Some required calculus skills for STAT 542 & STAT 543

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Outline

- Topics in STAT 542
- Topics in STAT 543
- Required calculus for STAT 542 & STAT 543

Topics in STAT 542

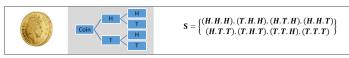
- Probability Theory
- Transformation and Expectations
- Common Families of Distributions
- Multiple Random Variables
- Properties of Random Sample

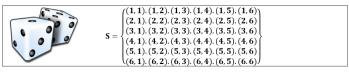
Topics in STAT 543

- Principles of Data Reduction
- Point Estimation
- Hypothesis Testing
- Interval estimation
- Asymptotic Evaluations

Sample Space

• The set of all possible outcomes of an experiment is called the sample space.



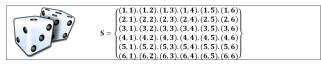


Random Variable

• A random variable is a real-valued function defined over the elements of a sample

space.

Roll a pair of dice and define X as a total sum of the points.



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if \omega \in \{(1,1)\}
                                                                                  {X = 2} = {(1,1)}
           if \omega \in \{(1,2)(2,1)\}
                                                                                  {X = 4} = {(1,3)(2,2)(3,1)}
       4 if \omega \in \{(1,3)(2,2)(3,1)\}
                                                                                  \{X = 5\} = \cdots
          if \omega \in \{(1,4)(2,3)(3,2)(4,1)\}
                                                                                  {X = 6} = \cdots
       6 if \omega \in \{(1,5)(2,4)(3,3)(4,2)(5,1)\}
                                                                                  {X = 7} = \cdots
                                                                    means
X = \{7 \text{ if } \omega \in \{(1,6)(2,5)(3,4)(4,3)(5,2)(6,1)\}
                                                                                  {X = 8} = \cdots
                                                                    Outcome
          if \omega \in \{(3,5)(5,3)(4,4)(2,6)(6,2)\}
                                                                                  {X = 9} = \cdots
            if \omega \in \{(3,6)(4,5)(5,4)(6,3)\}
                                                                                  {X = 10} = {(4,6)(5,5)(6,4)}
       10 if \omega \in \{(4,6)(5,5)(6,4)\}
                                                                                  {X = 11} = {(5,6)(6,5)}
       11 if \omega \in \{(5,6)(6,5)\}
                                                                                  {X = 12} = {(6.6)}
      12 if \omega \in \{(6.6)\}
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Figure: The RV X is defined as the sum of points (x, y)



Discrete & Continuous Random Variable



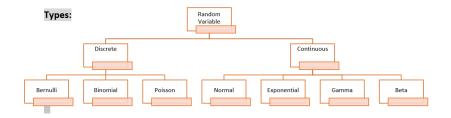


Figure: Some well-known Random Variables

PDF, PMF, CDF

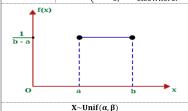
Discrete Random Variable Probability distribution of X		Continuous Random Variable Probability density function	
A function can serve as the probability distribution function of a discrete random variable X if and only if its values, f(x), satisfy the condition.	 f(x) ≥ 0 for each value within its domain; ∑_x f(x) = 1, where the summation extends over all the values within its domain. 	A function can serve as the probability density function of a continuous random variable X if and only if its values, f(x), satisfy the condition.	 f(x) ≥ 0 for each value within its domain; ∫_{-∞}[∞] f(x)dx = 1, where the summation extends over all the values within its domain
Distribution function of X (or cumulative distribution of X (CDF))		Distribution function of X (or cumulative distribution of X (CDF))	
$F(x) = P(X \le x) = \sum_{y \le x} P(X = y)$		$F(x) = P(X \le x) = \int_{-\infty}^{x} f(x) dx$ $for - \infty < x < \infty$	

PDF, CDF: Example

A random variable X has a Uniform Distribution and it is referred to as a continuous uniform random variable if and only if its probability density is given by

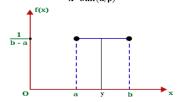
PDF:
$$f(x)$$

$$f(x) = \begin{cases} \frac{1}{\beta - \alpha}, & \text{for } \alpha < x < \beta \\ 0, & \text{elsewhere.} \end{cases}$$



Note: f(x) is a density function, because

$$\int_{\alpha}^{\beta} f(x) \cdot dx = \int_{\alpha}^{\beta} \frac{1}{\beta - \alpha} \cdot dx = \frac{1}{\beta - \alpha} \cdot \int_{\alpha}^{\beta} dx$$
$$= \frac{1}{\beta - \alpha} \cdot (\beta - \alpha) = 1$$



