

Let consider the 2-class case

Risk integral is

$$R(h) = \int_{S_1(h)} L(1,0) p_{y=0} p_{x|y=0} dx + \int_{S_0(h)} L(0,1) p_{y=1} p_{x|y=1} dx$$

We know that

$$p_{y=0} = 1 - p_{y=1}$$

$$R(h) = L(1,0) \int_{S_1(h)} p_{x|y=0} dx +$$

$$L(0,1) (1 - p_{y=0}) \int_{S_0(h)} p_{x|y=1} dx$$

$$= L(0,1) \int_{S_0(h)} p_{x|y=1} dx +$$

$$p_{y=0} \left[L(1,0) \int_{S_1(h)} p_{x|y=0} dx - L(0,1) \int_{S_0(h)} p_{x|y=1} dx \right]$$

This indicates for a fixed classifier risk varies linearly with prior probability.

Minmax Classifier:

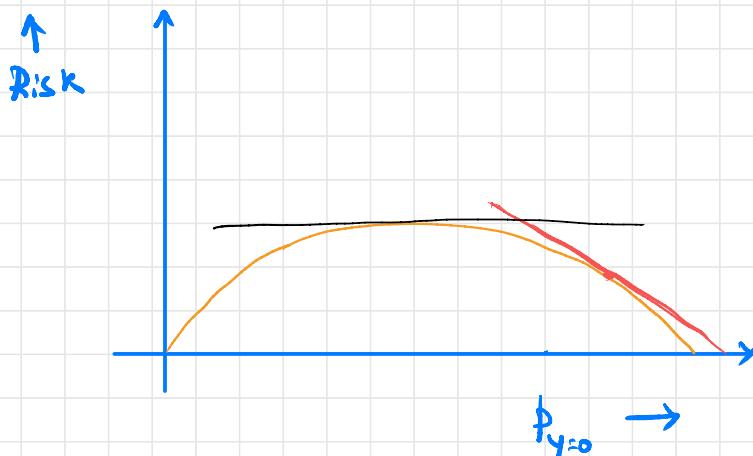
Consider a classifier $s \in$

$$L(1,0) \int_{S_1(h)} p_{x|y=0} dx = L(0,1) \int_{S_0(h)} p_{x|y=1} dx$$

* For this classifier the risk would be

$$R(h_m) = L(0,1) \int_{S_0(h)} p_{x|y=1} dx.$$

Risk would be independent of prior.



Minmax classifier.

Risk grows
Linearly with prior
for any fixed
Classifier.

Bayes Risk.

$$L(1,0) \int_{S_1(h)} p_{x|y=0} dx = L(0,1) \int_{S_0(h)} p_{x|y=1} dx$$

Suppose in the above condition

$$L(1,0) = L(0,1)$$

$$\int_{S_1(h)} p_{x|y=0} dx = \int_{S_0(h)} p_{x|y=1} dx.$$

Now, for a simple case of one dimensional feature
Normal Class Conditional densities.

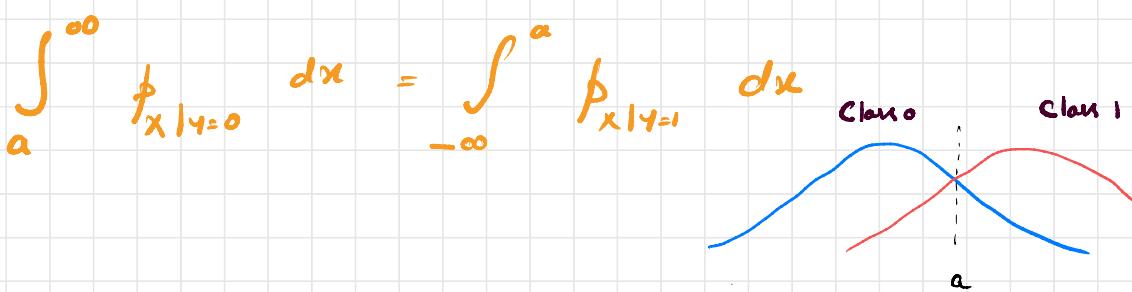
$p_{x|y=i}$ is normal with μ_i, σ_i^2

assume $\mu_0 < \mu_1$

$h(x) = 0$ iff

$x < a$

now we will fix the threshold "a", to satisfy
the minmax criterion.



These will become the integrals of std normal

$$z = \frac{x - \mu_0}{\sigma_0} \quad \text{for 1st integral}$$

$$z = \frac{x - \mu_1}{\sigma_1} \quad \text{for 2nd integral}$$

The threshold should satisfy

$$1 - \phi\left(\frac{a - \mu_0}{\sigma_0}\right) = \phi\left(\frac{a - \mu_1}{\sigma_1}\right)$$

wk

$$1 - \phi(z) = \phi(-z)$$

$$\phi\left(\frac{\mu_0 - a}{\sigma_0}\right) = \phi\left(\frac{a - \mu_1}{\sigma_1}\right)$$

$$\frac{\mu_0 - a}{\sigma_0} = \frac{a - \mu_1}{\sigma_1}$$

$$a = \frac{\mu_0 \sigma_1 + \mu_1 \sigma_0}{\sigma_0 + \sigma_1}$$

now here we can observe that

1) Minmax is a linear
while Bayes Classifier for the same case is
quadratic.

2) When $\sigma_0 = \sigma_1 = \sigma$

Minmax is same as Bayes in this
special case.