

DA-STATS

Topic 05: Univariate Sampling

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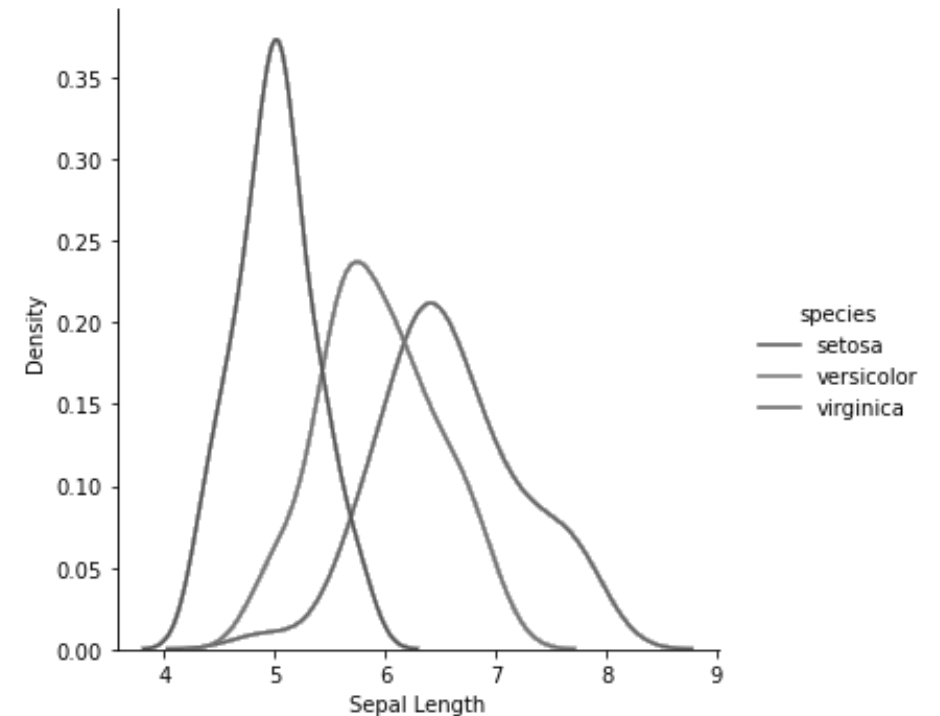
Outline



- Distributions
- Probability Density Function (PDF)
- Cumulative Distribution Function (CDF)
- Sampling from Univariate Distribution
 - Inverse CDF Sampling
 - Reject Sampling
 - Monte-Carlo Simulation

Sampling from univariate distributions

- How do we draw random numbers from a given distribution?
- What are distributions?



Distributions



- Mathematical function that gives the probabilities of occurrence of different possible outcomes for a random phenomenon
- Examples
 - Coin Toss
 - Dice
 - Someone getting sick from viral infection

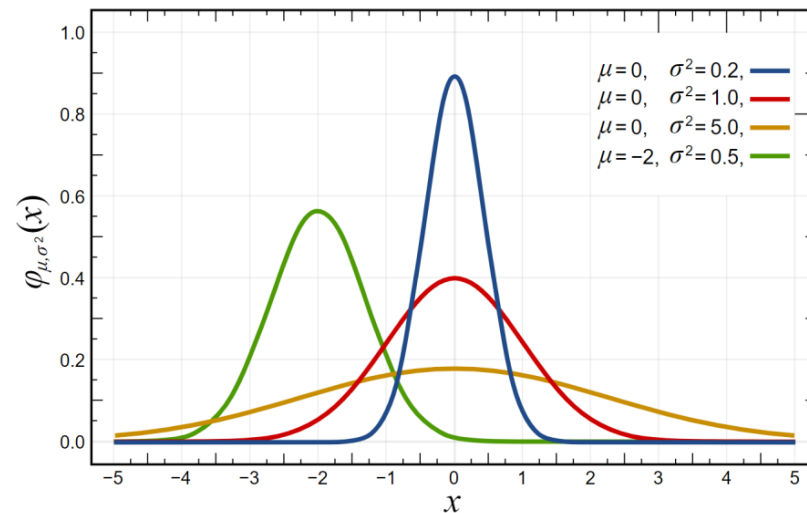
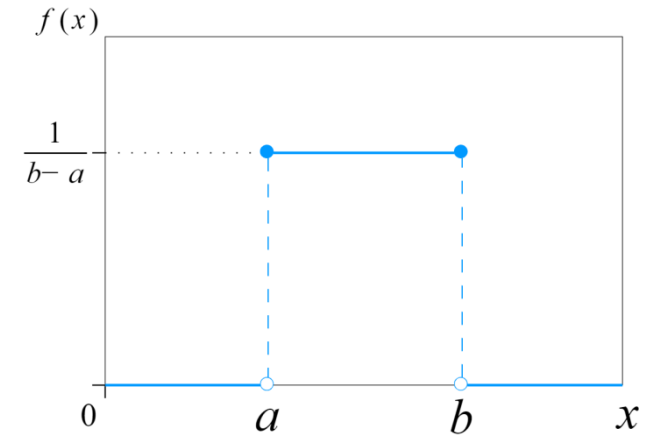
Distribution Examples

