

Solutions 10

Exercise 2.2.12

Part 1 The conditional Bayes Risk is

$$R(\delta|H_0) = C_{10}P_F(\delta) + C_{00}(1 - P_F(\delta)) \quad (1)$$

and

$$R(\delta|H_1) = C_{11}P_D(\delta) + C_{01}(1 - P_D(\delta)) \quad (2)$$

Hence the Bayes Risk is

$$\begin{aligned} R(\delta, \phi(\mathbf{R})) &= [C_{10}P_F(\delta) + C_{00}(1 - P_F(\delta))](1 - \phi(\mathbf{R})) \\ &\quad + [C_{11}P_D(\delta) + C_{01}(1 - P_D(\delta))]\phi(\mathbf{R}) \\ &= C_{00}(1 - \phi(\mathbf{R})) + C_{01}\phi(\mathbf{R}) \\ &\quad + (C_{10} - C_{00})P_F(\delta)(1 - \phi(\mathbf{R})) + (C_{11} - C_{01})P_D(\delta)\phi(\mathbf{R}), \end{aligned} \quad (3)$$

where $P_F(\delta) = \sum_{y \in \mathcal{Y}_1} P(y|H_0)$ and $P_D(\delta) = \sum_{y \in \mathcal{Y}_1} P(y|H_1)$.

Part 2 The LRT that minimize the Bayes Risk is $L(\delta, \phi(\mathbf{R})) \underset{H_0}{\overset{H_1}{\gtrless}} \tau$, where

$$\tau = \frac{(C_{10} - C_{00})(1 - \phi(\mathbf{R}))}{(C_{11} - C_{01})\phi(\mathbf{R})}. \quad (4)$$

It is a specific decision rule, and the case that $(\delta, \phi(\mathbf{R})) = \tau$ doesn't influence the Bayes Risk since $(C_{10} - C_{00})P_F(\delta)(1 - \phi(\mathbf{R})) + (C_{11} - C_{01})P_D(\delta)\phi(\mathbf{R}) = 0$. Hence a randomized test is not necessary.

Part 3 The slope of and straight-line segment on the ROC is a constant, which is exactly the LRT. For the same LRT, we have the same decision rule hence $P_F(\delta)$ and $P_D(\delta)$ are constant. Then the Bayes Risk is constant.

Part 4 If P_F is continuous function with respect to τ , then $P_F = \alpha$. If it is a discrete function, then we need to find p that

$$p(1 - \alpha^+) + (1 - p)(1 - \alpha^-) = 1 - \alpha \quad (5)$$

and hence $p = \frac{\alpha^- - \alpha}{\alpha^- - \alpha^+}$. Then $\phi(\mathbf{R}) = \frac{\alpha^- - \alpha}{\alpha^- - \alpha^+}$.

Exercise 2.2.15

Part 1

$$\begin{aligned}
\sqrt{2\pi}\operatorname{erfc}(x) &= \int_x^\infty e^{-\frac{t^2}{2}} dt \\
&= \frac{1}{x}e^{-\frac{x^2}{2}} - \int_x^\infty \frac{1}{t^2}e^{-\frac{t^2}{2}} dt \\
&= f_1(x) \\
&= \frac{1}{x}e^{-\frac{x^2}{2}} - \frac{1}{x^3}e^{-\frac{x^2}{2}} + \int_x^\infty \frac{3}{t^4}e^{-\frac{t^2}{2}} dt \\
&= f_2(x)
\end{aligned} \tag{6}$$

We know

$$f_1(x) \leq \frac{1}{x}e^{-\frac{x^2}{2}} \tag{7}$$

and

$$\begin{aligned}
f_2(x) &\geq \frac{1}{x}e^{-\frac{x^2}{2}} - \frac{1}{x^3}e^{-\frac{x^2}{2}} \\
&= \frac{1}{x}e^{-\frac{x^2}{2}} \left(1 - \frac{1}{x^2}\right)
\end{aligned} \tag{8}$$

Hence we finish the proof.

Part 2

Solution 1 By mathematical induction we can finish the proof.

