

REGRESSION

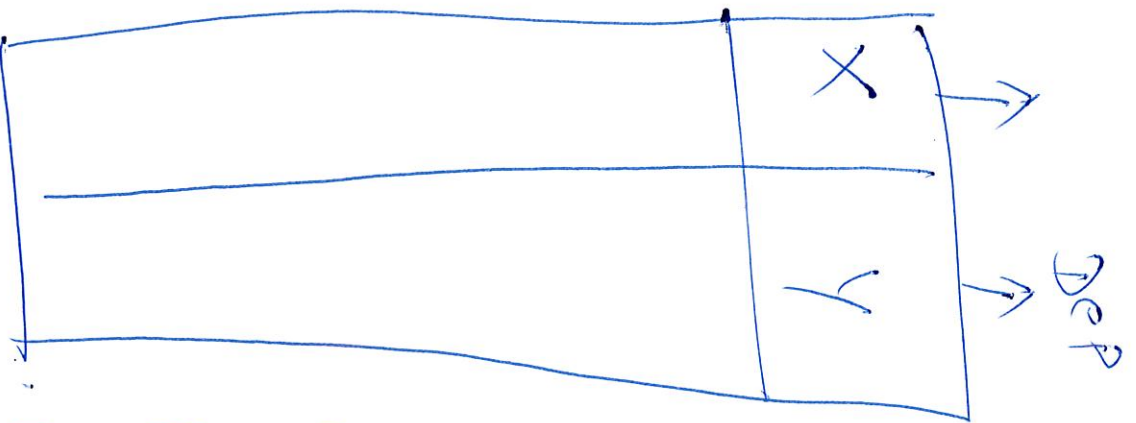
- * BASICS of REGRESSION
- * TESTS for SIGNIFICANCE
 - Adequacy of Model
- * ~~MULTIVARIATE~~ REGRESSION
- PLI
- * CHOICE of VARIABLES

