Working with the ought-to-know continuous rv distributions in R

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Normal

The Normal (μ, σ^2) pdf is

$$f_X(x;\mu,\sigma^2) = \frac{1}{\sqrt{2\pi}} \frac{1}{\sigma} \exp\left[-\frac{(x-\mu)^2}{2\sigma^2}\right], \quad -\infty < x < \infty.$$

In R, we can type ?dnorm into the console to pull up the R documentation about the Normal distribution functions. We see that we can compute the height of the density curve $f_X(x;\mu,\sigma^2)$ as dnorm(x = x, mean = mu, sd = sigma). Note that R takes the parameter σ , which is the standard deviation, instead of σ^2 .

Normal cdf

We can compute

$$F_X(x;\mu,\sigma^2) = \int_{-\infty}^x \frac{1}{\sqrt{2\pi}} \frac{1}{\sigma} \exp\left[-\frac{(t-\mu)^2}{2\sigma^2}\right] dt, \quad -\infty < x < \infty$$

in R as pnorm(q = x , mean = mu, sd = sigma).

Normal quantiles

We can compute

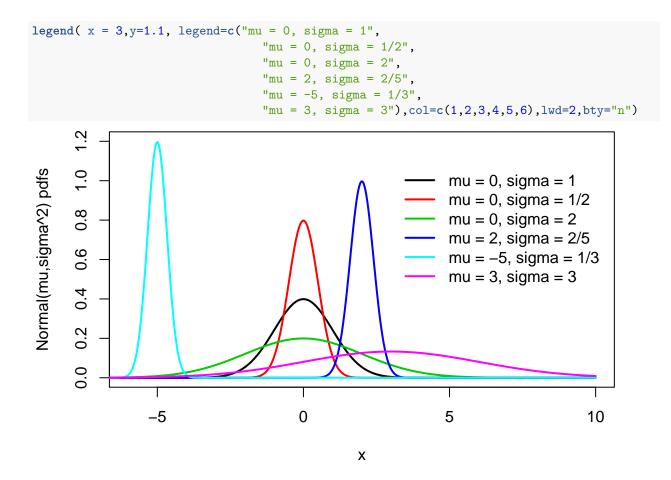
$$Q_X(u;\mu,\sigma^2) = \inf\{x: F_X(x;\mu,\sigma^2) \ge u\}, \quad 0 < u < 1$$

in R as qnorm(p = u, mean = mu, sd = sigma).

Plots of different Normal pdfs

Let's plot the pdfs of some different Normal distributions:

```
# create a sequence of x values at which to compute the Normal pdfs
x.seq <- seq(-10,10, length=500)
# create an empty plot to which we can add several lines
plot(NA,xlim=c(-6,10),ylim=c(0,1.2),ylab="Normal(mu,sigma^2) pdfs",xlab="x")
# plot Normal pdf against sequence of x values for different choices of mu and sigma
lines(dnorm(x.seq,mean = 0, sd = 1)~x.seq,col=1,lwd=2) # mu = 1, sigma = 1
lines(dnorm(x.seq,mean = 0, sd = 1/2)~x.seq,col=2,lwd=2) # mu = 1, sigma = 1/2
lines(dnorm(x.seq,mean = 0, sd = 2)~x.seq,col=3,lwd=2) # mu = 0, sigma = 2
lines(dnorm(x.seq,mean = 2, sd = 2/5)~x.seq,col=4,lwd=2) # mu = 2, sigma = 2/5
lines(dnorm(x.seq,mean = -5, sd = 1/3)~x.seq,col=5,lwd=2) # mu = -5, sigma = 1/3
lines(dnorm(x.seq,mean = 3, sd = 3)~x.seq,col=6,lwd=2) # mu = 3, sigma = 3
# add a legend to the plot
```



Gamma

The gamma(α, β) pdf is

$$f_X(x;\alpha,\beta) = \begin{cases} \frac{1}{\Gamma(\alpha)\beta^{\alpha}} x^{\alpha-1} \exp\left[-\frac{x}{\beta}\right] & x > 0\\ 0 & x \le 0. \end{cases}$$

In R, we can type ?dgamma into the console to pull up the R documentation about the gamma distribution functions. We see that we can compute the height of the density curve $f_X(x; \alpha, \beta)$ as dgamma(x = x, shape = alpha, scale = beta).

