(), dtype=float64)
loss tf.Tensor(0.6992543087946675, shape=(), dtype=float64) val loss tf.Tensor(0.676808101458747, shape=(), dtype=float64)
loss tf. Tensor (0.648066246893346, shape=(), dtvpe=float64) val loss tf. Tensor (0.6958530072446778, shape=(), dtvpe=float64)
loss tf.Tensor(0.6754348190736564, shape=(), dtype=float64) val loss tf.Tensor(0.6847898981616553, shape=(), dtype=float64)
loss tf.Tensor(0.7249233730152513, shape=(), dtype=float64) val loss tf.Tensor(0.698585324121069, shape=(), dtype=float64)
loss tf. Tensor (0.626890140903593, shape=(), dtvpe=float64) val loss tf. Tensor (0.6931068047203345, shape=(), dtvpe=float64)
loss tf. Tensor(0.643984107658615, shape=(), dtype=float64) val loss tf. Tensor(0.6918733068380882, shape=(), dtype=float64)
loss tf.Tensor(0.6727321368212407, shape=(), dtype=float64) val loss tf.Tensor(0.6415228835096695, shape=(), dtype=float64)
loss tf.Tensor(0.6604965854778566, shape=(), dtvpe=float64) val loss tf.Tensor(0.6750646103222412, shape=(), dtvpe=float64)
loss tf.Tensor(0.6721383177676291, shape=(), dtype=float64) val loss tf.Tensor(0.7031128630330268, shape=(), dtype=float64)
loss tf.Tensor(0.6627174108184721, shape=(), dtype=float64) val_loss tf.Tensor(0.7051839350263664, shape=(), dtype=float64)
loss tf.Tensor(0.7021666395689927, shape=(), dtype=float64) val loss tf.Tensor(0.6698325299178501, shape=(), dtype=float64)
loss tf.Tensor(0.710454896450808, shape=(), dtype=float64) val loss tf.Tensor(0.6686069624585107, shape=(), dtype=float64)
loss tf.Tensor(0.6647213582212509, shape=(), dtype=float64) val loss tf.Tensor(0.6593381449068024, shape=(), dtype=float64)
loss tf.Tensor(0.6369359593509291, shape=(), dtype=float64) val loss tf.Tensor(0.6598384277985323, shape=(), dtype=float64)
loss tf.Tensor(0.6192108211476626, shape=(), dtype=float64) val loss tf.Tensor(0.6664981779756161, shape=(), dtype=float64)
loss tf.Tensor(0.6869505253548251, shape=(), dtype=float64) val loss tf.Tensor(0.6069132964204423, shape=(), dtype=float64)
loss tf.Tensor(0.669179178976899, shape=(), dtype=float64) val loss tf.Tensor(0.6858540379882376, shape=(), dtype=float64)
loss tf.Tensor(0.6382145373617545, shape=(), dtype=float64) val_loss tf.Tensor(0.6861475467419955, shape=(), dtype=float64)
loss tf.Tensor(0.6702241097493398, shape=(), dtype=float64) val loss tf.Tensor(0.6910532454448759, shape=(), dtype=float64)
loss tf.Tensor(0.6882342601982667, shape=(), dtype=float64) val loss tf.Tensor(0.6716889833219419, shape=(), dtype=float64)