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- Uniform Distribution

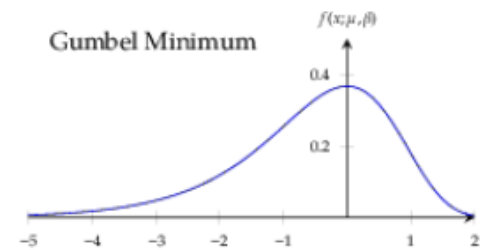
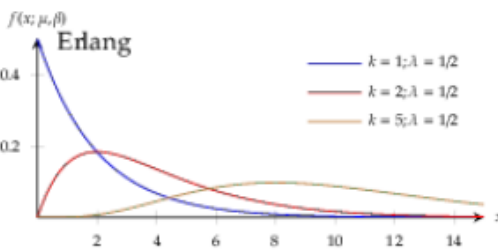
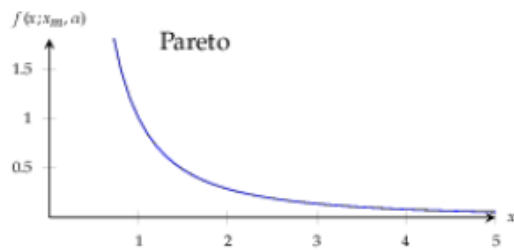
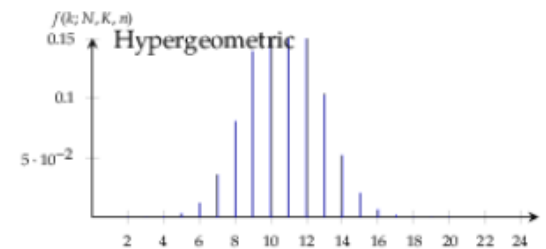
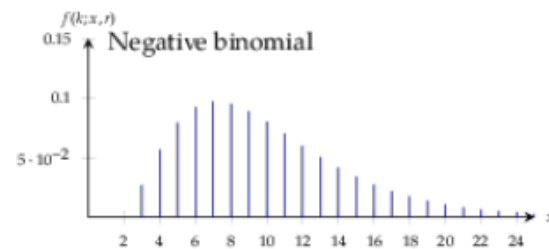
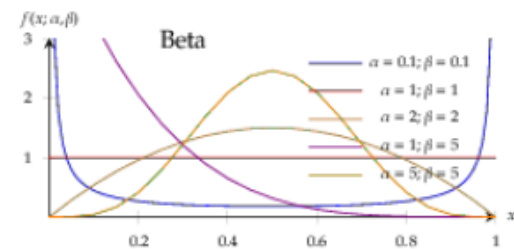
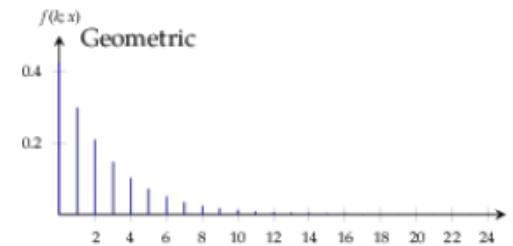
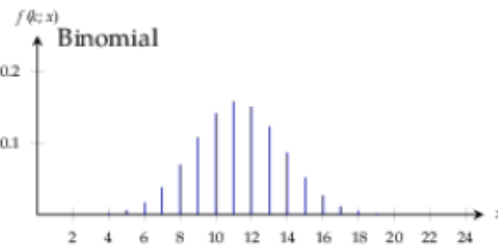
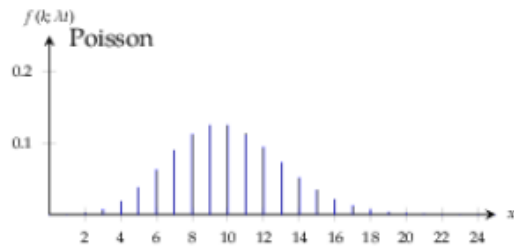
$$p(x) = \frac{1}{b-a} \text{ for } x \in [a, b] \text{ else } 0.0$$

- Normal Distribution
- Gaussian Distribution
- Characterization of distributions

