

# Simulations

STAT 133

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How do we use a computer to  
simulate a chance process?

# Case Study 1

# Simulation Probability examples

## For instance

- ▶ Simulate flipping a coin
- ▶ Simulate rolling a die
- ▶ Simulate drawing a card from a deck
- ▶ Simulate a probability experiment with balls in an urn
- ▶ Simulate the “Monty Hall Problem”

# Flipping a Coin

# Simulating a Coin

How to simulate tossing a coin?



# Simulating a coin

One way to simulate a coin

```
coin <- c("heads", "tails")
```

# Simulating a coin

One way to simulate a coin

```
coin <- c("heads", "tails")
```

One way to simulate flipping a coin

```
sample(coin, size = 1)
```

```
## [1] "heads"
```



# Probability

Probability allows us to quantify statements about the chance of an event taking place

For example, flip a fair coin

- ▶ What's the chance it lands heads?
- ▶ Flip it 4 times, what proportion of heads do you expect?
- ▶ Will you get exactly that proportion?
- ▶ What happens when you flip the coin 1000 times?

# Simulating a Coin

## When you flip a coin

- ▶ it may land heads
- ▶ it may land tails
- ▶ with what probability it lands heads?
- ▶ If it is a fair coin:  $p = 0.5$

# Simulating a Coin

Tossing a coin can be modeled with a random variable following a Bernoulli distribution:

- ▶ heads ( $X = 1$ ) with probability  $p$
- ▶ tails ( $X = 0$ ) with probability  $q = 1 - p$

The Bernoulli distribution is a special case of the Binomial distribution:

$$B(1, p)$$

# Simulating tossing a coin

Tossing a coin simulated with a **binomial** distribution:

```
# binomial distribution generator  
rbinom(n = 1, size = 1, prob = 0.5)  
  
## [1] 1
```

# Flipping a Coin function

Function that simulates flipping a coin

```
# flipping coin function
coin <- function(prob = 0.5) {
  rbinom(n = 1, size = 1, prob = prob)
}

coin()

## [1] 0
```

# Flipping a Coin function

It's better if we assign labels

```
# flipping coin function
coin <- function(prob = 0.5) {
  out <- rbinom(n = 1, size = 1, prob = prob)
  ifelse(out, "heads", "tails")
}

coin()

## [1] "heads"
```

# Flipping a Coin function

```
# 10 flips
for (i in 1:10) {
  print(coin())
}
```

```
## [1] "tails"
## [1] "tails"
## [1] "tails"
## [1] "tails"
## [1] "heads"
## [1] "tails"
## [1] "heads"
## [1] "heads"
## [1] "heads"
## [1] "tails"
```

# 4 Flips

## In 4 flips

- ▶ Possible outputs:
  - HHHH, THHH, HTHH, HHTH, HHHT, ...
- ▶ we can get 0, 1, 2, 3, 4 heads
- ▶ so the proportion of heads can be: 0, 0.25, 0.5, 0.75, 1
- ▶ we expect the proportion to be 0.5
- ▶ but a proportion of 0.25 is also possible



## 4 Flips

- ▶ we can think of the proportion of Heads in 4 flips as a **statistic** because it summarizes data
- ▶ this proportion is a random quantity: it takes on 5 possible values, each with some probability
  - $0 \rightarrow 1/16$
  - $0.25 \rightarrow 4/16$
  - $0.50 \rightarrow 8/16$
  - $0.75 \rightarrow 4/16$
  - $1.0 \rightarrow 1/16$

# Simulating flipping $n$ coins

Function that simulates flipping a coin  $n$  times (i.e. flipping  $n$  coins)

```
# generic function
flip_coins <- function(n = 1, prob = 0.5) {
  out <- rbinom(n = n, size = 1, prob = prob)
  ifelse(out, "heads", "tails")
}

flip_coins(5)

## [1] "heads" "heads" "heads" "heads" "tails"
```

# Proportion of Heads

```
# number of heads
num_heads <- function(x) {
  sum(x == "heads")
}

# proportion of heads
prop_heads <- function(x) {
  num_heads(x) / length(x)
}
```

# 1000 Flips

- ▶ when we flip the coin 1000 times, we can get many different possible proportions of Heads
- ▶ 0, 0.001, 0.002, 0.003, ..., 0.999, 1.000
- ▶ It's highly unlikely that we would get 0 for the proportion—how unlikely?
- ▶ what does the distribution of the proportion of heads in 1000 flips look like?

