

Introduction to Data Science With Probability and Statistics

Lecture 13: Variance of Discrete and Continuous Random Variables

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What will we learn today?

- ❑ Variance
- ❑ *A Modern Introduction to Probability and Statistics, Chapter 7*



Review from Last Time

The **expectation** or **expected value** of a discrete random variable X that takes the values a_1, a_2, \dots and with pmf p is given by:

$$E[X] = \sum_i a_i P(X = a_i) = \sum_i a_i p(a_i)$$

The **expectation** or **expected value** or **mean** of a continuous random variable X with probability density function f is:

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

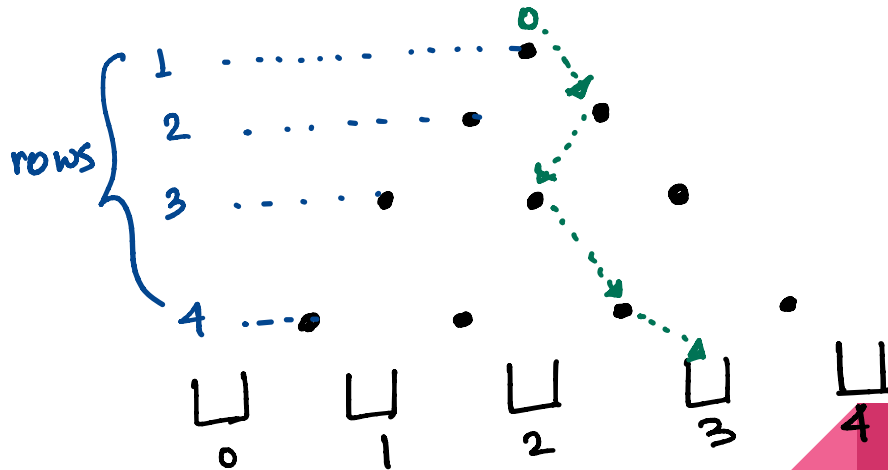
Change-of-Variables Formula: Let X be a random variable and $g: \mathbb{R} \rightarrow \mathbb{R}$ be a function. Then:

$$E[g(x)] = \sum_i g(a_i) P(X = a_i) \quad \text{and} \quad E[g(x)] = \int_{-\infty}^{\infty} g(x) f(x) dx$$

Plinko!

Let X be the random variable describing the result in each round of Plinko with n rows and probability p of moving to the right off of each peg. (Ignoring the edges for now)

<https://www.youtube.com/watch?v=naUppHrHJpI>



Plinko!

Let X be the random variable which denotes the final bin on which the ball will eventually end up. At each row, the ball is equally likely to go left or right.

What is the distribution that X follow?

for every row i , let $y_i \sim \text{Ber}(p)$. = $\begin{cases} 1 (\text{moving right}) , & p . \\ 0 (\text{moving left}) , & 1-p \end{cases}$

$X = y_1 + y_2 + y_3 + y_4 = \sum y_i$ } for the simplified example.

$\therefore X \sim \text{Bin}(4, p)$

$X \sim \text{Bin}(n, p)$ } General case with 'n' rows.

Plinko!

Let X be the random variable which denotes the final bin on which the ball will eventually end up. At each row, the ball is equally likely to go left or right.

What is the expected value of X ? $E[X] = \sum_i a_i p(a_i) \dots$

$$\begin{aligned} E[X] &= E\left[\sum y_i\right] = E[y_1 + y_2 + \dots + y_n] \\ &= E[y_1] + E[y_2] + \dots + E[y_n] \quad \left\{ \begin{array}{l} \because E[ax+b] = aE[x] + b \\ \therefore E[y_i] = 0 \cdot (1-p) + 1 \cdot p \\ \quad \quad \quad = p \\ \text{if } y_i \sim \text{Ber}(p) \end{array} \right. \\ &= p + p + \dots + p = n \cdot p \end{aligned}$$

$E[X] = np$

Plinko!

Let X be the random variable which denotes the final bin on which the ball will eventually end up. At each row, the ball is equally likely to go left or right.

What is the **variance** of X ?

what have we seen before?

