

Business Analytics

Lecture 7

Simple Linear Regression

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Review

- Session 6: Hypothesis Testing
 - Null Hypothesis vs. Alternative Hypothesis
 - Set confidence level and significance level
 - Computing the p-value, t-stats
 - Rejecting H_0 to accept H_A requires strong statistical evidence
- Session 7 and 8: simple statistical tool for studying relationships:
 - Regression analysis

Example: Armand's Pizza

Restaurant i	Student Population ('000) X_i	Annual Sales (\$ '000) Y_i
1	2	58
2	6	105
3	8	88
4	8	118
5	12	117
6	16	137
7	20	157
8	20	169
9	22	149
10	26	202

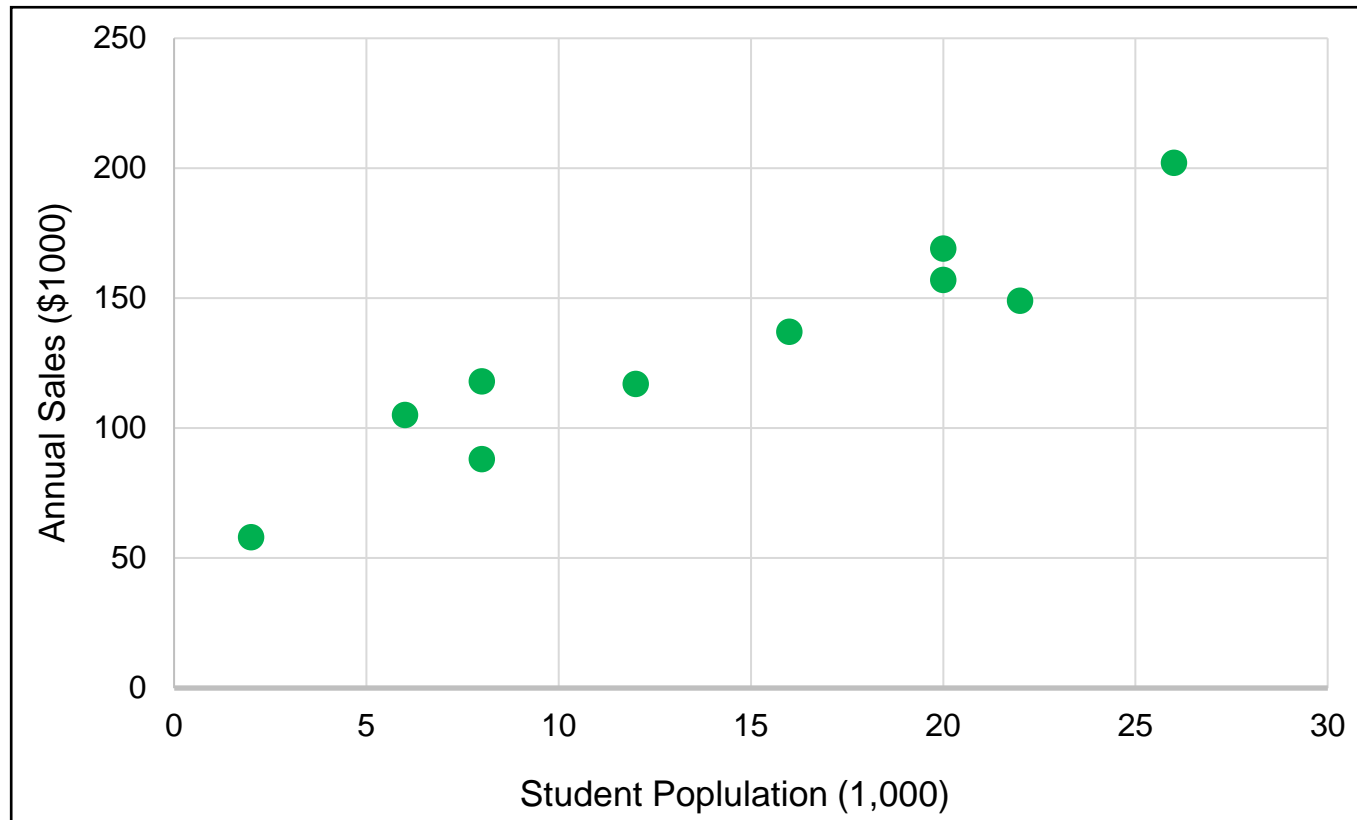
Introduction

- Regression refers to the statistical technique of **modeling the relationship between variables**.
- In simple linear regression, we model the relationship between two variables.
- One of the variables, denoted by Y , is called the **dependent variable** and the other, denoted by X , is called the **independent variable**.
- The model we will use to depict the relationship between X and Y will be a straight-line relationship.
- A graphical sketch of the pairs (X, Y) is called a scatter plot.

The Goal

- The basic idea in simple linear regression is to
 - (i) **establish** a relationship between a dependent variable Y and an independent variable X
 - (ii) **quantify** the magnitude of the impact of X on Y
 - (iii) **find** the 95% prediction interval for forecasting

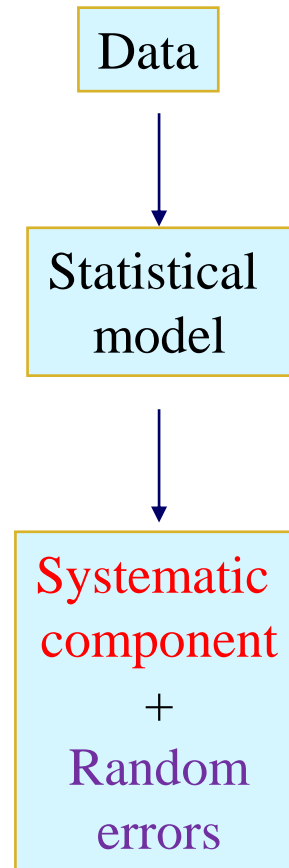
Armand's Pizza: Scatter Plot



Any relationship between Student Population and Annual Sales?
We need a statistical model to answer this question.

Model Building

