Suppose X is a vector of two random variables representing a roll of two balanced dice,

$$X = \left[\begin{array}{c} X_1 \\ X_2 \end{array} \right]$$

where X_1 and X_2 each take the values 1, 2, 3, 4, 5, 6.

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When a pair of dice are rolled, usually we are more interested in the sum of the individual outcomes than the outcomes themselves.

In this case, we define a new random variable Y as the sum of X_1 and X_2 :

$$Y = X_1 + X_2$$

We often need to determine the expected value E(Y) and variance V(Y) for this new variable.

We will assume that we know the expected value μ of the random vector X

$$\mu = \left[\begin{array}{c} \mu_1 \\ \mu_2 \end{array} \right] = \left[\begin{array}{c} E(X_1) \\ E(X_2) \end{array} \right]$$

and its variance-covariance matrix V:

$$V = \begin{bmatrix} \sigma_1^2 & \sigma_{12} \\ \sigma_{12} & \sigma_2^2 \end{bmatrix}$$

The most succinct way to determine E(Y) and V(Y) is with matrix algebra. Think of the sum as the matrix product

$$\begin{bmatrix} 1,1 \end{bmatrix} \begin{bmatrix} X_1 \\ X_2 \end{bmatrix} = 1 \cdot X_1 + 1 \cdot X_2 = X_1 + X_2$$

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More generally, we can represent any **linear combination** of X_1 and X_2

$$\begin{bmatrix} a_1, a_2 \end{bmatrix} \begin{bmatrix} X_1 \\ X_2 \end{bmatrix} = a \cdot X_1 + b \cdot X_2 = a_1 X_1 + a_2 X_2$$

If we write the coefficients $[a_1, a_2]$ as a column vector a,

$$a = \left[\begin{array}{c} a_1 \\ a_2 \end{array} \right]$$

then the linear combination $Y = a_1X_1 + a_2X_2$ can be written in matrix form

$$Y = a'X = \begin{bmatrix} a_1, a_2 \end{bmatrix} \begin{bmatrix} X_1 \\ X_2 \end{bmatrix}$$

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The symbol a' represents the **transpose** of a, which we obtain by writing the columns of a as rows:

if
$$a = \begin{bmatrix} a_1 \\ a_2 \end{bmatrix}$$
 then $a' = [a_1, a_2]$

The following important result gives us a way to compute E(Y) and V(Y): Suppose

$$Y = a'X$$
 with $a = \begin{bmatrix} a_1 \\ a_2 \end{bmatrix}$ and $X = \begin{bmatrix} X_1 \\ X_2 \end{bmatrix}$

and

$$E(X) = \mu = \begin{bmatrix} \mu_1 \\ \mu_2 \end{bmatrix} \text{ and } V = \begin{bmatrix} \sigma_1^2 & \sigma_{12} \\ \sigma_{12} & \sigma_2^2 \end{bmatrix}$$

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Then

$$E(Y) = a'\mu$$
 and $V(Y) = a'Va$

Example: Suppose *X* is a two-dimensional random vector:

$$X = \begin{bmatrix} X_1 \\ X_2 \end{bmatrix} \quad E(X) = \mu = \begin{bmatrix} 1 \\ 3 \end{bmatrix} \quad \text{and} \quad V = \begin{bmatrix} 2 & 1 \\ 1 & 3 \end{bmatrix}$$

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then

$$E(Y) = a'\mu = [1,1] \begin{bmatrix} 1\\ 3 \end{bmatrix} = 1 \cdot 1 + 1 \cdot 3 = 4$$

and

$$v(Y) = a'Va = \begin{bmatrix} 1,1 \end{bmatrix} \begin{bmatrix} 2 & 1 \\ 1 & 3 \end{bmatrix} \begin{bmatrix} 1 \\ 1 \end{bmatrix}$$

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$$= [3,4] \begin{bmatrix} 1\\1 \end{bmatrix} = 7$$

and

$$v(Y) = a'Va = \begin{bmatrix} 1, 1 \end{bmatrix} \begin{bmatrix} 2 & 1 \\ 1 & 3 \end{bmatrix} \begin{bmatrix} 1 \\ 1 \end{bmatrix}$$
$$= \begin{bmatrix} 3, 4 \end{bmatrix} \begin{bmatrix} 1 \\ 1 \end{bmatrix} = 7$$

Although the matrix notation can be confusing at first, the formulas

$$E(a'X) = a'\mu$$
 and $Variance(a'X) = a'Va$

are much simpler than their non-matrix counterparts.

Packages like MATLAB and GNU OCTAVE are designed to perform matrix computations and are optimized for it.

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While not primarily designed for this, R has some rudimentary facilities that will suffice for our purposes.

The preceding example would be programmed in R as:

```
mu=matrix(c(1,3),nrow=2)
V=matrix(c(2,1,1,3),nrow=2)
a=matrix(c(1,1),nrow=2)
t(a)%*%mu
t(a)%*%V%*%a
```

If X is a two-dimensional random vector and $Y = X_1 + X_2$, find E(Y) and V(Y) if

$$E(X) = \mu = \begin{bmatrix} 2\\4 \end{bmatrix}$$
 and $V = \begin{bmatrix} 4 & -2\\-2 & 5 \end{bmatrix}$

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Solution: In this case, a = [1, 1]' and

$$E(Y) = a'\mu = \begin{bmatrix} 1, 1 \end{bmatrix} \begin{bmatrix} 2\\ 4 \end{bmatrix} = 6$$
$$V(Y) = a'Va = \begin{bmatrix} 1, 1 \end{bmatrix} \begin{bmatrix} 4 & -2\\ -2 & 5 \end{bmatrix} \begin{bmatrix} 1\\ 1 \end{bmatrix} = 5$$

If X is a two-dimensional random vector and $Y = X_1 - X_2$, find E(Y) and V(Y) if

$$E(X) = \mu = \begin{bmatrix} 2\\4 \end{bmatrix}$$
 and $V = \begin{bmatrix} 4 & -2\\-2 & 5 \end{bmatrix}$

If X is a two-dimensional random vector and $Y = X_1 - X_2$, find E(Y) and V(Y) if

$$E(X) = \mu = \begin{bmatrix} 2\\4 \end{bmatrix}$$
 and $V = \begin{bmatrix} 4 & -2\\-2 & 5 \end{bmatrix}$

Solution: In this case, a = [1, -1]' and

$$E(Y) = a'\mu = \begin{bmatrix} 1, -1 \end{bmatrix} \begin{bmatrix} 2\\ 4 \end{bmatrix} = -2$$
$$V(Y) = a'Va = \begin{bmatrix} 1, -1 \end{bmatrix} \begin{bmatrix} 4 & -2\\ -2 & 5 \end{bmatrix} \begin{bmatrix} 1\\ -1 \end{bmatrix} = 13$$