Solutions to EE 3025 Recitation 10 Review Problems for Exam 2

- **1.** Let RV X have variance 100 and let RV Y have variance 400.
 - (a) If the correlation coefficient $\rho_{X,Y}$ is 1/2, compute Var(X + Y) and Var(X Y). Solution. We have

$$Cov(X,Y) = \rho\sigma_X\sigma_Y = (1/2)(10)(20) = 100.$$

Therefore,

$$Var(X+Y) = Var(X) + Var(Y) + 2Cov(X,Y) = 700$$
$$Var(X+Y) = Var(X) + Var(Y) - 2Cov(X,Y) = 300$$

(b) On the other hand, if Var(X + Y) = 400 and Var(X - Y) = 600, figure out what the correlation coefficient ρ_{X,Y} is. Then, compute Var(3X - 2Y).
 Solution. We have

Var(X+Y) = Var(X) + Var(Y) + 2Cov(X,Y) = 500 + 2Cov(X,Y) = 400and therefore

$$Cov(X,Y) = -50.$$

(The Var(X - Y) condition is not needed.) Then

$$\rho = \frac{Cov(X,Y)}{\sigma_X \sigma_Y} = \frac{-50}{(10)(20)} = -1/4.$$

Var(3X-2Y) = 9Var(X) + 4Var(Y) - 12Cov(X,Y) = 9*100 + 4*400 - 12*(-50) = 3100.

- **2.** Let RV X have variance 100 and let RV Y have variance 400. Let Cov(X, Y) be 100.
 - (a) Find the constant C that makes

$$Cov(X, X - CY) = 0.$$

In other words, you are making X and X - CY uncorrelated. Solution.

$$Cov(X, X - CY) = Cov(X, X) - C * Cov(X, Y) = 100 - C * 100 = 0.$$

Take C = 1.

(b) Find the constant D that makes

$$Var(X - DY)$$

a minimum.

Solution.

$$Var(X - DY) = Var(X) + D^{2}Var(Y) - 2D * Cov(X, Y) = 100 * 400D^{2} - 200D.$$

Set the derivative equal to zero:

$$800D - 200 = 0.$$

You see that D = 1/4.

3. Discrete RV's X, Y are independent. The PDF of X is

$$(0.2)\delta(x-1) + (0.4)\delta(x-3) + (0.4)\delta(x-4).$$

The PDF is Y is

 $(0.4)\delta(y+1) + (0.6)\delta(y-2).$

(a) Use convolution to find the PDF of U = X + Y.

Solution. Think of these as two functions of time that you are convoluting:

$$(0.2)\delta(t-1) + (0.4)\delta(t-3) + (0.4)\delta(t-4)$$

and

$$(0.4)\delta(t+1) + (0.6)\delta(t-2)$$

First, convolute the 1st signal with $(0.4)\delta(t+1)$:

$$0.4[(0.2)\delta(t) + (0.4)\delta(t-2) + (0.4)\delta(t-3)]$$

Then, convolute the 1st signal with $(0.6)\delta(t-2)$:

$$0.6[(0.2)\delta(t-3) + (0.4)\delta(t-5) + (0.4)\delta(t-6)]$$

Then add the two results, writing as a function of U:

$$0.08\delta(u) + 0.16\delta(u-2) + 0.28\delta(u-3) + 0.24\delta(u-5) + 0.24\delta(u-6)$$

(b) Use convolution to find the PDF of V = 2X - Y.

Solution. First, write down the PDF of 2X as a function of t:

$$(0.2)\delta(t-2) + (0.4)\delta(t-6) + (0.4)\delta(t-8).$$

Then, write down the PDF of -Y as a function of t:

$$(0.4)\delta(t-1) + (0.6)\delta(t+2).$$

Convolute the 1st signal with $(0.4)\delta(t-1)$:

$$0.4[(0.2)\delta(t-3) + (0.4)\delta(t-7) + (0.4)\delta(t-9)]$$

Convolute the 1st signal with $(0.6)\delta(t+2)$:

$$0.6[(0.2)\delta(t) + (0.4)\delta(t-4) + (0.4)\delta(t-6)]$$

Add the two results, writing as a function of v:

$$0.12\delta(v) + 0.08\delta(v-3) + 0.24\delta(v-4) + 0.24\delta(v-6) + 0.16\delta(v-7) + 0.16\delta(v-9).$$

