Advanced Statistical Methods for Modeling and Finance: Solutions for exercises

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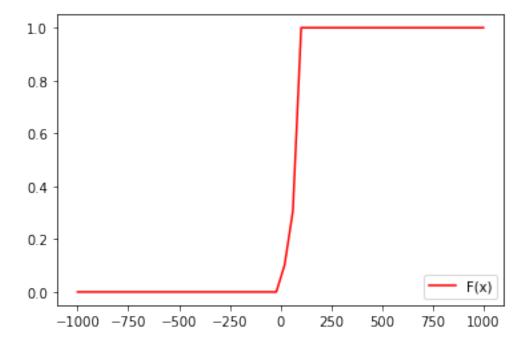
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Let us suppose a random variable X has it's CDF defined by:

$$F(x) = \begin{cases} 0 & \text{if } x < 0\\ \frac{x}{200} & \text{if } 0 < x < 100\\ 1 & \text{if } x \ge 100 \end{cases}$$

5.1 The graph of F

```
[2]: import numpy as np
import matplotlib.pyplot as plt
F=lambda x:np.where(0<x and x<100,x/200,np.where(x<0 ,0,1))
x=np.linspace(-1000,1000).tolist()
plt.plot(x,[F(i) for i in x],label='F(x)',color='r')
plt.legend(loc='lower right')
plt.show()</pre>
```



5.2 Let's compute the following probabilities

a)
$$P(-50 < X < 50)$$

$$P(-50 < X < 50) = F(50) - F(-50)$$

$$50 \in [0; 100] \Rightarrow F(50) = \frac{50}{200}$$

= $\frac{1}{4}$

$$-50 \in]-\infty; 0[\Rightarrow F(-50) = 0$$

Then we have

$$P(-50 < X < 50) = \frac{1}{4} - 0$$
$$= \frac{1}{4}$$

b)
$$P(X = 0)$$

$$P(X = 0) = F(0) - \lim_{x \to 0^{-}} F(x)$$
$$= \frac{0}{200} - \lim_{x \to 0^{-}} 0$$
$$= 0$$

c)
$$P(X = 100)$$

$$P(X = 100) = F(100) - \lim_{x \to 100^{-}} F(x)$$

$$= 1 - \lim_{x \to 100^{-}} \frac{x}{200}$$

$$= 1 - \frac{100}{200}$$

$$= 1 - \frac{1}{2}$$

$$= \frac{1}{2}$$

6 Exercise 6

Let us suppose a random variable X has it's PDF defined by:

$$f(x) = \begin{cases} kx^2 & \text{if } 0 \le x \le 10\\ 0 & \text{if } x > 10. \end{cases}$$

Let us find P(7 < X < 15)

Let us note F the CDF of X

$$P(7 < X < 15) = F(15) - F(7)$$

 $\forall x \geq 0$, we have:

$$F(x) = \int_{-\infty}^{x} f(t)dt$$
$$= \int_{-\infty}^{0} f(t)dt + \int_{0}^{x} f(t)dtdt$$
$$= 0 + \int_{0}^{x} f(t)dt$$
$$= \int_{0}^{x} f(t)dt$$

 $t \in [0; x] \iff 0 \le t \le x$

• If $x \le 10$, then $0 \le t \le x \le 10$ ie $0 \le t \le 10$ and then $f(t) = 2kt^2$ So

$$F(x) = \int_0^x 2kt^2 dt$$
$$= \frac{2}{3}k \left[t^3\right]_0^x$$
$$= \frac{2}{3}kx^3$$

• If $x \ge 10$, then we have:

$$F(x) = \int_0^{10} f(t)dt + \int_{10}^x f(t)dt$$

$$= \int_0^{10} f(t)dt + 0$$

$$= \int_0^{10} 2kt^2dt$$

$$= \frac{2}{3}k \left[t^3\right]_0^{10}$$

$$= \frac{2}{3}10^3k$$

$$= \frac{2000k}{3}$$

$$F(x) = \frac{2}{3}k \begin{cases} x^3 & \text{if } 0 \le x \le 10\\ 1000 & \text{if } x > 10. \end{cases}$$

