

# Inference in Multiple Regression: Part 1

- Test for Significant Overall Regression
- Type I Sum of Squares
- Type III Sum of Squares
- Partial F-Test
- Confidence Intervals About Regression Coefficients

Lecture 8  
Sections 9.1 – 9.3, 9.5

## Three Types of Tests in Multiple Regression

- 1. Overall Test:** Does the entire set of independent variables contribute significantly to the prediction of  $Y$ ?
- 2. Test for Addition of a Single Variable:** Does the addition of one particular independent variable add significantly to the prediction of  $Y$  after considering all other predictors already in the model?
- 3. Test for Addition of a Group of Variables:** Does the addition of some group of independent variables add significantly to the prediction of  $Y$  after considering all other predictors already in the model?

## Test for Significant Overall Regression

- **Goal:** Determine if the entire set of predictors  $X_1, X_2, \dots, X_k$  contributes significantly to the prediction of  $Y$ 
  - **Procedure:** Test the full model against the model with no predictors where  $\bar{Y}$  is the best prediction for all observations
  - **Hypotheses:**  $H_0: \beta_1 = \beta_2 = \dots = \beta_k = 0$  vs.  $H_A: \text{At least one } \beta_i \neq 0$
  - **Test Statistic:**  $F = \frac{MSR}{MSE} = \frac{(SSY - SSE)/k}{SSE/(n-k-1)}$ 
    - Has  $k$  and  $n - k - 1$  df
  - **Equivalent Test Statistic:**  $F = \frac{R^2/k}{(1-R^2)/(n-k-1)}$

## Example: Test for Significant Overall Regression

- **Scenario:** Examine how patient satisfaction (1-100) is related to age (in years), severity of illness (1-100), and anxiety level (1-10)
- **Task:** Perform the test for overall regression.

• **Model:** \_\_\_\_\_

•  $X_1$ : \_\_\_\_\_

•  $X_2$ : \_\_\_\_\_

•  $X_3$ : \_\_\_\_\_

• **Hypotheses:**

•  $H_0$ : \_\_\_\_\_

•  $H_A$ : \_\_\_\_\_

*	C1	C2	C3	C4
	Patient Satisfaction	Age	Severity	Anxiety Level
1	48	50	51	7.2
2	57	36	46	7.2
3	66	40	48	6.8
4	70	41	44	5.2
5	89	28	43	5.2
6	36	49	54	9.6
7	46	42	50	6.8
8	54	45	48	7.6
9	26	52	62	9.6
10	77	29	50	6.4

## Example: Test for Significant Overall Regression

- **Scenario:** Examine how patient satisfaction (1-100) is related to age (in years), severity of illness (1-100), and anxiety level (1-10)
- **Task:** Perform the test for overall regression.

Source	DF	Adj SS	Adj MS	F-Value	P-Value
Regression	3	9120			0.000
Error	42	4249			
Total	45	13369			

• **Test Statistic:**

$F =$  \_\_\_\_\_

• **Critical Value:** \_\_\_\_\_; **P-Value:** \_\_\_\_\_

• **Conclusion:** \_\_\_\_\_ and conclude that the model is \_\_\_\_\_

## Review: Fundamental Equation of Regression Analysis

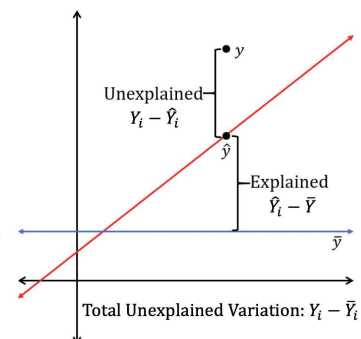
- Recall the Fundamental Equation of Regression Analysis:

$$SSY = SSR + SSE$$

where  $SSR$  is the amount of variability explained by the regression and  $SSE$  is the unexplained variability.

- In multiple regression, every time another predictor is inserted into the model, it helps to explain the response better.

