Mathematics 231

Lecture 7
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Announcements

Reading

■ Today	M&M 2.3	108-121
	1/1001/1 - 2.5	100-12

■ Next class M&M 2.3 117-119

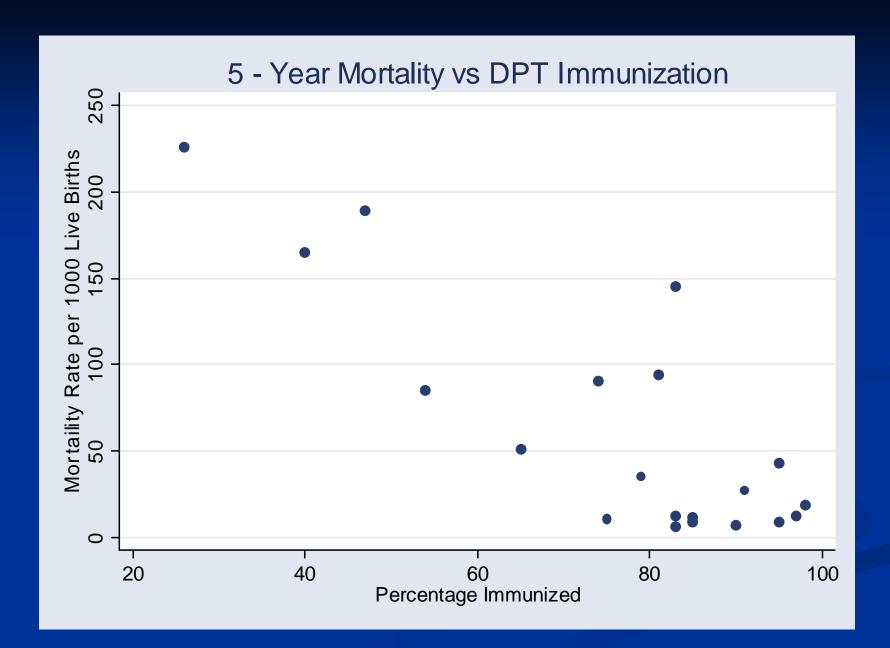
M&M 2.4 125-132

Linear Relationships & Regression

- Linear relationship between two variables.
- Response and explanatory variables.
- Regression line.
- Least squares criterion.

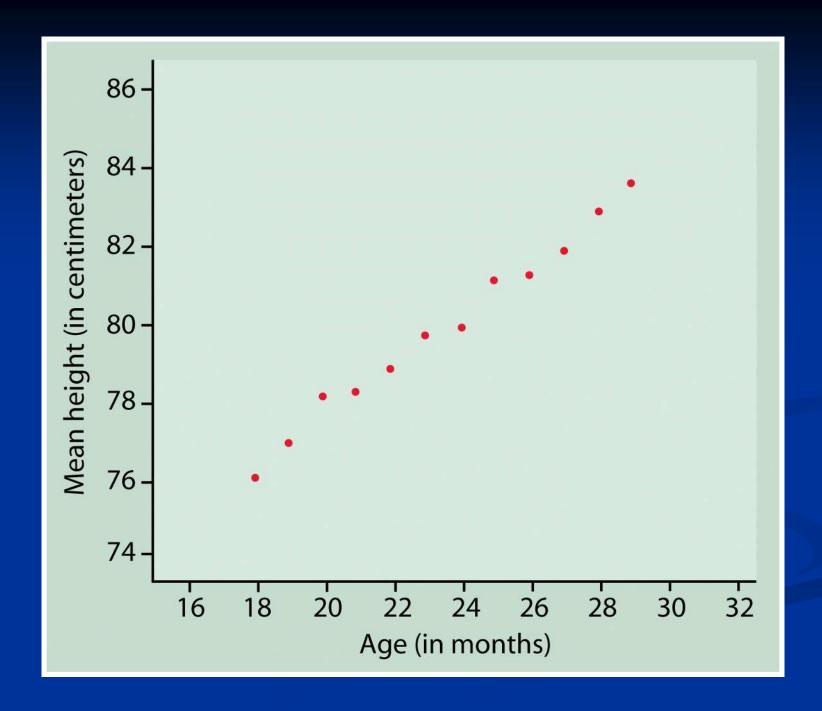
Response and Explanatory Variables

- A response variable, denoted as Y, measures the outcome of an experiment, survey, or study. Y is the variable we want to explain or predict.
- An explanatory variable, denoted as X, is a variable that may affect, explain or predict (but not necessarily cause) the response variable.



