### 05 - Normal distribution

HCI/PSYCH 522 Iowa State University

February 8, 2022

### Overview

- Normal distribution
  - Numerical data
- Inference for means
  - Estimating 1 mean
  - Comparing 2 means
  - Comparing 3+ means

### Normal

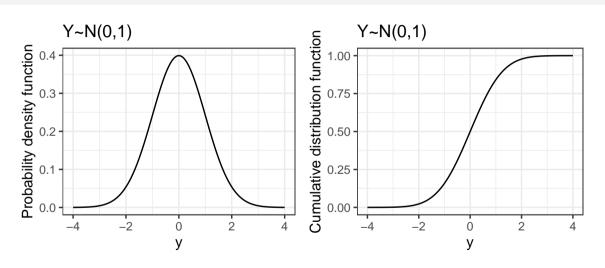
We typically model numerical data with a normal distribution. If  $Y \sim N(\mu, \sigma^2)$ , then

- the expected value  $E[Y] = \mu$ ,
- variance  $Var[Y] = \sigma^2$ ,
- standard deviation  $SD[Y] = \sigma$ ,
- probability density function (bell-shaped curve)

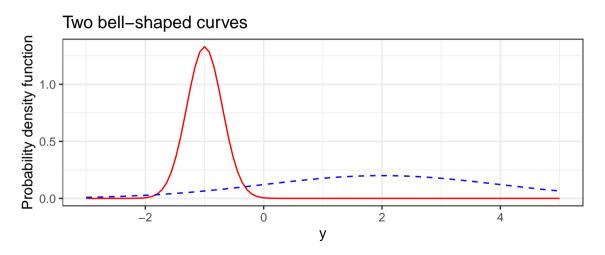
$$f(y) = (2\pi\sigma^2)^{-1/2} \exp\left(-\frac{1}{2\sigma^2}(y-\mu)^2\right),$$

• cumulative distribution function  $P(Y \leq y)$ .

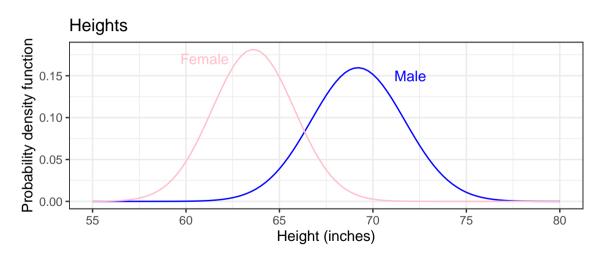
### Normal



## Normal

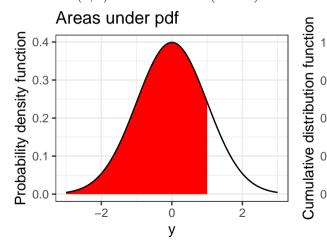


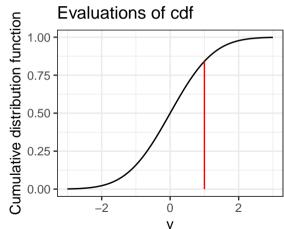
# Heights



### **Probabilities**

Let  $Y \sim N(0,1)$  and calculate P(Y < 1).

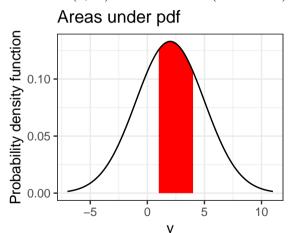


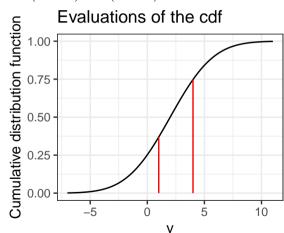


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### **Probabilities**

Let  $Y \sim N(2, 3^2)$  and calculate P(1 < Y < 4) = P(Y < 4) - P(Y < 1).





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### Probabilities in R

Let  $Y \sim N(-3, 4^2)$ .

$$mn < -3$$
 $s < -4$ 

Calculate P(Y < 0).

$$pnorm(0, mean = -3, sd = 4)$$

## [1] 0.7733726

Calculate P(Y > 1).

$$1-pnorm(1, mean = -3, sd = 4)$$

## [1] 0.1586553

### Probabilities in R

Let 
$$Y \sim N(-3, 4^2)$$
.

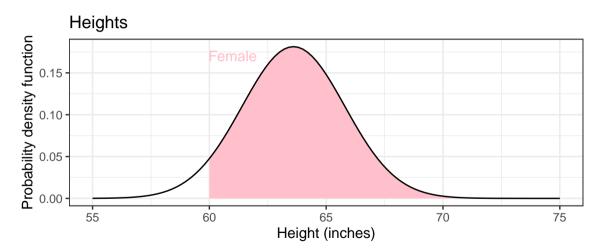
$$mn < -3$$

Calculate 
$$P(0 < Y < 1) = P(Y < 1) - P(Y < 0)$$
.

For continuous random variables, e.g. normal, P(Y=y)=0 for any value y. This is NOT true for discrete random variables, e.g. binomial.

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# Probability female height is above 60 inches?



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## Probability female height is above 60 inches?

Let  $Y \sim N(63.6, 2.2^2)$ . Calculate P(Y > 60).

```
1-pnorm(60, mean = 63.6, sd = 2.2)
```

## [1] 0.9491182

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## Estimating 1 mean

#### Suppose we have

- n numerical observations,
- ullet with the same population mean  $\mu$  and
- population standard deviation  $\sigma$ , and
- observations are independent.

Let  $Y_i$  be the value for the *i*th observation and assume  $Y_i \stackrel{ind}{\sim} N(\mu, \sigma^2)$ .