• **Result:** Some of the \_\_\_\_\_ becomes \_\_\_\_\_



## Type I Sum of Squares

- **Type I Sum of Squares:** the amount of explained variability contributed to SSR by a predictor when it is added into a model after considering the contributions to the SSR by the other predictors already added to the model
  - Referred to as **variables added in order** or **sequential** sums of squares
- The types of Type I sums of squares are dependent upon the order in which the variables enter the model. For example:
  - $SS(X_1)$ : Sum of squares explained using only  $X_1$ 
    - $X_1$  is the first predictor added to the model
  - $SS(X_2|X_1)$ : Sum of squares added to SSR by  $X_2$  after  $X_1$  has already been added into the model
  - $SS(X_3|X_1, X_2)$ : Sums of squares added to SSR by  $X_3$  after both  $X_1$  and  $X_2$  have been added into the model

## Type I Sum of Squares

- Every combination in which predictors are added has its own set of Type I sum of squares
- For a regression with 3 predictors, there are 12 Type I SS:
  - $SS(X_1), SS(X_2), SS(X_3)$
  - $SS(X_1|X_2), SS(X_1|X_3), SS(X_2|X_1), SS(X_2|X_3), SS(X_3|X_1), SS(X_3|X_2)$
  - $SS(X_1|X_2, X_3), SS(X_2|X_1, X_3), SS(X_3|X_1, X_2)$
- Regardless of the order the variables are added, the Type I SS will always sum to the total sum of squares for the regression.
  - $SSR = SS(X_1) + SS(X_2|X_1) + SS(X_3|X_1, X_2)$
- When another predictor is added, its Type I SS comes out of the SSE to guarantee that  $SSY = SSR + SSE$  always holds.

## Example: Type I Sums of Squares

- **Scenario:** Examine how patient satisfaction (1-100) is related to age (in years), severity of illness (1-100), and anxiety level (1-10)
- **Question:** How can we tell that these are Type I sums of squares?
- **Answer:** Sums of squares for individual predictors \_\_\_\_\_

• \_\_\_\_\_

- **Question:** In what order were the predictors added?

- **Answer:**

1. \_\_\_\_\_

2. \_\_\_\_\_

3. \_\_\_\_\_

Source	DF	Seq SS	Seq MS	F-Value	P-Value
Regression	3	9120.5	3040.15	30.05	0.000
Age	1	8275.4	8275.39	81.80	0.000
Severity	1	480.9	480.92	4.75	0.035
Anxiety Level	1	364.2	364.16	3.60	0.065
Error	42	4248.8	101.16		
Total	45	13369.3			

## Example: Type I Sums of Squares

- **Scenario:** Examine how patient satisfaction (1-100) is related to age (in years), severity of illness (1-100), and anxiety level (1-10)
- **Question:** What do the Type I sums of squares mean?

Source	DF	Seq SS	Seq MS	F-Value	P-Value
Regression	3	9120.5	3040.15	30.05	0.000
Age	1	8275.4	8275.39	81.80	0.000
Severity	1	480.9	480.92	4.75	0.035
Anxiety Level	1	364.2	364.16	3.60	0.065
Error	42	4248.8	101.16		
Total	45	13369.3			

- **Answer:**
  - **Age:** \_\_\_\_\_ of the 9120.5 SSR is explained by \_\_\_\_\_
  - **Severity:** After \_\_\_\_\_, severity of illness explains an \_\_\_\_\_ of the 9120.5 SSR (or \_\_\_\_\_ SSR)
  - **Anxiety Level:** After accounting for \_\_\_\_\_, anxiety level explains \_\_\_\_\_ of the 9120.5 SSR (which is \_\_\_\_\_)

## Example: Type I Sums of Squares

- **Scenario:** Compare the model that include only age and severity of illness with the model that includes all three predictors.
- **Question:** What happened to the sums of squares?

Source	DF	Seq SS	Seq MS	F-Value	P-Value
Regression	2	8756.3	4378.15	40.81	0.000
Age	1	8275.4	8275.39	77.14	0.000
Severity	1	480.9	480.92	4.48	0.040
Error	43	4613.0	107.28		
Total	45	13369.3			

Source	DF	Seq SS	Seq MS	F-Value	P-Value
Regression	3	9120.5	3040.15	30.05	0.000
Age	1	8275.4	8275.39	81.80	0.000
Severity	1	480.9	480.92	4.75	0.035
Anxiety Level	1	364.2	364.16	3.60	0.065
Error	42	4248.8	101.16		
Total	45	13369.3			

