

ECE 313: Solutions to the Final Exam

1. (a) A and B are two events such that $0 < P(A) < 1$ and $0 < P(B) < 1$.

TRUE $P(A \cup B) \geq \max\{P(A), P(B)\}$. ($A \subset A \cup B, B \subset A \cup B$)

TRUE $P(A \cap B) \leq \min\{P(A), P(B)\}$. ($A \cap B \subset A, A \cap B \subset B$)

TRUE If A and B are independent, $P(A|B) + P(A^c|B^c) = 1$.

TRUE $P(A|B)P(B) + P(A^c|B^c)P(B^c) = 1 - P(A \oplus B)$.

FALSE $P(A|B)P(B) + P(A^c|B)P(B) = P(A)$. (true if RHS were $P(B)$)

TRUE If $P(A) = P(B)$, then $P(A|B) = P(B|A)$.

FALSE If $P(A|B) = P(B|A)$, then $P(A) = P(B)$. Conclusion does not hold if $P(A|B) = P(B|A) = 0$.

TRUE If $P(A|B) = P(A)$, then $P(B^c|A) = 1 - P(B)$. A and B are independent events

- (b) \mathcal{X} and \mathcal{Y} are random variables such that $\text{var}(\mathcal{X}) = \text{var}(\mathcal{Y}) = \sigma^2 < \infty$.

$\text{var}(2\mathcal{X} + 3\mathcal{Y} + 4) = 4\sigma^2 + 9\sigma^2 + 12 \cdot \text{cov}(\mathcal{X}, \mathcal{Y})$

$= \text{var}(3\mathcal{X} - 2\mathcal{Y} + 1) = 9\sigma^2 + 4\sigma^2 - 12 \cdot \text{cov}(\mathcal{X}, \mathcal{Y})$ implies that $\text{cov}(\mathcal{X}, \mathcal{Y}) = 0$.

TRUE \mathcal{X} and \mathcal{Y} are *uncorrelated* random variables.

FALSE \mathcal{X} and \mathcal{Y} are *independent* random variables.

TRUE $\text{var}(2\mathcal{X} + 3\mathcal{Y} + 4) = \text{var}(2\mathcal{X} - 3\mathcal{Y} + 1)$.

TRUE $\text{cov}(\mathcal{X} + \mathcal{Y}, \mathcal{X} - \mathcal{Y}) = 0$.

2. At the Democratic National Convention (DNC), Hillary Clinton and Barack Obama have equal numbers of delegates committed to them, and neither candidate can win the nomination on a ballot. In desperation, the DNC decides to have a *series of debates* between the candidates to decide the Democratic nominee. Hillary wins a debate (event H) with probability p , and Barack wins a debate (event B) with probability $q = 1 - p$. *There are no draws.* The debates continue until one of the candidates wins *two debates in a row* and is declared the Democratic nominee. Successive debates can be regarded as independent trials of an experiment, and \mathcal{X} denotes the total number of debates.

Express the answers to the following questions in terms of p and q , that is, do not write (say) pq as $p(1-p)$ or multiply it out as $p - p^2$. On the other hand, feel free to simplify $p + q = 1$.

- (a) For $n > 0$, find the probability that *more than* $2n$ debates occur at the DNC.

Solution: If $2n$ debates have not produced a winner, then the winners *must* have alternated, that is, the result of these $2n$ debates *must* have been either $HBHB \cdots HB$ or $BHBH \cdots BH$. Hence,

$$P\{\mathcal{X} > 2n\} = pqpq \cdots pq + qpqp \cdots qp = 2(pq)^n.$$

Note that this formula does not hold for $n = 0$. (Why not? and what is the value of $P\{\mathcal{X} > 0\}$ anyway?)

- (b) For $n \geq 0$, find the probability that *more than* $2n + 1$ debates occur at the DNC.

Solution: By the same argument as in part (a) that the winners must alternate, we get that $P\{\mathcal{X} > 2n + 1\} = (pq)^n p + (qp)^n q = (pq)^n$ which does hold for $n = 0$.

(c) Find $E[\mathcal{X}]$. Hint: use the results of parts (a) and (b).