loss tf.Tensor(0.6311002346575305, shape=(), dtype=float64) val loss tf.Tensor(0.6742937294435515, shape=(), dtype=float64)
loss tf.Tensor(0.630272645983877, shape=(), dtype=float64) val loss tf.Tensor(0.6577118580876269, shape=(), dtype=float64)
loss tf.Tensor(0.6809905752431893, shape=(), dtype=float64) val loss tf.Tensor(0.7177009493918087, shape=(), dtype=float64)
loss tf.Tensor(0.6808002550548535, shape=(), dtype=float64) val loss tf.Tensor(0.6762249705740324, shape=(), dtype=float64)
loss tf.Tensor(0.6254514362833008, shape=(), dtype=float64) val loss tf.Tensor(0.6414428944930146, shape=(), dtype=float64)
loss tf.Tensor(0.6615450971489165, shape=(), dtype=float64) val loss tf.Tensor(0.6638778563990193, shape=(), dtype=float64)
loss tf.Tensor(0.666173578166893, shape=(), dtype=float64) val loss tf.Tensor(0.7030899125675601, shape=(), dtype=float64)
loss tf.Tensor(0.6072781613437183, shape=(), dtype=float64) val loss tf.Tensor(0.6808090278060256, shape=(), dtype=float64)
loss tf.Tensor(0.6454217675241385, shape=(), dtype=float64) val loss tf.Tensor(0.6973908345165002, shape=(), dtype=float64)
loss tf.Tensor(0.6200864000254587, shape=(), dtype=float64) val loss tf.Tensor(0.6516227430630405, shape=(), dtype=float64)
```



```
x=3.0
y=2.0
plt.scatter(x_label1, y_label1, c='r', marker='x', s=20)
plt.scatter(x_label2, y_label2, c='g', marker='1', s=20)
plt.scatter(x,y,c='b', marker='s')
plt.show()
```



```
x=7.0
y=6.0
plt.scatter(x_label1, y_label1, c='r', marker='x', s=20)
plt.scatter(x_label2, y_label2, c='g', marker='1', s=20)
plt.scatter(x,y,c='b', marker='s')
plt.show()
```



# → Number of epochs - 3000

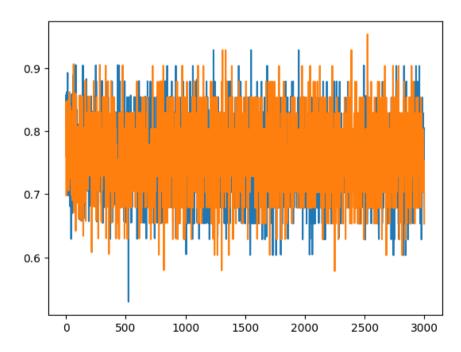
```
Loss = []
Val loss = []
epochs = 3000
learning rate = 0.1
batch size = 20
w = tf.Variable(np.random.random((2, 2)))
b = tf.Variable(np.random.random((2)))
data = np.column stack((xs.ys))
data train, label train, data val, label val = split dataset(data, labels, 0.2)
for _ in range(epochs):
 data batch, labels batch = subset dataset concatenated(data train, label train, batch size)
 data val batch, labels val batch = subset dataset concatenated(data val, label val, batch size)
 with tf.GradientTape() as tape:
    pred l=tf.nn.softmax(tf.matmul(data batch, w) + b)
    pred l val=tf.nn.softmax(tf.matmul(data val batch, w) + b)
    labels batch one hot = tf.one hot(labels batch, depth=2)
    labels_val_batch_one_hot = tf.one_hot(labels_val_batch, depth=2)
    loss = tf.keras.losses.BinaryCrossentropy(from_logits=True)(labels_batch_one_hot, pred_l)