Let us determine k.

$$\int_{\mathbb{R}} f(t)dt = \int_{-\infty}^{0} f(t)dt + \int_{0}^{10} f(t)dt + \int_{10}^{+\infty} f(t)dt$$

$$= 0 + \int_{0}^{10} 2kt^{2}dtdt + 0$$

$$= 2k \left[\frac{1}{3}t^{3}\right]_{0}^{10}$$

$$= \frac{2k}{3}[t^{3}]_{0}^{10}$$

$$= \frac{2k}{3}[t^{3}]_{0}^{10}$$

$$= \frac{2k}{3}(10^{3} - 0^{3})$$

$$= \frac{2k}{3}(1000)$$

$$= \frac{2000k}{3}$$

f is a PDF, then $\int_{\mathbb{R}} f(t)dt = 1$

$$\int_{\mathbb{R}} f(t)dt = 1 \Rightarrow \frac{2000k}{3} = 1$$
$$\Rightarrow 2000k = 3$$
$$\Rightarrow k = \frac{3}{2000}$$

And then

$$F(x) = \frac{2}{3} \times \frac{3}{2000} \begin{cases} x^3 & \text{if } 0 \le x \le 10\\ 1000 & \text{if } x > 10. \end{cases}$$
$$= \frac{1}{1000} \begin{cases} x^3 & \text{if } 0 \le x \le 10\\ 1000 & \text{if } x > 10. \end{cases}$$
$$= \begin{cases} \frac{1}{1000} x^3 & \text{if } 0 \le x \le 10\\ 1 & \text{if } x > 10. \end{cases}$$

15 > 10 then F(15) = 1

 $7 \in [0; 10]$ then $F(7) = \frac{7^3}{1000} = \frac{343}{1000}$

$$P(7 < X < 15) = F(15) - F(7)$$

$$' = 1 - \frac{343}{1000}$$

$$= \frac{1000 - 343}{1000}$$

$$= 3 \times \frac{16}{2000}$$

$$= \frac{357}{1000}$$

7 Exercise 7

Let

- X be a random variable with f as PDF, such that $f(t) = \frac{1}{\sqrt{2\pi}} \exp(-\frac{1}{2}t^2); t \in \mathbb{R}$
- $Y = X^2$ another random variable.
- F_X the CDF of X
- F_Y the CDF of Y
- f_Y the PDF of Y

Let us show that $f_Y(y) = \frac{1}{\sqrt{2\pi y}} \exp(-\frac{y}{2}) \forall t \in \mathbb{R} \ \forall t \in \mathbb{R}$, we have:

$$F_Y(t) = P(Y < t)$$

$$= P(X^2 < t)$$

$$= P(Y < \sqrt{t})$$

$$= P(|x| < \sqrt{t}) \text{ since } X^2 < t \Rightarrow t > 0$$

$$= P(-\sqrt{t} < \sqrt{t} < \sqrt{t})$$

$$= F_X(\sqrt{t}) - F_X(-\sqrt{t})$$

$$f_{Y}(t) = F'_{Y}(t)$$

$$= \frac{1}{2\sqrt{t}}F'_{X}(\sqrt{t}) - (-\frac{1}{2\sqrt{t}})F'_{X}(\sqrt{t})$$

$$= \frac{1}{2\sqrt{t}}f(\sqrt{t}) - (-\frac{1}{2\sqrt{t}})f(-\sqrt{t})$$

$$= \frac{1}{2\sqrt{t}}f(\sqrt{t}) + \frac{1}{2\sqrt{t}}f(-\sqrt{t})$$

$$= \frac{1}{2\sqrt{t}}[f(\sqrt{t}) + f(-\sqrt{t})]$$

$$= \frac{1}{2\sqrt{t}}(2)(\frac{1}{\sqrt{2\pi}}exp(-\frac{1}{2}t))$$

$$= \frac{1}{\sqrt{t}}\frac{1}{\sqrt{2\pi u}}exp(-\frac{1}{2}t)$$

$$= \frac{1}{\sqrt{2\pi u}}\exp(-\frac{t}{2})$$

8 Exercise 8

Let X be a random variable such that Var(X) exists.