Solution: Since \mathcal{X} is a positive integer-valued random variable, we have that

$$\begin{aligned}
 E[\mathcal{X}] &= \sum_{k=0}^{\infty} P(\mathcal{X} > k) = \sum_{n=0}^{\infty} P\{\mathcal{X} > 2n\} + \sum_{n=0}^{\infty} P\{\mathcal{X} > 2n+1\} \\
 &= 1 + \sum_{n=1}^{\infty} P\{\mathcal{X} > 2n\} + 1 + \sum_{n=1}^{\infty} P\{\mathcal{X} > 2n+1\} \\
 &= 1 + \sum_{n=1}^{\infty} 2(pq)^n + 1 + \sum_{n=1}^{\infty} (pq)^n = -1 + 3 \left[1 + (pq) + (pq)^2 + \dots \right] \\
 &= \frac{3}{1-pq} - 1 = \frac{2+pq}{1-pq}.
 \end{aligned}$$

Notice that $E[\mathcal{X}] = 2$ if $p = 0$ or $q = 0$ which makes perfect sense. On the other hand, symmetry indicates that the maximum value of $E[\mathcal{X}]$ should occur when $p = q = \frac{1}{2}$, and surprisingly, this maximum value is only 3; on average, three or fewer debates can be expected to occur at the DNC!

Alternatively, we can steal the result from part (d) that $P\{\mathcal{X} = 2n+1\} = (pq)^n$ for $n \geq 1$, and use the same argument as in part (d) to deduce that for $n \geq 1$, $P\{\mathcal{X} = 2n\} = P\{\mathcal{X} > 2n-1\} - P\{\mathcal{X} > 2n\} = (pq)^{n-1} - 2(pq)^n = (pq)^{n-1}(1-2pq) = (p^2+q^2)(pq)^{n-1}$. Since $\mathcal{X} \geq 2$, we get

$$\begin{aligned}
 E[\mathcal{X}] &= \sum_{k=2}^{\infty} k \cdot P\{\mathcal{X} = k\} = (p^2+q^2) \sum_{n=1}^{\infty} 2n \cdot (pq)^{n-1} + \sum_{n=1}^{\infty} (2n+1)(pq)^n \\
 &= 2(p^2+q^2) \sum_{n=1}^{\infty} n(pq)^{n-1} + 2(pq) \sum_{n=1}^{\infty} n(pq)^{n-1} + \sum_{n=1}^{\infty} (pq)^n \\
 &= 2(p^2+q^2+pq) \frac{1}{(1-pq)^2} + \frac{1}{1-pq} - 1 \\
 &= \frac{2p^2+2q^2+2pq+1-pq-1+2pq-(pq)^2}{(1-pq)^2} \\
 &= \frac{2(p^2+q^2+2pq)-pq-(pq)^2}{(1-pq)^2} = \frac{2(p+q)^2-pq-(pq)^2}{(1-pq)^2} \\
 &= \frac{2-pq-(pq)^2}{(1-pq)^2} = \frac{(2+pq)(1-pq)}{(1-pq)^2} = \frac{2+pq}{1-pq} \text{ just as before.}
 \end{aligned}$$

(d) Find $P\{\mathcal{X} = 2n+1\}$ and $P\{\mathcal{X} = 2n+1 | \mathcal{X} > 2n\}$.

Solution: For integer-valued \mathcal{X} , $P(\mathcal{X} = k+1) = P\{\mathcal{X} > k\} - P\{\mathcal{X} > k+1\}$. So, $P\{\mathcal{X} = 2n+1\} = P\{\mathcal{X} > 2n\} - P\{\mathcal{X} > 2n+1\} = 2(pq)^n - (pq)^n = (pq)^n$. Note that since $P\{\mathcal{X} > 0\} = 1 \neq 2(pq)^0$, the result $P\{\mathcal{X} = 2n+1\} = (pq)^n$ does not hold for $n = 0$. For $n = 0$, $P\{\mathcal{X} = 1\} = P\{\mathcal{X} > 0\} - P\{\mathcal{X} > 1\} = 1 - 1 = 0$ as should be obvious if we just think about the problem a bit! From this we get

that $P\{\mathcal{X} = 1 | \mathcal{X} > 0\} = 0$, while for $n > 0$,

$$\begin{aligned} P\{\mathcal{X} = 2n + 1 | \mathcal{X} > 2n\} &= \frac{P(\{\mathcal{X} = 2n + 1\} \cap \{\mathcal{X} > 2n\})}{P\{\mathcal{X} > 2n\}} \\ &= \frac{P\{\mathcal{X} = 2n + 1\}}{P\{\mathcal{X} > 2n\}} = \frac{(pq)^n}{2(pq)^n} = \frac{1}{2}. \end{aligned}$$

Let \bar{H} denote the event that Hillary wins the Democratic nomination. Note that this is not the same as the event H that she wins a debate.

