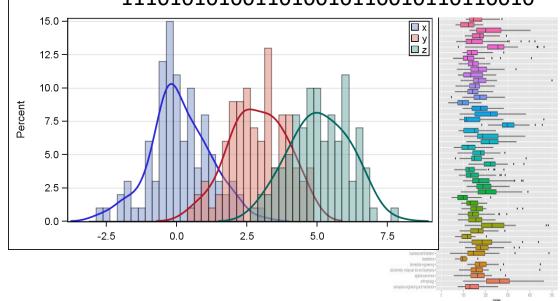
$\begin{cases} x_n \\ y_n \\ y_$

Inverse Transform Sampling





The distribution of F(X)

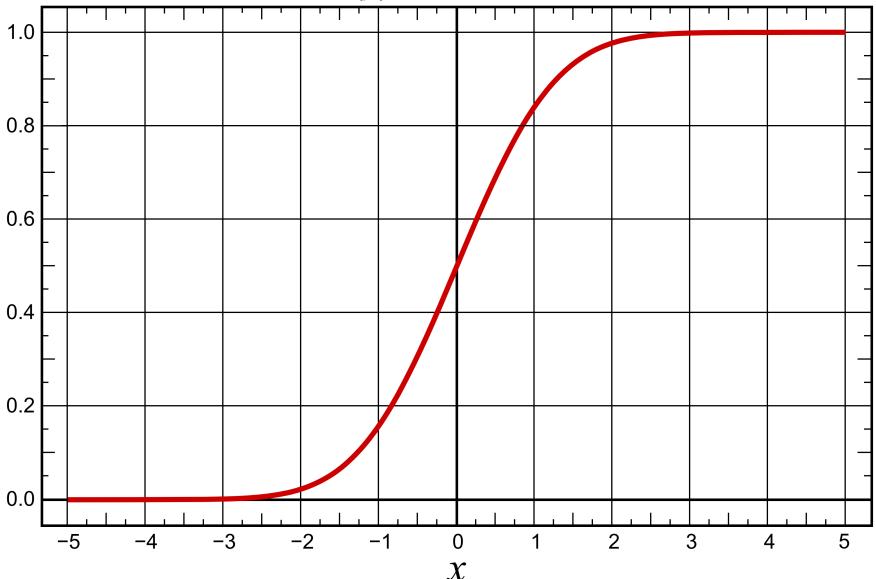
- Generate 1000 samples from a standard normal distribution: $x_1, ..., x_{1000}$
- Trace each sampled value to its CDF: $c_i = \Phi(x_i)$

- How many c_i s do we expect to see in [-5, -3]?
- How many c_i s do we expect to see in [0,1]?
- How many c_i s do we expect to see in [0,0.2] ?
- How many c_i s do we expect to see in [0.4,0.6] ?



Cumulative distribution function

for normal distribution, mean (μ) = 0 and standard deviation (σ) = 1





The distribution of F(X)

- Generate 1000 samples from a standard normal distribution: $x_1, ..., x_{1000}$
- Trace each sampled value to its CDF: $c_i = \Phi(x_i)$
- How many c_i s do we expect to see in [-5, -3]?
- How many c_i s do we expect to see in [0,1]?
- How many c_i s do we expect to see in [0,0.2]?
- How many c_i s do we expect to see in [0.4,0.6]?
- We can transform a sample from any distribution into a uniform sample, using the CDF of that distribution.
- By applying the inverse function of the CDF, we would be able to map the uniform distribution to a sample from any distribution we want

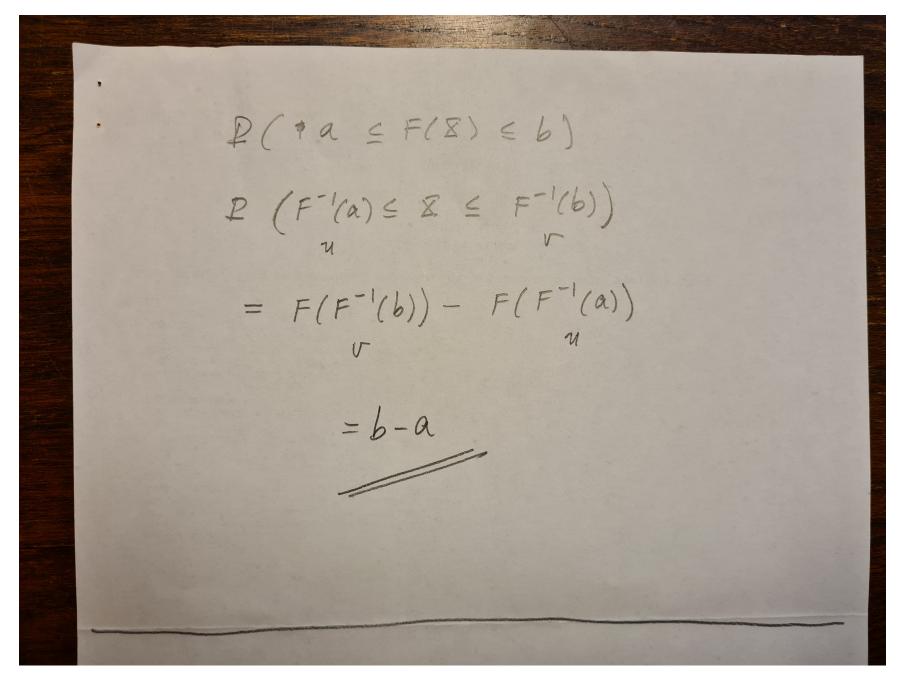


Inverse Mapping

• Formally, if X is a continuous random variable with cumulative distribution function F_X , then the random variable $U = F_X(X)$ has a uniform distribution on [0,1].

• If U has a uniform distribution on [0,1] and if X has a cumulative distribution F_X then the random variable $F_X^{-1}(U)$ has the same distribution as X.





$$P\left(a \leq F'(\overline{U}) \leq b\right)$$

$$P\left(F(a) \leq \overline{U} \leq F(b)\right)$$

$$= F(b) - F(a)$$

Practicalities

• Let X be a RV and let F_X be the CDF of X

- ullet Generate a random number u from the standard uniform distribution U
- Compute $x = F_X^{-1}(u)$
- This procedure yields sampling from X