Gamma cdf

We can compute

$$F_X(x;\alpha,\beta) = \begin{cases} \int_0^x \frac{1}{\Gamma(\alpha)\beta^{\alpha}} t^{\alpha-1} \exp\left[-\frac{t}{\beta}\right] dt & 0 < x < \infty \\ 0 & -\infty < x \le 0 \end{cases}$$

in R as pgamma(q = x , shape = alpha, scale = beta).

Gamma quantiles

We can compute

 $Q_X(u;\alpha,\beta) = \inf\{x : F_X(x;\alpha,\beta) \ge u\}, \quad 0 < u < 1$

in R as qgamma(p = u, shape = alpha, scale = beta).

Plots of different Gamma pdfs

Let's plot the pdfs of some different gamma distributions:

```
# create a sequence of x values at which to compute the gamma pdfs
x.seq <- seq(0,20, length=500)</pre>
# create an empty plot to which we can add several lines
plot(NA,xlim=c(0,20),ylim=c(0,.5),ylab="gamma(alpha,beta) pdfs",xlab="x")
# plot gamma pdf against sequence of x values for different choices of alpha and beta
lines(dgamma(x.seq,shape = 1, scale = 1/2)~x.seq,col=1,lwd=2) # alpha = 1, beta = 1/2
lines(dgamma(x.seq,shape = 1, scale = 2)~x.seq,col=2,lwd=2) # alpha = 1, beta = 2
lines(dgamma(x.seq,shape = 1, scale = 4)~x.seq,col=3,lwd=2) # alpha = 1, beta = 4
lines(dgamma(x.seq,shape = 2, scale = 2)~x.seq,col=4,lwd=2) # alpha = 2, beta = 2
lines(dgamma(x.seq,shape = 5, scale = 2)~x.seq,col=5,lwd=2) # alpha = 5, beta = 2
lines(dgamma(x.seq,shape = 10, scale = 1/2)~x.seq,col=6,lwd=2) # alpha = 10, beta = 1/2
# add a legend to the plot
legend( x = 10, y=.45, legend=c("alpha = 1, beta = 1/2",
                                "alpha = 1, beta = 2",
                                "alpha = 1, beta = 4",
                                "alpha = 2, beta = 2",
                                "alpha = 5, beta = 2",
                                "alpha = 10, beta = 1/2"),col=c(1,2,3,4,5,6),lwd=2,bty="n")
      0.5
gamma(alpha,beta) pdfs
      0.4
                                                         alpha = 1, beta = 1/2
                                                         alpha = 1, beta = 2
                                                         alpha = 1, beta = 4
      0.3
                                                         alpha = 2, beta = 2
                                                         alpha = 5, beta = 2
      0.2
                                                         alpha = 10, beta = 1/2
      0.1
      0.0
             0
                               5
                                                10
                                                                 15
                                                                                  20
```

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Exponential

The Exponential(λ) pdf is

$$f_X(x;\lambda) = \begin{cases} \frac{1}{\lambda} \exp\left[-\frac{x}{\lambda}\right] & x > 0\\ 0 & x \le 0. \end{cases}$$

In R, we can type ?dexp into the console to pull up the R documentation about the Normal distribution functions. We see that we can compute the height of the density curve $f_X(x;\lambda)$ as dexp(x = x, rate = 1/lambda). Note that R parameterizes the exponential distribution differently, such that we must put in $1/\lambda$ instead of λ .

Exponential cdf

We can compute

$$F_X(x;\lambda) = \begin{cases} 1 - \exp[-\frac{x}{\lambda}] & 0 < x < \infty\\ 0 & -\infty < x \le 0 \end{cases}$$

in R as pexp(q = x , rate = 1/lambda).