Suppose

$$\operatorname{erfc}(x) = \frac{1}{\sqrt{2\pi}x}e^{-x^2/2} \left[1 + \sum_{m=1}^{n-1} (-1)^m \frac{1 \cdot 3 \cdots (2m-1)}{x^{2m}} + R_n\right] \tag{9}$$

and

$$R_n = (-1)^n x e^{\frac{x^2}{2}} \int_x^{+\infty} \frac{1 \cdot 3 \cdots (2n-1)}{t^{2n}} e^{-\frac{t^2}{2}} dt \tag{10}$$

Firstly, from part 1 we have

$$\begin{aligned}
\operatorname{erfc}(x) &= \frac{1}{\sqrt{2\pi}} \int_x^\infty e^{-\frac{t^2}{2}} dt \\
&= \frac{1}{\sqrt{2\pi}x} e^{-\frac{x^2}{2}} \left[1 - x e^{\frac{x^2}{2}} \int_x^{+\infty} \frac{1}{t^2} e^{-\frac{t^2}{2}} dt\right] \\
&= \frac{1}{\sqrt{2\pi}x} e^{-\frac{x^2}{2}} [1 + R_1] \\
&= \frac{1}{\sqrt{2\pi}x} e^{-\frac{x^2}{2}} \left[1 - \frac{1}{x^2} + x e^{\frac{x^2}{2}} \int_x^\infty \frac{3}{t^4} e^{-\frac{t^2}{2}} dt\right] \\
&= \frac{1}{\sqrt{2\pi}x} e^{-\frac{x^2}{2}} \left[1 - \frac{1}{x^2} + R_2\right]
\end{aligned} \tag{11}$$

Then the assumption holds for $n = 1$ and $n = 2$. Suppose that it holds for the k -th item, then for $k + 1$ -th term we have

$$\operatorname{erfc}(x) = \frac{1}{\sqrt{2\pi}x} e^{-x^2/2} \left(1 + \sum_{m=1}^{k-1} (-1)^m \frac{1 \cdot 3 \cdots (2m-1)}{x^{2m}} + R_k \right) \quad (12)$$

where

$$\begin{aligned} R_k &= (-1)^k x e^{\frac{x^2}{2}} \int_x^{+\infty} \frac{1 \cdot 3 \cdots (2k-1)}{t^{2k}} e^{-\frac{t^2}{2}} dt \\ &= (-1)^k x e^{\frac{x^2}{2}} \left[\frac{1 \cdot 3 \cdots (2k-1)}{x^{2k+1}} e^{-\frac{x^2}{2}} - \int_x^{+\infty} \frac{1 \cdot 3 \cdots (2k-1)(2k+1)}{t^{2k+2}} e^{-\frac{t^2}{2}} dt \right] \\ &= (-1)^k \frac{1 \cdot 3 \cdots (2k-1)}{x^{2k}} + (-1)^{k+1} x e^{\frac{x^2}{2}} \int_x^{+\infty} \frac{1 \cdot 3 \cdots (2k-1)(2k+1)}{t^{2k+2}} e^{-\frac{t^2}{2}} dt \\ &= (-1)^k \frac{1 \cdot 3 \cdots (2k-1)}{x^{2k}} + R_{k+1} \end{aligned} \quad (13)$$

Hence

$$\operatorname{erfc}(x) = \frac{1}{\sqrt{2\pi}x} e^{-x^2/2} \left(1 + \sum_{m=1}^k (-1)^m \frac{1 \cdot 3 \cdots (2m-1)}{x^{2m}} + R_{k+1} \right) \quad (14)$$

and the assumption is right.

From (13) we can get R_k has the same sign with $(-1)^k \frac{1 \cdot 3 \cdots (2k-1)}{x^{2k}}$ when $x > 0$, i.e. R_k has the same sign as the $k + 1$ -th term. Notice that R_{k+1} has the different sign from R_k , then $|R_k| < \frac{1 \cdot 3 \cdots (2k-1)}{x^{2k}}$, i.e. the remainder is less than the magnitude of the $k + 1$ -th term.