- Given x_1, x_2, \dots, x_n their sample variance is $\frac{1}{n-1} \sum_{k=1}^n (x_k - \bar{x})^2$

which is basically : Avg of squared differences
from the mean.

$$: E[(X - E[X])^2] //$$

Variance

The **variance** $\text{Var}(X)$ of a random variable X is the number:

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

The **standard deviation** of a random variable X is the square root of the variance:

$$\text{SD}(X) = \sqrt{\text{Var}(X)}$$



Variance

The **variance** $\text{Var}(X)$ of a random variable X is the number:

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

Alternatively: $\text{Var}(X) = E[X^2] - E[X]^2$

Proof :
$$\begin{aligned}\text{Var}[X] &= E[(X - E[X])^2] = E[X^2 - 2XE[X] + (E[X])^2] \\ &= E[X^2] + E[-2X \underbrace{E[X]}_{\mu}] + E[(\underbrace{E[X]}_{\mu})^2] \\ &= E[X^2] - 2E[X] \cdot E[X] + (E[X])^2 \\ &= E[X^2] - E[X]^2 \quad \parallel\end{aligned}$$

μ : constant value.

Plinko!

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What is the **variance** of X ?



Plinko!

Let X be the random variable which denotes the final bin on which the ball will eventually end up. At each row, the ball is equally likely to go left or right.

What is the **variance** of X ?

$$\text{Var}[X] = \text{Var}[Y_1 + Y_2 + Y_3 + \dots + Y_n]$$

$$\begin{aligned}\text{Var}[Y_i] &= E[Y^2] - E[Y]^2 \\ &= (0^2 \cdot (1-p) + 1^2 p) - p^2 \\ &= p - p^2 = p(1-p)\end{aligned}$$

Variance

Let X be the random variable describing the result in each round of Plinko with n rows and probability p of moving to the right off of each peg. (Ignoring the edges for now.)

What is the **variance** of $X \sim \text{Bin}(n, p)$?

If X and Y are **independent**, then $\text{Var}(X+Y) = \text{Var}(X) + \text{Var}(Y)$

$$\text{So, } \text{Var}(X) = \text{Var}\left(\sum_{i=1}^n Y_i\right) = \text{Var}(Y_1) + \text{Var}(Y_2) + \dots + \text{Var}(Y_n)$$

$$= n \cdot p(1-p)$$

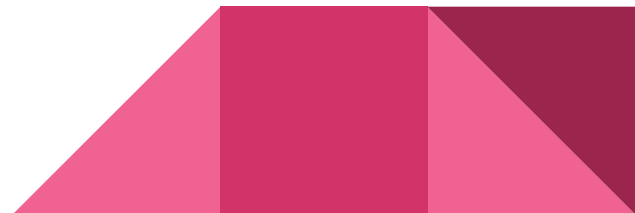
Quick Summary

If $X \sim \text{Ber}(p)$, then:

- $E[X] = p$
- $\text{Var}(X) = p(1 - p)$

If $X \sim \text{Bin}(n, p)$, then:

- $E[X] = np$
- $\text{Var}(X) = np(1 - p)$



The Binomial Distribution

Example: You are taking a 12-question quiz and believe there to be about a 75% chance that any one of your answers is correct. What is your expected grade on the quiz? What is the variance in your quiz grade?



The Binomial Distribution


Example: You are taking a 12-question quiz and believe there to be about a 75% chance that any one of your answers is correct. What is your expected grade on the quiz? What is the variance in your quiz grade?

Let X : # correct answers.

$$\therefore X \sim \text{Bin}(n=12, p=0.75)$$

$$E[X] = n \cdot p = 12 \times 0.75 = 9$$

$$\text{Var}(X) = n \cdot p \cdot (1-p) = 12 \times 0.75 \times 0.25 = 2.25$$

$$\text{S.D.}(X) = 1.5$$


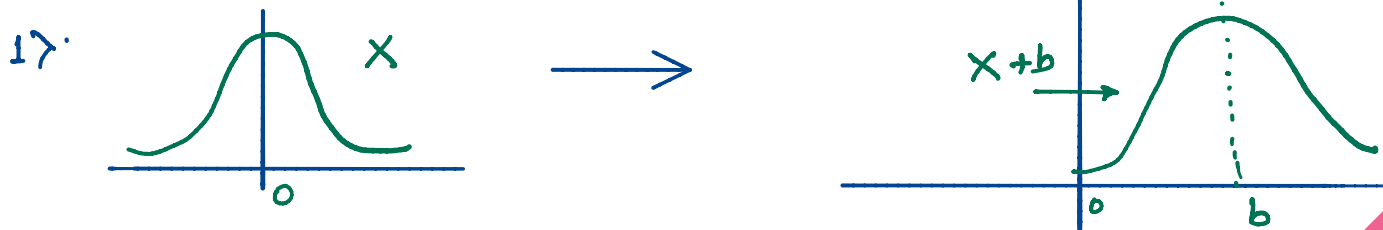
More Facts about Variance

Expectation is linear: $E[aX + b] = aE[X] + b$

What about variance?

