

# Tutorial 4

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# Overview

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## Confidence Intervals for $\hat{\beta}_0$ and $\hat{\beta}_1$

A 95% confidence interval for  $\hat{\beta}_0 \longrightarrow \hat{\beta}_0 \pm t(0.975) \times se(\hat{\beta}_0)$

A 95% confidence interval for  $\hat{\beta}_1 \longrightarrow \hat{\beta}_1 \pm t(0.975) \times se(\hat{\beta}_1)$

# Confidence Interval for parameters

- It is an **observed interval** (i.e., it is calculated from the observations), that potentially includes the unobservable **true parameter of interest**.
- How frequently the observed interval contains the true parameter **if the experiment is repeated** is called the confidence level.
- population mean  $\mu$  and population standard deviation  $\sigma$
- “We are 95% confident that the **true value of the parameter** is in our confidence interval.”

## Prediction Interval for future observations

- It is an **estimated interval** in which **future observations** will fall, with a certain probability, given what has already been observed.
- Prediction intervals predict the distribution of **observable** future points.
- observations or sample point  $X_{n+1}$
- “On repeated applications of this computation, the future observation  $X_{n+1}$  will fall in the predicted interval 95% of the time.”

## Confidence Intervals for Prediction

We often interested in predicting the **response value for future observations** at particular values of predictor (e.g.,  $x_0$ ).

$$\hat{Y}(x_0) = b_0 + b_1x_0$$

Associated standard error of the estimate

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

Substitute our estimator,  $s$ , for the scale parameter,  $\sigma$ , we then have an estimate of the standard error of prediction

$$s\{\hat{Y}(x_0)\} = s\sqrt{\frac{1}{n} + \frac{(x_0 - \bar{x})^2}{S_{xx}}}$$

## Confidence intervals for prediction

Assuming normal errors, a  $100(1-\alpha)\%$  interval for the **expected response**  $E(Y|x_0)$  is

$$\hat{Y}(x_0) \pm t_{n-2}(1 - \alpha/2)s\{\hat{Y}(x_0)\}$$

This interval is indeed a confidence interval, since it gives a plausible range for a **fixed, population quantity**.

## Prediction intervals for a future response value

We also make prediction for the value of a single future response value,  $Y_0$ , at a particular value of the predictor value,  $x_0$ . It is called a prediction interval since we want to find a likely range for a random quantity.

A future value  $Y_0$  is independent of the observed data used to estimate  $b_0$  and  $b_1$ . So,

$$\text{Var}\{Y_0 - \hat{Y}(x_0)\} = \text{Var}(Y_0) + \text{Var}\{\hat{Y}(x_0)\} = \sigma^2 + \sigma^2 \left\{ \frac{1}{n} + \frac{(x_0 - \bar{x})^2}{S_{xx}} \right\}$$



## Prediction intervals for a future response value

Since  $E\{Y_0 - \hat{Y}(x_0)\} = 0$ , the standard normal theory assumptions show that

$$\frac{Y_0 - \hat{Y}(x_0)}{s\sqrt{1 + \frac{1}{n} + \frac{(x_0 - \bar{x})^2}{S_{xx}}}} \sim t_{n-2}.$$

(see Page 19 of Lecture Notes)

# CI and PI

A  $100(1-\alpha)\%$  confidence interval for a given value  $x_0$ :

$$\hat{Y}(x_0) \pm t_{\alpha/2, n-2} \times s \sqrt{\frac{1}{n} + \frac{(x_0 - \bar{x})^2}{S_{xx}^2}}$$

A  $100(1-\alpha)\%$  prediction interval for a given value  $x_0$ :

$$\hat{Y}(x_0) \pm t_{\alpha/2, n-2} \times s \sqrt{1 + \frac{1}{n} + \frac{(x_0 - \bar{x})^2}{S_{xx}^2}}$$

- The further away from the  $\bar{x}$  we are predicting, the wider our prediction interval will be.
- PI is wider than CI for the same value of  $x_0$

## Hypothesis test of the $\rho_{x,y}$

**Step One:**  $H_0 : \rho_{x,y} = 0$  v.s.  $H_A : \rho_{x,y} \neq 0$

**Step Two:** Test Statistic  $= \frac{r-0}{se(r)} = \frac{r\sqrt{n-2}}{1-r^2}$

**Step Three:** Refers to the  $t$  distribution table with  $n - 2$  degrees of freedom and find the critical values.

**Step Four:** Compare the calculated test statistics with the critical values and make a decision.

**Step Five:** Conclusion

## Question 1 (d)-(g)

In R, we use "predict()" command to compute 95% confidence intervals and prediction intervals

- CI: `predict(object,newdata=as.data.frame(cbind(col.name=new.x)), interval="confidence")`
- PI: `predict(object,newdata=as.data.frame(cbind(col.name=new.x)), interval="prediction")`