Solution 2

$$\begin{aligned} \operatorname{erfc}(x) &= \frac{1}{\sqrt{2\pi}} \int_x^{\infty} e^{-\frac{u^2}{2}} du \\ &= \frac{1}{\sqrt{2\pi}} \frac{1}{x} e^{-\frac{x^2}{2}} - \int_x^{\infty} e^{-\frac{u^2}{2}} \frac{1}{u^2} du \\ &= \frac{1}{\sqrt{2\pi}} \frac{1}{x} e^{-\frac{x^2}{2}} - \frac{1}{x^3} e^{-\frac{x^2}{2}} + \int_x^{\infty} \frac{3}{u^4} e^{-\frac{u^2}{2}} du \\ &\vdots \\ &= \frac{1}{\sqrt{2\pi}} \frac{1}{x} e^{-\frac{x^2}{2}} \left[1 + \sum_{m=1}^{n-1} (-1)^m \frac{1 \cdot 3 \cdots (2m-1)}{x^{2m}} \right] \\ &\quad + \frac{1}{\sqrt{2\pi}} (-1)^n 1 \cdot 3 \cdots (2n-1) \int_x^{\infty} \frac{1}{u^{2n}} e^{-\frac{u^2}{2}} du \end{aligned} \quad (15)$$

If we want to derive R_n , we need prove that

$$\int_x^{\infty} \frac{1}{u^{2n}} e^{-\frac{u^2}{2}} du = \frac{1}{x} e^{-\frac{x^2}{2}} \frac{1}{x^{2n}} \theta \quad (16)$$

And the proof is:

$$\begin{aligned}
\int_x^\infty \frac{1}{u^{2n}} e^{-\frac{u^2}{2}} du &\stackrel{t=\frac{u^2-x^2}{2}}{\implies} \int_0^\infty \frac{1}{(2t+x^2)^n} e^{-(t+\frac{x^2}{2})} \frac{1}{\sqrt{2t+x^2}} dt \\
&= e^{-\frac{x^2}{2}} \cdot \frac{1}{x^{2n+1}} \int_0^\infty e^{-t} \left(1 + \frac{1}{x^2}\right)^{-n-\frac{1}{2}} dt \\
&= e^{-\frac{x^2}{2}} \cdot \frac{1}{x^{2n+1}} \theta
\end{aligned} \tag{17}$$

Therefore,

$$\operatorname{erfc}(x) = \frac{1}{\sqrt{2\pi}x} e^{-\frac{x^2}{2}} \left[1 + \sum_{m=1}^{n-1} (-1)^m \frac{1 \cdot 3 \cdots (2m-1)}{x^{2m}} + R_n \right] \tag{18}$$

where

$$R_n = \left[(-1)^n \cdot \frac{1 \cdot 3 \cdots (2n-1)}{2^{2n}} \right] \int_0^\infty e^{-t} \left(1 + \frac{2t}{x^2}\right)^{-n-\frac{1}{2}} dt \tag{19}$$

Exercise 2.2.17

Part 1 The LRT is

$$\begin{aligned}
L(X_1, X_2) &= \frac{p_{x_1, x_2|H_1}(x_1, x_2|H_1)}{p_{x_1, x_2|H_0}(x_1, x_2|H_0)} \\
&= \frac{\sigma_0 \left[\exp\left(\frac{X_1^2}{2\sigma_0^2} - \frac{X_1^2}{2\sigma_1^2}\right) + \exp\left(\frac{X_2^2}{2\sigma_0^2} - \frac{X_2^2}{2\sigma_1^2}\right) \right]}{2\sigma_1}
\end{aligned} \tag{20}$$

Part 2 We have

$$\begin{aligned}
P_D &= P(H_1|H_1) \\
&= \int_{L(x_1, x_2) \geq \tau} p_{x_1, x_2|H_1}(x_1, x_2|H_1) dx_1 dx_2 \\
&= \frac{1}{4\pi\sigma_1\sigma_0} \int_{L(x_1, x_2) \geq \tau} \left[\exp\left(-\frac{X_1^2}{2\sigma_1^2} - \frac{X_2^2}{2\sigma_0^2}\right) + \exp\left(-\frac{X_1^2}{2\sigma_0^2} - \frac{X_2^2}{2\sigma_1^2}\right) \right] dx_1 dx_2
\end{aligned} \tag{21}$$

and

$$\begin{aligned}
P_F &= P(H_1|H_0) \\
&= \int_{L(x_1, x_2) \geq \tau} p_{x_1, x_2|H_0}(x_1, x_2|H_0) dx_1 dx_2 \\
&= \frac{1}{2\pi\sigma_0^2} \int_{L(x_1, x_2) \geq \tau} \exp\left(-\frac{X_1^2}{2\sigma_0^2} - \frac{X_2^2}{2\sigma_0^2}\right) dx_1 dx_2
\end{aligned} \tag{22}$$

Let $L(x_1, x_2) \geq \tau$, we have

$$\exp\left(\frac{\sigma_1^2 - \sigma_0^2}{2\sigma_0^2\sigma_1^2} x_1^2\right) + \exp\left(\frac{\sigma_1^2 - \sigma_0^2}{2\sigma_0^2\sigma_1^2} x_2^2\right) \geq \frac{2\tau\sigma_1}{\sigma_0} \tag{23}$$

And here we need divide the problem into 2 cases to discuss.

