

AFRE 802

Statistical Methods for Agricultural, Food, & Resource Economists



Multivariate probability distributions – Part 1 of 3 (WMS Ch. 5.1-5.2)

October 5, 2017
Nicole Mason
Michigan State University
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GAME PLAN

Collect Ch. 4 HW

Review

Graded in-class exercise

Multivariate probability distributions (Part 1 of 3)

1. Bivariate probability distributions (discrete & continuous)
2. Multivariate probability distributions (discrete & continuous)

Review: standard normal distribution

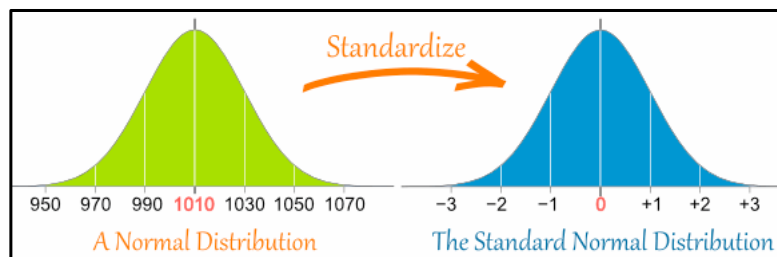
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Can convert any normal RV to standard normal

- Suppose $Y \sim N(\mu, \sigma^2)$, then

$$Z = \frac{Y - \mu}{\sigma}, \quad Z \sim N(0,1) = \text{standard normal}$$

- Once converted to standard normal \rightarrow use Table 4 ($P(Z > z)$)



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Review: gamma distribution

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- Non-negative** and **skewed to the right** (graphs on next slide)

$$f(y) = \begin{cases} \frac{y^{\alpha-1} e^{-y/\beta}}{\beta^\alpha \Gamma(\alpha)}, & 0 \leq y < \infty \\ 0, & \text{elsewhere} \end{cases}$$

where $\Gamma(\alpha) = \int_0^\infty y^{\alpha-1} e^{-y} dy$
and $\alpha > 0, \beta > 0$

$\Gamma(\alpha)$: "gamma function"

$$\Gamma(1) = 1$$

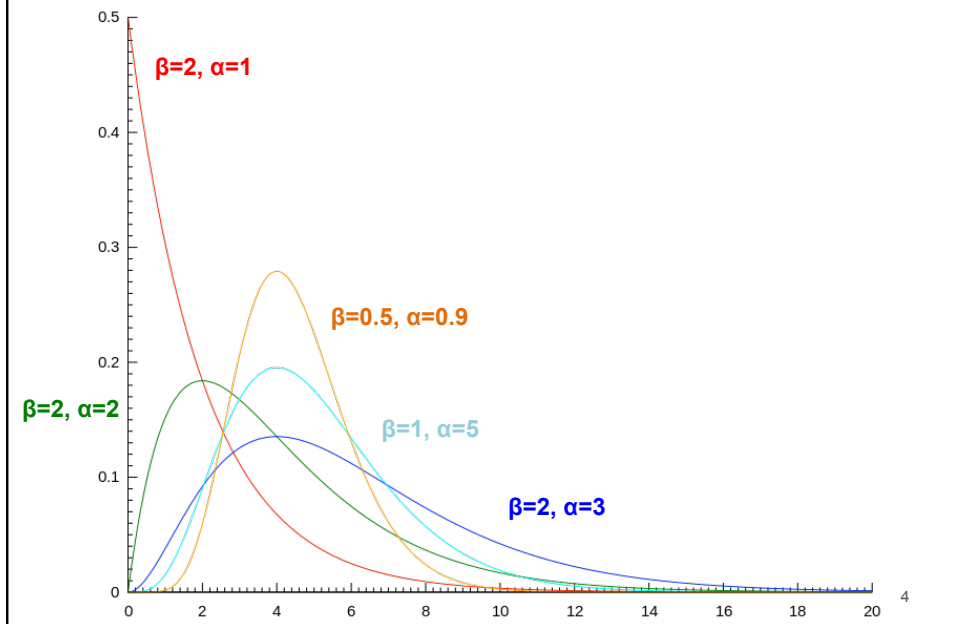
$$\Gamma(n) = (n-1)! \text{ if } n \text{ is an integer}$$

