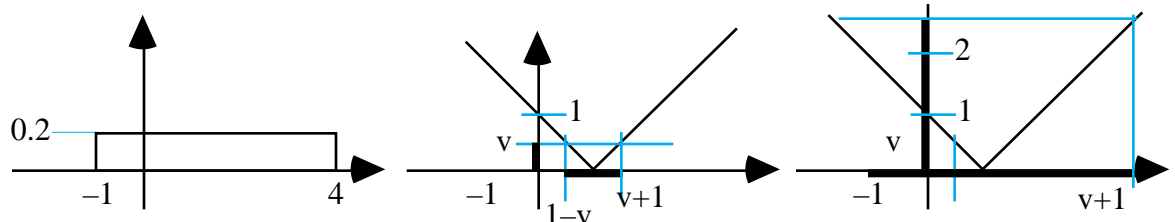


1. (a) $P\{X > b\} = 1 - F_X(b)$. If $F_X(a) < F_X(b)$, then $a < b$.
 If $a < b$, then $F_X(a) < F_X(b)$. $F_X(u) = 1/2$ for some $u, - < u < .$
 The first two properties hold for all CDFs since $F_X(u)$ is a *right-continuous non-decreasing* function. But, if $a < b$, it is possible that $F_X(a) = F_X(b)$ because the CDF did not increase between those points (it cannot decrease, of course.) As a counterexample for $\hat{}$, note that for a Bernoulli random variable with parameter $p = 1/2$, $F_X(u)$ takes on values 0, p , and 1 only. Thus, $\hat{}$ and $\hat{}$ are not properties of all CDFs.

- (b) $f_X(u) = 1$ for all $u, - < u < .$ $\lim_{u \rightarrow -} f_X(u) = 0$
 $\lim_{u \rightarrow +} f_X(u) = 1$ $P\{a < X < b\} = P\{a < X < b\}$
 The second and fourth properties are satisfied by all pdfs. The first need not hold: the value of the function is a probability *density*, not a probability mass, and can exceed 1 for some pdfs. On the other hand, **no** pdf can possibly satisfy the third property.

2. $P\{|X - 4| > 3\} = P\{X > 7\} + P\{X < 1\} = 1 - \left(\frac{7-2}{5}\right) + \left(\frac{1-2}{5}\right) = 1 - (1) + (-0.2)$
 $= 1 - (1) + 1 - (0.2) = 2 - 0.8413 - 0.5793 = 0.5794$
 $P\{X < 3 | X > 2\} = P\{2 < X < 3\} / P\{X > 2\} = 2 \left[\left(\frac{3-2}{5}\right) - \left(\frac{2-2}{5}\right) \right] = 2(0.2 - 0.5)$
 $= 1.1586 - 1 = 0.1586$

- 3.(a) The pdf of X has value 0.2 for $-1 \leq u \leq 4$, is as shown below in the left-hand figure below. We have that
 $E[Y] = E[|X-1|] = \int_{-1}^1 0.2(1-u) du + \int_1^4 0.2(u-1) du = 0.2(u - u^2/2) \Big|_{-1}^1 + 0.2(u^2/2 - u) \Big|_1^4 = 1.3$.

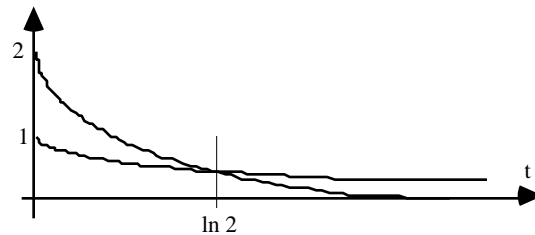


- (b) Y takes on values in the range $[0, 3]$. From the middle and right-hand figures above, we see that for any $v, 0 \leq v \leq 2$, $F_Y(v) = P\{Y \leq v\} = P\{1-v \leq X \leq v+1\} = 0.2(v+1 - (1-v)) = 0.4v$, while for any $v, 2 \leq v \leq 3$, $F_Y(v) = P\{Y \leq v+1\} = P\{X \leq v\} = 0.2(v+1 - (-1)) = 0.2(v+2)$.
 Hence, $f_Y(v) = \begin{cases} 0.4, & 0 \leq v \leq 2, \\ 0.2, & 2 \leq v \leq 3 \\ 0, & \text{elsewhere} \end{cases}$ which is easily verified to be a valid pdf.

- 4.(a) $N(3, 6]$ is a Poisson random variable with parameter 3μ . Hence, $P\{N(3, 6] = 2\} = \exp(-3\mu) \cdot (3\mu)^2 / 2!$.
 (b) If there are 2 arrivals in $(0, 6]$, and also 2 arrivals in $(3, 9]$, then it must be that in the *disjoint* intervals $(0, 3]$, $(3, 6]$, and $(6, 9]$, there are $(2, 0, 2)$ arrivals respectively, or $(1, 1, 1)$ arrivals respectively, or $(0, 2, 0)$ arrivals respectively, which correspond to values of $N(3, 6]$ of 0, 1, and 2 respectively.

