

## Part2:

### 1. Covariance and Correlation matrices of U:

Part II.

1. Find the correlation and covariance matrices of  $U$ .

Given:  $f_{xy}(x,y) = \begin{cases} x + \frac{3}{2}y^2, & 0 < x, y \leq 1 \\ 0, & \text{otherwise} \end{cases}$

Random vector,  $U = \begin{pmatrix} X \\ Y \end{pmatrix}$

Solution:

Marginal Pdfe of  $x$  and  $y$

$$\Rightarrow f_x(x) = \int_0^1 f_{xy}(x,y) dy$$

$$= \int_0^1 x + \frac{3}{2}y^2 dy$$

$$= x + \frac{3}{2} \left[ \frac{y^3}{3} \right]_0^1, \text{ for } 0 < x < 1$$

$$f_x(x) = x + \frac{1}{2}, \text{ for } 0 < x < 1$$

$$\Rightarrow f_y(y) = \int_0^1 f_{xy}(x,y) dx$$

$$= \int_0^1 x + \frac{3}{2}y^2 dx$$

$$= \left[ \frac{x^2}{2} \right]_0^1 + \frac{3}{2}y^2$$

$$f_y(y) = \frac{1}{2} + \frac{3}{2}y^2, \text{ for } 0 < y < 1$$

$$\begin{aligned}
 E[x] &= \int_0^1 x f_x(x) dx \\
 &= \int_0^1 x \left[ x + \frac{1}{2} \right] dx \\
 &= \int_0^1 x^2 + \frac{x}{2} dx \\
 &= \left[ \frac{x^3}{3} + \frac{x^2}{4} \right]_0^1 \\
 &= \frac{1}{3} + \frac{1}{4} = \frac{7}{12}
 \end{aligned}$$

$$E[x] = 7/12$$

$$\begin{aligned}
 E[y] &= \int_0^1 y f_y(y) dy \\
 &= \int_0^1 y \left[ \frac{1}{2} + \frac{3}{2} y^2 \right] dy \\
 &= \int_0^1 \frac{y}{2} + \frac{3y^3}{2} dy \\
 &= \left[ \frac{y^2}{4} + \frac{3y^4}{8} \right]_0^1 \\
 &= \frac{1}{4} + \frac{3}{8} = \frac{5}{8}
 \end{aligned}$$

$$E[y] = 5/8$$

$$\begin{aligned}
 E[x^2] &= \int_0^1 x^2 f_x(x) dx \\
 &= \int_0^1 x^3 + \frac{x^2}{2} dx \\
 &= \left[ \frac{x^4}{4} + \frac{x^3}{6} \right]_0^1 \\
 &= \frac{1}{4} + \frac{1}{6}
 \end{aligned}$$

$$E[x^2] = 5/12$$

$$\begin{aligned}
 E[y^2] &= \int_0^1 y^2 f_y(y) dy \\
 &= \int_0^1 \frac{y^2}{2} + \frac{3}{2} y^4 dy \\
 &= \left[ \frac{y^3}{6} + \frac{3y^5}{10} \right]_0^1
 \end{aligned}$$

$$\begin{aligned}
 &= \frac{1}{6} + \frac{3}{10} \\
 &= \frac{10+18}{60} = \frac{28}{60} \\
 &= \frac{14}{30}
 \end{aligned}$$

$$E[y^2] = 7/15$$

$$\text{Var}(x) = E[x^2] - [E[x]]^2$$

$$= \frac{5}{12} - \left[ \frac{7}{12} \right]^2$$

$$= \frac{5}{12} - \frac{49}{144}$$

$$= \frac{5 \times 12 - 49}{144}$$

$$= \frac{11}{144}$$

$$\text{Var}(y) = E[y^2] - [E[y]]^2$$

$$= \frac{7}{15} - \left( \frac{5}{8} \right)^2$$

$$= \frac{7}{15} - \frac{25}{64}$$

$$= \frac{448 - 375}{960}$$

$$= \frac{73}{960}$$

$$\begin{aligned}
E[XY] &= \int_0^1 \int_0^1 xy f_{XY}(x,y) dx dy \\
&= \int_0^1 \int_0^1 xy \left[ x + \frac{3}{2}y^2 \right] dx dy \\
&= \int_0^1 \int_0^1 x^2 y + \frac{3}{2} x y^3 dx dy \\
&= \int_0^1 \left[ \frac{x^3 y}{3} + \frac{3 x^2 y^4}{4} \right]_0^1 dy \\
&= \int_0^1 \frac{y}{3} + \frac{3 y^4}{4} dy \\
&= \left[ \frac{y^2}{6} + \frac{3 y^5}{16} \right]_0^1 \\
&= \frac{1}{6} + \frac{3}{16} = \frac{8+9}{48}
\end{aligned}$$

$$E[XY] = \frac{17}{48}$$

$$\begin{aligned}
\text{We know, } \text{cov}(x, y) &= E[XY] - E[X] E[Y] \\
&= \frac{17}{48} - \left[ \left( \frac{7}{12} \right) \left( \frac{5}{8} \right) \right] \\
&= -\frac{1}{96}
\end{aligned}$$

The correlation matrix is given as,

$$\begin{aligned}
R_U &= E[U U^T] = \begin{bmatrix} E[X^2] & E[XY] \\ E[XY] & E[Y^2] \end{bmatrix} \\
&= \begin{bmatrix} \frac{5}{12} & \frac{17}{48} \\ \frac{17}{48} & \frac{7}{15} \end{bmatrix}
\end{aligned}$$