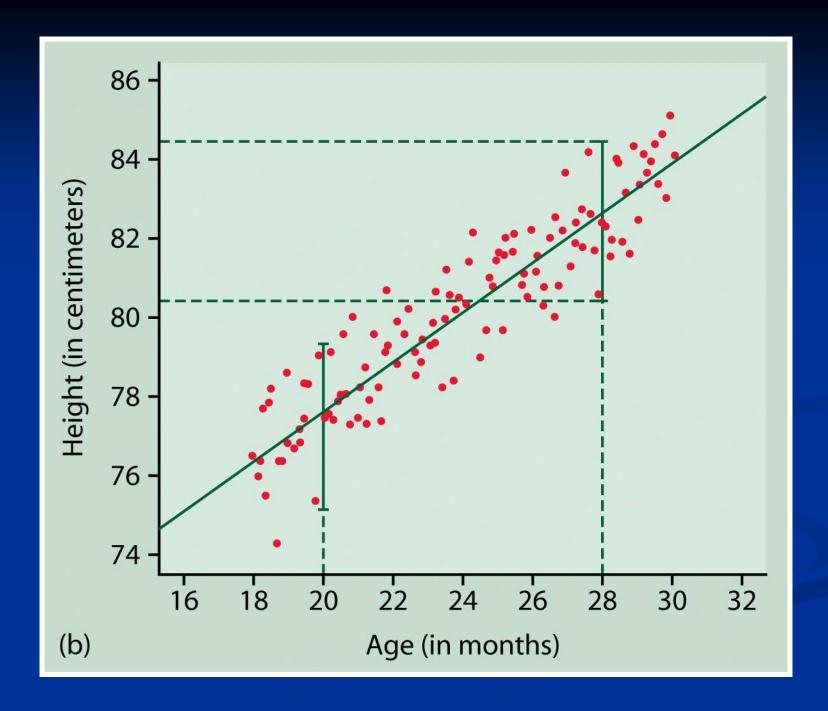
Example: Height and Age

TABLE 2.7	Mean height of Kalama children		
Age <i>x</i> in months	Height <i>y</i> in centimeters		
18	76.1		
19	77.0		
20	78.1		
21	78.2		
22	78.8		
23	79.7		
24	79.9		
25	81.1		
26	81.2		
27	81.8		
28	82.8		
29	83.5		



Conditional Distributions

- In general, we can consider the distribution of Y variables (e.g., height) for observations that satisfy some condition X = x (e.g., age equals 28 months).
- This is called the **conditional distribution of** Y given X = x.
- In a scatter plot, the conditional distribution of Y given X = x is the distribution of points in the vertical strip above a given value of x.



Conditional Mean

Conditional distributions have center, spread, and shape properties like all distributions.

The mean value of Y in the vertical strip above a given value x is called the **conditional mean of** Y given X = x.

Linear Regression

- Linear regression is used to explain or predict Y using X.
- It quantifies the relationship between the two variables in terms of a straight line.
- Suppose we have n pairs of Y and X,

$$(x_1, y_1), (x_2, y_2), (x_3, y_3), \dots, (x_n y_n)$$

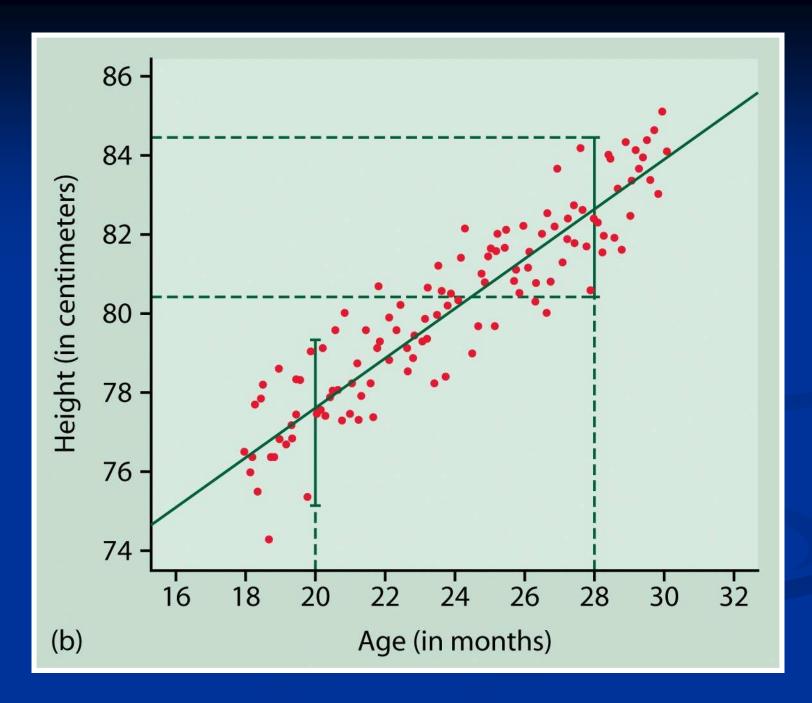
How can we find the straight line that best "fits" or describes these data?