$$Y = \beta_0 + \beta_1 x + \underbrace{\varepsilon}_{\text{Error}}$$

$$E(Y) = E(\beta_0 + \beta_1 x + \varepsilon)$$

$$E(Y) = \beta_0 + \beta_1 x \Rightarrow \text{REGRESSION EQUATION}$$

Linear Regression Equation



$$y = a x + c$$

~~$y = a x + c$~~

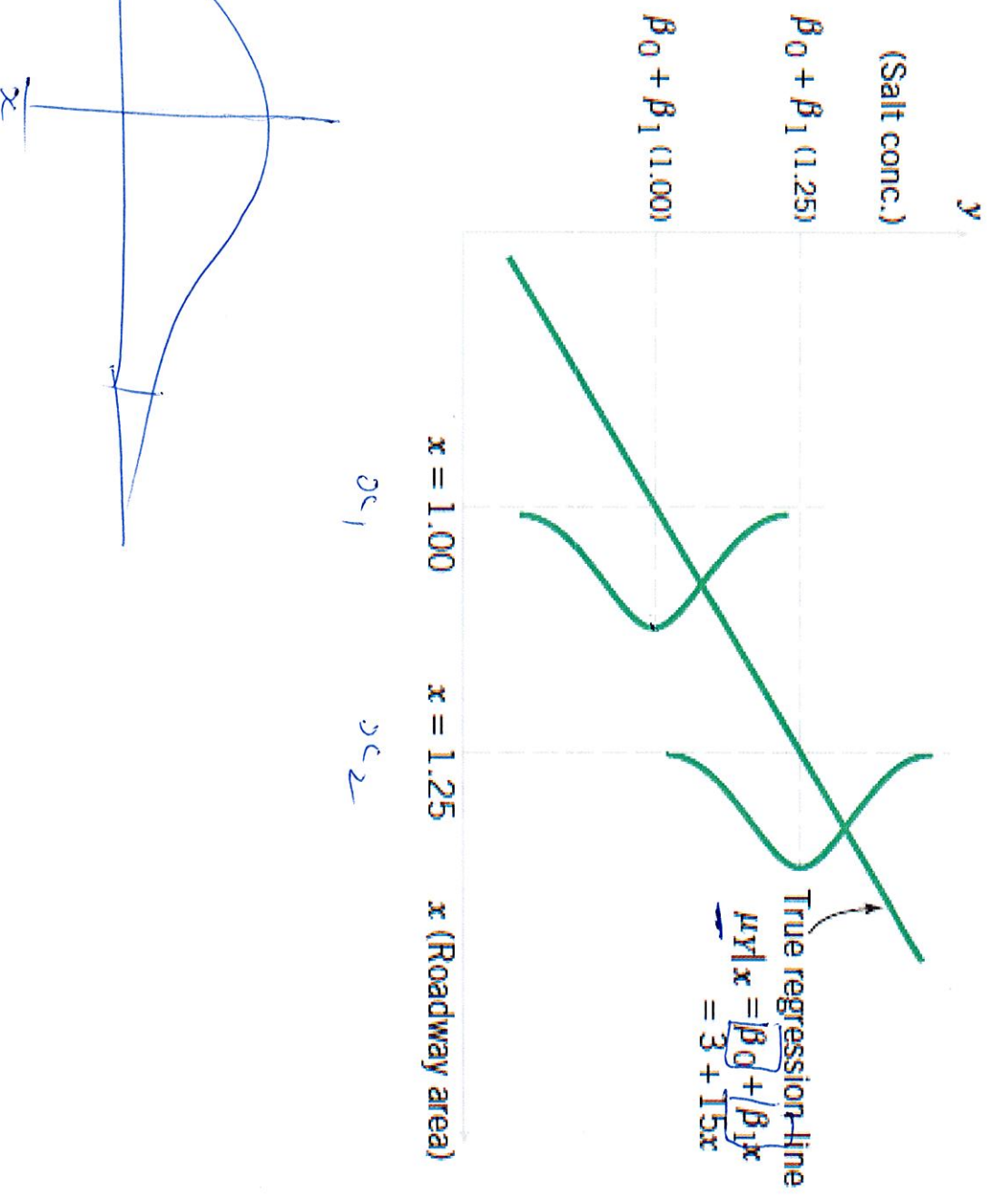
$$y = \beta_0 + \beta_1 x_1$$

$$y = \beta_0 + \beta_1 x_1 + \beta_2 x_2$$

$$y = \beta_0 + \beta_1 x_1 + \beta_2 x_1^2 + \beta_3 x_3$$

6-1 Introduction To Empirical Models

Figure 6-2 The distribution of Y for a given value of x for the salt concentration—roadway area data.



6-1 Introduction To Empirical Models

We think of the regression model as an empirical model.

Suppose that the mean and variance of ϵ are 0 and σ^2 , respectively, then

$$E(Y|x) = E(\beta_0 + \beta_1 x + \epsilon) = \beta_0 + \beta_1 x + E(\epsilon) = \beta_0 + \beta_1 x$$

The variance of Y given x is

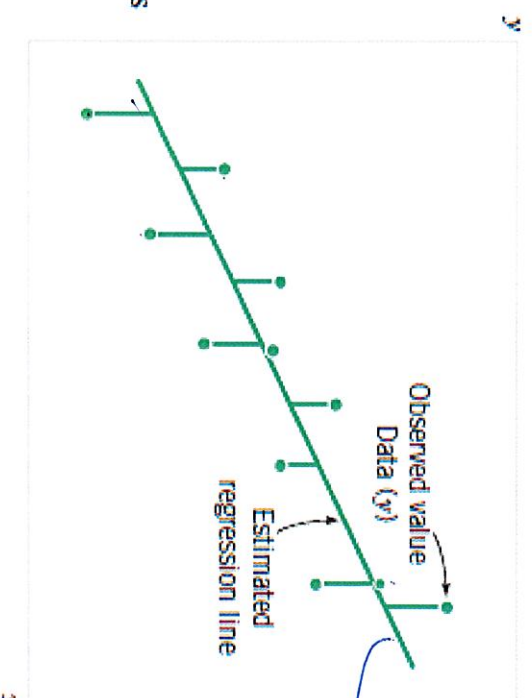
$$V(Y|x) = V(\beta_0 + \beta_1 x + \epsilon) = V(\beta_0 + \beta_1 x) + V(\epsilon) = 0 + \sigma^2 = \sigma^2$$

6-2 Simple Linear Regression

6-2.1 Least Squares Estimation

- Suppose that we have n pairs of observations $(x_1, y_1), (x_2, y_2), \dots, (x_n, y_n)$.
- The **method of least squares** is used to estimate the parameters, β_0 and β_1 by minimizing the sum of the squares of the vertical deviations in Figure 6-6.

Figure 6-6 Deviations of the data from the estimated regression model.

