

TP2

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EX5.1

on pose , $Y = |X|$ Donc , $dx = -dy$ quand $x < 0$ et $dx = dy$ quand $x > 0$

$$\begin{aligned}\mathbb{E}_X[h(x)] &= \int_{-\infty}^{+\infty} h(x)f_X(x)dx \\&= \int_{-\infty}^0 h(x)f_X(x)dx + \int_0^{\infty} h(x)f_X(x)dx \\&= \int_0^{\infty} h(-x)f_X(-x)d(-x) + \int_0^{\infty} h(x)f_X(x)dx \\&= \int_0^{\infty} h(y)f_X(y)d(y) + \int_0^{\infty} h(y)f_X(y)dy \\&= \int_0^{\infty} h(y)2f_X(y)dy = E_Y[h(y)]\end{aligned}$$

Donc

$$f_Y(y) = 2f_X(y)\mathbb{I}_{y>0}$$

EX5.2

EX5.3

```
rexpoentielle <- function(N,lambda)
{
  X <- rep(0,N)
  for(i in 1:N)
  {
    u <- runif(1)
    X[i] <- -log(u)/lambda
  }
  return(X)
}

r_abs_normal <- function(N)
{
  pi <- 3.1415926
  X <- rep(0,N)
  i <- 1
  for(i in 1:N)
  {
    u <- runif(1)
    Y <- rexpoentielle(1,1)
  }
}
```

```

c <- sqrt(2*exp(1)/pi)
fY <- 2/sqrt(2*pi)*exp((-1/2)*Y^2)
gY <- exp(-Y)
hY <- fY/(c*gY)
while(u>hY)
{
  u <- runif(1)
  Y <- rexp(1,1)
  c <- sqrt(2*exp(1)/pi)
  fY <- 2/sqrt(2*pi)*exp((-1/2)*Y^2)
  gY <- exp(-Y)
  hY <- fY/(c*gY)
}
X[i]=Y
}
return(X)
}

```

```
r_abs_normal(10)
```

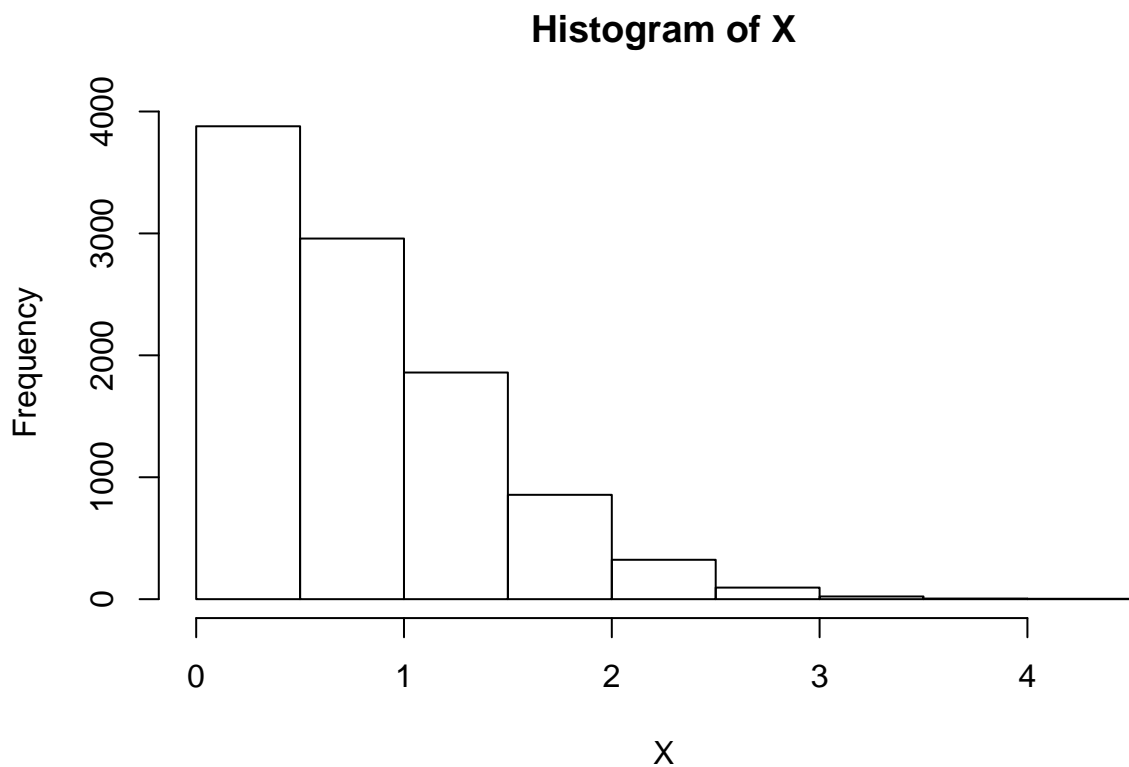
```
## [1] 0.80529503 1.17371905 0.03637641 0.14986678 1.39813868 1.06588321
## [7] 0.07328164 0.25873669 2.14929340 0.39294355
```

