

Lecture 6: Discrete & Continuous Probability and Random Variables

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Math Camp

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1 Finishing Basics of Expectation and Variance

- Variance of Discrete Random Variables

2 Continuous Probability

- Introduction
- Continuous Probability Densities
- Cumulative Density Function
- Expectation of Continuous Random Variables
- Variance of Continuous Random Variables

3 Properties of Expectation and Variance

- Properties of Expectation
- Properties of Variance

Variance of Discrete Random Variables

Definition

The *variance* of a random variable is the expected value of its squared deviations from μ . If X is discrete with pdf $p_X(k)$,

$$\text{VAR}[X] = \sigma^2 = E[(X - \mu)^2] = \sum_{\forall k} (k - \mu)^2 \cdot p_X(k).$$

Note that $\mu = E[X]$. Also note that the standard deviation σ is the square root of the variance.

$$\text{VAR}[X] = E[(X - E[X])^2]$$

Just as the expected value has a clear analogy to the “center of balance” of a distribution, the variance is analogous to the “moment of inertia.”

Imagine placing two distributions on a turntable and pushing on each with identical force δ ” from the center. A distribution with low variance would spin faster, and a distribution with high variance would spin slower.

Example

What is the variance of a six sided die roll?

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Answer

First calculate the expected value of a single die roll.

$$\begin{aligned}\sum_{1}^{6} x_i * p(x_i) &= 1(1/6) + 2(1/6) + 3(1/6) \\ &\quad + 4(1/6) + 5(1/6) + 6(1/6) \\ E[X] = \mu &= 3.5\end{aligned}$$

Answer

Then the Variance is:

$$\begin{aligned}\sum_{i=1}^6 (x_i - \mu)^2 p(x_i) &= (1 - 3.5)^2(1/6) + (2 - 3.5)^2(1/6) + (3 - 3.5)^2(1/6) \\ &\quad + (4 - 3.5)^2(1/6) + (5 - 3.5)^2(1/6) + (6 - 3.5)^2(1/6) \\ &= 17.5(1/6) \\ &= 2.92\end{aligned}$$

Example

An urn contains five chips, two red and three white. Two chips are drawn at random, *with replacement*. Let X denote the number of red chips drawn. What is $\text{VAR}[X]$?

Example of Discrete RV Variance

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Answer

X is binomial, $n = 2$ and $p = \frac{2}{5}$.

$$E[X] = np = (2)(2/5) = 4/5$$

$$\begin{aligned}\text{VAR}[X] &= \sum (x - E[X])^2 f(x) = \sum_{i=0}^2 (i - .8)^2 \binom{2}{i} \left(\frac{2}{5}\right)^i \left(\frac{3}{5}\right)^{2-i} \\ &= (0 - .8)^2 \cdot \binom{2}{0} (2/5)^0 (3/5)^2 + (1 - .8)^2 \cdot \binom{2}{1} (2/5)^1 (3/5)^1 + (2 - .8)^2 \binom{2}{2} (2/5)^2 (3/5)^0 \\ &= (16/25)(1)(9/25) + (1/25)(2)(6/25) + (36/25)(1)(4/25) = \frac{12}{25}\end{aligned}$$

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Example

Example

Some process has equal chance of generating a number between $[0,1]$.
What is the probability the process generate a number less that 0.5?

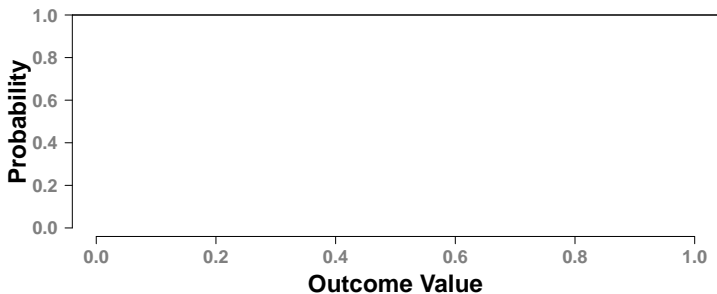
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Draw a number line.

