



ASSIGNMENT 9

(due at 8:30 AM Thursday, March 23 in class)

1. The random variables \mathbf{X} and \mathbf{Y} have a joint density function given by

$$f_{\mathbf{XY}}(x, y) = \begin{cases} \frac{2}{\pi} e^{-(x^2+y^2)}, & \text{if } xy > 0; \\ 0, & \text{otherwise;} \end{cases}$$

Find the joint density function of $\mathbf{Z} = \mathbf{XY}$ and $\mathbf{W} = \mathbf{X}/\mathbf{Y}$ and the marginal density function of \mathbf{W} . Are \mathbf{Z} and \mathbf{W} independent?

Hint: You don't need to explicitly find the density functions for both \mathbf{Z} and \mathbf{W} , but just consider $f_{\mathbf{Z}}(z | \mathbf{W} = w)$.

2. Suppose \mathbf{X} and \mathbf{Y} have the joint density function

$$f_{\mathbf{XY}}(x, y) = \begin{cases} (y+1)e^{-(x+y+xy)}, & x > 0 \text{ and } y > 0; \\ 0, & \text{otherwise,} \end{cases}$$

Find the joint density function of $\mathbf{Z} = \mathbf{X}^2\mathbf{Y}^2$ and $\mathbf{W} = \mathbf{Y}^2/\mathbf{X}^2$.

3. The random variables \mathbf{X} and \mathbf{Y} two independent $U(0, 1)$ random variables. Use the technique of an auxiliary random variable to find the density function of $\mathbf{Z} = \mathbf{XY}$.
4. Suppose that \mathbf{X} , \mathbf{Y} are two random variables with means 2 and -3 (respectively), variance 3 and 4 (respectively) and covariance -1.
- What is the correlation coefficient of these two random variables?
 - What is the correlation matrix of these two random variables?
 - What is the covariance matrix of these two random variables?
 - What is the mean, mean power and variance of the random variable $\mathbf{Z} = \mathbf{X} - 2\mathbf{Y}$.
5. Suppose $\theta \sim U(0, \pi)$, and $\mathbf{X} = \cos^2(\theta/2)$, $\mathbf{Y} = \sin^2(\theta/2)$. Determine if \mathbf{X} and \mathbf{Y} are uncorrelated.
6. For two real random variables \mathbf{X} and \mathbf{Y} that have finite second order moments, it is easy to see that for all real values α , $\mathcal{E}\{[\mathbf{X} + \alpha\mathbf{Y}]^2\} \geq 0$, and that

$$\mathcal{E}\{[\mathbf{X} + \alpha\mathbf{Y}]^2\} = \mathcal{E}\{\mathbf{X}^2 + \alpha^2\mathbf{Y}^2 + 2\alpha\mathbf{XY}\} = \alpha^2\mathcal{E}\{\mathbf{Y}^2\} + 2\alpha\mathcal{E}\{\mathbf{XY}\} + \mathcal{E}\{\mathbf{X}^2\}.$$

- Find the value of α that minimizes $\mathcal{E}\{[\mathbf{X} + \alpha\mathbf{Y}]^2\}$, expressing it in terms of the correlation between \mathbf{X} and \mathbf{Y} , $\text{cor}(\mathbf{X}, \mathbf{Y})$, $\mathcal{E}\{\mathbf{X}^2\}$ and $\mathcal{E}\{\mathbf{Y}^2\}$. Assume that $\mathcal{E}\{\mathbf{Y}^2\} > 0$.
- Using the fact that the minimum value of $\mathcal{E}\{[\mathbf{X} + \alpha\mathbf{Y}]^2\}$ must be non-negative, show that

$$[\mathcal{E}\{\mathbf{XY}\}]^2 \leq \mathcal{E}\{\mathbf{X}^2\}\mathcal{E}\{\mathbf{Y}^2\} \quad (\dagger)$$

when $\mathcal{E}\{\mathbf{Y}^2\} > 0$. Argue that we can drop the assumption that $\mathcal{E}\{\mathbf{Y}^2\} > 0$ so that (\dagger) holds for all \mathbf{X} and \mathbf{Y} which both have a finite mean square value. This is a very important result that is known as the *Cauchy-Schwarz inequality* (for random variables). Under what conditions will we have equality in this expression?