(e) Find $P\{\bar{H} | \mathcal{X} = 2n + 1\}$.

Solution: From the definition of conditional probability,

$$P\{\bar{H} | \mathcal{X} = 2n + 1\} = \frac{P(\bar{H} \cap \{\mathcal{X} = 2n + 1\})}{P\{\mathcal{X} = 2n + 1\}} = \frac{qpqp \cdots qp}{P(\mathcal{X} = 2n + 1)} = \frac{(qp)^n p}{(pq)^n} = p.$$

3. The joint pdf of random variables \mathcal{X} and \mathcal{Y} is given by

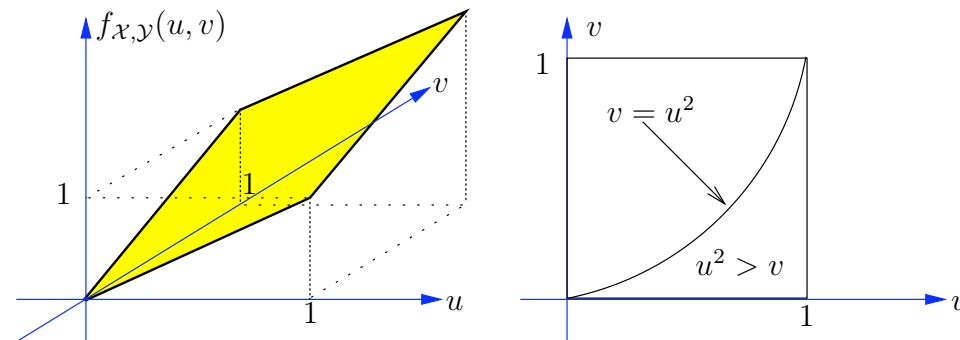
$$f_{\mathcal{X}, \mathcal{Y}}(u, v) = \begin{cases} u + v, & 0 < u < 1, 0 < v < 1, \\ 0, & \text{otherwise.} \end{cases}$$

(a) Find the marginal pdf $f_{\mathcal{X}}(u)$ of the random variable \mathcal{X} . Be sure to specify the value of $f_{\mathcal{X}}(u)$ for all real numbers u .

Solution: The pdf surface is a rhombus above the u - v plane as illustrated in the left-hand figure below.

$$f_{\mathcal{X}}(u) = \int_{-\infty}^{\infty} f_{\mathcal{X}, \mathcal{Y}}(u, v) dv = \int_0^1 u + v dv = uv + \frac{v^2}{2} \Big|_0^1 = u + \frac{1}{2}, \text{ for } 0 < u < 1,$$

and $f_{\mathcal{X}}(u) = 0$ otherwise. It is easily verified that this is a valid pdf.



(b) Find the probability that the solutions of the quadratic equation $\alpha^2 + 2\mathcal{X}\alpha + \mathcal{Y} = 0$ are real numbers.

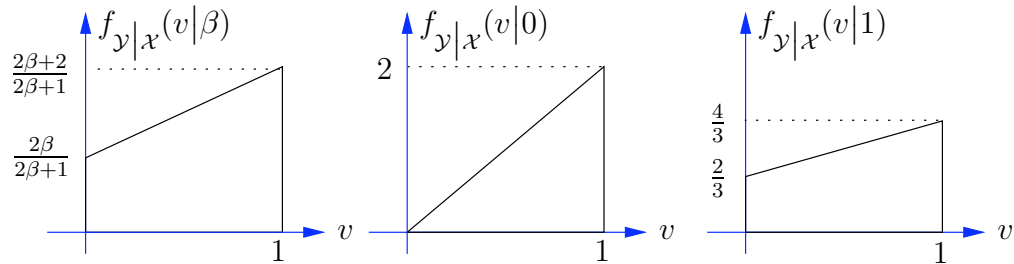
Solution: The solutions to the quadratic equation are $\frac{-2\mathcal{X} \pm \sqrt{4\mathcal{X}^2 - 4\mathcal{Y}}}{2}$ and are real numbers if $4\mathcal{X}^2 - 4\mathcal{Y} \geq 0 \Leftrightarrow \mathcal{X}^2 \geq \mathcal{Y}$, that is, if the random point $(\mathcal{X}, \mathcal{Y})$