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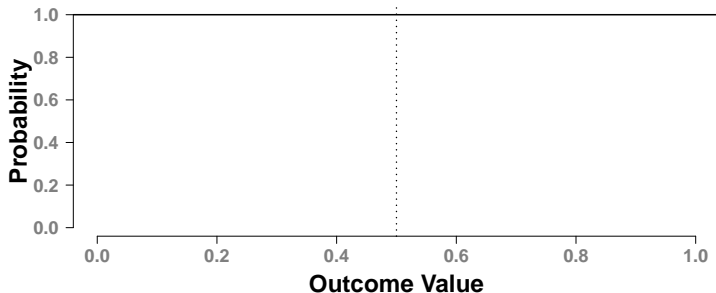
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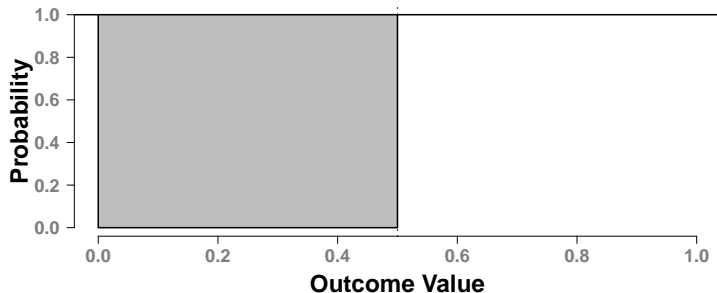
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Definition

A probability density function P on a set of real numbers S is called *Continuous* if there exists a function $f(t)$ such that for any closed interval $[a, b] \subset S$,

$$P([a, b]) = \int_a^b f(t) dt$$

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Comment

$f(t)$ must satisfy two properties:

- 1 $f(t) \geq 0$ for all t
- 2 $\int_{-\infty}^{\infty} f(t) dt = 1$

Frequently, we refer to the probability distribution function as the “pdf.”

Example

- What continuous probability function describes the uniform chance of drawing a number on the range $[0,10]$?
 - What is the probability that a randomly selected number falls between 4 and 7?
 - What is the probability that a randomly selected number falls between $[2,4]$ or $[6,9]$?
- What continuous probability function describes the uniform chance of drawing a number on the range $[a,b]$?

Continuous Probability Example 2

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- What is the probability that $2/3 \leq x \leq 1$?

Answer

- Yes. Recall the two properties $f(x)$ just satisfy: (A) $f(x)$ is always non-negative; (B) The integral of $f(x)$ across the function's domain = 1.
- $\int_0^{1/3} 3x^2 = x^3 \Big|_0^{1/3} = \mathbf{1/27}$
- $\int_{2/3}^1 3x^2 = x^3 \Big|_{2/3}^1 = 1 - 8/27 = \mathbf{19/27}$.

Example 3

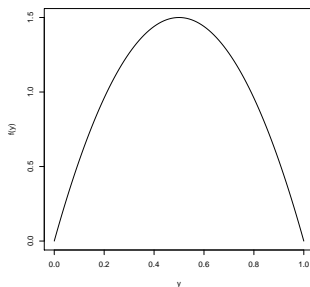
Example

Suppose we would like a continuous function to describe a variable Y in such a way that it more likely to produce y 's near the middle of the range more frequently than near the extremes of the range. What might this look like?

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I've drawn one where
 $f(y) = 6y(1 - y)$.

Example 3

Example

Does this meet the requirements of a probability density function?

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Answer

Yes!

- 1 $f_Y \geq 0$ for all y ;
- 2 $P(\Omega) = 1$.

Then, with this function, answer the following:

- What is the probability that a number is less than 0.5?
- Greater than 0.5?
- Less than 0.25?
- Greater than 0.25?
- Exactly 0.25?

Cumulative Density Function

Definition

The cumulative density function (called "cdf") of a continuous, real-valued random variable Y is an indefinite integral of its pdf.

$$F_Y(x) = \int_{-\infty}^x f_Y(x) dx = P(X \leq x)$$

Note the difference in notation between a cumulative distribution function (cdf) and a probability distribution function (pdf).