$$E(Y) = \alpha\beta$$

$$V(Y) = \alpha\beta^2$$

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Examples of gamma PDFs



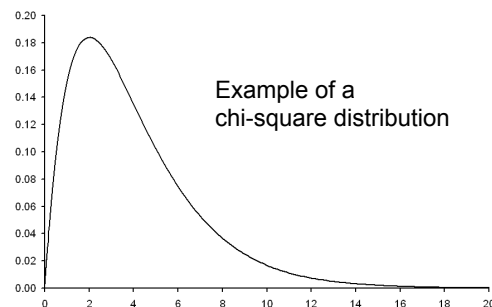
Review: gamma special case #1 – chi square

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1. **Chi-square** (χ^2) distribution: gamma w/ $\alpha = \frac{\nu}{2}$, $\beta = 2$
where ν is referred to as the number of degrees of freedom

$$E(Y) = \alpha\beta = \frac{\nu}{2} \cdot 2 = \nu$$

$$V(Y) = \alpha\beta^2 = \frac{\nu}{2} \cdot 2^2 = 2\nu$$



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Review: gamma special case #2 – exponential

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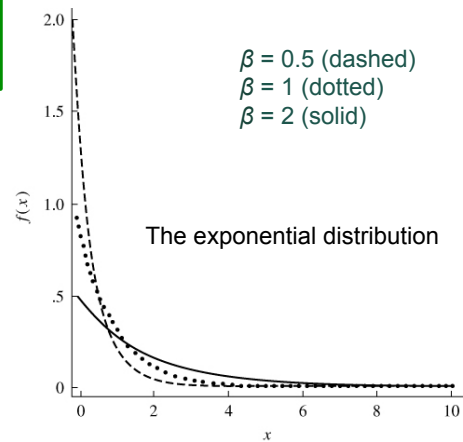
2. Exponential distribution: gamma w/ $\alpha = 1, \beta > 0$

$$f(y) = \begin{cases} \frac{e^{-y/\beta}}{\beta}, & 0 \leq y \leq \infty \\ 0, & \text{elsewhere} \end{cases}$$

$$F(y) = \begin{cases} 0, & y < 0 \\ 1 - e^{-y/\beta}, & y \geq 0 \end{cases}$$

$$E(Y) = \alpha\beta = 1\beta = \beta$$

$$V(Y) = \alpha\beta^2 = 1\beta^2 = \beta^2$$



Distribution	Probability Function	Mean	Variance
Uniform	$f(y) = \frac{1}{\theta_2 - \theta_1}; \theta_1 \leq y \leq \theta_2$ $F(y) = 0$ for $y < \theta_1$; $\frac{y - \theta_1}{\theta_2 - \theta_1}$ for $\theta_1 \leq y \leq \theta_2$; 1 for $y > \theta_2$	$\frac{\theta_1 + \theta_2}{2}$	$\frac{(\theta_2 - \theta_1)^2}{12}$
Normal	$f(y) = \frac{1}{\sigma\sqrt{2\pi}} \exp\left[-\left(\frac{1}{2\sigma^2}\right)(y - \mu)^2\right]$ $-\infty < y < +\infty$ $Z = \frac{Y - \mu}{\sigma}, \quad Z \sim N(0,1) = \text{standard normal}$	μ	σ^2
Exponential	$f(y) = \frac{1}{\beta} e^{-y/\beta}; \quad \beta > 0$ $0 < y < \infty$ $F(y) = \begin{cases} 0, & y < 0 \\ 1 - e^{-y/\beta}, & y \geq 0 \end{cases}$	β	β^2
Gamma	$f(y) = \left[\frac{1}{\Gamma(\alpha)\beta^\alpha}\right] y^{\alpha-1} e^{-y/\beta};$ $0 < y < \infty$ where $\Gamma(\alpha) = \int_0^\infty y^{\alpha-1} e^{-y} dy$ $\Gamma(n) = (n-1)!$ if n is an integer	$\alpha\beta$	$\alpha\beta^2$
Chi-square	$f(y) = \frac{(y)^{(v/2)-1} e^{-y/2}}{2^{v/2} \Gamma(v/2)};$ $y > 0$	v	$2v$