# 1000 Flips

- ▶ With some probability theory and math tools we can figure this out
- ▶ But we can also get a good idea using simulation
- ▶ In our simulation we'll assume that the chance of Heads is 0.5 (i.e. fair coin)
- ▶ we can find out what the possible values for the proportion of heads in 1000 flips look like

# Flipping coins

```
set.seed(99900)
flips <- flip_coins(1000)

num_heads(flips)

## [1] 494

prop_heads(flips)

## [1] 0.494
```

# Flipping coins

```
set.seed(76547)
a_flips <- flip_coins(1000)
b_flips <- flip_coins(1000)

num_heads(a_flips)

## [1] 493

num_heads(b_flips)

## [1] 507
```

# 1000 Flips 1000 times

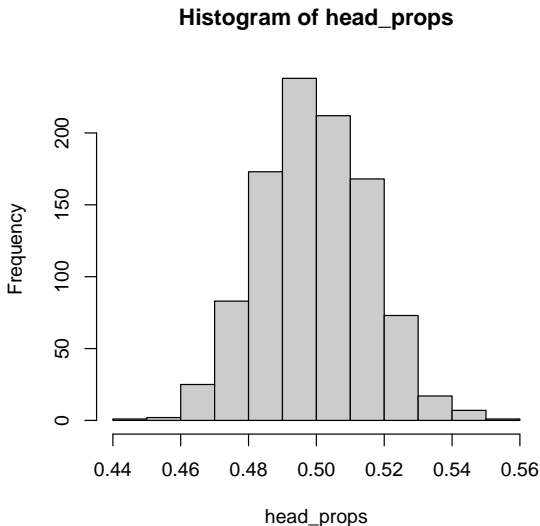
```
times <- 1000

head_props <- numeric(times)

for (s in 1:times) {
  flips <- flip_coins(1000)
  head_props[s] <- prop_heads(flips)
}
```



# Empirical distribution of 1000 flips



# Flipping coins

Experiment: flipping a coin 100 times and counting number of heads

You flip a coin 100 times and you get 65 heads. Is it a fair coin?

# Flipping coins

Experiment: flipping a coin 100 times and counting number of heads

You flip a coin 100 times and you get 65 heads. Is it a fair coin?

We could perform a hypothesis test, or we could perform resampling

# Flipping coins

```
# repeat experiment 100 times
times <- 10000
head_times <- numeric(times)
for (s in 1:times) {
  flips <- flip_coins(100)
  head_times[s] <- num_heads(flips)
}

sum(head_times >= 65)

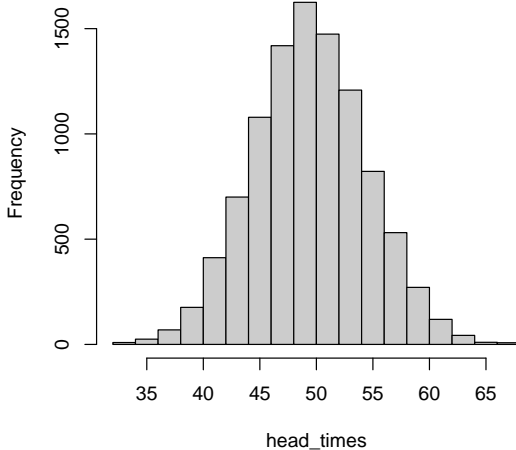
## [1] 18

sum(head_times >= 65) / times

## [1] 0.0018
```

