# R01b - Simple linear regression Uncertainty and prediction intervals

STAT 5870 (Engineering) Iowa State University

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# Uncertainty when explanatory variable is zero

Let

$$Y_i \stackrel{ind}{\sim} N(\beta_0 + \beta_1 X_i, \sigma^2),$$

then

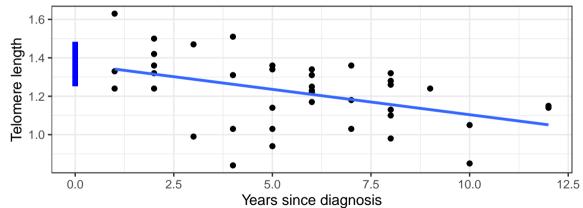
$$E[Y_i|X_i=0] = \beta_0$$

and a 100(1-a)% credible/confidence interval is

$$\hat{\beta}_0 \pm t_{n-2,1-a/2} \hat{\sigma} \sqrt{\frac{1}{n} + \frac{\overline{x}^2}{(n-1)s_x^2}}.$$

#### Telomere data: uncertainty

#### Telomere length vs years post diagnosis



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# Uncertainty when explanatory variable is x

Let

$$Y_i \stackrel{ind}{\sim} N(\beta_0 + \beta_1 X_i, \sigma^2),$$

then

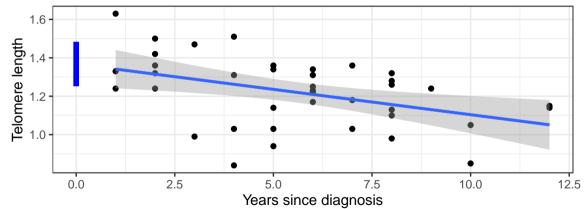
$$E[Y_i|X_i = x] = \beta_0 + \beta_1 x$$

and a 100(1-a)% credible/confidence interval is

$$\hat{\beta}_0 + \hat{\beta}_1 x \pm t_{n-2,1-a/2} \hat{\sigma} \sqrt{\frac{1}{n} + \frac{(\overline{x} - x)^2}{(n-1)s_x^2}}.$$

#### Telomere data: uncertainty

#### Telomere length vs years post diagnosis



#### Prediction intervals

Let

$$Y_i \stackrel{ind}{\sim} N(\beta_0 + \beta_1 X_i, \sigma^2),$$

then

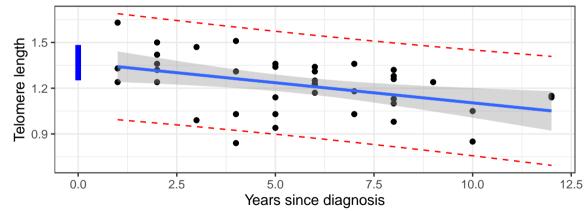
$$E[Y_i|X_i = x] = \beta_0 + \beta_1 x$$

and a 100(1-a)% prediction interval is

$$\hat{\beta}_0 + \hat{\beta}_1 x \pm t_{n-2,1-a/2} \hat{\sigma} \sqrt{1 + \frac{1}{n} + \frac{(\overline{x} - x)^2}{(n-1)s_x^2}}.$$

#### Telomere data: prediction intervals

#### Telomere length vs years post diagnosis



# Summary

Two main types of uncertainty intervals:

• where is the line?

$$\hat{\beta}_0 + \hat{\beta}_1 x \pm t_{n-2,1-a/2} \hat{\sigma} \sqrt{\frac{1}{n} + \frac{(\overline{x} - x)^2}{(n-1)s_x^2}}$$

• where will a new data point fall?

$$\hat{\beta}_0 + \hat{\beta}_1 x \pm t_{n-2,1-a/2} \hat{\sigma} \sqrt{\frac{1+\frac{1}{n} + \frac{(\overline{x}-x)^2}{(n-1)s_x^2}}}$$

Both intervals are confidence and credible intervals.