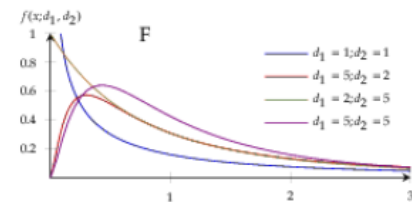
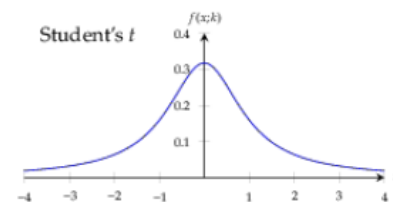
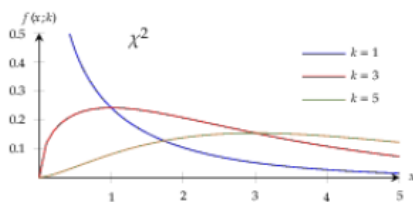
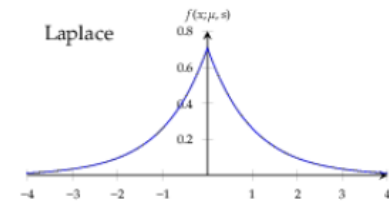
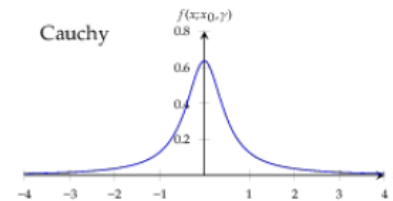
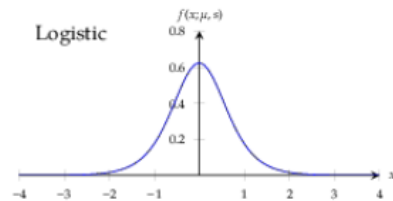
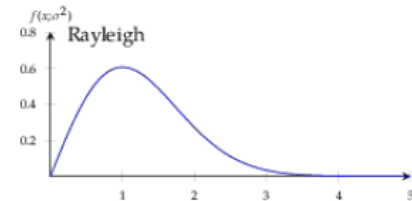
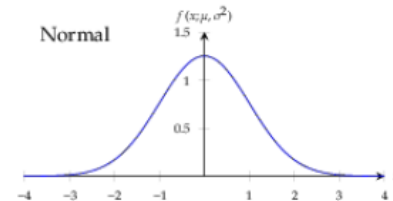
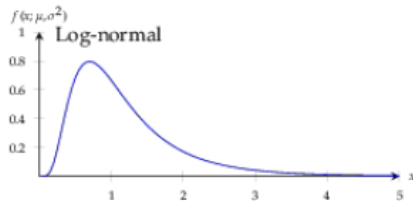
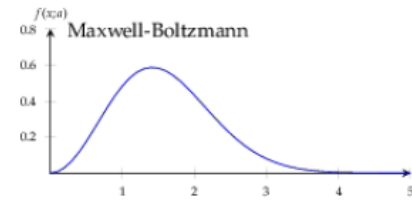
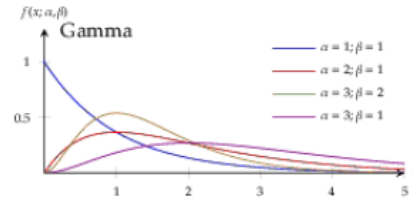
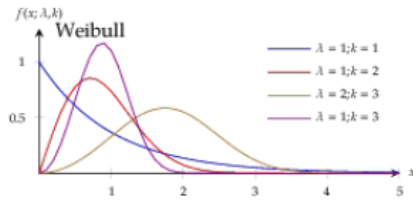


$$f(x) = \frac{1}{\sigma\sqrt{2\pi}} e^{-\frac{1}{2}\left(\frac{x-\mu}{\sigma}\right)^2}$$

Distribution Examples



Distribution Examples



Central Limit Theorem



- Given a population with mean μ and standard deviation σ and take sufficiently large random samples from the population with replacement
- The **distribution of the sample means** will be approximately normally distributed.
- Holds true regardless of whether the source population is normal or skewed, provided the sample size is sufficiently large.
- If the population is normal, then the theorem holds true even for samples smaller than 30.

https://sphweb.bumc.bu.edu/otlt/mph-modules/bs/bs704_probability/BS704_Probability12.html

Sampling from a univariate distribution



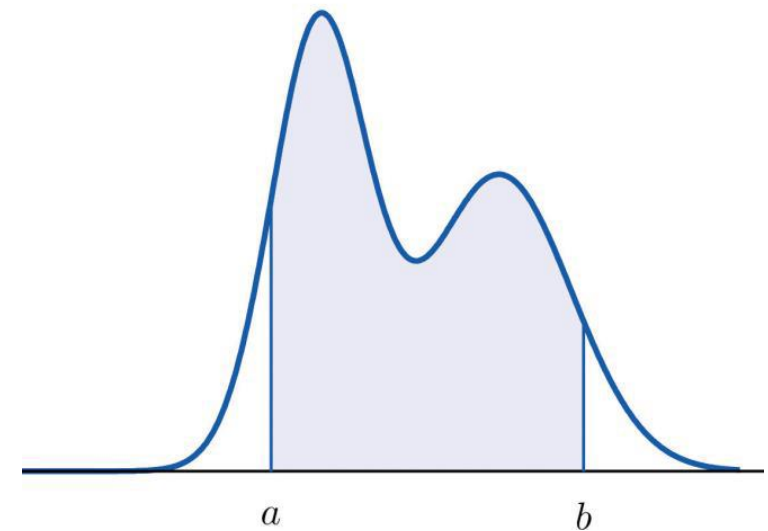
- How to generate random numbers that follow a distribution?
- Implementation
 - `np.random.uniform(0, 1, N)`
 - `np.random.gaussian`
 - More in lab exercises

<https://docs.scipy.org/doc/numpy 1.15.0/reference/routines.random.html>

How are these Random Numbers Generated?

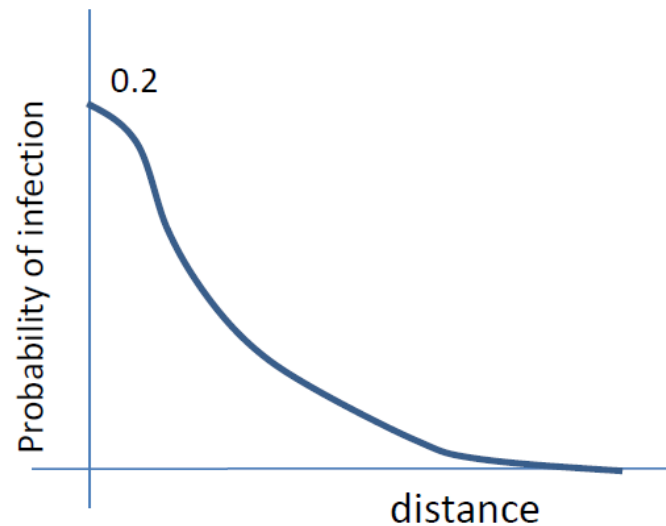


- How to sample from an arbitrary distribution?
- Let's say, you are required to generate random samples from a distribution like this on the right.
- Why would someone ask you to generate random samples that follow a distribution?

