

# Section 9: Uniform distribution, CDFs and sampling

STOR 155.02, Spring '21

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# What you will learn

- continuous distributions: uniform, normal
- CDFs and quantile functions
- sampling any r.v. using the quantile function and uniform r.v.

## Resources

- Textbook: ch 1.7, 3.5, 4.1

# Continuous distributions

# So far we've seen discrete probability distributions

'discrete' because

$$P(X = x_i) > 0$$

for distinct outcomes  $x_1 \dots$

which are often whole numbers

examples

$X$  is Bernoulli( $p$ ),  $x_1 = 1, x_2 = 0$

$X$  is Binomial( $N, p$ ),

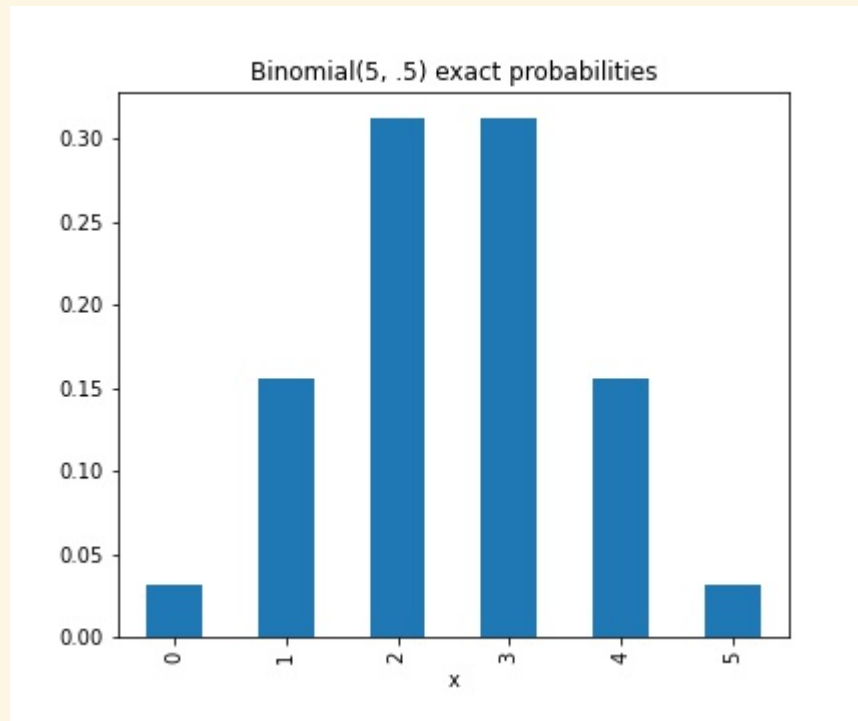
$$x_1 = 0, \dots, x_{N+1} = N$$

$X$  is Geometric( $p$ ),

$$x_1 = 1, x_2 = 2, \dots$$

Imagine this on a bar chart, with height as probability

$$P(X = k) = \binom{5}{k} p^k (1 - p)^{5-k}, \quad k = 0 \dots 5$$



# Continuous random variables

## key facts

takes a *range* of possible values

$$\text{e.g., } 0 \leq X \leq 1$$

$$\text{e.g., } -\infty < X < \infty$$

no gaps between values

$$P(X = x) = 0$$

for any particular value  $x$

instead have a **probability density function**  $p(x)$

$$\frac{P(x \leq X \leq x + \Delta x)}{\Delta x} \approx p(x)$$

for a small value  $\Delta x$ .

$$P(X \leq x)$$

= area under the curve  $p(y)$  for  $y \leq x$

$$P(a \leq X \leq b) = \int_a^b p(x) dx$$

if you remember calculus from high school.

# Uniform distribution

mother of all

# Uniform(0, 1) distribution

$p(x) = 1$  for  $0 \leq x \leq 1$  and 0 for all other  $x$ .