Exponential quantiles

We can compute

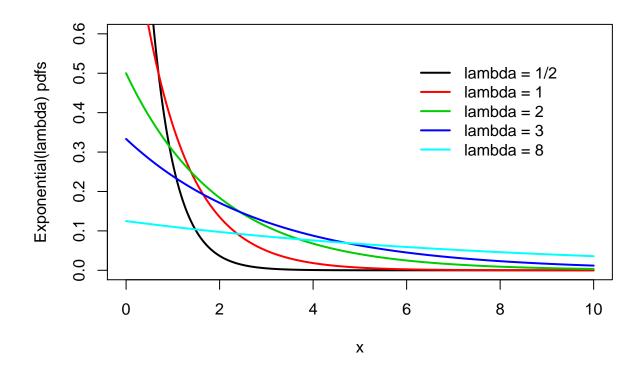
$$Q_X(u;\lambda) = \inf\{x : F_X(x;\lambda) \ge u\}, \quad 0 < u < 1$$

in R as qexp(p = u, rate = 1/lambda).

Plots of different Exponential pdfs

Let's plot the pdfs of some different exponential distributions:

```
# create a sequence of x values at which to compute the exponential pdfs
x.seq <- seq(0, 10, length=500)
# create an empty plot to which we can add several lines
plot(NA,xlim=c(0,10),ylim=c(0,.6),ylab="Exponential(lambda) pdfs",xlab="x")
# plot exponential pdf against sequence of x values for different choices of lambda
lines(dexp(x.seq,rate=1/(.5))~x.seq,col=1,lwd=2) # lambda = 1/2
lines(dexp(x.seq,rate=1/1)~x.seq,col=2,lwd=2) # lambda = 1
lines(dexp(x.seq,rate=1/2)~x.seq,col=3,lwd=2) # lambda = 2
lines(dexp(x.seq,rate=1/3)~x.seq,col=4,lwd=2) # lambda = 3
lines(dexp(x.seq,rate=1/8)~x.seq,col=5,lwd=2) # lambda = 4
# add a legend to the plot
legend( x = 6, y=.55, legend=c("lambda = 1/2",
                              "lambda = 1",
                              "lambda = 2",
                              "lambda = 3",
                              "lambda = 8"), col=c(1,2,3,4,5), lwd=2, bty="n")
```



Chi-squared

The Chi-squared(ν) pdf is

$$f_X(x;\nu) = \begin{cases} \frac{1}{\Gamma(\nu/2)2^{\nu/2}} x^{\nu/2-1} \exp\left[-\frac{x}{2}\right] & 0 < x < \infty \\ 0 & -\infty < x \le 0. \end{cases}$$

In R, we can type ?dchisq into the console to pull up the R documentation about the chi-squared distribution functions. We see that we can compute the height of the density curve $f_X(x;\nu)$ as dchisq(x = x, df = nu). The df stands for "degrees of freedom", which is the name of the parameter we have denoted by ν .

Chi-squared cdf

We can compute

$$F_X(x;\nu) = \begin{cases} \int_0^x \frac{1}{\Gamma(\nu/2)2^{\nu/2}} t^{\nu/2-1} \exp\left[-\frac{t}{2}\right] dt & 0 < x < \infty \\ 0 & -\infty < x \le 0 \end{cases}$$

in R as pchisq(q = x , df = nu).

Chi-squared quantiles

We can compute

$$Q_X(u;\nu) = \inf\{x : F_X(x;\nu) \ge u\}, \quad 0 < u < 1$$

in R as qchisq(p = u, df = nu).