Let's show that $Var(X) = E(X^2) - \mu^2$ where $\mu = E(X)$

We have by definition:

$$Var(X) = E([X - E(X)]^{2})$$

$$= E(X^{2} - 2XE(X) + (E(X))^{2})$$

$$= E(X^{2} - 2\mu X + \mu^{2})$$

$$= E(X^{2} - 2\mu X + \mu^{2}) \text{ with } \mu = E(X)$$

$$= E(X^{2}) - 2\mu E(X) + \mu^{2}$$

$$= E(X^{2}) - 2\mu \mu + \mu^{2} \text{ because } E(X) = \mu$$

$$= E(X^{2}) - 2\mu^{2} + \mu^{2}$$

$$= E(X^{2}) - \mu^{2}$$

Hence the result.

9 Exercise 9

Let X a discrete random variable with pmf f positive at -1, 0,1 and zero elsewhere.

9.1 a) Let us find E(X) if $p(0) = \frac{1}{4}$

$$E(X^{2}) = (-1)^{2}p(-1) + (0)^{2}p(0) + (1)^{2}p(1) + 0$$

= $p(-1) + p(1)$

We have

$$p(-1) + p(0) + p(1) + 0 = p(-1) + p(0) + p(1)$$
$$= p(-1) + \frac{1}{4} + p(1) \text{ since } p(0) = \frac{1}{4}$$

Then

$$p(-1) + p(1) = 1 - p(0)$$

$$= 1 - \frac{1}{4}$$

$$= \frac{3}{4}$$

And then $E(X^2) = \frac{3}{4}$

9.2 a) Let us find p(-1) and p(1) if $p(0) = \frac{1}{4}$ and if E(X) = 1

$$E(X) = (-1)p(-1) + (0)p(0) + (1)p(1) + 0$$

= $p(1) - p(-1)$

$$\begin{split} E(X) &= 1 \Rightarrow p(1) - p(-1) = 1 \\ &= \Rightarrow p(1) = 1 + p(-1) \text{ (1)} \end{split}$$

$$p(-1) + p(0) + p(1) = 1$$
 and $p(0) = \frac{1}{4}$

Then we get:

$$p(-1) + p(1) = 1 - \frac{1}{4}$$
$$= \frac{3}{4}$$

And then $p(1) = \frac{3}{4} - P(-1)$ ②

Hence

$$p(-1) = \frac{3}{4} - p(-1)$$

$$= \frac{3}{4} + \frac{1}{8}$$

$$= \frac{24 + 4}{4 \times 8}$$

$$= \frac{28}{32}$$

$$= \frac{7}{8}$$

10 Exercise 10

Let X a continous random variable with pdf f, such that $f(x) = \frac{1}{3}$ for -1 < x < 2

Let M_X the moment generating function of X

Let
$$t \in]-1;2[$$

$$M_X(t) = E(e^t X)$$

$$= \int_{\mathbb{R}} e^{tx} f(x) dx$$

$$= \int_{-\infty}^{-1} e^{tx} f(x) dx + \int_{-1}^{2} e^{tx} f(x) dx + \int_{2}^{+\infty} e^{tx} f(x) dx$$

$$= 0 + \int_{-1}^{2} \frac{1}{3} e^{tx} dx + 0$$

$$= \frac{1}{3} \int_{-1}^{2} e^{tx} dx$$

When $t \neq 0$, we have:

$$M_X(t) = \frac{1}{3} \left[\frac{1}{t} e^{tx} \right]_{-1}^2$$
$$= \frac{1}{3t} \left[e^{tx} \right]_{-1}^2$$
$$= \frac{1}{3t} (e^{2t} - e^{-t})$$

11 Exercise 11

Let $X \sim \mathcal{B}(n, p)$ and M_X the MGF of X

Let us show that $M_X(t) = (1 - p + pe^t)^n \ \forall t \in \mathbb{R}$

Let q = 1 - p

We have $\mathbb{P}(X=k) = \binom{n}{k} p^k q^{n-k} \forall k \in \{0,1,\ldots,n\}$

