

CIND-221: Taller 4

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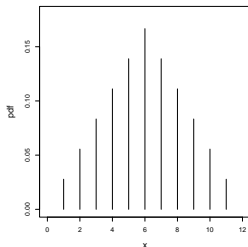
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Variable aleatoria

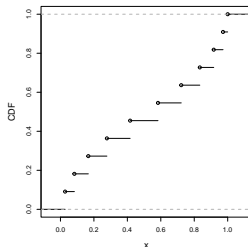
Ejemplo (lanzamiento de dos dados):

Considere el lanzamiento de 2 dados. Sea X la suma de sus caras. Entonces,

x	2	3	4	5	6	7	8	9	10	11	12
$P(X = x)$	$\frac{1}{36}$	$\frac{2}{36}$	$\frac{3}{36}$	$\frac{4}{36}$	$\frac{5}{36}$	$\frac{6}{36}$	$\frac{5}{36}$	$\frac{4}{36}$	$\frac{3}{36}$	$\frac{2}{36}$	$\frac{1}{36}$



(a) $p(x)$



(b) $F(x)$

Ejemplo:

Suponga

$$f(x) = \frac{1}{2} \exp(-x/2), \quad x > 0,$$

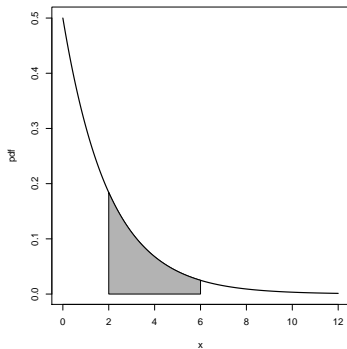
con $\lambda = 1/2$.

Tenemos que

$$F(x) = 1 - \exp(-x/2), \quad x > 0,$$

donde $F(x) = P(X \leq x)$.

Considera $P(2 < X < 6)$, es decir:



De este modo,

$$\begin{aligned}P(2 < X < 6) &= \frac{1}{2} \int_2^6 \exp(-x/2) \, dx = F(6) - F(2) \\&= [1 - \exp(-3)] - [1 - \exp(-1)] = 0.9502 - 0.6321 \\&= 0.3181\end{aligned}$$

Adicionalmente,

$$P(X < 8) = F(8) = 1 - \exp(-4) = 0.9817,$$

y

$$P(X \geq 8) = 1 - F(8) = 1 - 0.9817 = 0.0183.$$

Variable aleatoria

```
1 # evaluando la densidad exponencial
2 > x <- seq(0, 13, length = 300)
3 > length(x)
4 [1] 300
5 > x[1:10]
6 [1] 0.00000000 0.04347826 0.08695652 0.13043478 0.17391304
7 [6] 0.21739130 0.26086957 0.30434783 0.34782609 0.39130435
8
9 > y <- dexp(x, rate = .5)
10 > plot(x, y, type = "l", ylab = "pdf", lwd = 2)
11
12 # calculando probabilidades acumuladas
13 > pexp(6, rate = .5) # F(6)
14 [1] 0.9502129
15 > pexp(2, rate = .5) # F(2)
16 [1] 0.6321206
17 > pexp(6, rate = .5) - pexp(2, rate = .5) # F(6) - F(2)
18 [1] 0.3180924
19 > pexp(8, rate = .5) # F(8)
20 [1] 0.9816844
21 > pexp(8, rate = .5, lower = FALSE) # P(X > 8)
22 [1] 0.01831564
23 > 1 - pexp(8, rate = .5)
24 [1] 0.01831564
25
```

Cadenas de Markov

```
1 matrix.power <- function(a, pow = 2)
2 { ## computes the power of a square matrix
3   if (is.data.frame(a))
4     a <- as.matrix(a)
5   if (!is.matrix(a))
6     stop("supply a matrix-like 'a'")
7   if (!is.numeric(a))
8     stop("argument a is not a numeric matrix")
9
10  da <- dim(a)
11  n <- da[1]
12  p <- da[2]
13  if (n != p)
14    stop("argument a is not a square matrix")
15
16  if (pow < 0)
17    stop("only implemented for positive power")
18  z <- a
19  k <- pow
20  if (k == 0)
21    return (diag(n))
22  if (k == 1)
23    return (z)
24  if (k > 1)
25    return(z %*% matrix.power(z, pow = k - 1))
26 }
27
```