Note: If $X \sim Unif(\alpha, \beta)$ then

Derivative formulas

$$\frac{d}{dx}(c) = 0$$

$$\frac{d}{dx}(x) = 1$$

$$\frac{d}{dx}(cx) = c \cdot \frac{d}{dx}(x) = c$$

$$\frac{d}{dx}(u+v+\cdots) = \frac{d}{dx}(u) + \frac{d}{dx}(v) + \cdots$$

$$\frac{d}{dx}(uv) = u \cdot \frac{d}{dx}(v) + v \cdot \frac{d}{dx}(u)$$

$$\frac{d}{dx}\binom{u}{v} = \frac{v \cdot \frac{d}{dx}(u) - u \cdot \frac{d}{dx}(v)}{v^2}, \quad v$$

$$\frac{d}{dx}(uvw) = uv \cdot \frac{d}{dx}(vw) + uw \cdot \frac{d}{dx}(v) + vw \cdot \frac{d}{dx}(u)$$

$$\frac{d}{dx}(cu) = c \cdot \frac{d}{dx}(u), \quad \frac{d}{dx}\left(\frac{c}{u}\right) = c \cdot \frac{d}{dx}\left(\frac{1}{u}\right) = -\frac{c}{u^2} \cdot \frac{d}{dx}(u), \qquad u \neq 0$$

$$\frac{d}{dx}(x^m) = m \cdot x^{m-1}$$

$$\frac{d}{dx}(u^m) = m \cdot u^{m-1} \frac{d}{dx}(u)$$

$$z = y^m$$
$$y = u(x)$$

$$\frac{d}{dx}(e^x) = e^x$$
 , $\frac{d}{dx}lnx = \frac{1}{x}$

$$\frac{d}{dx}e^{g(x)} = g'(x) \cdot e^{g(x)}$$

$$\frac{d}{dx}a^{x} = a^{x} \cdot ln(a), \ \frac{d}{dx}ln[g(x)] = \frac{g'(x)}{g(x)}$$

$$\frac{d}{dx}f[g(x)] = g(x)' \cdot f'[g(x)]$$



Integration formulas

$\int a dx = a \cdot x$	$\int a f(x) dx = a \int f(x) dx$
$\int (u \pm v) dx = \int u dx \pm \int v dx$	$\int u dv = u \cdot v - \int v du$
	$\int \frac{dx}{x} = \ln x $
$\int \sqrt{x^m} \ dx = \frac{2x \cdot \sqrt{x^m}}{m+2} \qquad (m \neq -2)$	$\int \sqrt[p]{m} \ dx = \frac{px \cdot \sqrt[p]{x^m}}{m+p} \qquad (m+p \neq 0)$
$\int e^x dx = e^x$	$\int e^{ax} dx = \frac{e^{ax}}{a} \qquad (a \neq 0)$

Mean & Variance

Continuous Random Variable

If X is a continuous random variable and f(x) is the value of its probability density at x, the expected value of X is

$$E(X) = \int_{-\infty}^{\infty} x \cdot f(x) dx$$

If X is a continuous random variable and f(x) is the value of its probability density at x, the expected value of g(X) is given by

$$E[g(X)] = \int g(x) \cdot f(x) dx$$

The mean and variance of the uniform distribution X are:

$$E(X) = \int_{\alpha}^{\beta} x \cdot f(x) \cdot dx = \int_{\alpha}^{\beta} x \cdot \frac{1}{\beta - \alpha} \cdot dx = \frac{1}{\beta - \alpha} \cdot \int_{\alpha}^{\beta} x \cdot dx = \frac{1}{\beta - \alpha} \cdot \frac{x^2}{2} \Big|_{\alpha}^{\beta} = \frac{\alpha + \beta}{2}$$

$$\begin{aligned} \mathbb{E}[X^2] &= \int\limits_{\alpha}^{\beta} x^2 \cdot f(x) \cdot dx = \int\limits_{\alpha}^{\beta} x^2 \cdot \frac{1}{\beta - \alpha} \cdot dx = \frac{1}{\beta - \alpha} \cdot \int\limits_{\alpha}^{\beta} x^2 \cdot dx = \frac{1}{\beta - \alpha} \cdot \frac{x^3}{3} \Big|_{\alpha}^{\beta} \\ &= \frac{1}{3(\beta - \alpha)} \cdot (\beta^3 - \alpha^3) \end{aligned}$$

and the variance is following by

$$Var(X) = E[X^{2}] - [E(X)]^{2} = \frac{(\beta^{3} - \alpha^{3})}{3(\beta - \alpha)} - \left(\frac{\alpha + \beta}{2}\right)^{2} = \frac{1}{12}(\beta - \alpha)^{2}$$