    Loss.append(loss.numpy())
   val loss = tf.keras.losses.BinaryCrossentropy(from logits=True)(labels val batch one hot, pred l val)
    Val loss.append(val loss.numpy())
    print("loss",loss,"val loss",val loss)
 dloss_dw,dloss_db = tape.gradient(loss, [w, b])
 w.assign_sub(learning_rate*dloss_dw )
 b.assign_sub(learning_rate*dloss_db )
```

```
LUSS LI.IEIISUI(W.OWJZOJZZY44YJJJ, SIIdpe=(), ULYpe=ILOdLO4) Val LUSS LI.IEIISUI(W./JJJZO4/Z110YJJY, SIIdpe=(), ULYpe=ILodLO4)
loss tf.Tensor(0.8032780502616876, shape=(), dtype=float64) val loss tf.Tensor(0.7533234681799102, shape=(), dtype=float64)
loss tf.Tensor(0.7533483665090945, shape=(), dtype=float64) val loss tf.Tensor(0.8035295460389478, shape=(), dtype=float64)
loss tf.Tensor(0.7032946386007684, shape=(), dtype=float64) val loss tf.Tensor(0.7532561658746768, shape=(), dtype=float64)
loss tf.Tensor(0.7534063056010446, shape=(), dtype=float64) val loss tf.Tensor(0.7535546572500454, shape=(), dtype=float64)
loss tf.Tensor(0.6783898224984162, shape=(), dtype=float64) val loss tf.Tensor(0.8038315174852343, shape=(), dtype=float64)
loss tf.Tensor(0.8037070722928803, shape=(), dtype=float64) val loss tf.Tensor(0.7533070314333337, shape=(), dtype=float64)
loss tf.Tensor(0.9032382688459677, shape=(), dtype=float64) val loss tf.Tensor(0.7532743550626113, shape=(), dtype=float64)
loss tf.Tensor(0.7782687492560763, shape=(), dtype=float64) val loss tf.Tensor(0.7782871180483786, shape=(), dtype=float64)
loss tf.Tensor(0.7533693808743397, shape=(), dtype=float64) val loss tf.Tensor(0.7285701510759152, shape=(), dtype=float64)
loss tf.Tensor(0.7056159927654255, shape=(), dtype=float64) val loss tf.Tensor(0.8283929638248277, shape=(), dtype=float64)
loss tf.Tensor(0.7034835643612104, shape=(), dtype=float64) val loss tf.Tensor(0.7782697315447761, shape=(), dtype=float64)
loss tf.Tensor(0.8033632907523938, shape=(), dtype=float64) val loss tf.Tensor(0.7282472643100585, shape=(), dtype=float64)
loss tf.Tensor(0.7533752060699157, shape=(), dtype=float64) val loss tf.Tensor(0.8032764687065408, shape=(), dtype=float64)
loss tf.Tensor(0.7783331154594265, shape=(), dtype=float64) val loss tf.Tensor(0.7283546567142685, shape=(), dtype=float64)
loss tf.Tensor(0.6535131927492153, shape=(), dtype=float64) val loss tf.Tensor(0.7533490931411693, shape=(), dtype=float64)
loss tf.Tensor(0.8283786906555124, shape=(), dtype=float64) val loss tf.Tensor(0.7783017180758891, shape=(), dtype=float64)