- 1) If $\sigma_1 > \sigma_0$, then the upper bound region is

$$|x| \geq \sqrt{\frac{2\sigma_0^2\sigma_1^2}{\sigma_1^2 - \sigma_0^2} \ln\left(\frac{2\tau\sigma_1}{\sigma_0} - 1\right)} = C_1 \quad (24)$$

and the lower bound region is

$$|x| \geq \sqrt{\frac{2\sigma_0^2\sigma_1^2}{\sigma_1^2 - \sigma_0^2} \ln\left(\frac{\tau\sigma_1}{\sigma_0}\right)} = C_2 \quad (25)$$

The upper bound of P_D :

$$\begin{aligned} P_D &= P(H_1|H_1) \\ &= \int_{L(x_1, x_2) \geq \tau} p_{x_1, x_2|H_1}(x_1, x_2|H_1) dx_1 dx_2 \\ &= \frac{1}{4\pi\sigma_1\sigma_0} \int_{L(x_1, x_2) \geq \tau} \left[\exp\left(-\frac{X_1^2}{2\sigma_1^2} - \frac{X_2^2}{2\sigma_0^2}\right) + \exp\left(-\frac{X_1^2}{2\sigma_0^2} - \frac{X_2^2}{2\sigma_1^2}\right) \right] dx_1 dx_2 \\ &\leq \frac{1}{\pi\sigma_1\sigma_0} \int_{x_1=C_1}^{\infty} \int_{x_2=C_1}^{\infty} \left[\exp\left(-\frac{X_1^2}{2\sigma_1^2} - \frac{X_2^2}{2\sigma_0^2}\right) + \exp\left(-\frac{X_1^2}{2\sigma_0^2} - \frac{X_2^2}{2\sigma_1^2}\right) \right] dx_1 dx_2 \\ &= 2 \frac{1}{\sqrt{2\pi}\sigma_1} \frac{1}{\sqrt{2\pi}\sigma_0} \int_{x_1=C_1}^{\infty} \int_{x_2=C_1}^{\infty} \left[\exp\left(-\frac{X_1^2}{2\sigma_1^2} - \frac{X_2^2}{2\sigma_0^2}\right) + \exp\left(-\frac{X_1^2}{2\sigma_0^2} - \frac{X_2^2}{2\sigma_1^2}\right) \right] dx_1 dx_2 \\ &= 4Q\left(\frac{C_1}{\sigma_1}\right)Q\left(\frac{C_1}{\sigma_0}\right) \end{aligned} \quad (26)$$

The lower bound of P_D :

$$\begin{aligned} P_D &= P(H_1|H_1) \\ &= \int_{L(x_1, x_2) \geq \tau} p_{x_1, x_2|H_1}(x_1, x_2|H_1) dx_1 dx_2 \\ &= \frac{1}{4\pi\sigma_1\sigma_0} \int_{L(x_1, x_2) \geq \tau} \left[\exp\left(-\frac{X_1^2}{2\sigma_1^2} - \frac{X_2^2}{2\sigma_0^2}\right) + \exp\left(-\frac{X_1^2}{2\sigma_0^2} - \frac{X_2^2}{2\sigma_1^2}\right) \right] dx_1 dx_2 \\ &\geq \frac{1}{\pi\sigma_1\sigma_0} \int_{x_1=C_2}^{\infty} \int_{x_2=C_2}^{\infty} \left[\exp\left(-\frac{X_1^2}{2\sigma_1^2} - \frac{X_2^2}{2\sigma_0^2}\right) + \exp\left(-\frac{X_1^2}{2\sigma_0^2} - \frac{X_2^2}{2\sigma_1^2}\right) \right] dx_1 dx_2 \\ &= 2 \frac{1}{\sqrt{2\pi}\sigma_1} \frac{1}{\sqrt{2\pi}\sigma_0} \int_{x_1=C_2}^{\infty} \int_{x_2=C_2}^{\infty} \left[\exp\left(-\frac{X_1^2}{2\sigma_1^2} - \frac{X_2^2}{2\sigma_0^2}\right) + \exp\left(-\frac{X_1^2}{2\sigma_0^2} - \frac{X_2^2}{2\sigma_1^2}\right) \right] dx_1 dx_2 \\ &= 4Q\left(\frac{C_2}{\sigma_1}\right)Q\left(\frac{C_2}{\sigma_0}\right) \end{aligned} \quad (27)$$