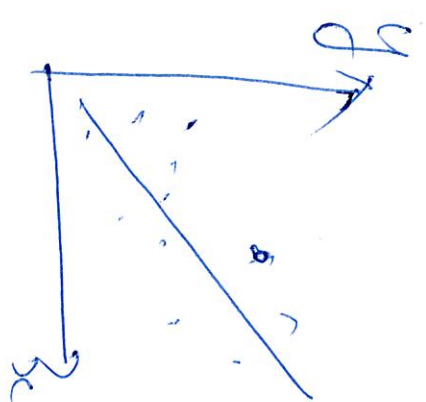


$$\begin{aligned} &\rightarrow E(y) \\ &= \beta_0 + \beta_1 x \end{aligned}$$

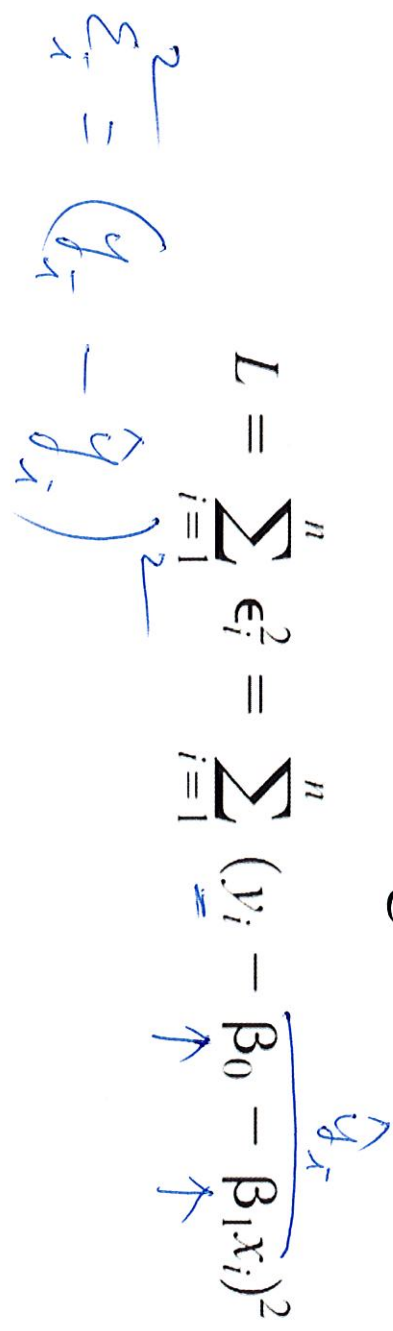
6-2 Simple Linear Regression

6-2.1 Least Squares Estimation

- Using Equation 6-8, the n observations in the sample can be expressed as

$$y_i = \beta_0 + \beta_1 x_i + \epsilon_i, \quad i = 1, 2, \dots, n$$


- The sum of the squares of the deviations of the observations from the true regression line is

$$L = \sum_{i=1}^n \epsilon_i^2 = \sum_{i=1}^n (y_i - \beta_0 - \beta_1 x_i)^2$$


6-2 Simple Linear Regression

6-2.1 Least Squares Estimation

The fitted or estimated regression line is therefore

$$\hat{y} = \hat{\beta}_0 + \hat{\beta}_1 x \quad (6-15)$$

Note that each pair of observations satisfies the relationship

$$y_i = \hat{\beta}_0 + \hat{\beta}_1 x_i + e_i, \quad i = 1, 2, \dots, n$$

where $e_i = y_i - \hat{y}_i$ is called the **residual**. The residual describes the error in the fit of the model to the i th observation y_i . Subsequently we will use the residuals to provide information about the **adequacy** of the fitted model.

6-2 Simple Linear Regression

6-2.1 Least Squares Estimation

EXAMPLE 6-1

Therefore, the least squares estimates of the slope and intercept are

$$\hat{\beta}_1 = \frac{S_{xy}}{S_x} = \frac{64.4082}{3.67068} = 17.5467$$

and

$$\hat{\beta}_0 = \bar{y} - \hat{\beta}_1 \bar{x} = 17.135 - (17.5467)(0.824) = 2.6765$$

The fitted simple linear regression model is

$$\hat{y} = 2.6765 + 17.5467x$$

↑ ↑

$\hat{\beta}_0$ $\hat{\beta}_1$

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6-2 Simple Linear Regression

6-2.1 Least Squares Estimation

Table 6-2 Minitab Regression Analysis Output for Salt Concentration and Roadway Data

Regression Analysis: Salt conc (y) versus Roadway area (x)

The regression equation is

Salt conc (y) = 2.68 + 17.5 Roadway area (x)

Predictor	Coef	SE Coef	T	P
Constant	2.6765 $\leftarrow \hat{\beta}_0$	0.8680	3.08	0.006
Roadway area	17.5467 $\leftarrow \hat{\beta}_1$	0.9346	18.77	0.000
S	1.791 $\leftarrow \hat{\sigma}$			
			R-Sq = 95.1%	
				R-Sq(adj) = 94.9%

t-test H_0
 H_1

Analysis of Variance

Source	DF	SS	MS	F	P
Regression	1	1130.1 $\leftarrow SS_R$	1130.1	352.46	0.000
Residual Error	(n-2) 18	57.7 $\leftarrow SS_E$	3.2 $\leftarrow \hat{\sigma}^2$		
Total	(n-2)+1 = 19	1187.9 $\leftarrow SS_T$			

$(n-3)+2$
 $(n-3)$
 2

$$y = \beta_0 + \beta_1 x_1 + \beta_2 x_2$$

$n = 20$

$$\hat{y} = \beta_0 + \beta_1 x$$

Eq 1.
Eq 2.