The sample can be summarized by the sample mean

$$\overline{Y} = \frac{Y_1 + Y_2 + \dots + Y_n}{n}$$

and sample variance

$$S^{2} = \frac{(Y_{1} - \overline{Y})^{2} + (Y_{2} - \overline{Y})^{2} + \dots + (Y_{n} - \overline{Y})^{2}}{n - 1}$$

(or the sample standard deviation  $S = \sqrt{S^2}$ .)

# Sample statistics in R

```
heights \leftarrow c(66.9, 63.2, 58.7, 64.2, 65.1)
length(heights) # number of observations
## [1] 5
mean(heights) # sample mean
## [1] 63.62
var(heights) # sample variance
## [1] 9.417
sd(heights) # sample standard deviation
## [1] 3.068713
```

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#### Parameter estimation

If we assume  $Y_i \overset{ind}{\sim} N(\mu, \sigma^2)$ , then we can use these sample statistics to estimate population parameters:

- $\hat{\mu} = \overline{Y}$ ,
- $\hat{\sigma} = S$ , and
- $\hat{\sigma}^2 = S^2$ .

Please remember that sample statistics are only estimates (not the true values).

# Posterior belief about population mean

Our posterior belief about the population mean is

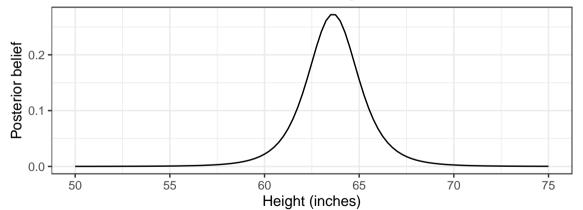
$$\mu|y \sim t_{n-1}(\overline{y}, s^2/n)$$

where

- $y = (y_1, \ldots, y_n)$  is the data,
- n is the sample size,
- ullet  $\overline{y}$  is the sample mean,
- $\bullet$   $s^2$  is the sample variance, and
- $t_{n-1}(\overline{y}, s^2/n)$  is a T distribution with
  - n-1 degrees of freedom,
  - location  $\overline{y}$ , and
  - scale s.

# Posterior belief about female mean height

## Posterior belief about female mean height



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### Credible interval in R

```
t.test(heights, conf.level = 0.95)
##
##
   One Sample t-test
##
## data: heights
## t = 46.358, df = 4, p-value = 1.295e-06
## alternative hypothesis: true mean is not equal to 0
## 95 percent confidence interval:
   59.80969 67.43031
##
## sample estimates:
## mean of x
##
       63.62
```

## Calculating posterior probabilities

What is our belief that mean female height is greater than 60 inches?

$$P(\mu > 60|y)$$

```
1-pt((60-mean(heights))/(sd(heights)/sqrt(length(heights))), df = length(heights)-1)
## [1] 0.9711426
```

or

```
plst <- function(q, df, location, scale) { # location-scale t distribution
  pt( (q-location)/scale, df = df)
}
1-plst(60, df = length(heights)-1, location = mean(heights), scale = sd(heights)/sqrt(length(heights)))
## [1] 0.9711426</pre>
```

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## Comparing 2 means

Suppose we have groups indexed by  $g = 1, \dots, G$ 

- ullet  $n_g$  numerical observations in group g,
- ullet the same population mean  $\mu_g$  within a group and
- ullet same population standard deviation  $\sigma_g$  within a group,
- all observations are independent.

Let  $Y_{ig}$  be the value for the *i*th observation in the *g*th group and assume  $Y_{ig} \stackrel{ind}{\sim} N(\mu_g, \sigma_g^2)$ .

When we collect data, we will have a sample mean and sample standard deviation for each group.

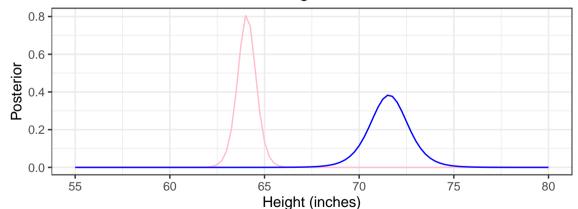
## Sample statistics in R

```
d <- read_csv("heights.csv")</pre>
d %>%
 group_by(sex) %>%
  summarize(n = n(),
           mean = mean(height),
           sd = sd(height))
## # A tibble: 2 \times 4
##
  sex n mean
                       sd
##
   <chr> <int> <dbl> <dbl>
## 1 female 11 64.1 1.59
## 2 male 7 71.6 2.66
```

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### Posterior beliefs

## Posterior beliefs about mean height



## Posterior probabilities

What is the probability that males are, on average, taller than females?

$$P(\mu_{\mathsf{male}} > \mu_{\mathsf{female}}|y)$$

We use a Monte Carlo approach

```
rlst <- function(n, df, location, scale) {
   location+scale*rt(n, df = df)
}
n_reps <- 100000
mu_female <- rlst(n_reps, df = 11-1, location = 64.1, scale = 1.59/sqrt(11))
mu_male <- rlst(n_reps, df = 7-1, location = 71.6, scale = 2.66/sqrt(7))
mean(mu_male > mu_female)
## [1] 0.99981
```

#### Credible interval for the difference

```
a <- 1 - 0.95
quantile(mu_male - mu_female, prob = c(a/2, 1-a/2))

## 2.5% 97.5%
## 4.822371 10.161489
```

# Using built in R functions

```
d <- read_csv("heights.csv")</pre>
t.test(height ~ sex, data = d)
##
##
   Welch Two Sample t-test
##
## data: height by sex
## t = -6.7492, df = 8.7839, p-value = 9.392e-05
## alternative hypothesis: true difference in means is not equal to 0
## 95 percent confidence interval:
## -10.033670 -4.981915
## sample estimates:
## mean in group female mean in group male
##
               64.06364
                                    71.57143
```