lies below the parabola in the right-hand figure above. Thus, we have that

$$\begin{aligned} P\{\text{real solutions}\} &= P\{\mathcal{X}^2 \geq \mathcal{Y}\} = \int_{u=0}^1 \int_{v=0}^{u^2} (u+v) dv du \\ &= \int_{u=0}^1 \left[uv + \frac{v^2}{2} \right]_{v=0}^{u^2} du = \int_{u=0}^1 \left[u^3 + \frac{u^4}{2} \right] du \\ &= \left[\frac{u^4}{4} + \frac{u^5}{10} \right]_0^1 = \frac{1}{4} + \frac{1}{10} = \frac{7}{20}. \end{aligned}$$

- (c) Find the conditional pdf $f_{\mathcal{Y}|\mathcal{X}}(v|\beta)$ of \mathcal{Y} given that $\mathcal{X} = \beta$, where $0 < \beta < 1$. Be sure to specify the value of $f_{\mathcal{Y}|\mathcal{X}}(v|\beta)$ for all real numbers v .

Solution: $f_{\mathcal{Y}|\mathcal{X}}(v|\beta) = \frac{f_{\mathcal{X},\mathcal{Y}}(\beta, v)}{f_{\mathcal{X}}(\beta)} = \frac{\beta + v}{\beta + \frac{1}{2}} = \frac{2(\beta + v)}{2\beta + 1}$ for $0 < v < 1$, and $f_{\mathcal{Y}|\mathcal{X}}(v|\beta) = 0$ otherwise. The pdf is illustrated in the left-hand figure below. Using the formula for the area of a trapezoid, it is easily verified that the area under the curve equals 1.



- (d) Find the minimum-mean-square-error (MMSE) estimate of \mathcal{Y} given that $\mathcal{X} = \beta$ where $0 < \beta < 1$.

Solution: The MMSE estimator of \mathcal{Y} given that $\mathcal{X} = \beta$ is the mean of the conditional pdf $f_{\mathcal{Y}|\mathcal{X}}(v|\beta)$, viz.,

$$E[\mathcal{X}|\mathcal{Y} = \beta] = \int_0^1 \frac{2v(\beta + v)}{2\beta + 1} dv = \frac{1}{2\beta + 1} \left[\beta v^2 + \frac{2v^3}{3} \right]_0^1 = \frac{\beta + 2/3}{2\beta + 1} = \frac{3\beta + 2}{6\beta + 3}.$$

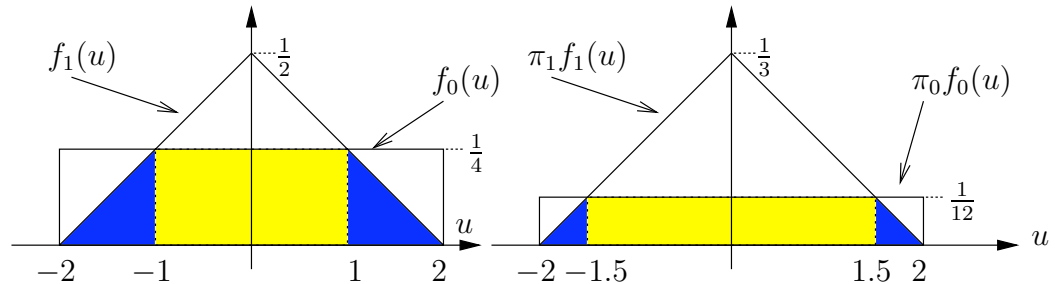
Note that the MMSE estimator is a *nonlinear* function of β , varying from value $\frac{2}{3}$ at $\beta = 0$ to $\frac{5}{9} < \frac{2}{3}$ at $\beta = 1$. Since both \mathcal{X} and \mathcal{Y} are more likely to be close to 1 than to 0, the estimated value of \mathcal{Y} is pretty large even when β , the observed value of \mathcal{X} , is close to 0. When β is close to 1, the estimated value of \mathcal{Y} is still large but closer to $\frac{1}{2}$ because the conditional pdf of \mathcal{Y} is much “more uniform” when β is large than when β is small (e.g., compare the middle and right-hand sketches in the figure on the previous page).