# Flipping coins

**Histogram of head\_times**



# Case Study 2

# Chevalier de Mere

- ▶ Antoine Gombaud (1607 - 1684)
- ▶ *Nom de plume* “Chevalier de Mere”
- ▶ French writer and gambler, but not a nobleman
- ▶ Amateur mathematician





# De Mere's games

## Game 1

- ▶ One die
- ▶ Four rolls
- ▶ Win: at least one 6

# De Mere's games

## Game 1

- ▶ One die
- ▶ Four rolls
- ▶ Win: at least one 6

## Game 2

- ▶ Two dice
- ▶ 24 rolls
- ▶ Win: at least one double 6

# De Mere's games

## Game 1

- ▶ One die
- ▶ Four rolls
- ▶ Win: at least one 6

## Game 2

- ▶ Two dice
- ▶ 24 rolls
- ▶ Win: at least one double 6

De Mere was making money with game 1, but losing money with game 2. He turned to Blaise Pascal for help.

# De Mere's faulty reasoning

## Game 1

- ▶ The chance of getting a six in one roll of a die is  $\frac{1}{6}$ .
- ▶ In four rolls of a die, the chance of getting one six would be  $\frac{4}{6} = \frac{2}{3}$ .

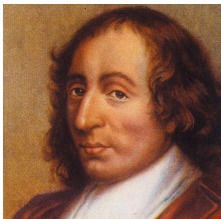
# De Mere's faulty reasoning

## Game 2

- ▶ The chance of getting a double six in rolling a pair of dice is  $\frac{1}{36}$ .
- ▶ In 24 rolls of a pair of dice, the chance of getting one double six would be  $\frac{24}{36} = \frac{2}{3}$ .

# Pascal and Fermat

- ▶ De Mere turned into Blaise Pascal
- ▶ Pascal consulted with Pierre de Fermat
- ▶ Beginning of Pascal's and Fermat's famous correspondence (1650's)
- ▶ Origin of combinatorial probability



Blaise Pascal  
(1623 - 1662)  
French mathematician



Pierre de Fermat  
(1601 - 1665)  
French lawyer and amateur  
mathematician

# Letter from Pascal to Fermat

*"If one undertakes to throw a six with one die, the advantage of undertaking it in 4 throws is as 671 to 625. If one undertakes to throw a double-six with two dice, there is a disadvantage of undertaking it in 24 throws. And nevertheless 24 is to 36 (which is the number of faces of two dice) as 4 to 6 (which is the number of faces of one die)."*



# De Mere's Game 1

Probability of no six in four rolls:

$$\frac{5}{6} \times \frac{5}{6} \times \frac{5}{6} \times \frac{5}{6} = \frac{625}{1296} = 0.482253$$

# De Mere's Game 1

Probability of no six in four rolls:

$$\frac{5}{6} \times \frac{5}{6} \times \frac{5}{6} \times \frac{5}{6} = \frac{625}{1296} = 0.482253$$

$P(\text{at least one six}) = 1 - \text{Probability of no six in four rolls:}$

$$1 - \frac{625}{1296} = \frac{671}{1296} = 0.517747$$

## De Mere's Game 2

Probability of no double six in 24 rolls:

$$P(\text{no double six in 24 rolls}) = \left(\frac{35}{36}\right)^{24} = 0.5086$$

## De Mere's Game 2

Probability of no double six in 24 rolls:

$$P(\text{no double six in 24 rolls}) = \left(\frac{35}{36}\right)^{24} = 0.5086$$

$P(\text{at least one double six in 24 rolls}) =$

$$1 - 0.5086 = 0.4914$$

Using a computer to simulate a  
chance process

# Some Questions

- ▶ Rather than solve the problem analytically, we can simulate 4 rolls of a die and count the number of 6's
- ▶ If we simulate rolling 4 dice many times, then the proportion of times we get 0, 1, 2, 3, or 4 sixes should be close to the chance of that many sixes on any 4 rolls

# Some Questions

- ▶ What is the chance of getting one 6 when rolling a die?
- ▶ What is the chance of getting one 6 when rolling two dice?
- ▶ What is the chance of getting at least one 6 when rolling 4 dice?