4. Let I, R be nonnegative continuously distributed RV's. Use the CDF method to show that V = IR has PDF

$$f_V(v) = \int_0^\infty \frac{1}{i} f_I(i) f_R(v/i) di, \quad v \ge 0 \text{ (zero elsewhere)}.$$

Solution. For $v \ge 0$,

$$F_V(v) = P[V \le v] = P[IR \le v] = \int_0^\infty P[IR \le v|I=i]f_I(i)di.$$

Now

$$P[IR \le v|I = i] = P[iR \le v|I = i] = P[R \le v/i] = F_R(v/i).$$

Therefore,

$$F_V(v) = \int_0^\infty F_R(v/i) f_I(i) di.$$

Differentiating both sides,

$$f_V(v) = dF_V(v)/dv = \int_0^\infty \partial F_R(v/i) f_I(i)/\partial v di.$$

Plugging in

$$\partial F_R(v/i)f_I(i)/\partial v = (1/i)f_R(v/i)f_I(i),$$

we obtain the desired result.

5. X, Y have the bivariate Gaussian density

$$f_{X,Y}(x,y) = C \exp(-0.5[x^2 - 2xy + 4y^2]).$$

(a) What are $\mu_X, \mu_Y, \sigma_X, \sigma_Y, \rho_{X,Y}$?

Solution. Since there are no x, y terms in the exponent,

$$\mu_X = \mu_Y = 0.$$

Note that

$$x^{2} - 2xy + 4y^{2} = (x \ y) \begin{pmatrix} 1 & -1 \\ -1 & 4 \end{pmatrix} (x \ y)^{T}$$

The inverse of the "matrix in the middle" is the covariance matrix, which is

$$\left(\begin{array}{cc} 4/3 & 1/3 \\ 1/3 & 1/3 \end{array}\right) = \left(\begin{array}{cc} \sigma_X^2 & \sigma_{X,Y} \\ \sigma_{X,Y} & \sigma_Y^2 \end{array}\right)$$

We have

$$\sigma_X^2 = 4/3, \ \sigma_Y^2 = 1/3, \ \sigma_{X,Y} = 1/3,$$

 $\rho = \frac{\sigma_{X,Y}}{\sigma_X \sigma_Y} = \frac{1/3}{\sqrt{4/3}\sqrt{1/3}} = 1/2.$

(b) What are E[X|Y = 1] and E[Y|X = 1]? Solution. For every y,

$$E[X|Y = y] = \mu_X + \rho(\sigma_X/\sigma_Y)(y - \mu_Y) = y.$$

Therefore,

$$E[X|Y=1] = 1.$$

For every x,

$$E[Y|X = x] = \mu_Y + \rho(\sigma_Y/\sigma_X)(x - \mu_X) = x/4.$$

Therefore

$$E[Y|X = 1] = 1/4.$$

- 6. Let X be Uniform(0,1).
 - (a) Find the PDF of Y = X³.
 Solution. Y ranges from 0 to 1. For each fixed y in this range,

$$F_Y(y) = P(Y \le y) = P(X^3 \le y) = P(X \le y^{1/3}) = y^{1/3}.$$

Differentiating,

$$f_Y(y) = (1/3)y^{-2/3}, \ 0 \le y \le 1 \ (zero \ elsewhere).$$

(b) Find the PDF of Y = X(1 - X).

Solution. First, plot the curve y = x(1-x) from x = 0 to x = 1. You'll see that Y ranges from 0 to 1/4. For each y between 0 and 1/4, there are two x values such that x(1-x) = y, namely, the values

$$x_1(y) = \frac{1 - \sqrt{1 - 4y}}{2}, \quad x_2(y) = \frac{1 + \sqrt{1 - 4y}}{2}.$$

From the curve y = x(1 - x), you will see that the event $\{Y \le y\}$ can be broken down as follows:

$$\{Y \le y\} = \{X \le x_1(y)\} \cup \{X \ge x_2(y)\}.$$

Therefore,

$$P(Y \le y) = P(X \le x_1(y)) + P(X \ge x_2(y)).$$

By symmetry, the two prob on the right side are equal and so

$$P(Y \le y) = 2P(X \le x_1(y)) = 2x_1(y) = 1 - \sqrt{1 - 4y}.$$

Differentiating,

$$f_Y(y) = \frac{2}{\sqrt{1-4y}}, \ 0 \le y \le 1/4 \ (zero \ elsewhere).$$

- 7. Let X_1, X_2 be independent discrete RV's each taking the values 1, 2, 3 with probability 1/3 each.
 - (a) Find the PMF of $U = \max(X_1, X_2)$.
 - (b) Find the PMF of $V = \min(X_1, X_2)$.