Graded in-class exercise

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Multivariate probability distributions

- We are often interested in the intersection of 2 (or more) events
 - EX) Blackjack
 - A: drawing an ace
 - B: drawing a face card
 - EX) Egg producer
 - A: A chicken produces n eggs
 - B: y of those eggs are bad
- Need to understand the joint probability distributions (bivariate = 2, multivariate ≥ 2)
- We'll discuss these for discrete & continuous RVs

Bivariate probability distribution for discrete RV

Example

- Roll a pair of dice. *How many ordered pairs of #s?*
 - $mn=6*6=36$
- Let (y_1, y_2) represent the (# of 1st die, # on 2nd die)
 - EX) Pair of 1s: (1, 1)
 - EX) 2 on 1st die, 3 on 2nd die: (2, 3)
- Consider all pairs (y_1, y_2) . *What is the probability of each (ordered) pair?*
 - $1/36$

→ Bivariate probability distribution for this example:

$$p(y_1, y_2) = P(Y_1 = y_1, Y_2 = y_2) = 1/36,$$

where $y_1 = 1, 2, 3, 4, 5, 6$, $y_2 = 1, 2, 3, 4, 5, 6$

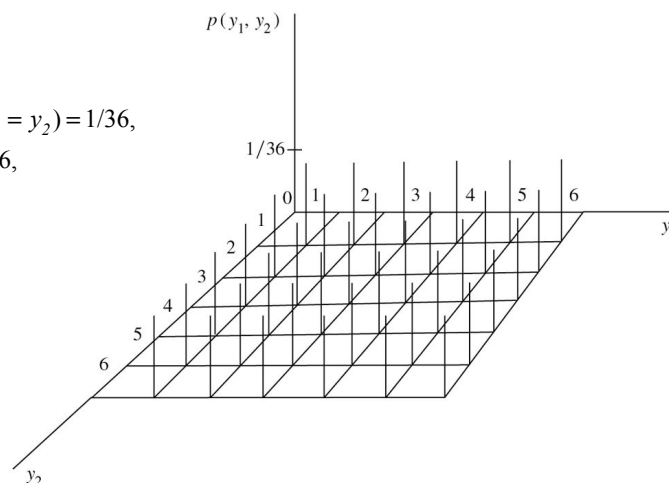
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Graphical representation of rolling a pair of die bivariate probability distribution

$$p(y_1, y_2) = P(Y_1 = y_1, Y_2 = y_2) = 1/36,$$

where $y_1 = 1, 2, 3, 4, 5, 6$,

$y_2 = 1, 2, 3, 4, 5, 6$



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Bivariate (or joint) probability distribution for discrete RVs (in general)

$$p(y_1, y_2) = P(Y_1 = y_1, Y_2 = y_2)$$

where $-\infty < y_1 < \infty, \quad -\infty < y_2 < \infty$

The usual rules for probabilities apply, but now they apply to the joint probability, $p(y_1, y_2)$. *What is the interpretation?*

1. $0 \leq p(y_1, y_2) \leq 1$
2. $\sum_{y_1, y_2} p(y_1, y_2) = 1$

Process for finding the bivariate probability distribution for discrete RVs: (1) find all (ordered) pairs of values; (2) assign a probability to each pair.