Hence, by the independence of arrivals in disjoint intervals, we get that
 $P(AB) = P\{N(3,6] = 0\} \cdot P\{\text{two arrivals in } (0, 3] \text{ and two arrivals in } (6, 9]\}$
 $+ P\{N(3,6] = 1\} \cdot P\{\text{one arrival in } (0, 3] \text{ and one arrival in } (6, 9]\}$
 $+ P\{N(3,6] = 2\} \cdot P\{\text{no arrivals in } (0, 3] \text{ and no arrivals in } (6, 9]\}$
 $= \exp(-3\mu) \cdot [\exp(-3\mu) \cdot (3\mu)^2 / 2!]^2 + [\exp(-3\mu) \cdot (3\mu)] \cdot [\exp(-3\mu) \cdot (3\mu)]^2 + [\exp(-3\mu) \cdot (3\mu)^2 / 2!] \cdot [\exp(-3\mu)]^2$
 $= \exp(-9\mu) \cdot [81\mu^4 / 4 + 27\mu^3 + 9\mu^2 / 2]$. Hence, we have
 $P\{N(3, 6] = 1 | AB\} = P\{\{N(3, 6] = 1\} \cap AB\} / P(AB) = \frac{27\mu^3}{81\mu^4/4 + 27\mu^3 + 9\mu^2/2} = \frac{12\mu}{9\mu^2 + 12\mu + 2}$

- 5.(a) X is an exponential random variable with parameter $\theta_0 = 1$ or $\theta_1 = 2$ according as hypothesis H_0 or H_1 is true. Thus, the pdfs $f_0(u)$ and $f_1(u)$ are as shown below where the crossing-point is easily found by equating $2 \cdot \exp(-2t) = \exp(-t)$ and solving to get $t = \ln 2$. Thus, we see that the maximum-likelihood decision rule chooses H_1 if $T < \ln 2$, and H_0 if $T > \ln 2$. Hence, $i = 0, \tau = \ln 2$ in the decision rule exhibited on the exam..



- (b) $P_{FA} = P\{X < \ln 2 \mid H_0 \text{ is true}\} = 1 - P\{X > \ln 2 \mid H_0 \text{ is true}\} = 1 - \exp(-\theta_0 \ln 2) = 1 - \exp(-\ln 2) = 1/2$.
 $P_{MD} = P\{X > \ln 2 \mid H_1 \text{ is true}\} = \exp(-\theta_1 \ln 2) = \exp(-2 \ln 2) = \exp(-4)$.
- (c) The MEP rule compares $\theta_1 f_1(T) / \theta_0 f_0(T) = (0.4) \cdot 2 \cdot \exp(-2T) / [(0.6) \cdot \exp(-T)] = (4/3) \cdot \exp(-T)$ to the threshold 1 and chooses H_1 if this ratio exceeds 1. Thus, the MEP decision rule chooses H_1 if $\exp(T) < 4/3$, i.e. if $T < \ln(4/3)$. Note that $\ln(4/3) < \ln 2$ and the MEP rule is thus playing the odds and choosing the more likely hypothesis H_0 whenever $\ln(4/3) < T < \ln 2$, in which cases the maximum-likelihood rule would be choosing H_1 . We have that $i = 0$ as before, and $\tau = \ln(4/3)$.
- (d) More generally, the MEP rule chooses H_1 if $\exp(T) < 2 \theta_1 / \theta_0 = 2(1 - \theta_0) / \theta_0$ and H_0 if $\exp(T) > 2 \theta_1 / \theta_0$. If $\theta_0 > 2/3$, then $2(1 - \theta_0) / \theta_0 < 1$, and hence $\exp(T) > 2 \theta_1 / \theta_0$ for all $T > 0$. It follows that if $\theta_0 > 2/3$, then the MEP rule always chooses H_0 . On the other hand, $2 \theta_1 / \theta_0 = 2(1 - \theta_0) / \theta_0$ is finite for all $\theta_0 > 0$, and hence there is always a T for which $\exp(T) > 2 \theta_1 / \theta_0$. Thus, for any $\theta_0 > 0$, the MEP rule *will* choose H_0 on those trials on which the outcome exceeds $\ln(2 \theta_1 / \theta_0)$, and will not always choose H_1 . The MEP rule will always choose H_1 only when $\theta_0 = 0$, in which case, it should be intuitively obvious that it makes no sense for an MEP rule to make any choice other than H_1 !!