Solution. There's a clever way to work (a),(b) in a unified way. Map (X_1, X_2) values into (U, V) values:

X1	Х2	U	V
1	1	1	1
1	2	2	1
1	3	3	1
2	1	2	1
2	2	2	2
2	3	3	2
3	1	3	1
3	2	3	2
3	3	3	3

Each (X_1, X_2) value has prob 1/9. This allows you to construct the following U, V joint prob table:

$$V = 1 \quad V = 2 \quad V = 3$$
$$U = 1 \quad \begin{pmatrix} 1/9 & 0 & 0 \\ 2/9 & 1/9 & 0 \\ 0 & 2/9 & 2/9 & 1/9 \end{pmatrix}$$
$$U = 3 \quad \begin{pmatrix} 2/9 & 1/9 & 0 \\ 2/9 & 2/9 & 1/9 \end{pmatrix}$$

The row sums and column sums are the PMF's of U, V, respectively.

8. Let X, Y be discrete RV's whose joint PMF takes the form

$$P[X = i, Y = j] = C|i - 2j|,$$

for i, j = 1, 2, 3 and 0 elsewhere.

(a) Find the constant C.

Solution. C must be the reciprocal of the sum

$$\sum_{i,j=1}^{3} |i - 2j|.$$

(b) Find the marginal PMF $p_X(x)$. Solution.

$$p_X(i) = C \sum_{j=1}^3 |i - 2j|, \quad i = 1, 2, 3.$$

(c) Find the conditional PMF of Y given X = 2. Solution.

$$p_{Y|X}(y|x=2) = \frac{C|2-2y|}{p_X(2)}, \quad y = 1, 2, 3.$$

(d) Compute E[Y|X = 2]. Solution.

$$E[Y|X=2] = \sum_{y=1}^{3} y[C|2-2y|/p_X(2)].$$

(e) Compute the correlation E[XY]. Solution.

$$E[XY] = \sum_{i,j=1}^{3} Cij|i-2j|.$$

9. Let X, Y be jointly continuous RV's whose joint PDF takes the form

$$f(x,y) = C(x+2y),$$

for $0 \le x \le 2$ and $0 \le y \le 1$.

(a) Find the constant C.Solution. C must be the reciprocal of the double integral

$$\int_0^1 \int_0^2 (x+2y) dx dy.$$

(b) Find the marginal PDF $f_X(x)$. Solution.

$$f_X(x) = \int_0^1 C(x+2y)dy,$$

for $0 \le x \le 2$.

(c) Find the conditional PDF of Y given X = 1. Solution.

$$f_{Y|X}(y|x=1) = \frac{C(1+2y)}{f_X(1)}, \quad 0 \le y \le 1.$$

(d) Compute P[Y < 1/2|X = 1]. Solution.

$$P(Y < 1/2 | X = 1) = \int_0^{1/2} \frac{C(1+2y)}{f_X(1)} dy.$$

(e) Compute the correlation E[XY].Solution.

$$E[XY] = \int_0^1 \int_0^2 Cxy(x+2y)dxdy.$$

- 10. Let (X, Y) be uniformly distributed over the triangular region inside the triangle in the xy-plane whose three vertices are (0,0), (5,0), (3,3)
 - (a) Find μ_X, μ_Y . Solution.

$$(\mu_X, \mu_Y) = (1/3)[(0,0) + (5,0) + (3,3)].$$

- (b) Find E[X|Y = y] as a function of y. This is a straight line function of y. What is this straight line in terms of the triangular region? Indicate on a plot.
 Solution. It's the median line that connects vertex (3,3) to the midpoint of the opposite side.
- **11.** Let X, Y be jointly continuously distributed with joint density

$$f_{X,Y}(x,y) = \begin{cases} 2, & 0 \le y \le x \le 1\\ 0, & elsewhere \end{cases}$$

(a) Set up the double integral for E[XY] with all 4 limits properly inserted. Do not integrate.

