

Inverse CDF Sampling

Troy Stribling

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1 Introduction

Inverse [CDF](#) sampling is a method for obtaining samples from both discrete and continuous probability distributions that requires the CDF to be invertible. The method generates a CDF value from a Uniform random variable on $[0, 1]$ that is then used as input into the inverted CDF to generate a sample with the desired distribution. Here examples for both cases are discussed. For the continuous case a proof is given that demonstrates the samples produced have the expected distribution.

2 Sampling Discrete Distributions

A discrete probability distribution consisting of a finite set of N probability values is defined by,

$$\{p_1, p_2, \dots, p_N\}$$

with $p_i \geq 0, \forall i$ and $\sum_{i=1}^N p_i = 1$.

The CDF specifies the probability that $i \leq n$ and is given by,

$$P(i \leq n) = P(n) = \sum_{i=1}^n p_i, \tag{1}$$

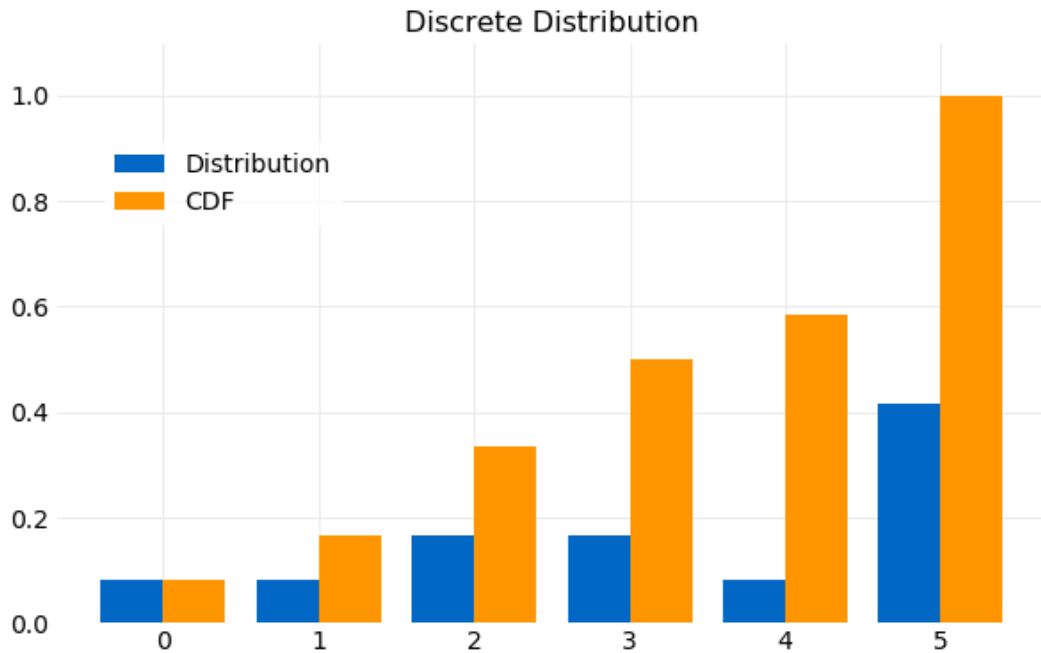
where $P(N) = 1$.

For a given generated CDF value, u , equation (1) can always be inverted by evaluating it for each n and searching for the value of n that satisfies, $P(n) \geq u$. It can be seen that the generated samples will have distribution $\{p_n\}$ since the intervals $P(n) - P(n-1) = p_n$ are Uniformly sampled.

Consider the distribution,

$$\left\{ \frac{1}{12}, \frac{1}{12}, \frac{1}{6}, \frac{1}{6}, \frac{1}{12}, \frac{5}{12} \right\} \quad (2)$$

It is shown in the following plot with its CDF.



