

1. Suppose we have a dataset that shows the number of bedrooms and the selling price for 20 houses in a particular neighborhood

Bedrooms = [1,1,1,2,2,2,2,3,3,3,3,3,3,3,4,4,4,5,5,6]

Prices =

[120,133,139,185,148,160,192,205,244,213,236,280,275,273,312,311,304,415,396,488]

2. Using the Bedrooms data points:
 - Extract a random sample of size 10
 - Compute the mean and standard deviation of the sample
 - Compute the the mean of the population
 - Computer the confidence interval of 95% using the sample
 - Check if the population mean lies between the upper and lower bounds of the interval

$$CI = \bar{x} \pm z \cdot \frac{s}{\sqrt{n}}$$

Mean value (points to \bar{x})
 Lower/Upper limit (points to \pm)
 z-value for the confidence level (points to z)
 Standard deviation (points to s)
 Sample size (points to \sqrt{n})

3. Using both of the Bedrooms and the Prices:
 - Compute the covariance and the correlation between the two variables
 - Build a regression model and estimate the regression parameters (slope and intercept).
 - Predict the house price for a house with 7 rooms
 - Estimate the upper bound and the lower bounds of the prediction interval of a house with 7 rooms
 - Alpha = 0.1
 - N = 20
 - Df = N-2

$$\hat{y} \pm t_{(1-\alpha/2, n-2)} \sqrt{\frac{\sum_{i=1}^n (y_i - \hat{y}_i)^2}{n-2} \left(1 + \frac{1}{n} + \frac{(\bar{x} - \bar{x})^2}{\sum_{i=1}^n (x_i - \bar{x}_i)^2} \right)}$$

4. Using as sample of houses that have only 3 bedrooms
5. Try to find the lower price that can be paid for a house with 3 rooms using Tolerance interval
6. Try to find the highest price that can be paid for a house with 3 rooms using Tolerance interval

$$\bar{x} \pm k_1 s$$

$$\bar{x} \pm k_2 s$$

Some Extra stuff:

Try to use these Formulas as well and compare the results

We use the following formula to calculate a **confidence interval**:

$$\hat{y}_0 \pm t_{\alpha/2, n-2} * S_{yx} \sqrt{((x_0 - \bar{x})^2 / SS_x + 1/n)}$$

We use the following formula to calculate a **prediction interval**:

$$\hat{y}_0 \pm t_{\alpha/2, n-2} * S_{yx} \sqrt{((x_0 - \bar{x})^2 / SS_x + 1/n + 1)}$$

where:

- **\hat{y}_0** : Estimated mean value of response variable
- **$t_{\alpha/2, n-2}$** : t-critical value with n-2 degrees of freedom
- **S_{yx}** : Standard error of response variable
- **x_0** : specific value of predictor variable
- **\bar{x}** : mean value of predictor variable
- **SS_x** : Sum of squares for predictor variable
- **n** : Total sample size