loss tf.Tensor(0.7533611493984773, shape=(), dtype=float64) val loss tf.Tensor(0.6784726110488796, shape=(), dtype=float64)
loss tf.Tensor(0.704116623233452, shape=(), dtype=float64) val loss tf.Tensor(0.7784643362784186, shape=(), dtype=float64)
loss tf.Tensor(0.703390756582842, shape=(), dtype=float64) val loss tf.Tensor(0.7032723412273862, shape=(), dtype=float64)
loss tf.Tensor(0.7284076625572362, shape=(), dtype=float64) val loss tf.Tensor(0.7784269106768464, shape=(), dtype=float64)
loss tf.Tensor(0.778221016721072, shape=(), dtype=float64) val loss tf.Tensor(0.8533494611654397, shape=(), dtype=float64)
loss tf.Tensor(0.7533284917844537, shape=(), dtype=float64) val loss tf.Tensor(0.7532790832719274, shape=(), dtype=float64)
loss tf.Tensor(0.8036342303892912, shape=(), dtype=float64) val loss tf.Tensor(0.7033634238899007, shape=(), dtype=float64)
loss tf.Tensor(0.7535390795082351, shape=(), dtype=float64) val loss tf.Tensor(0.7033744109229243, shape=(), dtype=float64)
loss tf.Tensor(0.6283602511956633, shape=(), dtype=float64) val loss tf.Tensor(0.8283129716652609, shape=(), dtype=float64)
loss tf.Tensor(0.7285388992589688, shape=(), dtype=float64) val loss tf.Tensor(0.7532914829058963, shape=(), dtype=float64)
loss tf.Tensor(0.6785633737398258, shape=(), dtype=float64) val loss tf.Tensor(0.753440876000771, shape=(), dtype=float64)
loss tf.Tensor(0.728778706116581, shape=(), dtype=float64) val loss tf.Tensor(0.7783316322685636, shape=(), dtype=float64)
loss tf.Tensor(0.7286722085457048, shape=(), dtype=float64) val loss tf.Tensor(0.7532500743220181, shape=(), dtype=float64)
loss tf.Tensor(0.753320398524011, shape=(), dtype=float64) val loss tf.Tensor(0.7533403383343072, shape=(), dtype=float64)
loss tf.Tensor(0.8533338582865444, shape=(), dtype=float64) val loss tf.Tensor(0.7784162706826778, shape=(), dtype=float64)
loss tf.Tensor(0.7034191225864005, shape=(), dtype=float64) val loss tf.Tensor(0.7784408751603162, shape=(), dtype=float64)
loss tf.Tensor(0.7784048084567537, shape=(), dtype=float64) val loss tf.Tensor(0.7035336624657285, shape=(), dtype=float64)
loss tf.Tensor(0.7032849380653443, shape=(), dtype=float64) val loss tf.Tensor(0.7533805142628731, shape=(), dtype=float64)
loss tf.Tensor(0.8033403917415273, shape=(), dtype=float64) val loss tf.Tensor(0.7532628579965575, shape=(), dtype=float64)
loss tf.Tensor(0.778274088483718, shape=(), dtype=float64) val_loss tf.Tensor(0.803258790836421, shape=(), dtype=float64)
loss tf.Tensor(0.7284473422875088, shape=(), dtype=float64) val loss tf.Tensor(0.8033537797825481, shape=(), dtype=float64)