# What are the steps?

- ▶ Simulate rolling one die
- ▶ Simulate rolling a pair of dice
- ▶ Simulate rolling four dice
- ▶ Count the number of sixes



# Simulating one die



- ▶ Let's start with one die
- ▶ What features?
- ▶ How to create a **die** object?

# One die

```
die <- 1:6
```

```
die
```

```
## [1] 1 2 3 4 5 6
```

# Your turn

Which option would produce an invalid die:

*# A*

```
die <- 1L:6L
```

*# B*

```
die <- seq(from = 1, to = 6)
```

*# C*

```
die <- c(1, 2, 3, 4, 5, 6)
```

*# D*

```
die <- seq(from = 1, to = 6, by = 6)
```

*# E*

```
die <- seq_len(6)
```

How to simulate the roll of a die?

# sample()

Very useful function for selecting from a discrete set (vector) of possibilities.

## sample() arguments

- ▶ x
- ▶ size
- ▶ replace
- ▶ prob

# Roll a die

```
# use of sample() to simulate the roll of a die  
sample(die, size = 1)  
  
## [1] 4
```

# Roll a die

```
# use of sample() to simulate the roll of a die
```

```
sample(die, size = 1)
```

```
## [1] 4
```

```
sample(die, size = 1)
```

```
## [1] 2
```

```
sample(die, size = 1)
```

```
## [1] 5
```

```
sample(die, size = 1)
```

```
## [1] 6
```



# Function: Roll one die

Write a function to make it more convenient

```
# function  
rolldie <- function() {  
  
}
```

# Function: Roll one die

Write a function to make it more convenient

```
# function  
rolldie <- function() {  
  die <- 1:6  
  sample(die, size = 1)  
}
```

# Function: Roll one die

Write a function to make it more convenient

```
# function  
rolldie <- function() {  
  die <- 1:6  
  sample(die, size = 1)  
}
```

```
rolldie()
```

```
## [1] 2
```

# Rolling one die

Let's simulate 10 rolls of a die

```
# roll 10 times  
for (i in 1:10) {  
  rolldie()  
}
```

# Rolling one die

Let's simulate 10 rolls of a die

```
# roll 10 times  
for (i in 1:10) {  
  rolldie()  
}
```

What's *wrong*?

Why nothing is shown on screen?

# Rolling one die

```
# roll 10 times  
for (i in 1:10) {  
  print(rolldie())  
}
```

```
## [1] 2  
## [1] 3  
## [1] 4  
## [1] 6  
## [1] 2  
## [1] 6  
## [1] 6  
## [1] 4  
## [1] 4  
## [1] 1
```

# Rolling one die

```
# roll until first 5  
repeat {  
  rol <- rolldie()  
  print(rol)  
  if (rol == 5) break  
}
```

```
## [1] 4  
## [1] 2  
## [1] 6  
## [1] 4  
## [1] 3  
## [1] 6  
## [1] 5
```

# Rolling one die

```
# roll until first 5  
rol <- 1
```

```
while (rol != 5) {  
  rol <- rolldie()  
  print(rol)  
}
```

```
## [1] 4  
## [1] 2  
## [1] 6  
## [1] 4  
## [1] 3  
## [1] 6  
## [1] 5
```



# Rolling one die

```
# roll 10,000 times
results <- numeric(10000)

for (i in 1:10000) {
  results[i] <- rolldie()
}
```