Solution.

$$E[XY] = \int_0^1 \int_0^x xy f(x, y) dy dx.$$

(b) Set up E[X|Y = 0.5] as the ratio of two one-dimensional integrals with all the limits properly inserted. Do not integrate.
 Solution.

$$E[X|Y = 0.5] = \frac{\int_{1/2}^{1} xf(x,y)dx}{\int_{1/2}^{1} f(x,y)dx},$$

in which you substitute y = 0.5.

(c) Set up P[X² + Y² ≤ 0.25] as a double integral in polar coordinates with all the limits properly inserted. Do not integrate.
 Solution.

$$\int_0^{\pi/4} \int_0^{1/2} 2r dr d\theta.$$

- 12. Bill eats X ice cream cones, where X is Poisson with parameter 2. Bill then runs Y miles, where Y is the number of heads obtains in flipping a fair coin X + 2 times. Use the law of iterated expectation to compute:
 - (a) E[Y]. Solution.

$$E[Y|X] = (X+2)/2.$$

$$E[Y] = E[E[Y|X]] = (1/2)(E[X]+2) = 2.$$

(b) Var[Y].

Solution.

$$Var[Y|X] = (X+2)/4.$$

$$E[Y^2|X] = E[Y|X]^2 + Var[Y|X] = (X+2)^2/4 + (X+2)/4.$$

$$E[Y^2] = E[(X+2)^2]/4 + E[X+2]/4.$$

Expand this using E[X] = 2, $E[X^2] = 6$. Then use the formula

$$Var[Y] = E[Y^2] - \mu_Y^2.$$

(c) E[XY]. Solution.

$$E[XY|X] = XE[Y|X] = X(X+2)/2.$$

 $E[XY] = E[X(X+2)]/2.$

- (d) Cov[X,Y]. Solution. It's $E[XY] - \mu_X \mu_Y$.
- (e) $\rho_{X,Y}$. Solution. It's Cov[X,Y] divided by $\sigma_X \sigma_Y$. Use $\sigma_X = \sqrt{2}$.

13. Discrete random variables N and K have the joint PMF

$$P_{N,K}(n,k) = \begin{cases} \frac{100^n e^{-100}}{(n+1)!}, & k = 0, 1, \cdots, n; \ n = 0, 1, 2, \cdots \\ 0, & elsewhere \end{cases}$$

(a) Find $P_N(n)$.

Solution.

$$P_N(n) = \sum_{k=0}^n P_{N,K}(n,k) = (100)^n \exp(-100)/n!, \ n = 0, 1, 2, 3, \cdots$$

This is the Poisson(100) distribution.

(b) Find $P_{K|N}(k|n)$. Solution.

$$P_{K|N}(k|n) = \frac{P_{N,K}(n,k)}{P_N(n)} = 1/(n+1), \ k = 0, 1, \cdots, n.$$

In other words, this cond dist is DiscreteUniform(0, n).

- (c) Compute E[K|N = n]. Solution. E[K|N = n] = (0 + n)/2, the mean of DiscreteUniform(0, n). (See Appendix A.)
- (d) Compute E[K] via law of iterated expectation formula

$$E[K] = E[E[K|N]].$$

Solution. E[K|N] = N/2. Therefore,

$$E[K] = E[E[K|N]] = E[N/2] = 50,$$

since N is Poisson with mean 100.

14. X_1, X_2, \dots, X_{10} are independent RV's which are each uniformly distributed bewteeen -2 and 2. Let

$$S = X_1 + X_2 + \dots + X_{10}.$$

(a) Determine the correlation between S and the RV

$$U = X_1 - X_2 + X_3 - X_4 + X_5 - X_6 + X_7 - X_8 + X_9 - X_{10}$$

(b) Determine the correlation between S and the RV

$$V = X_1 - X_2 + X_3 - X_4 + X_5$$

(c) Determine the correlation between S and the RV

$$W = X_6 + X_7 + X_8 + X_9 + X_{10}$$

Solution. For $i \neq j$,

$$E[X_i X_j] = E[X_i]E[X_j] = 0 * 0 = 0.$$

Therefore, if you have to compute the expected value of the product of two linear comb's of the X_i 's, you only have to pay attention to product terms of the form $X_i * X_i = X_i^2$. Solution to (a). The only X_i^2 terms appearing in the product of S times U are:

$$X_1^2 - X_2^2 + X_3^2 - X_4^2 + X_5^2 - X_6^2 + X_7^2 - X_8^2 + X_9^2 - X_{10}^2$$

When you take the expected value term by term, the expected values can cancel out:

$$EX_1^2$$
] - $E[X_2^2]$ + ... + $E[X_9^2]$ - $E[X_{10}^2]$ = 0.