EX5.4

```

X <- r_abs_normal(10000)
hist(X)

```

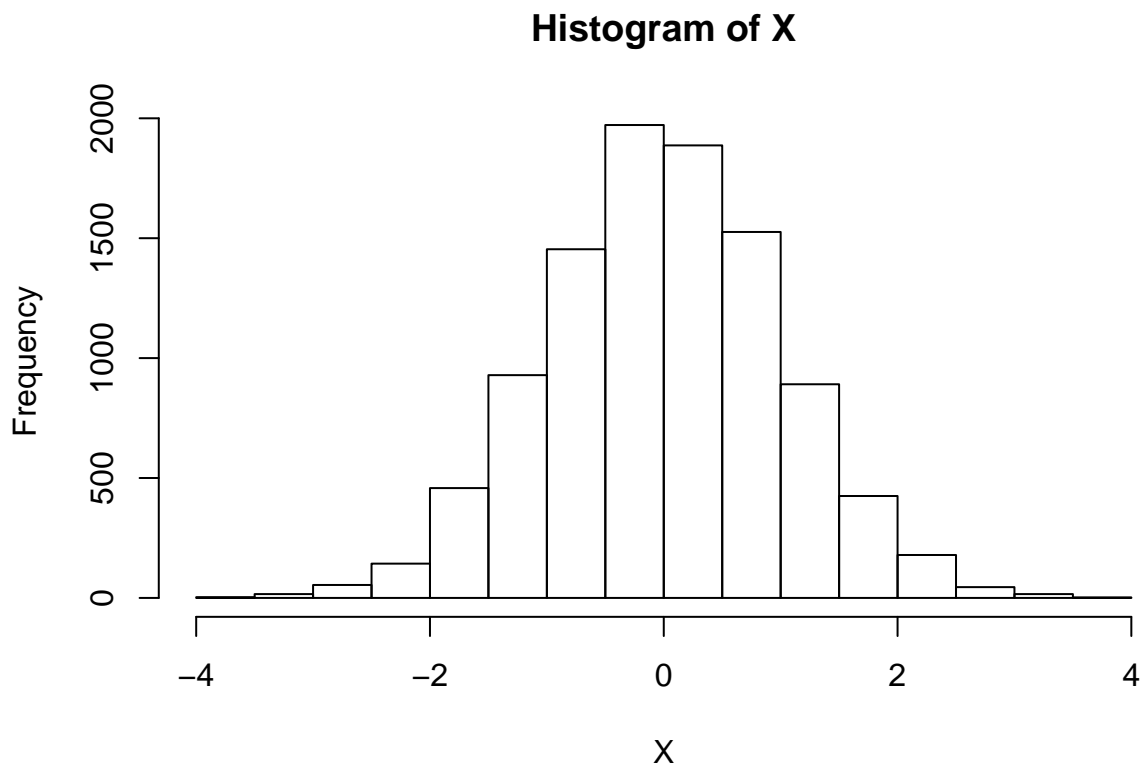


###EX5.5

EX5.6

```
r_normal <- function(N)
{
  X <- rep(0,N)
  for(i in 1:N)
  {
    u <- runif(1)
    if(u<1/2)
    {
      X[i] <- r_abs_normal(1)
    }
    else{
      X[i] <- -r_abs_normal(1)
    }
  }
  return (X)
}

X <- r_normal(10000)
hist(X)
```