# Rolling one die

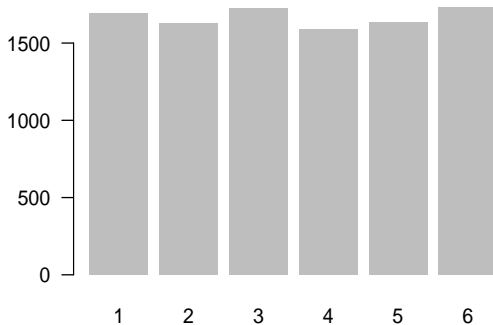
```
# frequencies  
table(results)
```

```
## results  
##      1      2      3      4      5      6  
## 1692 1631 1724 1586 1633 1734
```

```
# relative frequencies  
table(results) / 10000
```

```
## results  
##      1      2      3      4      5      6  
## 0.1692 0.1631 0.1724 0.1586 0.1633 0.1734
```

# Distribution of results



How to simulate a **loaded** die?

# Loaded Die

Changing argument **prob** to create a loaded die

```
probs <- c(1/21, 2/21, 3/21, 4/21, 5/21, 6/21)

# function
loaded <- function() {
  die <- 1:6
  sample(die, size = 1, prob = probs)
}
```

# Loaded Die

Changing argument **prob** to create a loaded die

```
probs <- c(1/21, 2/21, 3/21, 4/21, 5/21, 6/21)
```

```
# function
```

```
loaded <- function() {  
  die <- 1:6  
  sample(die, size = 1, prob = probs)  
}
```

```
loaded()
```

```
## [1] 6
```

# Rolling a loaded die

```
# roll 10 times  
for (i in 1:10) {  
  print(loaded())  
}
```

```
## [1] 6  
## [1] 5  
## [1] 4  
## [1] 2  
## [1] 6  
## [1] 2  
## [1] 2  
## [1] 4  
## [1] 4  
## [1] 6
```

# Rolling a loaded die

```
# roll 10,000 times
results <- numeric(10000)

for (i in 1:10000) {
  results[i] <- loaded()
}
```



# Rolling a loaded die

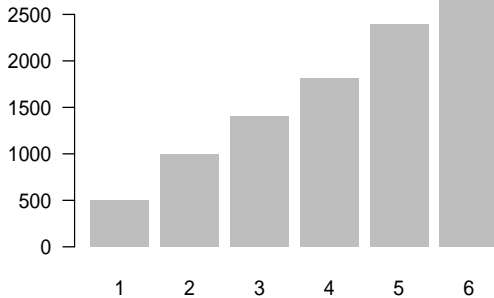
```
# frequencies  
table(results)
```

```
## results  
##      1      2      3      4      5      6  
## 505 1000 1405 1817 2391 2882
```

```
# relative frequencies  
table(results) / 10000
```

```
## results  
##      1      2      3      4      5      6  
## 0.0505 0.1000 0.1405 0.1817 0.2391 0.2882
```

# Loaded Distribution



# Simulating rolling a pair of dice



# Roll a pair of dice

```
# 1st die  
sample(die, size = 1)
```

```
## [1] 1
```

```
# 2nd die  
sample(die, size = 1)
```

```
## [1] 5
```

## Function: roll a pair of dice

```
# pair of dice function  
roll2 <- function() {  
  
  
  
  
  
  
  
  
  
}
```

# Function: roll a pair of dice

```
# pair of dice function
roll2 <- function() {
  die <- 1:6
  rol1 <- sample(die, size = 1)
  rol2 <- sample(die, size = 1)
  c(rol1, rol2)
}
```

# Various rolls

```
roll2()
```

```
## [1] 5 1
```

```
roll2()
```

```
## [1] 1 1
```

```
roll2()
```

```
## [1] 6 6
```



# Function

```
# pair of dice function
roll2 <- function() {
  die <- 1:6
  rol1 <- sample(die, size = 1)
  rol2 <- sample(die, size = 1) # repeated command!
  c(rol1, rol2)
}
```

# Roll a pair of dice

```
die <- 1:6  
  
# avoid repetition with one call of 'sample()'  
sample(die, size = 2)  
  
## [1] 6 1
```

