Explanation of algorithm for sampling from  $s_k(x) = u_k(x)/\int_D u_k(x')dx' = c \ u_k(x)$ . First we need to find the normalizing constant  $c := 1/\int_D u_k(x')dx'$ :

$$\int_{D} u_{k}(x')dx' = \sum_{j=1}^{k} \int_{z_{j-1}}^{z_{j}} \exp(h(x_{j}) + (x' - x_{j})h'(x_{j}))dx'$$

$$= \sum_{j=1}^{k} \frac{1}{h'(x_{j})} \exp(h(x_{j}) + (x' - x_{j})h'(x_{j})) \Big|_{x'=z_{j-1}}^{z_{j}}$$

$$= \sum_{j=1}^{k} \frac{\exp u_{k}(z_{j}) - \exp u_{k}(z_{j-1})}{h'(x_{j})}$$

Next, in order to use the inverse CDF method for sampling, we must find the CDF for  $s_k(x)$ ,  $S_k(x) = c \int_{z_0}^x u_k(x') dx'$ :

$$S_k(x) = c \left( \sum_{j=1}^{t-1} \int_{z_{j-1}}^{z_j} \exp(h(x_j) + (x' - x_j)h'(x_j))dx' + \int_{z_{t-1}}^x \exp(h(x_t) + (x' - x_t)h'(x_t))dx') \right)$$

$$= c \left( \sum_{j=1}^{t-1} \frac{\exp u_k(z_j) - \exp u_k(z_{j-1})}{h'(x_j)} + \frac{\exp u_k(x) - \exp u_k(z_{t-1})}{h'(x_t)} \right)$$

where t is the index of which interval of z's that x lies in. Formally, it is t(x) = i:  $x \in (z_{i-1}, z_i)$ . For convenience, we call the sum term above partialSums[t-1]. Note that our normalizing constant c = 1/partialSums[k]. Thus,

$$S_k(x) = c \left( \texttt{partialSums[t-1]} + \frac{\exp u_k(x) - \exp u_k(z_{t-1})}{h'(x_t)} \right)$$

Now we can determine the inverse transform  $S_k^{-1}(U)$ , where  $U \sim \text{Uniform}[0,1]$ . Because t is actually a function of x, it too must be inverted. Intuitively, we want to pick the biggest t such that  $S_k(z_{t-1}) < U$ . Formally,  $t(U) = i : U \in (c \text{ partialSums}[i - 1], c \text{ partialSums}[i])$ . Solving for the inverse, we have:

$$S_k^{-1}(U) = \frac{\log\left(h'(x_t)\left(\frac{U}{c} - \mathtt{partialSums[t-1]}\right) + \exp{u_k(z_{t-1})}\right) - h(x_t)}{h'(x_t)} + x_t$$