

# STA 674

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Regression Analysis And Design Of Experiments  
Fitting Simple Linear Regression Models – Lecture 5



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## Fitting Simple Linear Regression Models

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- Last time: we flirted with inference (about the LR parameters)
- Now we'll use the sampling distribution information to differentiate between the two facets of the estimators' properties:
  - Accuracy
  - Precision

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**Accuracy, precision of the linear regression parameters' estimators**

- IF the assumptions of the regression model are satisfied THEN ...

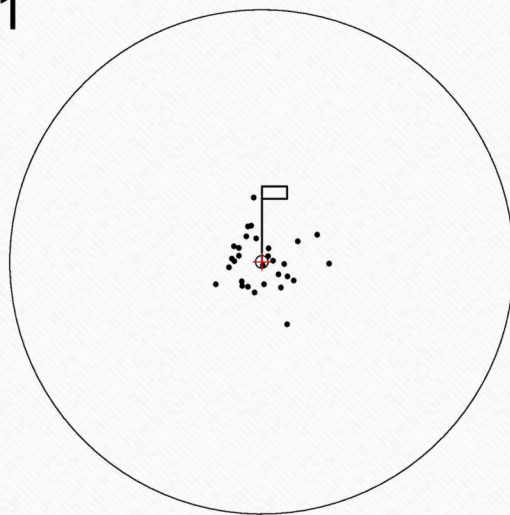
$$b_0 \sim \text{Normal} \left( \beta_0, \sigma_e^2 \left( \frac{1}{n} + \frac{\bar{x}^2}{(n-1)s_x^2} \right) \right)$$

$$b_1 \sim \text{Normal} \left( \beta_1, \frac{\sigma_e^2}{(n-1)s_x^2} \right)$$

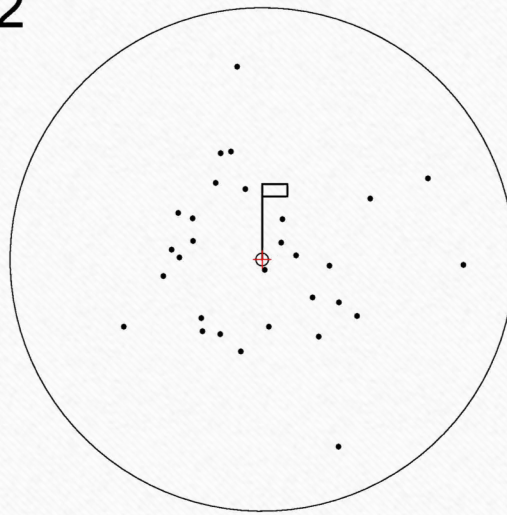
- **Accuracy and Precision**
  - **Accuracy:** a parameter estimator is accurate if its average value is equal to the population parameter.
  - **Precision:** a parameter estimator is precise if its sampling variation is small.



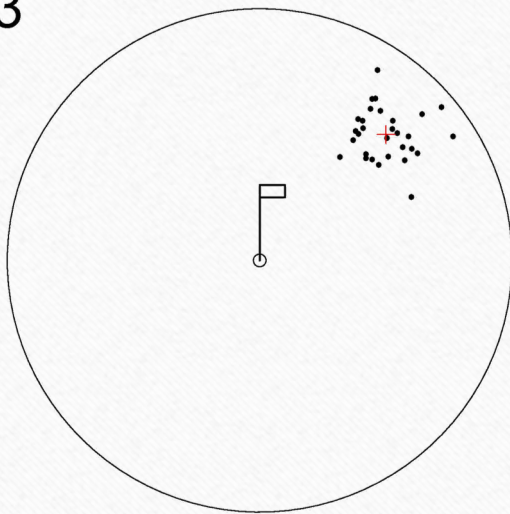
1



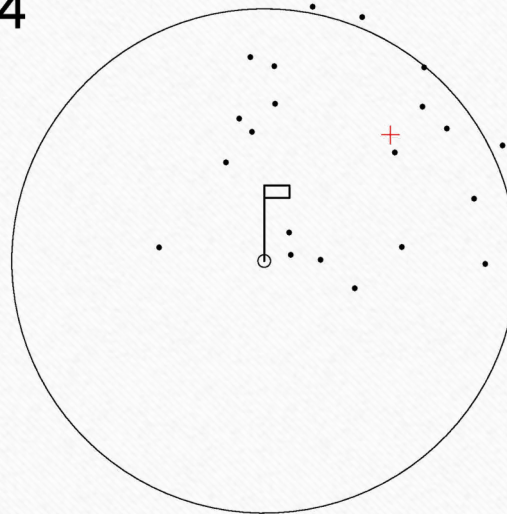
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- **Accuracy:** if the assumptions of the regression model are satisfied then the least squares estimates are accurate:
- the expected value of  $b_0$  is  $\beta_0$
- the expected value of  $b_1$  is  $\beta_1$
- **Precision:** if the assumptions of the regression model are satisfied then we can measure precision by the standard deviation of the sampling distributions:

- $SD(b_0) = \sqrt{\sigma_e^2 \left( \frac{1}{n} + \frac{\bar{x}^2}{(n-1)s_x^2} \right)}$  and  $SD(b_1) = \sqrt{\frac{\sigma_e^2}{(n-1)s_x^2}}$



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- **Standard errors:** if we approximate the values of  $SD(b_0)$  and  $SD(b_1)$  by replacing the unknown  $\sigma_e^2$  by *its* estimate,  $s_e^2$ , then we get the **standard errors** for these estimators:

- $SD(b_0) \approx \sqrt{s_e^2 \left( \frac{1}{n} + \frac{\bar{x}^2}{(n-1)s_x^2} \right)} = s_{b_0}$

- $SD(b_1) \approx \sqrt{\frac{s_e^2}{(n-1)s_x^2}} = s_{b_1}$

Standard deviation requires knowledge of population parameter...  
Standard error is far more practical and only requires knowledge of sample statistic

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- Example: Hooker's data:
- A (severely curtailed) SAS output is given at right.
- Identify the standard errors of  $b_0$  and  $b_1$ .
- Provide an interpretation of these values.

		Parameter	
Variable	DF	Parameter Estimate	Standard Error
Intercept	1	146.67290	0.77641
pressure	1	2.25260	0.03809



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### Summary

- The least squares parameter estimates will vary from sample-to-sample.
- Under certain conditions the least squares estimates are normally distributed.
- The mean of  $b_0$  is  $\beta_0$  and the mean of  $b_1$  is  $\beta_1$  – this means that the least squares estimates are accurate.
- The precision of the estimates can be measured by the standard deviations of the distributions of  $b_0$  and  $b_1$ .
- The variances of these distributions depend on the unknown value  $\sigma_e^2$ , but can be approximated by replacing  $\sigma_e^2$  by its estimate  $s_e^2$ . The new values are called the standard errors of  $b_0$  and  $b_1$ .