Cadenas de Markov

```
1 # leyendo fuentes R
2 source("matrix.power.R")
3
4 # creando una matriz
5 > p <- matrix(c(.7, .3, .4, .6), ncol = 2, byrow = TRUE)
6
7 # imprimiendo en pantalla
8 > p
9           [,1] [,2]
10 [1,]    0.7  0.3
11 [2,]    0.4  0.6
12
13 # sumando por filas (MARGIN = 1) o columnas (MARGIN = 2)
14 > apply(p, 1, sum)
15 [1] 1 1
16 > apply(p, 2, sum)
17 [1] 1.1 0.9
18
```


Cadenas de Markov

```
1 > matrix.power(p) # argumento 'pow' por defecto
2      [,1] [,2]
3 [1,] 0.61 0.39
4 [2,] 0.52 0.48
5
6 > matrix.power(p, pow = 0)
7      [,1] [,2]
8 [1,]    1    0
9 [2,]    0    1
10
11 > matrix.power(p, pow = 4)
12      [,1] [,2]
13 [1,] 0.5749 0.4251
14 [2,] 0.5668 0.4332
15
16 > matrix.power(p, pow = 12)
17      [,1] [,2]
18 [1,] 0.5714288 0.4285712
19 [2,] 0.5714283 0.4285717
20
21 > matrix.power(p, pow = 20)
22      [,1] [,2]
23 [1,] 0.5714286 0.4285714
24 [2,] 0.5714286 0.4285714
25
```

Cadenas de Markov

```
1 # ejemplo Slide 15
2 > p4 <- matrix.power(p, 4)
3 > p4
4           [,1]      [,2]
5 [1,] 0.5749 0.4251
6 [2,] 0.5668 0.4332
7
8 > init <- c(.4, .6)
9 > init # distribución inicial
10 [1] 0.4 0.6
11
12 > pred <- init %*% p4
13 > pred # distribución de X4
14           [,1]      [,2]
15 [1,] 0.57004 0.42996
16
17 # chequeo del tipo de datos
18 > is.matrix(pred)
19 [1] TRUE
20 > is.matrix(init)
21 [1] FALSE
22
```

Cadenas de Markov

```
1 # ejemplo Slide 3
2 > ex <- matrix(c(.75,.25,0,.25,.5,.75,0,.25,.25), ncol = 3)
3
4 > ex
5      [,1] [,2] [,3]
6 [1,] 0.75 0.25 0.00
7 [2,] 0.25 0.50 0.25
8 [3,] 0.00 0.75 0.25
9
10 > matrix.power(ex, pow = 10)
11      [,1]      [,2]      [,3]
12 [1,] 0.4320240 0.4265490 0.1414270
13 [2,] 0.4265490 0.4297562 0.1436949
14 [3,] 0.4242811 0.4310846 0.1446342
15
16 > matrix.power(ex, pow = 20)
17      [,1]      [,2]      [,3]
18 [1,] 0.4285936 0.4285585 0.1428480
19 [2,] 0.4285585 0.4285790 0.1428625
20 [3,] 0.4285439 0.4285875 0.1428685
21
22 > matrix.power(ex, pow = 40)
23      [,1]      [,2]      [,3]
24 [1,] 0.4285714 0.4285714 0.1428571
25 [2,] 0.4285714 0.4285714 0.1428571
26 [3,] 0.4285714 0.4285714 0.1428571
27
```

