

PCA finds a representation (through linear transformation) $\mathbf{z} = \mathbf{x}^\top \mathbf{W}$ where $\text{Var}[\mathbf{z}]$ is diagonal.

In section 2.12, we saw that the principal components of a design matrix \mathbf{X} are given by the eigenvectors of $\mathbf{X}^\top \mathbf{X}$. From this view,

$$\mathbf{X}^\top \mathbf{X} = \mathbf{W} \mathbf{\Lambda} \mathbf{W}^\top. \quad (5.86)$$

In this section, we exploit an alternative derivation of the principal components. The principal components may also be obtained via the singular value decomposition. Specifically, they are the right singular vectors of \mathbf{X} . To see this, let \mathbf{W} be the right singular vectors in the decomposition $\mathbf{X} = \mathbf{U} \mathbf{\Sigma} \mathbf{W}^\top$. We then recover the original eigenvector equation with \mathbf{W} as the eigenvector basis:

$$\mathbf{X}^\top \mathbf{X} = (\mathbf{U} \mathbf{\Sigma} \mathbf{W}^\top)^\top \mathbf{U} \mathbf{\Sigma} \mathbf{W}^\top = \mathbf{W} \mathbf{\Sigma}^2 \mathbf{W}^\top. \quad (5.87)$$

The SVD is helpful to show that PCA results in a diagonal $\text{Var}[\mathbf{z}]$. Using the SVD of \mathbf{X} , we can express the variance of \mathbf{X} as:

$$\text{Var}[\mathbf{x}] = \frac{1}{m-1} \mathbf{X}^\top \mathbf{X} \quad (5.88)$$

$$= \frac{1}{m-1} (\mathbf{U} \mathbf{\Sigma} \mathbf{W}^\top)^\top \mathbf{U} \mathbf{\Sigma} \mathbf{W}^\top \quad (5.89)$$

$$= \frac{1}{m-1} \mathbf{W} \mathbf{\Sigma}^\top \mathbf{U}^\top \mathbf{U} \mathbf{\Sigma} \mathbf{W}^\top \quad (5.90)$$

$$= \frac{1}{m-1} \mathbf{W} \mathbf{\Sigma}^2 \mathbf{W}^\top, \quad (5.91)$$

where we use the fact that $\mathbf{U}^\top \mathbf{U} = \mathbf{I}$ because the \mathbf{U} matrix of the singular value decomposition is defined to be orthogonal. This shows that if we take $\mathbf{z} = \mathbf{x}^\top \mathbf{W}$, we can ensure that the covariance of \mathbf{z} is diagonal as required:

$$\text{Var}[\mathbf{z}] = \frac{1}{m-1} \mathbf{Z}^\top \mathbf{Z} \quad (5.92)$$

$$= \frac{1}{m-1} \mathbf{W}^\top \mathbf{X}^\top \mathbf{X} \mathbf{W} \quad (5.93)$$

$$= \frac{1}{m-1} \mathbf{W}^\top \mathbf{W} \mathbf{\Sigma}^2 \mathbf{W}^\top \mathbf{W} \quad (5.94)$$

$$= \frac{1}{m-1} \mathbf{\Sigma}^2, \quad (5.95)$$

where this time we use the fact that $\mathbf{W}^\top \mathbf{W} = \mathbf{I}$, again from the definition of the SVD.