Hint: If $\mathcal{E}\{\mathbf{Y}^2\} = 0$ but $\mathcal{E}\{\mathbf{X}^2\} > 0$, interchanging \mathbf{X} and \mathbf{Y} gives us that (\dagger) holds in that case, leaving only to show (\dagger) holds when $\mathcal{E}\{\mathbf{X}^2\} = \mathcal{E}\{\mathbf{Y}^2\} = 0$. For that just consider $\mathcal{E}\{[\mathbf{X} + \mathbf{Y}]^2\} \geq 0$ and $\mathcal{E}\{[\mathbf{X} - \mathbf{Y}]^2\} \geq 0$ to show that if $\mathcal{E}\{\mathbf{X}^2\} = \mathcal{E}\{\mathbf{Y}^2\} = 0$, then $\mathcal{E}\{\mathbf{XY}\} = 0$ so (\dagger) holds then as well (with equality).

6. [continued]

(c) Based on (b), show that

$$[\text{cov}(\mathbf{X}, \mathbf{Y})]^2 \leq \text{var}(\mathbf{X})\text{var}(\mathbf{Y}),$$

and then that $[\rho_{\mathbf{X}\mathbf{Y}}]^2 \leq 1$.

7. Suppose that \mathbf{X} , \mathbf{Y} are two random variables with means 1 and -1 (respectively) and covariance matrix

$$\mathbf{C}_{\mathbf{X}\mathbf{Y}} = \begin{bmatrix} 2 & -1 \\ -1 & 9 \end{bmatrix}$$

Find the mean, correlation matrix and covariance matrix of $\mathbf{Z} = \mathbf{X} + 2\mathbf{Y} - 3$, $\mathbf{W} = 3\mathbf{X} - 7\mathbf{Y} + 12$.



ASSIGNMENT 9

SOLUTIONS

1/ First we solve the simultaneous equations $z = xy$, $w = x/y$. If $zw < 0$, there are no solutions, but if $zw > 0$, these simultaneous equations always have exactly two solutions and these vary with whether both z and w are positive or both are negative. If both $z > 0$ and $w > 0$, the solutions are:

$$x = \sqrt{zw}, y = \sqrt{z/w}, \quad \text{and} \quad x = -\sqrt{zw}, y = -\sqrt{z/w}$$

If both $z < 0$ and $w < 0$, the solutions are:

$$x = \sqrt{zw}, y = -\sqrt{z/w}, \quad \text{and} \quad x = -\sqrt{zw}, y = \sqrt{z/w}$$

[The case of $zw = 0$ can be ignored as \mathbf{X} and \mathbf{Y} are joint continuous.]

The Jacobian determinant of the transformation is

$$J(x, y) = \det \begin{bmatrix} y & x \\ 1/y & -x/y^2 \end{bmatrix} = -2x/y.$$

At the solution points the absolute value of this is $2|w|$. Hence,

$$f_{\mathbf{Z}\mathbf{W}}(z, w) = \begin{cases} \frac{f_{\mathbf{X}\mathbf{Y}}(\sqrt{zw}, \sqrt{z/w}) + f_{\mathbf{X}\mathbf{Y}}(-\sqrt{zw}, -\sqrt{z/w})}{2|w|}, & \text{if } z > 0 \text{ and } w > 0; \\ \frac{f_{\mathbf{X}\mathbf{Y}}(\sqrt{zw}, -\sqrt{z/w}) + f_{\mathbf{X}\mathbf{Y}}(-\sqrt{zw}, \sqrt{z/w})}{2|w|}, & \text{if } z < 0 \text{ and } w < 0; \\ 0, & \text{otherwise.} \end{cases}$$

$$= \begin{cases} \frac{2}{\pi|w|} e^{-(zw+[z/w])} = \frac{2}{\pi|w|} e^{-z(w+[1/w])}, & \text{if } z > 0 \text{ and } w > 0; \\ 0, & \text{otherwise.} \end{cases}$$

