

1 Prediction & Fit

- Once we have values for \hat{a} and \hat{b} , we can use them to predict values for Y: $\hat{Y}_i = a + bX_i$, for some given value of X: X_0 .
- Difference between sample and population regression line for $E(Y|X_0)$ is the forecasting or prediction error.
- Note that \hat{Y}_0 is a point estimator and therefore has a sampling distribution:

$$E(Y|X_0) = a + bX_0$$

$$\hat{\sigma}_{Y_0}^2 = \hat{\sigma}_e^2 \left[\frac{1}{n-2} + \frac{(X_0 - \bar{X})^2}{\sum_i x_i^2} \right]$$

- We can also estimate a confidence interval:

$$P \left[\left(\hat{a} + \hat{b}X_0 - t_{\alpha/2} \hat{\sigma}(Y_0) \right) \leq a + bX_0 \leq \left(\hat{a} + \hat{b}X_0 + t_{\alpha/2} \hat{\sigma}(Y_0) \right) \right] = (1 - \alpha)$$

- We can also look at the Predictive Ability of an estimated equation using the Chow test.
- The Chow test is a ‘within sample’ prediction test. Test is whether the observations in the sample, and in a truncated part of the sample are from the same population.
- If a sample of size n is drawn, you select a sub-sample of size n_1 . Estimate the same model: $\hat{Y}_i = a + bX_i$ with both n and n_1 and compute $\sum_i e_i^2$ for both models. Compute the F-test:

$$F^* = \frac{(\sum_i e_{ni}^2 - \sum_i e_{n_1i}^2) / (n - n_1)}{\sum_i e_{n_1i}^2 / (n_1 - k)}$$

- $F^* \sim F_{\alpha, (n-n_1, n_1-k)}$
- Predictive ability and general ‘goodness of fit’ for the regression equation can be summarized by the Coefficient of determination (R^2):

$$R^2 = \frac{\text{Explained Sum of Squares}}{\text{Total Sum of Squares}}$$

$$R^2 = \frac{\hat{b} \sum_i x_i y_i}{\sum_i x_i^2}$$