

THIS TIME: 2-independent samples

NEXT TIME: Correlation

→ Read LN pg L-214 → L-268

Today LN pg L-281

- Dichotomous Outcomes (redwood example)
 - 2 samples, CA & OR (2 diff designs)
 - Independent: 281 vs. 265 sample sizes
 - * like daphnia case study, except dichotomous
 - * The OR rate of infection is more than twice as high ($\hat{P}_1 = 3.4\%$ for CA & $\hat{P}_2 = 7.1\%$ for OR)

$$SE_{IID}(\hat{P}_1) = \frac{\sigma_1}{\sqrt{n_1}} = \frac{\sqrt{P_1(1-P_1)}}{\sqrt{n_1}} = \sqrt{\frac{P_1(1-P_1)}{n_1}}$$

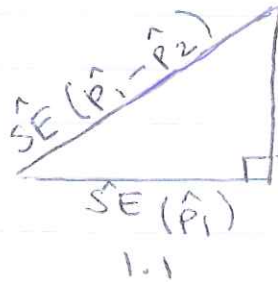
$$\hat{SE}_{IID}(\hat{P}_1) = \sqrt{\frac{\hat{P}_1(1-\hat{P}_1)}{n_1}} = 1.1\%$$

← Larger than we might like b/c dichotomous is causing more uncertainty

$$\hat{SE}_{IID}(\hat{P}_2) = \sqrt{\frac{\hat{P}_2(1-\hat{P}_2)}{n_2}} = 1.5\%$$

$$\hat{SE}_{\text{indep.}}(\hat{P}_1 - \hat{P}_2) = ?$$

$1.5 \leq \text{Hypotenuse} \leq 3.0$



$$\hat{SE}(\hat{P}_2) = 1.5$$

Shortcut formula

$$\hat{SE}_{\text{indep.}}(\hat{P}_1 - \hat{P}_2) = \sqrt{\left(\sqrt{\frac{\hat{P}_1(1-\hat{P}_1)}{n_1}}\right)^2 + \left(\sqrt{\frac{\hat{P}_2(1-\hat{P}_2)}{n_2}}\right)^2}$$

$$= \sqrt{\frac{\hat{P}_1(1-\hat{P}_1)}{n_1} + \frac{\hat{P}_2(1-\hat{P}_2)}{n_2}}$$

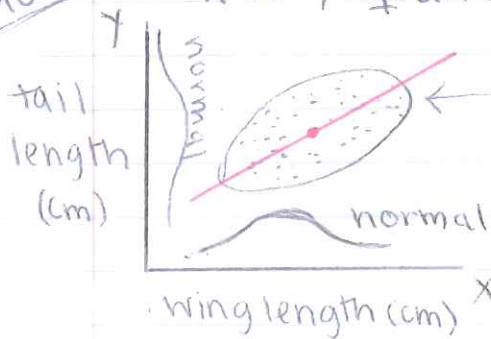
(2)

L-214

Simple Correlation & simple Linear Regression
Pg. L-221 (Bird's Tail & wing length)

* Studying the relationship b/w 2 variables
x & y quantitative continuous

1890's

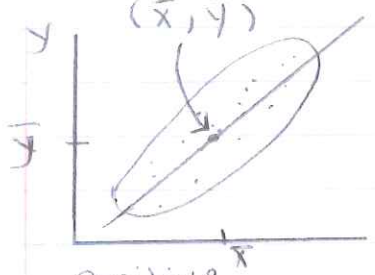


ellipse
Summary of bivariate normal distribution

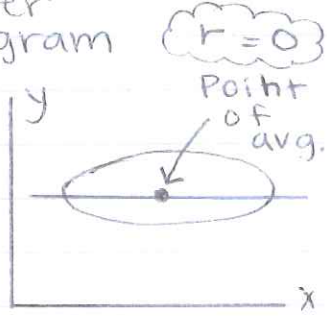
"two variables"

* WORKS OF Karl Pearson & Francis Galton

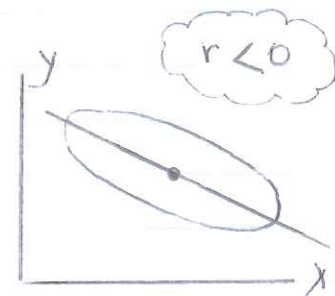
Scatter plot / scatter diagram



positive association



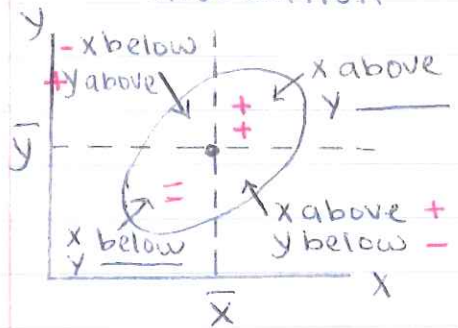
no (linear) association



negative association

$r = 0$
Point of avg.

$r < 0$



$\frac{x - \bar{x}}{s_x} > 0$ & $\frac{y - \bar{y}}{s_y} > 0$

$\frac{x - \bar{x}}{s_x} < 0$ & $\frac{y - \bar{y}}{s_y} < 0$

| | |
|----------|----------|
| y | x |
| y_1 | x_1 |
| y_2 | x_2 |
| \vdots | \vdots |
| \vdots | \vdots |

mean \bar{y} & \bar{x}
SD s_y & s_x

$\frac{1}{n} \sum_{i=1}^n \left(\frac{x_i - \bar{x}}{s_x^*} \right) \cdot \left(\frac{y_i - \bar{y}}{s_y^*} \right)$

$s_x^* = \sqrt{\frac{1}{n} \sum_{i=1}^n (x_i - \bar{x})^2}$
 $s_y^* = \sqrt{\frac{1}{n} \sum_{i=1}^n (y_i - \bar{y})^2}$

$\rightarrow r$

③

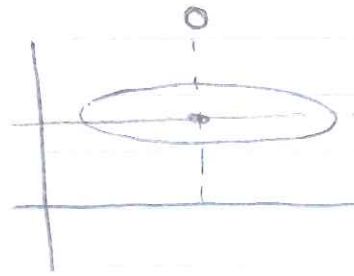
* r = Sample correlation (coefficient) b/w x & y

→ FACTS ABOUT (r)

$$-1 \leq r \leq +1$$

↑
Perfect
linearity
with (-) slope

↑
Perfect
linearity
with (+) slope



R-73