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EX) In the toss 2 dice experiment, what is $P(2 \leq Y_1 \leq 3, 1 \leq Y_2 \leq 2)$?

Rolls that satisfy this condition:

- (2, 1)
- (2, 2)
- (3, 1)
- (3, 2)

Probability of each is $1/36$

Mutually exclusive so can add up the probabilities

$$\rightarrow (1/36) + (1/36) + (1/36) + (1/36) = 4/36 = 1/9$$

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Bivariate probability distribution (discrete RVs) MICHIGAN STATE UNIVERSITY

EX) A supermarket has 3 checkout counters. Two customers arrive at the counters at different times when the counters are serving no other customers. (Both can choose the same counter.)

- Let Y_1 denote the number of customers (out of the two customers) who choose counter 1.
- Let Y_2 denote the number of customers who choose counter 2.

Find the joint probability function of Y_1 and Y_2 .

1. Define the sample space. Let $\{i, j\}$ denote that the 1st customer chooses counter i , 2nd chooses counter j .
E.g., $\{1, 3\}$ = 1st customer chooses counter 1, 2nd customer chooses counter 3. *How many pairs of counter choices are there?*
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2. Map these into (y_1, y_2) pairs (in a table is easiest). Assign probabilities to each (y_1, y_2) pair.

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Find the joint probability function of Y_1 and Y_2 .

1. Define the sample space. Let $\{i, j\}$ denote that the 1st customer chooses counter i , 2nd chooses counter j .
E.g., $\{1, 3\}$ = 1st customer chooses counter 1, 2nd customer chooses counter 3.
2. Map these into (y_1, y_2) pairs (in a table is easiest). Assign probabilities to each (y_1, y_2) pair.

Sample space
(counters chosen)

	y_1 (# of customers choosing counter 1)		
y_2 (# of customers choosing counter 2)	0	1	2
0			
1			
2			

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Bivariate CDF (discrete RVs)

$$F(y_1, y_2) = P(Y_1 \leq y_1, Y_2 \leq y_2) = \sum_{t_2 \leq y_2} \sum_{t_1 \leq y_1} p(t_1, t_2)$$

where $-\infty < y_1 < \infty$, $-\infty < y_2 < \infty$

EX) In the toss 2 dice experiment, what is $F(2, 3) = P(Y_1 \leq 2, Y_2 \leq 3)$? Recall that there are 36 possible ordered pairs, each w/ probability 1/36.

$$= p(1, 1) + p(1, 2) + p(1, 3) + p(2, 1) + p(2, 2) + p(2, 3) = 6 \cdot (1/36) = 1/6$$

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Bivariate cumulative distribution function (CDF)
(discrete RVs)

$$F(y_1, y_2) = P(Y_1 \leq y_1, Y_2 \leq y_2) = \sum_{t_2 \leq y_2} \sum_{t_1 \leq y_1} p(t_1, t_2)$$

EX) In the supermarket checkout counter example, find:

a. $F(-1, 2)$

b. $F(1.5, 2)$

c. $F(5, 7)$

y_2	y_1		
	0	1	2
0	1/9	2/9	1/9
1	2/9	2/9	0
2	1/9	0	0

Bivariate CDF (discrete RVs)

$$F(y_1, y_2) = P(Y_1 \leq y_1, Y_2 \leq y_2) = \sum_{t_2 \leq y_2} \sum_{t_1 \leq y_1} p(t_1, t_2)$$

Both discrete & continuous bivariate CDFs must satisfy similar properties to what we saw in the univariate case:

1. $F(-\infty, -\infty) = 0$, $F(-\infty, y_2) = 0$, $F(y_1, -\infty) = 0$
2. $F(\infty, \infty) = 1$