4. Consider the following binary hypothesis testing problem. If hypothesis H_0 is true, the continuous random variable $\mathcal{X} \sim \text{Uniform}(-2, 2)$, while if hypothesis H_1 is true, the pdf of \mathcal{X} is $f_1(u) = \begin{cases} \frac{1}{4}(2 - |u|), & |u| < 2, \\ 0, & \text{otherwise.} \end{cases}$

- (a) The *maximum-likelihood* decision rule can be stated in the form $|\mathcal{X}| \underset{\text{H}_{1-x}}{\overset{\text{H}_x}{\geq}} \eta$.

Specify whether x denotes 0 or 1, and find the values of η , the probability of false alarm P_{FA} , and the probability of missed detection P_{MD} .

Solution: The easiest way to solve this problem is to sketch the two pdfs as shown in the left-hand figure below.



It is obvious that the maximum-likelihood decision is in favor of H_1 if $|\mathcal{X}| < 1$, and hence $x = 0$, $\eta = 1$. By inspection, we get that $P_{\text{FA}} = 2 \times \frac{1}{4} = \frac{1}{2}$ while $P_{\text{MD}} = 2 \times \left(\frac{1}{2} \times 1 \times \frac{1}{4}\right) = \frac{1}{4}$.

The graphically-challenged can proceed as follows.

For $-2 < u < 2$, the likelihood ratio is $\Lambda(u) = \frac{f_1(u)}{f_0(u)} = \frac{0.25(2 - |u|)}{0.25} = 2 - |u|$.

When $\mathcal{X} = u$ is the observation, the *maximum-likelihood* decision rule decides in favor of H_1 if $\Lambda(u) > 1$. Hence $\Gamma_1 = \{u : |u| < 1\}$ and $\Gamma_0 = \{u : 1 < |u| < 2\}$, that is, the ML decision rule is that if $|\mathcal{X}| > 1$, the decision is that H_0 is the true hypothesis. Thus, we have $x = 0$, and $\eta = 1$.

$$P_{\text{FA}} = \int_{\Gamma_1} f_0(u) du = \int_{-1}^1 \frac{1}{4} du = \frac{1}{2}.$$

$$P_{\text{MD}} = \int_{\Gamma_0} f_1(u) du = 2 \int_1^2 \frac{1}{4}(2 - u) du = \frac{1}{2} \left(2u - \frac{u^2}{2}\right) \Big|_1^2 = \frac{1}{4}.$$

- (b) Suppose that the hypotheses have *a priori* probabilities $\pi_0 = 1/3$ and $\pi_1 = 2/3$. What is the error probability $P(E)$ of the maximum-likelihood decision rule?

Solution: The probability of error of the ML decision rule is

$$P(E) = \pi_0 P_{\text{FA}} + \pi_1 P_{\text{MD}} = \frac{1}{3} \times \frac{1}{2} + \frac{2}{3} \times \frac{1}{4} = \frac{1}{3}.$$

- (c) The MAP (also known as the minimum-error-probability or Bayesian) decision

rule can be stated in the form $|\mathcal{X}| \underset{\text{H}_{1-x}}{\overset{\text{H}_x}{\geq}} \xi$. Specify whether x denotes 0 or 1,

and find the values of ξ and the error probability $P(E)$.

Solution: Sketching $\pi_0 f_0(u)$ and $\pi_1 f_1(u)$ as in the right-hand figure above, we easily see that the MAP decision is in favor of H_1 if $|\mathcal{X}| < 1.5$, and hence $x = 0$, $\xi = 1.5$. By inspection, we get that $\pi_0 P_{\text{FA}} = 3 \times \frac{1}{12} = \frac{1}{4} = \frac{1}{3} \times \frac{3}{4}$ while $\pi_1 P_{\text{MD}} = 2 \times \left(\frac{1}{2} \times \frac{1}{2} \times \frac{1}{12}\right) = \frac{1}{24} = \frac{2}{3} \times \frac{1}{16}$, that is, $P_{\text{FA}} = \frac{3}{4}$, and $P_{\text{MD}} = \frac{1}{16}$.