# Roll a pair dice

```
for (i in 1:15) {  
  print(sample(die, size = 2))  
}
```

```
## [1] 1 3  
## [1] 1 4  
## [1] 5 4  
## [1] 4 3  
## [1] 3 1  
## [1] 2 6  
## [1] 4 6  
## [1] 3 4  
## [1] 4 6  
## [1] 6 3  
## [1] 2 5  
## [1] 2 5  
## [1] 2 3  
## [1] 3 2  
## [1] 2 5
```

*# Can you spot a problem?*

# Roll a pair of dice

```
die <- 1:6  
  
# sample with replacement  
sample(die, size = 2, replace = TRUE)  
  
## [1] 4 1
```

# Roll a pair of dice

```
for (i in 1:15) {  
  print(sample(die, size = 2, replace = TRUE))  
}
```

```
## [1] 6 6  
## [1] 6 5  
## [1] 5 4  
## [1] 3 6  
## [1] 1 2  
## [1] 1 5  
## [1] 1 2  
## [1] 3 1  
## [1] 6 1  
## [1] 1 2  
## [1] 1 6  
## [1] 3 2  
## [1] 3 1  
## [1] 2 2  
## [1] 1 1
```

# Roll a pair of dice

```
# rewrite roll2()  
roll2 <- function() {  
  die <- 1:6  
  sample(die, size = 2, replace = TRUE)  
}
```

# Roll a pair of dice

```
# rewrite roll2()  
roll2 <- function() {  
  die <- 1:6  
  sample(die, size = 2, replace = TRUE)  
}
```

```
roll2()
```

```
## [1] 5 4
```

Rolling a die any number of  
times



# Rolling several dice

More general function to roll a die any number of times

```
roll <- function(times = 1) {  
  die <- 1:6  
  sample(die, size = times, replace = TRUE)  
}
```

# Rolling several dice

More general function to roll a die any number of times

```
roll <- function(times = 1) {  
  die <- 1:6  
  sample(die, size = times, replace = TRUE)  
}
```

```
roll()
```

```
## [1] 4
```

# Rolling several dice

```
# default (one roll)
```

```
roll()
```

```
## [1] 4
```

```
# two rolls
```

```
roll(2)
```

```
## [1] 1 5
```

```
# 4 rolls
```

```
roll(4)
```

```
## [1] 6 4 6 5
```

# De Mere's Game 1

# De Mere's Game 1

## Game 1

- ▶ One die
- ▶ Four rolls
- ▶ Win: at least one 6

# De Mere's Game 1

```
# play 100 times
results <- matrix(0, nrow = 100, ncol = 4)

for (i in 1:100) {
  results[i, ] <- roll(times = 4)
}

head(results)
```

##	[,1]	[,2]	[,3]	[,4]
## [1,]	2	5	4	2
## [2,]	6	6	1	6
## [3,]	3	4	4	2
## [4,]	5	2	3	6
## [5,]	6	2	3	1
## [6,]	4	3	6	1

# De Mere's Game 1

```
counts <- 0

for (i in 1:100) {
  if (any(results[i, ] == 6))
    counts <- counts + 1
}

# proportion of wins
counts / 100

## [1] 0.56
```

# De Mere's Game 1

```
sixes <- apply(results, 1, function(x) sum(x == 6))  
  
table(sixes)  
  
## sizes  
## 0  1  2  3  
## 44 41 13  2
```



# De Mere's Game 1

```
sixes <- apply(results, 1, function(x) sum(x == 6))
```

```
table(sixes)
```

```
## sixes
```

```
## 0 1 2 3
```

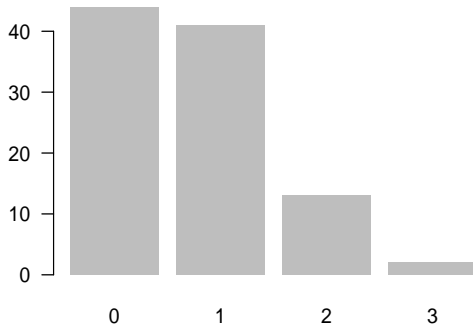
```
## 44 41 13 2
```

```
# rolls with at least one six
```

```
sum(table(sixes)[-1])
```

```
## [1] 56
```