- **Answer:** Sum of squares that were \_\_\_\_\_ in the model with two predictors got \_\_\_\_\_ in the model with all three predictors

## Type III Sum of Squares

- **Type III Sum of Squares:** the amount of explained variability contributed to SSR by a predictor assuming it is the last predictor added and all others are already in the model
  - Referred to as **variables added last** or **adjusted** sums of squares
- For a regression with 3 predictors, there are 3 Type III SS:
  - $SS(X_1|X_2, X_3)$ : Sum of squares added to SSR by  $X_1$  after  $X_2$  and  $X_3$  are already in the model
  - $SS(X_2|X_1, X_3)$ : Sum of squares added to SSR by  $X_2$  after  $X_1$  and  $X_3$  are already in the model
  - $SS(X_3|X_1, X_2)$ : Sum of squares added to SSR by  $X_3$  after  $X_2$  and  $X_3$  are already in the model
- **Important:**  $SSR \neq SS(X_1|X_2, X_3) + SS(X_2|X_1, X_3) + SS(X_3|X_1, X_2)$

## Example: Type III Sum of Squares

- **Scenario:** Examine how patient satisfaction (1-100) is related to age (in years), severity of illness (1-100), and anxiety level (1-10)
- **Question:** How can we tell that these are Type III sums of squares?

Source	DF	Adj SS	Adj MS	F-Value	P-Value
Regression	3	9120.5	3040.15	30.05	0.000
Age	1	2857.6	2857.55	28.25	0.000
Severity	1	81.7	81.66	0.81	0.374
Anxiety Level	1	364.2	364.16	3.60	0.065
Error	42	4248.8	101.16		
Total	45	13369.3			

- **Answer:** Sums of squares for individual predictors sum to \_\_\_\_\_
- \_\_\_\_\_
- Each adjusted sum of squares assumes the other variables are \_\_\_\_\_

## Example: Type III Sum of Squares

- **Scenario:** Want to investigate how weight is related to height, age, and age squared for children with a nutritional deficiency.
- **Question:** What do the Type III sums of squares mean?

Source	DF	Adj SS	Adj MS	F-Value	P-Value
Regression	3	9120.5	3040.15	30.05	0.000
Age	1	2857.6	2857.55	28.25	0.000
Severity	1	81.7	81.66	0.81	0.374
Anxiety Level	1	364.2	364.16	3.60	0.065
Error	42	4248.8	101.16		
Total	45	13369.3			

- **Answer:**
  - **Age:** After \_\_\_\_\_ are inserted into the model, age explains \_\_\_\_\_
  - **Severity:** After \_\_\_\_\_ are inserted into the model, severity of illness explains \_\_\_\_\_
  - **Anxiety Level:** After \_\_\_\_\_ are inserted into the model, anxiety level explains \_\_\_\_\_

## Partial F-Test

- **Goal:** Determine if adding a single variable  $X^*$  significantly improves the prediction of  $Y$  given that  $X_1, X_2, \dots, X_p$  are already in the model
  - **Procedure:** Test the full model against the reduced model
    - **Full Model:** Includes  $X_1, X_2, \dots, X_p$  as well as  $X^*$
    - **Reduced Model:** Includes only  $X_1, X_2, \dots, X_p$  (but not  $X^*$ )
  - **Hypotheses:**  $H_0: \beta^* = 0$  vs.  $H_A: \beta^* \neq 0$ 
    - *Null Hypothesis:* " $X^*$  does not significantly add to the prediction of  $Y$  given that  $X_1, X_2, \dots, X_p$  are already predictors in the model so the regression  $Y = \beta_0 + \beta_1 X_1 + \dots + \beta_p X_p + E$  is sufficient."
    - *Alternative Hypothesis:* " $X^*$  significantly adds to the prediction of  $Y$  given that  $X_1, X_2, \dots, X_p$  are already predictors in the model so the regression  $Y = \beta_0 + \beta_1 X_1 + \dots + \beta_p X_p + \beta^* X^* + E$  is better than the one without  $X^*$ ."