Cadenas de Markov

```
1 > p <- matrix(c(.2,.1,.1,.6,.8,.6,.2,.1,.3), ncol = 3)
2 > p
3      [,1] [,2] [,3]
4 [1,]  0.2  0.6  0.2
5 [2,]  0.1  0.8  0.1
6 [3,]  0.1  0.6  0.3
7
8 # decomposición espectral de 'p'
9 > rs <- eigen(p)
10 > rs
11 eigen() decomposition
12 $values
13 [1] 1.0 0.2 0.1
14
15 $vectors
16      [,1]      [,2]      [,3]
17 [1,] -0.5773503  0.6882472 -0.9847319
18 [2,] -0.5773503 -0.2294157  0.1230915
19 [3,] -0.5773503  0.6882472  0.1230915
20
```

```
1 # vector cuyos elementos son todos 1
2 > ones <- rep(1, 3)
3 > ones
4 [1] 1 1 1
5
6 # norma del vector
7 > sqrt(sum(ones^2)) # sqrt(3)
8 [1] 1.732051
9
10 # normalizando
11 > ones / sqrt(sum(ones^2))
12 [1] 0.5773503 0.5773503 0.5773503
13
14 # vectores propios
15 > rs$vectors
16           [,1]      [,2]      [,3]
17 [1,] -0.5773503  0.6882472 -0.9847319
18 [2,] -0.5773503 -0.2294157  0.1230915
19 [3,] -0.5773503  0.6882472  0.1230915
20
```

Cadenas de Markov

```
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3 > ones
4 [1] 1 1 1
5
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7 > sqrt(sum(ones^2)) # sqrt(3)
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10 # normalizando
11 > ones / sqrt(sum(ones^2))
12 [1] 0.5773503 0.5773503 0.5773503
13
14 # vectores propios
15 > rs$vectors
16           [,1]      [,2]      [,3]
17 [1,] -0.5773503  0.6882472 -0.9847319
18 [2,] -0.5773503 -0.2294157  0.1230915
19 [3,] -0.5773503  0.6882472  0.1230915
20
```

Cadenas de Markov

```
1 > z <- t(p) # matriz transpuesta
2 > eigen(z)
3 eigen() decomposition
4 $values
5 [1] 1.0 0.2 0.1
6
7 $vectors
8           [,1]           [,2]           [,3]
9 [1,] 0.1441500 -1.345794e-16 7.071068e-01
10 [2,] 0.9730125 -7.071068e-01 -1.296942e-16
11 [3,] 0.1801875 7.071068e-01 -7.071068e-01
12
13 > eigen(p)
14 eigen() decomposition
15 $values
16 [1] 1.0 0.2 0.1
17
18 $vectors
19           [,1]           [,2]           [,3]
20 [1,] -0.5773503 0.6882472 -0.9847319
21 [2,] -0.5773503 -0.2294157 0.1230915
22 [3,] -0.5773503 0.6882472 0.1230915
23
```

Cadenas de Markov

```
1 stationary <- function(x)
2 { ## finds the stationary distribution of a Markov
3   ## chain with transition matrix 'x'
4   z <- t(x)
5   x <- eigen(z)$vectors[,1] # 1st column
6   x <- as.vector(x / sum(x))
7   x
8 }
9
```

```
1 # obteniendo la distribución estacionaria
2 > u <- stationary(p)
3 > u
4 [1] 0.1111111 0.7500000 0.1388889
5
6 > u > 0
7 [1] TRUE TRUE TRUE
8 > sum(u)
9 [1] 1
10
```


Usando fastmatrix¹

```
1 # otra alternativa, usando 'fastmatrix'
2 # disponible en: https://github.com/faosorios/fastmatrix
3 > z <- power.method(t(p))
4 > z
5 $value
6 [1] 1
7
8 $vector
9 [1] 0.1441500 0.9730125 0.1801875
10
11 attr(,"iterations")
12 [1] 13
13
14 # distribución debe sumar 1
15 > st <- z$vector / sum(z$vector)
16 > st
17 [1] 0.1111111 0.7500000 0.1388889
18
```

¹Disponible en CRAN: <https://cran.r-project.org/package=fastmatrix>