Then

$$f_{\mathbf{W}}(w) = \int_{-\infty}^{\infty} f_{\mathbf{Z}\mathbf{W}}(z, w) dz = \begin{cases} \frac{2}{\pi|w|} \int_0^{\infty} e^{-z(w+[1/w])} dz, = \frac{2}{\pi w(w+[1/w])}, & \text{if } w > 0; \\ 0, & \text{otherwise.} \end{cases}$$

$$= \frac{2}{\pi(1+w^2)} u(w).$$

For $w > 0$, we thus have that

$$f_{\mathbf{Z}}(z | \mathbf{W} = w) = \frac{f_{\mathbf{Z}\mathbf{W}}(z, w)}{f_{\mathbf{W}}(w)} = \frac{1+w^2}{|w|} e^{-z(w+[1/w])} u(z),$$

which varies with both z and w [for $w = 1$ this is $2e^{-2z}u(z)$ and for $w = 2$ this is $\frac{5}{2}e^{-5z/2}u(z)$] and so cannot be $f_{\mathbf{Z}}(z)$. Thus $f_{\mathbf{Z}}(z)f_{\mathbf{W}}(w) \neq f_{\mathbf{Z}\mathbf{W}}(z, w)$ for some z, w , giving us that \mathbf{Z} and \mathbf{W} are *not* independent.

- 2/ First we solve the simultaneous equations $z = x^2y^2$, $w = y^2/x^2$. If $zw < 0$, there are no solutions, but if $zw > 0$, these simultaneous equations always have exactly two solutions and these vary with whether both z and w are positive or both negative. If both $z > 0$ and $w > 0$, the solutions are:

$$x = \sqrt[4]{z/w}, y = \sqrt[4]{zw}; \quad x = \sqrt[4]{z/w}, y = -\sqrt[4]{zw}; \quad x = -\sqrt[4]{z/w}, y = \sqrt[4]{zw}; \quad \text{and} \quad x = -\sqrt[4]{z/w}, y = -\sqrt[4]{zw}.$$

Otherwise there are no solutions unless both $z = 0$ and $w = 0$ in which case there are an uncountable number of solutions and then x can have any value as long as $y = 0$.

The Jacobian determinant of the transformation is

$$J(x, y) = \det \begin{bmatrix} 2xy^2 & 2x^2y \\ -2y^2/x^3 & 2y/x^2 \end{bmatrix} = 8y^3/x.$$

At the solution points, the absolute value of this is $8\sqrt[4]{z^2w^4} = 8w\sqrt{z}$. Hence,

$$f_{\mathbf{Z}\mathbf{W}}(z, w) = \begin{cases} \frac{f_{\mathbf{X}\mathbf{Y}}(\sqrt[4]{z/w}, \sqrt[4]{zw}) + f_{\mathbf{X}\mathbf{Y}}(-\sqrt[4]{z/w}, \sqrt[4]{zw}) + f_{\mathbf{X}\mathbf{Y}}(\sqrt[4]{z/w}, -\sqrt[4]{zw}) + f_{\mathbf{X}\mathbf{Y}}(-\sqrt[4]{z/w}, -\sqrt[4]{zw})}{8w\sqrt{z}}, & z, w > 0; \\ 0, & \text{otherwise.} \end{cases}$$

But as this $f_{\mathbf{X}\mathbf{Y}}(x, y)$ is only nonzero if both $x > 0$ and $y > 0$, this simplifies to

$$\begin{aligned} f_{\mathbf{Z}\mathbf{W}}(z, w) &= \begin{cases} \frac{f_{\mathbf{X}\mathbf{Y}}(\sqrt[4]{z/w}, \sqrt[4]{zw})}{8w\sqrt{z}}, & \text{if } z > 0, w > 0; \\ 0, & \text{otherwise.} \end{cases} \\ &= \begin{cases} \frac{(\sqrt[4]{zw} + 1)e^{-(\sqrt[4]{z/w} + \sqrt[4]{zw} + \sqrt{z})}}{8w\sqrt{z}}, & \text{if } z > 0, w > 0; \\ 0, & \text{otherwise.} \end{cases} \end{aligned}$$

We have ignored the case that $z = w = 0$ as the event that $\{\mathbf{X} = 0 \text{ or } \mathbf{Y} = 0\}$ is a zero probability event.