Bivariate CDF (continuous RVs)

$$F(y_1, y_2) = P(Y_1 \leq y_1, Y_2 \leq y_2) = \int_{-\infty}^{y_2} \int_{-\infty}^{y_1} f(t_1, t_2) dt_1 dt_2$$

where $-\infty < y_1 < \infty$, $-\infty < y_2 < \infty$

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Bivariate PDFs (continuous RVs)

$$f(y_1, y_2)$$

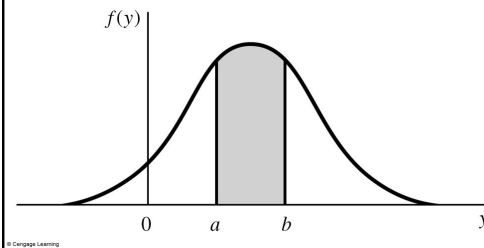
Like PDFs for the univariate case (and bivariate probability distributions for discrete RVs), bivariate PDFs must satisfy similar properties:

1. $0 \leq f(y_1, y_2) \leq 1$ for all y_1, y_2
2. $\int_{-\infty}^{\infty} \int_{-\infty}^{\infty} f(y_1, y_2) dy_1 dy_2 = 1$

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Graphical representation of univariate vs. bivariate PDFs

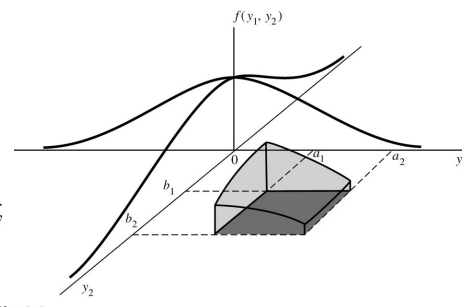
Univariate



Recall that in the univariate case, area under the PDF between a and $b = P(a \leq Y \leq b)$

$$P(a \leq Y \leq b) = \int_a^b f(y) dy$$

Bivariate



Bivariate PDFs are 2-dimensional surfaces, so the volume under the surface corresponds to a probability – e.g., above:

$$P(a_1 \leq Y_1 \leq a_2, b_1 \leq Y_2 \leq b_2) = \int_{b_1}^{b_2} \int_{a_1}^{a_2} f(y_1, y_2) dy_1 dy_2$$

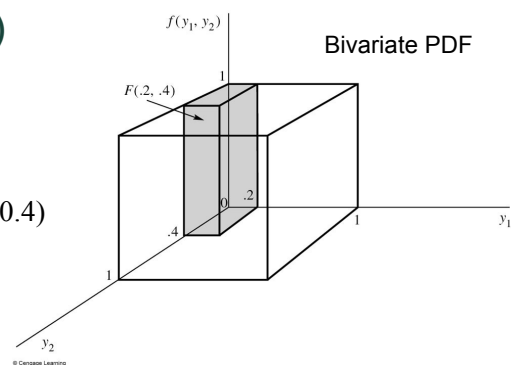
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EX1) Finding probabilities from a bivariate PDF (continuous RVs)

$$f(y_1, y_2) = \begin{cases} 1, & 0 \leq y_1 \leq 1, 0 \leq y_2 \leq 1 \\ 0, & \text{elsewhere} \end{cases}$$

a. Find $F(0.2, 0.4) = P(Y_1 \leq 0.2, Y_2 \leq 0.4)$

b. Find $P(0.1 \leq y_1 \leq 0.3, 0 \leq y_2 \leq 0.5)$



$$P(a_1 \leq Y_1 \leq a_2, b_1 \leq Y_2 \leq b_2) = \int_{b_1}^{b_2} \int_{a_1}^{a_2} f(y_1, y_2) dy_1 dy_2$$