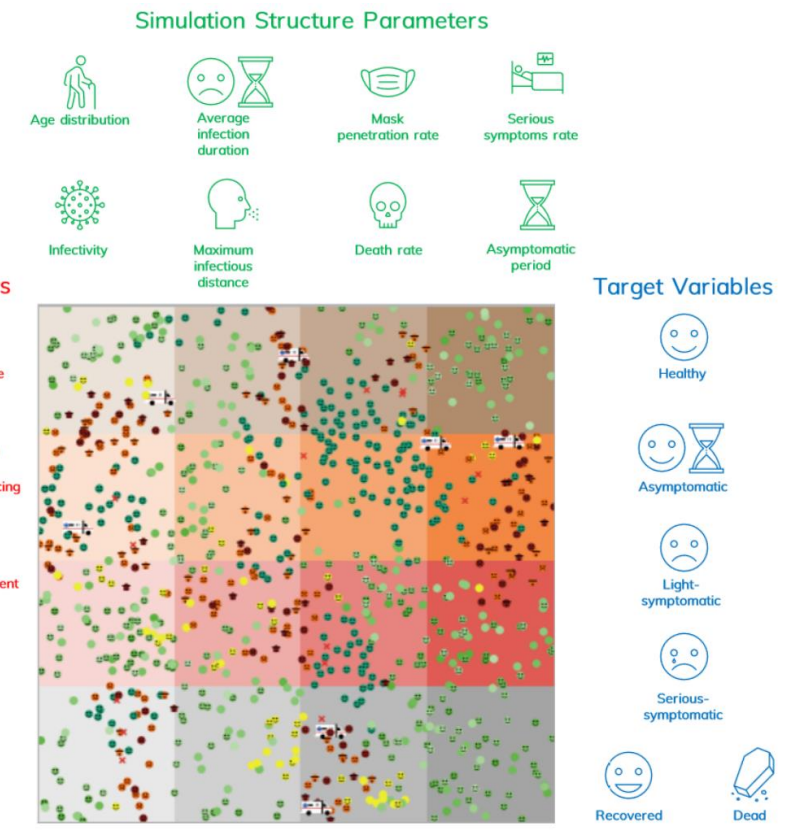


Application Example

- Let's say, we want to simulate how a virus transmits between two people given that one of them is infected.



<http://arxiv.org/abs/2011.11381>

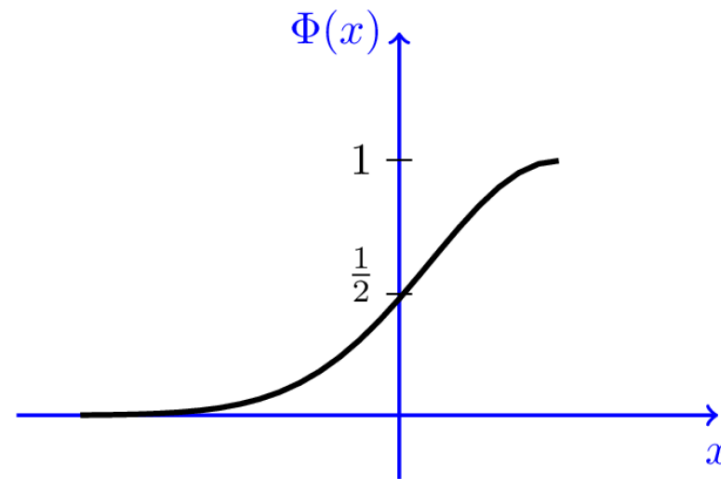
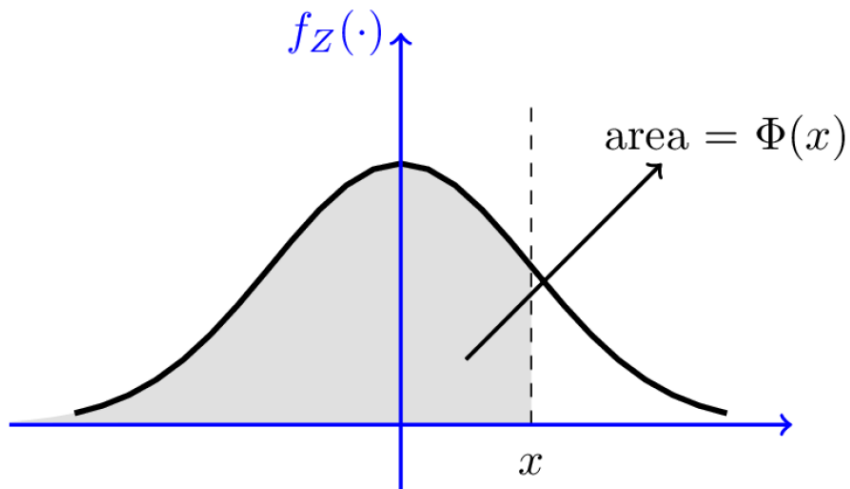
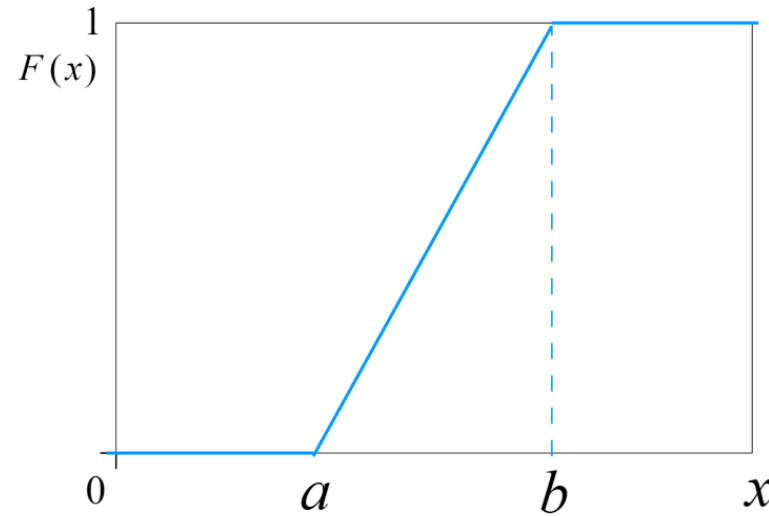
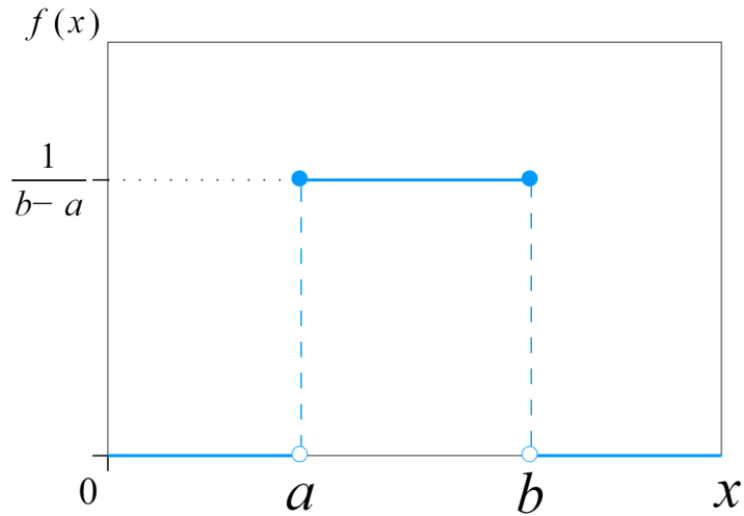