- 3/ Let us select as an auxiliary random variable, $\mathbf{W} = \mathbf{X}$. Now let us find the joint density function of \mathbf{Z} , \mathbf{W} . First we solve the simultaneous equations $z = xy$, $w = x$, which have a unique solution $x = w$, $y = z/w$ (if $w \neq 0$). The Jacobian determinant of the transformation is

$$J(x, y) = \det \begin{bmatrix} y & x \\ 1 & 0 \end{bmatrix} = -x.$$

At the solution points the absolute value of this is $|w|$. Hence, since

$$f_{\mathbf{X}\mathbf{Y}}(x, y) = \begin{cases} 1, & \text{if } 0 < x < 1, 0 < y < 1; \\ 0, & \text{otherwise,} \end{cases}$$

we have

$$f_{\mathbf{Z}\mathbf{W}}(z, w) = f_{\mathbf{X}\mathbf{Y}}(w, z/w) = \begin{cases} \frac{1}{|w|}, & \text{if } 0 < z < w < 1; \\ 0, & \text{otherwise.} \end{cases}$$

Then

$$f_{\mathbf{Z}}(z) = \int_{-\infty}^{\infty} f_{\mathbf{Z}\mathbf{W}}(z, w) dw = \begin{cases} \int_z^1 \frac{1}{w} dw = -\ln z, & \text{if } 0 < z < 1; \\ 0, & \text{otherwise.} \end{cases}$$

4/ (a) $\rho_{\mathbf{XY}} \triangleq \frac{\text{cov}(\mathbf{X}, \mathbf{Y})}{\sqrt{\text{var}(\mathbf{X})\text{var}(\mathbf{Y})}} = \frac{-1}{\sqrt{12}} \simeq -0.288675.$

(b) In general for real random variables \mathbf{A} and \mathbf{B} , $\text{cor}(\mathbf{A}, \mathbf{B}) = \text{cov}(\mathbf{A}, \mathbf{B}) + \mathcal{E}\{\mathbf{A}\}\mathcal{E}\{\mathbf{B}\}$, so

$$\text{cor}(\mathbf{X}, \mathbf{X}) = \text{var}(\mathbf{X}) + [\mathcal{E}\{\mathbf{X}\}]^2 = 3 + 2^2 = 7, \quad \text{cor}(\mathbf{Y}, \mathbf{Y}) = \text{var}(\mathbf{Y}) + [\mathcal{E}\{\mathbf{Y}\}]^2 = 4 + (-3)^2 = 13,$$

$$\text{cor}(\mathbf{X}, \mathbf{Y}) = \text{cor}(\mathbf{Y}, \mathbf{X}) = \text{cov}(\mathbf{X}, \mathbf{Y}) + \mathcal{E}\{\mathbf{X}\}\mathcal{E}\{\mathbf{Y}\} = -1 + 2 \times (-3) = -7.$$

Thus

$$\mathbf{R}_{\mathbf{XY}} = \begin{bmatrix} \text{cor}(\mathbf{X}, \mathbf{X}) & \text{cor}(\mathbf{X}, \mathbf{Y}) \\ \text{cor}(\mathbf{Y}, \mathbf{X}) & \text{cor}(\mathbf{Y}, \mathbf{Y}) \end{bmatrix} = \begin{bmatrix} 7 & -7 \\ -7 & 13 \end{bmatrix}.$$

(c) $\mathbf{C}_{\mathbf{XY}} = \begin{bmatrix} \text{cov}(\mathbf{X}, \mathbf{X}) & \text{cov}(\mathbf{X}, \mathbf{Y}) \\ \text{cov}(\mathbf{Y}, \mathbf{X}) & \text{cov}(\mathbf{Y}, \mathbf{Y}) \end{bmatrix} = \begin{bmatrix} 3 & -1 \\ -1 & 4 \end{bmatrix}.$

(d) $\mathcal{E}\{\mathbf{Z}\} = \mathcal{E}\{\mathbf{X}\} - 2\mathcal{E}\{\mathbf{Y}\} + 1 = 2 + 6 + 1 = 9.$

$$\text{var}(\mathbf{Z}) = \text{var}(\mathbf{X}) + (-2)^2\text{var}(\mathbf{Y}) + 2 \times (-2) \times 1\text{cov}(\mathbf{X}, \mathbf{Y}) = 3 + 16 + 4 = 23.$$

$$\text{mean power} = \mathcal{E}\{\mathbf{Z}^2\} = \text{var}(\mathbf{Z}) + (\mathcal{E}\{\mathbf{Z}\})^2 = 23 + 81 = 104.$$