The upper bound of P_F :

$$\begin{aligned}
P_F &= P(H_1|H_0) \\
&= \int_{L(x_1, x_2) \geq \tau} p_{x_1, x_2|H_0}(x_1, x_2|H_0) dx_1 dx_2 \\
&= \frac{1}{2\pi\sigma_0^2} \int_{L(x_1, x_2) \geq \tau} \exp\left(-\frac{X_1^2}{2\sigma_0^2} - \frac{X_2^2}{2\sigma_0^2}\right) dx_1 dx_2 \\
&\leq \frac{1}{2\pi\sigma_0^2} \int_{x_1=C_1}^{\infty} \int_{x_2=C_1}^{\infty} \exp\left(-\frac{X_1^2}{2\sigma_0^2} - \frac{X_2^2}{2\sigma_0^2}\right) dx_1 dx_2 \\
&= \frac{1}{\sqrt{2\pi}\sigma_0} \frac{1}{\sqrt{2\pi}\sigma_0} \int_{x_1=C_1}^{\infty} \int_{x_2=C_1}^{\infty} \exp\left(-\frac{X_1^2}{2\sigma_0^2} - \frac{X_2^2}{2\sigma_0^2}\right) dx_1 dx_2 \\
&= Q^2\left(\frac{C_1}{\sigma_0}\right)
\end{aligned} \tag{28}$$

The lower bound of P_F :

$$\begin{aligned}
P_F &= P(H_1|H_0) \\
&= \int_{L(x_1, x_2) \geq \tau} p_{x_1, x_2|H_0}(x_1, x_2|H_0) dx_1 dx_2 \\
&= \frac{1}{2\pi\sigma_0^2} \int_{L(x_1, x_2) \geq \tau} \exp\left(-\frac{X_1^2}{2\sigma_0^2} - \frac{X_2^2}{2\sigma_0^2}\right) dx_1 dx_2 \\
&\geq \frac{1}{2\pi\sigma_0^2} \int_{x_1=C_2}^{\infty} \int_{x_2=C_2}^{\infty} \exp\left(-\frac{X_1^2}{2\sigma_0^2} - \frac{X_2^2}{2\sigma_0^2}\right) dx_1 dx_2 \\
&= \frac{1}{\sqrt{2\pi}\sigma_0} \frac{1}{\sqrt{2\pi}\sigma_0} \int_{x_1=C_2}^{\infty} \int_{x_2=C_2}^{\infty} \exp\left(-\frac{X_1^2}{2\sigma_0^2} - \frac{X_2^2}{2\sigma_0^2}\right) dx_1 dx_2 \\
&= Q^2\left(\frac{C_2}{\sigma_0}\right)
\end{aligned} \tag{29}$$

- 2) If $\sigma_1 < \sigma_0$, then the upper bound region is

$$|x| \leq \sqrt{\frac{2\sigma_0^2\sigma_1^2}{\sigma_1^2 - \sigma_0^2} \ln\left(\frac{2\tau\sigma_1}{\sigma_0} - 1\right)} = C_1 \tag{30}$$

and the lower bound region is

$$|x| \leq \sqrt{\frac{2\sigma_0^2\sigma_1^2}{\sigma_1^2 - \sigma_0^2} \ln\left(\frac{\tau\sigma_1}{\sigma_0}\right)} = C_2 \tag{31}$$

The upper bound of P_D :

