

Joint Distributions, Independence  
Covariance and Correlation  
18.05 Spring 2014  
Jeremy Orloff and Jonathan Bloom

$X \setminus Y$	1	2	3	4	5	6
1	1/36	1/36	1/36	1/36	1/36	1/36
2	1/36	1/36	1/36	1/36	1/36	1/36
3	1/36	1/36	1/36	1/36	1/36	1/36
4	1/36	1/36	1/36	1/36	1/36	1/36
5	1/36	1/36	1/36	1/36	1/36	1/36
6	1/36	1/36	1/36	1/36	1/36	1/36

## Central Limit Theorem

**Setting:**  $X_1, X_2, \dots$  i.i.d. with mean  $\mu$  and standard dev.  $\sigma$ .

For each  $n$ :

$$\bar{X}_n = \frac{1}{n}(X_1 + X_2 + \dots + X_n)$$

$$S_n = X_1 + X_2 + \dots + X_n.$$

**Conclusion:** For large  $n$ :

$$\bar{X}_n \approx N\left(\mu, \frac{\sigma^2}{n}\right)$$

$$S_n \approx N(n\mu, n\sigma^2)$$

Standardized  $S_n$  or  $\bar{X}_n \approx N(0, 1)$

## Board Question: CLT

1. Carefully write the statement of the central limit theorem.
2. To head the newly formed US Dept. of Statistics, suppose that 50% of the population supports Erika, 25% supports Ruthi, and the remaining 25% is split evenly between Peter, Jon and Jerry.

A poll asks 400 random people who they support. What is the probability that at least 55% of those polled prefer Erika?

3. (Not for class. Solution will be on posted slides.)

An accountant rounds to the nearest dollar. We'll assume the error in rounding is uniform on  $[-.5, .5]$ . Estimate the probability that the total error in 300 entries is more than \$5.

## Joint Distributions

$X$  and  $Y$  are *jointly distributed* random variables.

Discrete: Probability mass function (pmf):

$$p(x_i, y_j)$$

Continuous: probability density function (pdf):

$$f(x, y)$$

Both: cumulative distribution function (cdf):

$$F(x, y) = P(X \leq x, Y \leq y)$$

## Discrete joint pmf: example 1

Roll two dice:  $X = \#$  on first die,  $Y = \#$  on second die

$X$  takes values in  $1, 2, \dots, 6$ ,  $Y$  takes values in  $1, 2, \dots, 6$

**Joint probability table:**

$X \backslash Y$	1	2	3	4	5	6
1	1/36	1/36	1/36	1/36	1/36	1/36
2	1/36	1/36	1/36	1/36	1/36	1/36
3	1/36	1/36	1/36	1/36	1/36	1/36
4	1/36	1/36	1/36	1/36	1/36	1/36
5	1/36	1/36	1/36	1/36	1/36	1/36
6	1/36	1/36	1/36	1/36	1/36	1/36

pmf:  $p(i, j) = 1/36$  for any  $i$  and  $j$  between 1 and 6.