Multiple Random Variables

Joint PMF: Two RVs X & Y.		Joint PDF: Two RVs X & Y.		
$f_{X,Y}(x,y) = P(X \le x, Y \le y)$		$f_{X,Y}(x,y) = P(X \le x, Y \le y)$		
A bivariate function can serve as the joint probability density function of a pair of discrete random variables X and Y if and only if its values, f(x, y), satisfy the conditions	 f(x,y) ≥ 0 for each pair of values (x, y) within its domain; ∑_x Σ_y f(x,y) = 1, where the double summation extends over all possible pairs (x, y) within its domain. 	A bivariate function can serve as the joint probability density function of a pair of continues random variables X and Y if and Only if its values, f(x, y), satisfy the conditions	 f(x,y) ≥ 0 for each pair of values (x, y) within its domain: ∫_{-∞} ∫_{-∞} ∫_{-∞} f(x,y) dx dy = 1 , where the double summation extends over all possible pairs (x, y) within its domain. 	
Joint CDF: Two RVs X & Y.		Joint CDF: Two RVs X & Y.		
$F_{X,Y}(x,y) = P(X \le x, Y \le y) = \sum_{s \le x} \sum_{t \le y} f(s,t)$ For $-\infty < x < \infty, -\infty < y < \infty$		$F_{X,Y}(x,y) = P(X \le x, Y \le y) = \int_{-\infty}^{y} \int_{-\infty}^{x} f(s,t)ds dt$		
Marginal Distribution:		For $-\infty < x < \infty, -\infty < y < \infty$ Marginal Density Function:		
If X and Y are discrete random variables and f(x, y) is the value of their joint probability distribution at (x, y), the function given by		If X and Y are continuous random variables and $f(x, y)$ is the value of their joint probability density at (x, y) , the function given by		
For each x within	$P(X = x) = \sum_{y} P(X = x, Y)$ $= y$	$f_X(x) = \int_{-\infty}^{\infty} f(x,y) dy$ $f_X(x) = \int_{-\infty}^{\infty} f(x,y) dy$ $f_X(x) \text{ is called the marginal density of X.}$		



Multiple Random Variables: Example

Let X and Y be continuous random variables having joint probability density function:

$$f_{X,Y}(\mathbf{x},\mathbf{y}) = \begin{cases} k(x^2 + y^2), & 0 \le \mathbf{x} \le 1, \\ 0, & 0. \end{cases} \quad 0 \le \mathbf{y} \le 1$$

- a) Find the constant k.
- b) Compute $P\left(X > \frac{1}{2}, Y < \frac{1}{2}\right)$.
- c) Are X and Y independent? Justify your answer.