EX6.1

```
r_unif_2dim <- function(N)
{
  U <- array(dim = c(N,2))
  for(i in 1:N)
```

```

{
  u1 <- 2*runif(1)-1
  u2 <- 2*runif(1)-1
  while(u1^2 + u2^2 > 1)
  {
    u1 <- 2*runif(1)-1
    u2 <- 2*runif(1)-1
  }
  U[i,1] <- u1
  U[i,2] <- u2
}
return(U)
}

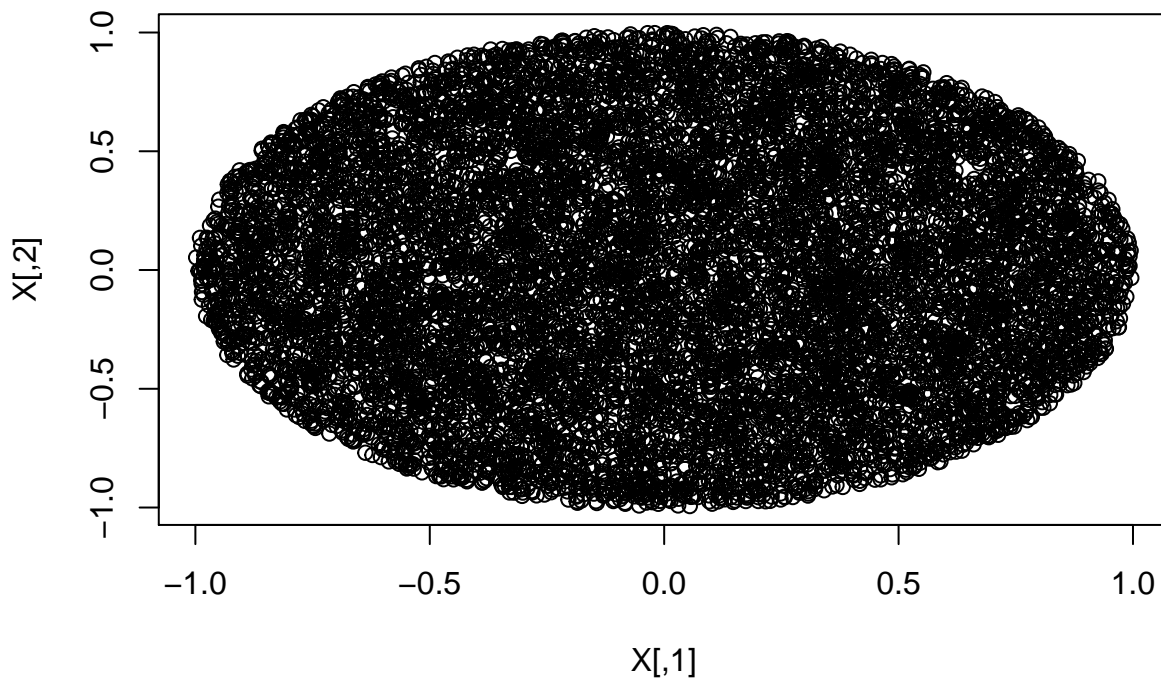
```

EX6.2 (Méthode de Box-Muller)

```

X <- r_unif_2dim(10000)
plot(X)

```



EX7.1

On a $X_1 = \sqrt{R}\cos(\Theta)$ et $X_2 = \sqrt{R}\sin(\Theta)$

Par consequence , on peut déduire que $R = X_1^2 + X_2^2$ et $\Theta = \arctan(X_2/X_1)$

On sais que $f_{(X_1, X_2)}(X) = f_{(R, \Theta)}(R, \Theta) * \det(J(X)) = f_R(R) * f_{\Theta}(\Theta) * \det(J(X))$

Donc, on dois juste calculer J(X)

$$J(X) = \begin{pmatrix} \frac{\partial R}{\partial X_1} & \frac{\partial R}{\partial X_2} \\ \frac{\partial \Theta}{\partial X_1} & \frac{\partial \Theta}{\partial X_2} \end{pmatrix}$$

$$= \begin{pmatrix} 2X_1 & 2X_2 \\ \frac{-X_2}{X_1^2+X_2^2} & \frac{X_1}{X_1^2+X_2^2} \end{pmatrix}$$

Alors , $\det(J(X)) = \frac{2(X_1^2+X_2^2)}{X_1^2+X_2^2} = 2$

Finalement , on a

$$f_{(X_1, X_2)}(x) = f_R(r) * f_{\Theta}(\theta) * \det(J(X)) = \frac{1}{2\pi} \exp(-(x_1^2 + x_2^2)/2) = \frac{1}{\sqrt{2\pi}} \exp(-x_1^2/2) * \frac{1}{\sqrt{2\pi}} \exp(-x_2^2/2)$$

ils sont independentes car $f_{(X_1, X_2)}(x) = f_{X_1}(x_1)f_{X_2}(x_2)$