

Problem 5.1 (2-6.2)

(a) $Y = g(X) = X^3$, $g'(x) = 3x^2$.

Solving x in terms of y , we get only one root:

$$x_1 = y^{1/3}, \quad -\infty < y < \infty.$$

From the theorem,

$$f_Y(y) = \frac{f_X(x_1)}{|g'(x_1)|} = \frac{\frac{1}{\sqrt{2\pi}} e^{-\frac{(y^{1/3})^2}{2}}}{|3(y^{1/3})^2|} = \frac{1}{3\sqrt{2\pi}} y^{-2/3} e^{-y^{2/3}/2}$$

(b) $E[Y] = \int_{-\infty}^{\infty} y f_Y(y) dy = \frac{1}{3\sqrt{2\pi}} \int_{-\infty}^{\infty} y^{1/3} e^{-\frac{y^{2/3}}{2}} dy$

Note that $y^{1/3} e^{-\frac{y^{2/3}}{2}}$ is an odd function \Rightarrow its integral from $-\infty$ to ∞ is zero.

$$\Rightarrow \boxed{E[Y] = 0}$$

(c) $\text{Var}[Y] = E[Y^2] - E^2[Y] = E[Y^2] = E[X^6]$

$$\begin{aligned} \text{From (2-27) in text, we have } E[(X-\bar{X})^6] &= E[X^6] = 1.3.5 \sigma^6 \\ &= 15 \cdot 1^6 \\ &= \boxed{15} \end{aligned}$$

Problem 5.2 (2-6.6) let Y_i , $i=1..5$ denote the independent Gaussian outcomes.

(a) $E[X^2] = E\left[\sum_{i=1}^5 Y_i^2\right] = \sum_{i=1}^5 E[Y_i^2] = \sum_{i=1}^5 (\sigma_i^2 + \mu_i^2) = 5 \cdot \sigma^2 = \boxed{5}$ ($= n\sigma^2$)

(b) $\text{Var}[X^2] = 2n\sigma^4 = 2 \cdot 5 \cdot 1^2 = \boxed{10}$

(c) The answer depends on the interpretation of "most probable value":

(Cont'd)

i) Value of y where $f_{X^2}(y)$ attains maximum (mode of the distribution)

ii) Expected value of X^2 , which is $\int_{-\infty}^{\infty} y f_{X^2}(y) dy$ (mean of the distribution)

iii) Value of y where $\Pr\{\tilde{X}^2 < y\} = \Pr\{\tilde{X}^2 > y\}$, which is the solution of the eqn:

$$\int_{-\infty}^y f_{X^2}(z) dz = \int_y^{\infty} f_{X^2}(z) dz \quad (\text{median of the distribution})$$

For interpretation (i), the answer is calculated as follows:

$$\frac{\partial}{\partial y} f_{X^2}(y) = \frac{\partial}{\partial y} \left[\frac{y^{3/2}}{2^{5/2} \Gamma(5/2)} e^{-y/2} \right] = \frac{3/2 y^{1/2}}{2^{5/2} \Gamma(5/2)} e^{-y/2} - \frac{1}{2} \frac{y^{3/2}}{2^{5/2} \Gamma(5/2)} e^{-y/2} = 0$$
$$\Rightarrow \frac{3}{2} = \frac{y}{2} \Rightarrow \boxed{y = 3}$$

For interpretation (ii), the answer is

$$E(X^2) = \boxed{5}$$

Interpretation (iii) does not yield a closed-form solution. It is solved numerically using the following relation for y :

$$F_X(y) = 1 - e^{-y/2} \sum_{k=0}^4 \frac{1}{k!} \left(\frac{y}{2}\right)^k = \frac{1}{2} \Rightarrow \boxed{e^{-y/2} \left(1 + \frac{y}{2} + \frac{y^2}{8} + \frac{y^3}{48} + \frac{y^4}{384}\right) = \frac{1}{2}}$$

All three of the answers are correct. They all find usage in different applications where the optimization criterion is stated in different ways. For example, it can be shown that:

i) The mode maximizes the probability that the r.v. X^2 is within an interval of infinitesimal width:

$$\text{mode}(X^2) = \max_y \Pr\{y - \epsilon < \tilde{X}^2 < y + \epsilon\}, \quad \epsilon \text{ infinitesimally small.}$$

ii) The mean minimizes the average squared distance:

$$\text{mean}(X^2) = \min_y E[|y - \tilde{X}^2|^2]$$

iii) The median minimizes the average absolute distance:

$$\text{median}(X^2) = \min_y E[|y - \tilde{X}^2|]$$

Problem 5.3 (2-7.3) $f_t(t) = \frac{1}{6} e^{-t/6}, t \geq 0$

(a) $\int_0^6 f_t(t) dt = \frac{1}{6} (-6 e^{-t/6})_0^6 = 1 - e^{-1} = \boxed{0,632}$

(b) $\int_{10}^{\infty} f_t(t) dt = \frac{1}{6} (-6 e^{-t/6})_{10}^{\infty} = e^{-10/6} = \boxed{0,189}$

(c) $\int_5^6 f_t(t) dt = \frac{1}{6} (-6 e^{-t/6})_5^6 = e^{-1} - e^{-6/5} = \boxed{0,0667}$