## Expectation, variance

$$E(X) = 1/2$$

$$Var(X) = 1/12$$

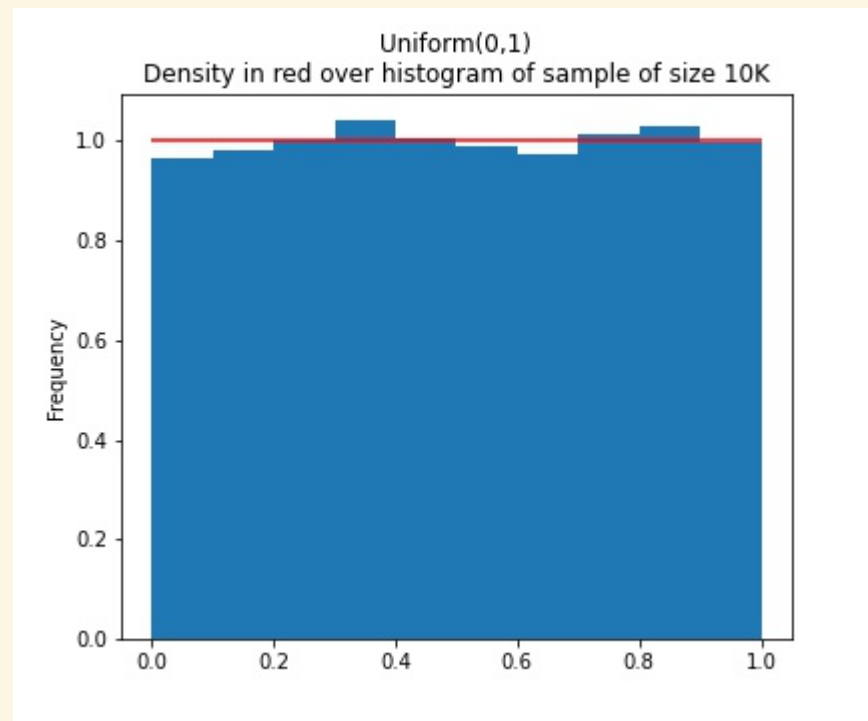
for those who have taken calculus

$$E(X) = \int_0^1 x dx, \quad Var(X) = \int_0^1 x^2 dx - \left( \int_0^1 x dx \right)^2$$



# Exact density vs. histogram

Q: Why does it make sense that  $E(X) = 1/2$ ?





# Uniform( $a, b$ ) for $a < b$

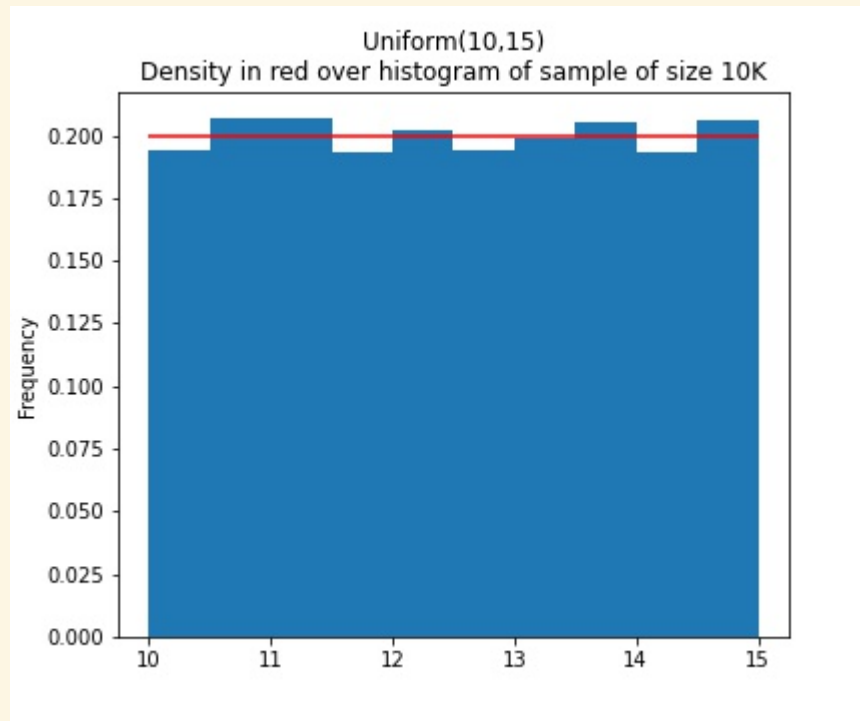
Density is a horizontal line between  $a$  and  $b$ .

$$p(x) = \frac{1}{b-a} \quad \text{for } a \leq x \leq b \text{ and } 0 \text{ for all other } x.$$

## Expectation, variance

$$E(X) = (a + b)/2 = \text{midpoint between } a, b$$

$$Var(X) = (b - a)^2/12$$



# Definition: Cumulative distribution function (CDF)

$$F_X(x) = P(X \leq x) = \text{CDF of } X \text{ at } x$$

## Discrete distributions

$$F_X(x_K) = \sum_{i=1}^K P(X = x_i)$$

sum of probabilities for possible outcomes  $x_1 \leq x_2 \leq \dots \leq x_K$  where  $x_K$  is one of the possible outcomes of  $X$ .

