

Lecture 7: Generating random variables – acceptance-rejection sampling

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Previously

- ▶ Generate $U \sim \text{Uniform}(0, 1)$ with random number generator (LCG, Mersenne twister, etc)
- ▶ Some other random variables can be generated by transformations
 - ▶ Inverse transform method (if inverse cdf is easy to get)
 - ▶ Other transformations for certain variables

Today: What is another alternative?

Example

Suppose we would like to generate $X \sim \text{Beta}(\alpha, \beta)$

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

- ▶ Inverse transform method: $F_X(t) = ?$
- ▶ Other transformations / relationships with other distributions?

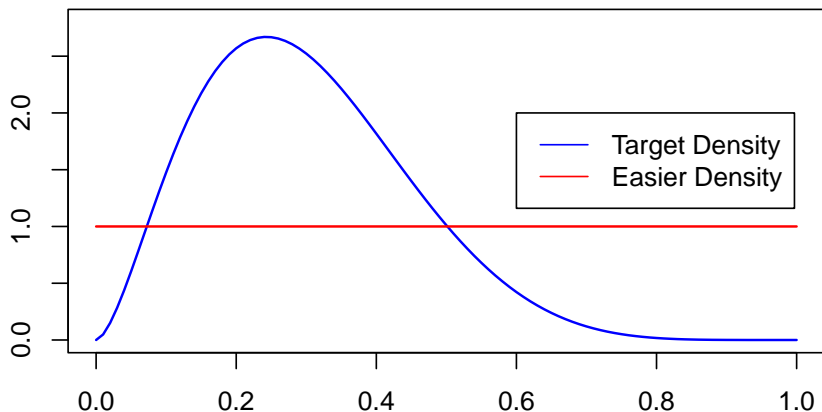
Acceptance-rejection sampling: motivation

► **Want:** $X \sim \text{Beta}(\alpha, \beta)$, $f_X(x) = \frac{\Gamma(\alpha+\beta)}{\Gamma(\alpha)\Gamma(\beta)} x^{\alpha-1} (1-x)^{\beta-1}$

However, it may be difficult to directly simulate from this distribution. Can you think of another distribution on $(0, 1)$ which is *easier* to simulate?

Acceptance-rejection sampling: motivation

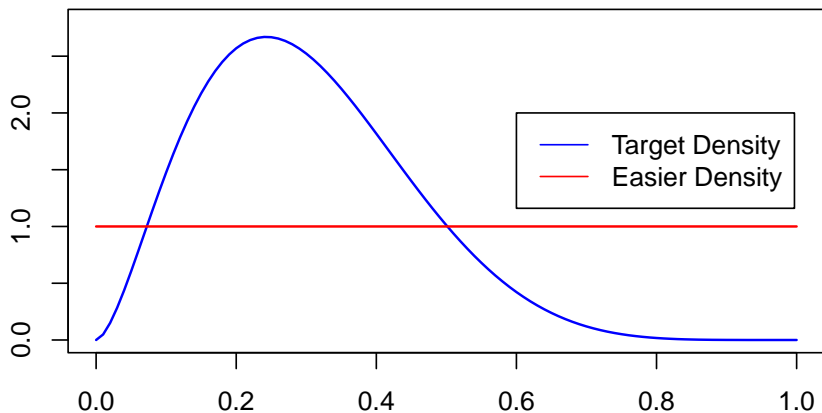
- ▶ **Want:** $X \sim \text{Beta}(\alpha, \beta)$, $f_X(x) = \frac{\Gamma(\alpha+\beta)}{\Gamma(\alpha)\Gamma(\beta)} x^{\alpha-1} (1-x)^{\beta-1}$
- ▶ **Can get:** $Y \sim \text{Uniform}(0, 1)$



Suppose we sample $Y \sim \text{Uniform}(0, 1)$ and observe $y = 0.9$. Is it likely we would observe that draw from the Beta distribution shown here?

Acceptance-rejection sampling: motivation

- **Want:** $X \sim \text{Beta}(\alpha, \beta)$, $f_X(x) = \frac{\Gamma(\alpha+\beta)}{\Gamma(\alpha)\Gamma(\beta)} x^{\alpha-1} (1-x)^{\beta-1}$
- **Can get:** $Y \sim \text{Uniform}(0, 1)$



Suppose we sample $Y \sim \text{Uniform}(0, 1)$ and observe $y = 0.25$. Is it likely we would observe that draw from the Beta distribution shown here?

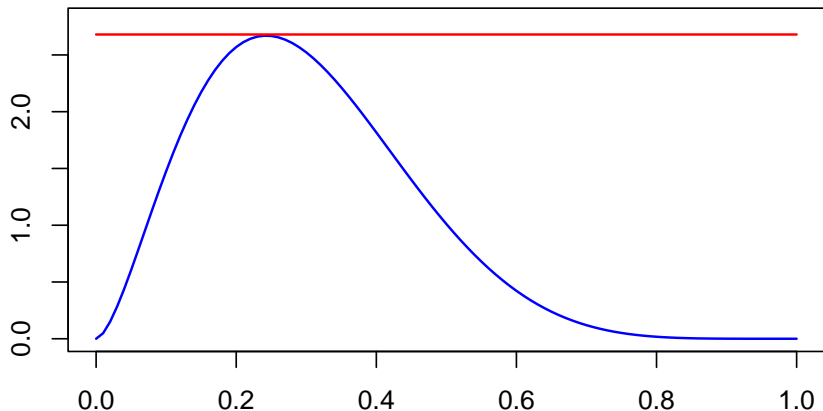
Acceptance-rejection sampling

Suppose we would like to generate a continuous random variable X with pdf f .

Illustration

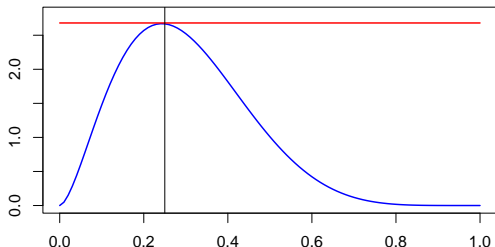
Target density: $f(t) = \frac{\Gamma(\alpha+\beta)}{\Gamma(\alpha)\Gamma(\beta)} x^{\alpha-1} (1-x)^{\beta-1}$, $t \in (0, 1)$

Candidate density: $g(t) = 1$, $t \in (0, 1)$



Illustration

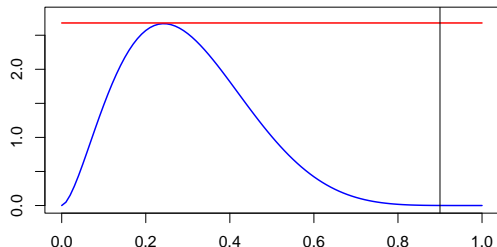
Now sample $Y \sim g$ and $U \sim \text{Uniform}(0, 1)$. Accept Y if $U \leq \frac{f(Y)}{cg(Y)}$



Suppose we observe $Y = 0.25$. Are we likely to *accept* or *reject* 0.25 as a sample from f ?

Illustration

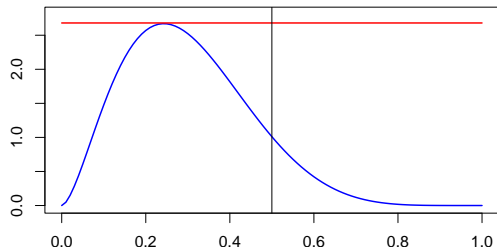
Now sample $Y \sim g$ and $U \sim \text{Uniform}(0, 1)$. Accept Y if $U \leq \frac{f(Y)}{cg(Y)}$



Suppose we observe $Y = 0.9$. Are we likely to *accept* or *reject* 0.9 as a sample from f ?

Illustration

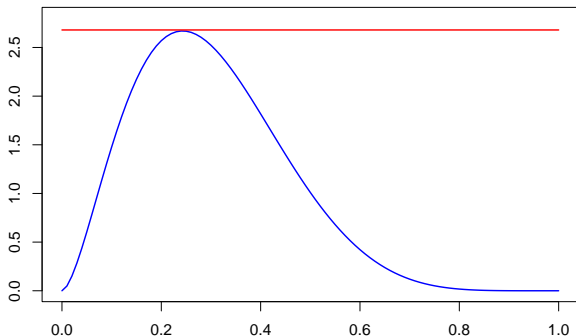
Now sample $Y \sim g$ and $U \sim \text{Uniform}(0, 1)$. Accept Y if $U \leq \frac{f(Y)}{cg(Y)}$



Suppose we observe $Y = 0.5$. Are we likely to *accept* or *reject* 0.5 as a sample from f ?

Illustration

- ▶ $Y \sim g$ and $U \sim \text{Uniform}(0, 1)$
- ▶ Accept Y if $U \leq \frac{f(Y)}{cg(Y)}$



Your turn

Practice questions on the course website:

https://sta379-s25.github.io/practice_questions/pq_7.html

- ▶ Implement acceptance-rejection sampling for the beta example
- ▶ Start in class. You are welcome to work with others
- ▶ Practice questions are to help you practice. They are not submitted and not graded
- ▶ Solutions are posted on the course website

Next time: formal proof that this sampling procedure works!