## Discrete joint pmf: example 2

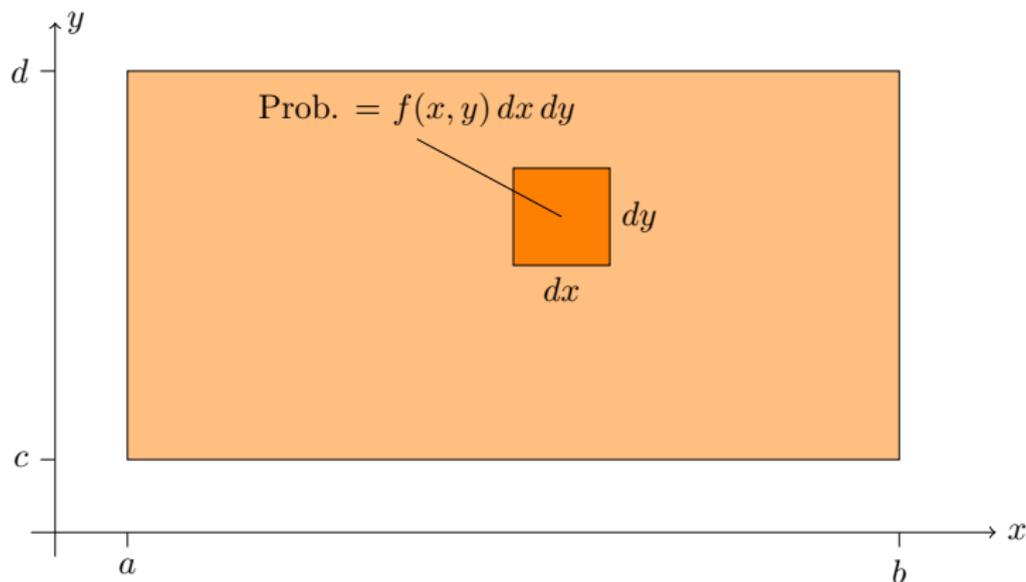
Roll two dice:  $X = \#$  on first die,  $T =$  total on both dice

$X \backslash T$	2	3	4	5	6	7	8	9	10	11	12
1	1/36	1/36	1/36	1/36	1/36	1/36	0	0	0	0	0
2	0	1/36	1/36	1/36	1/36	1/36	1/36	0	0	0	0
3	0	0	1/36	1/36	1/36	1/36	1/36	1/36	0	0	0
4	0	0	0	1/36	1/36	1/36	1/36	1/36	1/36	0	0
5	0	0	0	0	1/36	1/36	1/36	1/36	1/36	1/36	0
6	0	0	0	0	0	1/36	1/36	1/36	1/36	1/36	1/36

## Continuous joint distributions

- $X$  takes values in  $[a, b]$ ,  $Y$  takes values in  $[c, d]$
- $(X, Y)$  takes values in  $[a, b] \times [c, d]$ .
- *Joint probability **density** function* (pdf)  $f(x, y)$

$f(x, y) dx dy$  is the probability of being in the small square.



## Properties of the joint pmf and pdf

### Discrete case: probability mass function (pmf)

1.  $0 \leq p(x_i, y_j) \leq 1$
2. Total probability is 1.

$$\sum_{i=1}^n \sum_{j=1}^m p(x_i, y_j) = 1$$

### Continuous case: probability density function (pdf)

1.  $0 \leq f(x, y)$
2. Total probability is 1.

$$\int_c^d \int_a^b f(x, y) dx dy = 1$$

Note:  $f(x, y)$  can be greater than 1: it is a density *not* a probability.

## Example: discrete events

Roll two dice:  $X = \#$  on first die,  $Y = \#$  on second die.

Consider the event:  $A = 'Y - X \geq 2'$

Describe the event  $A$  and find its probability.

## Example: discrete events

Roll two dice:  $X = \#$  on first die,  $Y = \#$  on second die.

Consider the event:  $A = 'Y - X \geq 2'$

Describe the event  $A$  and find its probability.

**answer:** We can describe  $A$  as a set of  $(X, Y)$  pairs:

$$A = \{(1, 3), (1, 4), (1, 5), (1, 6), (2, 4), (2, 5), (2, 6), (3, 5), (3, 6), (4, 6)\}.$$

Or we can visualize it by shading the table:

$X \backslash Y$	1	2	3	4	5	6
1	1/36	1/36	1/36	1/36	1/36	1/36
2	1/36	1/36	1/36	1/36	1/36	1/36
3	1/36	1/36	1/36	1/36	1/36	1/36
4	1/36	1/36	1/36	1/36	1/36	1/36
5	1/36	1/36	1/36	1/36	1/36	1/36
6	1/36	1/36	1/36	1/36	1/36	1/36

$$P(A) = \text{sum of probabilities in shaded cells} = 10/36.$$

## Example: continuous events

Suppose  $(X, Y)$  takes values in  $[0, 1] \times [0, 1]$ .