Inverse Transform Sampling

- Also known as: Inverse CDF Sampling
- What is a CDF?
 - PDF (or PMF) Formally, probability density (mass) function
 - $p_X(x)$: probability of observing the random variable X with value x
 - CDF (Cumulative Distribution Function)
 - The cumulative probability

$$F_X(x) = p_X(X \leq x) = \int_{-\infty}^x p_X(x)$$

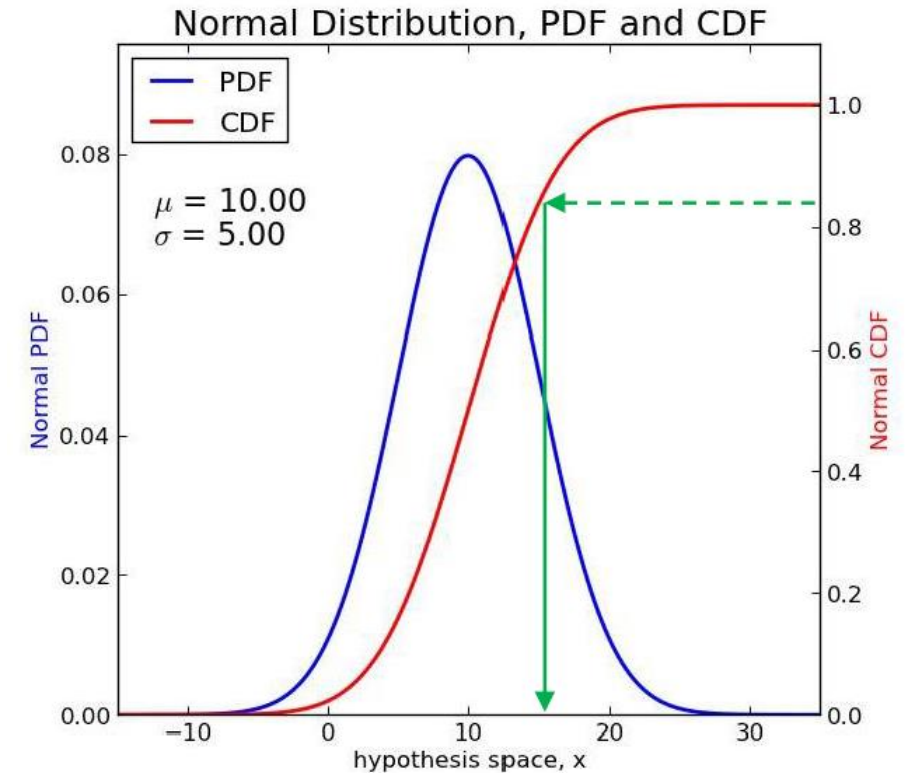
CDF Example