5/ $\cos^2(x/2) = \frac{1}{2} + \frac{1}{2} \cos(x), \quad \sin^2(x/2) = \frac{1}{2} - \frac{1}{2} \cos(x), \quad \cos^2(x/2) \sin^2(x/2) = \frac{1}{4}(1 - \cos^2(x)) = \frac{1}{8} - \frac{1}{8} \cos(2x).$

$$\mathcal{E}\{\mathbf{X}\} = \frac{1}{2} + \int_0^\pi \frac{\cos(\phi)}{2\pi} d\phi = \frac{1}{2}, \quad \mathcal{E}\{\mathbf{Y}\} = \frac{1}{2} - \int_0^\pi \frac{\cos(\phi)}{2\pi} d\phi = \frac{1}{2}.$$

$$\text{cor}(\mathbf{X}, \mathbf{Y}) = \int_0^\pi \frac{\sin^2(\phi/2) \cos^2(\phi/2)}{\pi} d\phi = \frac{1}{8} - \int_0^\pi \frac{\cos(2\phi)}{18\pi} d\phi = \frac{1}{8}.$$

$$\text{cov}(\mathbf{X}, \mathbf{Y}) = \mathcal{E}\{\mathbf{XY}\} - \mathcal{E}\{\mathbf{X}\}\mathcal{E}\{\mathbf{Y}\} = \frac{1}{8} - \frac{1}{4} = -\frac{1}{8} \neq 0.$$

Since $\text{cov}(\mathbf{X}, \mathbf{Y}) \neq 0$, the two random variables are not uncorrelated.

6/ (a) The value of α that minimizes $\mathcal{E}\{[\mathbf{X} + \alpha\mathbf{Y}]^2\}$ must satisfy the condition for a local maximum which is that $\frac{\partial}{\partial \alpha} \mathcal{E}\{[\mathbf{X} + \alpha\mathbf{Y}]^2\} = 0$. As $\mathcal{E}\{[\mathbf{X} + \alpha\mathbf{Y}]^2\} = \mathcal{E}\{\mathbf{X}^2\} + \alpha^2\mathcal{E}\{\mathbf{Y}^2\} + 2\alpha\mathcal{E}\{\mathbf{XY}\}$, we have

$$\frac{\partial}{\partial \alpha} \mathcal{E}\{[\mathbf{X} + \alpha\mathbf{Y}]^2\} = 2\alpha\mathcal{E}\{\mathbf{Y}^2\} + 2\mathcal{E}\{\mathbf{XY}\} = 0. \tag{*}$$

By assumption $\mathcal{E}\{\mathbf{Y}^2\} > 0$, so (*) has a unique solution:

$$\alpha = \frac{-\mathcal{E}\{\mathbf{XY}\}}{\mathcal{E}\{\mathbf{Y}^2\}} = \frac{-\text{cor}(\mathbf{X}, \mathbf{Y})}{\mathcal{E}\{\mathbf{Y}^2\}}. \tag{**}$$

(b) At the minimum, the value of $\mathcal{E}\{[\mathbf{X} + \alpha\mathbf{Y}]^2\}$ is

$$\mathcal{E}\{\mathbf{X}^2\} + \frac{\mathcal{E}^2\{\mathbf{XY}\}}{\mathcal{E}\{\mathbf{Y}^2\}} - \frac{2\mathcal{E}^2\{\mathbf{XY}\}}{\mathcal{E}\{\mathbf{Y}^2\}} = \mathcal{E}\{\mathbf{X}^2\} - \frac{\mathcal{E}^2\{\mathbf{XY}\}}{\mathcal{E}\{\mathbf{Y}^2\}}.$$

Since $\mathcal{E}\{[\mathbf{X} + \alpha\mathbf{Y}]^2\} \geq 0$ for any α , we must conclude $\mathcal{E}\{\mathbf{X}^2\} \geq \frac{\mathcal{E}^2\{\mathbf{XY}\}}{\mathcal{E}\{\mathbf{Y}^2\}}$. Since $\mathcal{E}\{\mathbf{Y}^2\} > 0$, this gives us that

$$\mathcal{E}^2\{\mathbf{XY}\} \leq \mathcal{E}\{\mathbf{X}^2\}\mathcal{E}\{\mathbf{Y}^2\} \quad \text{if } \mathcal{E}\{\mathbf{Y}^2\} > 0.$$