A statistical model separates the **systematic component** of a relationship from the **random component**.



In regression, the **systematic component** is the overall linear relationship, and the **random component** is the variation around the line.

The Simple Linear Regression Model

The population simple linear regression model:

$$Y = \beta_0 + \beta_1 X + \varepsilon$$

Nonrandom or
Systematic
Component
Random
Component

where

- Y is the **dependent variable**, the variable we wish to explain or predict
- X is the **independent variable**, also called the predictor variable
- ε is the error term, the only random component in the model, and thus, the only source of randomness in Y
- β_0 is the intercept of the systematic component of the regression relationship
- β_1 is the slope of the systematic component

Assumptions of the Model

$$Y = \beta_0 + \beta_1 X + \varepsilon$$

- β_0 Y-intercept of the line
- β_1 the slope of the line
- ε the error

1. The error ε is a random variable with mean 0.
2. The variance of ε , denoted as σ^2 , is the same for all values of X .
3. The values of ε are independent.
4. The error term ε is Normally distributed.

How to Estimate?

Estimation of a simple linear regression relationship involves finding estimated or predicted values of the intercept and slope of the linear regression line.

The **estimated regression equation**:

$$Y = b_0 + b_1X + \varepsilon$$

where

- b_0 estimates the intercept of the population regression line, β_0 ;
- b_1 estimates the slope of the population regression line, β_1 ;
- ε stands for the observed errors - the residuals from fitting the estimated regression line $b_0 + b_1X$ to a set of n points.

The estimated regression line:

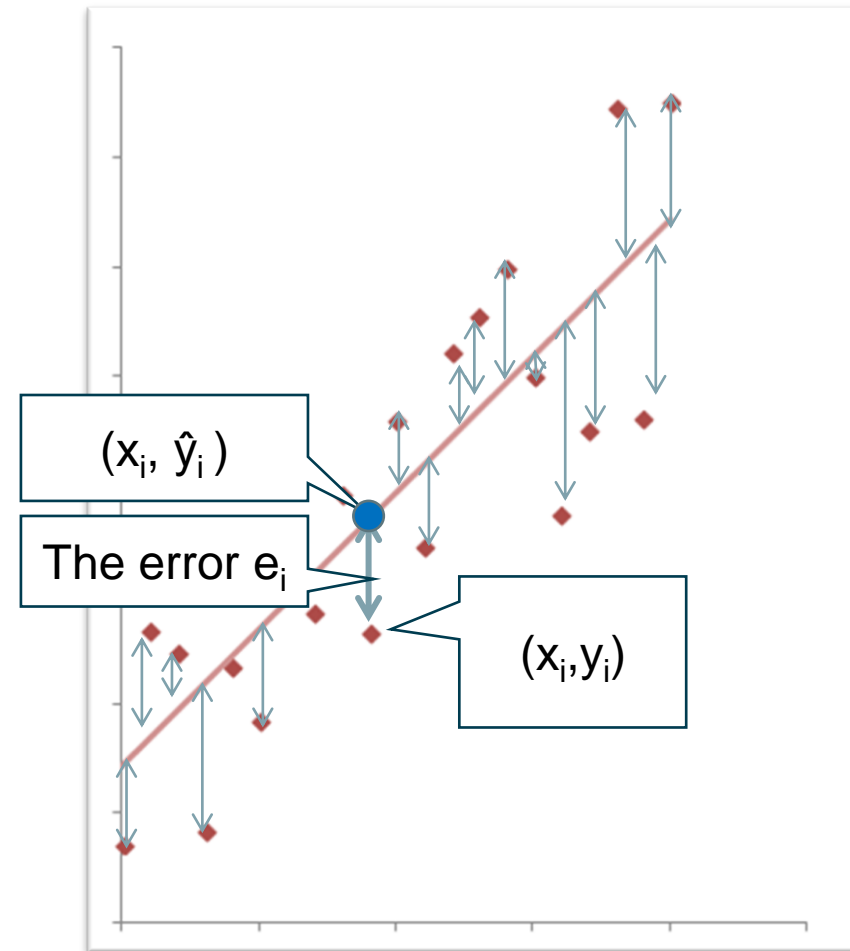
$$\hat{Y} = b_0 + b_1X$$

where \hat{Y} (Y-hat) is the value of Y lying on the fitted regression line for a given value of X .

The method of **least squares**

- To find coefficients b_0, b_1 ,
- we denote each data point by (x_i, y_i) .
- The line gives us an approximated value:
 $\hat{y}_i = b_0 + b_1 x_i$.
- The approximation error of each point is
 $e_i = |y_i - \hat{y}_i|$.
- **The Sum of Squares for Errors in regression is:**

$$SSE = \sum_{i=1}^n e_i^2 = \sum_{i=1}^n (y_i - \hat{y}_i)^2 = \sum_{i=1}^n (y_i - (b_0 + b_1 x_i))^2$$



To find b_0 , b_1 , which **minimise** SSE

$$SSE = \sum_{i=1}^n e_i^2 = \sum_{i=1}^n (y_i - \hat{y}_i)^2 = \sum_{i=1}^n (y_i - (b_0 + b_1 x_i))^2$$

Theorem. The following b_0 and b_1 minimise SSE :
(Least Squares Estimator)

$$b_1 = \frac{SS_{xy}}{SS_x},$$

$$b_0 = \bar{y} - b_1 \bar{x},$$

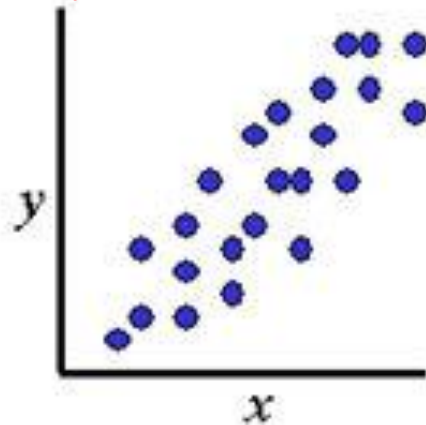
where $\bar{x} = \text{mean}(X)$, $\bar{y} = \text{mean}(Y)$

$$SS_x = \sum_{i=1}^n (x_i - \bar{x})^2 = \sum_{i=1}^n x_i^2 - \frac{1}{n} \left(\sum_{i=1}^n x_i \right)^2$$