<https://demonstrations.wolfram.com/ConnectingTheCDFAndThePDF/>

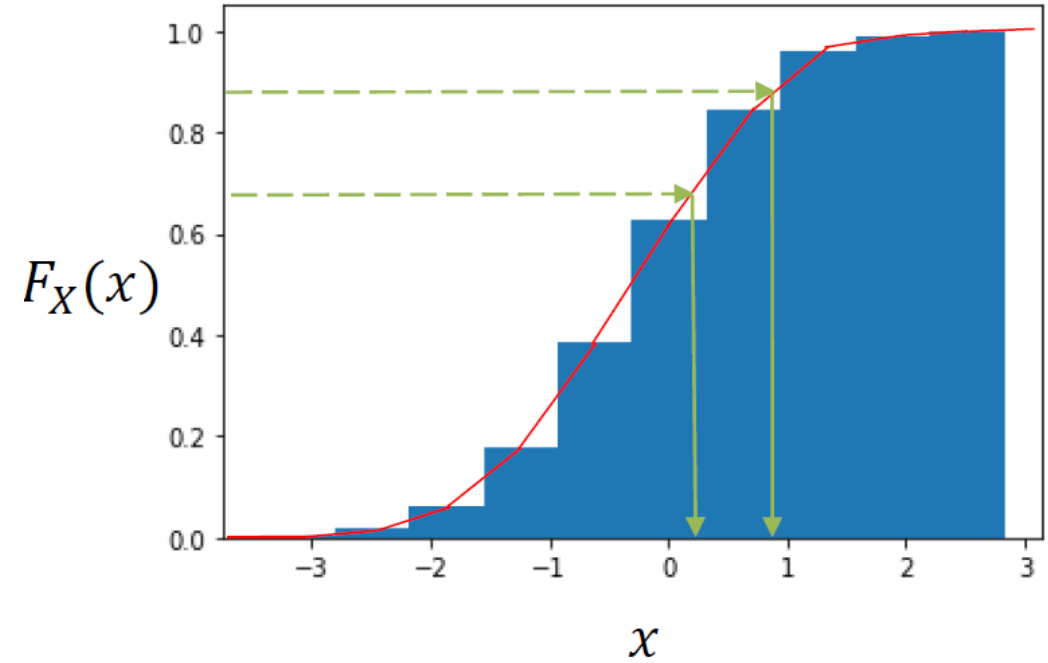
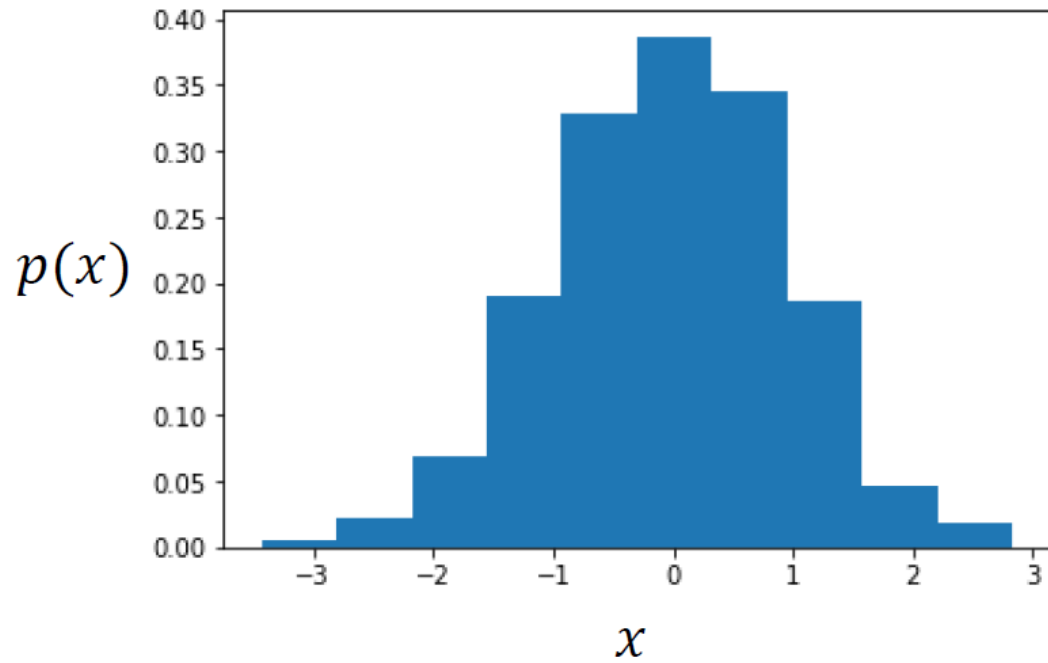
Inverse CDF Sampling

- Generate a random number u from the standard uniform distribution in the interval $[0, 1]$, e.g., from $U \sim \text{Unif}[0,1]$.
- Find the inverse of the desired CDF, e.g., $F_X^{-1}(x)$.
- Compute $X = F_X^{-1}(u)$. The computed random variable X has distribution $F_X(x)$.



