

# Simple Linear Regression Model – Part IV

Johns Hopkins Engineering

## **625.461 Statistical Models and Regression**

Module 2 – Lecture 2C



## Interval estimation: 100(1- $\alpha$ )% Confidence Interval (CI)

Following the basics in last lecture, we can obtain

$$\hat{\beta}_1 - t_{\alpha/2; n-2} se(\hat{\beta}_1) \leq \beta_1 \leq \hat{\beta}_1 + t_{\alpha/2; n-2} se(\hat{\beta}_1)$$

$$\hat{\beta}_0 - t_{\alpha/2; n-2} se(\hat{\beta}_0) \leq \beta_0 \leq \hat{\beta}_0 + t_{\alpha/2; n-2} se(\hat{\beta}_0)$$

## Interval estimation: 100(1-α)% Confidence Interval (CI)

Variance is 2<sup>nd</sup> moment. If  $y$  is normally distributed,

$$P\left\{\chi_{1-\alpha/2, n-2}^2 \leq \frac{(n-2)MS_{\text{Res}}}{\sigma^2} \leq \chi_{\alpha/2, n-2}^2\right\} = 1 - \alpha$$

$$\frac{(n-2)MS_{\text{Res}}}{\chi_{\alpha/2, n-2}^2} \leq \sigma^2 \leq \frac{(n-2)MS_{\text{Res}}}{\chi_{1-\alpha/2, n-2}^2}$$

# Rocket Propellant Data (Ex 2.5, page 30 of Textbook)

95% CI

$$\hat{\beta}_1 - t_{0.025,18} \text{se}(\hat{\beta}_1) \leq \beta_1 \leq \hat{\beta}_1 + t_{0.025,18} \text{se}(\hat{\beta}_1)$$
$$-37.15 - (2.101)(2.89) \leq \beta_1 \leq -37.15 + (2.101)(2.89)$$

$$\frac{18(9244.59)}{31.5} \leq \sigma^2 \leq \frac{18(9244.59)}{8.23}$$

# Estimation of Mean Response $E(y | x_0)$ at $x = x_0$

$$E(y | x_0) = \mu_{y|x_0} = \beta_0 + \beta_1 x_0 \quad , \quad \widehat{E(y|x_0)} = \hat{\mu}_{y|x_0} = \hat{\beta}_0 + \hat{\beta}_1 x_0$$

$$\begin{aligned} \text{Var}(\hat{\mu}_{y|x_0}) &= \text{Var}(\hat{\beta}_0 + \hat{\beta}_1 x_0) = \text{Var}\left[\bar{y} + \hat{\beta}_1(x_0 - \bar{x})\right] \\ &= \frac{\sigma^2}{n} + \frac{\sigma^2(x_0 - \bar{x})^2}{S_{xx}} = \sigma^2 \left[ \frac{1}{n} + \frac{(x_0 - \bar{x})^2}{S_{xx}} \right] \end{aligned}$$

$$\begin{aligned} \hat{\mu}_{y|x_0} - t_{\alpha/2, n-2} \sqrt{MS_{\text{Res}} \left( \frac{1}{n} + \frac{(x_0 - \bar{x})^2}{S_{xx}} \right)} \\ \leq E(y|x_0) \leq \hat{\mu}_{y|x_0} + t_{\alpha/2, n-2} \sqrt{MS_{\text{Res}} \left( \frac{1}{n} + \frac{(x_0 - \bar{x})^2}{S_{xx}} \right)} \end{aligned}$$

## How to predict new $y_0$ at $x = x_0$ ?

At  $x = x_0$  , the new value  $y_0 = \beta_0 + \beta_1 x_0 + \varepsilon_0$  . Thus, the predicted value should be  $\hat{y}_0 = \hat{\beta}_0 + \hat{\beta}_1 x_0$  , unbiased for  $y_0$  . Its variance is:

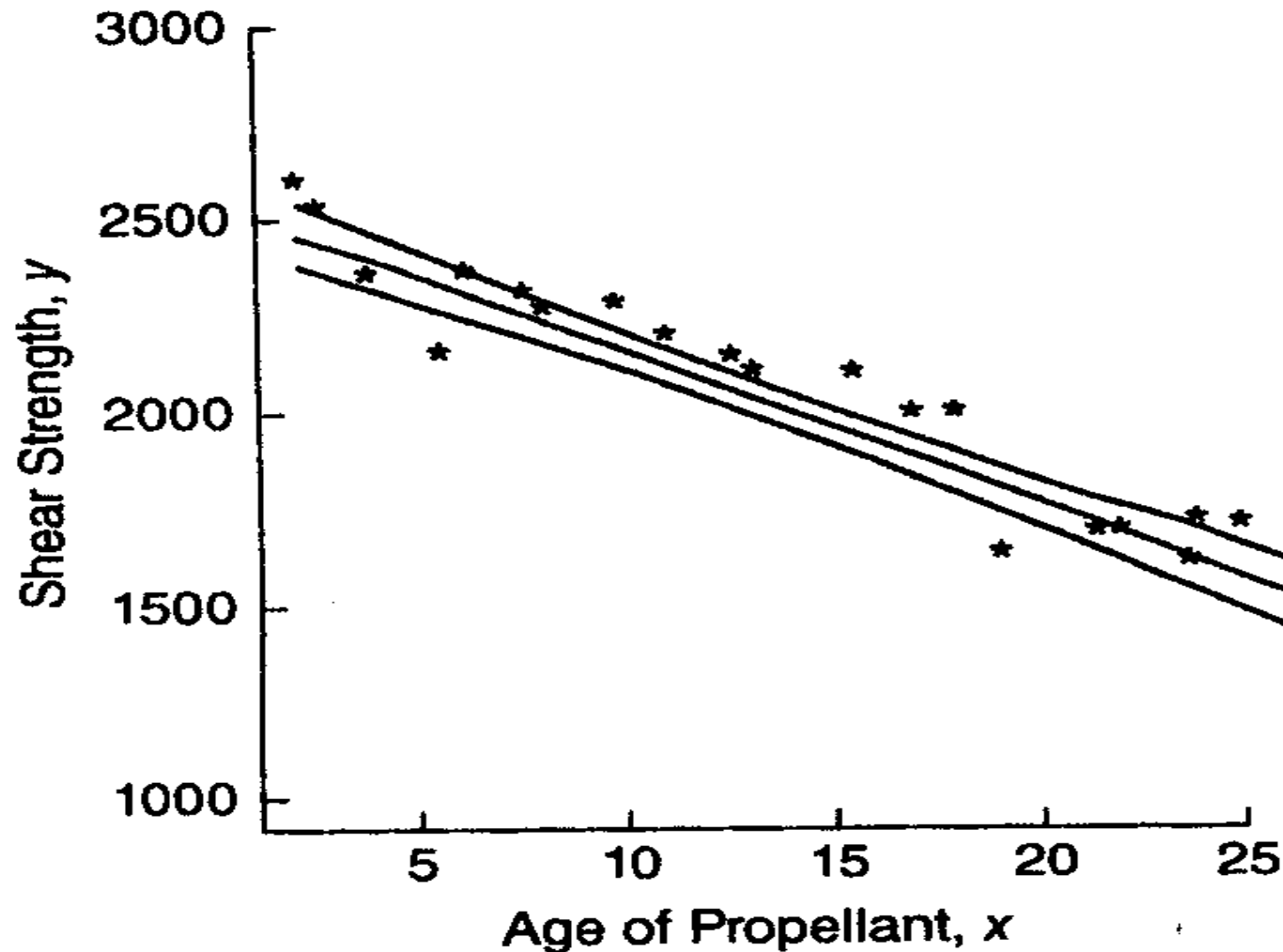
$$\text{Var}(y_0 - \hat{y}_0) = \sigma^2 \left[ 1 + \frac{1}{n} + \frac{(x_0 - \bar{x})^2}{S_{xx}} \right]$$

$$\begin{aligned} \hat{y}_0 - t_{\alpha/2, n-2} \sqrt{MS_{\text{Res}} \left( 1 + \frac{1}{n} + \frac{(x_0 - \bar{x})^2}{S_{xx}} \right)} \\ \leq y_0 \leq \hat{y}_0 + t_{\alpha/2, n-2} \sqrt{MS_{\text{Res}} \left( 1 + \frac{1}{n} + \frac{(x_0 - \bar{x})^2}{S_{xx}} \right)} \end{aligned}$$