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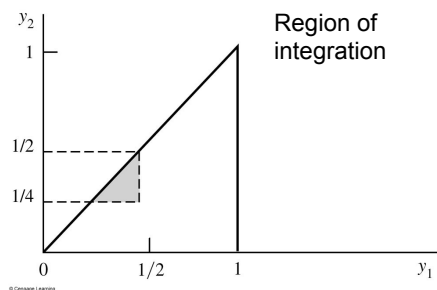
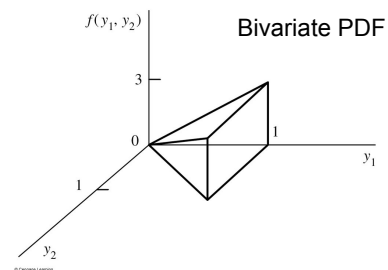
EX1) Finding probabilities from a bivariate PDF (continuous RVs)

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EX2) Finding probabilities from a bivariate PDF (continuous RVs)

$$f(y_1, y_2) = \begin{cases} 3y_1, & 0 \leq y_2 \leq y_1 \leq 1 \\ 0, & \text{elsewhere} \end{cases}$$

Find $P(0 \leq Y_1 \leq 0.5, Y_2 > 0.25)$

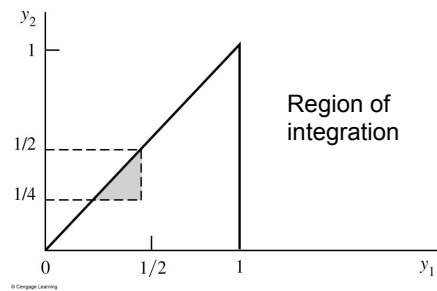


$$P(a_1 \leq Y_1 \leq a_2, b_1 \leq Y_2 \leq b_2) = \int_{b_1}^{b_2} \int_{a_1}^{a_2} f(y_1, y_2) dy_1 dy_2$$

EX2) Finding probabilities from a bivariate PDF (continuous RVs)

$$f(y_1, y_2) = \begin{cases} 3y_1, & 0 \leq y_2 \leq y_1 \leq 1 \\ 0, & \text{elsewhere} \end{cases}$$

Find $P(0 \leq Y_1 \leq 0.5, Y_2 > 0.25)$

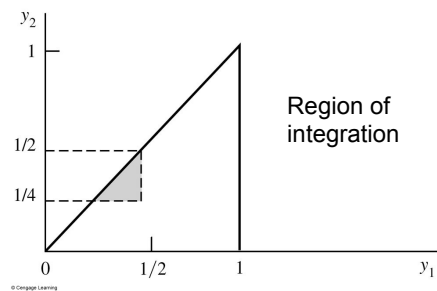


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EX2) Finding probabilities from a bivariate PDF (continuous RVs)

$$f(y_1, y_2) = \begin{cases} 3y_1, & 0 \leq y_2 \leq y_1 \leq 1 \\ 0, & \text{elsewhere} \end{cases}$$

Find $P(0 \leq Y_1 \leq 0.5, Y_2 > 0.25)$



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Moving beyond the bivariate case

Joint probability distributions for discrete RVs :

$$p(y_1, y_2, \dots, y_n) = P(Y_1 = y_1, Y_2 = y_2, \dots, Y_n = y_n)$$

Joint probability density function (PDF) for continuous RVs :

$$f(y_1, y_2, \dots, y_n)$$

Joint cumulative distribution function (CDF) for discrete & continuous RVs :

$$F(y_1, y_2, \dots, y_n) = P(Y_1 \leq y_1, Y_2 \leq y_2, \dots, Y_n \leq y_n)$$

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Homework:

- WMS Ch. 5 (part 1 of 3)
 - Bivariate & multivariate probability distributions: 5.1, 5.2, 5.4, 5.7, 5.8
- Ch. 5 will not be collected (b/c would be due the class before the midterm)

Next class:

- Multivariate probability distributions, cont'd (Part 2 of 3)
 - Marginal & conditional probability distributions
 - Independent RVs

Reading for next class:

- WMS Ch. 5 (sections 5.3-5.4)

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