Solution to (b). Just paying attention to the expected values of the X_i^2 terms in SV:

$$= E[X_1^2] - E[X_2^2] + E[X_3^2] - E[X_4^2] + E[X_5^2] = E[X_5^2] = Var(X_5) = \frac{4^2}{12} = \frac{4}{3}$$

Solution to (c). By inspection, you'll get 5 times $E[X_1]^2$, which is 20/3.

15. X and Y are jointly Gaussian RV's with E[X] = E[Y] = 0 and Var[X] = Var[Y] = 1. Furthermore,

$$E[Y|X] = X/2.$$

What is the joint PDF of X and Y?

Solution. We have

$$\mu_{Y|X=x} = x/2.$$

The coefficient of x, which is 1/2, must be equal to

$$1/2 = \rho \sigma_Y / \sigma_X = \rho$$

Therefore,

$$Var(Y|X) = \sigma_Y^2(1 - \rho^2) = 3/4.$$

The conditional density of Y given X = x is therefore

$$\frac{1}{\sqrt{2\pi}\sqrt{3/4}}\exp(-0.5\frac{(y-\mu_{Y|X=x})^2}{3/4}).$$

Multiplying this by the Gaussian (μ_X, σ_X^2) density, we get the joint density

$$\frac{1}{\sqrt{2\pi}}\exp(-0.5x^2)\frac{1}{\sqrt{2\pi}\sqrt{3/4}}\exp(-2(y-x/2)^2/3),$$

which simplifies to

$$\frac{1}{2\pi\sqrt{3/4}}\exp(-2\{x^2+y^2-xy\}/3).$$

16. Three RV's X, Y, Z have a "trivariate" Gaussian distribution in which their joint density f(x, y, z) takes the form

$$f(x, y, z) = C \exp\left(-0.5[3x^2 + 6y^2 + 4z^2 + 4xy - 2xz + 2yz]\right),$$

where C is a (unique) positive constant that one does not need to know in order to work this problem.

(a) Compute the three variances σ_X^2 , σ_Y^2 , σ_Z^2 . (Hint: Invert a certain 3×3 matrix.) Solution. You get the covariance matrix by inverting the matrix

3	2	-1		
2	6	1		
-1	1	4		
whi	ch g	gives		
23	/43		-9/43	8/43
-9	/43		11/43	-5/43
8	/43		-5/43	14/43

Therefore,

$$\sigma_X^2 = 23/43, \ \ \sigma_Y^2 = 11/43, \ \ \sigma_Z^2 = 14/43.$$

(b) Compute the three correlation coefficients $\rho_{X,Y}$, $\rho_{X,Z}$, $\rho_{Y,Z}$. Solution.

$$\rho_{X,Y} = \frac{Cov(X,Y)}{\sigma_X \sigma_Y} = \frac{-9/43}{\sigma_X \sigma_Y} = -0.5658$$

$$\rho_{X,Z} = \frac{Cov(X,Z)}{\sigma_X \sigma_Z} = \frac{8/43}{\sigma_X \sigma_Z} = -0.4458$$

$$\rho_{Y,Z} = \frac{Cov(Y,Z)}{\sigma_Y \sigma_Z} = \frac{-5/43}{\sigma_Y \sigma_Z} = -0.4029$$

(c) Determine the joint density of Y and Z. (Hint: This is a *bivariate* Gaussian density.)