loss tf.Tensor(0.8284304491868875, shape=(), dtype=float64) val loss tf.Tensor(0.7033828352609082, shape=(), dtype=float64)
loss tf.Tensor(0.8282936526883351, shape=(), dtype=float64) val loss tf.Tensor(0.7784444209687258, shape=(), dtype=float64)
loss tf.Tensor(0.7533774959323084, shape=(), dtype=float64) val loss tf.Tensor(0.7283087381297554, shape=(), dtype=float64)
loss tf.Tensor(0.8032914581688239, shape=(), dtype=float64) val loss tf.Tensor(0.7033245631931291, shape=(), dtype=float64)
loss tf.Tensor(0.803367037992361, shape=(), dtype=float64) val loss tf.Tensor(0.7034229251091851, shape=(), dtype=float64)
loss tf.Tensor(0.703674891322912, shape=(), dtype=float64) val_loss tf.Tensor(0.7034470461711948, shape=(), dtype=float64)
loss tf.Tensor(0.7532525550179352, shape=(), dtype=float64) val loss tf.Tensor(0.7033006697212171, shape=(), dtype=float64)
loss tf.Tensor(0.7536761168345661, shape=(), dtype=float64) val loss tf.Tensor(0.7532538054208489, shape=(), dtype=float64)
loss tf.Tensor(0.7536546781899522, shape=(), dtype=float64) val loss tf.Tensor(0.7033189374061586, shape=(), dtype=float64)
loss tf.Tensor(0.8037083499016298, shape=(), dtype=float64) val loss tf.Tensor(0.7034466137543968, shape=(), dtype=float64)
loss tf.Tensor(0.8044590001714678, shape=(), dtype=float64) val loss tf.Tensor(0.65332372774462, shape=(), dtype=float64)
```

print(data train.size, label train.size, data val.size, label val.size)

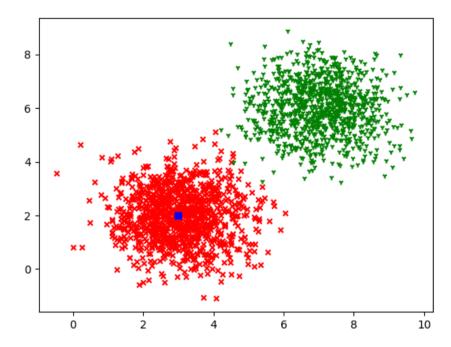
plt.show()

```
3200 1600 800 400
```

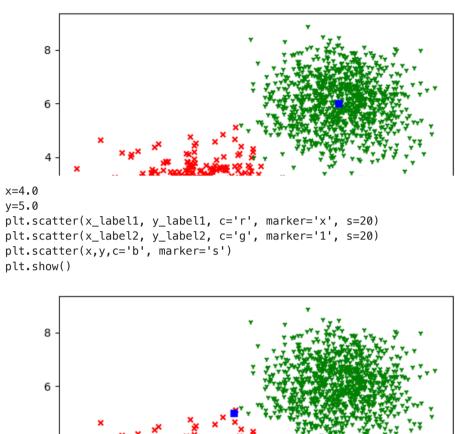


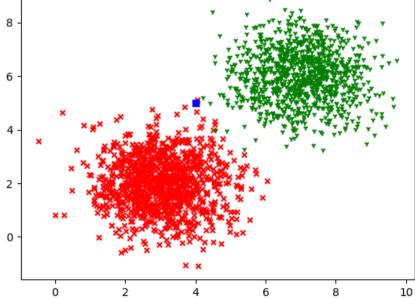
x=3.0  $https://colab.research.google.com/drive/1mZQPaqVV_6ntAh21eJvK45fUhy_jrvw_?hl=pl\#scrollTo=QKymyRl3m_b6\&printMode=true$ 

```
y=2.0
plt.scatter(x_label1, y_label1, c='r', marker='x', s=20)
plt.scatter(x_label2, y_label2, c='g', marker='1', s=20)
plt.scatter(x,y,c='b', marker='s')
plt.show()
```



```
x=7.0
y=6.0
plt.scatter(x_label1, y_label1, c='r', marker='x', s=20)
plt.scatter(x_label2, y_label2, c='g', marker='1', s=20)
plt.scatter(x,y,c='b', marker='s')
plt.show()
```





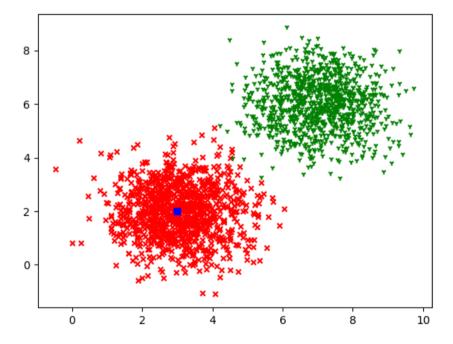
## → Minibatch

#### ▼ Batch size - 10

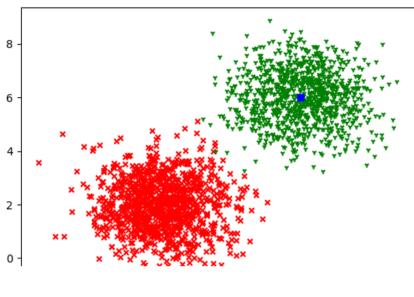
```
Loss = []
Val loss = []
epochs = 5000
learning rate = 0.1
batch_size = 10
w = tf.Variable(np.random.random((2, 2)))
b = tf.Variable(np.random.random((2)))
data = np.column stack((xs,ys))
data train, label train, data val, label val = split dataset(data, labels, 0.2)
