

Figure 23.3: The centered data points of Example 23.9.

Therefore, the vector  $Y \in \mathbb{R}^n$  consisting of the coordinates of the projections of  $X_1, \ldots, X_n$  onto the line spanned by v is given by Y = Xv, and this is the linear combination

$$Xv = v_1C_1 + \cdots + v_dC_d$$

of the columns of X (with  $v = (v_1, \ldots, v_d)$ ).

Observe that because  $\mu_j$  is the mean of the vector  $C_j$  (the jth column of X), we get

$$\overline{Y} = \overline{Xv} = v_1 \mu_1 + \dots + v_d \mu_d,$$

and so the centered point  $Y - \overline{Y}$  is given by

$$Y - \overline{Y} = v_1(C_1 - \mu_1) + \dots + v_d(C_d - \mu_d) = (X - \mu)v.$$

Furthermore, if Y = Xv and Z = Xw, then

$$cov(Y, Z) = \frac{((X - \mu)v)^{\top}(X - \mu)w}{n - 1}$$
$$= v^{\top} \frac{1}{n - 1} (X - \mu)^{\top} (X - \mu)w$$
$$= v^{\top} \Sigma w,$$

where  $\Sigma$  is the covariance matrix of X. Since  $Y - \overline{Y}$  has zero mean, we have

$$\operatorname{var}(Y) = \operatorname{var}(Y - \overline{Y}) = v^{\top} \frac{1}{n-1} (X - \mu)^{\top} (X - \mu) v.$$