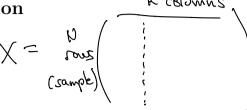
Variance, Covariance, and Correlation

Terminology review

- Samples and Features
- Tidy Data Matrix



Mean

The sample mean of a feature is

$$\mu_X = \frac{1}{N} \sum_{i=1}^{N} x_i$$

Variance

Definition: The (sample) variance of the feature measurements x_1, \ldots, x_n is

$$\sigma_X^2 = \frac{1}{N} \sum_{i=1}^N (x_i - \mu_X)^2 = \frac{1}{N} \left(\sum_{i=1}^N x_i^2 \right) - \mu_X^2$$

$$X_2 = \begin{bmatrix} X_i \\ \vdots \\ X_N \end{bmatrix}$$

$$E = \begin{bmatrix} 1 \\ 1 \\ \vdots \\ \vdots \\ X_N \end{bmatrix}$$

Covariance

Definition: If $X = (x_1, ..., x_N)$ and $Y = (y_1, ..., y_N)$ are two feature vectors then the (sample) covariance is

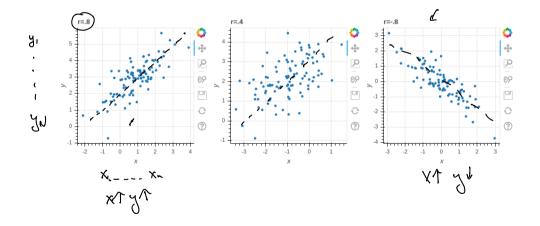
Correlation

Definition: Given feature vectors X and Y, the (sample) correlation coefficient r_{XY} is

$$r_{XY} = \frac{\sigma_{XY}}{\sigma_{XX}\sigma_{YY}} = \frac{(x - \mu_{x}) \cdot (x - \mu_{y} \in Y)}{||(x - \mu_{x} \in Y)||(x - \mu_{x} \in Y)||(x - \mu_{x} \in Y)||}$$

$$\sigma_{xY} = \frac{1}{N} \cdot (x - \mu_{x}) \cdot (Y - \mu_{Y})$$

$$\sigma_{xY} = \frac{1}{N} ||(x - \mu_{x} \in Y)||$$



The covariance matrix

Definition: Let X be an $N \times k$ data matrix, and let X_0 be its centered version. The (sample) covariance matrix is the $k \times k$ symmetric matrix

$$X_{0} = \begin{bmatrix} X_{0}^{T} X_{0} & X_{0} \\ X_{0}^{T} X_{0} & X_{0} \\ X_{0}^{T} X_{0} & X_{0}^{T} \\ X_{0}^{T} X_{0}^{T} X_{0} & X_{0}^{T} \\ X_{0}^{T} X_{0}^{T} X_{0} & X_{0}^{T} \\ X_{0}^{T} X_{0}^{T} X_{0}^{T} \\ X_{0}^{T} X_{0}^{T} X_{0}^{T} \\ X_{0}^{T} X_{0}^{T} X_{0}^{T} \\ X_{0}^{T} X_{0}^{T} X_{0}^{T} X_{0}^{T} X_{0}^{T} X_{0}^{T} \\ X_{0}^{T} X_{0}^{T} X_{0}^{T} X_{0}^{T} X_{0}^{T} \\ X_{0}^{T} X_$$

X = centered data matrix columns of Xo have mean X = & X, but subtract

the average of

each olumn from

that column. di diagonal entres of Do $d_{ii} = \frac{1}{N} \sum_{j=1}^{N} x_{ji}$ $d_{ii} = \frac{1}{N} \sum_{j=1}^{N} x_{ji}^{2} = \sigma_{ii}^{2}$