Uniform density  $f(x, y) = 1$ .

Visualize the event ' $X > Y$ ' and find its probability.

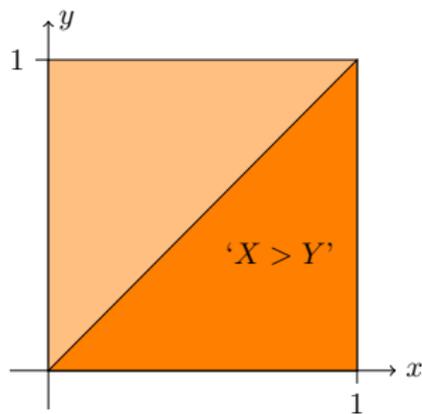
## Example: continuous events

Suppose  $(X, Y)$  takes values in  $[0, 1] \times [0, 1]$ .

Uniform density  $f(x, y) = 1$ .

Visualize the event ' $X > Y$ ' and find its probability.

answer:



The event takes up half the square. Since the density is uniform this is half the probability. That is,  $P(X > Y) = .5$

## Cumulative distribution function

$$F(x, y) = P(X \leq x, Y \leq y) = \int_c^y \int_a^x f(u, v) du dv.$$

$$f(x, y) = \frac{\partial^2 F}{\partial x \partial y}(x, y).$$

### Properties

1.  $F(x, y)$  is non-decreasing. That is, as  $x$  or  $y$  increases  $F(x, y)$  increases or remains constant.
2.  $F(x, y) = 0$  at the lower left of its range.  
If the lower left is  $(-\infty, -\infty)$  then this means

$$\lim_{(x, y) \rightarrow (-\infty, -\infty)} F(x, y) = 0.$$

3.  $F(x, y) = 1$  at the upper right of its range.

## Marginal pmf

Roll two dice:  $X = \#$  on first die,  $T =$  total on both dice.

The marginal pmf of  $X$  is found by summing the rows. The marginal pmf of  $T$  is found by summing the columns

$X \setminus T$	2	3	4	5	6	7	8	9	10	11	12	$p(x_i)$
1	1/36	1/36	1/36	1/36	1/36	1/36	0	0	0	0	0	1/6
2	0	1/36	1/36	1/36	1/36	1/36	1/36	0	0	0	0	1/6
3	0	0	1/36	1/36	1/36	1/36	1/36	1/36	0	0	0	1/6
4	0	0	0	1/36	1/36	1/36	1/36	1/36	1/36	0	0	1/6
5	0	0	0	0	1/36	1/36	1/36	1/36	1/36	1/36	0	1/6
6	0	0	0	0	0	1/36	1/36	1/36	1/36	1/36	1/36	1/6
$p(t_j)$	1/36	2/36	3/36	4/36	5/36	6/36	5/36	4/36	3/36	2/36	1/36	1

## Marginal pdf

**Example.** Suppose  $X$  and  $Y$  take values on the square  $[0, 1] \times [1, 2]$  with joint pdf  $f(x, y) = \frac{8}{3}x^3y$ .

The marginal pdf  $f_X(x)$  is found by *integrating out* the  $y$ . Likewise for  $f_Y(y)$ .

**answer:**

$$f_X(x) = \int_1^2 \frac{8}{3}x^3y \, dy = \left[ \frac{4}{3}x^3y^2 \right]_1^2 = \boxed{4x^3}$$

$$f_Y(y) = \int_0^1 \frac{8}{3}x^3y \, dx = \left[ \frac{2}{3}x^4y \right]_0^1 = \boxed{\frac{2}{3}y}$$

## Board question

Suppose  $X$  and  $Y$  are random variables and

- $(X, Y)$  takes values in  $[0, 1] \times [0, 1]$ .
- the pdf is  $\frac{3}{2}(x^2 + y^2)$ .