To calculate  $F_X(x)$  for  $x$  **between** possible values, use the next possible value below  $x$ .

## Continuous distributions

Area under the curve given by density function  $p(\cdot)$  to the left of  $x$

## CDF of Uniform(0, 1)

$$F_X(x) = P(X \leq x) = x$$

since this is the area of the rectangle of height 1, width  $x$ .

## CDF of Uniform(a, b)

$$F_X(x) = \frac{x - a}{b - a}$$

## CDF of Binomial(10, 1/2)

$$F_X(4) = \sum_{i=1}^4 \binom{10}{i} p^i (1-p)^{10-i}$$

**Use a computer!**

Calculating CDFs by hand is either tedious or not really possible

# Definition: Inverse CDF, aka quantile function

$F_X^{-1}(u)$  =  $u$ th theoretical quantile of distribution for  $X$

## Remember

When looking at data, the  $u$ th quantile is the

- value of a variable
- so that  $u$  proportion of observations
- are less than or equal to that value.

Same idea here

# Sample from any distribution!

$X$  is some random variable with CDF  $F_X$ .

$$U \sim \text{Uniform}(0, 1)$$

then

$F_X^{-1}(U)$  has the same distribution as  $X$

## how to use this

Sample independent  $\text{Uniform}(0, 1)$  random variables  $U_1 \dots U_N$ , then

