(1) [20 points] Bayesian Chicken Farmer

Probabilities:

$$P(\omega_{Healthy}) = 0.99, P(\omega_{Sick}) = 0.01$$

$$P(t_{Normal} \mid \omega_{Healthy}) = 0.99, P(t_{High} \mid \omega_{Healthy}) = 0.01$$

$$P(t_{High} | \omega_{Sick}) = 0.99, P(t_{Normal} | \omega_{Sick}) = 0.01$$

State ω Action a	healthy	sick
sell	-\$1	\$500
kill	\$0.5	\$0.5

The Cost Table: $\lambda(a|\omega)$

(1.1) [10 points] Compute the total risks of the following three decision functions

- 1. Sell regardless of t: $f_1(t) = sell$
- 2. *Kill* regardless of t: $f_2(t) = kill$
- 3. $f_3(t_{Normal}) = sell$, $f_3(t_{High}) = kill$
- (1.2) [10 points) Assuming you are a policy maker and the above f_3 is chosen, to allow this industry to be profitable, what is the upper limit of $\lambda(a=sell \mid \omega=sick)$ if the rest costs remain the same? Explain your choice.
- (2) [20 points] Consider the following decision rule for a two-category one-dimensional problem: Classify \vec{x} into class 1 if $x > \tau_0$; otherwise classify x into class 2.

(2.1) [**5 points**] Show that the probability of error for this rule is given by:
$$P(error) = P(\omega_1) \int_{-\infty}^{\tau_0} p(x|\omega_1) dx + P(\omega_2) \int_{\tau_0}^{\infty} p(x|\omega_2) dx$$

(2.2) [10 points] Show that a necessary condition to minimize P(error) is that τ_0 satisfies:

$$P(\omega_1)p(\tau_0|\omega_1) = p(\tau_0|\omega_2)P(\omega_2)$$

- (2.3) [5 points] Give an example where a value of τ_0 satisfying the condition defined in (2.2) actually maximizes P(error).
- (3) [40 points] In many pattern classification problems, one has the option either to assign the pattern to one of c classes, or to reject it as being unrecognizable. Let

$$\lambda(a_i|\omega_j) = \begin{cases} 0 & i=j; 1 \leq i, j \leq c \\ \lambda_0 & reject \\ \lambda_s & otherwise \end{cases}$$

- (3.1) [10 points] Show the minimal risk is obtained if we classify x into the i-th class if $P(\omega_i|x) \ge$ $P(\omega_i|x)$ for all j and if $P(\omega_i|x) \ge 1 - \lambda_0/\lambda_s$, and reject otherwise.
- (3.2) [15 points] Show the following discriminant functions are optimal for such problems:

$$g_i(x) = \begin{cases} p(x|\omega_i)P(\omega_i) & i = 1, ..., c \\ \frac{\lambda_s - \lambda_r}{\lambda_s} \sum_{j=1}^c p(x|\omega_j)P(\omega_j) & reject \end{cases}$$

- (3.3) [**5 points**] What happens if $\lambda_r = 0$?
- (3.4) [**5 points**] What happen if $\lambda_0 > \lambda_s$?
- (3.5) [5 points] Describe qualitatively what happens as λ_0/λ_s is increase from 0 to 1.
- (4) [20 points] Calculate the entropy of the following distribution

HW2

(4.1) [10 points] $p(x) = \frac{1}{(2\pi)^{1/2}\sigma} \exp \left[-\frac{1}{2} \left(\frac{x-\mu}{\sigma} \right)^2 \right]$

(4.2) [10 points] $p(\vec{x}) = \frac{1}{(2\pi)^{D/2} |\Lambda|^{1/2}} \exp\left[-\frac{1}{2} (\vec{x} - \vec{\mu})^T \Lambda^{-1} (\vec{x} - \vec{\mu})\right]$