$$\begin{aligned}
P_D &= P(H_1|H_1) \\
&= \int_{L(x_1, x_2) \geq \tau} p_{x_1, x_2|H_1}(x_1, x_2|H_1) dx_1 dx_2 \\
&= \frac{1}{4\pi\sigma_1\sigma_0} \int_{L(x_1, x_2) \geq \tau} \left[\exp\left(-\frac{X_1^2}{2\sigma_1^2} - \frac{X_2^2}{2\sigma_0^2}\right) + \exp\left(-\frac{X_1^2}{2\sigma_0^2} - \frac{X_2^2}{2\sigma_1^2}\right) \right] dx_1 dx_2 \\
&\leq \frac{1}{4\pi\sigma_1\sigma_0} \int_{x_1=-C_1}^{C_1} \int_{x_2=-C_1}^{C_1} \left[\exp\left(-\frac{X_1^2}{2\sigma_1^2} - \frac{X_2^2}{2\sigma_0^2}\right) + \exp\left(-\frac{X_1^2}{2\sigma_0^2} - \frac{X_2^2}{2\sigma_1^2}\right) \right] dx_1 dx_2 \\
&= \frac{1}{2} \frac{1}{\sqrt{2\pi}\sigma_1} \frac{1}{\sqrt{2\pi}\sigma_0} \int_{x_1=-C_1}^{C_1} \int_{x_2=-C_1}^{C_1} \left[\exp\left(-\frac{X_1^2}{2\sigma_1^2} - \frac{X_2^2}{2\sigma_0^2}\right) + \exp\left(-\frac{X_1^2}{2\sigma_0^2} - \frac{X_2^2}{2\sigma_1^2}\right) \right] dx_1 dx_2 \\
&= \left(1 - 2Q\left(\frac{C_1}{\sigma_1}\right)\right) \left(1 - 2Q\left(\frac{C_1}{\sigma_0}\right)\right)
\end{aligned} \tag{32}$$

The lower bound of P_D :

$$\begin{aligned}
P_D &= P(H_1|H_1) \\
&= \int_{L(x_1, x_2) \geq \tau} p_{x_1, x_2|H_1}(x_1, x_2|H_1) dx_1 dx_2 \\
&= \frac{1}{4\pi\sigma_1\sigma_0} \int_{L(x_1, x_2) \geq \tau} \left[\exp\left(-\frac{X_1^2}{2\sigma_1^2} - \frac{X_2^2}{2\sigma_0^2}\right) + \exp\left(-\frac{X_1^2}{2\sigma_0^2} - \frac{X_2^2}{2\sigma_1^2}\right) \right] dx_1 dx_2 \\
&\leq \frac{1}{4\pi\sigma_1\sigma_0} \int_{x_1=-C_2}^{C_2} \int_{x_2=-C_2}^{C_2} \left[\exp\left(-\frac{X_1^2}{2\sigma_1^2} - \frac{X_2^2}{2\sigma_0^2}\right) + \exp\left(-\frac{X_1^2}{2\sigma_0^2} - \frac{X_2^2}{2\sigma_1^2}\right) \right] dx_1 dx_2 \\
&= \frac{1}{2} \frac{1}{\sqrt{2\pi}\sigma_1} \frac{1}{\sqrt{2\pi}\sigma_0} \int_{x_1=-C_2}^{C_2} \int_{x_2=-C_2}^{C_2} \left[\exp\left(-\frac{X_1^2}{2\sigma_1^2} - \frac{X_2^2}{2\sigma_0^2}\right) + \exp\left(-\frac{X_1^2}{2\sigma_0^2} - \frac{X_2^2}{2\sigma_1^2}\right) \right] dx_1 dx_2 \\
&= \left(1 - 2Q\left(\frac{C_2}{\sigma_1}\right)\right) \left(1 - 2Q\left(\frac{C_2}{\sigma_0}\right)\right)
\end{aligned} \tag{33}$$

The upper bound of P_F :

$$\begin{aligned}
P_F &= P(H_1|H_0) \\
&= \int_{L(x_1, x_2) \geq \tau} p_{x_1, x_2|H_0}(x_1, x_2|H_0) dx_1 dx_2 \\
&= \frac{1}{2\pi\sigma_0^2} \int_{L(x_1, x_2) \geq \tau} \exp\left(-\frac{X_1^2}{2\sigma_0^2} - \frac{X_2^2}{2\sigma_0^2}\right) dx_1 dx_2 \\
&\leq \frac{1}{2\pi\sigma_0^2} \int_{x_1=-C_1}^{C_1} \int_{x_2=-C_1}^{C_1} \exp\left(-\frac{X_1^2}{2\sigma_0^2} - \frac{X_2^2}{2\sigma_0^2}\right) dx_1 dx_2 \\
&= \frac{1}{\sqrt{2\pi}\sigma_0} \frac{1}{\sqrt{2\pi}\sigma_0} \int_{x_1=-C_1}^{C_1} \int_{x_2=-C_1}^{C_1} \exp\left(-\frac{X_1^2}{2\sigma_0^2} - \frac{X_2^2}{2\sigma_0^2}\right) dx_1 dx_2 \\
&= \left(1 - 2Q\left(\frac{C_1}{\sigma_0}\right)\right)^2
\end{aligned} \tag{34}$$

