Covariance and Correlation Are the two variables independent?

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Covariance and Correlation

Are taller people also heavier?

- Two stochastic variables: height *H* weight *W*
- We can observe them in pairs (H_i, W_i)
 - each individual i has a height and a weight



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THE 1 A

Scatter plot



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What do you expect?





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Measuring of Correlation

- Recall the variance $\operatorname{var}(X) = E((X \mu_X)^2)$
 - measures the spread of one variable
- Covariance $Cov(X, Y) = E((X \mu_X)(Y \mu_Y))$
 - not normalised
 - if σ_X or σ_Y is large, the covariance will have high absolute value
- Normalise → Correlation Coefficient

$$\rho = \frac{\operatorname{Cov}(X, Y)}{\sigma_X \sigma_Y}$$

Interpreting the Covariance



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Sample Correlation Coefficient

Population correlation coefficent

$$\rho = \frac{\operatorname{Cov}(X, Y)}{\sigma_X \sigma_Y}$$

Sample correlation coefficient

$$\hat{\rho} = \frac{\sum_{i=1}^{n} (X_{i} - \bar{X})(Y_{i} - \bar{Y})/(n-1)}{s_{X}s_{Y}}$$
$$\hat{\rho} = \frac{\sum_{i=1}^{n} (X_{i} - \bar{X})(Y_{i} - \bar{Y})}{\sqrt{\left[\sum_{i=1}^{n} (X_{i} - \bar{X})^{2}\right]\left[\sum_{i=1}^{n} (Y_{i} - \bar{Y})^{2}\right]}}$$

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Summary

$$\rho = \frac{E((X - \mu_X)(Y - \mu_Y))}{\sigma_X \sigma_Y}$$
$$\hat{\rho} = \frac{\sum_{i=1}^n (X_i - \bar{X})(Y_i - \bar{Y})}{\sqrt{\left[\sum_{i=1}^n (X_i - \bar{X})^2\right]\left[\sum_{i=1}^n (Y_i - \bar{Y})^2\right]}}$$

- Correlation is dependence between two variables
- The Correlation Coefficient ρ is the principal measure
 - $\rho = 0 \Rightarrow$ independent variables
 - $\rho > 0 \Rightarrow X$ tends to be large when Y is
 - *ρ* < 0 ⇒ X tends to be large when Y is small