Let $t \in \mathbb{R}$

$$\begin{split} M_X(t) &= E(e^{tX}) \\ &= \Sigma_{i=1}^n e^{tk} P(Y=k) \\ &= \Sigma_{i=1}^n e^{tk} \binom{n}{k} p^k q^{n-k} \\ &= \Sigma_{i=1}^n p^k (e^t)^k \binom{n}{k} q^{n-k} \\ &= \Sigma_{i=1}^n (pe^t)^k \binom{n}{k} q^{n-k} \\ &= \Sigma_{i=1}^n a^k \binom{n}{k} b^{n-k} \text{ with } a = pe^t \text{ and } b = q \\ &= (a+b)^n \text{ from Newten's binomial formula} \\ &= (pe^t + 1 - p)^n \text{ since } a = pe^t \text{ and } b = q \\ &= (1-p+pe^t)^n \end{split}$$

Hence the result.

12 Exercise 12

Let X a random variable that has a Poisson ditribution with λ as parameter

We have:

$$P(X = k) = \frac{\lambda^k e^{-\lambda}}{k!}$$
 for k in N

Let M_X the moment generating function of X and $t \in \mathbb{R}$

12.1 Let us show that $M_X(t) = e^{\lambda(e^t-1)}$

$$\begin{split} M_X(t) &= E(e^t X) \\ &= \Sigma_{k \in \mathbb{N}} e^{tk} \mathbb{P}(X = k) \\ &= \Sigma_{k \in \mathbb{N}} e^{tk} \frac{\lambda^k e^{-\lambda}}{k!} \\ &= \Sigma_{k \in \mathbb{N}} \frac{e^{tk} \lambda^k e^{-\lambda}}{k!} \\ &= \Sigma_{k \in \mathbb{N}} \frac{(e^t)^k \lambda^k e^{-\lambda}}{k!} \\ &= \Sigma_{k \in \mathbb{N}} \frac{(\lambda e^t)^k e^{-\lambda}}{k!} \end{split}$$

$$\forall x \in \mathbb{R}, e^x = \sum_{n \in \mathbb{N}} \frac{x^n}{n!}$$

Then we have:

$$M_X(t) = e^{-\lambda} e^{\lambda e^t}$$

$$= e^{-\lambda + \lambda e^t}$$

$$= e^{\lambda e^t - \lambda}$$

$$= e^{\lambda (e^t - 1)}$$

Hence $M_X(t) = e^{\lambda(e^t - 1)}$

12.2 Let us find it's mean and variance

Let $t \in \mathbb{R}$

• Mean E(X)

$$E(X) = M_X'(0)$$

$$M_X(t) = e^{\lambda(e^t - 1)}$$

$$M'_X(t) = \lambda e^t e^{\lambda(e^t - 1)}$$
$$= \lambda e^t M_X(t)$$

$$M_X(0) = e^{e^0 - 1} = e^0 = 1$$

$$M'_X(0) = \lambda e^0 M_X(0)$$
$$= \lambda(1)$$
$$= \lambda$$

Then
$$E(X) = \lambda$$

• Variance Var(X)

$$var(X) = E(X^2) - (E(X))^2$$
 with $E(X^2) = M_X''(0)$

$$M_X''(t) = (M_X')'(t)$$

$$M_X'(t) = \lambda e^t M_X(t)$$

$$M_X''(t) = \lambda [e^t M_X(t) + e^t M_X'(t)]$$
$$= \lambda e^t [M_X(t) + M_X'(t)]$$

$$M_X''(0) = \lambda e^0 [M_X(0) + M_X'(0)]$$
$$= \lambda(1)[1 + \lambda]$$
$$= \lambda(1 + \lambda)$$

Then $E(X^2) = \lambda(1+\lambda)$ and then

$$Var(X) = \lambda(1+\lambda) - \lambda^{2}$$
$$= \lambda + \lambda^{2} - \lambda^{2}]$$
$$= \lambda$$