Linear Regression

■ This line has an equation of the form:

$$\hat{y}_i = a + bx_i$$

where \hat{y}_i (y-hat) is the predicted
value of Y ,
 a is the y-intercept (the value of Y
when $X = 0$),
and b is the slope of the line.

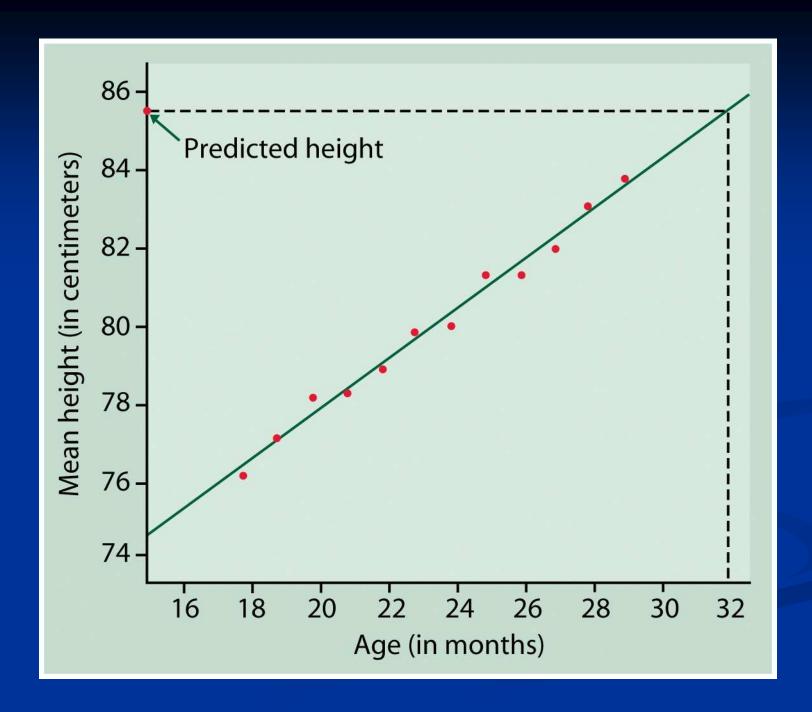


Definition 1

For any particular value, x_i, the **predicted** (or **fitted**) value is:

$$\hat{y}_i = a + bx_i$$

and is the y-value of the line at x_i .

