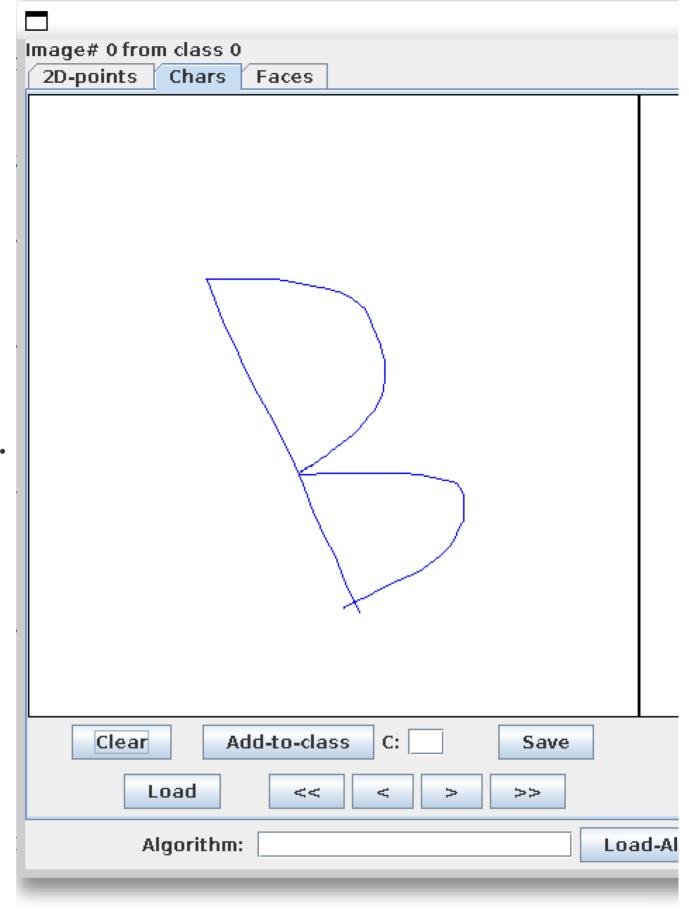
# **Exercise 1**

Yes, google

## **Exercise 2**

• It's harder because the split-uniform dataset overlaps



• The first guy have more varitey

## **Exercise 3**

The screen output has 5 point in each class corresponding to the train set

### **Exercise 4**

- We can use either pixel of the char or vector(start end point of a line segment) of the char
- g.drawLine (x1,y1, x2,y2);
- The char with more curves will create more line segments. Like B or C.

### **Exercise 5**

- The size of the image by the color depth.
- No, they are not the same. it depends on the image size.

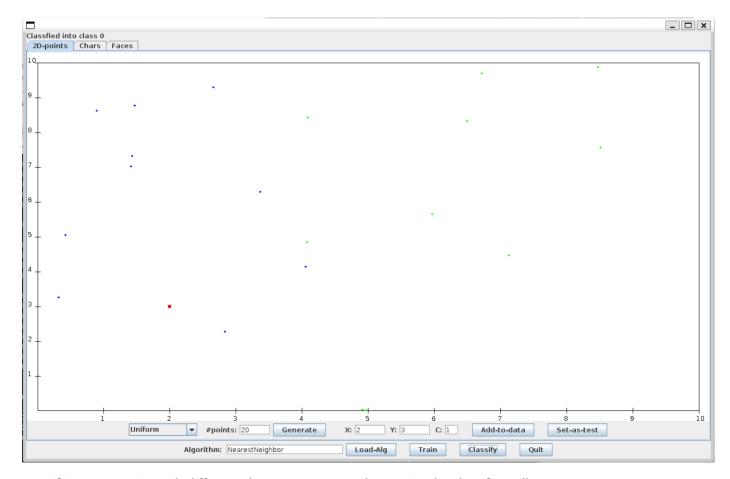
#### **Exercise 6**

The arm problem

### **Exercise 7**

- The third input dimension
- The first value of third input dimension
- The label corresponding to the third input dimension

### **Exercise 8**



- If I create a point with different class near a wrong class point, the classifier will go wrong.
- Distance is meaningless in char problem

### **Exercise 9**

train time: M\*M

run time: M

A priority queue is best for storing nearest neighbor.

A 2d hash map representing approx location of each point is best for calculation.

### **Exercise 10**

1,000,000 \* 0.1 M = 100G Calculations

MSE between two images can represent it's distance

### **Exercise 11**

0.24

After change, Pr[Error]=0.135

Maybe there are a lot of class 0 samples to be tested. So changing answer to class 0 when distance is equal will increase the correctness.

### **Exercise 12**

class0: 0.397

class1: 0.603

### **Exercise 13**

Pr[X=3 | C=0]= 0.091

Pr[X=3 | C=1]= 0.609

- because this two number is the value of prob X3 in C=0. All prob value of X in C=0 add up to 1
- More likely to come from class1

#### **Exercise 14:**

```
1 Pr[X=0|C=1]=0
2 Pr[X=1|C=1]=0.05
3 Pr[X=2|C=1]=0.1
4 Pr[X=3|C=1]=0.6
5 Pr[X=4|C=1]=0.1
6 Pr[X=5|C=1]=0.15
```

## **Exercise 15:**

Pr[E]=0.13099

## **Exercise 16:**

Pr[E3]=0.09967510726420475

After change to 0.9

Pr[E3]=0.5937170010559663

## **Exercise 17:**

• PR =0.4

Pr[X=x | C=0]Pr[C=0] = 0.04

Pr[X=x | C=1]Pr[C=1] = 0.36

• Pr = 0.9

Pr[X=x | C=0]Pr[C=0] = 0.09

Pr[X=x | C=1]Pr[C=1] = 0.06

## **Exercise 18:**

When data is not 1d, like class 0 are in a circle. And outside the circle are all class 1

### **Exercise 19:**

It is decided by different wi but not a single wi

### **Exercise 20:**

We can apply derivative on each wi so it's differentiable

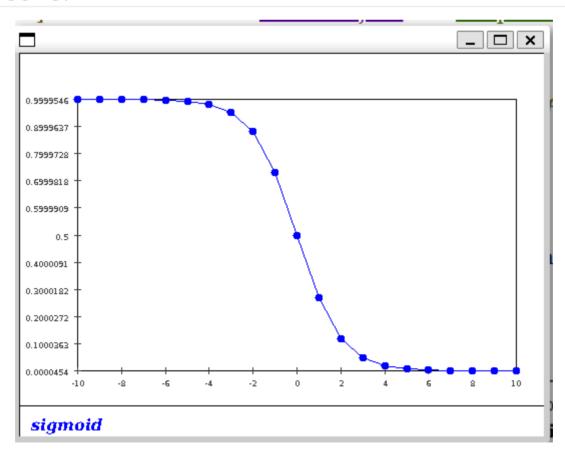
## **Exercise 21:**

• Result doens't changes much

## **Exercise 22:**

To check if the new trained parameter is better or worse than the previous ones.

## **Exercise 23:**



We can compress the x axies of the sigmoid to something like  $f(s)=rac{1}{1+e^{-10s}}$ 

## **Exercise 25:**

#### x1+x2>0

x1\x2	0	1
0	0	1
1	1	0

# **Exercise 26:**

If the points are generated by x axies, then y axies could be ignored.