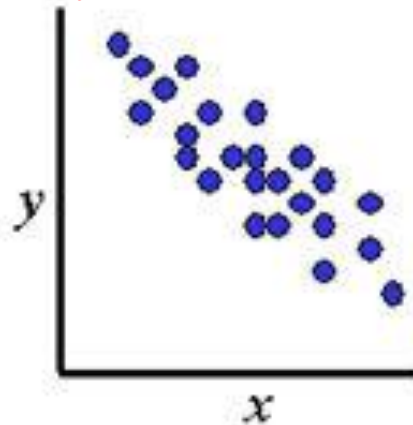
$$SS_{xy} = \sum_{i=1}^n (x_i - \bar{x})(y_i - \bar{y}) = \sum_{i=1}^n x_i y_i - \frac{1}{n} \left(\sum_{i=1}^n x_i \right) \left(\sum_{i=1}^n y_i \right).$$

What is b_1 's sign in the following relationships?

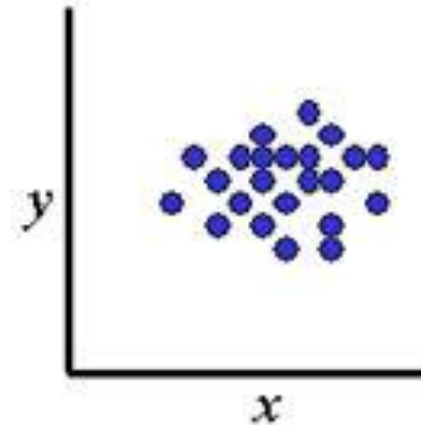
Positive b_1
As x increases,
 y increases



Negative b_1
As x increases,
 y decreases



$b_1=0$
No relation
between x and y



- It is important to check whether b_1 is significantly different that 0.
- How? Hypothesis testing.

Hypothesis testing for a linear relationship

Hypotheses:

$$H_0: b_1 = 0$$

$$H_1: b_1 \neq 0.$$



The test statistic for the existence of a linear relationship between X and Y can be calculated in Excel.

Armand's Pizza: Excel Output

Regression Statistics	
Multiple R	0.950122955
R Square	0.90273363
Adjusted R Square	0.890575334
Standard Error	13.82931669
Observations	10

Standard error for Y

Sample size

ANOVA					
	<i>df</i>	<i>SS</i>	<i>MS</i>	<i>F</i>	<i>Significance F</i>
Regression	1	14200	14200	74.24837	2.54887E-05
Residual	8	1530	191.25		
Total	9	15730			

	<i>Coefficients</i>	<i>Standard Error</i>	<i>t Stat</i>	<i>P-value</i>	<i>Lower 95%</i>	<i>Upper 95%</i>
Intercept	60	9.22603481	6.503336	0.000187	38.72471182	81.27528818
X Variable	5	0.580265238	8.616749	2.55E-05	3.661905096	6.338094904

Estimated b1

Standard error for b1

Test statistic based on confidence level defined

Confidence Interval for b1

Regression Results

$$Y = 60 + 5 * X$$

Interpretation of coefficients:

- $b_0 = 60$, is the Y-intercept of the line
- $b_1 = 5$, is the slope of the line
- $b_1 = 5$ means that for a unit increase in X-value, the value of Y increases by 5 units

Forecasting: fit a line using the Least Squares Method:

- $Y = 60 + 5X$
- Forecast sales for $X = 10$: $y = 60 + 5 * 10 = 110$

Significant Relationship

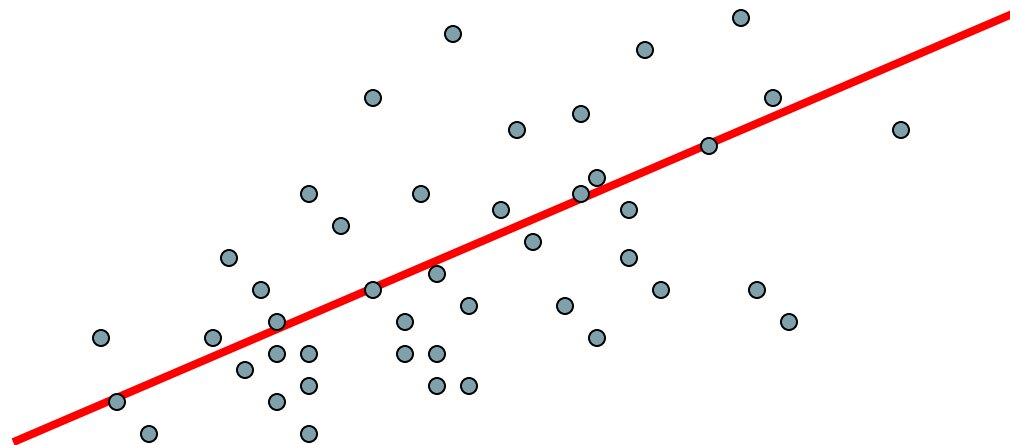
The coefficient is deemed significant at **95% confidence level**:

- If the p-value associated with a coefficient is less than **0.05** (the significance level)
- If the t-stat associated with a coefficient is larger than **1.96** (normal distribution) or **$t(n-2, 0.025)$** (for t distribution)
- If 0 is outside the 95% confidence interval

Then we can **reject the null hypothesis** ($b_1=0$), namely there is a relationship between X and Y

Is there a relationship?

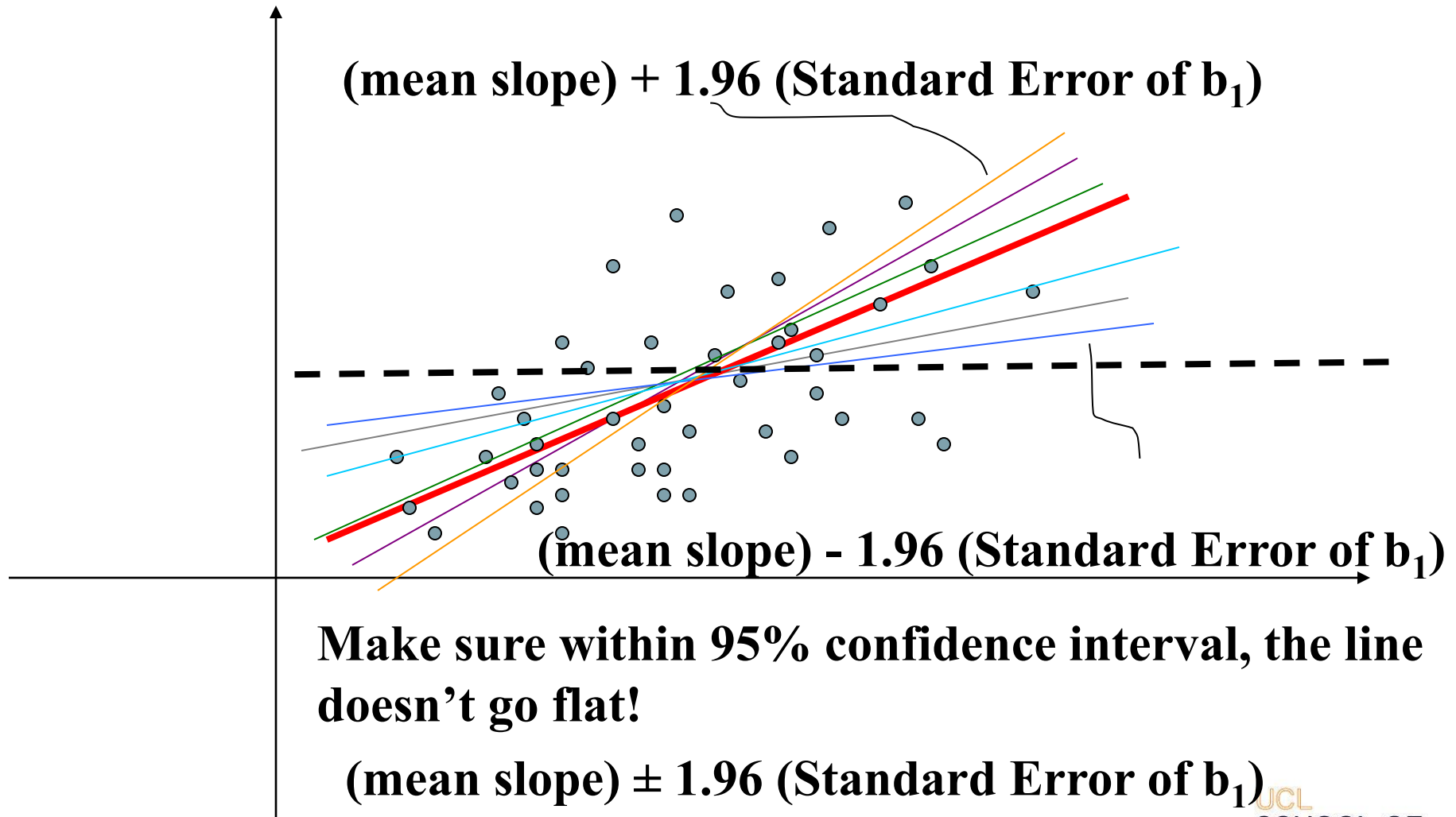
b_1 is the slope of the line.