$P(E) = \pi_0 P_{\text{FA}} + \pi_1 P_{\text{MD}} = \frac{1}{4} + \frac{1}{24} = \frac{7}{24} < \frac{1}{3}$, where $\frac{1}{3}$ is the error probability of the ML rule (with the same *a priori* probabilities) that we found in part (c).

Without using any graphical aids, we have that when $\mathcal{X} = u$ is the observation, the MAP decision rule decides in favor of H_1 if $\Lambda(u) = 2 - |u| > \pi_0/\pi_1 = 1/2$. Hence, $\Gamma_1 = \{u : |u| < \frac{3}{2}\}$ and $\Gamma_0 = \{u : \frac{3}{2} < |u| < 2\}$ for the MAP decision

rule. Once again, $x = 0$ while $\xi = \frac{3}{2}$. We get $P_{\text{FA}} = \int_{\Gamma_1} f_0(u) du = \int_{-\frac{3}{2}}^{\frac{3}{2}} \frac{1}{4} du = \frac{3}{4}$.

$$P_{\text{MD}} = \int_{\Gamma_0} f_1(u) du = 2 \int_{\frac{3}{2}}^2 \frac{1}{4}(2-u) du = \frac{1}{2} \left(2u - \frac{u^2}{2} \right) \Big|_{\frac{3}{2}}^2 = \frac{1}{16}.$$

$$\text{Hence, } P(E) = \pi_0 \cdot P_{\text{FA}} + \pi_1 \cdot P_{\text{MD}} = \frac{1}{3} \times \frac{3}{4} + \frac{2}{3} \times \frac{1}{16} = \frac{1}{4} + \frac{1}{24} = \frac{7}{24} < \frac{1}{3}.$$

- (d) For what range (if any) of values of π_0 , does the MAP decision rule always choose H_0 ?

Solution: The MAP decision rule compares the likelihood ratio $\Lambda(u) = 2 - |u|$ and decides in favor of H_0 if $\Lambda(u) < \pi_0/\pi_1$. Since $\Lambda(u)$ takes on values in $(0, 2]$, it cannot exceed π_0/π_1 if $\pi_0 > \frac{2}{3}$. Thus, the MAP decision rule always decides in favor of H_0 if $\pi_0 > 2/3$, and achieves average error probability $P(E) = \pi_1 < 1/3$.

- (e) For what range (if any) of values of π_0 , does the MAP decision rule always choose H_1 ?

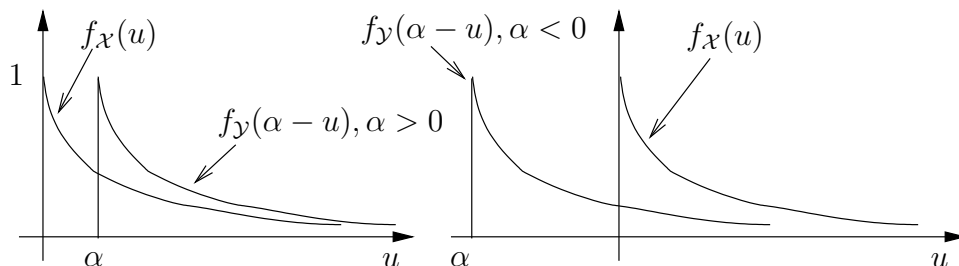
Solution: On the other hand, there is *no* value of π_0 , $0 < \pi_0 < 1$ for which the MAP rule *always* decide in favor of H_1 , because no matter how large π_1 is, the ratio $\pi_0/\pi_1 > 0$. Thus, for *some* values of \mathcal{X} , the likelihood ratio will be smaller than the threshold, and the decision will be in favor of H_0 .

5. \mathcal{X} and \mathcal{Y} are independent random variables with pdfs as specified below:

$$f_{\mathcal{X}}(u) = \begin{cases} \exp(-u), & u > 0, \\ 0, & \text{otherwise,} \end{cases} \quad \text{and} \quad f_{\mathcal{Y}}(v) = \begin{cases} \exp(v), & v < 0, \\ 0, & \text{otherwise.} \end{cases}$$

Find $f_{\mathcal{Z}}(\alpha)$, the pdf of the random variable $\mathcal{Z} = \mathcal{X} + \mathcal{Y}$. Be sure to specify the value of $f_{\mathcal{Z}}(\alpha)$ for all real numbers α .