https://en.wikipedia.org/wiki/Inverse_transform_sampling

Exercise



Sampling: Reject Sampling

- Keep samples that follow the distribution and reject others



Finding the value of π

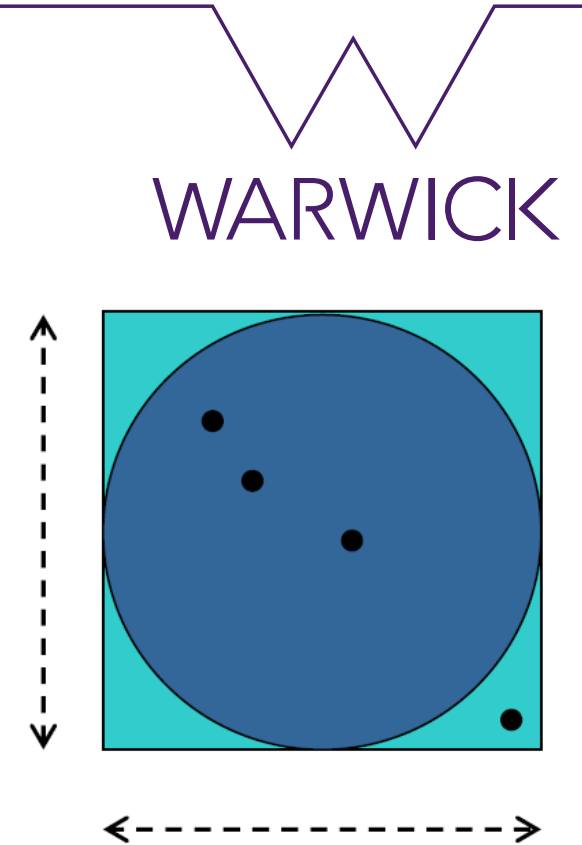
- $r = 1.0 \Rightarrow A_{square} = (2r)^2 = 4.0, A_{circle} = \pi r^2 = \pi$

- $\frac{A_{circle}}{A_{square}} = \frac{\pi}{4.0} \Rightarrow \pi = 4 \left(\frac{A_{circle}}{A_{square}} \right)$

- If we throw darts at random at an object, the percentage of darts that land inside it is proportional to its area relative.
- Throw darts at random within the square and calculate the percentage of darts that land inside the circle. This gives an approximate value of the ratio

$$\frac{A_{circle}}{A_{square}} = \frac{N_{circle}}{N_{square}}$$

- Use the above equation to calculate π



Throwing darts

- Generate coordinates using “random” module
 - $x = \text{random.random}$
 - $y = \text{random.random}$
- How would you check if (x, y) is inside the circle?
 - If $x^2 + y^2 < r^2$
- Throw a large number of darts and test what proportion lands inside. You can use it to estimate the value of π
- Further details in the lab session

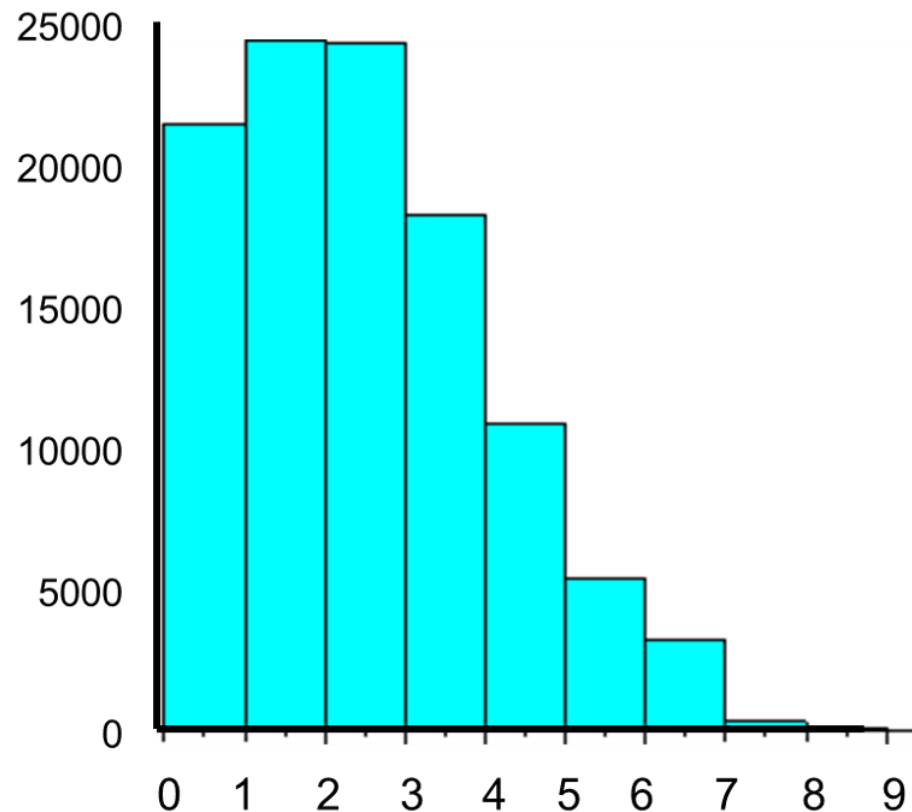
Monte Carlo Simulation

- What you did is a monte carlo simulation of estimating the value of π
- It can be used for arbitrary simulation of systems
- Can be used for integrating arbitrary functions
- Finding risks, robustness, etc.