# De Mere's Game 1



# De Mere's Game 1

## Considerations

- ▶ How would you make the code more flexible?
- ▶ What type of “parameters”?
- ▶ Avoid Repetition

# De Mere's Game 1

```
games <- 10000
results <- matrix(0, nrow = games, ncol = 4)

for (i in 1:games) {
  results[i, ] <- roll(times = 4)
}

counts <- 0
for (i in 1:games) {
  if (any(results[i, ] == 6))
    counts <- counts + 1
}

counts / games  # proportion of wins

## [1] 0.5204
```

# De Mere's Game 1

```
sixes <- apply(results, 1, function(x) sum(x == 6))
```

```
table(sixes)
```

```
## sizes
```

```
##      0      1      2      3      4
```

```
## 4796 3858 1171  166      9
```

# De Mere's Game 1

```
sixes <- apply(results, 1, function(x) sum(x == 6))
```

```
table(sixes)
```

```
## sixes
```

```
##      0      1      2      3      4
```

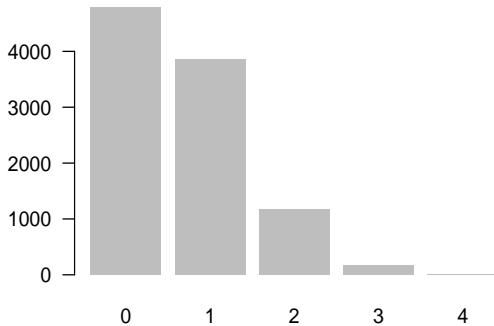
```
## 4796 3858 1171  166    9
```

```
# rolls with at least one six
```

```
sum(table(sixes)[-1])
```

```
## [1] 5204
```

# De Mere's Game 1



# De Mere's Game 2



# De Mere's Game 2

## Game 2

- ▶ Two dice
- ▶ 24 rolls
- ▶ Win: at least one double 6

## De Mere's Game 2

```
roll2 <- function(times = 1) {  
  dice2 <- unlist(lapply(1:6, function(x) x + 1:6))  
  sample(dice2, size = times, replace = TRUE)  
}
```

```
roll2()
```

```
## [1] 10
```

```
roll2(24)
```

```
## [1] 8 4 4 3 8 6 7 6 9 9 5 4 5 7 3 11 9 11 10
```

# De Mere's Game 2

## Game 2

- ▶ It's better if we sum the points of rolling two dice
- ▶ Possible outcomes:  $\{2, 3, 4, \dots, 10, 11, 12\}$
- ▶ Double six is equivalent to 12 points

## De Mere's Game 2

```
games <- 10000
results <- matrix(0, nrow = games, ncol = 24)

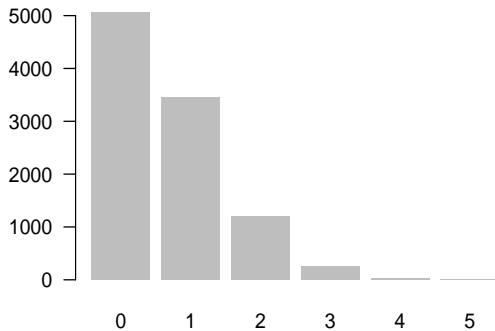
for (i in 1:games) {
  results[i, ] <- roll2(24)
}

doublesix <- apply(results, 1, function(x) sum(x == 12))

sum(table(doublesix)[-1])

## [1] 4930
```

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```
counts <- 0
for (i in 1:games) {
  if (any(results[i, ] == 12))
    counts <- counts + 1
}

counts / games  # proportion of wins

## [1] 0.493
```