1. Show  $f(x, y)$  is a valid pdf.
2. Visualize the event  $A = 'X > .3 \text{ and } Y > .5'$ . Find its probability.
3. Find the cdf  $F(x, y)$ .
4. Find the marginal pdf  $f_X(x)$ . Use this to find  $P(X < .5)$ .
5. Use the cdf  $F(x, y)$  to find the marginal cdf  $F_X(x)$  and  $P(X < .5)$ .
6. See next slide

## Board question continued

6. (New scenario) From the following table compute  $F(3.5, 4)$ .

$X \setminus Y$	1	2	3	4	5	6
1	1/36	1/36	1/36	1/36	1/36	1/36
2	1/36	1/36	1/36	1/36	1/36	1/36
3	1/36	1/36	1/36	1/36	1/36	1/36
4	1/36	1/36	1/36	1/36	1/36	1/36
5	1/36	1/36	1/36	1/36	1/36	1/36
6	1/36	1/36	1/36	1/36	1/36	1/36

# Independence

Events  $A$  and  $B$  are independent if

$$P(A \cap B) = P(A)P(B).$$

Random variables  $X$  and  $Y$  are independent if

$$F(x, y) = F_X(x)F_Y(y).$$

Discrete random variables  $X$  and  $Y$  are independent if

$$p(x_i, y_j) = p_X(x_i)p_Y(y_j).$$

Continuous random variables  $X$  and  $Y$  are independent if

$$f(x, y) = f_X(x)f_Y(y).$$

## Concept question: independence I

Roll two dice:  $X$  = value on first,  $Y$  = value on second

$X \setminus Y$	1	2	3	4	5	6	$p(x_i)$
1	1/36	1/36	1/36	1/36	1/36	1/36	1/6
2	1/36	1/36	1/36	1/36	1/36	1/36	1/6
3	1/36	1/36	1/36	1/36	1/36	1/36	1/6
4	1/36	1/36	1/36	1/36	1/36	1/36	1/6
5	1/36	1/36	1/36	1/36	1/36	1/36	1/6
6	1/36	1/36	1/36	1/36	1/36	1/36	1/6
$p(y_j)$	1/6	1/6	1/6	1/6	1/6	1/6	1

Are  $X$  and  $Y$  independent?      1. Yes      2. No

## Concept question: independence II

Roll two dice:  $X$  = value on first,  $T$  = sum

$X \setminus T$	2	3	4	5	6	7	8	9	10	11	12	$p(x_i)$
1	1/36	1/36	1/36	1/36	1/36	1/36	0	0	0	0	0	1/6
2	0	1/36	1/36	1/36	1/36	1/36	1/36	0	0	0	0	1/6
3	0	0	1/36	1/36	1/36	1/36	1/36	1/36	0	0	0	1/6
4	0	0	0	1/36	1/36	1/36	1/36	1/36	1/36	0	0	1/6
5	0	0	0	0	1/36	1/36	1/36	1/36	1/36	1/36	0	1/6
6	0	0	0	0	0	1/36	1/36	1/36	1/36	1/36	1/36	1/6
$p(y_j)$	1/36	2/36	3/36	4/36	5/36	6/36	5/36	4/36	3/36	2/36	1/36	1

Are  $X$  and  $Y$  independent?      1. Yes      2. No

## Concept Question

Among the following pdf's which are independent? (Each of the ranges is a rectangle chosen so that  $\int \int f(x, y) dx dy = 1$ .)

i)  $f(x, y) = 4x^2y^3$ .

ii)  $f(x, y) = \frac{1}{2}(x^3y + xy^3)$ .

iii)  $f(x, y) = 6e^{-3x-2y}$

Put a 1 for independent and a 0 for not-independent.