The covariance matrix is given as,

$$C_U = \begin{bmatrix} \text{var}(X) & \text{cov}(X, Y) \\ \text{cov}(Y, X) & \text{var}(Y) \end{bmatrix} = \begin{bmatrix} 1/144 & -1/96 \\ -1/96 & 73/960 \end{bmatrix}$$

**2and 3) Generation of 1000-sample vector series Xs with same covariance matrix as U:**

```
clear all;

fxy = @(x,y) x+(3/2)*y.^2;
% marginal pdf
fx = @(x)integral(@(y)fxy,0,1);
fy = @(y)integral(@(x)fxy,0,1);

%calculate E[X]
fx1= @(x) x.^2.+(x./2);
Ex= integral(fx1,0,1);

%calculate E[X^2]
fx2= @(x) x.*x.^2.+(x./2));
Ex2= integral(fx2,0,1);

%calculate var[X]
Var_X=Ex2-(Ex^2);

%calculate E[Y]
fy1= @(y) y./2 +(3/2)*y.^3;
Ey= integral(fy1,0,1);

%calculate E[Y^2]
fy2= @(y) y.*y./2 +(3/2)*y.^3);
Ey2= integral(fy2,0,1);

%calculate var[Y]
Var_y=Ey2-(Ey^2);

%calculate E[XY]
fxy1=@(x,y) x.^2.*y+(3/2)*x.*y.^3;
Exy= integral2(fxy1,0,1,0,1);
cov_xy= Exy-(Ex*Ey);

%correlation matrix of U
R_U=[Ex2 Ey; Exy Ey2];
%covariance matrix of U
cov_U=[Var_X cov_xy; cov_xy Var_y];
disp(cov_U);
%cholesky factorization and Xs generation
Upper_tri=chol(cov_U,"upper");
Lower_tri=chol(cov_U,"lower");
Xs=Upper_tri'*randn(2,500);
Cov_Xs=(cov(Xs'));
disp(Xs);

diff=cov_U-Cov_Xs;
```

**Covariance of U:**

Editor - part2\_final.m

Variables - cov\_U

Workspace

Name	Value
cov_U	[0.0764, -0.0104; -0.0104, 0.0760]
Cov_Xs	[0.0732, -0.0140; -0.0140, 0.0794]
cov_xy	-0.0104
diff	[0.0032, 0.0036; 0.0036, 0.0032]
Ex	0.5833
Ex2	0.4167
Exy	0.3542
Ey	0.6250
Ey2	0.4667
fx	$\int x^2 \cdot f(x) dx$
fx1	$\int x^2 \cdot f(x) dx / 2$
fx2	$\int x^2 \cdot f(x) dx + (3/2) \cdot \int x \cdot f(x) dx$
fxy	$\int x \cdot y \cdot f(x, y) dx dy$
fxy1	$\int x \cdot y \cdot x^2 \cdot f(x, y) dx dy$
fy	$\int y^2 \cdot f(y) dy$
fy1	$\int y^2 \cdot y \cdot f(y) dy$
fv1	$\int v^2 \cdot f(v) dv$
fv2	$\int v^2 \cdot v \cdot f(v) dv$

#### Covariance of Xs:

Editor - part2\_final.m

Variables - Cov\_Xs

Workspace

Name	Value
cov_U	[0.0764, -0.0104; -0.0104, 0.0760]
Cov_Xs	[0.0732, -0.0140; -0.0140, 0.0794]
cov_xy	-0.0104
diff	[0.0032, 0.0036; 0.0036, 0.0032]
Ex	0.5833
Ex2	0.4167
Exy	0.3542
Ey	0.6250
Ey2	0.4667
fx	$\int x^2 \cdot f(x) dx$
fx1	$\int x^2 \cdot f(x) dx / 2$
fx2	$\int x^2 \cdot f(x) dx + (3/2) \cdot \int x \cdot f(x) dx$
fxy	$\int x \cdot y \cdot f(x, y) dx dy$
fxy1	$\int x \cdot y \cdot x^2 \cdot f(x, y) dx dy$
fy	$\int y^2 \cdot f(y) dy$
fy1	$\int y^2 \cdot y \cdot f(y) dy$
fv1	$\int v^2 \cdot f(v) dv$
fv2	$\int v^2 \cdot v \cdot f(v) dv$

4)

By comparing the covariance matrix of U and Xs, we can see that both are almost similar with negligible variations. When we calculate the difference matrix for these two, we get very minute difference. The difference matrix is given by:

Editor - part2\_final.m

Variables - diff

Workspace

	1	2	3	4	5	6	7	8
1	0.0057	-0.0011						
2	-0.0011	7.2403e-04						
3								
4								
5								
6								
7								

Now taking the sample size as 2000: We have:

$$\text{Cov}_Xs =$$

$$\begin{bmatrix} 0.0752 & -0.0079 \\ -0.0079 & 0.0789 \end{bmatrix}$$

```
>> cov_U
```

```
cov_U =
```

```
0.0764 -0.0104  
-0.0104 0.0760
```

We can clearly see that as the sample size is increased, the covariance matrix of U and Xs are even more similar with very minute difference. Thus, by increasing the sample number, we can improve the covariance matrices based on requirements.

Program files for Part2:

Part2\_final.m