

**Definition.** If  $X$  and  $Y$  are jointly distributed random variables with means  $\mu_X$  and  $\mu_Y$ , respectively, then  $E[(X - \mu_X)(Y - \mu_Y)]$  is called the *covariance* of  $X$  and  $Y$  and is denoted  $\sigma_{XY}$ ,  $\text{cov}(X, Y)$ , or  $C(X, Y)$ .

**Theorem 1.**  $C(X, Y) = E(XY) - E(X)E(Y)$

1. Calculate  $C(X, Y)$  for  $X$  and  $Y$  with joint probability density  $f(x, y) = \begin{cases} 4xy & \text{if } 0 < x < 1, 0 < y < 1 \\ 0 & \text{elsewhere} \end{cases}$

2. Calculate  $C(X, Y)$  for  $X$  and  $Y$  with joint probability density  $f(x, y) = \begin{cases} \frac{6}{5}(x^2 + y) & \text{if } 0 < x < 1, 0 < y < 1 \\ 0 & \text{elsewhere} \end{cases}$

**3.** Let  $X$  be a random variable and let  $a$  and  $b$  be constants.

a) Show that  $V(aX + b) = a^2V(X)$

b) If  $M_{aX+b}(t) = E \left[ e^{(aX+b)t} \right]$ , show that  $M_{aX+b}(t) = e^{bt} M_X(at)$ .