Massachusetts Institute of Technology

Department of Electrical Engineering & Computer Science

6.041/6.431: Probabilistic Systems Analysis (Fall 2011)

Recitation 10 Solutions October 13, 2011

1. (a) We know that the PDF must integrate to 1. Therefore we have

$$\int_{-\infty}^{\infty} f_Z(z) dz = \int_{-2}^{1} \gamma (1+z^2) = \gamma \left(z + \frac{1}{3}z^3\right) \Big|_{-2}^{1} = 6\gamma.$$

From this we conclude $\gamma = 1/6$.

(b) To find the CDF, we integrate:

$$F_{Z}(z) = \int_{-\infty}^{z} f_{Z}(t) dt = \begin{cases} 0, & \text{if } z < -2, \\ \frac{1}{6} \left(t + \frac{1}{3} t^{3} \right) \Big|_{-2}^{z}, & \text{if } -2 \le z \le 1, \\ 1, & \text{if } z > 1 \end{cases}$$

$$= \begin{cases} 0, & \text{if } z < -2, \\ \frac{1}{6} (z + \frac{1}{3} z^{3} + \frac{14}{3}), & \text{if } -2 \le z \le 1, \\ 1, & \text{if } z > 1. \end{cases}$$

- 2. See textbook, Problem 3.9, page 187.
- 3. (a) For $x \ge 0$,

$$F_X(x) = \int_{-\infty}^x f_X(t) dt = \int_0^x \lambda e^{-\lambda t} dt = \left[-e^{-\lambda t} \right]_0^x = 1 - e^{-\lambda x}.$$

For x < 0, we have $F_X(x) = \int_{-\infty}^x f_X(t) dt = 0$. Thus we conclude

$$F_X(x) = \begin{cases} 0, & \text{if } x < 0, \\ 1 - e^{-\lambda x}, & \text{if } x \ge 0. \end{cases}$$

(b) The key step in the following computation uses integration by parts, whereby

$$\int_0^\infty u \, dv = uv \Big|_0^\infty - \int_0^\infty v \, du$$

is applied with u = x and $v = -e^{-\lambda x}$:

$$\mathbf{E}[X] = \int_{-\infty}^{\infty} x f_X(x) \, dx = \int_0^{\infty} x \lambda e^{-\lambda x} \, dx = \left[-x e^{-\lambda x} \right]_0^{\infty} + \int_0^{\infty} e^{-\lambda x} \, dx = \frac{1}{\lambda}.$$

(c) Integrating by parts with $u = x^2$ and $v = -e^{-\lambda x}$ in the second line below gives

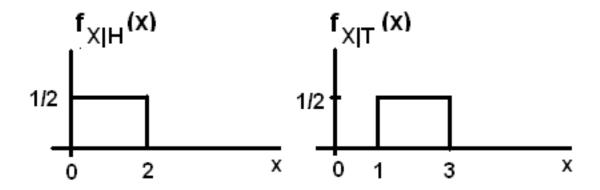
$$\mathbf{E}[X^2] = \int_{-\infty}^{\infty} x^2 f_X(x) dx = \int_{0}^{\infty} x^2 \lambda e^{-\lambda x} dx$$
$$= \left[-x^2 e^{-\lambda x} \right]_{0}^{\infty} + 2 \int_{0}^{\infty} x e^{-\lambda x} dx = \frac{2}{\lambda} \mathbf{E}[X] = \frac{2}{\lambda^2}.$$

Combining with the previous computation, we obtain

$$var(X) = \mathbf{E}[X^2] - (\mathbf{E}[X])^2 = \frac{2}{\lambda^2} - \left(\frac{1}{\lambda}\right)^2 = \frac{1}{\lambda^2}.$$

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4. Let H be the event that the flip results in heads and T be the event that the flip results in tails. Then, $\mathbf{P}(H) = \frac{3}{4}$ and $\mathbf{P}(T) = \frac{1}{4}$. We are told $f_{X|H}(x)$ and $f_{X|T}(x)$ looks like:



From the total probability theorem, we have

$$f_X(x) = \mathbf{P}(H)f_{X|H}(x) + \mathbf{P}(T)f_{X|T}(x),$$

from which we conclude $f_X(x)$ looks like:

