9 Stochastic simulation

- Maybe talk about number generation
- Define cdf
- List requirements for inverse cdf
- Prove the below proposition
- Prove the Inverse Transformation Method

Det ønskes at simulere $U \sim \text{unif}[0, 1]$.

- Approksimér [0,1] ved $\{0,\frac{1}{m},\ldots,\frac{m-1}{m},1\}$ for store m
- Ønskes at simulere $U = \frac{Y}{m}$, $Y \sim \text{unif}\{0, 1, \dots, m\}$
- Lad Y_0 være givet ved et seed Lad $Y_{n+1} = aY_n + b$
- Så er sekvensen Y_1, Y_2, \ldots en sekvens af i.i.d. RVer $\sim \text{unif}\{0, 1, \ldots, m\}$

Any discrete distribution X which assumes values x_1, x_2, \ldots with probability p_1, p_2, \ldots can be simulated by parting $U \sim \text{unif}[0, 1]$ into subintervals of the p's and assigning $X = x_k$ if U lands in the interval $[p_{k-1}, p_k]$.

Proposition 5.1 Consider the pmf p on the range $\{x_1, x_2, \ldots\}$ and let

$$F_0 = 0, \quad F_k = \sum_{j=1}^k p(x_k), \qquad k = 1, 2, \dots$$
 (1)

Let $U \sim \text{unif}[0,1]$ and let $X = x_k$ if $Fk - 1 < U \le F_k$. Then X has pmf p. Bevis

Note that $X = x_k$ if and only if $U \in (F_{k-1}F_k]$, which has probability

$$P(X = x_k) = P(F_{k-1} < U \le F_k) = F_k - F_{k-1} = p(x_k), \quad k = 1, 2, \dots (2)$$

If the range is finite $\{x_1, x_2, \dots, x_n\}$ we get $F_n = 1$.

Proposition 5.2 (The Inverse Transformation Method) Let F be a distribution function that is continuous and strictly increasing. Further, let $U \sim \text{unif}[0,1]$ and define then random variable $Y = F^{-1}(U)$. Then Y has distribution function F.

Bevis

Let F_y be distribution function of Y and let x be in range of Y:

$$F_Y(x) = P(F^{-1}(U) \le x)$$

= $P(U \le F(x)) = F_U(F(x)) = F(x)$

Proposition 5.3 (The Rejection Method)

- 1. Generate Y and $U \sim \text{unif}[0,1]$ independent of each other.
- 2. If $U \leq \frac{f(Y)}{cg(Y)}$, set X = Y. Otherwise return to step 1.

The random variable X generated by the algorithm has pdf f.

Bevis

Make sure the algorithm terminates. Calculate the probability of succes:

$$P\left(U \le \frac{f(Y)}{cg(Y)}\right) = \int_{R} P\left(U \le \frac{f(Y)}{cg(Y)}\right) g(y) dy$$
$$= \int_{R} \frac{f(y)}{cg(y)} g(y) dy = \frac{1}{c} \int_{R} f(y) dy = \frac{1}{c}$$

where we used the independence of U and Y and the fact that $U \sim \text{unif}[0,1]$. The probability of succes is 1/c. Now we want to show that the conditional distribution of Y is the same as the distribution of X. Use conditional probability:

$$P\left(Y \le x | U \le \frac{f(Y)}{cg(Y)}\right) = cP\left(Y \le x, U \le \frac{f(Y)}{cg(Y)}\right) \tag{3}$$

With independence we get joint pdf f(u, y) = g(y).

$$\begin{split} cP\left(Y \leq x, U \leq \frac{f(Y)}{cg(Y)}\right) &= c\int_{-\infty}^{x} \int_{0}^{f(y)/cg(y)} g(y) \, dy \\ &= c\int_{-\infty}^{x} \frac{f(y)}{cg(y)} g(y) \, dy = P(X \leq x) \end{split}$$