Solution: Since \mathcal{X} and \mathcal{Y} are independent random variables, $f_{\mathcal{Z}} = f_{\mathcal{X}} \star f_{\mathcal{Y}}$, that is, $f_{\mathcal{Z}}(\alpha) = \int_{-\infty}^{\infty} f_{\mathcal{X}}(u) f_{\mathcal{Y}}(\alpha - u) du = \int_0^{\infty} \exp(-u) f_{\mathcal{Y}}(\alpha - u) du$ since $f_{\mathcal{X}}(u) = 0$ for $u < 0$. Now, $f_{\mathcal{Y}}(\alpha - u) = 0$ if $\alpha - u > 0$, that is, if $u < \alpha$. As illustrated in the figure below, we have to consider two cases.



- If $\alpha > 0$, then the integrand is 0 for $0 < u < \alpha$, and we get

$$f_Z(\alpha) = \int_{\alpha}^{\infty} \exp(-u) \exp(\alpha - u) du = \exp(\alpha) \frac{1}{2} \exp(-2u) \Big|_{\alpha}^{\infty} = \frac{1}{2} \exp(-\alpha).$$
- If $\alpha < 0$, then the integrand is nonzero for $0 < u < \infty$, and we get

$$f_Z(\alpha) = \int_0^{\infty} \exp(-u) \exp(\alpha - u) du = \exp(\alpha) \frac{1}{2} \exp(-2u) \Big|_0^{\infty} = \frac{1}{2} \exp(\alpha).$$

We can combine these two cases and write $f_Z(\alpha) = \frac{1}{2} \exp(-|\alpha|)$, $-\infty < \alpha < \infty$.

6. Suppose \mathcal{X} and \mathcal{Y} are jointly Gaussian random variables with means 0 and 2 respectively, variances 4 and 16 respectively, and correlation coefficient $\frac{1}{2}$. Let $\mathcal{W} = -\mathcal{X} + b\mathcal{Y}$ and $\mathcal{Z} = 7\mathcal{X} - \mathcal{Y}$ where b is a number whose value you will determine below.

- (a) For what value(s) of b does $E[\mathcal{W}]$ equal 0?

Solution: $E[\mathcal{W}] = -E[\mathcal{X}] + bE[\mathcal{Y}] = 0 + 2b = 0 \Rightarrow b = 0$.

- (b) [10 points] For what value(s) of b does $\text{var}(\mathcal{W})$ equal 3?

Solution: $3 = \text{var}(\mathcal{W}) = \text{var}(\mathcal{X}) + b^2 \text{var}(\mathcal{Y}) - 2b \text{cov}(\mathcal{X}, \mathcal{Y}) = 4 + 16b^2 - 2b \frac{1}{2} \cdot 4$ implies that $16b^2 - 8b + 1 = 0 \Rightarrow (4b - 1)^2 = 0 \Rightarrow b = \frac{1}{4}$.

- (c) For what value(s) of b are \mathcal{W} and \mathcal{Z} independent random variables?

Solution: Since \mathcal{X} and \mathcal{Y} are jointly Gaussian random variables, \mathcal{W} and \mathcal{Z} are also jointly Gaussian random variables, and hence independent if $\text{cov}(\mathcal{W}, \mathcal{Z}) = 0$. But,

$$\begin{aligned} \text{cov}(\mathcal{W}, \mathcal{Z}) &= \text{cov}(-\mathcal{X} + b\mathcal{Y}, 7\mathcal{X} - \mathcal{Y}) \\ &= -7 \cdot \text{cov}(\mathcal{X}, \mathcal{X}) + 7b \cdot \text{cov}(\mathcal{Y}, \mathcal{X}) + \text{cov}(\mathcal{X}, \mathcal{Y}) - b \cdot \text{cov}(\mathcal{Y}, \mathcal{Y}) \\ &= -7 \cdot 4 + 7b \cdot 4 + 4 - b \cdot 16 \\ &= 12b - 24 \end{aligned}$$

and we see that $\text{cov}(\mathcal{W}, \mathcal{Z}) = 0$ when $b = 2$.