

# 15: INFERENCE FOR SLR

Larsen & Marx 11.3

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## 1 Properties of MLEs for Simple Linear Regression

1.  $\hat{\beta}_0$  and  $\hat{\beta}_1$  are normal RV's
2.  $\hat{\beta}_0$  and  $\hat{\beta}_1$  are unbiased
3.  $V(\hat{\beta}_1) = \frac{\sigma^2}{\sum(x_i - \bar{X})^2}$
4.  $V(\hat{\beta}_0) = \frac{\sigma^2 \sum x_i^2}{n \sum (x_i - \bar{X})^2} = \sigma^2 \left[ \frac{1}{n} + \frac{\bar{x}^2}{\sum (x_i - \bar{x})^2} \right]$
5.  $\hat{\beta}_1$ ,  $\bar{Y}$  and  $\hat{\sigma}^2$  are mutually independent
6.  $\frac{n\hat{\sigma}^2}{\sigma^2} \sim \chi_{n-2}^2$
7.  $S^2 = \frac{n}{n-2} \hat{\sigma}^2$  is an unbiased estimator for  $\sigma^2$

**Proof:** ( $\hat{\beta}_1$  is a normal RV)

**Proof:** ( $V(\hat{\beta}_1)$ )

## 2 Inference for Simple Linear Regression

### 2.1 Inference for $\beta_1$

#### Test statistic for $\beta_1$

Let  $(x_1, Y_1), (x_2, Y_2), \dots, (x_n, Y_n)$  be a set of points satisfying  $E(Y|X = x) = \beta_0 + \beta_1 x$  and let  $S^2 = \frac{1}{n-2} \sum (Y_i - (\hat{\beta}_0 + \hat{\beta}_1 x_i))^2$ . Then,

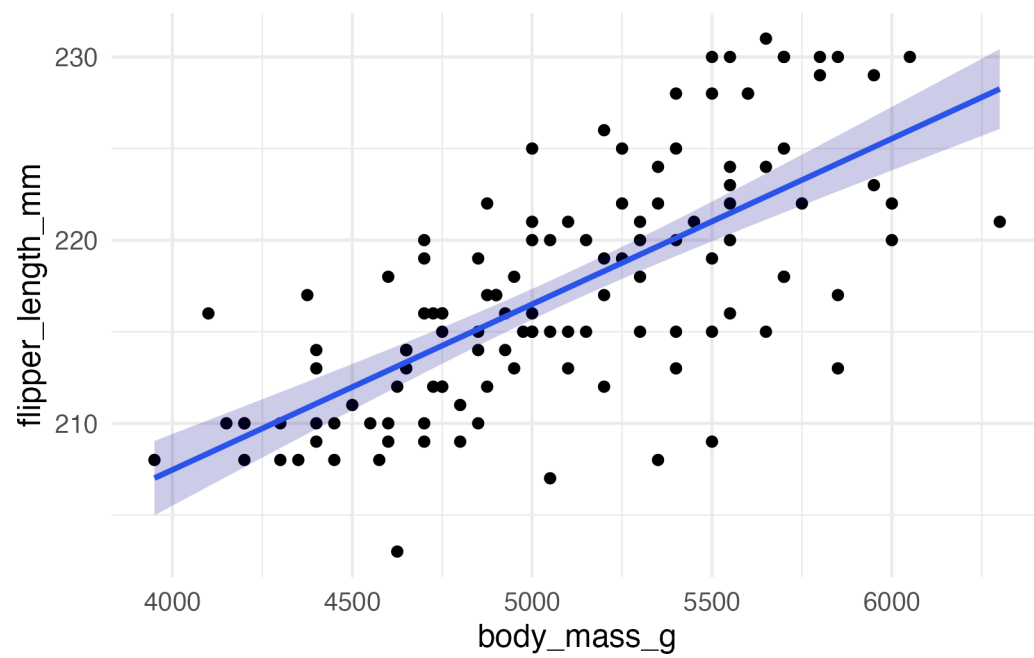
$$T = \frac{\hat{\beta}_1 - \beta_1}{S / \sqrt{\sum (x_i - \bar{x})^2}}$$

**Proof:**

*Note:* Hypothesis tests based on  $T$  are GLRTs!

## 2.2 Inference for $\sigma^2$

### 2.3 Inference for $E(Y|x)$



## 2.4 Inference for new $Y_i$ 's