Theorem

$$F(y) = \int_{-\infty}^y f(t) dt; \text{ conversely,}$$
$$\frac{d}{dy} F(y) = f(y)$$

Cumulative Density Examples

Example

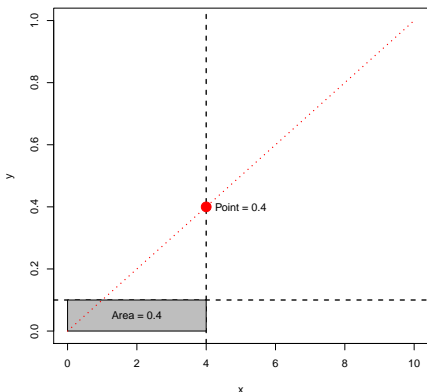
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Cumulative Density Examples

Example

A uniform distribution covers the range $[0,10]$. What is the probability that a random number drawn is less than 4?

Illustration of Cumulative Density Function



$$\begin{aligned}\int_0^4 f(x)dx &= .1x \Big|_0^4 \\ &= .1(4) - .1(0)\end{aligned}$$

$$\int f(x) = F_X(x)$$

$$F_X(x) = 0.1x$$

$$F_X(4) = 0.1(4)$$

$$= .4$$

Theorem

Let Y be a continuous random variable with cdf $F_Y(y)$, Then,

- 1 $P(Y > s) = 1 - F_Y(s)$
- 2 $P(r < Y \leq s) = F_Y(s) - F_Y(r)$
- 3 $\lim_{x \rightarrow +\infty} F_Y(y) = 1$
- 4 $\lim_{x \rightarrow -\infty} F_Y(y) = 0$

Additionally:

- The CDF is a monotonically nondecreasing continuous function.
- Will be useful for easily/quickly calculating the probability a variable takes a value on some interval.

An Example

In Encinitas the distribution of waves a broski, Tad, takes every hour, Y , is described by the pdf:

$$f_Y(y) = y * e^{-y}, y \geq 0.$$

What is the median of Encinitas' wave-income distribution?

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What is the median of Encinitas' wave-income distribution?

In looking for the median, we are looking for the number where the probability below is equal to the probability above.

$$\begin{aligned} F_Y(y) &= \int_0^{\infty} ye^{-y} \\ &= -e^{-y} \\ 0.5 &= -e^{-y} \Big|_y^{\infty} \\ 0.5 &= 0 - (-e^{-y}) \\ 0.5 &= e^{-y} \\ \ln(0.5) &= -y \rightarrow \mathbf{0.69} \end{aligned}$$

Definition

If Y is a continuous random variable with pdf $f_Y(y)$,

$$E[Y] = \int_{-\infty}^{\infty} y \cdot f_Y(y) dy$$

In the same way as the discrete case, one may think of this as a center of gravity of the distribution.

Example

A continuous uniform density function, $f(x) = 1$ is defined on the range $[0,1]$. What is the expected value of this function?

Answer

$$\begin{aligned} E[X] &= \int_0^1 x \cdot f(x) dx \\ &= \int_0^1 1x dx \\ &= \frac{1}{2} x^2 \Big|_0^1 \\ &= \frac{1}{2} (1 - 0) \\ &= \mathbf{0.5} \end{aligned}$$

Definition

Let X be a real-valued random variable with density function $f_X(x)$. Then, the *variance*, σ^2 , is defined by

$$\begin{aligned}\sigma^2 &= E[(X - \mu)^2] \\ &= \int_{-\infty}^{\infty} (x - \mu)^2 f(x) dx\end{aligned}$$

Expectation Identities

Theorem

For any constant c , $E[c] = c$

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For any constant c and random variable X ,

$$E[cX] = cE[X]$$

Theorem

For any X and Y that are random variables that have finite expectations

$$E[X + Y] = E[X] + E[Y]$$

Expected Value of a Function of a Random Variable

Theorem

Suppose X is a random variable with pdf $p_X(k)$. Let $g(X)$ be any function of X . Then the expected value of the random variable $g(X)$ is given by

$$E[g(X)] = \int_{-\infty}^{\infty} g(x) \cdot f_X(x) dx$$

Example

Suppose $X = f_X(x)$ is the uniform density function defined on the range $[0,1]$. Then $f_X(x) = 1$. Suppose $g(x) = 1/5x$.

$$\begin{aligned} E[g(X)] &= \int_0^1 \frac{1}{5}x \cdot f_X(x) dx \\ &= \frac{1}{5} \left[\frac{1}{2}x^2 \right]_0^1 = \frac{1}{10} \end{aligned}$$

Example

The Zetas want to burn their gang-signs into a wall. Suppose the amount of fuel in the torch, Y , is a random variable with pdf

$$f_Y(y) = 3y^2, 0 < y < 1.$$

In the past they have been able to burn a circle whose radius is 20 times the size of y . How much area can they expect to burn?