## Rocket Propellant Data (Ex 2.6, page 31 of Textbook)

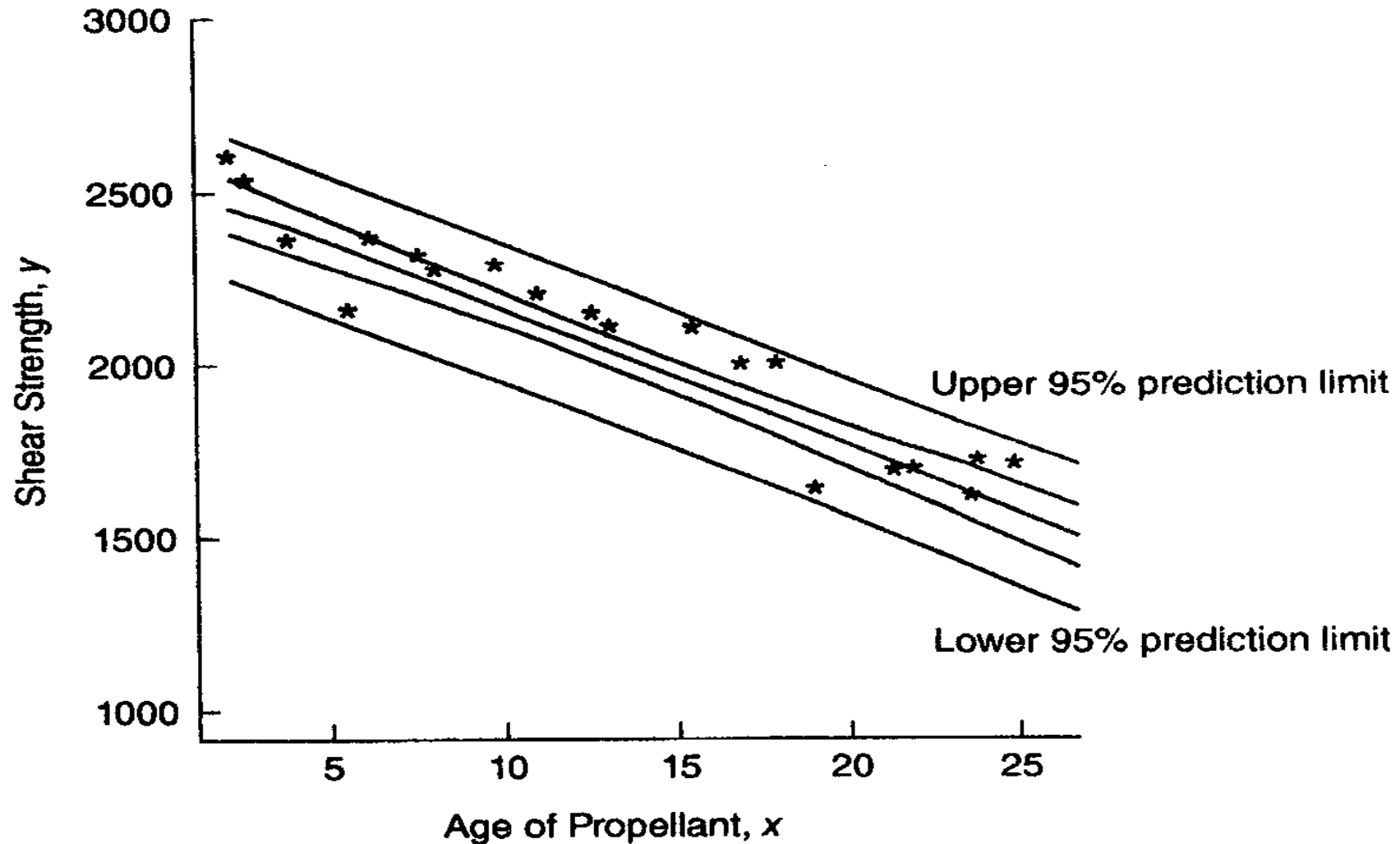
$$\begin{aligned} & \hat{\mu}_{y|x_0} - (2.101) \sqrt{9244.59 \left( \frac{1}{20} + \frac{(x_0 - 13.3625)^2}{1106.56} \right)} \\ & \leq E(y|x_0) \leq \hat{\mu}_{y|x_0} + (2.101) \sqrt{9244.59 \left( \frac{1}{20} + \frac{(x_0 - 13.3625)^2}{1106.56} \right)} \end{aligned}$$

# Confidence Limits of Estimated Mean Shear Strength (Fig 2.4, page 32 of Textbook)





# Prediction Limits of New Shear Strength (Fig 2.5, page 35 of Textbook)





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