

# Statistics Review – Part 2

## Sample mean

$$\bar{y} = \frac{1}{n} \sum_{i=1}^n y_i \quad (3.1)$$

- From heights example:  $\bar{y} = 174.1$ ,  $\mu_y = 176.8$
- The sample (the  $y_i$ ) were drawn randomly
- $y$  is random  $\rightarrow \bar{y}$  is a random variable!
- $\bar{y}$  has a *sampling distribution* (probability function)
- The mean and variance of  $\bar{y}$  will determine if  $\bar{y}$  is:

- Unbiased
- Efficient
- Consistent

$$\begin{aligned}\mathbf{E} [\bar{y}] &= \mathbf{E} \left[ \frac{1}{n} \sum_{i=1}^n y_i \right] \\ &= \frac{1}{n} \mathbf{E} \left[ \sum_{i=1}^n y_i \right] \\ &= \frac{1}{n} \mathbf{E} [y_1 + y_2 + \cdots + y_n] \\ &= \frac{1}{n} (\mathbf{E} [y_1] + \mathbf{E} [y_2] + \cdots + \mathbf{E} [y_n]) \\ &= \frac{1}{n} (\mu_y + \mu_y + \cdots + \mu_y) \\ &= \frac{n\mu_y}{n} = \mu_y\end{aligned}\tag{3.2}$$

### 3.2.4 Efficiency

An estimator is efficient if it has the smallest variance among all other potential estimators (for us, potential = linear, unbiased)

Need to get the variance of  $\bar{y}$ .

$$\begin{aligned}
\text{Var} [\bar{y}] &= \text{Var} \left[ \frac{1}{n} \sum_{i=1}^n y_i \right] \\
&= \frac{1}{n^2} \text{Var} \left[ \sum_{i=1}^n y_i \right] \\
&= \frac{1}{n^2} \text{Var} [y_1 + y_2 + \cdots + y_n] \\
&= \frac{1}{n^2} (\text{Var} [y_1] + \text{Var} [y_2] + \cdots + \text{Var} [y_n]) \\
&= \frac{1}{n} (\sigma_y^2 + \sigma_y^2 + \cdots + \sigma_y^2) \\
&= \frac{n\sigma_y^2}{n^2} = \frac{\sigma_y^2}{n}
\end{aligned} \tag{3.3}$$

- Gauss-Markov theorem proves this is minimum variance
- We'll also need this to prove consistency, and for hyp. testing

### 3.2.5 Consistency

Suppose we had a lot of information. ( $n \rightarrow \infty$ )

What value should we get for our estimator?

How would state this mathematically?

Q) Prove that the sample mean is a consistent estimator for the population mean.

Q) Define the terms unbiasedness, efficiency, and consistency.

### 3.3 Hypothesis tests (known $\sigma_y^2$ )

$$H_0 : \mu_y = \mu_{y,0} \tag{3.4}$$

$$H_A : \mu_y \neq \mu_{y,0}$$

- Estimate  $\mu_y$  (using  $\bar{y}$  for example)
- See if  $\bar{y}$  appears “close” to  $\mu_{y,0}$ 
  - Remember,  $\bar{y}$  is random! (and Normal)
- If it’s close  $\rightarrow$  fail to reject
- If it’s far  $\rightarrow$  reject

Example:

- Hypothesize that mean height of a U of M student is 173cm

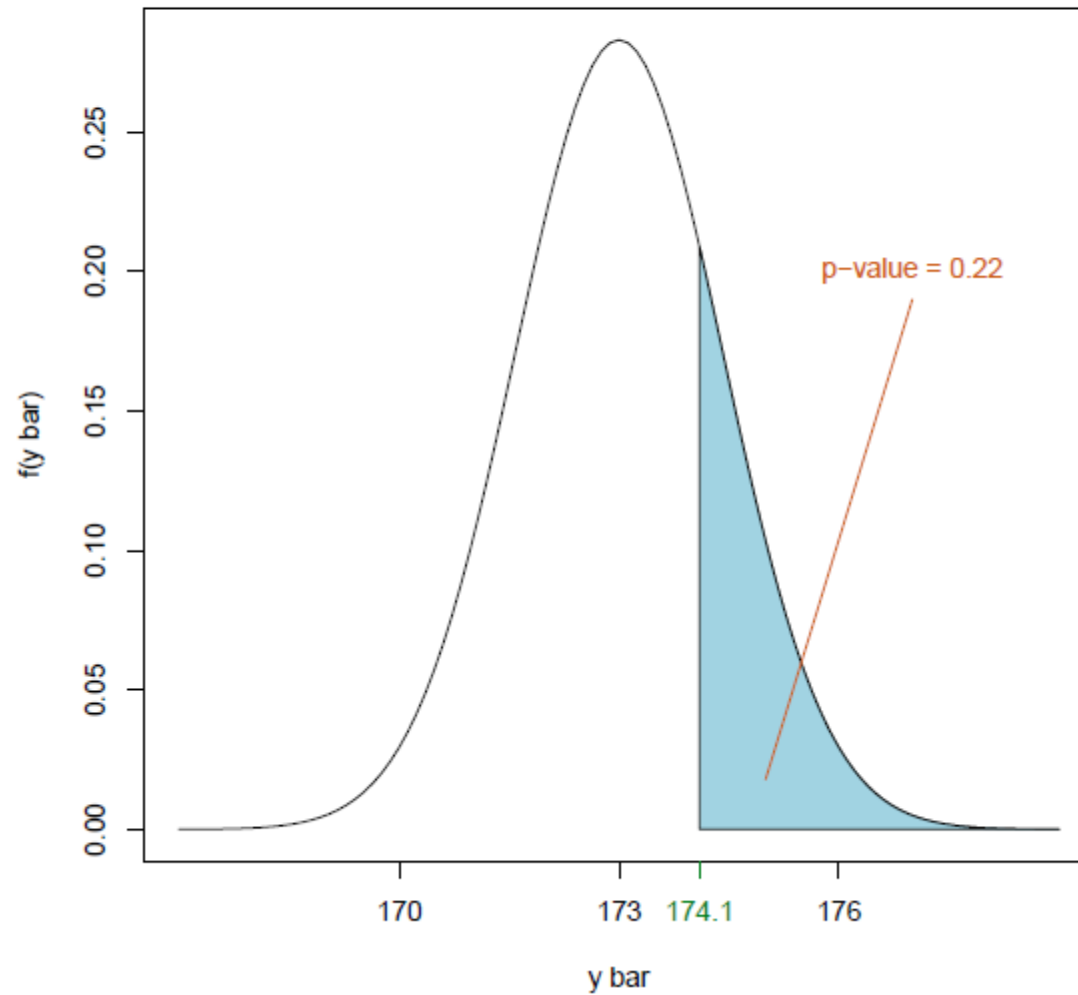
$$H_0 : \mu_y = 173 \tag{3.5}$$

$$H_A : \mu_y \neq 173$$

- Collect a sample:  $y = \{173.9, 171.7, \dots, 172.0\}$
- Calculate  $\bar{y} = 174.1$
- Suppose (very unrealistically) that  $\sigma_y^2 = 39.7$
- What now?



Figure 3.2: Normal distribution with  $\mu = 173$  and  $\sigma^2 = 39.7/20$ . Shaded area is the probability that the normal variable is greater than 174.1.



The p-value for the above test is 0.44. How to interpret this?

3.3.1 Significance of a test

3.3.2 Type I error

3.3.3 Type II error (and power)

### 3.3.4 Test statistics

- Just a more convenient way of getting the p-value for the test
- Each hypothesis test would present us with a new normal curve that we would have to draw, and calculate a new area (see fig. 3.2)
- Instead: *standardize*
- This gives us *one curve for all testing problems* (the standard normal curve)
- Calculate a bunch of areas under the curve, and tabulate them
- Not an issue with modern computers, but this is still the way we do things

### 3.3.5 Critical values

### 3.3.6 Confidence intervals

What is the probability that our  $z$  statistic will be within a certain interval, if the null hypothesis is true? For example, what is the following probability?

$$\Pr(-1.96 \leq z \leq 1.96)? \quad (3.12)$$

$$\Pr\left(-1.96 \leq \frac{\bar{y} - \mu_{y,0}}{\sqrt{\sigma_y^2/n}} \leq 1.96\right) = 0.95 \quad (3.13)$$

Finally, we solve equation 3.13 so that the null hypothesis  $\mu_{y,0}$  is in the middle of the probability statement:

$$\Pr\left(\bar{y} - 1.96 \times \sqrt{\frac{\sigma_y^2}{n}} \leq \mu_{y,0} \leq \bar{y} + 1.96 \times \sqrt{\frac{\sigma_y^2}{n}}\right) = 0.95 \quad (3.14)$$