

### SML Homework 3

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3.

$$p_k(n) = \frac{\pi_k \frac{1}{\sqrt{2\pi}\sigma_k} \exp\left(-\frac{1}{2\sigma_k^2} (n-\mu_k)^2\right)}{\sum \pi_k \frac{1}{\sqrt{2\pi}\sigma_k} \exp\left(-\frac{1}{2\sigma_k^2} (n-\mu_k)^2\right)}$$

$$\log(p_k(n)) = \log(\pi_k) + \log\left(\frac{1}{\sqrt{2\pi}\sigma_k}\right) + (-)\left(\frac{1}{2\sigma_k^2}\right)(n-\mu_k)^2$$

$$\log\left(\sum \pi_k \frac{1}{\sqrt{2\pi}\sigma_k} \exp\left(-\frac{1}{2\sigma_k^2} (n-\mu_k)^2\right)\right)$$

$$\log(p_k(n)) - \log\left(\sum \pi_k \frac{1}{\sqrt{2\pi}\sigma_k} \exp\left(-\frac{1}{2\sigma_k^2} (n-\mu_k)^2\right)\right) = \log(\pi_k) + \log\left(\frac{1}{\sqrt{2\pi}\sigma_k}\right) - \frac{1}{2\sigma_k^2} (n-\mu_k)^2$$

$$\delta(n) = \log(\pi_k) + \log\left(\frac{1}{\sqrt{2\pi}\sigma_k}\right) - \frac{1}{2\sigma_k^2} (n-\mu_k)^2$$

and thus  $\delta(n)$  is quadratic function of  $n$ .

4. (a)

on average  $\cdot 10^4$  for simplicity, ignoring cases when  $x < 0.05$  and  $x > 0.95$ .

(b)

on average  $\cdot 11$ .

(c)

On average,  $0.10^{100} * 100 = 10^{-98}$ .

(d) As  $p$  increases linearly, observations that are geometrically near decrease exponentially.

c.

$$p=1, \lambda=0.10$$

$$p=2, \lambda=\sqrt{0.10}=0.32$$

$$p=3, \lambda=0.10^{1/3}=0.46$$

$$p=N=0.10^{1/N}$$

5. (a) if the Bayes decision boundary is linear, we expect QDA to outperform on the training set because its greater flexibility will result in a better fit. we expect LDA to outperform QDA on the test set because QDA may overfit the linearity of Bayes decision boundary.

b) if the Bayes decision boundary is non-linear, we anticipate that QDA will outperform on both the training and test sets.

c) we predict that the test prediction accuracy of QDA relative to LDA will improve as sample size  $n$  increases, because a more flexible method will yield a better fit as more samples can be fit and variance is offset by larger sample sizes.

d) False, with fewer sample points, the variance from using a more flexible method, such as QDA, would lead to overfit, yielding a higher test rate than LDA.

6.

$$p(x) = \frac{\exp(\beta_0 + \beta_1 x_1 + \beta_2 x_2)}{1 + \exp(\beta_0 + \beta_1 x_1 + \beta_2 x_2)}$$

$x_1$  = hours studied,  $x_2$  = undergrad GPA

$$\beta_0 = -6, \beta_1 = 0.05, \beta_2 = 1$$



a)

$$x = [40 \text{ hour}, 3.5 \text{ GPA}]$$

$$\begin{aligned} p(x) &= \frac{\exp(-6 + 0.05x_1 + x_2)}{1 + \exp(-6 + 0.05x_1 + x_2)} \\ &= \frac{\exp(-6 + 0.0540 + 3.5)}{1 + \exp(-6 + 0.0540 + 3.5)} \\ &= \frac{\exp(-0.5)}{1 + \exp(-0.5)} \\ &= 37.75\% \end{aligned}$$

b)

$$x = [x_1 \text{ hour}, 3.5 \text{ GPA}]$$

$$\begin{aligned} p(x) &= \frac{\exp(-6 + 0.05x_1 + x_2)}{1 + \exp(-6 + 0.05x_1 + x_2)} \\ 0.50 &= \frac{\exp(-6 + 0.05x_1 + 3.5)}{1 + \exp(-6 + 0.05x_1 + 3.5)} \end{aligned}$$

$$\begin{aligned} 0.50(1 + \exp(-2.5 + 0.05x_1)) &= \exp(-2.5 + 0.05x_1) \\ 0.50 + 0.50\exp(-2.5 + 0.05x_1) &= \exp(-2.5 + 0.05x_1) \\ 0.50 &= 0.50\exp(-2.5 + 0.05x_1) \end{aligned}$$

$$\log(1) = -2.5 + 0.05x_1$$

$$x_1 = 2.5 / 0.05 = 50 \text{ hours.}$$

7.

$$P_X(x) = \frac{\pi_1 \frac{1}{\sqrt{2\pi}\sigma} \exp\left(-\frac{1}{2\sigma^2} (x-\mu_1)^2\right)}{\sum \pi_i \frac{1}{\sqrt{2\pi}\sigma} \exp\left(-\frac{1}{2\sigma^2} (x-\mu_i)^2\right)}$$

$$P_{\text{yes}}(x) = \frac{\pi_{\text{yes}} \exp\left(-\frac{1}{2\sigma^2} (x-\mu_{\text{yes}})^2\right)}{\sum \pi_i \exp\left(-\frac{1}{2\sigma^2} (x-\mu_i)^2\right)}$$

$$= \frac{\pi_{\text{yes}} \exp\left(-\frac{1}{2\sigma^2} (x-\mu_{\text{yes}})^2\right)}{\pi_{\text{yes}} \exp\left(-\frac{1}{2\sigma^2} (x-\mu_{\text{yes}})^2\right) + \pi_{\text{no}} \exp\left(-\frac{1}{2\sigma^2} (x-\mu_{\text{no}})^2\right)}$$

$$= \frac{0.80 \exp\left(-\frac{1}{2 \times 36} (x-10)^2\right)}{0.80 \exp\left(-\frac{1}{2 \times 36} (x-10)^2\right) + 0.20 \exp\left(-\frac{1}{2 \times 36} x^2\right)}$$

$$P_{\text{yes}}(4) = \frac{0.80 \exp\left(-\frac{1}{2 \times 36} (4-10)^2\right)}{0.80 \exp\left(-\frac{1}{2 \times 36} (4-10)^2\right) + 0.20 \exp\left(-\frac{1}{2 \times 36} 4^2\right)} = 75.2\%$$

8.

Given,

logistic regression: 20% training error rate, 30% test error rate  
 KNN (k=1) average error rate of 18%.

for KNN with k=1, the training error rate is 0% because for any training observation, its nearest neighbour will be the response itself. So, KNN has a test error rate of 36%. I would choose logistic regression because of its lower test error rate of 30%.



9.

(a)

$$\frac{P(x)}{1-P(x)} = 0.37$$

$$P(x) = 0.37 (1-P(x))$$

$$1.37 P(x) = 0.37$$

$$P(x) = 0.37 / 1.37 = 0.27$$

(b)

$$\text{odds} = \frac{P(x)}{1-P(x)} = 0.16 / 0.84 = 0.19$$