for in range(epochs):
 data_batch,labels_batch = subset_dataset_concatenated(data_train,label_train,batch_size)
 data val batch, labels val batch = subset dataset concatenated(data val, label val, batch size)
 with tf.GradientTape() as tape:
    pred l=tf.nn.softmax(tf.matmul(data batch, w) + b)
    pred_l_val=tf.nn.softmax(tf.matmul(data_val_batch, w) + b)
    labels batch one hot = tf.one hot(labels batch, depth=2)
    labels val batch one hot = tf.one hot(labels val batch, depth=2)
    loss = tf.keras.losses.BinaryCrossentropy(from logits=True)(labels batch one hot, pred l)
    Loss.append(loss.numpy())
    val_loss = tf.keras.losses.BinaryCrossentropy(from_logits=True)(labels_val_batch_one_hot, pred_l_val)
    Val loss.append(val loss.numpy())
    print("loss",loss,"val_loss",val_loss)
 dloss_dw,dloss_db = tape.gradient(loss, [w, b])
 w.assign sub(learning rate*dloss dw )
 b.assign sub(learning rate*dloss db )
```

```
LOSS LI.IENSOI(W.314113/92920203, Shape=(), Glype=Float04) Val_LOSS LI.IENSOI(W.3143/40092, Shape=(), Glype=Float04)
    loss tf.Tensor(0.5096694293137294, shape=(), dtype=float64) val loss tf.Tensor(0.5181277790226497, shape=(), dtype=float64)
    loss tf.Tensor(0.5154135442723452, shape=(), dtype=float64) val loss tf.Tensor(0.52041582878872, shape=(), dtype=float64)
    loss tf.Tensor(0.5176992249696943, shape=(), dtype=float64) val loss tf.Tensor(0.5211597038437501, shape=(), dtype=float64)
print(data train.size, label train.size, data val.size, label val.size)
    3200 1600 800 400
np.max(Loss),np.min(Loss)
    (0.8840225828888627, 0.505522010036187)
np.max(Val loss),np.min(Val loss)
    (0.8200778241943407, 0.5052412980904848)
print(w.numpy())
print(b.numpy())
    [[0.22663319 1.15123026]
     [0.20432411 1.49084049]]
    [ 5.1085927 -4.24261619]
plt.plot(Loss)
plt.plot(Val loss)
plt.show()
```

```
x=3.0
y=2.0
plt.scatter(x_label1, y_label1, c='r', marker='x', s=20)
plt.scatter(x_label2, y_label2, c='g', marker='1', s=20)
plt.scatter(x,y,c='b', marker='s')
plt.show()
```



```
x=7.0
y=6.0
plt.scatter(x_label1, y_label1, c='r', marker='x', s=20)
plt.scatter(x_label2, y_label2, c='g', marker='1', s=20)
plt.scatter(x,y,c='b', marker='s')
plt.show()
```

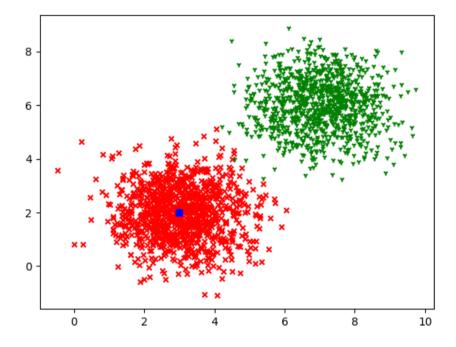


#### → Batch size - 100

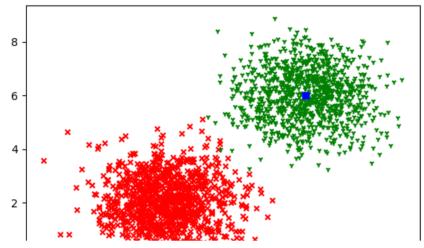
```
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                                                                                                           - TENTINE PROPERTY.