Consider 2 cases :

1>. Effect of adding a constant ($X + \underline{b}$)



$$\therefore \text{Var}(X+b) = \text{Var}(X)$$

More Facts about Variance

Expectation is linear: $E[aX + b] = aE[X] + b$

What about variance?

2> Effect Multiplying a constant to the random variable. (aX)

$$\begin{aligned}\text{Var}(aX) &= E[(aX)^2] - (E[aX])^2 = E[a^2X^2] - (aE[X])^2 \\ &= a^2 E[X^2] - a^2 E[X]^2 \\ &= a^2 (E[X^2] - E[X]^2) = a^2 \text{Var}(X)\end{aligned}$$

\therefore

$$\boxed{\text{Var}(aX) = a^2 \text{Var}(X)}$$

Mean and Variance of a Uniform Random Variable

Example: Let $X \sim U[\alpha, \beta]$. What are $E[X]$ and $\text{Var}(X)$?



Mean and Variance of a Uniform Random Variable

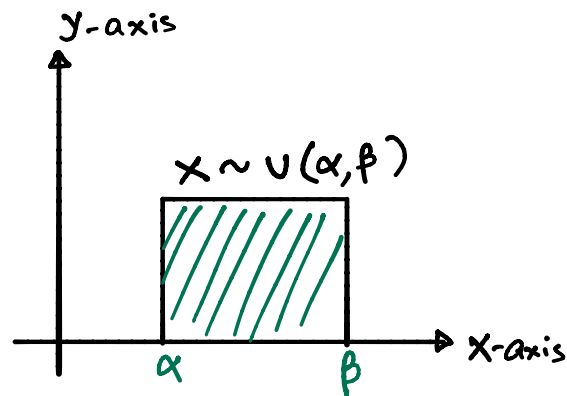
Example: Let $X \sim U[\alpha, \beta]$. What are $E[X]$ and $\text{Var}(X)$?

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

$$\therefore E[X] = \int_{-\infty}^{\infty} x f(x) dx = \int_{\alpha}^{\beta} x \cdot \frac{1}{\beta - \alpha} \cdot dx$$

$$= \frac{1}{\beta - \alpha} \int_{\alpha}^{\beta} x dx = \frac{1}{\beta - \alpha} \cdot \frac{x^2}{2} \Big|_{\alpha}^{\beta}$$

$$= \frac{1}{\beta - \alpha} \cdot \frac{1}{2} (\beta^2 - \alpha^2) = \boxed{\frac{1}{2} (\alpha + \beta)}$$



Mean and Variance of a Uniform Random Variable

Example: Let $X \sim U[\alpha, \beta]$. What are $E[X]$ and $\text{Var}(X)$?

$$\text{Var}(X) = \underbrace{E[X^2]} - \underbrace{E[X]^2}_{\text{already done.}}$$

$$E[X^2] = \int_{-\infty}^{\infty} x^2 \cdot f(x) = \int_{\alpha}^{\beta} x^2 \cdot \frac{1}{\beta - \alpha} \cdot dx = \frac{1}{\beta - \alpha} \cdot \frac{x^3}{3} \Big|_{\alpha}^{\beta} = \frac{1}{\beta - \alpha} \cdot \frac{1}{3} \cdot \beta^3 - \alpha^3$$

$$= \frac{1}{\beta - \alpha} \cdot \frac{1}{3} \cdot (\beta - \alpha) (\beta^2 + \alpha^2 + \alpha\beta)$$

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

Quick Summary

If $X \sim \text{Ber}(p)$, then:

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If $X \sim \text{Bin}(n, p)$, then:

- $E[X] = np$
- $\text{Var}(X) = np(1-p)$

If $X \sim U[\alpha, \beta]$, then:

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



Next Time:

- ❖ The Normal Distribution !