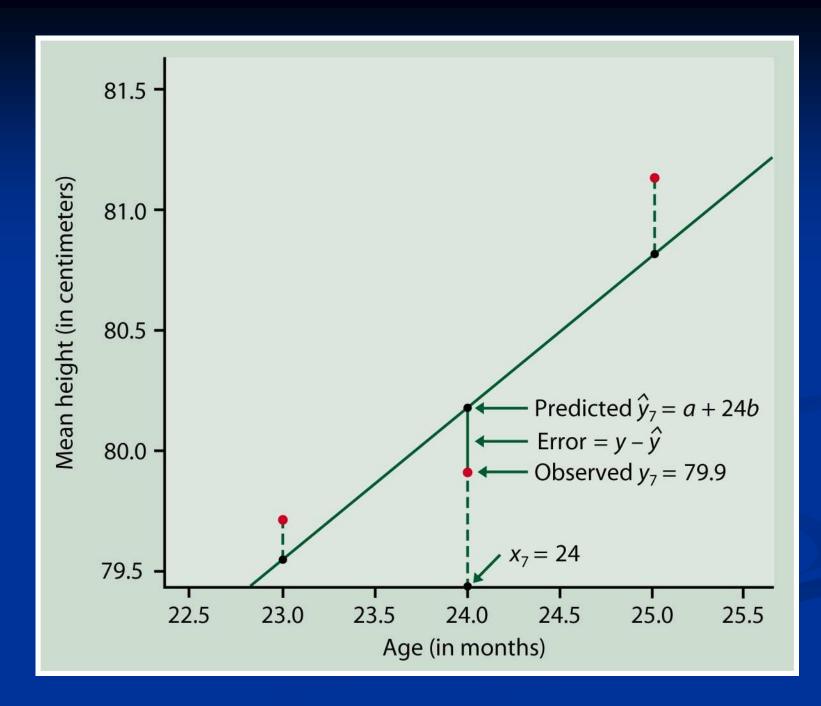


Definition 2

The vertical deviation from a point to the line (the difference between the observed and predicted values of *Y*, or the error) is called the residual.

$$y_i - \hat{y}_i = y_i - (a + bx_i)$$

residual = ε_i = observed y_i - predicted y_i



Least Squares Criterion

- The "best fit" line is defined as the line that minimizes the sum of the squared residuals.
- We want the values of *a* and *b* that minimizes the following quantity,

$$\sum_{i=1}^{n} (y_i - \hat{y}_i)^2 = \sum_{i=1}^{n} (y_i - (a + bx_i))^2$$

Least Squares Intercept and Slope

The values of *a* and *b* that minimize this quantity are,

$$b = r \frac{s_y}{s_x}$$

$$a = \overline{y} - b\overline{x}$$

where r is the correlation coefficient.

Assumptions

- The regression line estimates the conditional mean of Y given X=x for any point x if the following assumptions are met.
 - 1. Conditional mean of Y is a linear function of X.
 - 2. Conditional SD of Y is constant for all X.
- We often make an additional assumption:
 - 3. The conditional distribution of Y is a normal distribution for any value of x.

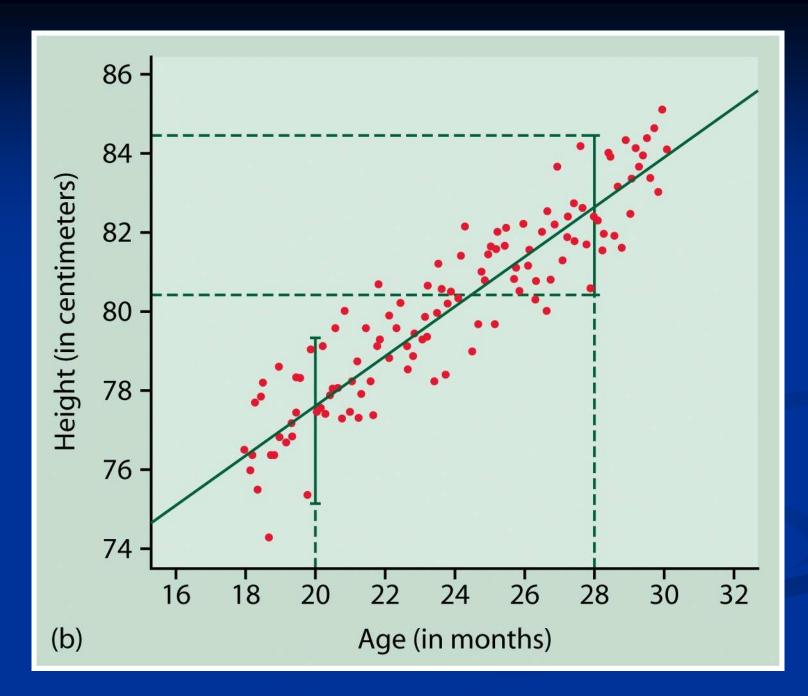


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Example: Height and Age

. regress height age

Source	SS	df 		MS		Number of obs F(1, 10)	= 12 = 880.00
Model Residual	57.6548678 .655171562	1 10		5548678 5517156		Prob > F	= 0.0000 = 0.9888
Total	58.3100394	11	5.30	091267			= .25596
height	Coef.	Std.	Err.	t	P> t	[95% Conf.	Interval]
age _cons	.6349653 64.92832	.0214		29.66 127.71	0.000 0.000	.5872726 63.79551	.682658 66.06112

