Assignment 4 Computer simulation course

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Part 1:Generating random variants

1 About Beta Distribution

1.1 Beta Function

- Fixed parameters: $\alpha > 0, \beta > 0$
- We define the beta function

 $B(\alpha, \beta) := \int_0^1 y^{\alpha - 1} (1 - y)^{\beta - 1} dy = \text{some constant}$

• Relationship between the gamma and beta functions:

$$B(\alpha, \beta) = \frac{\Gamma(\alpha)\Gamma(\beta)}{\Gamma(\alpha + \beta)}$$

• This makes it easy to compute for whole numbers, since $\Gamma(n) = (n-1)!$.

(Plug in values for α and β , and it spits out a number.) Don't mix up the beta function and the beta distribution!

1.2 Beta Distribution

• We want to define a distribution on [0,1] using the function

$$y^{\alpha-1}(1-y)^{\beta-1}$$

- Recall that a probability density function must satisfy $\int_0^1 f(y)dy = 1$. To make this work, we divide by $B(\alpha, \beta)$:
- Density function for the beta distribution:

$$f(y) := \frac{y^{\alpha - 1}(1 - y)^{\beta - 1}}{B(\alpha, \beta)}, 0 \le y \le 1$$

Don't mix up the beta function and the beta distribution! Key Properties of the Beta Distribution

• Mean:

$$\mu = E(Y) = \frac{\alpha}{\alpha + \beta}$$

• Variance:

$$\sigma^{2} = V(Y) = \frac{\alpha\beta}{(\alpha + \beta)^{2}(\alpha + \beta + 1)}$$

• Standard deviation:

$$\sigma = \sqrt{V(Y)} = \sqrt{\frac{\alpha\beta}{(\alpha+\beta)^2(\alpha+\beta+1)}}$$

2 Generating Beta variants

2.1 Finding PDF Beta Function

Alpha and beta based on my student ID : $\alpha=3,\beta=4$

Calculate B(3,4)

$$B(\alpha, \beta) = \frac{\Gamma(\alpha)\Gamma(\beta)}{\Gamma(\alpha + \beta)}$$

$$B(3, 4) = \frac{\Gamma(3)\Gamma(4)}{\Gamma(7)}$$

$$= \frac{2! \cdot 3!}{6!}$$

$$= \frac{2 \cdot 6}{720}$$

$$= \frac{1}{60}$$

$$f(x) = \frac{x^{\alpha - 1}(1 - x)^{\beta - 1}}{B(\alpha, \beta)} = 60 * (x^2) * ((1 - x)^3), 0 \le x \le 1$$

2.2The acceptance-Rejection method

The Acceptance-Rejection Method

Suppose that X and Y are r.v. with density or pmf f and g respectively, and there exists a constant c such that $\frac{f(t)}{g(t)} \leq c$ for all t such that f(t) > 0. Then the acceptance-rejection method can be applied to generate the r.v. X.

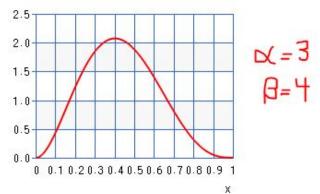
- 1. Find a r.v. Y with density g satisfying $f(t)/g(t) \leq c$, for all t such that f(t) > 0
 - 2. Generate a random y from the dist. with density g
 - 3. Generate a random u from the U(0,1) dist.
- 4. If u < f(y)/(cg(y)) accept y and deliver x = y; otherwise reject y and

Note that in step 4, $P(\text{ accept } \mid Y) = P\left(U < \frac{f(Y)}{cg(Y)} \mid Y\right) = \frac{f(Y)}{cg(Y)}$. The total prob. of acceptance for any iteration is

$$\sum_{y} P(\text{ accept } \mid y)P(Y=y) = \sum_{y} \frac{f(y)}{cg(y)}g(y) = \frac{1}{c}$$

2.2.2 Finding appropriate C

We have two ways to find a suitable C; the first and best one is calculating derivative of Beta's PDF function (which will tell us the maximum of PDF occurs at y = 0.4 and the second method is finding approximate c by the plot of PDF:



So, $c = f(0.4) = 2.07360 \approx 2.1$.

Implementation in R language

```
n \leftarrow 1000; k \leftarrow 0 # counter for accepted
y <- numeric(n); j <- 0 # iterations
while (k < n) \{u < - runif (1); j < -j+1;
x <- runif (1) # random variate from q
#we accept x if:
if ((60 * (1/2.1) * (x^2) * ((1-x)^3)) > u){k <-k +1; y[k] <- x}}
print(j)
## [1] 2118
# compare empirical and theoretical percentiles
p <-seq(.1 ,.9 ,.1); Qhat <- quantile(y,p) # quantiles of sample</pre>
Q <- qbeta(p, 3, 4) # theoretical quantiles
se \leftarrow sqrt((60 * (p^2) * ((1-p)^3)) / (n * dbeta(Q, 3, 4)))
# Now we compare
round( rbind (Qhat , Q, se), 3)
          10%
                20% 30% 40%
                                   50%
                                          60%
                                                70%
                                                      80%
## Qhat 0.185 0.257 0.322 0.377 0.428 0.477 0.527 0.591 0.675
        0.201 0.269 0.323 0.373 0.421 0.471 0.524 0.585 0.667
## se 0.019 0.027 0.031 0.032 0.030 0.026 0.021 0.014 0.007
```

Part 2: Evaluation of the result

As we can see the result of Q and Qhat, they are approximately the same. Therefore, we can be statistically sure that the generated numbers are of Beta distribution.