https://en.wikipedia.org/wiki/Monte_Carlo_method

Central Limit Theorem

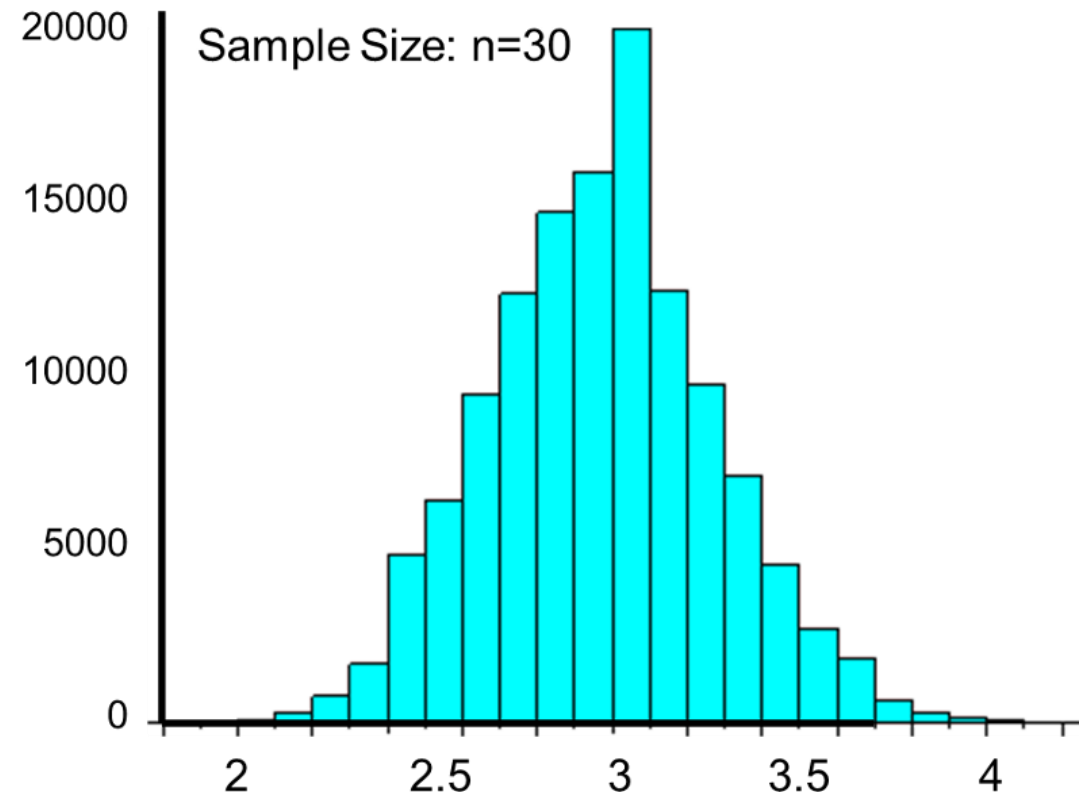
- Consider a Poisson distribution with $\mu = 3$ and $\sigma = 1.73$



https://sphweb.bumc.bu.edu/otlt/mph-modules/bs/bs704_probability/BS704_Probability12.html

Central Limit Theorem

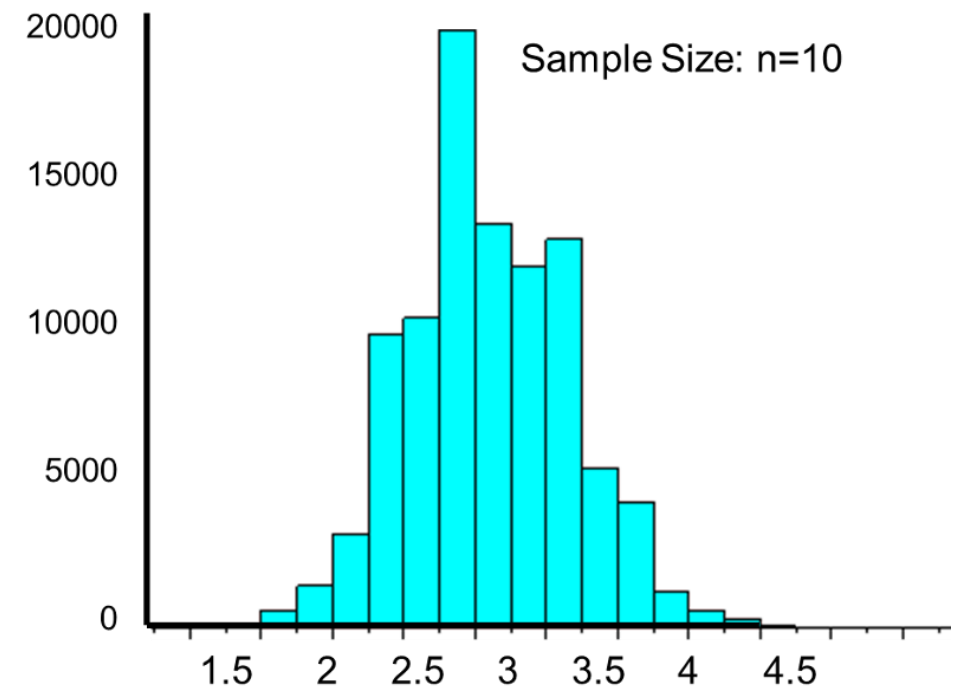
- This population is not normally distributed, but the Central Limit Theorem will apply if $n > 30$.
- If we take samples of size $n = 30$, we obtain samples distributed as shown in the graph with a $\mu = 3$ and $\sigma = 0.32$



https://sphweb.bumc.bu.edu/otlt/mph-modules/bs/bs704_probability/BS704_Probability12.html

Central Limit Theorem

- In contrast, with small samples of $n = 10$, we obtain samples distributed as shown in the graph.



https://sphweb.bumc.bu.edu/otlt/mph-modules/bs/bs704_probability/BS704_Probability12.html