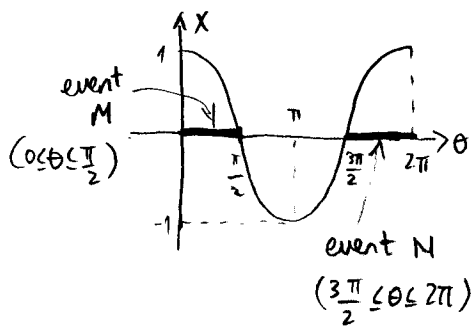
Problem 5.4 (2-8.2)

(a) From the continuous version of Bayes' Theorem:

$$f(x|M) = \frac{\Pr\{M | \tilde{x}=x\} f_x(x)}{\Pr\{M\}}$$

$$\Rightarrow f(x|0 \leq \theta \leq \frac{\pi}{2}) = \frac{\Pr\{0 \leq \theta \leq \frac{\pi}{2} | \tilde{x}=x\}}{\Pr\{0 \leq \theta \leq \frac{\pi}{2}\}} \cdot \frac{1}{\pi \sqrt{1-x^2}} = \frac{\Pr\{0 \leq \theta \leq \frac{\pi}{2} | \tilde{x}=x\}}{\frac{\pi}{4} \sqrt{1-x^2}}$$

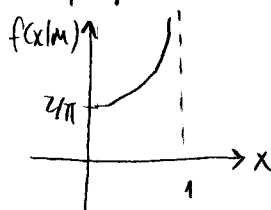
Let's examine the term $\Pr\{0 \leq \theta \leq \frac{\pi}{2} | \tilde{x}=x\}$:



It is obvious that the above probability is zero when $-1 \leq x \leq 0$, since no θ between $0 \leq \theta \leq \frac{\pi}{2}$ yields a negative value.

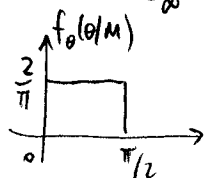
When $0 \leq x \leq 1$, the probability that θ is between 0 and $\frac{\pi}{2}$ is equal to $1/2$, since event M and event N are equally probable because of the uniformity of θ .

$$\Rightarrow f(x|M) = \begin{cases} \frac{2}{\pi} \frac{1}{\sqrt{1-x^2}}, & 0 \leq x \leq 1 \\ 0, & \text{else} \end{cases}$$



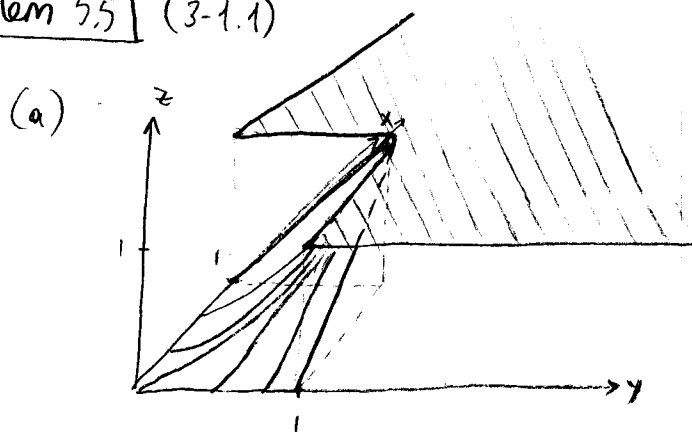
(b) $E[X|M] = E[\cos \theta | M] = \int_{-\infty}^{\infty} \cos \theta f_{\theta}(\theta|M) d\theta$

Since θ is uniform,

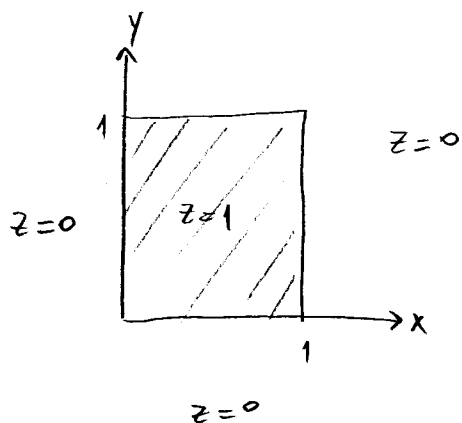


$$\Rightarrow E[X|M] = \int_0^{\pi/2} \cos \theta \cdot \frac{2}{\pi} d\theta = \boxed{\frac{2}{\pi}}$$

Problem 5.5 (3-1.1)



$$(b) f(x,y) = \frac{\partial^2 F(x,y)}{\partial x \partial y} = \begin{cases} 0, & x < 0, y < 0 \\ 1, & 0 \leq x \leq 1, 0 \leq y \leq 1 \\ 0, & x > 1, y > 0 \end{cases}$$



$$(c) \Pr\left\{x \leq \frac{3}{4}, y > \frac{1}{4}\right\} = \int_{x=0}^{3/4} \int_{y=1/4}^1 dy dx = \left(\frac{3}{4}\right) \left(\frac{3}{4}\right) = \boxed{\frac{9}{16}}$$