

# Development Statistics

## S11 Correlation

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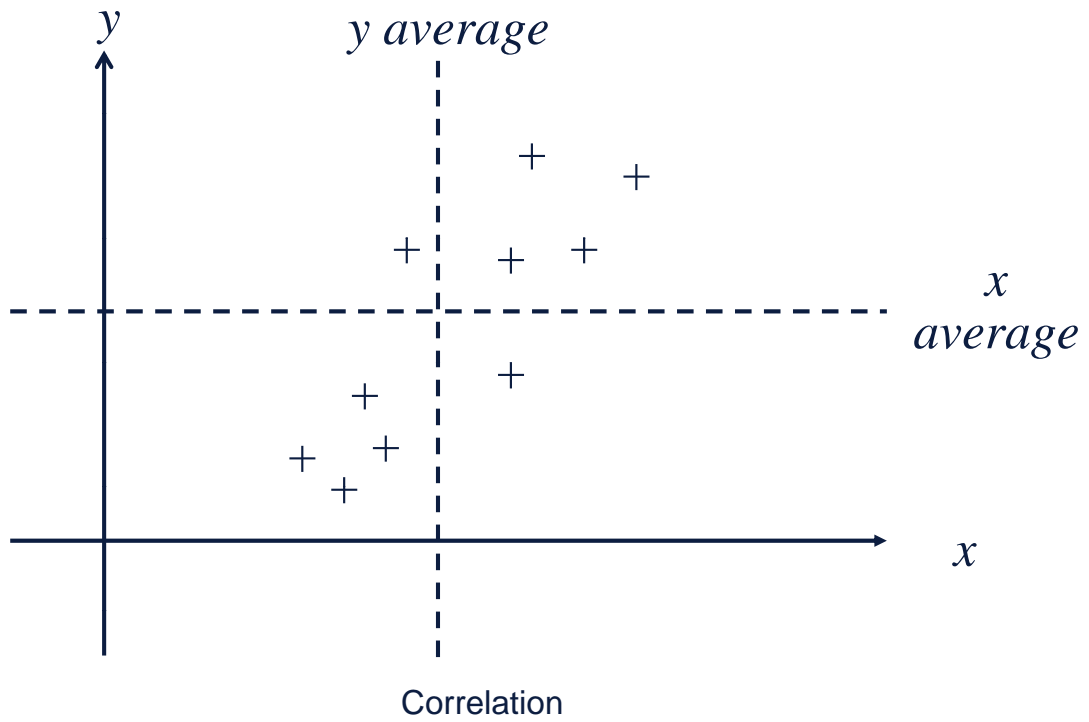
### Sample Covariance

- Positive correlation
- Negative correlation

$$\text{cov}(x, y) = s(x, y) = \frac{\sum_i (x_i - \bar{x})(y_i - \bar{y})}{n}$$

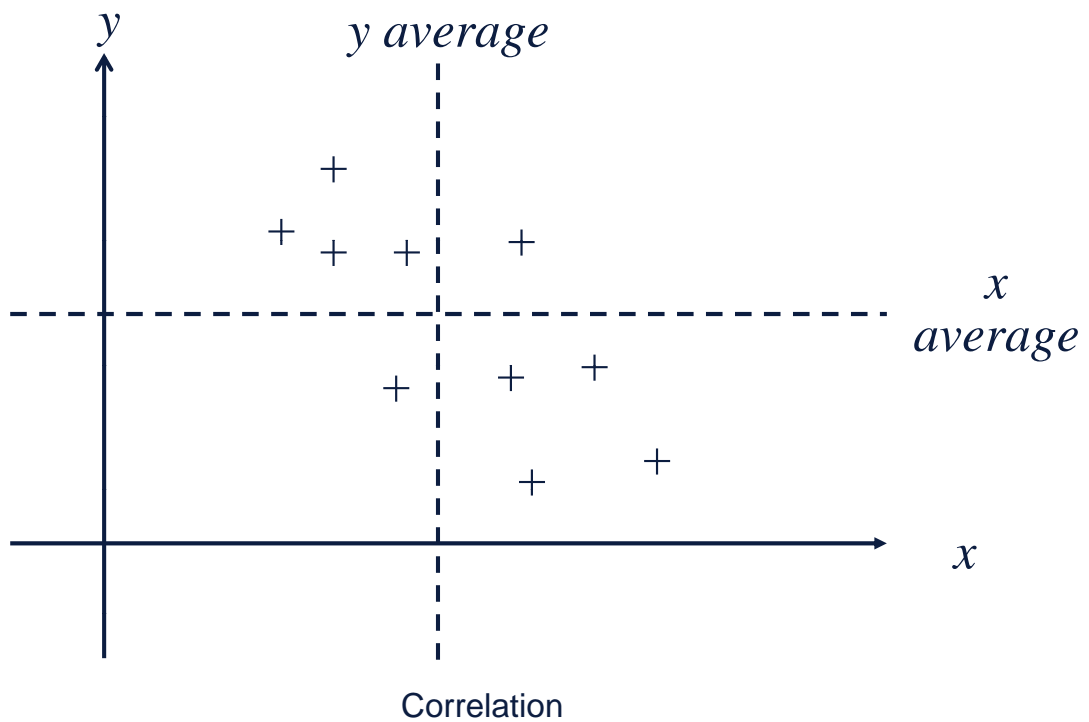
- Variance is a special case that  $x=y$

# Illustration 1



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# Illustration 2



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# Sample Correlation

- However, variance and covariance are unit-dependent
- Normalization of covariance

$$r(x, y) = \frac{\sum(x - \bar{x})(y - \bar{y})}{\sqrt{\sum(x - \bar{x})^2} \sqrt{\sum(y - \bar{y})^2}} = \frac{S(x, y)}{S(x)S(y)}$$

$$-1.0 \leq r(x, y) \leq 1.0$$

# Test of the correlation of the population

- Suppose the population follows bivariate normal distribution
- If the correlation of the population is 0
- $r$  : sample correlation,  $n$  : sample size

$$T(r) = \frac{r\sqrt{n-2}}{\sqrt{1-r^2}} \sim t(n-2)$$

## Correlation & t-value

- T-test of regression analysis

$$y_i = \alpha + \beta x_i + u_i$$

$$\frac{r\sqrt{n-2}}{\sqrt{1-r^2}} = \frac{b - \beta}{SE(b)} \sim t(n-2)$$

## Standard error of regression coefficient

- Regression line  $\hat{y}_i = a + bx_i$
- Residual  $y_i - \hat{y}_i = y_i - (a + bx_i) = e_i$
- Estimator of variance  $s^2 = \sum_i e_i^2 / (n-2)$
- Standard error  $s = \sqrt{\sum_i e_i^2 / (n-2)}$