

## Announcements

- Reading
- Today M\&M 2.3 108-121
- Next class M\&M 2.3 117-119

M\&M 2.4 125-132

## Linear Relationships \& Regression

## Response and Explanatory Variables

- Linear relationship between two variables.
- Response and explanatory variables.
- Regression line.
- Least squares criterion.
- 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 <br> Kalama children |
| :---: | :---: |
| Age $x$ <br> in months | Height $y$ <br> 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 | 88.8 |
| 28 | 88.8 |
| 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 $\boldsymbol{Y}$ given $\boldsymbol{X}=\boldsymbol{x}$.
- 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 $\boldsymbol{Y}$ given $\boldsymbol{X}=\boldsymbol{x}$.


## 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 $X$,

$$
\left(x_{1}, y_{1}\right),\left(x_{2}, y_{2}\right),\left(x_{3}, y_{3}\right), \ldots,\left(x_{n} y_{n}\right)
$$

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


## Linear Regression

- This line has an equation of the form:
$\hat{y}_{i}=a+b x_{i}$
where $\hat{y}_{i}(\mathrm{y}$-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, $\mathrm{x}_{\mathrm{i}}$, the predicted (or fitted) value is:
$\hat{y}_{i}=a+b x_{i}$
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.
$y_{i}-\hat{y}_{i}=y_{i}-\left(a+b x_{i}\right)$
residual $=\varepsilon_{i}=$ 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,

$$
\sum_{i=1}^{n}\left(y_{i}-\hat{y}_{i}\right)^{2}=\sum_{i=1}^{n}\left(y_{i}-\left(a+b x_{i}\right)\right)^{2}
$$

## Least Squares Intercept and Slope

- The values of $a$ and $b$ that minimize this quantity are,
$b=r \frac{s_{y}}{s_{x}}$
$a=\bar{y}-b \bar{x}$
where $r$ is the correaltion 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 $x$.


## Example: Height and Age