Now if we interchange \mathbf{X} and \mathbf{Y} , we would get

$$\mathcal{E}^2\{\mathbf{XY}\} \leq \mathcal{E}\{\mathbf{X}^2\}\mathcal{E}\{\mathbf{Y}^2\} \quad \text{if } \mathcal{E}\{\mathbf{X}^2\} > 0,$$

6/ (b) continued)

establishing that (†) holds when at least one of $\mathcal{E}\{\mathbf{X}^2\} > 0$ or $\mathcal{E}\{\mathbf{Y}^2\} > 0$ holds. We can also show it holds in the final case where $\mathcal{E}\{\mathbf{X}^2\} = 0$ and $\mathcal{E}\{\mathbf{Y}^2\} = 0$. If $\mathcal{E}\{\mathbf{X}^2\} = \mathcal{E}\{\mathbf{Y}^2\} = 0$, then

$$\mathcal{E}\{[\mathbf{X} + \mathbf{Y}]^2\} = \mathcal{E}\{\mathbf{X}^2\} + \mathcal{E}\{\mathbf{Y}^2\} + 2\mathcal{E}\{\mathbf{XY}\} = 2\mathcal{E}\{\mathbf{XY}\} \geq 0 \quad \implies \mathcal{E}\{\mathbf{XY}\} \geq 0; \quad (A)$$

$$\mathcal{E}\{[\mathbf{X} - \mathbf{Y}]^2\} = \mathcal{E}\{\mathbf{X}^2\} + \mathcal{E}\{\mathbf{Y}^2\} - 2\mathcal{E}\{\mathbf{XY}\} = -2\mathcal{E}\{\mathbf{XY}\} \geq 0 \quad \implies \mathcal{E}\{\mathbf{XY}\} \leq 0. \quad (B)$$

For (A) and (B) to hold, we must have $\mathcal{E}\{\mathbf{XY}\} = 0$, allowing us to state

$$\mathcal{E}^2\{\mathbf{XY}\} \leq \mathcal{E}\{\mathbf{X}^2\}\mathcal{E}\{\mathbf{Y}^2\} \quad \text{if } \mathcal{E}\{\mathbf{X}^2\} = \mathcal{E}\{\mathbf{Y}^2\} = 0$$

So we know that (†) holds in all cases where \mathbf{X} and \mathbf{Y} that have finite second order moments.

To have equality hold in the inequality when one of $\mathcal{E}\{\mathbf{X}^2\} > 0$ or $\mathcal{E}\{\mathbf{Y}^2\} > 0$ holds, we see we must have for some α either $\mathcal{E}\{[\mathbf{X} + \alpha\mathbf{Y}]^2\} = 0 \implies \mathbf{X} = -\alpha\mathbf{Y}$ or $\mathcal{E}\{[\mathbf{Y} + \alpha\mathbf{X}]^2\} = 0 \implies \mathbf{Y} = -\alpha\mathbf{X}$ that is to say, one random variable is a scaled version of the other. Thinking about a random variable as a vector, this can be stated as the random variables are colinear.

(c) If we replace \mathbf{X} by $\mathbf{X} - \mathcal{E}\{\mathbf{X}\}$ and \mathbf{Y} by $\mathbf{Y} - \mathcal{E}\{\mathbf{Y}\}$ in the above result gives us the required result.

7/ $\mathcal{E}\{\mathbf{Z}\} = \mathcal{E}\{\mathbf{X}\} + 2\mathcal{E}\{\mathbf{Y}\} - 3 = 1 - 2 - 3 = -4$, $\mathcal{E}\{\mathbf{W}\} = 3\mathcal{E}\{\mathbf{X}\} - 7\mathcal{E}\{\mathbf{Y}\} + 12 = 3 + 7 + 12 = 22$.

$$\mathbf{C}_{\mathbf{ZW}} = \begin{bmatrix} 1 & 2 \\ 3 & -7 \end{bmatrix} \mathbf{C}_{\mathbf{XY}} \begin{bmatrix} 1 & 3 \\ 2 & -7 \end{bmatrix} = \begin{bmatrix} 34 & -119 \\ -119 & 501 \end{bmatrix}$$

$$\mathbf{R}_{\mathbf{ZW}} = \mathbf{C}_{\mathbf{ZW}} + \begin{bmatrix} -4 \\ 22 \end{bmatrix} \begin{bmatrix} -4 & 22 \end{bmatrix} = \begin{bmatrix} 50 & -207 \\ -207 & 985 \end{bmatrix}$$