Loss = []
Val loss = []
epochs = 5000
learning rate = 0.1
batch size = 100
w = tf.Variable(np.random.random((2, 2)))
b = tf.Variable(np.random.random((2)))
data = np.column stack((xs.ys))
data train,label train,data val,label val = split dataset(data,labels,0.2)
 for in range(epochs):
     data batch, labels batch = subset dataset concatenated(data train, label train, batch size)
     data val batch, labels val batch = subset dataset concatenated(data val, label val, batch size)
     with tf.GradientTape() as tape:
           pred_l=tf.nn.softmax(tf.matmul(data_batch, w) + b)
           pred l val=tf.nn.softmax(tf.matmul(data val batch, w) + b)
           labels_batch_one_hot = tf.one_hot(labels_batch, depth=2)
           labels val batch one hot = tf.one hot(labels val batch, depth=2)
           loss = tf.keras.losses.BinaryCrossentropy(from logits=True)(labels batch one hot, pred l)
           Loss.append(loss.numpy())
           val loss = tf.keras.losses.BinaryCrossentropy(from logits=True)(labels val batch one hot, pred l val)
           Val loss.append(val loss.numpy())
           print("loss", loss, "val_loss", val_loss)
     dloss_dw,dloss_db = tape.gradient(loss, [w, b])
     w.assign sub(learning rate*dloss dw )
     b.assign_sub(learning_rate*dloss_db )
```

```
LUSS LI.IENSUI(W.D1421/WDD110411, Shape=(), ULype=Floato4) Val_LUSS LI.IENSUI(W.D2190W9020D90204, Shape=(), ULype=Floato4)
    loss tf.Tensor(0.5175655631292828, shape=(), dtype=float64) val loss tf.Tensor(0.5187147889281255, shape=(), dtype=float64)
    loss tf.Tensor(0.5166561068743636, shape=(), dtype=float64) val loss tf.Tensor(0.5181992710894627, shape=(), dtype=float64)
print(data train.size, label train.size, data val.size, label val.size)
    3200 1600 800 400
np.max(Loss),np.min(Loss)
    (0.8007723652234441, 0.5113890253787418)
np.max(Val_loss),np.min(Val_loss)
    (0.8183081053596879, 0.511993478640346)
print(w.numpy())
print(b.numpy())
    [-0.00202623 1.30862547]]
    [ 5.07338022 -4.22840152]
plt.plot(Loss)
plt.plot(Val_loss)
plt.show()
```

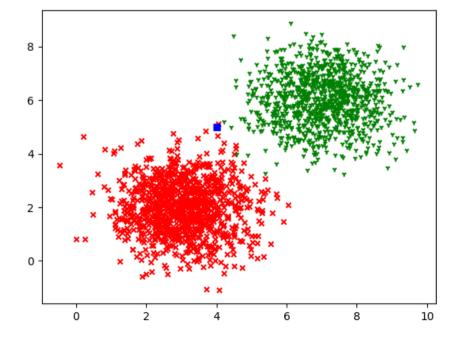
```
x=3.0
y=2.0
plt.scatter(x_label1, y_label1, c='r', marker='x', s=20)
plt.scatter(x_label2, y_label2, c='g', marker='1', s=20)
plt.scatter(x,y,c='b', marker='s')
plt.show()
```



```
x=7.0
y=6.0
plt.scatter(x_label1, y_label1, c='r', marker='x', s=20)
plt.scatter(x_label2, y_label2, c='g', marker='1', s=20)
plt.scatter(x,y,c='b', marker='s')
plt.show()
```



x=4.0
y=5.0
plt.scatter(x\_label1, y\_label1, c='r', marker='x', s=20)
plt.scatter(x\_label2, y\_label2, c='g', marker='1', s=20)
plt.scatter(x,y,c='b', marker='s')
plt.show()



Najlepsze wyniki otrzymałem dla współczynnika uczenia 0.1, liczby epok 5000, batcha równego 20, najgorsze dla współczynnika uczenia 0.001, liczby epok 100, batcha równego 100.

I got the best results for a learning rate of 0.1, a number of epochs of 5000, a batch of 20, and the worst for a learning rate of 0.001, a number of