CS 107, Probability, Spring 2019 Lecture 23

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AUA

18 March 2019

Content

• Examples of Important Discrete R.V.s

Let us prove by Induction that all horses are the same color

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Figure: Same color horses $\ddot{\ }$

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By the induction assumption, horses in each set have the same color. Hence, all (n+1) horses will have the same color!



Discrete Uniform Distribution

Recall the definition of the Discrete Uniform Distribution:

Discrete Uniform Distribution

We say that the r.v. X has a Discrete Uniform Distribution with (over) the values $x_1, x_2, ..., x_n$ ($x_i \neq x_j$, $i \neq j$), and we will write $X \sim DiscreteUnif(x_1, ..., x_n)$, if

Values of
$$X \mid x_1 \mid x_2 \mid \dots \mid x_n$$

$$\mathbb{P}(X = x) \quad \frac{1}{n} \mid \frac{1}{n} \mid \dots \mid \frac{1}{n}$$

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$$\mathbb{P}(X = x) \qquad \frac{1}{n} \mid \frac{1}{n} \mid \dots \mid \frac{1}{n}$$

that is, $\mathbb{P}(X = x_k) = \frac{1}{n}$, for any k = 1, 2, ..., n.



Discrete Uniform Distribution: R Examples

R Code

```
#Discrete Uniform on data
data <- c(-1,2,4)
sample(data, size = 10, replace = T)

#Generating a sample of size 100
s <- sample(data, size = 100, replace = T)
#Calculating the number of -1 in the sample s
length(s[s == -1])</pre>
```

Bernoulli Distribution

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We will say that the r.v. X has a Bernoulli Distribution with the parameter (probability) $p \in [0,1]$, and we will write $X \sim Bernoulli(p)$, if it has the following PMF:

$$\begin{array}{c|cccc} \text{Values of } X & 0 & 1 \\ \hline \mathbb{P}(X = x) & 1 - p & p \\ \hline \end{array}$$

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that is, $\mathbb{P}(X=1)=p$, and $\mathbb{P}(X=0)=1-p$. The same can be written in the form (for $p \in (0,1)$):

$$\mathbb{P}(X = x) = p^{x} \cdot (1 - p)^{1-x}, \quad x \in \{0, 1\}.$$

^aTo avoid numbers like 0⁰.



Bernoulli Distribution models the binary outcomes case - Success/Failure, Yes/No, Boy/Girl, Pass/Fail, Healthy/III, Smoker/Non-Smoker, Flan/Fstan, Some examples:

• We are tossing a fair coin, and X=0 if Heads appears, and X=1 in the other case. Then $X \sim Bernoulli\left(\frac{1}{2}\right)$;

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- Say, X=1 in case the (randomly chosen Prob) student will pass the course, and X=0 otherwise. Say, we can model X as a Bernoulli r.v.: $X \sim Bernoulli(0.76)$;

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- Say, X=1 in case the (randomly chosen Prob) student will pass the course, and X=0 otherwise. Say, we can model X as a Bernoulli r.v.: $X \sim Bernoulli(0.76)$;
- We are interested if a specific student will get the A+ grade from Probability: we can take X=1 to indicate that he/she will get A+, and X=0 otherwise. Then $X \sim Bernoulli(p)$, and p depends on the student: if he/she is working well, then p is close to 1.

• We are interested if the patient has a specific disease or not: we can take, for example, X = 1, if he/she has that disease, and X = 0 otherwise. p will be the probability of having that disease, and $X \sim Bernoulli(p)$;

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- An insurance company is interested if the claim size for today will exceed 10Mln AMD: we can take X=1, if it will exceed and X=0 if not.
- We are interested if the max stock price for FB next month will be higher than the stock price today: we can take X=1, if that max price next month will be larger, and X=0 otherwise.

- We are interested if some specific Football team will win the next game: X=1 means will win, and X=0 otherwise. Then $X \sim Bernoulli(p)$;
- We are interested if some specific Football team will win the next game with the difference in the scores more than
 3: X = 1 if that will be the case, and X = 0 otherwise.
 Then X ~ Bernoulli(p);

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- And you can make infinitely many examples of Bernoulli distributed r.v.s

R Code

```
#Bernoulli Distribution p <- 0.3
#In fact, we use the Binomial Distr with size = 1 x <- rbinom(10, size = 1, prob = p)
```

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We will say that the r.v. X has a Binomial Distribution with the parameters n and p, $p \in [0,1]$, and we will write $X \sim Binom(n,p)$, if it has the following PMF:

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that is (in the case $p \in (0,1)$),

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Exercise: Check that the sum of probabilities is 1.



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Binomial Distribution gives the model for the repeated trials. If $X \sim Binom(n, p)$, then

- we are doing n independent repetitions of some Simple Experiment,
- in each Simple Experiment some Event can happen (Success) with the Probability p;
- X shows the number of Successes we will have during that n trials

Binomial Distribution: Examples

 We can toss a fair coin 12 times, and we are interested in how many heads we will get during that tosses. If we will denote by X the number of heads we will have, then X ~ Binom(12, 0.5).