Make sure within 95% confidence interval, the line doesn't go flat!

(mean slope) \pm 1.96 (Standard Error of b_1)

Is there a relationship?



Make sure within 95% confidence interval, the line doesn't go flat!

$(\text{mean slope}) \pm 1.96 (\text{Standard Error of } b_1)$

Uncertainty in Forecast

- Prediction Interval
 - With a 95% confidence level, the individual value of y for a given value of x will lie in the interval:

$$\hat{y} \pm 1.96 \times \text{standard error of the estimate}$$

When **t-distribution** is used (i.e., for **small sample size**), 1.96 needs to be replaced by $t_{(n-2, 0.025)}$

- For $x = 10$, the 95% prediction interval is:

$$110 \pm 2.306 \times 13.829$$

How Good Is the Fit?

- R^2 measures **how well the regression line fits the data**. In the pizza example, $R^2 = 0.90$. This means that **90% of the variation in sales is due to the variation in student population**. The other 10% of the variation remains unexplained. ($0 \leq R^2 \leq 1$)
- R^2 is one of several statistics that should be used in evaluating the quality of the regression model.

Summary

- Regression is useful in testing the relationship between two variables and in forecasting. Excel can generate the regression results.
- How to interpret them:
 1. Write the equation of the estimated line
 - $\text{Sales} = b_0 + b_1 * (\text{student population}) + \varepsilon$
 2. Is the coefficient, b_1 , significant? Check,
 - $p\text{-value} < 0.05$?
 - $t\text{-stats} > Z\text{-value from normal distribution (or } t\text{-value from } t\text{-distribution)}$
 - does the 95% interval for the coefficient contain 0?
 3. What is the point forecast for the mean and the 95% prediction interval?

$$\hat{y} \pm 1.96 \text{ standard error of the estimate}$$

When **t-distribution** is used (i.e., for **small sample size**), 1.96 needs to be replaced by $t_{(n-2, 0.025)}$

4. How good is the fit? Look at the R^2 .

Excel Example: Armand's Pizza

- Download data file from Moodle: Armand's Pizza.xlsx
- Draw scatter plot
- Run regression and interpret the results
- Plot predicted value and draw regression line.

*Hints. 1. For scatter charts in excel, go to INSERT -> Charts -> Scatter
2. For regression in excel, go to DATA -> Data Analysis -> Regression
3. Tick "Line Fit Plots" for the fitted line in regression .*

Mini Case: 2016 Rio Olympic Games

- Download Mini Case: 2016 Rio Olympic Games and the related data file from Moodle, and follow the instructions.

*Hints. 1. For scatter charts in excel, go to INSERT -> Charts -> Scatter
2. For regression in excel, go to DATA -> Data Analysis -> Regression
3. Tick “Line Fit Plots” for the fitted line in regression .*

Reference

Chapter 10 of:

Aczel, A., & J. Sounderpandian. 2008. Complete Business Statistics. McGraw-Hill/Irwin, Seventh Edition