## Partial F-Test

- **Goal:** Determine if adding a single variable  $X^*$  significantly improves the prediction of  $Y$  given that  $X_1, X_2, \dots, X_p$  are already in the model

- **Sum of Squares:** Regression  $SS(X^* | X_1, X_2, \dots, X_p) =$   
Regression  $SS(X_1, X_2, \dots, X_p, X^*) -$  Regression  $SS(X_1, X_2, \dots, X_p)$

- “Extra sum of squares from adding  $X^*$  into the model **equals** regression sum of squares when  $X_1, X_2, \dots, X_p$ , and  $X^*$  are all in the model **minus** regression sum of squares when only  $X_1, X_2, \dots, X_p$  are in the model.”

- **Test Statistic:**  $F(X^* | X_1, X_2, \dots, X_p) = \frac{\text{Regression } SS(X^* | X_1, X_2, \dots, X_p)}{MSE(X_1, X_2, \dots, X_p, X^*)}$

- Has 1 and  $n - p - 2$  df
- “Test statistic **equals** the sum of squares added by  $X^*$  given  $X_1, X_2, \dots, X_p$  are already in the model **divided by** the mean squared error from the full model that includes  $X_1, X_2, \dots, X_p$ , and  $X^*$ .”

## Example: Partial F-Test

- **Scenario:** Examine how patient satisfaction (1-100) is related to age (in years), severity of illness (1-100), and anxiety level (1-10)
- **Task:** Test if age ( $X_1$ ) contributes significantly to the model after severity ( $X_2$ ) and anxiety level ( $X_3$ ) have already been included.

- **Models:**

- **Full Model:** \_\_\_\_\_

- **Reduced Model:** \_\_\_\_\_

- **Hypotheses:**  $H_0$ : \_\_\_\_\_ vs.  $H_A$ : \_\_\_\_\_

- **Need:**

- \_\_\_\_\_
- \_\_\_\_\_
- \_\_\_\_\_

}  $F =$  \_\_\_\_\_

## Example: Partial F-Test

- **Outputs with Type I SS:**

Source	DF	Seq SS	Seq MS	F-Value	P-Value
Regression	3	9120.5	3040.15	30.05	0.000
Age	1	8275.4	8275.39	81.80	0.000
Severity	1	480.9	480.92	4.75	0.035
Anxiety Level	1	364.2	364.16	3.60	0.065
Error	42	4248.8	101.16		
Total	45	13369.3			

Source	DF	Seq SS	Seq MS	F-Value	P-Value
Regression	2	6262.9	3131.5	18.95	0.000
Severity	1	4860.3	4860.3	29.41	0.000
Anxiety Level	1	1402.7	1402.7	8.49	0.006
Error	43	7106.4	165.3		
Total	45	13369.3			

- **Test Statistic:**  $F =$  \_\_\_\_\_

- **Critical Value:** \_\_\_\_\_; **P-Value:** \_\_\_\_\_

- **Conclusion:** \_\_\_\_\_ and conclude that \_\_\_\_\_  
\_\_\_\_\_ after \_\_\_\_\_  
have been included.

## Example: Partial F-Test

- **Scenario:** Examine how patient satisfaction (1-100) is related to age (in years), severity of illness (1-100), and anxiety level (1-10)
- **Task:** Test if severity ( $X_2$ ) contributes significantly to the model after age ( $X_1$ ) and anxiety level ( $X_3$ ) have already been included.
- **Models:**
  - **Full Model:** \_\_\_\_\_
  - **Reduced Model:** \_\_\_\_\_
- **Hypotheses:**  $H_0$ : \_\_\_\_\_ vs.  $H_A$ : \_\_\_\_\_
- **Question:** Rather than finding the sums of squares from two different models, how could we find the test statistic?
- **Answer:** Use \_\_\_\_\_

## Example: Partial F-Test

- **Scenario:** Examine how patient satisfaction (1-100) is related to age (in years), severity of illness (1-100), and anxiety level (1-10)
- **Question:** How can the Type III sum of squares be found for severity of illness?
- **Answer:** Change to \_\_\_\_\_
  - $SS(S|A, AL) =$  \_\_\_\_\_

Type I (Sequential)

Type III (Adjusted)

Source	DF	Seq SS	Seq MS
Regression	3	9120.5	3040.15
Age	1	8275.4	8275.39
Severity	1	480.9	480.92
Anxiety Level	1	364.2	364.16
Error	42	4248.8	101.16
Total	45	13369.3	