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- We can toss a fair coin 12 times, and we are interested in how many heads we will get during that tosses. If we will denote by X the number of heads we will have, then X ~ Binom(12, 0.5).
- Assume we have 5 red and 10 green balls in the box. We are drawing a ball at random, fixing its color, returning it to the box, and then doing that again 100 times. We are interested in the number of green balls we will pick. If we will denote by X the number of green balls taken out of the box during this 100 trials, then $X \sim Binom(100, 2/3)$, since the probability to get a green ball in each trial is $p = \frac{10}{15} = \frac{2}{3}$ (we assume drawing each ball has the same probability for all balls).
- Your examples?



Binomial Distribution: R Examples

R Code

```
#Binomial Distribution
p < -0.45
#Size is our n
x \leftarrow rbinom(20, size = 3, prob = p)
Х
#Plotting the PMF
size = 21
x \leftarrow 0:size
PMF <- dbinom(x, size = size, prob = p)
plot(x, PMF, pch = 19)
```

Supplementary: Very short R Intro

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Note: Please run the commands below to see what they are doing $\ddot{\ }$

Basic Commands of R

Some assignments (The Sharp(Diez $\stackrel{\smile}{\smile}$, #) sign denotes that after it the whole line is a comment):

```
x <- 10 # x=10
x = 10 # the same as above, x=10
x
print(x)
y <- 20
z <- x+y
z
3 -> x
```

```
x = c(2,3,5,7,2,1) \# c means concatenation,
        \#so x is an array with elements 2,3,...
v = c(10, 15, 12)
z = c(x,y) concatenates, joins the arrays
X
x[2] #the second element in x
У
7.
z^2 #squares the elements of z
x*x #squares the elements of x
```

Try now:

```
x+\lambda
```

Can you guess what is doing **R** in this case?

Ways to create some regular-pattern vectors

```
x \leftarrow 1:5 #The same as x \leftarrow c(1,2,3,4,5)

x \leftarrow seq(from = 1, to = 10, by = 2)

x \leftarrow seq(from = 1, to = 10, length = 3)

y \leftarrow rep(x,2) #repeats (replicates) x = 2 times

y \leftarrow rep(x, each = 2)
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Operations with vectors:

Choosing elements of a vector:

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```
x \leftarrow c(2, -1, 9, 0, 4, 1)
x[3] #the third element of x
x[2:4] #elements no. 2,3,4, of x
x[c(2,4,5)] #elements no. 2,4,5 of x
x > 0
x[x > 0] #Selects all positive elements of x
x[-c(2,3)] #x, without elements no 2,3
x[-\{2:4\}] #x, without elements no 2,3,4
x[-(2:4)] #x, without elements no 2,3,4
x[seq(from = 2, to = length(x), by = 2)]
        # will give the even-indexed elements of x
```

```
x \leftarrow c(2, -1, 9, 0, -4, 1, 2)

x[x < 0] \leftarrow 0 #assigns 0 to all negative elements of x

which(x == 2) #gives the index of elements equal to 2

which(x > 0) #gives the indices of positive elements

which(x == max(x)) #returns the index (indices) of

#max element in x

which.max(x) #same thing as above

which.min(x)
```

Defining Functions

```
square = function(x) x^2
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The usage is:

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```
f <- function(x) {x+1}
f(4)
f <- function(x) {return(x+1)}
f(4)
f <- function(x,y) {x*y}
f(2,3)
f <- function(x,y=10) {x*y}
f(2,3); f(2)</pre>
```

Plotting 2D graphs

```
square = function(x) x^2
plot(square)
plot(square, -1,3)
plot(square, xlim = c(-2,3))
plot(square, lwd=3, xlim = c(-2,3))
plot(square, lwd=3, col="red", xlim = c(-2,3))
plot(square, lwd=3, xlim = c(-2,3), ylim = c(0,1))
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```

Another method is

```
curve(sin, -pi, pi)
curve(sin(x), -pi, 2*pi)
curve(sin, -pi, 2*pi, lwd = 4)
f <- function(x) {x+2*x^2-4*x^3}
curve(f, -2,2)</pre>
```

Vector Functions in R

```
x <- c(2,2,3,4,3,5,7,2,1)
max(x)
min(x)
range(x)
length(x)
sum(x)
prod(x)
cumsum(x)
cumprod(x)</pre>
```

The following names are used in \mathbf{R} for distributions:

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Distribution	R name	Distribution	R name
Beta	beta	Lognormal	Inorm
Binomial	binom	Negative Binomial	nbinom
Cauchy	cauchy	Normal	norm
Chisquare	chisq	Poisson	pois
Exponential	exp	Student t	t
F	f	Uniform	unif
Gamma	gamma	Tukey	tukey
Geometric	geom	Weibull	weib
Hypergeometric	hyper	Wilcoxon	wilcox
Logistic	logis		

Usage:

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d name()	PD(M)F of distribution	
p name()	CDF	
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Example:

```
rnorm(10)
dunif(1)
pexp(3, rate = 2)-pexp(-3, rate = 2)
qchisq(0.3, df = 1)
```

Note that the distributions above have some parameters (say, min and max for Uniform, mean and sd for Normal etc.), and some parameters have default values you can keep (say, for Normal Distribution one has mean=0 and sd=1).

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Plotting the Graph of Normal PDF and CDF

```
curve(dnorm, -4, 4, 1wd = 2) curve(pnorm, -4, 4, 1wd = 2)
```

```
curve(dnorm(x, 1,3), -4, 4, 1wd = 2)
```