a)
$$\frac{\text{Verify } \int_{0}^{1} \int_{0}^{1} f_{X,Y}(x,y) \, dx \, dy = 1}{\int_{0}^{1} \int_{0}^{1} f_{X,Y}(x,y) \, dx \, dy = \int_{0}^{1} \int_{1}^{1} k(x^{2} + y^{2}) \, dx \, dy = \int_{0}^{1} \int_{1}^{1} (kx^{2} + ky^{2}) \, dx \, dy = \int_{0}^{1} \left[\int_{1}^{1} (kx^{2} + ky^{2}) \, dx \, dy = \int_{0}^{1} \left[\int_{1}^{1} (kx^{2} + ky^{2}) \, dx \, dy = \int_{0}^{1} \left[\int_{1}^{1} (kx^{2} + ky^{2}) \, dx \, dy = \int_{0}^{1} \left[\int_{1}^{1} (kx^{2} + ky^{2}) \, dx \, dy = \int_{0}^{1} \left[\int_{1}^{1} (kx^{2} + ky^{2}) \, dx \, dy = \int_{0}^{1} \left[\int_{1}^{1} (kx^{2} + ky^{2}) \, dx \, dy = \int_{0}^{1} \left[\int_{1}^{1} (kx^{2} + ky^{2}) \, dx \, dy = \int_{0}^{1} \left[\int_{1}^{1} (kx^{2} + ky^{2}) \, dx \, dy = \int_{0}^{1} \left[\int_{1}^{1} (kx^{2} + ky^{2}) \, dx \, dy = \int_{0}^{1} \left[\int_{1}^{1} (kx^{2} + ky^{2}) \, dx \, dy = \int_{0}^{1} \left[\int_{1}^{1} (kx^{2} + ky^{2}) \, dx \, dy = \int_{0}^{1} \left[\int_{1}^{1} (kx^{2} + ky^{2}) \, dx \, dy = \int_{0}^{1} \left[\int_{1}^{1} (kx^{2} + ky^{2}) \, dx \, dy = \int_{0}^{1} \left[\int_{1}^{1} (kx^{2} + ky^{2}) \, dx \, dy = \int_{0}^{1} \left[\int_{1}^{1} (kx^{2} + ky^{2}) \, dx \, dy = \int_{0}^{1} \left[\int_{1}^{1} (kx^{2} + ky^{2}) \, dx \, dy = \int_{0}^{1} \left[\int_{1}^{1} (kx^{2} + ky^{2}) \, dx \, dy = \int_{0}^{1} \left[\int_{1}^{1} (kx^{2} + ky^{2}) \, dx \, dy = \int_{0}^{1} \left[\int_{1}^{1} (kx^{2} + ky^{2}) \, dx \, dy = \int_{0}^{1} \left[\int_{1}^{1} (kx^{2} + ky^{2}) \, dx \, dy = \int_{0}^{1} \left[\int_{1}^{1} (kx^{2} + ky^{2}) \, dx \, dy = \int_{0}^{1} \left[\int_{1}^{1} (kx^{2} + ky^{2}) \, dx \, dy + \int_{0}^{1} \left[\int_{1}^{1} (kx^{2} + ky^{2}) \, dx \, dy + \int_{0}^{1} \left[\int_{1}^{1} (kx^{2} + ky^{2}) \, dx \, dy + \int_{0}^{1} \left[\int_{1}^{1} (kx^{2} + ky^{2}) \, dx \, dy + \int_{0}^{1} \left[\int_{1}^{1} (kx^{2} + ky^{2}) \, dx \, dy + \int_{0}^{1} \left[\int_{1}^{1} (kx^{2} + ky^{2}) \, dx \, dy + \int_{0}^{1} \left[\int_{1}^{1} (kx^{2} + ky^{2}) \, dx \, dy + \int_{0}^{1} \left[\int_{1}^{1} (kx^{2} + ky^{2}) \, dx \, dy + \int_{0}^{1} \left[\int_{1}^{1} (kx^{2} + ky^{2}) \, dx \, dy + \int_{0}^{1} \left[\int_{1}^{1} (kx^{2} + ky^{2}) \, dx \, dx + \int_{0}^{1} \left[\int_{1}^{1} (kx^{2} + ky^{2}) \, dx \, dx + \int_{0}^{1} \left[\int_{1}^{1} (kx^{2} + ky^{2}) \, dx \, dx + \int_{0}^{1} \left[\int_{1}^{1} (kx^{2} + ky^{2}) \, dx \, dx + \int_{0}^{1} \left[\int_{1}^{1} (kx^{2} +$$

b)
$$\int_{0}^{\frac{1}{2}} \int_{\frac{1}{2}}^{1} f_{X,Y}(x,y) \, dx \, dy = \int_{0}^{\frac{1}{2}} \int_{\frac{1}{2}}^{1} k(x^2 + y^2) \, dx \, dy = \int_{0}^{\frac{1}{2}} \int_{\frac{1}{2}}^{1} (kx^2 + ky^2) \, dx \, dy = \int_{0}^{\frac{1}{2}} \left[\int_{\frac{1}{2}}^{1} (kx^2 + ky^2) \, dx \, dy \right]$$

$$= \int_{0}^{\frac{1}{2}} \left[\int_{\frac{1}{2}}^{1} kx^2 \, dx + \int_{\frac{1}{2}}^{1} ky^2 \, dx \right] dy = \int_{0}^{1} \left[k \cdot \frac{x^3}{3} \left[\frac{1}{2} + k \cdot y^2 \cdot x \right] \frac{1}{2} \right] dy$$

$$= \int_{0}^{\frac{1}{2}} \left[\frac{1}{3} k - \frac{1}{24} k + k \cdot y^2 - \frac{1}{2} k \cdot y^2 \right] dy = 1/4$$

$$f_{X}(x) = \int_{0}^{1} f_{X,Y}(x,y) \, dy$$

$$= \int_{0}^{1} \left[(kx^{2} + ky^{2}) \, dy \right]$$

$$= \int_{0}^{1} \left[\frac{3}{2}x^{2} + \frac{3}{2}y^{2} \right] \, dy$$

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

$$f_{X}(x) \cdot f_{Y}(y) = \left(\frac{3}{2}x^{2} + \frac{1}{2} \right) \cdot \left(\frac{1}{2} + \frac{3}{2}y^{2} \right) = \frac{9}{4}x^{2}y^{2} + \frac{1}{4}x^{2} + \frac{3}{4}y^{2} + \frac{1}{4} \neq f_{X,Y}(x,y)$$
So the random variable X and Y are not independent to each other.

Some General Advice

- Be confident.
- Be in contact with faculty members.
- Be in touch with your classmates.
- Be organized in your academic life.