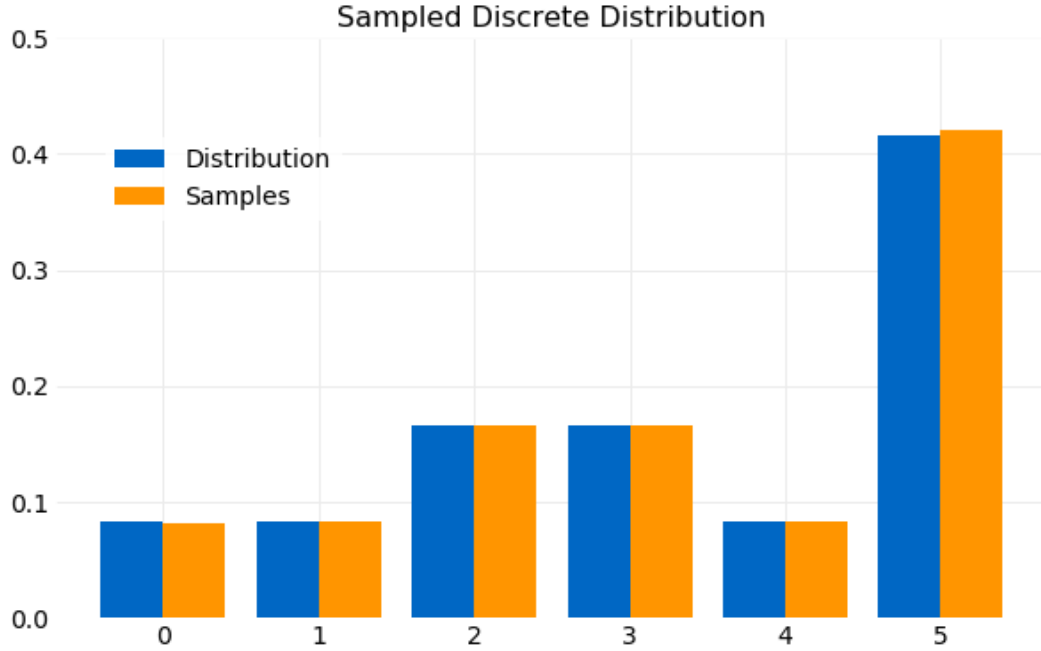
A sampler using the Inverse CDF method can be implemented in Python in a few lines of code,

```
import numpy

n = 10000
df = numpy.array([1/12, 1/12, 1/6, 1/6, 1/12, 5/12])
cdf = numpy.cumsum(df)

cdf_star = numpy.random.rand(n)
samples = [numpy.flatnonzero(cdf >= cdf_star[i])[0] for i in range(n)]
```

The figure below favorably compares samples generated by the Inverse CDF sampler and distribution (2),



It is also possible to directly sample $\{p_n\}$ using the `multinomial` sampler from `numpy`,

```
import numpy

n = 10000
df = numpy.array([1/12, 1/12, 1/6, 1/6, 1/12, 5/12])
samples = numpy.random.multinomial(n, df, size=1)/n
```

3 Sampling Continuous Distributions

A continuous probability distribution is defined by the [PDF](#),

$$f_X(x),$$

where $f_X(x) \geq 0, \forall x$ and $\int f_X(x)dx = 1$.

The CDF is a monotonically increasing function that specifies probability that $X \leq x$,

$$P(X \leq x) = F_X(x) = \int^x f_X(w)dw \quad (3)$$

To prove that Inverse CDF sampling works is necessary to show that,

$$P[F_X^{-1}(u) \leq x] = F_X(x),$$

where $F_X^{-1}(x)$ is the inverse of $F_X(x)$ and u is Uniform on $[0, 1]$. Since $F_X(x)$ is monotonically increasing,

$$P[F_X^{-1}(u) \leq x] = P[F_X(F_X^{-1}(u)) \leq F_X(x)] \quad (4)$$

$$= P[u \leq F_X(x)] \quad (5)$$

$$= \int_0^{F_X(x)} dw \quad (6)$$

$$= F_X(x) \quad (7)$$

and the desired result is obtained.