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Answer

By problem setup, $g(Y) = 20\pi Y^2$

$$\begin{aligned} E[g(Y)] &= \int_0^1 20\pi y^2 \cdot 3y^2 dy = 60\pi \int_0^1 y^4 \\ &= \frac{60\pi y^5}{5} \Big|_0^1 \\ &= 12\pi \text{ft}^2 \end{aligned}$$

Recall the definition of Variance:

Definition

The *variance* of a random variable is the expected value of its squared deviations from μ . If X is discrete with pdf $p_X(k)$,

$$\text{VAR}[X] = \sigma^2 = E[(X - \mu)^2] = \sum_{\forall k} (k - \mu)^2 \cdot p_X(k).$$

Note that $\mu = E[X]$. Also note that the standard deviation σ is the square root of the variance.

$$\text{VAR}[X] = E[(X - E[X])^2]$$

A Variance Identity

Theorem

Let X be any random variable (discrete or continuous), having mean μ and for which $E[X^2]$ is finite. Then,

$$\text{VAR}[X] = E[(X - E[X])^2] = E[X^2] - E[X]^2$$

Proof.

$$\begin{aligned}\text{VAR}[X] &= E[(X - E[X])^2] \\ &= E[X^2 - 2XE[X] + E[X]^2] \\ &= E[X^2] - E[2XE[X]] + E[E[X]^2] \\ &= E[X^2] - E[2X]E[X] + E[X]^2 \\ &= E[X^2] - 2E[X]E[X] + E[X]^2 \\ &= E[X^2] - E[X]^2\end{aligned}$$

Example

Given the following pdf, what is the variance?

$$f(x) = 3(1 - x)^2, 0 < x < 1$$

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Answer

$$E[X] = \int_0^1 x \cdot 3(1 - x)^2 dx = 3 \int_0^1 x - 2x^2 + x^3 dx = 3$$

$$E[X^2] = \int_0^1 x^2 \cdot 3(1 - x)^2 dx = 3 \int_0^1 x^2 - 2x^3 + x^4 dx = \frac{1}{10}$$

$$\text{VAR}[X] = E[X^2] - E[X]^2 = \frac{3}{80}$$

Theorem

For any random variable X and constant c

$$\text{VAR}[cX] = c^2 \text{VAR}[X]$$

$$\text{VAR}[X + c] = \text{VAR}[X]$$

Proof.

For ease, let $E[X] = \mu$. Then $E[cX] = c\mu$, and

$$\begin{aligned} V[cX] &= E[(cX - c\mu)^2] = E[c^2(X - \mu)^2] \\ &= c^2 E[(X - \mu)^2] \\ &= c^2 \text{VAR}[X] \end{aligned}$$



Theorem

Let X and Y be two independent random variables. Then,

$$\text{VAR}[X + Y] = \text{VAR}[X] + \text{VAR}[Y]$$

Proof.

Let $E[X] = a$ and $E[Y] = b$.

$$\begin{aligned}\text{VAR}[X + Y] &= E[(X + Y)^2] - (a + b)^2 \\ &= E[X^2] + 2E[XY] + E[Y^2] - a^2 - 2ab - b^2 \\ &= E[X^2] - a^2 + E[Y^2] - b^2 = \text{VAR}[X] + \text{VAR}[Y]\end{aligned}$$



What do we require of a probability distribution/density model?

- It must be non-negative for all outcomes in its sample space
- It must sum or integrate to one across the sample space.

By this definition, $f(y) = \frac{y}{4} + 7\frac{y^3}{2}, 0 \leq y \leq 1$ is a valid pdf.

- $f(y) \geq 0, \forall y, 0 \leq 1$, and;
- $\int_0^1 \frac{y}{4} + 7\frac{y^3}{2} dy = 1$

But, how useful is this model?

- A pdf has utility as a probability model if it actually models the behavior of the real (or political) world.
- A surprisingly small number of distributions describe these real world outcomes.
- Many measurements/outcomes are the result of the same set of assumptions about the data generating process
 - Number of fraud incidents
 - Number of latrines built
 - Feeding patterns of zebra muscles