(a) 111    (b) 110    (c) 101    (d) 100

(e) 011    (f) 010    (g) 001    (h) 000

## Covariance

Measures the degree to which two random variables vary together, e.g. height and weight of people.

$X$ ,  $Y$  random variables with means  $\mu_X$  and  $\mu_Y$

$$\text{Cov}(X, Y) = E((X - \mu_X)(Y - \mu_Y)).$$

# Properties of covariance

## Properties

1.  $\text{Cov}(aX + b, cY + d) = ac\text{Cov}(X, Y)$  for constants  $a, b, c, d$ .
2.  $\text{Cov}(X_1 + X_2, Y) = \text{Cov}(X_1, Y) + \text{Cov}(X_2, Y)$ .
3.  $\text{Cov}(X, X) = \text{Var}(X)$
4.  $\text{Cov}(X, Y) = E(XY) - \mu_X\mu_Y$ .
5. If  $X$  and  $Y$  are independent then  $\text{Cov}(X, Y) = 0$ .
6. **Warning:** The converse is not true, if covariance is 0 the variables might not be independent.

## Concept question

Suppose we have the following joint probability table.

$Y \setminus X$	-1	0	1	$p(y_j)$
0	0	1/2	0	1/2
1	1/4	0	1/4	1/2
$p(x_i)$	1/4	1/2	1/4	1

We compute the covariance using the following steps.

1.  $E(X) = 0$      $E(Y) = 1/2$
2.  $E(XY) = -1/4 + 1/4 = 0$
3. Therefore  $\text{Cov}(X, Y) = E(XY) - E(X)E(Y) = 0$ .

Because the covariance is 0 we know that  $X$  and  $Y$  are independent

1. True
2. False

## Concept question

Suppose we have the following joint probability table.

$Y \setminus X$	-1	0	1	$p(y_j)$
0	0	1/2	0	1/2
1	1/4	0	1/4	1/2
$p(x_i)$	1/4	1/2	1/4	1

We compute the covariance using the following steps.

1.  $E(X) = 0$      $E(Y) = 1/2$
2.  $E(XY) = -1/4 + 1/4 = 0$
3. Therefore  $\text{Cov}(X, Y) = E(XY) - E(X)E(Y) = 0$ .

Because the covariance is 0 we know that  $X$  and  $Y$  are independent

1. True
2. False

Key point: covariance measures the linear relationship between  $X$  and  $Y$ . It can completely miss a quadratic or higher order relationship.

## Board question: computing covariance

Flip a fair coin 3 times.

Let  $X$  = number of heads in the first 2 flips

Let  $Y$  = number of heads on the last 2 flips.

Compute  $\text{Cov}(X, Y)$ ,

## Stop and rest

We'll stop here and finish the remainder of these slides in the next class.

## Correlation

Like covariance, but removes scale.

The *correlation coefficient* between  $X$  and  $Y$  is defined by

$$\text{Cor}(X, Y) = \rho = \frac{\text{Cov}(X, Y)}{\sigma_X \sigma_Y}.$$

Properties:

1.  $\rho$  is the covariance of the standardized versions of  $X$  and  $Y$ .
2.  $\rho$  is dimensionless (it's a ratio).
3.  $-1 \leq \rho \leq 1$ .  $\rho = 1$  if and only if  $Y = aX + b$  with  $a > 0$  and  $\rho = -1$  if and only if  $Y = aX + b$  with  $a < 0$ .

## Real-life correlations

- Over time, amount of Ice cream consumption is correlated with number of pool drownings.
- In 1685 (and today) being a student is the most dangerous profession.
- In 90% of bar fights ending in a death the person who started the fight died.
- Hormone replacement therapy (HRT) is correlated with a lower rate of coronary heart disease (CHD).

## Correlation is not causation

Edward Tufte: "Empirically observed covariation is a necessary but not sufficient condition for causality."