Plots of different Chi-squared pdfs

Let's plot the pdfs of some different gamma distributions:

```
# create a sequence of x values at which to compute the chi-squared pdfs
x.seq <- seq(0,20, length=500)</pre>
# create an empty plot to which we can add several lines
plot(NA,xlim=c(0,20),ylim=c(0,.5),ylab="chi-squared(nu) pdfs",xlab="x")
# plot chi-squared pdf against sequence of x values for different choices of nu.
lines(dchisq(x.seq, df = 1)~x.seq,col=1,lwd=2) # nu = 1
lines(dchisq(x.seq, df = 2)~x.seq,col=2,lwd=2) # nu = 1
lines(dchisq(x.seq, df = 3)~x.seq,col=3,lwd=2) # nu = 1
lines(dchisq(x.seq, df = 4)~x.seq,col=4,lwd=2) # nu = 1
lines(dchisq(x.seq, df = 5)~x.seq,col=5,lwd=2) # nu = 1
lines(dchisq(x.seq, df = 6)~x.seq,col=6,lwd=2) # nu = 1
lines(dchisq(x.seq, df = 7)~x.seq,col=7,lwd=2) # nu = 1
lines(dchisq(x.seq, df = 8)~x.seq,col=8,lwd=2) # nu = 1
# add a legend to the plot
legend( x = 12,y=.45, legend=c("nu = 1",
                                "nu = 2",
                                "nu = 3",
                                "nu = 4".
                                "nu = 5".
                                "nu = 6",
                                "nu = 7",
                                "nu = 8"),col=c(1,2,3,4,5,6,7,8),lwd=2,bty="n")
      0.5
      0.4
                                                                 nu = 1
chi-squared(nu) pdfs
                                                                 nu = 2
                                                                 nu = 3
      0.3
                                                                 nu = 4
                                                                 nu = 5
      0.2
                                                                 nu = 6
                                                                 nu = 7
                                                                 nu = 8
      0.1
      0
      o.
              0
                               5
                                                10
                                                                  15
                                                                                   20
                                                 Х
```

Beta

The $\text{Beta}(\alpha,\beta)$ pdf is

$$f_X(x; \alpha, \beta) = \frac{\Gamma(\alpha + \beta)}{\Gamma(\alpha)\Gamma(\beta)} x^{\alpha - 1} (1 - x)^{\beta - 1}, \quad 0 < x < 1.$$

In R, we can type ?dbeta into the console to pull up the R documentation about the chi-squared distribution functions. We see that we can compute the height of the density curve $f_X(x; \alpha, \beta)$ as dbeta(x = x, shape1 = alpha, shape2 = beta).

Beta cdf

We can compute

$$F_X(x;\alpha,\beta) = \begin{cases} 1 & 1 \le x < \infty \\ \int_0^x \frac{\Gamma(\alpha+\beta)}{\Gamma(\alpha)\Gamma(\beta)} t^{\alpha-1} (1-t)^{\beta-1} dt & 0 < x < 1 \\ 0 & -\infty < x \le 0 \end{cases}$$

in R as pbeta(q = x, shape1 = alpha, shape2 = beta).

Beta quantiles

We can compute

 $Q_X(u;\alpha,\beta) = \inf\{x : F_X(x;\alpha,\beta) \ge u\}, \quad 0 < u < 1$

in R as qbeta(p = u, shape1 = alpha, shape2 = beta).

Plots of different Beta pdfs

Let's plot the pdfs of some different beta distributions:

```
# create a sequence of x values at which to compute the beta pdfs
x.seq <- seq(0,1, length=500)
# create an empty plot to which we can add several lines
plot(NA,xlim=c(0,1),ylim=c(0,4.5),ylab="Beta(alpha,beta) pdfs",xlab="x")
# plot beta pdf against sequence of x values for different choices of alpha and beta.
lines(dbeta(x.seq, shape1 = 1/2, shape2 = 1/2)~x.seq, col=1, lwd=2) # alpha = 1/2, beta = 1/2
lines(dbeta(x.seq, shape1 = 4, shape2 = 4)~x.seq, col=2, lwd=2) # alpha = 5, beta = 5
lines(dbeta(x.seq, shape1 = 2, shape2 = 1)~x.seq, col=3, lwd=2) # alpha = 2, beta = 1
lines(dbeta(x.seq, shape1 = 1, shape2 = 10)~x.seq,col=4,lwd=2) # alpha = 1, beta = 10
lines(dbeta(x.seq, shape1 = 10, shape2 = 3)~x.seq,col=5,lwd=2) # alpha = 10, beta = 3
# add a legend to the plot
legend( x = .2,y=4.5, legend=c("alpha = 1/2, beta = 1/2",
                             "alpha = 4, beta = 4",
                             "alpha = 2, beta = 1",
                             "alpha = 1, beta = 10",
                             "alpha = 10, beta = 3"), col=c(1,2,3,4,5), lwd=2, bty="n")
```