Solution. The means of X, Y, Z are each zero (no linear terms in the exponent of the trivariate density). The means of Y and Z are therefore zero, and the covariance matrix of Y and Z is

11/43 -5/43 -5/43 14/43 The inverse of this covariance matrx is 14/3 5/3 5/3 11/3

The Y, Z joint density is therefore of the form

$$C\exp(-0.5\{14y^2/3 + 10yz/3 + 11z^2/3\}),$$

where

$$C = \frac{1}{2\pi\sqrt{1-\rho_{Y,Z}^2}} = 0.1739.$$

17. Let random variables X, Y be jointly continuously distributed in the square

 $\{(x,y): 0 \le x \le 3; 0 \le y \le 3\}.$

Suppose their joint CDF satisfies

$$F_{X,Y}(1,1) = 1/81$$

$$F_{X,Y}(2,1) = 4/81$$

$$F_{X,Y}(3,1) = 1/9$$

$$F_{X,Y}(1,2) = 4/81$$

$$F_{X,Y}(2,2) = 16/81$$

$$F_{X,Y}(3,2) = 4/9$$

$$F_{X,Y}(1,3) = 1/9$$

$$F_{X,Y}(2,3) = 4/9$$

Compute

$$P[(X,Y) \in R],$$

where R is the cross-shaped region in the first quadrant of the xy-plane given by

 $R = \{(x, y) : 0 \le x \le 3; \ 1 \le y \le 2\} \cup \{(x, y) : 1 \le x \le 2; \ 0 \le y \le 3\}.$

(Hint: Partition R into 3 rectangular pieces and find the prob (X, Y) belongs to each piece.)

Solution. Abbreviate $F_{X,Y}(x,y)$ as F(x,y). For the rectangle (square) with four corners

(1, 1), (1, 2), (0, 1), (0, 2),

the prob (X, Y) falls inside is

$$F(1,2) + F(0,1) - F(1,1) - F(0,2) = 3/81.$$

For the rectangle with four corners

the prob (X, Y) falls inside is

$$F(2,3) + F(1,0) - F(2,0) - F(1,3) = 1/3.$$

For the rectangle (square) with four corners

the prob (X, Y) falls inside is

F(3,2) + F(2,1) - F(3,1) - F(2,2) = 15/81.

Therefore, the prob (X, Y) falls in the union of these 3 rectangles is

(3/81) + (1/3) + (15/81) = 5/9.

18. Problem 5.6.7, page 241, of your textbook.

Solution. The pairs (Y_1, Y_2) and (Y_3, Y_4) have the same probability distribution; they are also independent pairs. Therefore the mean vector of (Y_1, Y_2, Y_3, Y_4) will have the form

abab

where a, b are the means of Y_1, Y_2 , respectively, and the covariance matrix of (Y_1, Y_2, Y_3, Y_4) will have the form

c d 0 0 e f 0 0 0 0 c d 0 0 e f where c d e f

is the covariance matrix of (Y_1, Y_2) .

By factorization, (Y_1, Y_2) has joint PDF equal to 2 over the region

$$\{0 \le y_1 \le y_2 \le 1\},\$$

and so

$$\begin{split} E[Y_1] &= \int_0^1 \int_{y_1}^1 y_1(2) dy_2 dy_1 = 1/3 \\ E[Y_2] &= \int_0^1 \int_{y_1}^1 y_2(2) dy_2 dy_1 = 2/3 \\ E[Y_1Y_2] &= \int_0^1 \int_{y_1}^1 y_1 y_2(2) dy_2 dy_1 = 1/4 \\ Cov(Y_1, Y_2) &= 1/4 - (1/3)(2/3) = 1/36 \\ E[Y_1^2] &= \int_0^1 \int_{y_1}^1 y_1^2(2) dy_2 dy_1 = 1/6 \\ E[Y_2^2] &= \int_0^1 \int_{y_1}^1 y_2^2(2) dy_2 dy_1 = 1/2 \\ Var(Y_1) &= 1/6 - (1/3)^2 = 1/18 \\ Var(Y_2) &= 1/2 - (2/3)^2 = 1/18 \end{split}$$

The mean vector of (Y_1, Y_2, Y_3, Y_4) is therefore

$$(1/3, 2/3, 1/3, 2/3)^T$$
,

and the covariance matrix is

1/18	1/36	0	0
1/36	1/18	0	0
0	0	1/18	1/36
0	0	1/36	1/18

We used the equation

$$R_Y = C_Y + \mu_Y \mu_Y^T$$

to compute the correlation matrix of (Y_1, Y_2, Y_3, Y_4) :

1/6	1/4	1/9	2/9
1/4	1/2	2/9	4/9
1/9	2/9	1/6	1/4
2/9	4/9	1/4	1/2