## Overlapping sums of uniform random variables

We made two random variables  $X$  and  $Y$  from overlapping sums of uniform random variables

For example:

$$X = X_1 + X_2 + X_3 + X_4 + X_5$$

$$Y = X_3 + X_4 + X_5 + X_6 + X_7$$

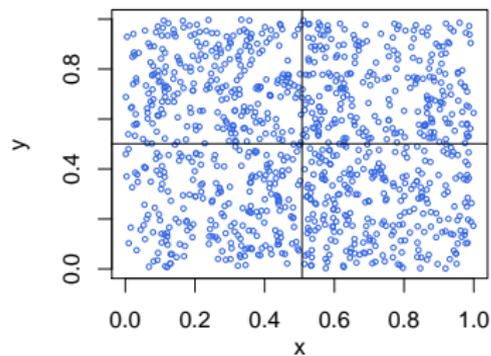
These are sums of 5 of the  $X_i$  with 3 in common.

If we sum  $r$  of the  $X_i$  with  $s$  in common we name it  $(r, s)$ .

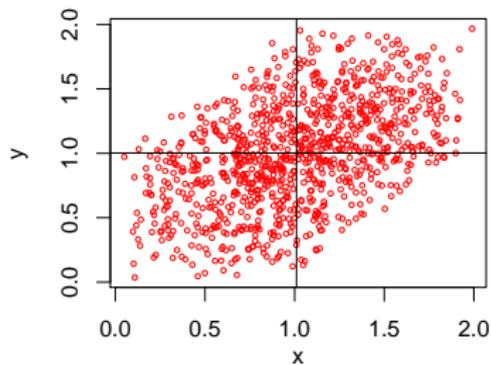
Below are a series of scatterplots produced using R.

# Scatter plots

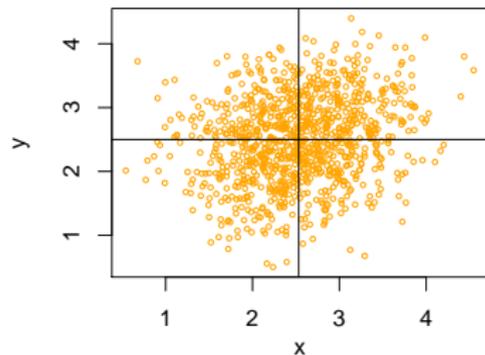
**(1, 0) cor=0.00, sample\_cor=-0.07**



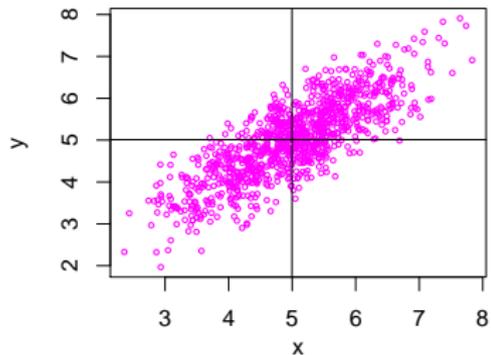
**(2, 1) cor=0.50, sample\_cor=0.48**



**(5, 1) cor=0.20, sample\_cor=0.21**



**(10, 8) cor=0.80, sample\_cor=0.81**



## Concept question

Toss a fair coin  $2n + 1$  times. Let  $X$  be the number of heads on the first  $n + 1$  tosses and  $Y$  the number on the last  $n + 1$  tosses.

If  $n = 1000$  then  $\text{Cov}(X, Y)$  is:

- (a) 0      (b)  $1/4$       (c)  $1/2$       (d) 1
- (e) More than 1      (f) tiny but not 0

## Board question

Toss a fair coin  $2n + 1$  times. Let  $X$  be the number of heads on the first  $n + 1$  tosses and  $Y$  the number on the last  $n + 1$  tosses.

Compute  $\text{Cov}(X, Y)$  and  $\text{Cor}(X, Y)$ .

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