The lower bound of P_F :

$$\begin{aligned}
P_F &= P(H_1|H_0) \\
&= \int_{L(x_1, x_2) \geq \tau} p_{x_1, x_2|H_0}(x_1, x_2|H_0) dx_1 dx_2 \\
&= \frac{1}{2\pi\sigma_0^2} \int_{L(x_1, x_2) \geq \tau} \exp\left(-\frac{X_1^2}{2\sigma_0^2} - \frac{X_2^2}{2\sigma_0^2}\right) dx_1 dx_2 \\
&\leq \frac{1}{2\pi\sigma_0^2} \int_{x_1=-C_2}^{C_2} \int_{x_2=-C_2}^{C_2} \exp\left(-\frac{X_1^2}{2\sigma_0^2} - \frac{X_2^2}{2\sigma_0^2}\right) dx_1 dx_2 \\
&= \frac{1}{\sqrt{2\pi}\sigma_0} \frac{1}{\sqrt{2\pi}\sigma_0} \int_{x_1=-C_2}^{C_2} \int_{x_2=-C_2}^{C_2} \exp\left(-\frac{X_1^2}{2\sigma_0^2} - \frac{X_2^2}{2\sigma_0^2}\right) dx_1 dx_2 \\
&= \left(1 - 2Q\left(\frac{C_2}{\sigma_0}\right)\right)^2
\end{aligned} \tag{35}$$

Exercise 2.2.19

Part 1 The LRT is

$$\begin{aligned}
L(R) &= \prod_{i=1}^N \frac{p_{r_i|H_1}(R_i|H_1)}{p_{r_i|H_0}(R_i|H_0)} \\
&= \left(\frac{\sigma_0}{\sigma_1}\right)^N \prod_{i=1}^N \exp\left[\frac{(R_i - m_0)^2}{2\sigma_0^2} - \frac{(R_i - m_1)^2}{2\sigma_1^2}\right] \\
&= \left(\frac{\sigma_0}{\sigma_1}\right)^N \prod_{i=1}^N \exp\left[\frac{\sigma_1^2(R_i - m_0)^2 - \sigma_0^2(R_i - m_1)^2}{2\sigma_0^2\sigma_1^2}\right] \\
&= \left(\frac{\sigma_0}{\sigma_1}\right)^N \prod_{i=1}^N \exp\left[\frac{(\sigma_1^2 - \sigma_0^2)R_i^2 - 2R_i(\sigma_1^2 m_0 - \sigma_0^2 m_1) + \sigma_1^2 m_0^2 - \sigma_0^2 m_1^2}{2\sigma_0^2\sigma_1^2}\right] \\
&= \left(\frac{\sigma_0}{\sigma_1}\right)^N \exp\left[\frac{(\sigma_1^2 - \sigma_0^2)l_\beta - 2l_\alpha(\sigma_1^2 m_0 - \sigma_0^2 m_1) + N\sigma_1^2 m_0^2 - N\sigma_0^2 m_1^2}{2\sigma_0^2\sigma_1^2}\right]
\end{aligned} \tag{36}$$

Part 2 If $2m_0 = m_1$, $2\sigma_1 = \sigma_0$, we can derive the LRT is

$$\begin{aligned}
L(R) &= 2^N \exp\left(\frac{-3l_\beta + 14l_\alpha m_0 - 15Nm_0^2}{8\sigma_1^2}\right) \underset{H_0}{\overset{H_1}{\geq}} \tau \\
&\implies l_\beta \underset{H_1}{\overset{H_0}{\geq}} -\frac{8}{3}\sigma_1^2(\ln \tau - N \ln 2) + \frac{14}{3}l_\alpha m_0 - 5Nm_0^2.
\end{aligned} \tag{37}$$

Therefore, there is a line separate the l_α, l_β -plane to two parts. The slope and the intercept of the line are

$$\begin{cases} k = \frac{14}{3}m_0 \\ b = -\frac{8}{3}(\ln \tau - N \ln 2)\sigma_1^2 - 5Nm_0^2 \end{cases} \tag{38}$$

The region above the line is H_0 and the other is H_1 .