Review of Linear Models Class 1-3

Robert Gulotty

University of Chicago

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Often the sample (co)variance appears in an alternative form:

$$\widehat{Var}(X) = \frac{1}{N} \sum_{i} (x_i - \bar{x})(x_i - \bar{x})$$

$$= \frac{1}{N} \sum_{i} (x_i - \bar{x})x_i$$

$$\widehat{Cov}(X, Y) = \frac{1}{N} \sum_{i} (x_i - \bar{x})(y_i - \bar{y})$$

$$= \frac{1}{N} \sum_{i} (x_i - \bar{x})y_i$$

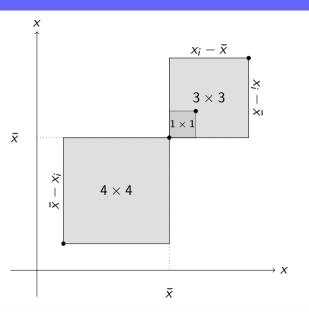
Algebra Demonstration

$$\widehat{Var}(X) = \frac{1}{N} \sum_{i} (x_{i} - \bar{x})(x_{i} - \bar{x})$$

$$= \frac{1}{N} \sum_{i} x_{i}^{2} - \bar{x}x_{i} - x_{i}\bar{x} + \bar{x}\bar{x}$$

$$= \frac{1}{N} \sum_{i} x_{i}^{2} - \bar{x} \frac{1}{N} \sum_{i} x_{i} - \bar{x} \frac{1}{N} \sum_{i} x_{i} + \frac{1}{N} \sum_{i} \bar{x}\bar{x}$$
(Expand the square)
$$= \frac{1}{N} \sum_{i} x_{i}^{2} - \bar{x} \frac{1}{N} \sum_{i} x_{i} - \bar{x} \frac{1}{N} \sum_{i} x_{i} + \bar{x}\bar{x}$$

$$= \frac{1}{N} \sum_{i} x_{i}^{2} - \bar{x} \frac{1}{N} \sum_{i} x_{i}$$
(Distribute the \sum)
$$= \frac{1}{N} \sum_{i} x_{i}^{2} - \bar{x} \frac{1}{N} \sum_{i} x_{i}$$
($\bar{x}\bar{x}$ cancel)
$$= \frac{1}{N} \sum_{i} (x_{i} - \bar{x})x_{i}$$

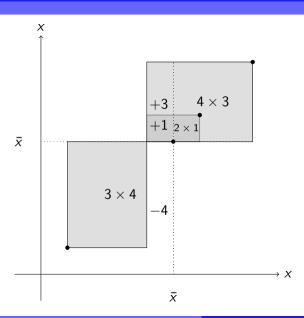


$$\widehat{Var}(X) = \frac{1}{N} \sum_{i} (x_i - \bar{x})(x_i - \bar{x})$$

$$= \frac{(-4*-4) + (1*1) + (3*3)}{4}$$

$$= \frac{26}{4}$$

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$$\widehat{Var}(X) = \frac{1}{N} \sum_{i} (x_i - \bar{x})(x_i - (\bar{x} - 1))$$

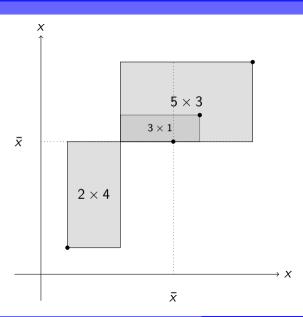
$$= \frac{(-4 * -3) + (1 * 2) + (3 * 4)}{4}$$

$$= \frac{26}{4}$$

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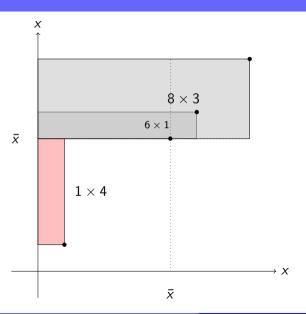
$$\widehat{Var}(X) = \frac{1}{N} \sum_{i} (x_i - \bar{x})(x_i - (\bar{x} - 2))$$

$$= \frac{(-4 * -2) + (1 * 3) + (3 * 5)}{4}$$

$$= \frac{26}{4}$$

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$$\widehat{Var}(X) = \frac{1}{N} \sum_{i} (x_i - \bar{x})(x_i - (\bar{x} - \bar{x}))$$

$$\widehat{Var}(X) = \frac{1}{N} \sum_{i} (x_i - \bar{x})(x_i)$$

$$= \frac{(-4 * 1) + (1 * 6) + (3 * 8)}{4}$$

$$= \frac{26}{4}$$

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How does y_k affect Cov(X, Y)?

$$\widehat{Cov}(X,Y) = \frac{1}{N} \sum_{i} (x_i - \bar{x})(y_i - \bar{y})$$

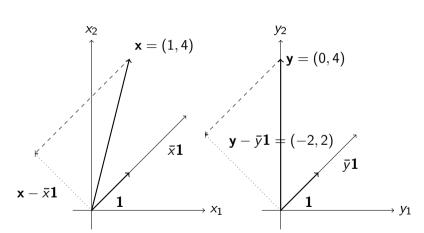
$$= \frac{1}{N} \sum_{i} (x_i - \bar{x})(y_i - \frac{1}{N} \sum_{j} y_j)$$

$$= \frac{1}{N} \sum_{i} (x_i - \bar{x})y_i$$

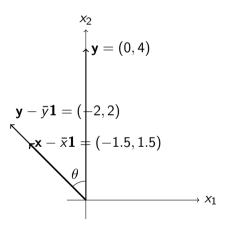
We can take the derivative with respect to y_k in only one term of the summation.

Geometry and the Dot product

Suppose $\mathbf{x} = (1,4), \ \bar{\mathbf{x}} = 2.5, \ \mathbf{y} = (0,4), \ \bar{\mathbf{y}} = 2$ $(\mathbf{x} - \bar{\mathbf{x}}\mathbf{1}) \cdot (\mathbf{y}) =$ -1.5 * 0 + 1.5 * 4 = 6 $(\mathbf{x} - \bar{\mathbf{x}}\mathbf{1}) \cdot (\mathbf{y} - \bar{\mathbf{y}}\mathbf{1}) =$ -1.5 * (-2) + 1.5 * 2 = 6



$\mathbf{a} \cdot \mathbf{b} = ||\mathbf{a}|| ||\mathbf{b}|| \cos \theta$



Note, $\mathbf{y} - \bar{\mathbf{y}}\mathbf{1} \perp \bar{\mathbf{y}}\mathbf{1}$, so [(0,4) - (0,0) - (-2,2)] is a right triangle and cos $\theta = \frac{\text{Adjacent}}{\text{Hypotenuse}}$

$$||\boldsymbol{y} - \bar{y}\mathbf{1}|| = ||\boldsymbol{y}||\cos\theta$$

Any vector minus its means lies on a hyperplane $\perp \mathbf{1}$ and so are parallel to one another ($\theta = 0$ or π)

$$\mathbf{y} - \bar{y}\mathbf{1} \parallel \mathbf{x} - \bar{x}\mathbf{1} \to \cos(0) = 1 \to$$

$$||\mathbf{x} - \bar{x}\mathbf{1}||||\mathbf{y}||\cos(\theta) =$$

$$\sqrt{4.5}\sqrt{16} * \cos(\pi/4) = 6$$

$$||\mathbf{x} - \bar{x}\mathbf{1}||||\mathbf{y} - \bar{y}\mathbf{1}||\cos(\theta) =$$

$$\sqrt{4.5} * \sqrt{8}\cos(0) = 6$$