Source	DF	Seq SS	Seq MS
Regression	2	9038.8	4519.40
Age	1	8275.4	8275.39
Anxiety Level	1	763.4	763.42
Error	43	4330.5	100.71
Total	45	13369.3	

Source	DF	Adj SS	Adj MS
Regression	3	9120.5	3040.15
Age	1	2857.6	2857.55
Severity	1	81.7	81.66
Anxiety Level	1	364.2	364.16
Error	42	4248.8	101.16
Total	45	13369.3	

## Example: Partial F-Test

- **Scenario:** Examine how patient satisfaction (1-100) is related to age (in years), severity of illness (1-100), and anxiety level (1-10)
- **Task:** Test if severity ( $X_2$ ) contributes significantly to the model after age ( $X_1$ ) and anxiety level ( $X_3$ ) have already been included.

- **Test Statistic:** \_\_\_\_\_
- **Critical Value:** \_\_\_\_\_
- **P-Value:** \_\_\_\_\_

Source	DF	Adj SS	Adj MS	F-Value	P-Value
Regression	3	9120.5	3040.15	30.05	0.000
Age	1	2857.6	2857.55	28.25	0.000
Severity	1	81.7	81.66	0.81	0.374
Anxiety Level	1	364.2	364.16	3.60	0.065
Error	42	4248.8	101.16		
Total	45	13369.3			

- **Conclusion:** \_\_\_\_\_ and conclude that severity of illness \_\_\_\_\_ after age and anxiety level have \_\_\_\_\_.



## t-Test and Confidence Interval

- **Goal:** Determine if adding a single variable  $X^*$  significantly improves the prediction of  $Y$  given that  $X_1, X_2, \dots, X_p$  are already in the model

- **Hypotheses:**  $H_0: \beta^* = 0$  vs.  $H_0: \beta^* \neq 0$

- **Test Statistic:**  $t = \frac{\hat{\beta}^*}{S_{\hat{\beta}^*}}$  which has  $n - p - 1$  df

- Equivalent to the partial F-test because only one parameter is being tested and  $t_{n-p-1}^2 = F_{1,n-p-1}$ .

- **Confidence Interval:** A  $100(1 - \alpha)\%$  confidence interval for the coefficient  $\beta^*$  for the predictor  $X^*$  after  $X_1, X_2, \dots, X_p$  have been added is:

$$\hat{\beta}^* \pm t_{n-p-1, 1-\alpha/2} \times S_{\hat{\beta}^*}$$

## Example: t-Test

- **Task:** Test if anxiety level ( $X_3$ ) contributes significantly to the model after age ( $X_1$ ) and severity ( $X_2$ ) have already been included.

Term	Coef	SE Coef	T-Value	P-Value	VIF
Constant	151.8	18.3	8.30	0.000	
Age	-1.142	0.215	-5.31	0.000	1.63
Severity	-0.442	0.492	-0.90	0.374	2.00
Anxiety Level	-3.37	1.77	-1.90	0.065	2.01

- **Models:**

- **Full Model:** \_\_\_\_\_

- **Reduced Model:** \_\_\_\_\_

- **Hypotheses:**  $H_0$ : \_\_\_\_\_ vs.  $H_A$ : \_\_\_\_\_

- **Test Statistic:** \_\_\_\_\_

## Example: t-Test

- **Task:** Test if anxiety level ( $X_3$ ) contributes significantly to the model after age ( $X_1$ ) and severity ( $X_2$ ) have already been included.

Term	Coef	SE Coef	T-Value	P-Value	VIF
Constant	151.8	18.3	8.30	0.000	
Age	-1.142	0.215	-5.31	0.000	1.63
Severity	-0.442	0.492	-0.90	0.374	2.00
Anxiety Level	-3.37	1.77	-1.90	0.065	2.01

- **Critical Values:** \_\_\_\_\_; **P-Value:** \_\_\_\_\_

- **Confidence Interval:** \_\_\_\_\_

- **Conclusion:** \_\_\_\_\_ and conclude that anxiety level \_\_\_\_\_ after age and severity of illness have been included.

- Confidence interval \_\_\_\_\_  $\rightarrow$  \_\_\_\_\_