$F_X^{-1}(U_1), \dots, F_X^{-1}(U_N)$  gives  $N$  independent samples of  $X$

**This is the best (only?) way to sample random variables in Excel**

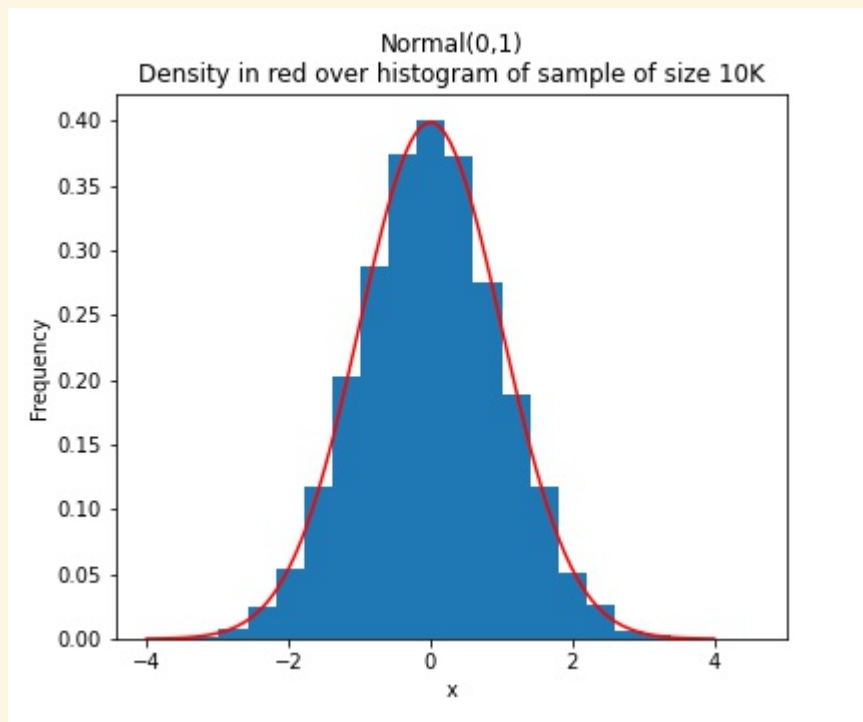


# Normal distribution

# Normal(0, 1) distribution

The famous **bell curve**

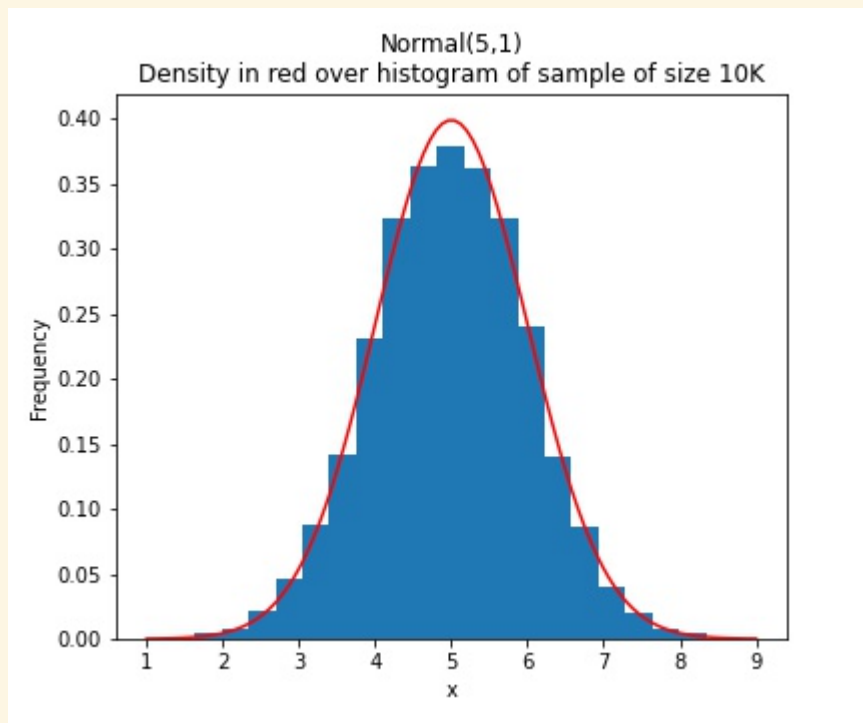
$$E(X) = 0, \quad \sqrt{\text{Var}(X)} = SD(X) = 1, \quad -\infty < X < \infty$$

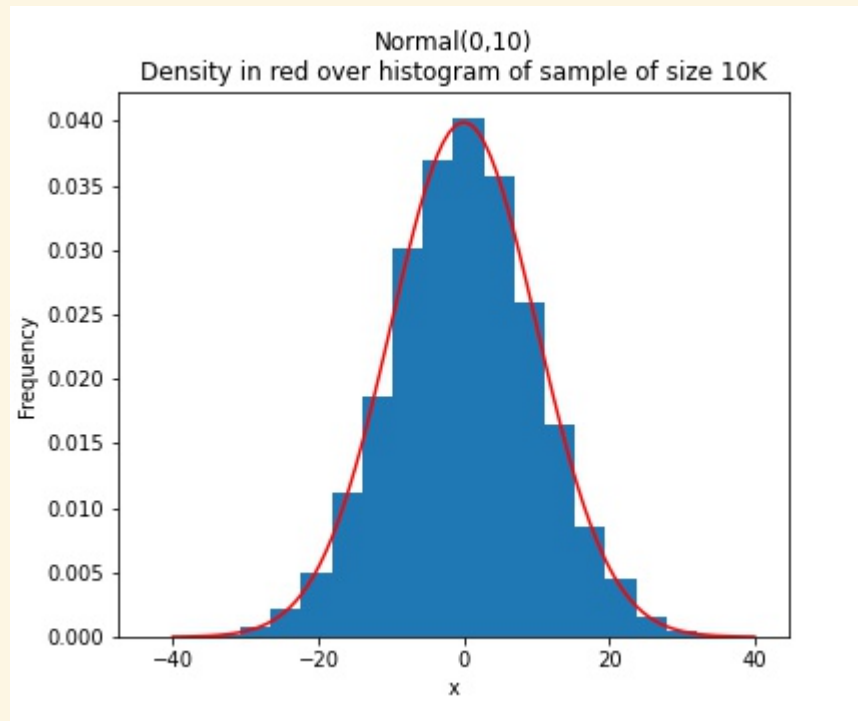


# Normal( $m$ , $s$ )

The **mean shifts the middle of the bell** and the **standard deviation determines how wide the bell is**.

$$E(X) = m, \quad SD(X) = s, \quad -\infty < X < \infty$$





# Convert Normal( $m$ , $s$ ) to Normal(0, 1)

If  $X$  is Normal( $m$ ,  $s$ ) then

$$Z = \frac{X - m}{s} \text{ is Normal}(0, 1)$$

**This is called a 'z-score'**

these will come back later in hypothesis testing

often used in statistics classes to calculate CDFs of Normal( $m$ ,  $s$ )

using a 'z-table'

But we won't do that in this class because ... mumble mumble ... computers exist

**PollEv.com/brendanbrown849**

**poll closes at** 

Five more minutes

